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## PREDICTORS OF STUDENT OUTCOMES IN DEVELOPMENTAL MATH AT A PUBLIC COMMUNITY AND TECHNICAL COLLEGE

A Dissertation submitted to the Graduate College of Marshall University

In partial fulfillment of the requirements for the degree of Doctor of Education

> Curriculum and Instruction by Linda Darlene Hunt

Approved by Dr. Carl Johnson, Chairperson Dr. Paula Lucas Dr. Laura Adkins Dr. Linda Hankins

> Marshall University December 2011

#### ABSTRACT

With the wide range of abilities of community college students, proper course placement is crucial. Therefore, having better predictors of success can help improve placement of students for their achievement. This study analyzed student predictors, instructor predictors, and classroom predictors in relation to student final exam score and student final grade in Elementary Algebra and Intermediate Algebra classes. Student predictors included gender, ACT math score, SAT math score, community college enrollment, math pretest score, and ASC grade. Instructor predictors included gender, employment status, Mozart music use, and ALEKS software use. Classroom predictors included time of day, number of class meetings per week, and class size. The Elementary Algebra and Intermediate Algebra data sets were analyzed with simple and stepwise multiple regression as well as simple and stepwise binary logistic regression.

The study of specific predictors that impact student achievement in developmental mathematics revealed the following. When analyzed individually, Elementary and Intermediate Algebra shared ACT Math score, community college enrollment, and math pretest score as predictors of final exam score. When analyzed individually, Elementary and Intermediate Algebra shared ACT Math score, math pretest score, and ASC grade as predictors of final grade. When analyzed in combinations, Elementary and Intermediate Algebra shared ACT Math score and math pretest score as common predictors of final exam score. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act. When analyzed in combinations, Elementary and Intermediate Algebra shared act.

#### DEDICATION

Math has never been easy for me; I've always had to work at it. I think this helps me relate to my developmental mathematics students who struggle to understand. This dissertation is dedicated to all of the teachers along the way who had the patience to make sure I finally got it. A special thanks goes to the elementary school teachers who worked diligently at teaching long division and to my dad for helping me with homework, especially how to remember my nines in the multiplication table. I was blessed to have a strong math curriculum in high school and teachers such as Jerry Schultz who continued to help me even when I was no longer his official student. I continued to be blessed in college with great teachers such as James Allison, Mary Ellen Komorowski, and David Brown who spent many office hours helping me with homework and extra exercises I wanted to do. A special thank you goes to Bob Bickel who started helping me in graduate school in 1992 and who encouraged me to complete my dissertation. None of my accomplishments would be possible if it weren't for my third grade teacher Alta McNinch who helped me learn to read. If it hadn't been for her, I would have never been able to read to learn. Mostly, I'd like to dedicate this to my developmental mathematics students for whom I work diligently the way my teachers worked for me.

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#### ACKNOWLEDGMENTS

First, I'd like to acknowledge my dissertation committee: Carl Johnson, who served as chair into his retirement; Paula Lucas, who served as my minor chair; Laura Adkins, who assisted with all aspects of the design and analysis of data; and Linda Hankins, who helped me with my minor comprehensive exams on math anxiety. Special acknowledgments go to my friend Gemmie who really is a gem and my colleague Pam Bird Duelley, aka vent sister and proofreader extraordinaire. Thanks to David Childress, Jim Fox, and Sherri Ritter for their technical assistance with formatting.

Second, I'd like to acknowledge my fellow doctoral students who inspired and supported me through the coursework and dissertation process.

Third, I'd like to acknowledge my work colleagues for their inspiration: my longtime friend Kay Thompson for telling me this is something I must complete; my Delta Kappa Gamma sorority sister Dr. Barbara Walters for relighting my fire when I'd all but given up; my college president Dr. Greg Adkins, academic dean Dr. Janie Kitchen, and my division chair Dr. Keith Brammell, who had faith that I'd complete my dissertation.

Fourth, I'd like to acknowledge my best friend Mike Beck, Ph.D., who helped me every step of the way with unending support.

Fifth, I'd like to acknowledge my parents for their variety of support.

Finally, and most importantly, I acknowledge God because "With God all things are possible" (Mark 10:27)!

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## QUOTATION

My favorite quotation is from Jesse Stuart's (1949) book *A Thread That Runs So True*:

And I am firm in my belief that a teacher lives on and on through his students. I will live if my teaching is inspirational, good, and stands firm for good values and character training. Tell me how can good teaching ever die? Good teaching is forever and the teacher is immortal. (p. 7)

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## PREDICTORS OF STUDENT OUTCOMES IN DEVELOPMENTAL MATH AT A COMMUNITY AND TECHNICAL COLLEGE CHAPTER ONE

#### INTRODUCTION

Today there is much interest in the preparation of students for postsecondary education. People focused on this issue range from philanthropists Bill and Melinda Gates (Jaschik, 2008) to the President of the United States (Obama, 2009). The Gates Foundation is worried about "poor graduation and retention rates" (Jaschik, 2008) while President Obama is concerned about remediating basic mathematics skills and finding out what is preventing students from being successful.

According to the American Mathematical Association of Two-Year Colleges, AMATYC (Writing Team and Task Force of the Standards for Introductory College Mathematics Project, 1995, p. ix), "Of 1,295,000 students studying mathematics in two-year college mathematics departments, 724,000 (56%) were studying at the remedial level. By contrast, in four-year college and university mathematics departments, 261,000 (15% of the mathematics enrollment) were studying at the remedial level."

The National Center for Education Statistics, NCES, reports that in the fall of 2000, 71% of two- and four-year postsecondary institutions offered remedial courses in mathematics (Parsad & Lewis, 2003). Additionally, in the fall of 2000, 22% of entering freshmen enrolled in remedial mathematics classes (Parsad & Lewis, 2003). McCabe notes that "sixty-two percent of remedial education

students are deficient in mathematics..." (2000, p. 41). Clery and Topper (2008, p. 1) asserted that convincing students who require developmental coursework to enroll in the requisite courses without delay, "and then doing everything possible to help them succeed, will improve degree completion rates" because approximately half of students fail to complete their developmental courses.

As noted in the study by Parsad and Lewis (2003), nearly three-fourths of the institutions in the United States provided courses in developmental mathematics with almost one quarter of entering freshmen enrolling in these courses. This occurred at a cost to the nation of one billion dollars (Saxon & Boylan, 2001). Clearly, the issue of students being ill-prepared for college-level mathematics is a serious problem. Because student success will encourage retention rates, predictors of student achievement will be useful in enhancing the success of all students (Heath, 1995; Huber, 2006). A method to predict the success of these students could help save considerable time and money. In this study, the association between potential predictors (student characteristics, instructor characteristics, and classroom characteristics) and the student outcomes (as measured by their final exam scores as well as their final grades) were investigated.

#### Definition of Key Terms

Key terms in this study include the following: developmental mathematics, outcomes, potential predictors, and the community and technical college for this study. Developmental mathematics includes Elementary Algebra and Intermediate Algebra. Outcomes are defined as the students' final grades and

students' final exam scores. Potential predictors are in three categories: student (gender, ACT Math score, SAT Math score, college, math pretest, and ASC grade), instructor (gender, employment status, Mozart for Your Mind tape use, and ALEKS software use), and classroom (time of day, number of class meetings per week, and class size). The Marshall University Community and Technical College was the institutional site for this study. More detailed definitions follow.

#### Developmental Mathematics

Traditionally, developmental mathematics, which cannot be applied toward post-secondary degree requirements or graduation, has included arithmetic, elementary algebra, intermediate algebra, and geometry. For the purpose of this study, developmental mathematics is restricted to Elementary Algebra and Intermediate Algebra that are prerequisite skills for college-level mathematics. This is consistent with the definition used in the field of developmental mathematics (Parsad & Lewis, 2003). Although some of the literature still uses the term remedial mathematics, developmental mathematics is the preferred term (Saxon & Boylan, 2001) and was the term used in this study. Note that in the following descriptions from the 1999 – 2001 Marshall University Undergraduate Catalog (Marshall Community and Technical College, 2001) the preferred term, developmental mathematics, is used.

#### Elementary Algebra

MAT 096 Developmental Mathematics. 4 hrs. CR/NC.

To help students develop mathematical and elementary algebra skills with labs. Topics include fractions, decimals, percents, real numbers,

equations, algebraic expressions, and ratios and proportion. The graduation requirement is increased four hours for students who complete this course. (PR: ASSET or ACT; CR: ASC 099 1 hr.)

Intermediate Algebra

MAT 097 Developmental Algebra. 4 hrs. CR/NC.

To help students develop algebra skills. Topics include factoring, rational expressions, quadratics, logarithms, graphing, systems of equations/inequalities. Graduation requirement is increased four hours for students upon completion of course. (PR: MAT 096 or ASSET or ACT score; CR: ASC 099 1 hr.)

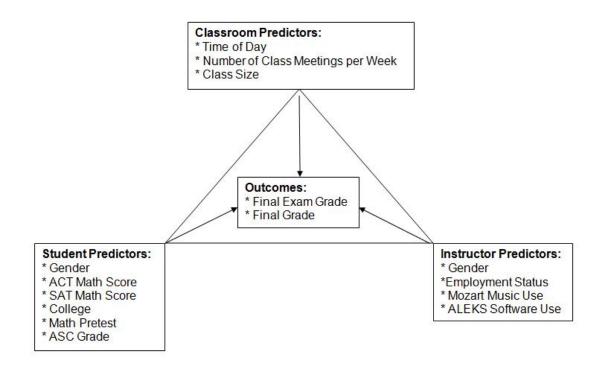
#### Outcomes

Two types of outcomes were predicted. One was the student's final grade for the class. The other was the student's grade on the final exam.

#### Potential Predictors

Although there are a multitude of possible predictors for success, this researcher chose 13 potential predictors. The 13 potential predictors fell into three categories. They included student characteristics, instructor characteristics, and classroom characteristics. Student potential predictors were gender, ACT Math score, SAT Math score, college in which student was enrolled (Community College compared to other colleges on the University campus), math pretest, and Academic Skills Center (ASC) grade (Credit/No Credit). The ASC was a form of supplemental instruction. Four instructor potential predictors were gender, employment status (full time or part time), use of Mozart for Your

Mind tape, and use of ALEKS software. Three classroom potential predictors included time of day, number of class meetings per week, and class size. Figure 1 shows the interaction between the independent and dependent variables. Figure 1 Predictors of Student Success



A list of the independent variables (possible predictors) with their category and

the source of their data can be found in Table 1.

## Table 1 Independent Variables

Potential Predictor	Category	Source of Data
Gender	Student	Student Record
ACT Math Score	Student	Student Record
SAT Math Score	Student	Student Record
College	Student	Student Record
Math Pretest	Student	In-house test
ASC Grade	Student	Student Record
Gender	Instructor	Dept. Records
Employment Status	Instructor	Dept. Records
Mozart for Your Mind Tape Use	Instructor	Instructor
ALEKS Software Use	Instructor	Instructor
Time of Day	Classroom	Class Schedule
No. of Meetings per Week	Classroom	Class Schedule
Class Size	Classroom	Class Schedule

#### Community and Technical College

The site of this study was Marshall University Community and Technical College (MUCTC), which has now become Mountwest Community and Technical College (MCTC), located in Huntington, West Virginia. At the time this study was conducted, MUCTC was a two-year public community and technical college located on the Marshall University campus where students could earn associate degrees. At that time, MUCTC provided all of the developmental course offerings for Marshall University.

#### **Problem Statement**

This study selected 13 characteristics of developmental mathematics students and related them to student achievement. Through multiple regression, models were developed to predict the final exam score as well as final grade based on characteristics of developmental mathematics students, their instructors, and classrooms.

Before specifying a predictive model, stepwise regression was used to reduce the number of predictors to a smaller, more manageable number of independent variables. The data are a deidentified set of existing records for elementary and intermediate algebra students for the fall 2001 semester at Marshall University Community and Technical College (MUCTC). Therefore, this work was exempt from human subject research requirements (see Appendix A).

#### Purpose of the Study

The purpose of this study was to analyze specific independent variables for their ability in predicting student achievement in developmental mathematics. With this information, decisions can be made to improve the student success rate in developmental mathematics.

#### **Research Questions**

Each of the following research questions was considered separately for Elementary Algebra and for Intermediate Algebra.

#### Research Question One

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

#### Research Question Two

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

#### Research Question Three

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

#### Research Question Four

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

In two of the research questions, students' scores on the departmental comprehensive multiple choice final exams were used as the dependent variable. In the other two research questions, the students' final grades (Credit/No Credit) were used as the dependent variable. The departmental policy stated that 25% of the student's final grade was comprised of the final exam. By using final grades as the dependent variable, the remaining part of the students' grades (75%) was influenced by the instructors' choices. See Appendices B and C for copies of the

Elementary and Intermediate Algebra departmental syllabi for full descriptions of the instructors' choices in the remaining 75% of the students' grades. For this study, all analyses were performed using the final exam score (which is a numerical value from 0 to 100) and then all analyses were repeated using the final course grade (which is a dichotomous variable).

#### Significance of the Study

The significance of analyzing predictors of student achievement in developmental mathematics is important in light of the number of college students who require remediation in basic math skills. Students with a wide range of abilities attend college, including students who experience difficulty achieving success in developmental mathematics (Armington, 2002).

#### Procedure

This study utilized an ex-post facto research design with data from a deidentified set of existing records for elementary and intermediate algebra students from the fall 2001 semester at Marshall University Community and Technical College. The analysis used step-wise regression and multiple regression to determine whether student factors (gender, ACT Math score, SAT Math score, college, math pretest, and ASC grade), instructor factors (gender, employment status, Mozart music use, and ALEKS software use) and classroom factors (time of day, number of class meetings per week, and class size) served as predictors of student achievement, both individually and in combinations.

#### **Risks and Potential Benefits of the Research**

There was no risk of identifying specific students used in this study because the results were not reported per individual student, instructor, or classroom. The potential benefits of the research are to learn what independent variables can predict student outcomes in developmental mathematics, thereby adding to the body of literature and providing suggestions for further research.

#### Assumptions of the Study

The researcher made no assumptions about class rank or number of times the student had attempted to complete a course. However, it was assumed that students had graduated from high school or had a GED.

#### **Complete Inclusion/Exclusion Criteria**

Missing data commonly occur when analyzing data. PASW Statistics GradPack (Version 18) was used for the multiple regression analysis, and this software uses list-wise deletion; therefore, a student's information is deleted when there are missing data.

#### Summary

The purpose of this study was to identify specific predictors of student outcomes in developmental mathematics classes. In this study, developmental mathematics consisted of Elementary Algebra and Intermediate Algebra which were prerequisite skills for college-level mathematics. Thirteen potential predictors were selected from three different categories: student characteristics, instructor characteristics, and classroom characteristics. Multiple regression and binary logistic regression were used to identify characteristics that predict student outcomes. These potential predictors were analyzed individually and in combinations. Final exam scores as well as final grades were used as dependent variables in this study.

#### CHAPTER TWO

#### **REVIEW OF THE RELATED LITERATURE**

The literature of the following three areas was considered: student outcomes, potential predictors of student outcomes, and regression analyses of these outcomes. They will be considered separately in that order.

#### Student Outcomes

Grades were frequently the student outcome studied. Kenison (1986), McFadden (1986), Faro-Schroeder (1995), and Penny (1996) used student grades in remedial/developmental mathematics as the outcome. Fleming (2003) and Shalyefu (2004) used student grades in Beginning/Elementary Algebra as the outcome. Baxter and Smith (1998), Sandruck (2003), Fike (2005) as well as Fike and Fike (2007) used student grades in Intermediate Algebra as the outcome. Stephens (2005) used student grades in Elementary and Intermediate Algebra as the outcome.

Lawrence (1988) defined outcomes as a dichotomy (i.e., earning a grade of A or B in Basic Algebra was successful, whereas earning a C, D, or NC was considered unsuccessful). Long (2003) also defined outcomes as a dichotomy; earning a grade of A, B, C, or pass was successful while earning a D, F, W, or making progress but must re-enroll was considered unsuccessful. Echenique (2007) also used dichotomous outcomes; earning a grade of A, B, or C in developmental math was successful, while earning a D, F, W, X, or AU was unsuccessful. Another type of dichotomy was studied by Sundeen (2000, p. v)

who looked at outcomes in two ways: first, "the normalized difference between pretest and posttest scores," and second, the math course retention rate.

In addition to studying final grade as an outcome, Sandruck (2003) and Shalyefu (2004) also viewed outcome as a dichotomy. Sandruck (2003) viewed success as students earning an A, B, or C; otherwise, they were counted as unsuccessful. Shalyefu viewed students whose final course grade was 80% or above as successful, whereas students whose final course grade was less than 80% or who dropped out were unsuccessful.

Bershinsky (1993) defined outcomes in remedial math as a trichotomy completion (with an A, B, C, or S), unsuccessful completion (with a D, F, or U), or noncompletion (due to dropping or withdrawing). Similarly, Autrey (1998) defined outomes in developmental math as a trichotomy: completion (with an A, B, or C), nonsuccessful completion (with a D or F), or noncompletion (with a W or F due to nonattendance). Another type of trichotomy was used by Marwick (2002, p. iii) who analyzed outcomes in terms of math course completion, math course grade, "and persistence to enroll in a subsequent mathematics course." Yet another type of trichotomy was utilized by Summerlin (2003) who used final class grades, the Texas Assessment of Skills Program test score, and success in the first college math course after remediation.

Haehl (2007) defined outcome as a trichotomy for her multinomial regression. Two of the three categories involved grades. The three categories were Pass (grade of A, B, or C), Fail (grade of D or F), and withdrawal from the course. She also used a dichotomy in which earning a final grade of A, B, or C

was defined as pass whereas earning a D or F was defined as fail. Incompletes were removed from Haehl's study.

Standardized test scores were utilized as outcomes in several studies. The Descriptive Tests of Mathematical Skills in Elementary Algebra Skills scores served as Betshahbazadeh's (2001) dependent variable. Krzemien (2003) used the pretest-posttest gain score of the Asset Numerical Skills Test. For Huber (2006), the Asset Test score for Algebra I was the dependent variable that defined outcomes.

Roueche and Roueche (1999) were proponents of using departmental test scores to determine student outcomes. Walters (2003) used the final exam score on departmental tests in developmental math classes as her dependent variable. Carter (2004) utilized the pretest-posttest (both departmental tests) gain score in Basic Math as her dependent variable. In studies by Fleming (2003) and Keleher (2005), the final departmental exam score in Beginning/Elementary Algebra was the dependent variable. The final departmental exam score in Intermediate Algebra was the dependent variable for Spradlin and Ackerman (2010).

#### Potential Predictors of Student Outcomes

Various criteria have been used as potential predictors of student outcomes. These criteria fall into three categories as follows: student-level, instructor-level, and classroom-level predictors. They will be considered in order.

#### Student-level Potential Predictors

Student-level predictors are those predictors that are specific to an individual student. These included student gender, ACT Math score, SAT Math score, college, math placement/pretest score, and supplemental instruction. *Student Gender* 

Depending on the study, student gender was either a significant or nonsignificant predictor of developmental mathematics achievement. In studies by Hudson, McPhee, and Petrosko (1993), Barker (1994), Mitchell (1999), Little (2002), Fike (2005), Echenique (2007), Fike and Fike (2007), Donovan and Wheland (2008), as well as Spradlin and Ackerman (2010), females achieved higher developmental scores in mathematics than males.

Conversely, Shalyefu (2004) and Knowlton (2011) found females achieved lower developmental scores in mathematics achievement than males. MdFadden (1986), Goolsby, Dwinell, Higbee, and Bretscher (1988), Lawrence (1988), Burgess (1992), Bershinsky (1993), Penny (1996), Penny and White (1998), Hutson (1999), Sundeen (2000), Hoyt and Sorensen (2001), Krzemien (2003), Sandruck (2003), Summerlin (2003), Shonkwiler (2004), Keleher (2005), Huber (2006) and Taylor (2006) discovered gender to be a non-significant predictor of developmental mathematics achievement.

Walker and Plata (2000, p. 25) wrote, "Fewer than the expected number of males failed fundamental math and elementary algebra. Females failed fundamental math at less than the expected rate but failed elementary and intermediate algebra in greater than expected numbers." Echenique (2007)

calculated that females (70.3%) were more than twice as successful as males (29.7%) in successful completion of their developmental mathematics course. Similarly, Haehl (2007, p. 64) cited in Basic Math, "females were almost twice as likely as males to pass versus fail," but gender was not a significant predictor of achievement in Introductory Algebra.

The majority gender plays a role in developmental math. Based on her classroom visits, Waycaster (2002) observed that the majority gender had higher class participation. For example, if a class had a majority of females, the females had higher class participation.

Another relationship involves majority gender and course level. Long (2003) noted an inverse relationship between gender majority and course level. For example, females comprised 70%, 62%, 54%, and 46% of Arithmetic, Prealgebra, Beginning Algebra, and Intermediate Algebra classes, respectively. *ACT Math Score* 

Bershinsky (1993), Baxter and Smith (1998) as well as Hutson (1999) showed ACT Math score to be a non-significant predictor of success. Hudson, McPhee, and Petrosko (1993, p. 10) quoted that "although male students had significantly higher ACT mathematics and placement test scores, female students earned higher grades in mathematics courses." Stephens (2005), recorded no significant correlation between a student's final grade in Elementary Algebra and ACT Math score. However, he found a significant correlation (.322 with p < .001) between a student's final grade in Intermediate Algebra and ACT

Math score. Donovan and Wheland (2008) revealed ACT Math score to be significant.

We have found that the ACT Mathematics and COMPASS Domain I (Algebra) Placement scores both correlate well with success in the Intermediate Algebra course and that, although females have lower placement test scores than males, they have a higher success rate in the course. (Donovan & Wheland, 2008, p. 2)

Shonkwiler (2004) used the College Board's concordance table to convert SAT Math scores to ACT Math scores. She acknowledged ACT Math score to be a significant predictor (p = .01) of student grade in first college-level math course after completing developmental math. Stephens (2005) converted SAT scores to "equivalent" ACT scores; how the conversion was made was not disclosed. Dorans (1999, p. 14), principal measurement specialist at Educational Testing Service, states that even though there is a correlation between the ACT Math and the SAT Math, and a concordance has been developed, they "should not be used interchangeably." He contends that different skills are being tested in each standardized test. For example, the SAT Math covers reasoning in arithmetic, algebra, and geometry, while the ACT Math covers pre-algebra, elementary algebra, intermediate algebra, coordinate geometry, plane geometry, and trigonometry.

#### SAT Math Score

Kenison's (1986) private college and Lawrence's (1988) university studies ascertained SAT Math score to be a significant predictor of developmental

mathematics student achievement. In their regression analyses, both used grades as their dependent variable; Kenison's (1986) level of significance was p = .0219, while Lawrence's (1988) level of significance was p < .01.

McFadden (1986) concluded that SAT Math score was not a significant predictor of developmental math course grade. Keleher (2005) found that SAT Math score was not a significant predictor of final exam score in Elementary Algebra. The SAT Math score was used for placement into Intermediate Algebra in Taylor's (2006) study, rather than being one of her independent variables.

In Howland's study comparing self-paced, small classes, and large classes, he converted ACT scores to SAT scores using a conversion chart "prepared by the Chief of Naval Education and Training, United States Navy, Pensacola, Florida" (1991, pp. 71-72). Again, Dorans (1999) at the Educational Testing Service states that ACT and SAT scores are not interchangeable. *College* 

Long's (2003) data source included major, but major was not one of her independent variables. Taylor's (2006) study included participants from three colleges and two universities. The college students formed the control group which received traditional lecture instruction, and the university students formed the experimental group which received self-paced computer instruction. However, Taylor (2006) did not compare the students within their respective institutions.

#### Math Placement/Pretest Score

Several researchers used math placement tests or math pretests as their predictor variables. Baxter and Smith (1998) as well as Hutson (1999) indicated that placement test scores were non-significant predictors of developmental mathematics students' achievement. Conversely, studies by Lawrence (1988) and Little (2002) have shown in-house placement test scores to be significant predictors of students' achievement. Similarly, Shalyefu (2004) has shown inhouse pretest scores to be significant predictors of students' achievement. In her study, Carter (2004) used pretest-posttest gain score but demonstrated no significant difference in her control group, which was a traditional lecture course, and her experimental group, which combined lecture with ALEKS software.

Little's (2002) in-house placement test, known as the Algebra Basic Skills Test (ABST) had 25 questions. This multiple-choice test had four choices and was administered on the first day of class. The results of the test could be used to change placement. Little (2002, pp. 8-9) makes the following summary, "Given that the majority of the students enrolled in Introductory Algebra for the Fall semester of 2001 took the computerized in-house placement test, it can serve as a common measure for prior mathematics achievement for these students." *Supplemental Instruction* 

The literature reports that institutions often utilize some form of supplemental instruction in their classes. For example, Penny (1996) suggested including student support services as an independent variable to explain differences in student achievement. It was also asserted that andragogy which

included discussion, out-of-class tests, as well as use of calculators, manipulatives, and "...discovery-interactions with the instructor..." (Baxter & Smith, 1998, p. 278) were significant predictors for Intermediate Algebra grades. Little (2002, p. 86) said, "Quantity of instruction should not only be measured by the amount of time in class, but also by the time spent out of class in computer labs, tutoring sessions, study skills courses, or other intervention programs" in her suggestions for future research.

According to Casazza and Silverman (1996, p. 109), "...when tutoring is delivered by trained tutors, it is the strongest correlate of student success; however, when tutors are not trained, there is no correlation with academic performance." Webster (2005, p. v) stated "Students who participated in tutoring at the math tutoring center had much closer to an equal chance of passing the course than students who did not participate in interventions." Conversely, Barker (1994, pp. x-xi), proposed that studying in a math lab and viewing math videos "were associated with lower achievement."

# Instructor-level Potential Predictors

Instructor-level predictors are those factors that are specific to a particular instructor. These include instructor gender, instructor employment status, instructor's use of music in class, and instructor requiring use of computer software.

### Instructor Gender

Penny (1996), Penny and White (1998), and Hewitt (2001) accounted for female instructors' positive effect on the final course grades of developmental mathematics students. Although Waycaster (2002) observed that the majority student gender had higher class participation (i.e., if a class had a male majority, the males had higher class participation) she observed two exceptions. In one class, there were a majority of males with a female instructor and a higher participation by female students. In another class, there were a majority of females with a male instructor and a higher participation by male students. *Instructor Employment Status* 

In studies in which instructor employment status was used as a predictor of developmental mathematics students' success, Burgess (1992) and Penny (1996) contended that part-time instructors had a significant positive effect on developmental mathematics student achievement whereas Fike (2005) as well as Fike and Fike (2007) named employment status to be a non-significant predictor of developmental mathematics students' success.

Maxwell (1997, p. 271) stated that "although students earn higher grades from part-time instructors who teach remedial courses, they do less well in mainstream math courses than those who complete remedial courses taught by full-time math teachers." Conversely, Hewitt (2001) disclosed that although developmental mathematics students earned a greater percent of passing grades (A, B, or C) from their adjunct instructors, they do the same in college-level mathematics courses compared to the developmental mathematics students taught by full-time mathematics instructors.

### Music

It has been suggested that listening to music may help students perform better on math tests. Betshahbazadeh (2001) compared students who listened to 15 minutes of Mozart's music before a standardized exam, students who listened to 15 minutes of Tejano music before a standardized exam, and students who did not listen to music before a standardized exam. The standardized exam was "The Descriptive Tests of Mathematics Skills in Elementary Algebra Skills from the College Board's Multiple Assessment Programs and Services" (Betshahbazadeh, 2001, p. iv). He surmised that listening to music before an exam did not impact performance.

Walters (2003) compared the final exam results of students who listened to music during the final exam to students who did not listen to music during the final exam. She too showed that there was no difference between the experimental and control groups.

### Computer Software Use

Krzemien (2003) pointed out that students who had a lecture-discussion Basic Arithmetic course had a significantly higher gain score on their pretestposttest Asset Numerical Skills Test than students who had only computer-based instruction. Conversely, Sandruck (2003, p. 94) made the distinction that 78% of only Computer-Based Instruction students passed Intermediate Algebra compared to 66% of Teacher-Led Instruction; however, "this difference was not statistically significant for this sample."

The ALEKS Web site defines ALEKS: "Assessment and LEarning in Knowledge Spaces is a Web-based, artificially intelligent assessment and learning system (What is ALEKS?, 2010)." For Beginning Algebra students, Fleming (2003) denoted no statistical difference between grades for Beginning Algebra students who learned with ALEKS and those who learned in a traditional lecture format. However, when final exam scores were used as the dependent variable, Fleming (2003) noticed that students experiencing Beginning Algebra with traditional instruction earned higher final exam scores compared to students using ALEKS as their mode of instruction.

For Intermediate Algebra students, Carter (2004) as well as Spradlin and Ackerman (2010) agreed that there was no statistically significant difference between students in the lecture group and students who received lecture and computer-assisted instruction outside of class. Taylor (2006) presented no statistical difference between Intermediate Algebra students who learned with ALEKS and Intermediate Algebra students who learned in a traditional lecture setting.

### Potential Classroom-level Predictors

Classroom-level predictors are those predictors that are specific to the classroom and are part of the learning environment. These included class time of day, number of class meetings, and class size. Each will be considered separately.

### Class Time of Day

Although the focus of the study by Burgess (1992) was adjunct instructors compared to full-time instructors, he also studied classroom factors such as daytime students compared to nighttime students. He argued that, compared to full-time instructors, adjunct instructor daytime Introductory Algebra students had a greater chance of completion and had a greater chance of taking Intermediate Algebra. Conversely, Sandruck (2003) designated a negative correlation between daytime students and final grade in Intermediate Algebra.

Although daytime sections and nighttime sections were mentioned by Stephens (2005), data on this variable were not collected. In his recommendations for future research, he suggested using time of day as a factor. Fike (2005) also made specific mention of adding class time of day to the predictors in his suggestions for further research.

#### Number of meetings per week

Fike (2005) considered the number of class meetings to be a significant predictor of student achievement. In his study, Intermediate Algebra students who met once a week for a total of 150 minutes performed better than students who met twice a week for 75 minutes each time. Conversely, Sundeen (2000) examined no difference in student achievement between students who had a class that met twice a week compared to students who had a class that met three times per week. In a study by Sandruck (2003), day classes met two, three, or four times per week while evening classes met twice a week. However her predictor was time of day rather than number of class meetings.

According to Dr. Hunter Boylan (personal communication, July 10, 2000), Director of the National Center for Developmental Education, "A class that meets five days per week will have a higher absence rate than a class that meets three days per week. This situation could be corrected with a well-enforced attendance policy." These statements are based on Boylan's visits to college campuses and his interpretations of their statistical data. Conversely, keeping course credit hours constant at five credit hours, Waycaster (2002) observed classes meeting two, three, four, and five days per week; class meetings that included a break at the halfway point had poor attendance after the break.

Carter (2004) mentioned that one control and one treatment group met three days a week while another control and treatment group met two days a week. However, she did not use number of class meetings as one of her independent variables in her study of supplementing lecture with ALEKS software.

### Class Size

Howland's (1991) study compared three groups: self-paced/self-study, small (less than 30 students) classes, and large (between 200 and 300 students) classes and found significant differences between total credit hours earned to date and overall GPA. The self-paced/self-study students were the worst performers who, on average, completed 13.1 credit-hours and earned a 0.857 GPA. On average, students in large classes completed 19.8 credit-hours and earned a 1.555 GPA, while students in small classes completed 29.4 credit-hours and earned a 1.598 GPA.

Baxter and Smith (1998) as well as Little (2002) were in agreement on class size as non-significant. Baxter and Smith's (1998) study involved four groups: small (32 students) traditional class, large (100 students) traditional class, small (32 students) class with daily quizzes and 10 hourly exams outside of class, and a large (240 students) class which included a workshop with discovery learning using calculators and manipulatives. Little (2002) reported a mean class size of 34 with a standard deviation of 7. Her minimum class size was 19 while her maximum was 50.

Smith, O'Hear, Baden, Hayden, Gorham, Ahuja, and Jacobsen (1996) made the following recommendation about class size.

The bigger the class, the less the opportunity to engage students, the greater the opportunity to use the lecture method, the easier for students to remain loners. Sections of 20 are far more likely to produce significant involvement than sections of 40. (p. 41)

Although Waycaster (2002) observed class size ranging from 12 – 30 students, she did not report on the success of these students in relation to the size of their class.

### **Regression Analyses**

Several researchers (see Appendix D) have used regression to study achievement predictors in developmental mathematics. The simplest regression is linear regression, which involves fitting a line to a set of data so that the equation of the line will predict the outcome. According to Allen and Bennett (2010, p. 177), the purpose of multiple regression is, "To examine the linear

relationship between one continuous criterion (or dependent) variable, and two or more predictor (or independent) variables. Predictor variables can be either continuous or dichotomous." Logistic regression is a type of multiple regression. Binary logistic regression is used when there are two possible outcomes (such as pass or fail), and multinomial logistic regression is used when there are more than two possible categorical outcomes such as the trichotomies that were discussed at the beginning of this chapter: pass, fail, and fail due to nonattendance.

According to Allen and Bennett (2010, p. 185) in multiple regression, " $R^2$  represents the proportion of variance in the criterion that can be accounted for by the predictor variables in combination." The higher the value of  $R^2$ , the better the multiple regression equation is at modeling the relationship between the significant independent variables and the dependent variable.

In Kenison's (1986) study of a private nonprofit co-ed business college,  $R^2 = 17\%$ . This means that 83% of the variation in remedial mathematics grades cannot be explained by high school grade point average (p = .0096) and SAT Math scores (p = .0219), which were significant independent variables. She suggested exploring other predictors such as age, self-discipline, determination, gender, motivation, learning style, and attitude toward mathematics to explain the variation.

In Lawrence's (1988) study of university students,  $R^2 = 25\%$ . This means that 75% of the variation in Basic Algebra grades cannot be explained by the significant (*p* < .01) independent variables of placement test scores, SAT Math

scores, and high school grade point average. Although she gave recommendations for further study, such as replicating her study, she gave no suggestions for additional predictor variables.

In Penny's (1996) study of four-year institutions,  $R^2 = 16\%$  means teacher attributes predict students' performance in developmental mathematics. This means that 84% of the variation in grades cannot be explained by the significant independent variables of employment status and teacher gender. Penny (1996) also found  $R^2$  to be 21% for student attributes predicting students' performance in developmental mathematics, meaning that 79% of the variation in grades cannot be explained by the significant independent variables of race, age, and enrollment status. In the conclusion of the study, Penny (1996) states

The unexplained variance in the model suggests that there are other variables that were not included in the model that may explain the variance in students' performance in developmental mathematics...a revised model could include variables such as student preparation, goals and institutional commitments and characteristics, faculty-student ratio, and student support services. These variables may account for the unexplained variance. (pp. 72-73)

In Little's (2002) study of urban community college students,  $R^2 = 40\%$ . This means that 60% of the variation in introductory algebra grades cannot be explained by the significant (*p* < .01) independent variables of cumulative college grade point averages, math prerequisite status, in-house placement test, student gender, math attitude, race, or instructor. She suggested exploring other

independent variables such as high school math course grade point average, high school class rank, high school grade point average, learning styles, learning environments, interaction between students, interaction between students and instructor, interaction between students and the institution, instructor andragogy, instructor grading policies, instructor qualifications, and instructor experience. Furthermore, she suggested using departmental exams as the dependent variable because "…more standardized grading practices might help alleviate some of the inconsistency in grading procedures and some of the research limitations associated with relying on final grades as the dependent variable" (p. 88)

For first college-level math course after completing developmental math, Shonkwiler (2004) found R<sup>2</sup> to be 17%. This means that 83% of the variation in final grades cannot be explained by the significant independent variables of Math 1 grade (p = .01), ACT Math score (p = .01), and high school grade point average (p = .004).

In Fike's (2005) study of community college students,  $R^2 = 5\%$ . This means that 95% of the variation in Intermediate Algebra grades cannot be explained by his independent variables of number of class meetings (*p* = .031), race (*p* = .029), gender (*p* = .036), or age (*p* < .001).

Stephens (2005) used stepwise multiple regression in his study of state university students. For Elementary Algebra,  $R^2 = 25.6\%$  means that 74.4% of the variation in Elementary Algebra grades cannot be explained by his independent variables, which are the number of high school math classes at or

above Algebra I and overall high school grade point average. For Intermediate Algebra,  $R^2 = 17.2\%$ , means 82.8% of the variation in Intermediate Algebra grades cannot be explained by his significant independent variables, which are ACT Math score and overall high school grade point average.

Haehl (2007) used multinomial logistic regression, so R<sup>2</sup> is not reported. For Basic Math, Haehl (2007) found that 7% of the total variance in Basic Math grade was attributed to Compass Reading score, Compass Math score, ethnicity, and gender. For students in Introductory Algebra, 19% of the variance in the Introductory Algebra grade could be predicted by Basic Math grade.

The eight researchers above surmised that independent variables were missing from their analyses. They thought model specification error was due to lacking appropriate predictors to explain some of the variance.

### Summary

The researcher reviewed literature concerning definitions of student outcomes, potential predictors of student outcomes, and types of regression analyses. Student outcomes were defined in the literature several ways. Studies reviewed used grades, dichotomy, trichotomy, completion, persistence, Asset Test score for Algebra I, pretest-posttest gain score of the ASSET Numerical Skills Test, Descriptive Tests of Mathematics Skills in Elementary Algebra Skills, and the final exam score on a departmental test. Potential predictors of student outcomes or independent variables included variables at the student level, the instructor level, and the classroom level. Student-level variables such as gender, ACT Math score, SAT math score, college in which student is enrolled, math

placement/pretest score, and forms of supplemental instruction were discussed. Instructor-level variables of gender, employment status, use of music, and use of computer software were reviewed. Classroom-level variables of time of day, number of meetings per week, and class size were also reviewed. Regression analyses reviewed included linear regression, multiple regression, multinomial logistic regression, and logistic regression.

### CHAPTER 3

### METHODOLOGY

This study utilized an ex-post facto research design (Kerlinger, 1986; Tuckman, 1988). Rather than designing an experiment to match the data, the data were analyzed after the study was designed, in order to determine what impact the predictors had on the outcome.

#### **Course Descriptions**

Elementary and Intermediate Algebra were four credit hour/five contact hour courses because one of the credit hours was considered to be a lab hour that met for two contact hours. Both Elementary and Intermediate Algebra incorporated lecture and cooperative learning activities. Students were assigned homework from their respective textbooks: *Principles of Elementary Algebra with Applications* (Nustad & Wesner, 1991a) and *Principles of Intermediate Algebra with Applications* (Nustad & Wesner, 1991b). In addition, students were required to register for a co-requisite course, ASC 099: Independent Study Skills, a one hour credit/no credit course.

In addition to the co-requisite course, ASC 099, developmental algebra courses also included an in-class lab component. The students were charged a lab fee to cover the cost of the departmentally-developed lab manuals and lab materials. The lab manuals included the syllabi which included course objectives and a weekly course outline, a math study skills section, study sheets on vocabulary and main concepts, hands-on/cooperative learning activities, problem sheets for videos, puzzles, and final exam review questions.

A minimum of five lab activities were chosen for use by the instructors from the departmentally-developed lab manuals. These manuals drew on cooperative learning activities from a similar text by the authors Harry L. Nustad and Terry H. Wesner: *Cooperative Learning for the College Mathematics Classroom: Elementary Algebra* (1996a), and *Cooperative Learning for the College Mathematics Classroom: Intermediate Algebra* (1996b). Both books were written by Signe Kastberg. Data were not collected for the lab variable.

In-class math lab activities included study skills, manipulatives, and games. In addition, the labs included small-group analyses of common student errors, critical thinking, and real-life application problems. Some instructors incorporated technology into their labs by using graphing calculators and computer software.

#### Placement

Students were placed into developmental mathematics classes based on their ACT Math scores. An ACT Math score between 12 and 15 placed a student in Elementary Algebra, while a score between 16 and 18 placed a student in Intermediate Algebra.

### **Student Outcomes**

Although there were common elements to the syllabi for Elementary and Intermediate Algebra, instructors varied in how they computed final grades. The number of tests, as well as their percent of the final grade, differed. A number of hands-on mathematics activities from departmentally-developed manuals comprised between 10 and 15 percent of the final grade. Students were required

to complete a minimum of five activities from the aforementioned manual. Thus, the number of activities, as well as the percent of the final grade, varied. Daily work or quizzes were optional and were worth between zero and five percent of the final grade. Thus, daily work or quizzes, as well as their percent, were optional in computing the final grade. The common comprehensive departmental final exam represented 25% of the student's final grade. Students who earned a 75% or higher final average received Credit (CR); otherwise, they received No Credit (NC). Part of this study's purpose was to examine the impact of the instructors' decisions on the remaining 75% part of students' grades.

Elementary Algebra and Intermediate Algebra at MUCTC were designated as Credit or No Credit. Credit was awarded when a student's overall average was a 75% or above; otherwise, the student earned No Credit. In this study, the final grade was coded as a dummy variable, credit was coded as a 1, and noncredit was coded as a 0. The final exam score was based on a 100 point scale. This provided more detail in analysis than a simple Credit / No Credit system. The final exam score was coded as a number between 0 and 100.

### Potential Predictors of Student Outcomes

The potential predictors of student outcomes investigated were grouped in one of the following categories: student, instructor, or classroom. Each category of predictor is considered in detail below.

### Potential Student-level Predictors

The potential student-level predictors used in this study included student gender, ACT Math score, SAT Math score, college, math pretest score, and ASC grade. Each will be considered separately.

### Student Gender

Student gender, coded as a dummy variable, was gathered from department records. Females were coded as 0 and males were coded as 1.

# ACT Math scores

SAT Math scores were not converted to ACT Math scores. The ACT Math score was coded as a number between 6 and 22. These data were obtained from student records.

# SAT Math Score

ACT Math scores were not converted to SAT Math scores. The SAT Math score was coded as a number between 200 and 530. These data were obtained from student records.

### College

In this study, community college students were compared with students in other colleges from Marshall University. College, coded as a dummy variable, was obtained from department records. Community College students were coded as 1 and University students were coded as 0.

### Math Pretest

The MUCTC pretests were based on the "Pretest of Prerequisite Skills" from the texts *Elementary Algebra: A Prerequisite for Functions* (Abney, Mowers, Calland, & Crowley, 1999), and Intermediate Algebra: An Introduction to Functions Through Applications (Abney, Crowley, Mowers, & Calland, 1999). They were administered on the first day of class. The data were collected to determine if there was a correlation between pretest scores and final exam scores (see Appendix E). The 20 question objectives along with descriptive statistics for both the Elementary Algebra and Intermediate Algebra pretests may be found in appendices F through K. Due to security issues, the pretests are not included in the appendices.

### Supplemental Instruction

MUCTC's ASC course was a form of supplemental instruction. ASC 099: Independent Study Skills was a one hour, Credit/No Credit, co-requisite course for Elementary Algebra as well as Intermediate Algebra. In order for students to receive credit, they had to spend at least 15 hours (approximately one hour per week) in the Academic Skills Center (ASC) during the semester. At the ASC, students watched instructional videos by Elayn Martin-Gay (Videotape series to accompany Nustad and Wesner, Principles of Elementary Algebra with Applications, 2nd ed., 1991, 1990a; Videotape series to accompany Principles of Intermediate Algebra with Applications, 2nd. ed., 1991, 1990b) which accompanied their text, used computer tutorials, worked cooperatively with other students, or received assistance from mathematics tutors.

### Potential Instructor-level Predictors

Potential instructor predictors for student outcomes included gender, employment status, use of Mozart for Your Mind tape, and use of ALEKS software. Each is considered separately below.

#### Instructor Gender

Instructor gender came from department records and was coded as a dummy variable. It was coded the same as student gender, 1 for males and 0 for females.

### Instructor Employment Status

Instructor employment status was taken from department records and was coded as a dummy variable. Full-time instructors were coded as a 0 and adjunct instructors were coded as 1.

### Mozart for Your Mind Tape Use

One instructor played the tape Mozart for Your Mind (1990) before, between, and after classes. Music tape use was coded as a dummy variable. Using the tape was coded as a 1, and not using the tape was coded as a 0.

### ALEKS Software Use

Software use was coded as a dummy variable. Using the software was coded as a 1, and not using the software was coded as a 0.

### Potential Classroom-level Predictors

Potential classroom predictors of student outcomes included class time of day, number of class meetings, and class size. Each will be described separately.

# Time of Day

Although the beginning time of class was in the original data set, in order to simplify the categorical variable, the classes were coded as a dichotomous variable (i.e., a.m. or p.m. classes). Classes starting between 8 a.m. and noon were considered to be a.m. classes and were coded as 1 whereas classes starting between 12:15 p.m. and 6:30 p.m. were considered p.m. classes and were coded as 0.

### Number of Class Meetings

For the Elementary Algebra data set, the number of class meetings was two, four, or five times per week. For the Intermediate Algebra data set, the number of class meetings ranged from two to five times per week.

### Class Size

Class size was coded according to the number of students who took the pretest on the first day of class. Class size ranged from 11 to 44 students.

### **Research Design**

The census of all developmental mathematics students from the fall 2001 semester from MUCTC was used. Additionally, the Elementary and Intermediate Algebra groups were compared for similarities and differences in predicting final exam scores and final grades using stepwise regression and binary logistic regression to analyze potential predictors of student achievement individually, as well as in combinations.

### **Data Collection**

The following student data were collected from student records: gender, ACT Math score, SAT Math score, college in which student was enrolled, and ASC grade. The departmentally-developed pretest and final exam scores were obtained from the SAS reports from Marshall University. The instructor data, gender, and employment status were garnered from departmental records. The classroom data for time of day, number of class meetings, and class size were extracted from the class schedule.

One full-time instructor used the Mozart for Your Mind tape in one of her Elementary Algebra and two of her Intermediate Algebra classes. One full-time, male instructor used ALEKS in two of his Intermediate Algebra classes.

#### Instrumentation

In addition to ACT Math and SAT Math scores, four departmentallydeveloped, multiple-choice assessments were used. Elementary Algebra and Intermediate Algebra each used 20-question pretests which were administered and scored electronically on the first day of class. Therefore, no make-up tests were given which resulted in missing data. Students had no more than 45 minutes to complete the pretest.

The Elementary Algebra final exam had 50 questions, which were drawn from two textbook supplements: the book *Quiz Item File* (Nustad & Wesner, 1987a), the book *Test Item File* (Nustad & Wesner, 1987b) and MUCTC faculty. The Intermediate Algebra final exam had 40 questions which were drawn from the textbook supplement *Test Item File* and *Quiz Item File* (Smith H. M., 1987),

MUCTC faculty, and Marshall University faculty. The questions for the two pretests and the two final exams were multiple choice with four possible answer choices for each question. Because the final exam was approximately twice as long as the pretests, students had two hours to complete the final exam. The final exam was administered during final exam week.

These departmentally-developed pretests and final exams are criterionreferenced (objective-referenced) tests, with scores representing the percent of questions answered correctly. Due to security issues, copies of these assessments are not included in the appendices. The pretest skills being tested for Elementary and Intermediate Algebra are included in Appendices H and K. The Elementary and Intermediate Algebra topics, number of class days, and number of questions on the final exam are included in Appendices P and Q.

### Departmentally-Developed Test Validity

The pretests and final exams possessed criterion validity because the questions were written by math instructors. The Elementary Algebra and Intermediate Algebra pretests and final exams were written by full-time Elementary Algebra and Intermediate Algebra instructors. The Intermediate Algebra final exam also had questions written by Marshall University full-time mathematics faculty. Therefore, they possess content validity. Additionally, content outlines were created for the pretests and final exams, which was another way to ensure content validity. The Elementary Algebra and Intermediate Algebra pretests also have concurrent validity, which relates achievement on the pretest with achievement on the Math portion of the ACT.

# Data Analysis

The data were analyzed using linear regression, stepwise regression, multiple regression, and binary logistic regression using PASW Statistics GradPack (Version 18) (2009). The first and third research questions were analyzed with multiple regression due to the final exam score being a continuous dependent variable. The second and fourth research questions employed binary logistic regression due to the dependent variable (final grade) being a dichotomy. A summary of the coding of the independent variables can be found in Table 2.

Table 2	Independent	Variables
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Predictor	Category	Coded as
Gender	Student	1 for males, 0 for females
ACT Math Score	Student	a number between 6 and 22
SAT Math Score	Student	a number between 200 and 530
College	Student	1 for community college student, 0 otherwise
Math Pretest	Student	a number between 5 and 90
ASC Grade	Student	1 for Credit, 0 for No Credit
Gender	Instructor	1 for males, 0 for females
Employment Status	Instructor	1 for adjuncts, 0 for full time
Mozart Tape	Instructor	1 for tape use, 0 for not using tape
ALEKS Software	Instructor	1 for software use, 0 for not using software
Time of Day	Classroom	1 for a.m., 0 for p.m.
Number of meetings/wk	Classroom	a number between 2 and 5
Class size	Classroom	a number between 11 and 49

A summary of the coding of the dependent variables can be found in Table 3.

Table 3 Student Dependent Variables

Variable	Coded as
Final Exam	a number between 0 and 100
Final Grade	1 for Credit, 0 for No Credit

# **Errors in Hypothesis Testing**

Two types of incorrect decisions occur in hypothesis testing. A Type I error occurs when a true null hypothesis is rejected whereas a Type II error occurs when a false null hypothesis is erroneously accepted. In this study, a Type I error corresponds to acknowledging the independent variables (student, classroom, or instructor) as being predictors of the student's final exam score or final grade when in fact they are not predictors. A Type II error corresponds to not acknowledging the independent variables (student, classroom, or instructor) as being predictors of the student's final exam score, or instructor) as being predictors of the student of the student, classroom, or instructor) as being predictors of the final exam score or final grade when in fact they are.

### Level of significance

The level of significance, as denoted by  $\alpha$ , is the maximum probability of making a Type I error. In this study,  $\alpha$  = .05. This corresponds to a 5% chance of making a Type I error.

### Limitations

Care should be taken when generalizing this study to other institutions such as private institutions or institutions offering a different developmental mathematics curriculum. Students in this study may be more representative of the Appalachian region of the country as opposed to students from urban areas.

Additionally, MUCTC was a community and technical college located on a university campus.

Because the dependent variables in this study included final exam score and final grade, only student records that included these two measures could be used in the multiple regression analysis. These reasons may limit the generalizability of this study.

#### Summary

The data groups are Elementary Algebra and Intermediate Algebra students from MUCTC for the fall 2001 semester. The data were collected from student records, Marshall University SAS reports, departmental records, and class schedules. Students were placed in Elementary Algebra and Intermediate Algebra by their ACT Math and SAT Math scores, although this was not strictly enforced. Students who were not properly placed were informed on the first day of class that they needed to drop the incorrect course and add the correct course. Some students were not able to add the correct class to their schedules, so they stayed in their incorrect placement. The instrumentation included departmentally developed pretests and final exams. Departmentally-developed pretests and final exams exhibit criterion validity, content validity, and concurrent validity. The data were analyzed using PASW Statistics GradPack (Version 18). Type I and Type II errors in hypothesis testing were defined and related to this study. The level of significance was chosen as  $\alpha$  = .05. Limitations of this study are due to missing data. This study analyzed data using multiple regression and binary logistic regression.

### **CHAPTER 4**

### FINDINGS

The major purpose of this study was to analyze specific predictors that impact student achievement in developmental mathematics. The data group used in this study was comprised of 11 sections of Elementary Algebra with a total of 198 students and 28 sections of Intermediate Algebra with a total of 526 students for the fall 2001 semester. The study identified characteristics of developmental mathematics students as related to achievement. Through multiple regression, models were developed to predict the final exam score as well as final grade based on characteristics of developmental mathematics students, their instructors, and classrooms.

# **Research Questions**

Each of the following research questions will be discussed separately in this chapter for Elementary Algebra and for Intermediate Algebra.

### Research Question One

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

# Research Question Two

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

### Research Question Three

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

### Research Question Four

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

### **Discussion of Research Questions**

Each of the four research questions was considered separately for Elementary Algebra and Intermediate Algebra. The research questions are restated along with their results. Descriptive statistics for the predictors and outcomes can be found in Appendix R. Final course grades for Elementary and Intermediate Algebra were broken down in terms of the percent of passing and failing grades. For student gender, the number and percent were given for both courses. A distribution of math ACT scores was given for both groups of students. A comparison was given for number of students who had both SAT and ACT math scores verses the number of students who had only a math SAT score. The number of Elementary and Intermediate Algebra student community college enrollment was presented in terms of percent. Detailed pretest result summaries and descriptive statistics are included in Appendices F through K. ASC final course grades were explained in terms of number and percent of passing and failing grades.

Tables for instructor-level potential predictors compared the number and percent of instructors in terms of gender, employment status, Mozart music use, and ALEKS software use. Potential classroom-level predictors compared the number and percent of sections in terms of a.m., p.m., and self-paced as well as number of class meetings. Additionally, the ranges were given for class size.

# Discussion of Research Question One

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

Because the final exam score is a continuous variable, simple regression was used to analyze the data. Simple regression results with final exam score as the dependent variable for Elementary Algebra is detailed in Table 4. Statistically significant predictors are in bold. Interpretation of the results follows. SPSS output is included in Appendix S.

Predictor	R	$R^2$	F	Constant	В
			(sig.)	(sig.)	(sig.)
Student					
Gender	.035	.001	.204 (.652)	65.582 (.000)	985 (.652)
ACT Math	.238	.057	7.616 (.007)	28.430 (.032)	2.465 (.007)
SAT Math	.067	.004	.036 (.854)	73.473 (.173)	027 (.854)
MUCTC Student	.174	.030	5.223 (.024)	67.792 (.000)	-4.879 (.024)
Pretest	.370	.137	26.489 (.000)	52.349 (.000)	.307 (.000)
ASC Grade	.036	.001	.214 (.644)	66.500 (.000)	-1.547 (.644)
Instructor					
Gender	.147	.022	3.687 (.057)	63.615 (.000)	4.285 (.057)
Adjunct	.192	.037	6.373 (.013)	62.659 (.000)	5.366 (.013)
Mozart Use	.042	.002	.296 (.587)	64.863 (.000)	1.537 (.587)
Classroom					
a.m.	.018	.000	.051 (.821)	65.056 (.000)	.511 (.821)
No. of Meetings	.172	.030	5.121 (.025)	71.618 (.000)	-1.82 (.025)
Class Size	.106	.011	1.896 (.170)	60.972 (.000)	.122 (.170)

Table 4 Elementary Algebra Simple Regression

Simple regression results with final exam score as the dependent variable for Intermediate Algebra is detailed in Table 5. Statistically significant predictors are in bold. Interpretation of the results follows. SPSS output is included in Appendix S.

In each of these simple regressions, there is only one predictor. Thus, R represents the simple correlation between a predictor variable and the final exam score. The value of  $R^2$  accounts for the percent of variation in the final exam score explained by the predictor. For example, in Elementary Algebra, the predictor variable ACT math score, the value of  $R^2$  = .057, means that ACT math score can account for 5.7% of the variation in final exam score. The constants and B values provide the coefficients for the regression equations. Again, using Elementary Algebra, ACT math score predictor, predicted Final Exam Score = 2.465(ACT Math Score) + 28.430. The interpretation of this equation is that for each point that the ACT math score increases, the final exam score increases by 2.465 points.

Predictor	R	$R^2$	nple Regress F	Constant	В
			(sig.)	(sig.)	(sig.)
Student Gender	.149	.022	11.425 (.001)	70.214 (.000)	-4.327 (.001)
ACT Math	.240	.058	28.724 (.000)	18.577 (.047)	2.96 (.000)
SAT Math	.285	.081	5.493 (.022)	30.707 (.065)	.092 (.022)
MUCTC Student	.088	.008	4.018 (.046)	69.108 (.000)	-3.248 (.046)
Pretest	.294	.087	40.966 (.000)	55.026 (.000)	.294 (.000)
ASC Grade	.170	.029	15.179 (.000)	62.583 (.000)	6.942 (.000)
Instructor <b>Gender</b>	.098	.010	4.980 (.026)	69.412 (.000)	-3.006 (.026)
Adjunct	.017	.000	.147 (.701)	68.732 (.000)	479 (.701)
Mozart Use	.124	.015	8.111 (.005)	67.872 (.000)	5.594 (.005)
ALEKS Use	.036	.001	.686 (.408)	68.607 (.000)	-2.561 (.408)
Classroom a.m.	.020	.000	.196 (.658)	68.422 (.000)	.578 (.658)
No. of Meetings	.068	.005	2.380 (.124)	66.176 (.000)	.759 (.124)
Class Size	.085	.007	3.776 (.053)	63.314 (.000)	.202 (.053)

Table 5 Intermediate Algebra Simple Regression

Elementary Algebra and Intermediate Algebra share three statistically significant student predictor variables: ACT math scores, Marshall University Community and Technical College (MUCTC) students, and math pretest. Additionally, Elementary Algebra had instructor employment status of part time and number of class meetings as statistically significant predictors. In addition to the student predictors of ACT math score, MUCTC students, and math pretest, Intermediate Algebra also had student gender, SAT math score, and ASC grade as statistically significant predictors. For Intermediate Algebra instructor employment status was not a statistically significant predictor, but instructor gender and use of Mozart music were statistically significant predictors. Although Elementary Algebra had number of class meetings as a significant predictor, none of the classroom predictors was statistically significant for Intermediate Algebra. The relationship of commonalities and differences between statistically significant predictors in Elementary Algebra and Intermediate Algebra can be best shown via a Venn diagram as shown in Figure 2. Statistically significant predictors are in bold.

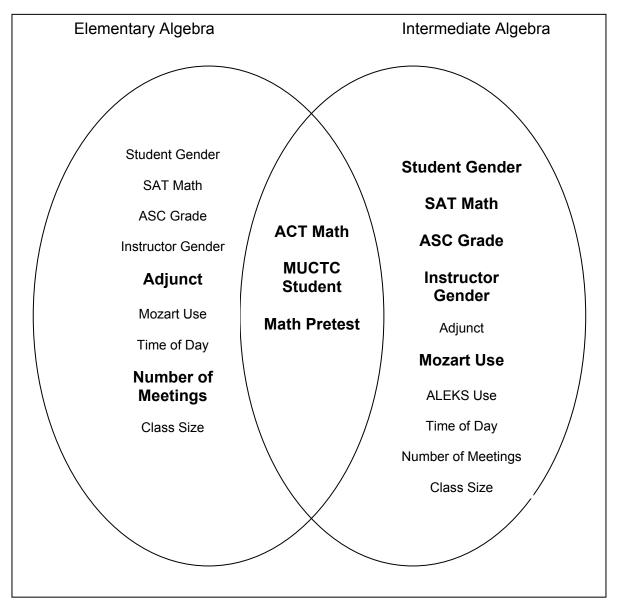


Figure 2 Elementary Algebra and Intermediate Algebra Simple Regression Commonalities and Differences

# Discussion of Research Question Two

When taken individually, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

Because the dependent variable, student's final grade, was pass fail, simple binary logistic regression was used to answer research question two. The results for Elementary Algebra and Intermediate Algebra are detailed in Tables 6 and 7. Statistically significant predictors are in bold. Interpretation of the results follows. Exp (B) is the factor that one multiplies the odds by when the predictor is increased by one. SPSS output is included in Appendix T.

Predictor	В	Sig.	Exp (B)
Student			
Gender	.003	.990	1.003
ACT Math	.327	.031	1.386
SAT Math	.005	.602	1.005
MUCTC Student	332	.213	.717
Pretest	.036	.000	1.037
ASC Grade	2.177	.000	8.818
Instructor			
Gender	.738	.009	2.092
Adjunct	.933	.001	2.541
Mozart Use	-1.063	.003	.345
Classroom			
a. m.	675	.017	.509
Number of Meetings	330	.002	.719
Class Size	001	.902	.999

Table 6 Elementary Algebra Simple Binary Logistic Regression

There is a slightly different interpretation of significant continuous variables compared to significant dichotomous variables; thus, the continuous variables will be interpreted first, followed by the interpretation of the dichotomous variables. For ACT Math score, the odds ratio of 1.386 means that one unit increase in the ACT Math score increases the odds of an Elementary

Algebra student passing the course by 38.6%. For the Pretest score, the odds ratio of 1.037 means that one unit increase in the Pretest score increases the odds of an Elementary Algebra student passing the course by 3.7%. For Number of Meetings, the parameter value of .719 means that one unit increase in the number of class meetings decreases the odds of passing Elementary Algebra by 28.1%.

The interpretation of the dichotomous variables follows. The ASC Grade, shows that the odds of passing Elementary Algebra for a student who earned a passing grade in the ASC co-requisite increased by nearly 9 fold compared to students who failed the ASC co-requisite. The Teacher Gender variable shows that the odds of passing Elementary Algebra for a student taking the class from a male instructor increased by over two times compared to the odds for a student taking Elementary Algebra from a female instructor. The Adjunct faculty variable shows that the odds of passing Elementary Algebra for a student taking the class from an adjunct faculty member increased by over 2.5 times compared to the odds for a student taking the class from a full-time faculty member. The Mozart Use variable shows that the odds of an Elementary Algebra student passing the course for a student who listened to Mozart music in the minutes before and after class decreased by 65.5% compared to the odds for a student who did not listen to Mozart music before and after class. The variable a.m. shows that the odds of passing Elementary Algebra for a student taking the class between 8 a.m. and noon decreased by 49.1% compared to the odds for a student taking the class between12:15 p.m. 6:30 p.m.

Predictor	В	Sig.	Exp (B)
Student			
Gender	746	.000	.474
ACT Math	.448	.000	1.565
SAT Math	.003	.559	1.003
MUCTC Student	857	.000	.425
Pretest	.034	.000	1.034
ASC Grade	2.662	.000	14.321
Instructor			
Gender	.182	.303	1.199
Adjunct	206	.210	.814
Mozart Use	424	.092	.654
ALEKS Use	460	.146	.632
Classroom			
a. m.	.002	.991	1.002
Number of Meetings	.063	.345	1.065
Class Size	.011	.413	1.011

Table 7 Intermediate Algebra Simple Binary Logistic Regression

As mentioned previously, there is a slightly different interpretation of significant continuous variables compared to significant dichotomous variables, thus, the continuous variables will be interpreted first, followed by the interpretation of the dichotomous variables. For ACT Math score, the odds ratio

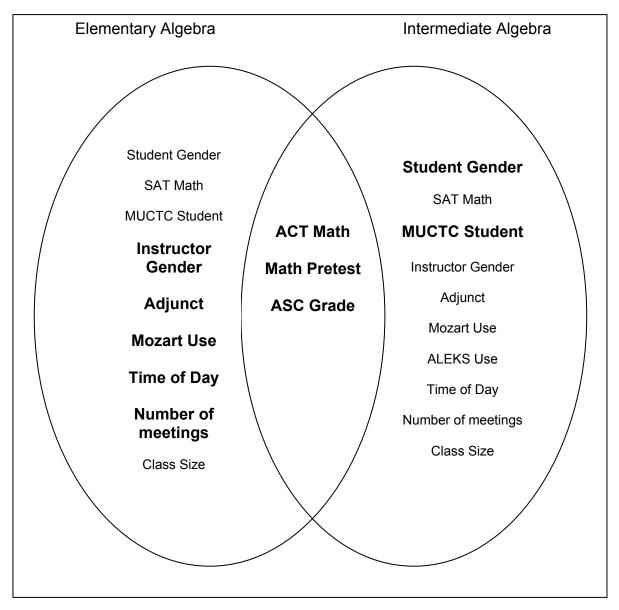
of 1.565 means that one unit increase in the ACT Math score increases the odds of an Intermediate Algebra student passing the course by 56.5%. For the Pretest score, the odds ratio of 1.034 means that one unit increase in the Pretest score increases the odds of an Intermediate Algebra student passing the course by 3.4%.

The interpretation of the dichotomous variables follows. The Student Gender variable shows that the odds of passing Intermediate Algebra for a male student decreased by 52.6% compared to the odds of passing Intermediate Algebra for a female student. The MUCTC Student variable shows that the odds of passing Intermediate Algebra for community college students decreased by 57.5% compared to the odds of passing Intermediate Algebra for other students attending Marshall University. The ASC grade, shows that the odds of passing Intermediate Algebra for a student who earned a passing grade in the ASC corequisite increased more than 14 times compared to students who failed the ASC co-requisite.

Elementary Algebra and Intermediate Algebra share three statistically significant student predictor variables: ACT Math score, math pretest, and ASC grade. Elementary Algebra had five additional significant predictors: the instructor characteristics of instructor gender, instructor employment status, instructor's use of Mozart music, and the classroom characteristics of classes meeting between 8 a.m. and noon, and number of class meetings. Intermediate Algebra had two additional statistically significant student predictors: student gender and MUCTC student. The relationship of commonalities and differences between statistically

significant predictors in Elementary Algebra and Intermediate Algebra can be best shown via a Venn diagram as shown in Figure 3. Statistically significant predictors are in bold.

Figure 3 Elementary Algebra and Intermediate Algebra Simple Binary Logistic Regression Commonalities and Differences



#### Discussion of Research Question Three

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final exam score, and to what extent?

To obtain a multiple regression equation, statistically significant predictors for simple regressions for Elementary Algebra and Intermediate Algebra were tried in combinations. For Elementary Algebra, the statistically significant predictors from the simple regression were added to the model starting with the two most statistically significant predictors of math pretest and ACT math score. The results are in Table 8. Statistically significant predictors are in bold.

Table 8 Elementary Algebra Multiple Regression with Math Pretest and ACT Math Score

R	R <sup>2</sup>	F	Constant	Pretest B	ACT B
		(sig.)	(sig.)	(sig.)	(sig.)
.377	.142	10.415	26.810	.264	1.844
		(.000)	(.034)	(.001)	(.037)

These results yield the following equation:

predicted final exam score = .264math pretest + 1.844ACT math score + 26.810. The interpretation of this equation is that holding the ACT math score constant, each point the math pretest increases, the final exam score increases by .264 points. Similarly, holding the math pretest score constant, for each point the ACT math score increases, the final exam score increases by 1.844 points. The value of  $R^2$  = .142 means that math pretest and ACT math score account for 14.2% of

the variation in final exam score, which is an improvement over the simple regression models.

Next the instructor employment status predictor was added to the two other statistically significant predictors of math pretest and ACT math score. The results are in Table 9. Statistically significant predictors are in bold. Table 9 Elementary Algebra Multiple Regression with Math Pretest, ACT Math Score and Adjunct

R	R <sup>2</sup>	F	Constant	Pretest B	ACT B	Adjunct B
		(sig.)	(sig.)	(sig.)	(sig.)	(sig.)
.430	.185	9.457	19.474	.232	2.260	6.100
.400	.100	(.000)	(.124)	(.002)	(.010)	(.011)

These results yield the following equation:

predicted final exam score =

.232math pretest + 2.260ACT math score + 6.100adjunct + 19.474. The interpretation of this equation is that holding the ACT math score and Adjunct constant, each point the math pretest increases, the final exam score increases by .232 points. Similarly, holding the math pretest score and adjunct predictors constant, for each point the ACT math score increases, the final exam score increases by 2.260 points. Finally, holding the math pretest and ACT math scores constant, students who had an adjunct instructor had an increase of 6.100 points in their final exam scores. The value of  $R^2$  = .185 means that math pretest, ACT math score, and instructor employment status account for 18.5% of the variation in final exam score, which is an improvement over both the simple regression models and the multiple regression equation with two predictor variables.

Adding the community college predictor to the three statistically significant predictors of math pretest, ACT math score, and adjunct, did not yield a statistically significant result. Adding the number of class meetings to the three statistically significant predictors of math pretest, ACT math score, and adjunct, did not yield a statistically significant result. The complete SPSS output for these multiple regressions can be found in Appendix U.

The multiple regression assumptions are discussed in Appendix U. In summary, the sample sizes were large enough because for each regression N > 50 + 8k, where k is the number of predictors. Although in each case the maximum Mahalanobis distance was greater than the critical chi-square value for the degrees of freedom equal to the number of predictors at  $\alpha = .001$ , the maximum Cook's distance was always less than one meaning outliers should not be a concern. Tolerance and VIF (variable inflation factor) are used to measure multicollinearity (high correlations between predictor variables). In each case the Tolerance was < 0.1 and VIF (the inverse of Tolerance) was < 5 which means that multicollinearity was not a problem. For each multiple regression, the Normal P-P Plot of Regression Standardized Residuals had points that clustered fairly close to the line which means that the residuals are normally distributed. For each case, the scatterplot of Regression Standardized Residual against

Regression Standardized Predicted Value showed no pattern, which means the assumptions of normality, linearity, and homoscedsaticity of the residuals has been met.

For Intermediate Algebra the statistically significant predictors from the simple regression were added to the model starting with the two most statistically significant predictors of math pretest and ACT math score. The results are in Table 10. Statistically significant predictors are in bold.

Math Score

Table 10 Intermediate Algebra Multiple Regression with Math Pretest and ACT

R	$R^2$	F	Constant	Pretest B	ACT B
		(sig.)	(sig.)	(sig.)	(sig.)
.342	.117	25.902	20.067	.271	2.136
		(.000)	(.050)	(.000)	(.001)

These results yield the following equation:

predicted final exam score = .271math pretest + 2.136ACT math score + 20.067. The interpretation of this equation is that holding the ACT math score constant, for each point the math pretest increases the final exam score increases by .271 points. Similarly, holding the math pretest score constant, for each point the ACT math score increases, the final exam score increases by 2.136 points. The value of  $R^2$  = .117 means that the math pretest and ACT math score account for 11.7% of the variation in final exam score, which is an improvement over the simple regression models.

Next the ASC grade predictor was added to the other two statistically significant predictors of math pretest and ACT math score. The results are in Table 11. Statistically significant predictors are in bold.

R	R <sup>2</sup>	F (sig.)	Constant (sig.)	Pretest B (sig.)	ACT B (sig.)	ASC B (sig.)
.377	.142	21.389	15.716	.273	2.059	6.407
		(.000)	(.126)	(.000)	(.001)	(.002)

Table 11 Intermediate Algebra Multiple Regression with Math Pretest, ACT Math Score, and ASC Grade

These results yield the following equation:

predicted final exam score =

.273math pretest score + 2.059ACT math score + 6.407ASC grade + 15.716.

The interpretation of this equation is that holding the ACT math score and ASC grade constant, each point the math pretest increases, the final exam score increases by .273 points. Similarly, holding the math pretest score and ASC grade predictors constant, for each point the ACT math score increases, the final exam score increases by 2.059 points. Finally, holding the math pretest and ACT math scores constant, students who earned a pass rather than a fail as their ASC grade had an increase of 6.407 points in their final exam scores. The  $R^2 = .142$  means that math pretest, ACT math score, and ASC grade account for 14.2% of the variation in the final exam score, which is an improvement over both the

simple regression models and the multiple regression equation with two predictor variables.

Next the student gender predictor was added to the three other statistically significant predictors of math pretest, ACT math score, and ASC grade. The results are in Table 12. Statistically significant predictors are in bold.

Table 12 Intermediate Algebra Multiple Regression with Math Pretest, ACT Math Score, ASC Grade, and Student Gender

R	R <sup>2</sup>	F (sig.)	Constant (sig.)	Pretest B (sig.)	ACT B ASC B Gender B (sig.) (sig.)
.409	.168	19.029 (.000)	17.532 (.084)	.269 (.000)	2.068 6.158 -3.678 (.001) (.002) (.010)

These results yield the following equation:

predicted final exam score =

.269math pretest + 2.068ACT math score + 6.158ASC grade -3.678student

gender + 17.532.

The interpretation of this equation is similar to the interpretation of the previous two equations. It is interesting to note that holding all variables constant except for student gender, that males' predicted final exam scores will be 3.678 points lower than females' final exam scores (gender is a dichotomous variable in which male was coded as 1). The value of  $R^2$  = .168 means that math pretest, ACT math score, ASC grade, and student gender account for 16.8% of the variation in final exam score. This is an improvement over the previous two regression equations.

Finally the Mozart use predictor was added to the other four statistically significant predictors of math pretest, ACT math score, ASC grade, and student gender. The results are in Table 13. Statistically significant predictors are in bold. Table 13 Intermediate Algebra Multiple Regression with Math Pretest, ACT Math Score, ASC Grade, Student Gender, and Mozart Use

R	R <sup>2</sup>	F	Constant	Pretest B	ACT B	ASC B	Gender B	Mozart B
.423	.179		16.521	.267	2.101	6.062	-3.574	4.862
		(.000)	(.102)	(.000)	(.001)	(.003)	(.012)	(.021)

These results yield the following equation:

predicted final exam score = .267math pretest + 2.101ACT math score +

6.062ASC grade - 3.574student gender + 4.862Mozart use + 16.521.

The interpretation of this equation is similar to the interpretation of the previous three equations. It is interesting to note that holding all variables constant except for Mozart use means that students who listened to Mozart music before and after class had predicted final exam scores that were 4.862 higher than students who did not listen to Mozart music before or after class. It is important to note that there were only 27 cases for this equation. The value of  $R^2 = .179$  means that math pretest, ACT math score, ASC grade, student gender, and Mozart use account for 17.9% of the variation in final exam score, which is an improvement over the multiple regression equation with four predictor variables.

Adding the SAT math score to the five statistically significant predictors of math pretest, ACT math score, ASC grade, student gender, and Mozart use did not yield a statistically significant result. Adding the instructor gender to the aforementioned five statistically significant predictors did not yield a statistically significant result. Finally, adding the community college predictor to the five statistically significant predictors did not yield a statistically significant result. The complete SPSS output for these multiple regressions can be found in Appendix V.

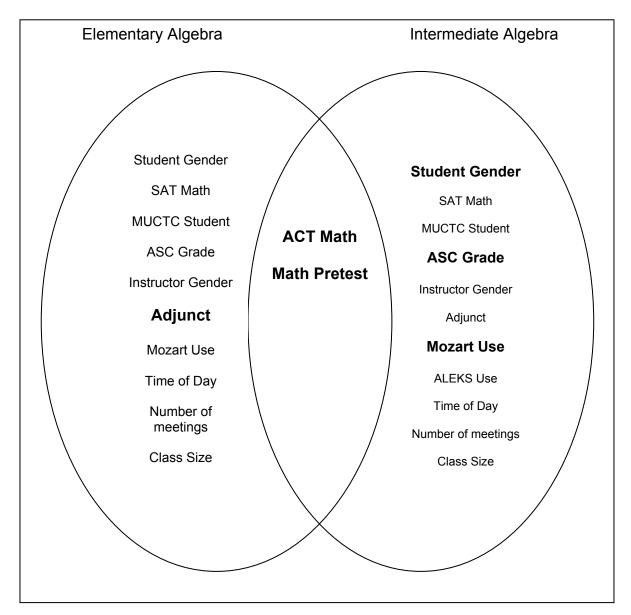
The multiple regression assumptions are discussed in Appendix V. In summary, the sample sizes were large enough because for each regression N > 50 + 8k, where k is the number of predictors. Although in each case the maximum Mahalanobis distance was greater than the critical chi-square value for the degrees of freedom equal to the number of predictors at  $\alpha = .001$ , the maximum Cook's distance was always less than one meaning outliers should not

be a concern. Tolerance and VIF (variable inflation factor) are used to measure multicollinearity (high correlations between predictor variables). In each case the Tolerance was < 0.1 and VIF (the inverse of Tolerance) was < 5 which means that multicollinearity was not a problem. For each multiple regression, the Normal P-P Plot of Regression Standardized Residuals had points that clustered fairly close to the line which means that the residuals are normally distributed. For each case, the scatterplot of Regression Standardized Residual against Regression Standardized Predicted Value showed no pattern which means the assumptions of normality, linearity, and homoscedsaticity of the residuals has been met.

Elementary Algebra and Intermediate Algebra share two statistically significant student predictor variables math pretest and ACT math score. The relationship of commonalities and differences between statistically significant predictors in Elementary Algebra and Intermediate Algebra can be best shown via a Venn diagram as shown in Figure 4. Statistically significant predictors are in bold.

Figure 4 Elementary Algebra and Intermediate Algebra Multiple Regression

Commonalities and Differences



Discussion of Research Question Four

When taken in combinations, which independent variables for student characteristics, instructor characteristics, and classroom characteristics predict the dependent variable of the student's final grade, and to what extent?

To obtain a multiple binary logistic regression equation, the statistically significant predictors for simple binary regressions for Elementary Algebra and Intermediate Algebra were tried in combinations. For Elementary Algebra, the statistically significant predictors from the simple binary logistic regression were added to the model starting with the two most statistically significant predictors of math pretest and ASC grade. The results are in Table 14. Statistically significant predictors are in bold.

Table 14 Elementary Algebra Multiple Binary Logistic Regression with Math Pretest and ASC Grade

Predictor	В	Sig.	Exp (B)
Math Pretest	.048	.000	1.049
ASC Grade	2.439	.000	11.464
Constant	-3.315	.000	.036

For math pretest, the odds ratio of 1.049 means that one unit increase in the math pretest score increases the odds of an Elementary Algebra student passing the course by 4.9%. The ASC grade shows that the odds of passing Elementary Algebra for a student who earned a passing grade in the ASC corequistite increased by 11.464 times compared to students who failed the ASC co-requisite. Next the instructor employment status predictor was added to the two

other statistically significant predictors of math pretest and ASC grade. The

results are in Table 15. Statistically significant predictors are in bold.

 Table 15 Elementary Algebra Multiple Binary Logistic Regression with Math

 Pretest, ASC Grade, and Instructor Employment Status

Predictor	В	Sig.	Exp (B)
Math Pretest	.045	.000	1.046
ASC Grade	2.993	.000	19.952
Instructor Employment Status	1.659	.000	5.255
Constant	-4.398	.000	.014

The interpretation of math pretest and ASC grade are similar to previous statements. The adjunct faculty variable shows that the odds of passing Elementary Algebra for a student taking the class from an adjunct faculty member increased by 5.255 times compared to the odds for a student taking a class from a full-time faculty member.

Next the number of class meetings was added to the three statistically significant variables. The results are in Table 16. Statistically significant predictors are in bold.

Table 16 Elementary Algebra Multiple Binary Logistic Regression with Math Pretest, ASC Grade, Instructor Employment Status, and Number of Meetings						
Predictor	В	Sig.	Exp (B)			
Math Pretest	.049	.000	1.050			
ASC Grade	3.079	.000	21.747			
Instructor Employment Status	1.015	.018	2.760			
Number of Meetings	433	.007	.649			
Constant	-2.577	.003	.076			

The interpretation of math pretest, ASC grade, and instructor employment status are as previously stated. For number of meetings, the parameter value of .649 means that one unit increase in the number of class meetings decreases the odds of passing Elementary Algebra by 35.1%.

Next the Mozart use predictor was added to the four other statistically significant predictors. The results are in Table 17. Statistically significant predictors are in bold.

Mozant Use			
Predictor	В	Sig.	Exp (B)
Math Pretest	.049	.000	1.050
ASC Grade	3.054	.000	21.199
Instructor Employment Status	.87 <b>9</b>	.048	2.408
Number or Meetings	349	.042	.705
Mozart Use	674	.175	.510
Constant	-2.676	.002	.069

Table 17 Elementary Algebra Multiple Binary Logistic Regression with Math Pretest, ASC Grade, Instructor Employment Status, Number of Meetings, and Mozart Use

Note that Mozart use is not statistically significant. The Mozart use predictor was removed from the model.

Next the instructor gender predictor was added to the four statistically significant predictors of math pretest, ASC grade, instructor employment status, and number of class meetings. The results are in Table 18. Statistically significant predictors are in bold.

Predictor	В	Sig.	Exp(B)
Math Pretest	.048	.000	1.050
ASC Grade	3.090	.000	21.974
Instructor Employment Status	1.213	.006	3.365
Number of Meetings	271	.126	.762
Instructor Gender	.949	.025	2.584
Constant	-3.530	.000	.029
Constant	-3.530	.000	.029

Table 18 Elementary Algebra Multiple Binary Logistic Regression with Math Pretest, ASC Grade, Instructor Employment Status, Number of Meetings, and Instructor Gender

Note that number of meetings is not statistically significant so it is removed from the model. The instructor gender variable shows that the odds of passing Elementary Algebra for a student taking the class from a male instructor increased by 2.584 times compared to the odds for a student taking Elementary Algebra from a female instructor.

Next the time of day predictor was added to the model. The results are in Table 19. Statistically significant predictors are in bold.

Predictor	В	Sig.	Exp (B)
Math Pretest	.053	.000	1.055
ASC Grade	3.223	.000	25.095
Instructor Employment Status	1.471	.001	4.352
Instructor Gender	1.229	.002	3.417
Time of Day	871	.025	.418
Constant	-4.623	.000	.010

Table 19 Elementary Algebra Multiple Binary Logistic Regression with Math Pretest, ASC Grade, Instructor Employment Status, Instructor Gender, and Time of Day

The time of day predictor shows that the odds of passing Elementary Algebra for a student taking the class between 8 a.m. and noon decreased by 58.2% compared to the odds for a student taking the class between 12:15 p.m. and 6:30 p.m.

Next ACT math score was added to the model. It was not statistically significant, so it was deleted from the model. SPSS output is included in Appendix W. The assumptions for binomial logistic regression are the same as the assumptions for multiple regression.

For Intermediate Algebra, the five statistically significant predictors from the simple binary logistic regression were all less than .001 so they were all added to the model. The results are in Table 20. Statistically significant predictors are in bold.

Predictor	В	Sig.	Exp (B)
Student Gender	419	.085	.658
ACT Math Score	.288	.010	1.334
Community College Student	604	.038	.547
Math Pretest Score	.027	.003	1.027
ASC Grade	2.527	.000	12.512
Constant	-6.666	.000	.001

Table 20 Intermediate Algebra Multiple Binary Logistic Regression with Student Gender, ACT Math Score, Community College Student, Math Pretest, and ASC Grade

Note that all of the predictors are statistically significant except for student

gender, so it was removed from the model. The results are summarized in Table

21. Statistically significant predictors are in bold.

Math Score, Community College Student, Math Pretest, and ASC Grade						
Predictor	В	Sig.	Exp (B)			
			(			
ACT Math Score	.292	.009	1.339			
Community College Student	681	.018	.506			
Math Pretest Score	.027	.002	1.028			
ASC Grade	2.579	.000	13.183			
Constant	-6.935	.000	.001			

Table 21 Intermediate Algebra Multiple Binary Logistic Regression with ACT

The interpretation of this model follows. Holding all of the predictors constant except for ACT Math Score, the odds ratio of 1.339 means that one unit increase in the ACT Math score increases the odds of an Intermediate Algebra student passing the course by 33.9%. Holding all of the predictors constant except for community college student means the odds of passing Intermediate Algebra for community college students decreased by 49.4% compared to the odds of passing Intermediate Algebra for other students attending Marshall University. Holding all the predictors constant except for math pretest score means the odds of an Intermediate Algebra student passing the course by 2.8%. Holding all the predictors constant except for ASC grade means that the odds of passing Intermediate Algebra for a student who earned a passing grade in the ASC co-requisite increased 13.183 times compared to students who failed the ASC co-requisite.

Elementary Algebra and Intermediate Algebra share two statistically significant student predictor variables: Math Pretest and ASC Grade. The relationship of commonalities and differences between statistically significant predictors in Elementary Algebra and Intermediate Algebra can be best shown via a Venn diagram as shown in Figure 5. Statistically significant predictors are in bold. SPSS output is included in Appendix X. The assumptions for binomial logistic regression are the same as the assumptions for multiple regression.

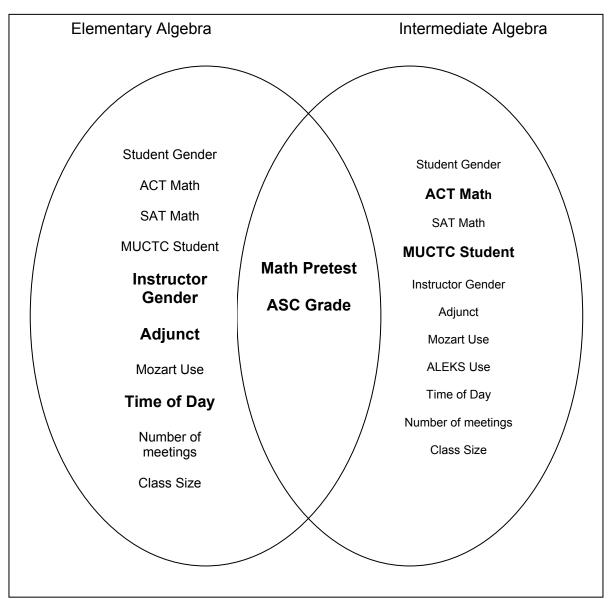


Figure 5 Elementary Algebra and Intermediate Algebra Multiple Binary Logistic Regression Commonalities and Differences

#### Summary

The study of specific predictors that impact student achievement in developmental mathematics revealed the following:

1. When taken individually ( $\alpha$  = .05), the student characteristics of ACT math score, community college students, and math pretest were statistically significant predictors of student's final exam score for both Elementary Algebra and Intermediate Algebra. Additionally, Elementary Algebra had the instructor characteristic of adjunct (p = .013) and the classroom predictor of number of class meetings (p = .025) as statistically significant predictors. Intermediate Algebra had additional statistically significant student predictors of gender (p = .001), SAT math (p = .022), ASC grade (p < .001). Intermediate Algebra had additional statistically significant predictors of gender (p = .001), SAT math (p = .022), ASC grade (p < .001). Intermediate Algebra had additional statistically significant instructor predictors of gender (p = .026) and Mozart music use (p = .005).

2. When taken individually ( $\alpha$  = .05), the student characteristics of ACT math score, math pretest, and ASC grade were statistically significant predictors of student's final grade (pass/fail) for both Elementary Algebra and Intermediate Algebra. Additionally, Elementary Algebra had the instructor characteristics of gender (p = .009), adjunct (p = .001), and Mozart music use (p = .003) as well as the classroom characteristics of time of day (p = .017) and number of class meetings (p = .002) as statistically significant predictors of student's final grade. Intermediate Algebra had additional statistically significant student predictors of gender (p < .001) and community college student (p < .001).

3. When taken in combinations ( $\alpha$  = .05), the student characteristics of ACT math score and math pretest were statistically significant predictors of student's final exam score for both Elementary Algebra and Intermediate Algebra. Additionally, Elementary Algebra had the instructor characteristic of adjunct (p = .011) as a statistically significant predictor of final grade. Intermediate Algebra had two additional student predictors, gender (p = .012) and ASC grade (p = .003), and one additional instructor predictor Mozart music use (p = .021) as statistically significant predictors of student's final grade.

4. When taken in combinations ( $\alpha$  = .05), the student characteristics of math pretest and ASC grade were statistically significant predictors of student's final grade for both Elementary Algebra and Intermediate Algebra. Additionally, Elementary Algebra had statistically significant instructor predictors of gender (p = .002) and adjunct (p = .001) and the classroom predictor of time of day (p = .025). Intermediate Algebra had two additional statistically significant student predictors of ACT math score (p = .009), and community college student (p = .018).

#### **CHAPTER 5**

#### SUMMARY

With the wide range of abilities of community college students, proper course placement is crucial. Therefore, having better predictors of success can help improve placement of students for their achievement. This study analyzed student predictors, instructor predictors, and classroom predictors in relation to student final exam score and student final grade in Elementary Algebra and Intermediate Algebra classes. Student predictors included gender, ACT math score, SAT math score, community college enrollment, math pretest score, and ASC grade. Instructor predictors included gender, employment status, Mozart music use, and ALEKS software use. Classroom predictors included time of day, number of class meetings per week, and class size. A review of the literature was conducted on the two outcome measures of final exam score and final grade as well as the 13 predictor variables. The Elementary Algebra and Intermediate Algebra data sets were analyzed with simple regression, simple binary logistic regression.

The study of specific predictors that impact student achievement in developmental mathematics revealed the following. When analyzed individually, for Elementary Algebra, ACT math score, community college enrollment, math pretest, instructor employment status, and number of class meetings were statistically significant as predictors of final exam scores. When analyzed individually, for Intermediate Algebra, ACT math score, community college enrollment, math pretest, student gender, SAT math score, ASC grade, instructor

gender, and Mozart music use were statistically significant predictors of final exam scores.

When analyzed individually, for Elementary Algebra, ACT math score, math pretest, ASC grade, instructor gender, instructor employment status, Mozart music use, time of day, and number of class meetings were statistically significant predictors of final grade. When analyzed individually, for Intermediate Algebra, ACT math score, math pretest, ASC grade, student gender, and community college enrollment were statistically significant predictors of final grade.

When analyzed in combinations, for Elementary Algebra, ACT math score, math pretest and instructor employment status were statistically significant predictors of final exam score. When analyzed in combinations, for Intermediate Algebra, ACT math score, math pretest, student gender, ASC grade, and Mozart music use were statistically significant predictors of final exam score.

When analyzed in combinations, for Elementary Algebra, math pretest, ASC grade, instructor gender, instructor employment status, and time of day were statistically significant predictors of final grade. When analyzed in combinations, for Intermediate Algebra, math pretest, ASC grade, ACT math score, and community college enrollment were statistically significant predictors of final grade.

#### Conclusions

The inhouse-developed math pretest was the strongest predictor of student math achievement. It was significant individually and in combinations for

Elementary Algebra and Intermediate Algebra for all four research questions involving simple regression, simple binary logistic regression, multiple regression, and multiple binary logistic regression. ACT math score was a strong predictor of student math achievement. It was a statistically significant predictor for both Elementary Algebra and Intermediate Algbebra for the first three research questions involving simple regression, simple binary logistic regression, and multiple regression. In both the simple binary logistic regression and the multiple binary logistic regression ASC grade was a statistically significant predictor for both Elementary Algebra and Intermediate Algebra. For research question one, involving simple regression, the fact that a student was a community college student was a statistically significant predictor of achievement.

Because math pretest was the strongest predictor of final exam score and final grade, instructors should consider administering this test to all students and interpreting the results. For example, students with a high Elementary Algebra pretest score could be moved up to Intermediate Algebra whereas students with a low Intermediate Algebra pretest score could be moved down to Elementary Algebra.

The student's ACT math score was the second strongest preditor. Although Dorans (1999) at the Educational Testing Service states that ACT and SAT scores are not interchangeable, a concordance could be used for students with an SAT score, but no ACT score, to complete a data set.

The ASC grade (credit/no credit) was the third strongest predictor of final grade. Students who passed the ASC co-requisite had odds of passing

Elementary Algebra or Intermediate Algebra varying from 8, 13, 14, or 25 times compared to students who failed the ASC co-requisite. The ASC co-requisite was a form of supplemental instruction in which students spent at least an hour per week in the Academic Skills Center (ASC). At the ASC, students could watch videos that accompanied their texts, receive tutoring, or work in small groups doing homework.

It is interesting to note that community college students in both Elementary Algebra and Intermediate Algebra performed worse on their respective final exams. Elementary Algebra students, on average, scored 4.879 points lower than their university counterparts. On average, Intermediate Algebra students scored 3.248 points lower than other students at the university.

#### **Recommendations for Further Research**

Better predictors are needed for these models. Because the R<sup>2</sup> values were 18.5% and 17.9%, for Elementary and Intermediate Algebra respectively, there are predictors missing that would explain approximately 80% of the outcome measure. Age, high school GPA, financial need, student enrollment type (part-time/full-time), reading ability (as measured by ACT or SAT) are some of the many predictors that could be explored. Another way results could be improved would be to increase the sample size. For example, rather than studying a semester, study an academic year.

# APPENDICES

Appendix A: Letter of Exemption from Institutional Review Board



Office of Research Integrity

June 29, 2010

Linda Hunt 450 Private Drive 10463 Proctorville, OH 45669

Dear Ms. Hunt:

This letter is in response to the submitted abstract concerning your study to analyze factors that impact student success in developmental math. After assessing the abstract it has been deemed not to be human subject research and therefore exempt from oversight of the Marshall University Institutional Review Board (IRB). The Code of Federal Regulations (45CFR46) has set forth the criteria utilized in making this determination. Since the information in this study is a deidentified data set of existing records it is not considered human subject research. If there are any changes to the abstract you provided then you would need to resubmit that information for review and determination.

I appreciate your willingness to submit the abstract for clarification. Please feel free to contact the Office of Research Integrity if you have any questions regarding future protocols that may require IRB review.

Sincerely,

mr. Bruce F. Day, Th.M., CIP

Director Office of Research Integrity

#### WEARE ... MARSHALL

401 11th Street, Suite 1300 • Huntington, West Virginia 25701 • Tel 304/696-7320 A State University of West Virginia • An Affirmative Action/Equal Opportunity Employer Appendix B: Elementary Algebra Syllabus

#### MATH 096 COURSE OVERVIEW SHEET

## COURSE TITLE: DEVELOPMENTAL MATH

#### COURSE NUMBER: MAT 096

#### SEMESTER AND YEAR:

Section:

Time:

## REQUIRED TEXT:

<u>Principles of Elementary Algebra With Applications</u> by Nustad and Wesner, 2nd edition

## REQUIRED MATERIALS: Lab Manual

## **ADDITIONAL MATERIALS:**

## **COMPUTER REQUIREMENTS:**

## **INSTRUCTOR:**

Name:

Office:

Office Hours:

Telephone:

E-Mail Address:

#### **COURSE DESCRIPTION:**

Math 096: A course designed to improve students' skills in: algebraic expressions, integers, fractions, decimals, real numbers, first-degree equations, ratio, proportion, and percent. Emphasis will be placed on skill mastery in preparation for future math courses.

#### CREDITS:

MAT 096 is a four credit hour course which counts toward full-time enrollment status and financial aid eligibility, but does not count toward the number of hours required in any college degree program. The graduation requirement is increased four hours for students who complete this course.

**PREREQUISITES:** Placement in MAT 096 is determined by ASSET/ACT score 12 - 15

#### COREQUISITE:

ASC 099: Independent Study Skills - A 1 hour CR/NC course to be taken concurrently with MAT 096. This consists of at least 15 hours in the Academic Skills Center for the semester.

## LEARNER OUTCOMES/OBJECTIVES

See attached.

#### ASSESSMENT/EVALUATION OF LEARNER OUTCOMES

1. MAT 096 is a not-for-credit course which is provided as a service by the college to those who have been out of school for a while or whose ACT/SAT scores prevented enrollment in a Freshman math course. This course is designed to offer those students an opportunity to brush up on their math skills and become better prepared for success in MAT 097 or subsequent college courses requiring solid math skills.

- 2. Students will demonstrate their understanding of the mathematical skills outlined in the general description of the course. This mastery of skills will be demonstrated by evaluation of tests, math labs, a common comprehensive final exam, and possibly daily work/quizzes (depending on instructor).
- 3. Students will demonstrate their knowledge of material on tests. There will be at least five and no more than nine tests. These tests will constitute 60% to 65% of the final grade.
- 4. Students will demonstrate and practice toward skill mastery through math labs and possibly daily work (depending on instructor). Labs will constitute10% to 15% of the final grade. Students will complete a minimum of five activities from the lab manual. Daily work/quizzes will constitute 0% to 5% of the final grade.
- 5. Since MAT 096 is a not-for-credit course, students will not receive a conventional letter grade (A, B, C, D, F), but will instead receive a final grade of "CR" (Credit) or "NC" (No Credit) for the course.
- 6. Students will demonstrate course mastery through a common comprehensive final exam. The final exam will constitute 25% of the final grade. Students will be given one opportunity to take the exit exam.
- 7. The grades from the six areas described above will determine whether students pass the course. Students must have at least a 75% final average to receive credit (CR) for the course. Students who have a final average of less than 75% will not receive credit (NC) for the course. Those who receive a grade of "NC" must repeat MAT 096 and earn a "CR" before being allowed to advance to MAT 097.
- 8. There is no extra credit available in this course.

## DAILY WORK AND POINTS

There will sometimes be an assignment due for each class session to ensure student preparation and class participation. Sometimes, there also will be in-class work completed for credit.

#### **OUTSIDE ASSIGNMENTS (HOMEWORK)**

Outside assignments will consist of all exercises listed on the assignment sheet. These assignments are the MINIMUM. You may need to do more; to study review sections; to use materials in the Academic Skills Center; to get help from another student, teacher or tutor; etc. Each student is expected to do the problems, check the answers (in the back of the book), and ask questions about any problems to which the solutions are unclear. These problems normally will NOT be collected but are useful in the building of a solid foundation of knowledge.

## **CHEATING**

Academic cheating will NOT be tolerated. Review procedures as outlined in the Student Handbook. Do not pretend to get an education...

## **GRADING POLICY**

The following grading scale is standard for this course in the Community and Technical College:

- CR = Grade for the course earned by a student with at least a 75% final ave.
- NC = Grade for the course earned by a student with less than a 75% final ave.

These requirements will be met from the eight categories outlined in the evaluation/assessment sections.

## **DUE DATES:**

See attached syllabus.

## MAKE-UP WORK POLICY

## <u>TESTS</u>

Tests <u>cannot</u> be made up unless prior arrangements have been made with the instructor. You must call me if you will be absent. The missed exam is arranged at the instructor's convenience.

Excused absences will be accepted for the following reasons only:

- 1. An illness that requires seeing a physician; a written medical excuse must be provided to me on the day that you return to class in order for the absence to be considered excused. I will call the doctor's office to verify each excuse.
- 2. The death of a parent, guardian, sibling, spouse, grandparent, aunt, uncle, or child. Provide an obituary or program from the funeral upon your return to class.
- 3. A Marshall University excused absence; field trip letter from instructor sports event letter from coach.

#### DAILY WORK AND LABS

There is **NO** make-up work for daily work missed. The instructors have built in some flexibility allowing for an occasional missed daily grade; no exceptions will be made. Also, the number of possible points for daily work will vary from instructor to instructor.

#### CLASS ATTENDANCE POLICY

The fact that classes are scheduled is evidence that attendance is important and students should, therefore, maintain regular attendance if they are to attain maximum success in the pursuit of their studies. Daily grades will be checked at most class sessions, so accumulated absences will affect grades. <u>There is no make-up for daily work. Absent</u> <u>students are responsible for all information and assignments given by the instructor.</u>

The student is expected to be in class on time and to stay for the entire class. (This is no more than an employer would ask of an employee.) This syllabus is a contract. Attendance in this class implies acceptance of these policies.

#### DAILY/WEEKLY OUTLINE:

See attached syllabus.

#### **CLASSROOM ETIQUETTE**

MAT 096 is a course that is designed to cover a great deal of material in a short period of time. It is therefore necessary that this course be offered to all students in an "Equal Opportunity" classroom where all students are entitled to the opportunity to do their best work without unnecessary and distracting disruptions. It is for the common good that all students come to class prepared for the day's activities and ready to concentrate and participate fully. Students are asked to refrain from sleeping, doing homework, excessive talking, unnecessary trips outside the classroom, or showing disrespect for their instructor or classmates. In order to preserve a quiet classroom environment, all pagers, cell phones, electronic games, radios, tape or CD players, wrist watches with alarms, or other devices that generate sound must be turned off when you enter the classroom. Disruption of class, whether by latecomers, noisy devices or inconsiderate behavior, will not be tolerated. Behavior problems will be referred to the assistant provost's office for appropriate action. Expulsion from the class is a possible result of this meeting.

#### **INSTRUCTOR'S RESPONSIBILITIES**

The role of the instructor will be one involving the explanation of new material and review of previously discussed material when questions are raised by students after they have attempted to do the material as an outside assignment.

#### **STUDENT'S RESPONSIBILITIES**

The student's responsibilities are a major consideration in this course. After material has been discussed during a class session, it is the student's responsibility to complete the outside assignment(s) associated with the material prior to the next class meeting so that material that remains unclear may be re-explained during the next class meeting. IN ADDITION, the student is expected to read through the NEW MATERIAL that is scheduled to be presented that class period so that the material will be generally familiar and so that preliminary questions may be asked.

#### **ADDITIONAL GUIDELINES**

Students may have coffee/soft drinks in class but should not use any smoke-less tobacco products. No Whining!!

Revised 6 June 2001 by Linda Hunt

# LEARNER OUTCOMES / OBJECTIVES

Numbers and the Number Line

- 1. Determine which of two numbers is greater
- 2. Find absolute value

Addition of Real Numbers

3. Add real numbers

Subtraction of Real Numbers

- 4. Subtract real numbers
- 5. Add and subtract in order from left to right

Multiplication of Real Numbers

6. Multiply real numbers

Division of Real Numbers

- 7. Perform division of real numbers
- 8. Perform division involving 0

Properties of Real Numbers and Order of Operations

- 9. Perform multiple operations in the proper order
- 10. Use exponents

Algebraic Notation and Terminology

- 11. Identify terms in an expression
- 12. Identify like terms
- 13. Use the distributive property
- 14. Combine like terms
- 15. Write an algebraic expression
- 16. Remove grouping symbols

Evaluating Algebraic Expressions

- 17. Evaluate an algebraic expression
- 18. Evaluate a formula
- 19. Write an algebraic expression

Addition and Subtraction Property of Equality

- 20. Determine if a given number is a root of an equation
- 21. Use the addition and subtraction property of equality
- 22. Simplify equations
- 23. Solve for an unknown
- 24. Check your answer

Multiplication and Division Property of Equality

- 25. Use the multiplication and division property of equality to form equivalent equations where the coefficient of the unknown is 1
- 26. Check your answer

Solving Linear Equations

- 27. Solve linear equations
- 28. Check your answers

Verbal Problems

- 29. Write an equation for a verbal problem
- 30. Solve for the unknown quantities

Solving Literal Equations and Formulas

31. Solve literal equations and formulas for a specified variable

Solving Linear Inequalities - The Addition and Subtraction Property of Inequalities

- 32. Graph inequalities
- 33. Graph compound inequalities
- 34. Solve inequalities
- 35. Solve compound inequalities
- 36. Write inequality statements for word statements

Solving Linear Inequalities - The Multiplication & Division Property of Inequalities

- 37. Solve linear inequalities in one variable using the multiplication and division property of inequalities
- 38. Solve linear inequalities in one variable using the addition and subtraction along with the multiplication and division properties of inequalities
- 39. Set up an inequality for a word problem and then solve it

Exponents - I

- 40. Write a product in exponential form
- 41. Multiply factors with like bases
- 42. Raise a group of factors to a power
- 43. Raise a power to a power
- 44. Raise a fraction to a power

Algebraic Addition and Subtraction

- 45. Identify like terms
- 46. Perform addition and subtraction of algebraic expressions
- 47. Remove grouping symbols

Products of Algebraic Expressions

- 48. Multiply monomials
- 49. Multiply a monomial with a multinomial
- 50. Multiply multinomials
- 51. Use the special products of the square of a binomial or the difference of two squares

Exponents - II

- 52. Perform division involving exponents
- 53. Perform operations involving negative exponents
- 54. Perform operations involving zero as an exponent

Scientific Notation

- 55. Express a number in scientific notation
- 56. Convert a number from scientific notation to standard form
- 57. Do computations using scientific notation

Ratio and Proportion

- 58. Write ratios
- 59. Reduce ratios
- 60. Write proportions
- 61. Solve proportions for the unknowns
- 62. Set up proportions to solve problems

Ordered Pairs and the Rectangular Coordinate System

- 63. Determine whether or not an ordered pair is a solution of a given equation
- 64. Find the value of one variable, given the value of the other variable
- 65. Plot ordered pairs in the rectangular coordinate plane
- 66. Plot ordered pair solutions of linear equations

Graphs of Linear Equations

- 67. Plot the graph of linear equations using ordered pairs
- 68. Find the x- and y-intercepts of a linear equation
- 69. Plot the graph of linear equations using the x- and y-intercepts
- 70. Plot graphs of the equations y = a constant and x = a constant

The Slope of a Line

- 71. Find the slope of a line given two points on the line
- 72. Determine the slope of a horizontal and vertical line

The Equation of a Line

- 73. Write the equation of a line in standard form
- 74. Find the equation of a line knowing the slope and a point or two points on the line
- 75. Find the slope and y-intercept of a line knowing the equation of the line
- 76. Graph a linear equation in two variables using the slope and yintercept
- 77. Graph a linear equation in two variables using the slope and a point on the line
- 78. Find the equation of a line given the slope and the y-intercept
- 79. Determine whether two lines are parallel

**Principal Roots** 

- 80. Find the principal square root of a perfect-square integer
- 81. Find the principal root of a number

# MATH 096 WEEKLY OUTLINE AND OBJECTIVES

The labs suggested may be replaced by your instructor. The pace and order in which the content and course objectives are covered may also vary from your instructor's pace and order.

Week 1

- A. Pretest
- B. Schedule a time for a member of the ASC staff to visit your class during the first week
- C. Explain course overview and syllabus
- D. Senses to Learn and Math Study Skills
- E. M & M lab to review fractions, decimals, and percents
- F. Section 1-1 (Numbers and the Number Line)
  - a. Determine which of 2 numbers is greater
  - b. Find absolute value

## Week 2

- A. Integer Addition with Algebra Tiles Worksheet in lab manual
- B. Adding Integers Exploration with Integer Counters in lab manual
- C. Section 1-2 (Addition of Real Numbers)
- D. Section 1-3 (Subtraction of Real Numbers)
  - a. Add and subtract in order from left to right.
- E. Integer subtraction review sheet, Subtracting Integers Exploration & checkbook cooperative activity in lab manual

- A. Multiplying Integers Exploration in lab manual
- B. Section 1-4 (Multiplication of Real Numbers)
- C. Section 1-5 (Division of Real Numbers) a. Perform division involving 0
- D. Section 1-6 (Properties of Real Numbers and Order of Operations)
  - a. Perform multiple operations in the proper order
    - b. Use exponents
- E. Cross Number Puzzle in lab manual
- F. Integer Puzzle in lab manual
- G. Understanding Exponents and Order of Operations Cooperative Activities in lab manual
- H. Order of Operations lab in lab manual
- I. Order of Operations Review sheet in lab manual

- A. Polynomial addition and subtraction using algebra tiles in lab manual
- B. Section 1-7 (Algebraic Notation and Terminology)
  - a. Identify terms in an expression
  - b. Identify like terms
  - c. Use the distributive property
  - d. Combine like terms
  - e. Write an algebraic expression
  - f. Remove grouping symbols
- C. Understanding Algebraic Addition and Subtraction cooperative activity in lab manual
- D. Section 1-8 (Evaluating Algebraic Expressions)
  - a. Evaluate a formula
  - b. Write and algebraic expression
- E. Evaluation Formulas lab in lab manual

# Week 5

- A. Test on Chapter 1 (Operations with Real Numbers and Intro to Algebra)
- B. Section 2-1 (Addition and Subtraction Property of Equality)
  - a. Determine if a given number is a root of an equation.
  - b. Simplify equations
  - c. Solve for an unknown
  - d. Check your answer
- C. Section 2-2 (Multiplication and Division Property of Equality)
- D. Section 2-3 (Solving Linear Equations)
- E. Equation puzzle in lab manual

- A. Section 2-4 (Verbal Problems)
  - a. Write an equation for an application problem
  - b. Solve for the unknown quantities
- B. Section 2-5 (Solving Literal Equations and Formulas)
- C. Section 2-6 (Solving Linear Inequalities Addition & Subtraction Property)
  - a. Graph inequalities
  - b. Graph compound inequalities
  - c. Solve compound inequalities
  - d. Write inequality statements for word statements

- A. Section 2-7 (Solving Linear Inequalities Multiplication & Division Property)
  - a. Set up an inequality for a word problem and then solve it
- B. Test on Chapter 2 (Solving Equations and Inequalities)
- C. Exponent Manipulatives Activity in lab manual

Week 8

C.

- A. Section 3-1 (Exponents I)
  - a. Write a product in exponential form
  - b. Multiply factors with like bases
  - c. Raise a group of factors to a power
  - d. Raise a power to a power
  - e. Raise a fraction to a power
- B. Section 3-2 (Algebraic Addition and Subtraction)
  - a. Identify like terms
  - b. Perform addition and subtraction of algebraic expressions.
  - c, Remove grouping symbols
  - Section 3-3 (Products of Algebraic Expressions)
    - a. Multiply monomials
    - b. Multiply a monomial with a multinomial
    - c. Multiply multinomials
    - d. Use the special product of the square of a binomial
    - e. Use the special product of the difference of two squares
- D. Multiplying Polynomials in Table Form in lab manual
- E. Section 3-4 (Exponents II)
  - a. Perform division involving exponents
  - b. Perform operations involving negative exponents
  - c. Perform operations involving zero as an exponent

- A. Understanding Polynomials and Exponents cooperative activity in lab manual
- B. Understanding Multiplication of Polynomials cooperative activity in lab manual
- C. Section 3-5 (Scientific Notation)
  - a. Express a number in scientific notation
  - b. Convert a number from scientific notation to standard form
  - c. Do computations using scientific notation
- D. Test on Chapter 3 (Algebraic Expressions)

- A. Ordered Pairs and the Rectangular Coordinate System lab
- B. Can we predict a person's height from their arm length activity
- C. Section 7-1 (Ordered Pairs and the Rectangular Coordinate System)
  - a. Determine whether or not an ordered pair is a solution of a given equation
  - b. Find the value of one variable given the value of the other variable
  - c. Plot ordered pairs in the rectangular coordinate plane
  - d. Plot ordered pair solutions of linear equations
- D. 7-1 plotting points activity in lab manual

- A. Section 7-2 (Graphs of Linear Equations)
  - a. Plot the graphs of linear equations using ordered pairs
  - b. Find the x- and y-intercepts of a linear equation
  - c. Plot the graph of linear equations using the x- and y-intercepts
  - d. Plot the graphs of the equations y = a constant and x = a constant
- B. Section 7-3 (The Slope of a Line)
  - a. Find the slope of a line given two points on the line
  - b. Determine the slope of a horizontal and vertical line
- C. Understanding Slope cooperative activity
- D. Section 7-4 (The Equation of a Line)
  - a. Write the equation of a line in standard form
  - b. Find the equation of a line knowing the slope and a point of 2 points on the line
  - c. Find the slope and y-intercept of a line knowing the equation of the line
  - d. Graph a linear equation in two variables using the slope and yintercept
  - e. Graph a linear equation in two variables using the slope and a point on the line
  - f. Find the equation of a line given the slope and y-intercept
  - g. Determine whether two lines are parallel
- E. Understanding the Equation of a Line cooperative activity

- A. Test on Chapter 7 (Linear Equations in Two Variables)
- B. Section 5-4 (Ratio and Proportion)
  - a. Write ratios
  - b. Reduce ratios
  - c. Write proportions
  - d. Solve proportions for the unknowns
  - e. Set up proportions to solve problems

Week 13

- A. Section 9-1 (Principal Roots)
  - a. Find the principal square root of a perfect-square integer
  - b. Find the principal root of a number
- B. Test on Sections 5-4 and 9-1

Week 14

- A. Review for Final Exam
- B. Make Up work

- A. Find out from your instructor where your final exam will be
- B. Finish reviewing for the final exam

Appendix C: Intermediate Algebra Syllabus

## MATH 097 COURSE OVERVIEW SHEET

# **COURSE TITLE:** DEVELOPMENTAL ALGEBRA

## COURSE NUMBER: MAT 097

## SEMESTER AND YEAR:

Section:

Time:

## REQUIRED TEXT:

<u>Principles of Intermediate Algebra With Applications</u> by Nustad and Wesner, 2nd edition

## **REQUIRED MATERIALS:**

Lab Manual

## **ADDITIONAL MATERIALS:**

## **COMPUTER REQUIREMENTS:**

## **INSTRUCTOR:**

Name:

Office:

Office Hours:

Telephone:

E-Mail Address:

#### **COURSE DESCRIPTION:**

Math 097: A course designed to improve students' skills in: first-degree equations and inequalities, polynomials, rational expressions, exponents, roots, and radicals, quadratic equations, linear equations in two variables, systems of linear equations, functions, exponential and logarithmic functions. Emphasis will be placed on skill mastery in preparation for future math courses.

#### CREDITS:

MAT 097 is a four credit hour course that counts toward full-time enrollment status and financial aid eligibility, but does not count toward the number of hours required in any college degree program. The graduation requirement is increased four hours for students who complete this course.

#### PREREQUISITES:

Placement in MAT 097 is determined by ACT score (16 – 18) or math placement test score.

#### COREQUISITE:

ASC 099: Independent Study Skills - A 1 hour CR/NC course to be taken concurrently with MAT 097. This consists of at least 15 hours in the Academic Skills Center for the semester.

## ASSESSMENT/EVALUATION OF LEARNER OUTCOMES

- 1. MAT 097 is a course that is provided as a service by the college to those who have been out of school for a while or whose ACT/SAT scores prevented enrollment in a freshman math course. This course is designed to offer those students an opportunity to brush up on their math skills and become better prepared for success in subsequent college courses requiring solid math skills.
- 2. Students will demonstrate their understanding of the mathematical skills outlined in the general description of the course. This mastery of skills will be demonstrated by evaluation of tests, math labs, a common comprehensive final exam, and possibly daily work/quizzes (depending on instructor).

- 3. Students will demonstrate their knowledge of material on tests. There will be at least five and no more than nine tests. These tests will constitute 60% to 65% of the final grade.
- 4. Students will demonstrate and practice toward skill mastery through math labs and possibly daily work (depending on instructor). Labs will constitute 10% to 15% of the final grade. Students will complete a minimum of five activities from the lab manual. Daily work/quizzes will constitute 0% to 5% of the final grade.
- 5. Students will not receive a conventional letter grade (A, B, C, D, F), but will receive a final grade of "CR" (Credit) or "NC" (No Credit) for the course.
- 6. Students will demonstrate course mastery through a common comprehensive final exam. The final exam will constitute 25% of the final grade. Students will be given one opportunity to take the exit exam.
- 7. The grades from the six areas described above will determine whether students pass the course. Students must have at least a 75% final average to receive credit (CR) for the course. Students who have a final average of less than 75% will not receive credit (NC) for the course. Those who receive a grade of "NC" must repeat MAT 097 and earn a "CR" before being allowed to advance to their next math course.
- 8. There is no extra credit available in this course.

## DAILY WORK AND POINTS

There will sometimes be an assignment due for each class session to ensure student preparation and class participation. Sometimes, there also will be inclass work completed for credit.

#### **OUTSIDE ASSIGNMENTS (HOMEWORK)**

Outside assignments will consist of all exercises listed on the assignment sheet. These assignments are the MINIMUM. You may need to do more; to study review sections; to use materials in the Academic Skills Center; to get help from another student, teacher or tutor; etc. Each student is expected to do the problems, check the answers (in the back of the book), and ask questions about any problems to which the solutions are unclear. These problems normally will NOT be collected but are useful in the building of a solid foundation of knowledge.

## **CHEATING**

Academic cheating will NOT be tolerated. Review procedures as outlined in the Student Handbook. Do not pretend to get an education...

## **GRADING POLICY**

The following grading scale is standard for this course in the CTC:

CR = Grade for the course earned by a student with at least a 75% final average

NC = Grade for the course earned by a student with less than a 75% final average.

These requirements will be met from the six categories outlined in the evaluation/assessment sections.

## MAKE-UP WORK POLICY

# <u>TESTS</u>

Tests cannot be made up unless prior arrangements have been made with the instructor. Your instructor must be notified if you will be absent. The missed exam is arranged at the instructor's convenience.

Excused absences will be accepted for the following reasons only:

- 1. An illness that requires seeing a physician; a written medical excuse must be provided to the instructor on the day that you return to class in order for the absence to be considered excused. A call to the doctor's office to verify each excuse may be made by the instructor.
- 2. The death of a parent, guardian, sibling, spouse, grandparent, aunt, uncle, or child. Provide an obituary or program from the funeral upon your return to class.
- 3. A Marshall University excused absence; field trip letter from instructor sports event letter from coach.

## DAILY WORK AND LABS

There is **NO** make-up work for daily work missed. The instructors have built in some flexibility allowing for an occasional missed daily grade; no exceptions will be made. Also, the number of possible points for daily work will vary from instructor to instructor.

#### **CLASS ATTENDANCE POLICY**

The fact that classes are scheduled is evidence that attendance is important and students should, therefore, maintain regular attendance if they are to attain maximum success in the pursuit of their studies. Daily grades will be checked at most class sessions, so accumulated absences will affect grades. <u>There is no make-up for daily work. Absent</u> <u>students are responsible for all information and assignments given</u> <u>by the instructor.</u> The student is expected to be in class on time and to stay for the entire class. (This is no more than an employer would ask of an employee.) This syllabus is a contract. Attendance in this class implies acceptance of these policies.

## **CLASSROOM ETIQUETTE**

MAT 097 is a course that is designed to cover a great deal of material in a short period of time. It is therefore necessary that this course be offered to all students in an "Equal Opportunity" classroom where all students are entitled to the opportunity to do their best work without unnecessary and distracting disruptions. It is for the common good that all students come to class prepared for the day's activities and ready to concentrate and participate fully. Students are asked to refrain from sleeping, doing homework, excessive talking, unnecessary trips outside the classroom, or showing disrespect for their instructor or classmates. In order to preserve a learning classroom environment, all pagers, cell phones, electronic games, radios, tape or CD players, wrist watches with alarms, or other devices that generate sound must be turned off when you enter the classroom. Disruption of class, whether by latecomers, noisy devices or inconsiderate behavior, will not be tolerated. Behavior problems will be referred to the division director's office for appropriate action. Expulsion from the class is a possible result of this meeting.

#### **INSTRUCTOR'S RESPONSIBILITIES**

The role of the instructor will be one involving the explanation of new material and review of previously discussed material when questions are raised by students after they have attempted to do the material as an outside assignment.

#### **STUDENT'S RESPONSIBILITIES**

The student's responsibilities are a major consideration in this course. After material has been discussed during a class session, it is the student's responsibility to complete the outside assignment(s) associated with the material prior to the next class meeting so that material that remains unclear may be re-explained during the next class meeting. IN ADDITION, the student is expected to read through the NEW MATERIAL scheduled for that class period so that the material will be generally familiar and so that preliminary questions may be asked.

## ADDITIONAL GUIDELINES

Students may have coffee/soft drinks in class but should not use any smoke-less tobacco products. No Whining!!

Revised 16 August 2001 by Linda Hunt and Jon Blatt

## Math 097 Objectives

- 2-1 Solving Equations
  - a. Solve linear equations by applying the addition and multiplication properties of equality.
  - b. Determine when a linear equation has no solution.
  - c. Check solution.
  - d. Write algebraic expressions for verbal statements.
- 2-2 Formulas and Literal Equations
  - a. Solve formulas and literal equations for the specified variable in terms of the other variables.
- 2-3 Verbal Problems
  - a. Translate verbal sentences into equations.
  - b. Solve for the unknown quantities.
- 2-4 Linear Inequalities
  - a. Solve linear inequalities.
  - b. Solve compound inequalities.
  - c. Represent the solution set of an inequality in set-builder notation or graphical notation.
- 2-5 Equations Involving Absolute Value
  - a. Solve an absolute value equation.

- 3-5 Greatest Common Factors and Factoring By Grouping
  - a. Factor the GCF from a polynomial.
  - b. Factor a four-term polynomial by grouping.
- 3-6 Factoring Trinomials of the Form  $x^2 + bx + c$  and Perfect Square Trinomials
  - a. Recognize when the trinomial  $x^2 + bx + c$  will factor and when it will not.
  - b. Factor trinomials of the form  $x^2 + bx + c$ .
  - c. Factor perfect square trinomials.
- 3-7 Factoring Trinomials of the Form  $ax^2 + bx + c$ 
  - a. Recognize when the trinomial  $ax^2 + bx + c$  will factor and when it will not.
  - b. Factor trinomials of the form  $ax^2 + bx + c$ .
- 3-8 Other Methods of factoring
  - a. Factor the difference of two squares.

- 4-1 Fundamental Properties of Rational Expressions
  - a. Determine the domain of a rational expression.
  - b. Reduce a rational expression to its lowest terms.
- 4-2 Multiplication and Division of Rational Expressions
  - a. Multiply rational expressions.
  - b. Divide rational expressions.
- 4-3 Addition and Subtraction of Rational Expressions
  - a. Find the LCD of a set of two or more rational expressions.
- 4-6 Equations Containing Rational Expressions
  - a. Solve a rational equation in one variable.
- 4-7 Problem Solving with Rational Equations
  - a. Set up and solve work problems.
  - b. Set up and solve uniform motion problems.

- 5-1 Roots and Rational Exponents
  - a. Find the principal root of a number
  - b. Express expressions with rational exponents in radical form.
  - c. Express radicals in rational exponent form.
- 5-2 Operations with Rational Exponents
  - a. Apply the properties of exponents to rational exponents.
- 5-3 Simplifying Radicals I
  - a. Simplify radicals by using the product property for radicals.
  - b. Multiply radicals with the same indices.
  - c. Simplify radicals by reducing the index of the radical.
- 5-5 Sums and Differences of Radicals
  - a. Identify like radicals.
  - b. Add and subtract like radicals.

- 6-1 Quadratic Equations and Solution by Factoring and by Extracting Roots
  - a. Solve a quadratic equation by factoring.
- 6-3 Solutions of Quadratic Equations by the Quadratic Formula
  - a. Solve a quadratic equation by using the quadratic formula.
- 6-4 Applications of Quadratic Equations
  - a. Substitute and solve physical formulas that are quadratic.
  - b. Solve verbal problems involving the use of a right angle and the Pythagorean Theorem.
  - c. Solve verbal problems involving the areas of geometric figures.
  - d. Solve verbal work problems.
- 6-5 Equations Involving Radicals
  - a. Identify extraneous solutions of a radical equation.
  - b. Find the solution set of a radical equation.

- 7-1 The Rectangular Coordinate System
  - a. Find the x- and y-intercepts of a linear equation in two variables.
  - b. Sketch the graph of a linear equation in two variables.
  - c. Sketch the graph of x = k and y = k.
- 7-2 The Distance Formula and the Slope of a Line
  - a. Find the distance between two points in the rectangular coordinate plane.
  - b. Find the slope of a straight line given two points on the line.
  - c. Determine if two lines are parallel.
- 7-3 Finding the Equation of a Line
  - a. Find the equation of a line using point-slope form.
  - b. Write the equation of a line in standard form.
  - c. Write the equation of a line in slope-intercept form.
  - d. Find the slope, m, and the y-intercept, b, of a line given its equation.
  - e. Sketch the graph of a linear equation in two variables using the slope and y-intercept.
- 8-1 Systems of Linear Equations in Two Variables
  - a. Determine whether an ordered pair is a solution of a system.
  - b. Solve a system of two linear equations in two variables.

- 10-1 Relations and Functions
  - a. Determine if a relation defines a function.
  - b. Find the domain of a relation.
  - c. Determine if a given graph of a relation represents a function by using the vertical line test.
- 10-2 Functional Notation
  - a. Evaluate f(x) for any value of x given the function f.
- 11-1 The Exponential Function
  - a. Sketch the graph of an exponential function.
  - b. Find the solution set of exponential equations.
- 11-2 The Logarithm
  - a. Given an exponential equation, write the equivalent logarithmic equation.
  - b. Given a logarithmic equation, write the equivalent exponential equation.
  - c. Evaluate a logarithmic expression.
- 11-4 The Common Logarithms
  - a. Find the common logarithm of a number using a calculator.
  - b. Find the antilogarithm of a number using a calculator.

Revised August 2k

Math 097 Weekly Schedule

Week 1

- A. Schedule a time for a member of the ASC staff to come to your class during the first week.
- B. Explain course overview and syllabus.
- C. 2-1 (Review Linear Equations)
  - a. Understanding 3 types of equations cooperative activity from lab manual
- D. 2-4 (Linear Inequalities)
  - a. First Degree Equations and Inequalities lab from manual
- E. 2-5 (Absolute Value Equations)
- F. 2-2 (Formulas and Literal Equations)
  - a. Understanding Literal Equations activity from lab manual
  - b. Errors commonly made by algebra students lab from manual

# Week 2

A. 2-3 (Verbal Problems)

- A. Test on Chapter 2 (First Degree Equations and Inequalities)
- B. 3-5 (GCF and Factor by Grouping)
- C. 3-6 (Factoring Trinomials Coefficient of the Squared term is one)
- D. 3-7 (Factoring Trinomials Coefficient of Squared term is not one)
  - a. Factoring trinomials when the coefficient of the squared term is not one activity sheet from lab manual

- A. 3-8 (Factoring the Difference of Two Squares)
- B. Factor completely worksheet from lab manual
- C. Understanding Equivalent Forms of Polynomials cooperative activity from lab manual
- D. Factoring Polynomials worksheet from lab manual
- E. 6-1 (quadratic equations and solution by factoring)
- F. Understanding Solving Quadratic Equations cooperative activity from lab manual
- G. The relationship between the solution set and the factored form of a quadratic equation activity from lab manual

Week 5

- A. Test on 3-5, 3-6, 3-7, 3-8, and 6-1
- B. 4-1 (Fundamental Properties of Rational Expressions)
  - a. Understanding Finding the Domain of a Rational Expression cooperative activity from lab manual
  - b. Domains of Rational Expressions activity from lab manual
- C. 4-2 (Multiplication and Division of Rational Expressions)
  - a. Understanding Multiplying Rational Expressions cooperative activity from lab manual

- A. 4-3 (Find the LCD for Rational Expressions)
- B. 4-6 (Equations Containing Rational Expressions)
- C. 4-7 (Problem Solving with Rational Expressions)
  - a. Rational Equations work problems lab from manual

- A. Test on 4-1, 4-2, 4-3, 4-6, and 4-7
- B. 5-1 (Roots and Rational Exponents)
- C. 5-2 (Operations with Rational Exponents)
  - a. Review of exponent properties from lab manual
  - b. Understanding Fractional Exponents cooperative activity from lab manual
- D. 5-3 (Simplifying Radicals)
  - a. Understanding Simplifying Radicals cooperative activity from lab manual

Week 8

- A. 5-5 (Sums and Differences of Radicals)
- B. Explore to Learn cooperative activity from lab manual
- C. Exponents and Radicals worksheet from lab manual
- D. 6-3 (Solutions of Quadratic Equations by the Quadratic Formula)
- E. 6-4 (Applications of Quadratic Equations)

Week 9

- A. 6-5 (Equations Involving Radicals)
- B. Quadratic and Radical Equations lab from manual
- C. Test on 5-1, 5-2, 5-3, 5-5, 6-3, 6-4, and 6-5

- A. 7-1 (The Rectangular Coordinate System)
- B. 7-2 (The Distance Formula and the Slope of a Line)
- C. 7-3 (Finding the Equation of a Line)
- D. Matching Exercise from lab manual
- E. Understanding the Equation of a Line from lab manual
- F. Graphing lab from manual
- G. Finding the equation of a line lab from manual
- H. 7-1, 7-2, 7-3 lab from manual

- A. 8-1 (Systems of Linear Equations in Two Variables)
  - a. 8-1 worksheet from lab manual
  - b. Practice Worksheet from lab manual
  - c. 8-1 lab from manual
- B. Test on 7-1, 7-2, 7-3, and 8-1
- C. Graphing Functions Worksheet from lab manual
- D. Choose the graph to fit data for a physical situation from lab manual
- E. Cooperative Learning Activity on functions from lab manual

Week 12

- A. 10-1 (Relations and Functions)
- B. 10-2 (Functional Notation)
- C. REQUIRED LAB FOR MATH 097 FROM CHAPTER 10: EITHER LABORATORY 8 GRAPHING FUNCTIONS WORKSHEET OR CHOOSING THE GRAPH TO FIT DATA FOR A PHYSICAL SITUATION.
- D. 11-1 (The exponential Function)
  - a. The sex riddle from lab manual

Week 13

- A. 11-2 (The Logarithm)
  - a. Understanding Logarithmic Equations from lab manual
- B. 11-4 (The Common Logarithms)
  - a. Log lab from manual
- C. Review for test on 10-1, 10-2, 11-1, 11-2, 11-4 from lab manual
- D. Test on 10-1, 10-2, 11-1, 11-2, and 11-4

Week 14

A. Review for final exam

Week 15

A. Finish reviewing for the final exam

Revised 14 September 2001

# Appendix D: Summary of Research Studies on Developmental Mathematics

Achievement

searcher te)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
nison 86)	College (Private Nonprofit Co-ed	Multiple Regression Stepwise	High School GPA (p = .0096) <b>SAT Math</b>	Remedial Math Grade (R squared = 17%)
	Business)	Regression Chi-Square (n = 63)	(p = .0219)	
vrence 88)	University	Regression Analysis (n = 357)	High School GPA (p < .01) <b>SAT Math</b> (p < .01)	A or B in Basic Algebra (R squared = 25%)
			In-house Algebra Placement Test (p < .01)	

mmary of Research Studies on Developmental Mathematics Achievement

searcher te)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
nny 96)	Four-year Institutions	Stepwise Regression Analysis (n = 1475)	Adjunct Instructor (p < .0001) Female Instructor (p < .0001)	Developmental Math Grade (R squared between 16% and 21%)
			Student Age (p < .0001)	
cter ith 98)	University	Analysis of Covariance (n = 435)	Andragogy (p < .05)	Intermediate Algebra Grade
			Grade in Elementary M (p = .0001)	ith
son 00)	College (Two-year	Multiple Regression	HS GPA (p = .03)	Intermediate Algebra Grade
	Technical Rural)	(n = 373)	Math HS GPA (p = .03)	
			Age (p = .01)	

mmary of Research	Studies on De	evelopmental	Mathematics	Achievement page 2	

searcher te)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
e 02)	Community College (Urban)	Multiple Regression (Setwise & Stepwise Forward) (n = 498)	Cumulative College GPA (p < .01) Math Prerequisite Status (p < .01) In-House Placement Test (p < .01)	Final Grade As A Numerical Average (R squared = 40%)
			<b>Female</b> (p < .01)	
			Math Attitude (p < .01)	
			Asian/Pacific Islanders (p < .01)	
			Individual Instructor (p < .01)	

mmary of Research Studies on Developmental Mathematics Achievement page 3

searcher ite)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
onkwiler	University	Independent	Math 1 grade	Grade in First
04)	••••••••••	Sample t-tests	(p = .01)	College-level Math Course
		1 10313	ACT Math	After Completing
		ANOVA	(p = .01)	Developmental Math
		Kruskal-	High School GPA	(R squared = 17%)
		Wallis	p = .004	
		H-tests	Gender	
		Pair Wise	Ochidei	
		Mann- Whitney tests	Ethnicity	
		Multiple Regression		
		(n = 744)		

# mmary of Research Studies on Developmental Mathematics Achievement page 4

searcher te)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
e 05)	Community College	Multiple Regression	<b>No. of Class Meetings</b> (p = .031)	Intermediate Algebra Final Grade
		Logistic Regression	White (p = .029)	(R squared = 5%)
		(n = 534)		
			<b>Female</b> (p = .036)	
			Age (p < .001)	
oer 06)	Community College	Descriptive Analysis	-	ASSET Test Score for Developmental Algebra I
		Logistic/Multiple Regression		
		Chi-Square Analysis		
		ANCOVA		

mmary of Research Studies on Developmental Mathematics Achievement page 5

earcher	Institution	Analysis		Dependent Variable
te)	(Type)	(n)	(p-value)	(R squared)
enique	Community Logistic			Developmental
07)	College (Urban	Regression (n = 451)	(p = .025)	Math Grade
	Multi-campus)		CPT Reading Score	(When all variables are entered R <sup>2</sup>
			5	between 6.5% and 8.8%
			Ethnicity and Age	
			<b>Gender</b> (p = .004)	
			Traditional College Student Status	
			Enrollment Status	
			CPT Math Score (p= .026)	Developmental Math Grade
			CPT Reading Score (p= .04	
			CPT Writing Score	Sepwise and
			Ethnicity and Age	Backward Stepwise
			<b>Gender</b> (p = .002)	R <sup>2</sup> between 6.3% and
			Traditional College Student Status (p = .004)	8.5%)
			Enrollment Status	

mmary of Research Studies on Developmental Mathematics Achievement page 6

searcher te)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
e, e 07)	Community College (Urban Public)	Multiple Regression (n = 1318)	Faculty Education Level (p = .007)	Intemediate Algebra Final Grades
			<b>Female</b> (p = .027)	
			White (p = .025)	
			Age (p < .001)	
ehl 07)	Community College (Urban Suburban multi- campus Metropolitan)Multinomial Regression CrosstabsBasic Math DV (n = 3501) p < .01 Intro Algebra DV (n = 1049)	Regression Gender	Ethnicity	DV = Basic Math (7%)
			Gender	$D_{i}$ = latroductor (
		Regression	Financial Aid	DV = Introductory Algebra (19%)
		Compass Reading Score		
		p < .01	Numerical Skills Placement Score	
		p < .01	When IA is the DV, BM	l is a predictor

mmary of Research Studies on Developmental Mathematics Achievement page 7

searcher ite)	Institution (Type)	Analysis (n)	Predictors (p-value)	Dependent Variable (R squared)
novan	University	Analysis	Mean ACT Math	C or better in
eland ebra	(Public Metropolitan	of Variance	(p < .05)	Intermediate
08)	Open- Enrollment)	Logistic Regression (n = 1694)	Mena ACT Math for Males (p < .001)	

mmary of Research Studies on Developmental Mathematics Achievement page 8

### Appendix E: Pretest Administration Memo

Marshall University Community and Technical College

General Studies Division

# Memo

 To:
 Math 096 and 097 Developmental Math Instructors

 From:
 Carol Perry, Director of General Studies Division

Date: 19 August 2001

Re: Pretests

In the past, giving a pretest was at the instructor's discretion. Starting this semester, the MCTC developmental math instructors need to collect statistical information on our developmental math students' prerequisite skills. We are going to see if there is a correlation between pretest scores and final exam scores/completion. Thank you in advance for your cooperation as we strive to serve our students better.

Attached you will find an envelope containing enough copies of the appropriate exam and scantron sheets. Be sure to take scrap paper to give to the students, and collect the scrap paper from the students when they turn in their pretest. We need for each student to bubble in a scantron sheet with their name and their answers to the exam. Additionally, they should write Math 096 or 097 at the top of their scan sheet. **Do not bubble in**: sex, birth date, ID#, or special codes. Please paperclip the 096 and 097 scan sheets to the provided instructor sheet.

There are only 20 questions on each pretest. The pretest must be administered on the first day of class. Although students drop and add the first week of class, we are only going to be concerned with getting scores for students who are present on the first day. <u>Students are not to use calculators on the pretest.</u> After administering the pretest, return the following items to the labeled area below the full-time faculty mailboxes: pretests, completed scan sheets, and unused scan sheets.

The pretests will be scored for you and you will get your results in your mailbox by Monday, August 27. Please do not go over any of the pretest questions, either in class or in your office, with your students since we will use this pretest again. By Monday, August 27, you will be provided with pretest prerequisite skills objective sheets. You may distribute these to your students.

Refer Math 096 students to Chapter R of their text for a review of fractions, decimals, and percents. The M & M lab in the Math 096 lab manual is an excellent way to reinforce fractions, decimals, and percents. Math 097 students should be referred to Chapter 1 sections 2 through 5 and Chapter 3 sections 1 through 4 of the intermediate algebra text. Students may also be referred to the Academic Skills Center to view videos on the skills they are lacking.

CAP:Idh

# Appendix F: Elementary Algebra Pretest Results

Elementary Algebra Pretest Results as reported by The SAS System Marshall University Exam Analysis System – EXAM 95 Test Score Analysis 10:01 Tuesday, September 4, 2001

FIRST-GLANCE STATISTICS (for non-blank sheets):	
NUMBER of sheets scored	249
MEAN score	40.60
MEDIAN score	40.00
STANDARD DEVIATION	15.90
Obtained LOW score	5
Obtained HIGH scoe	90

#### SUMMARY:

Number of sheets	259
Number of sheets with at least one response	259
Number of sheets with no responses (blank)	0
First question	1
Last question	20
Number of questions	20
Sheets with a "No Response"	20
Total "No Response"	32
Sheets with a "Multiple Response"	5
Total "Multiple Response"	9

### RAW SCORE STATISTICS (for non-blank sheets):

Mean score	8.12
Median score	8.00
Standard deviation	3.18
Obtained low score	1
Obtained high score	18

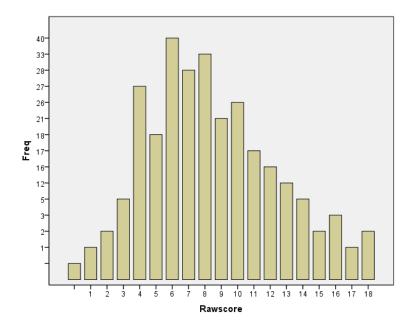
QUANTILES	Raw Score	%
99 <sup>th</sup> percentile	17	85
95 <sup>th</sup> percentile	14	70
90 <sup>th</sup> percentile	12	60
75 <sup>th</sup> percentile (3 <sup>rd</sup> quartile)		50
50 <sup>th</sup> percentile (median)	8	40
25 <sup>th</sup> percentile (1 <sup>st</sup> quartile)	6	30
10 <sup>th</sup> percentile.		
5 <sup>th</sup> percentile	4	20
1 <sup>st</sup> percentile	2	10

Appendix G: Elementary Algebra Pretest Distribution of Raw Scores

Raw	Frequency	Cumulative	Percent	Cumulative Percent
Score		Frequency		
18	2	2	0.77	0.77
17	1	3	0.39	1.16
16	3	6	1.16	2.32
15	2	8	0.77	3.09
14	5	13	1.93	5.02
13	12	25	4.63	9.65
12	16	41	6.18	15.83
11	17	58	6.56	22.39
10	26	84	10.04	32.43
9	21	105	8.11	40.54
8	33	138	12.74	53.28
7	28	166	10.81	64.09
6	40	206	15.44	79.54
5	18	224	6.95	86.49
4	27	251	10.42	96.91
3	5	256	1.93	98.84
2	2	258	0.77	99.61
1	1	259	0.39	100.00

Elementary Algebra Pretest Distribution of Raw Scores from SAS report

Histogram of the frequency of raw scores for the Elementary Algebra pretest



# Appendix H: Elementary Algebra Pretest of Prerequisite Skills

Question	Skill Being Tested	Percent who	answered co (Index of Dif	
1	Decimal addition		73%	(27%)
2	Decimal subtraction		54%	(46%)
3	Decimal multiplication		65%	(35%)
4	Decimal division		23%	(77%)
5	Rounding decimals		66%	(34%)
6	Ordering decimals (Choose the largest)		56%	(44%)
7	Order of operations		49%	(51%)
8	Decimal application		33%	(67%)
9	Adding fractions		24%	(76%)
10	Subtracting fractions		31%	(69%)
11	Subtracting mixed numbe	rs	21%	(79%)
12	Multiplying fractions		43%	(57%)
13	Dividing fractions		29%	(71%)
14	Dividing mixed numbers		28%	(72%)
15	Convert a decimal to a pe	rcent	27%	(73%)
16	Convert a fraction to a per	rcent	36%	(64%)
17	Convert a percent to a fra-	ction	21%	(79%)
18	Convert a percent to a dee	cimal	33%	(67%)
19	Find the amount in a perc	ent equation	62%	(38%)
20	Find the percent in a perc	ent equation	42%	(58%)

# Elementary Algebra Pretest of Prerequisite Skills from SAS report

# Appendix I: Intermediate Algebra Pretest Results

Intermediate Algebra Pretest Results as reported by the SAS System Marshall University Exam Analysis System – EXAM 95 Test Score Analysis 14:11 Friday, August 31, 2001

FIRST-GLANCE STATISTICS (for non-blank sheets):	
NUMBER of sheets scored	647
MEAN score	45.38
MEDIAN score	45.00
STANDARD DEVIATION	14.58
Obtained LOW score	5
Obtained HIGH score	90

#### SUMMARY:

Number of sheets	650
Number of sheets with at least one response	647
Number of sheets with no responses (blank)	3
First question	1
Last question	20
Number of questions	20
Sheets with a "No Response"	57
Total "No Response"	252
Sheets with a "Multiple Response"	8
Total "Multiple Response"	8

### RAW SCORE STATISTICS (for non-blank sheets):

Mean score	9.08
Median score	9.00
Standard deviation	2.92
Obtained low score	1
Obtained high score	18

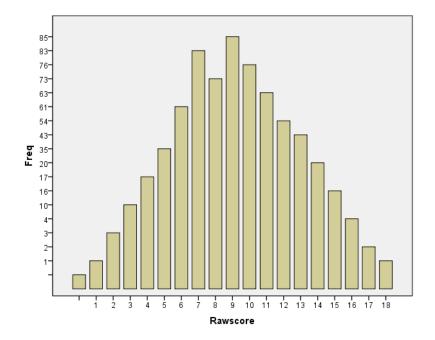
QUANTILES	Raw Score	
99 <sup>th</sup> percentile		80
95 <sup>th</sup> percentile	14	70
90 <sup>th</sup> percentile		
75 <sup>th</sup> percentile (3 <sup>rd</sup> quartile)		55
50 <sup>th</sup> percentile (median)		
25 <sup>th</sup> percentile (1 <sup>st</sup> quartile)	7	35
10 <sup>th</sup> percentile.		
5 <sup>th</sup> percentile	5	25
1 <sup>st</sup> percentile		

Appendix J: Intermediate Algebra Pretest Distribution of Raw Scores

Raw	Frequency	Cumulative	Percent	Cumulative Percent
Score		Frequency		
18	1	1	0.15	0.15
17	2	3	0.31	0.46
16	4	7	0.62	1.08
15	16	23	2.47	3.55
14	20	43	3.09	6.65
13	43	86	6.65	13.29
12	54	140	8.35	21.64
11	63	203	9.74	31.38
10	76	279	11.75	43.12
9	85	364	13.14	56.26
8	73	437	11.28	67.54
7	83	520	12.83	80.37
6	61	581	9.43	89.80
5	35	616	5.41	95.21
4	17	633	2.63	97.84
3	10	643	1.55	99.38
2	3	646	0.46	99.85
1	1	647	0.15	100.00

Intermediate Algebra Pretest Distribution of Raw Scores from SAS report

Histogram of the frequency of raw scores for the Intermediate Algebra pretest



# Appendix K: Intermediate Algebra Pretest of Prerequisite Skills

# Intermediate Algebra Pretest of Prerequisite Skills from SAS report

Question	Skill Being Tested	Percent who	answered co (Index of Dif	
1	Combine like terms		39%	(61%)
2	Solve a linear equation		43%	(57%)
3	Solve a literal equation		52%	(48%)
4	Change an equation of a line from standard form to S-I form		48%	(52%)
5	Zero exponent property		19%	(81%)
6	Like bases add exponents		52%	(48%)
7	Group of factors to a power		77%	(23%)
8	Use scientific notation		52%	(48%)
9	Determine whether an ordered p is a solution of an equation	air	41%	(59%)
10	Recognize the equation of a vert	ical line	25%	(75%)
11	Solve a linear inequality in one v	ariable	38%	(62%)
12	Plot an ordered pair		82%	(18%)
13	Find the perimeter of a rectangle	•	45%	(55%)
14	Find the slope of a line		21%	(79%)
15	Given y, find x		65%	(35%)
16	Solve an application problem		59%	(41%)
17	Quotient property for exponents		63%	(37%)
18	Square a binomial		13%	(87%)
19	Multiply a binomial times a binon	nial	46%	(54%)
20	Graph the solution of a linear ine	quality	41%	(59%)

# Appendix L: Elementary Algebra Final Exam Results

Elementary Algebra Final Exam Results as reported by The SAS System Marshall University Exam Analysis System – EXAM 95 Test Score Analysis 08:32 Monday, December 10, 2001

FIRST-GLANCE STATISTICS (for non-blank sheets):	
NUMBER of sheets scored	198
MEAN score	64.48
MEDIAN score	64.00
STANDARD DEVIATION	14.14
Obtained LOW score	
Obtained HIGH score	100

#### SUMMARY:

Number of sheets	198
Number of sheets with at least one response	
Number of sheets with no responses (blank)	0
First question	1
Last question	50
Number of questions	50
Sheets with a "No Response"	13
Total "No Response"	16
Sheets with a "Multiple Response"	4
Total "Multiple Response"	5

### RAW SCORE STATISTICS (for non-blank sheets):

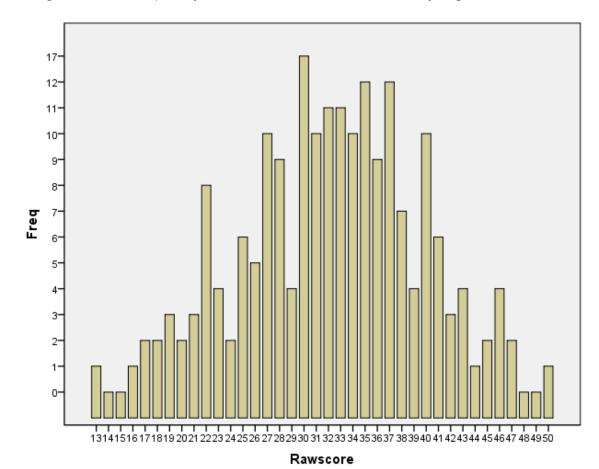
Mean score	32.24
Median score	32.00
Standard deviation	7.07
Obtained low score	13
Obtained high score	50

QUANTILES	Raw Score	
99 <sup>th</sup> percentile		94
95 <sup>th</sup> percentile		88
90 <sup>th</sup> percentile		82
75 <sup>th</sup> percentile (3 <sup>rd</sup> quartile)		74
50 <sup>th</sup> percentile (median)		
25 <sup>th</sup> percentile (1 <sup>st</sup> quartile)		56
10 <sup>th</sup> percentile.		
5 <sup>th</sup> percentile		
1 <sup>st</sup> percentile		32

Appendix M: Elementary Algebra Final Exam Distribution of Raw Scores

Frequency	Cumulative	Percent	Cumulative Percent
	Frequency		
-	1		0.51
	1		0.51
			0.51
		1.01	1.52
			3.54
2	9	1.01	4.55
1	10	0.51	5.05
4	14	2.02	7.07
3	17	1.52	8.59
6	23	3.03	11.62
10	33	5.05	16.67
4	37	2.02	18.69
7	44	3.54	22.22
12	56	6.06	28.28
9	65	4.55	32.83
12	77	6.06	38.89
10	87	5.05	43.94
11	98	5.56	49.49
11	109		55.05
			60.10
17	136		68.69
4	140		70.71
			75.25
10			80.30
			82.83
			85.86
			86.87
			88.89
			92.93
			94.44
			95.45
			96.97
			97.98
			98.99
			99.49
			99.49
			99.49
			100.00
	1         0         2         4         2         1         4         3         6         10         4         7         12         9         12         10         11         10         11         10	Frequency           1         1           0         1           0         1           2         3           4         7           2         9           1         10           4         14           3         17           6         23           10         33           4         37           7         44           12         56           9         65           12         77           10         87           11         98           11         109           10         119           17         136           4         140           9         149           10         159           5         164           6         170           2         172           4         176           8         184           3         187           2         196           1         197           0         197	Frequency           1         1         0.51           0         1         0.00           0         1         0.00           2         3         1.01           4         7         2.02           2         9         1.01           1         10         0.51           4         14         2.02           3         17         1.52           6         23         3.03           10         33         5.05           4         37         2.02           7         44         3.54           12         56         6.06           9         65         4.55           12         77         6.06           10         87         5.05           11         98         5.56           11         109         5.56           10         119         5.05           17         136         8.59           4         140         2.02           9         149         4.55           10         159         5.05           5         164         2.53

Elementary Algebra Final Exam Distribution of Raw Scores from SAS report



Histogram of the frequency of raw scores for the Elementary Algebra final exam

# Appendix N: Intermediate Algebra Final Exam Results

Intermediate Algebra Final Exam Results as reported by The SAS System Marshall University Exam Analysis System – EXAM 95 Test Score Analysis 09:42 Monday, December 10, 2001

FIRST-GLANCE STATISTICS (for non-blank sheets):	
NUMBER of sheets scored	526
MEAN score	68.40
MEDIAN score	70.00
STANDARD DEVIATION	14.07
Obtained LOW score	20
Obtained HIGH score	100

#### SUMMARY:

Number of sheets	526
Number of sheets with at least one response	526
Number of sheets with no responses (blank)	0
First question	1
Last question	40
Number of questions	40
Sheets with a "No Response"	40
Total "No Response"	104
Sheets with a "Multiple Response"	4
Total "Multiple Response"	5

### RAW SCORE STATISTICS (for non-blank sheets):

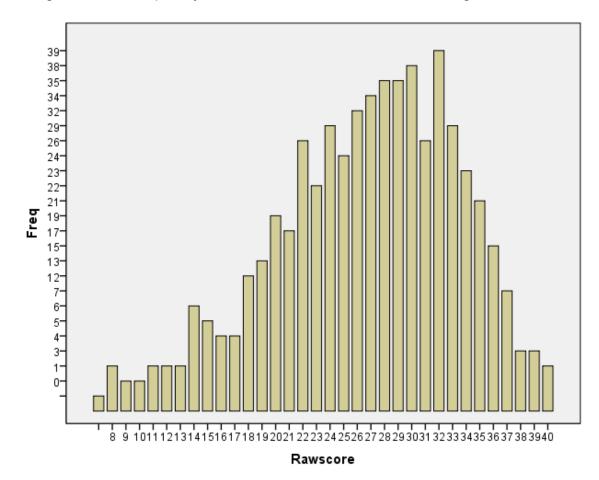
Mean score	27.36
Median score	28.00
Standard deviation	5.63
Obtained low score	8
Obtained high score	40

QUANTILES	Raw Score	
99 <sup>th</sup> percentile		95
95 <sup>th</sup> percentile		90
90 <sup>th</sup> percentile		85
75 <sup>th</sup> percentile (3 <sup>rd</sup> quartile)		80
50 <sup>th</sup> percentile (median)		70
25 <sup>th</sup> percentile (1 <sup>st</sup> quartile)		58
10 <sup>th</sup> percentile.		50
5 <sup>th</sup> percentile		45
1 <sup>st</sup> percentile		

Appendix O: Intermediate Algebra Final Exam Distribution of Raw Scores

Raw	Frequency	Cumulative	Percent	Cumulative Percent
Score		Frequency		
40	1	1	0.19	0.19
39	3	4	0.57	0.76
38	3	7	0.57	1.33
37	7	14	1.33	2.66
36	15	29	2.85	5.51
35	21	50	3.99	9.51
34	23	73	4.37	13.88
33	29	102	5.51	19.39
32	39	141	7.41	26.81
31	26	167	4.94	31.75
30	38	205	7.22	38.97
29	35	240	6.65	45.63
28	35	275	6.65	52.28
27	34	309	6.46	58.75
26	32	341	6.08	64.83
25	24	365	4.56	69.39
24	29	394	5.51	74.90
23	22	416	4.18	79.09
22	26	442	4.94	84.03
21	17	459	3.23	87.26
20	19	478	3.61	90.87
19	13	491	2.47	93.35
18	12	503	2.28	95.63
17	4	507	0.76	96.39
16	4	511	0.76	97.15
15	5	516	0.95	98.10
14	6	522	1.14	99.24
13	1	523	0.19	99.43
12	1	524	0.19	99.62
11	1	525	0.19	99.81
10	0	525	0.00	99.81
9	0	525	0.00	99.81
8	1	526	0.19	100.00

Intermediate Algebra Final Exam Distribution of Raw Scores from SAS report



Histogram of the frequency of raw scores for the Intermediate Algebra final exam

### Appendix P: Elementary Algebra Topics, Class Days, and Number of

Questions on the Final Exam

Elementary Algebra Topics

Topics		Questions on Final Exam
Determine which of 2 numbers is greater	1	1
Find absolute value	1	1
Addition of real numbers	3	1
Subtraction of real numbers	2	2
Multiplication of real numbers	1	1
Division of real numbers	1	2
Exponentiation	1	1
Order of operations	2	1
Combine like terms	1	1
Use the distributive property	1	1
Write an algebraic expression	1	1
Evaluate a formula	1	1
Algebraic addition and subtraction	1	1
Determine if a given number is a root of an equation	1	1
Addition and subtraction property of equality	1	1
Multiplication and division property of equality	1	1
Solve linear equations	1	1
Write an equation for an application problem	1	1
Solve application problems	1	1
Solve inequalities	1	1
Graph compound inequalities	1	1
Solve compound inequalities	1	1
Solve linear inequalities using multiplication and division	1	1
Linear inequality word problems	1	0
Exponent properties	4	7
Products of algebraic expressions	4	4
Scientific notation	1	1
Graph an ordered pair	3	1
Determine whether an ordered pair is a solution to an eqn.	1	1
Given x, find y	1	1
Graph a line	2	2
Graph horizontal and vertical lines	2	2
Given 2 points on a line, find the slope	1	1
Given a point and a slope, find the equation of the line	1	1
Determine whether 2 lines are parallel	1	1
Write and reduce ratios	2	1
Solve proportions	1	1

Class Number of

Write and reduce ratios2Solve proportions1Find the principal square root of a perfect square1Find the principal root of a number1

### Appendix Q: Intermediate Algebra Topics, Class Days, and Number of

Questions on the Final Exam

Intermediate Algebra Topics

Topics		Number of Questions on final exam
Solve a linear equation	1	1
Determine when a linear equation has no solution	1	1
Solve a formula for a specified variable	1	1
Translate verbal sentences into equations	1	2
Solve a linear equation verbal problem	1	2
Solve a linear inequality	1	1
Solve a compound inequality	1	1
Solve an absolute value equation	1	1
Factor a greatest common factor from a polynomial	1	1
Factor a four term polynomial by grouping	1	1
	1	1
Factor a trinomial with leading coefficient equal to 1	1	2
Factor a trinomial with leading coefficient not equal to 1	1	1
Factor the difference of two squares	-	-
Divide rational expressions	1	1
Solve a rational equation in one variable	1	1
Express radicals in rational exponent form	1	1
Simplify an expression with a rational exponent	1	1
Apply exponent properties to rational exponents	1	1
Multiply radicals with the same indices	1	1
Add like radicals	1	1
Solve a quadratic equation by factoring	1	1
Solve a quadratic eqn. by using the quadratic formula	1	1
Find the solution set of a radical equation	1	1
Find the x- and y-intercepts of a linear equation	1	1
Sketch the graph of a linear equation in two variables	1	1
Sketch the graph of a vertical line	1	1
Given two points on a line, find the slope of the line	1	1
Find the distance between two points	1	1
Determine if two lines are parallel	1	1
Given the eqn. of a line, state its slope and y-intercept	1	1
Given two points, write the equation of the line	1	1
Solve a system of linear equations in two variables	1	1
Determine if a graph of a relation is a function	1	1
Determine if a relation defines a function	1	1
Find the domain of a relation	1	1
Find the domain of a rational function	1	1
Given the function f, evaluate $f(x)$ for a value of x	1	1
Find the solution set of an exponential equation	1	1
Evaluate a logarithmic expression	1	1
Transform a logarithmic eqn. into into an exp. eqn.	1	1

# Appendix R: Descriptive Statistics for the Predictors and Outcome Variables for Elementary Algebra and Intermediate Algebra

#### **Student Outcomes**

#### Final Exam Grade

The final exam grade results are in appendices L through Q. Appendices L and N contain the final exam results including the means, medians, standard deviations, high scores, low scores, quantiles, raw scores, and percentages, for Elementary Algebra and Intermediate Algebra respectively. Appendices M and O contain the distribution of raw scores including the frequencies, cumulative frequencies, percents, and cumulative percents for Elementary and Intermediate Algebra respectively. Appendices P and Q contain the topics, number of class days, and number of questions on the final exam for Elementary Algebra and Intermediate Algebra respectively.

#### Final Course Grade

For Elementary Algebra 134 (53.2%) passed, 101 failed (40.1%), and there were 17 missing scores (6.7%). For Intermediate Algebra 450 (66.7%) passed, 221 (33%) failed, and there were two missing scores (0.3%). These data are summarized in the following tables.

#### Elementary Algebra Final Grades

Grade	Ν	Percent	
Pass	134	53.2%	
Fail	101	40.1%	
Missing	17	6.7%	
Totals	252	100%	

### Intermediate Algebra Final Grades

Grade	Ν	Percent
Pass	450	66.7%
Fail	223	33 %
Missing	2	.3%
Totals	675	100%

### **Potential Predictors of Student Outcomes**

Potential predictors of student outcomes occur at the student-level,

instructor-level, and classroom-level. They will be considered in order.

### Student-level Potential Predictors

Student-level predictors are specific to an individual student which included: student gender, ACT Math score, SAT Math score, college, math pretest score, and supplemental instruction as measured by the ASC grade. These will be considered in order.

#### Student Gender

The data group used in this study was comprised of 11 sections of Elementary Algebra with a total of 198 students and 28 sections of Intermediate Algebra with a total of 526 students for the fall 2001 semester. Seventy-seven of the 198 Elementary Algebra students (39%) were male, while 91 of the 198 Elementary Algebra students (46%) were female, with the gender data missing for 30 of the 198 Elementary Algebra students (15%). One hundred ninety-five of the 526 Intermediate Algebra students (37%) were male, while 313 of the 526 Intermediate Algebra students (60%) were female, with the gender data missing for 18 of the 526 Intermediate Algebra students (3%). There were more females than males in both classes. These data are summarized in the following tables. Student Gender for Elementary Algebra

Gender	Ν	Percent
Males	77	39%
Female	91	46%
Missing	30	15%
Totals	198	100%

Gender	Ν	Percent	
Males	195	37%	
Females	313	60%	
Missing	18	3%	
Totals	526	100%	

#### Student Gender for Intermediate Algebra

## ACT Math Score

An ACT Math score between 12 and 15 placed students in Elementary Algebra while an ACT Math score between 16 and 18 placed students in Intermediate Algebra (see syllabi in Appendices B and C). The actual range of ACT Math scores for Elementary Algebra was 10 to 21 while the range for Intermediate Algebra ACT Math scores was 6 to 22. Although "research has consistently shown that mandatory assessment and placement contributes to student success (Boylan, 2002, p. 36)",MUCTC did not enforce their placement policy (Hunt L. D., 2000). Twelve of the Elementary Algebra students were not properly placed while 72 of the Intermediate Algebra students were not properly placed. Sixty-eight of the 72 students should have been enrolled in Elementary Algebra instead of Intermediate Algebra. The ACT Math Scores and the enrollment in each category for Elementary and Intermediate Algebra are summarized in the following tables.

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Math ACT Score	Ν
10	2
11	0
12	8
13	12
14	50
15	97
16	6
17	1
18	1
19	0
20	0
21	2
Total	179

# Elementary Algebra Math ACT Scores

The ACT mean, median, and mode for Elementary Algebra was 15. The standard deviation was 1.233, the range was 11, and there were 72 missing pieces of data.

Math ACT Score	Ν
6	1
7	0
8	0
9	0
10	0
11	0
12	2
13	10
14	15
15	40
16	170
17	205
18	161
19	2
20	0
21	1
22	1
Total	608

# Intermediate Algebra Math ACT Scores

The ACT mean, median, and mode for Intermediate Algebra was 17. The standard deviation was 1.252, the range was 16, and there were 66 missing pieces of data.

## SAT Math Score

For the Elementary Algebra data set, 17 students had the SAT math score as part of their records. Eleven students had the SAT math score but no ACT math score which resulted in list-wise deletion of these 11 records with multiple regression. The remaining six students had both ACT math and SAT math scores which would not result in the deletion of the student records. These results are summarized in the following table.

Elementary Algebra SAT Math Scores

Math Test	Ν	
SAT Math Score only	11	
SAT and ACT Math Scores	6	
Total	17	

For the Intermediate Algebra data set, 80 students had the SAT math score as part of their records. Thirty-seven had the SAT math score but no ACT math score which resulted in list-wise deletion of these 37 records with multiple regression. The remaining 43 Intermediate Algebra students had both ACT math and SAT math scores which would not result in the deletion of the student record. These results are summarized in Table 11.

# Intermediate Algebra SAT Math Scores

Math Test	Ν
SAT Math Score only	37
SAT and ACT Math Scores	43
Total	80

## College

The number of Community and Technical College students compared to University students for Elementary Algebra and Intermediate Algebra are summarized in the following tables.

Comparison of Community and Technical College and University Students for

Elementary Algebra

Institution	Number	Percent
Community and Technical College Students	132	43%
University Students	107	52%
Missing	13	5%
Totals	252	100%

Comparison of Community and Technical College and University Students for Intermediate Algebra

Institution	Number	Percent
Community and Technical College Students	141	21%
University Students	529	78%
Missing	5	1%
Totals	674	100%

#### Pretest

Results of the Marshall University Exam Analysis System by SAS are included for the pretests in the appendices. Test result summaries and descriptive statistics are in appendices F through K.

## ASC Grade

For Elementary Algebra 166 passed (65.9%), 72 failed (28.6%), and there were 14 missing scores (5.6%). For Intermediate Algebra 485 passed (71.9%), 180 (26.7%) failed, and there were 10 missing scores (1.5%). These data are summarized in the following tables.

#### Elementary Algebra ASC Final Grades

Grade	Ν	Percent
0	72	28.6%
1	166	65.9%
Missing	14	5.6%
Totals	252	100.1% (due to round-off error)

#### Intermediate Algebra ASC Final Grades

Grade	Ν	Percent
0	180	26.7%
1	485	71.9%
Missing	10	1.5%
Totals	675	100.1% (due to round-off error)

## Instructor-level Potential Predictors

Although all developmental algebra instructors (Elementary Algebra and Intermediate Algebra) had at least a Bachelor's degree, and none of the developmental algebra instructors had a doctorate, instructor education was not included as one of the independent variables studied. Some studies (Fike & Fike, 2007) included faculty education level, but in this study information about individual instructors' educational levels was not collected.

Instructor-level predictors are specific to a particular instructor which included: instructor gender, instructor employment status, instructor playing Mozart for Your Mind tape, and instructor requiring use of ALEKS software. They will be considered in order.

#### Elementary Algebra

The 11 sections of Elementary Algebra were taught by 9 different instructors, of which two (22%) were males and seven (78%) were females. Four (44%) of these instructors were full-time and five (56%) were adjunct. In one of these sections (9%), the Mozart for Your Mind tape was played at the start and end of each class; the female instructor who played this music was full time. ALEKS software was not used in the Elementary Algebra course.

#### Intermediate Algebra

The 28 sections of Intermediate Algebra were taught by 15 different instructors, of which four (27%) were males and 11 (73%) were females. Five (33%) of these instructors were full-time and 10 (67%) were adjunct. The same instructor who played the Mozart for Your Mind tape for her Elementary Algebra students also played said tape for her two sections of Intermediate Algebra or 7% of the 28 sections. One full-time instructor used ALEKS software in two of the three classes that he taught. One ALEKS class had regular class meetings; the other was self paced. ALEKS software was used in two of the 28 sections (7%) of the Intermediate Algebra sections. This information is summarized in the following tables.

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# Elementary Algebra Instructor Gender

Gender	Ν	Percent
Males	2	22%
Females	7	78%
Totals	9	100%

# Elementary Algebra Instructor Employment Status

Employment Status	Ν	Percent
Full-Time	4	44%
Adjunct	5	56%
Totals	9	100%

# Elementary Algebra Mozart Music Use Per Section

Mozart for Your Mind Tape Use	Ν	Percent
Tape Used	1	9%
Tape Not Used	10	91%
Totals	11	100%

# Intermediate Algebra Instructor Gender

l	Percent
4	27%
1	73%
5	100%
4	4

# Intermediate Algebra Instructor Employment Status

Employment Status	Ν	Percent
Full-Time	5	33%
Adjunct	10	67%
Totals	15	100%

# Intermediate Algebra Mozart Music Use per Section

Mozart for Your Mind Use	Ν	Percent
Tape Used	2	7%
Tape Not Used	26	93%
Totals	28	100%

#### Intermediate Algebra ALEKS Software Use per Section

ALEKS Software Use	Ν	Percent
Software Used	2	7%
Software Not Used	26	93%
Totals	28	100%

For both classes the female instructors outnumbered the male instructors in numbers and percents. While there was nearly a 50-50 split on the employment status for Elementary Algebra instructors, two-thirds of Intermediate Algebra instructors were adjunct and only one-third were full-time.

## Potential Classroom-level Predictors

Classroom-level predictors are specific to the classroom and are an integral part of the learning environment. These included: class time of day, number of class meetings, and class size. They will be considered in order. *Time of Day* 

Although the beginning time of class was in the original data set, in order to simplify this categorical variable, the classes were coded as a dichotomous variable (i.e., a.m. or p.m. classes). Classes starting between 8 a.m. and noon were considered to be a.m. classes while classes starting between 12:15 p.m. and 6:30 p.m. were considered p.m. classes. For the 11 classes in the Elementary Algebra data set, there were four (36%) a.m. classes and seven (64%) p.m. classes. Three out of the four (75%) a.m. classes were taught by full-

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time faculty, while one out of the four (25%) a.m. classes was taught by an adjunct faculty member. This is summarized in the following table.

Elementary Algebra Time of Day

Time	Ν	Percent	
a.m.	4	36%	
p.m.	7	64%	
Totals	11	100%	

For the 28 classes in the Intermediate Algebra data set, there were nine (32%) a.m. classes, 18 (64%) p.m. classes, and one (4%) self-paced class which did not have a time. It is interesting, but not unusual to note, for the Intermediate Algebra classes all nine of the a.m. classes were taught by full-time faculty. For the p.m. classes, 15 of the 18 classes (83%) were taught by adjunct instructors while three of the 18 classes (17%) were taught by full-time instructors. This is summarized in the following table.

Time	Ν	Percent	
a.m.	9	32%	
p.m.	18	64%	
Self-paced	1	4%	
Totals	28	100%	

Intermediate Algebra	Time	of Day
----------------------	------	--------

#### Number of Class Meetings

For the Elementary Algebra data group the number of class meetings ranged from twice a week to four or five times per week. Five of the 11 (45%) classes met twice a week; four out of these five (80%) classes were taught by adjunct faculty. Three of the 11 (27%) classes met four times per week; one out of the three classes was taught by an adjunct faculty member. Three of the 11 (27%) classes met five times per week; one out of the three classes was taught by an adjunct faculty member. This is summarized in the following table. Elementary Algebra Number of Class Meetings

Number of Class Meetings	Ν	Percent	-
2	5	45%	-
4	3	27%	
5	3	27%	
Totals	11	100%	

For the Intermediate Algebra data group the number of class meetings ranged from two to five times per week. Fourteen of the 27 (52%) classes met twice a week; all of these classes were taught by adjunct faculty, and all of them were p.m. classes. One of the 27 (4%) classes met three times a week in the p.m. with an adjunct instructor. Seven of the 27 (26%) classes met four times a week; one of these seven (14%) classes was taught by an adjunct instructor, the remaining six out of seven (86%) were taught by full-time instructors. Five of the

27 (19%) classes met five times per week. All five of these classes were a.m.

classes taught by full-time faculty. This is summarized in the following table.

Intermediate Algebra	Number of Class	s Meetinas
internetiate Aigebra		s meetings

# Class Size

In the 11 Elementary Algebra sections, the class size ranged from 11 to 44 students. In the 28 Intermediate Algebra sections, the class size ranged from 19 to 39 students.

# Appendix S: SPSS Simple Regression Output for Elementary Algebra and

Intermediate Algebra

REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /<u>DEPENDENT fnlexam</u> /<u>METHOD=ENTER gender</u> /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 12:04:09
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER gender
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:02.043
	Elapsed Time	00:00:47.729
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet2] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

## **Descriptive Statistics**

	Mean	Std. Deviation	N
Fnlexam	65.13	14.036	168
Gender	.46	.500	168

Correlations				
		fnlexam	gender	
Pearson Correlation	Fnlexam	1.000	035	
	Gender	035	1.000	
Sig. (1-tailed)	Fnlexam		.326	
	Gender	.326		
Ν	Fnlexam	168	168	
	Gender	168	168	

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	gender <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model					Char	nge Statistics	
	R	R Square	Adjusted R Square	Std. Error of the Estimate	R Square Change	F Change	df1
1	.035 <sup>a</sup>	.001	005	14.070	.001	.204	1

a. Predictors: (Constant), gender

b. Dependent Variable: fnlexam

Model	Summary <sup>b</sup>
-------	----------------------

Model	Change Statistics		Change Statistics		
	df2	Sig. F Change	Durbin-Watson		
1	166	.652	1.702		

#### b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>					
Mode	el	Sum of Squares	Df	Mean Square	F	Sig.
1	Regression	40.468	1	40.468	.204	.652 <sup>a</sup>
	Residual	32860.651	166	197.956		
	Total	32901.119	167			

a. Predictors: (Constant), gender

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>					
Mode	el			Standardized		
		Unstandardize	ed Coefficients	Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	65.582	1.475		44.466	.000
	gender	985	2.179	035	452	.652

a. Dependent Variable: fnlexam

#### Coefficients<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	62.670	68.494			
	gender	-5.286	3.316	035	035	035

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model		Collinearity	Statistics	
		Tolerance	VIF	
1	(Constant)			
	gender	1.000	1.000	

a. Dependent Variable: fnlexam

## Coefficient Correlations<sup>a</sup>

Model			gender
1	Correlations	Gender	1.000
	Covariances	Gender	4.746

a. Dependent Variable: fnlexam

#### **Collinearity Diagnostics**<sup>a</sup>

Model	Dimension			Variance Proportions			
		Eigenvalue	Condition Index	(Constant)	gender		
1	1	1.677	1.000	.16	.16		
	<sup></sup> 2	.323	2.279	.84	.84		

Model	Dimension			Variance Proportions			
		Eigenvalue	Condition Index	(Constant)	gender		
1	1	1.677	1.000	.16	.16		
	- 2	.323	2.279	.84	.84		

**Collinearity Diagnostics**<sup>a</sup>

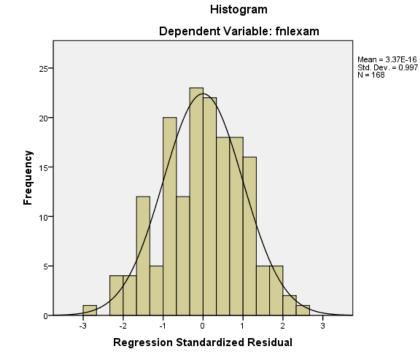
a. Dependent Variable: fnlexam

## **Residuals Statistics**<sup>a</sup>

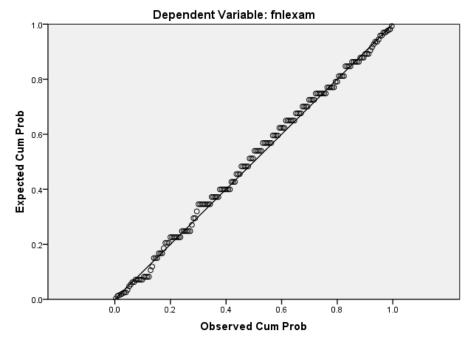
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	64.60	65.58	65.13	.492	168
Residual	-38.597	34.418	.000	14.027	168
Std. Predicted Value	-1.084	.917	.000	1.000	168
Std. Residual	-2.743	2.446	.000	.997	168

a. Dependent Variable: fnlexam

# Charts







#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N

/MISSING LISTWISE

/STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP

/CRITERIA=PIN(.05) POUT(.10)

/NOORIGIN

/DEPENDENT fnlexam

/METHOD=ENTER act

/RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 12:14:20
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER act
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:01.092
	Elapsed Time	00:00:01.190
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

# [DataSet2] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

	Descriptive Statistics						
	Mean	Std. Deviation	N				
fnlexam	64.37	14.173	129				
act	14.58	1.368	129				

Correlations						
		fnlexam	act			
Pearson Correlation	fnlexam	1.000	.238			
	act	.238	1.000			
Sig. (1-tailed)	fnlexam		.003			
	act	.003				
Ν	fnlexam	129	129			
	act	129	129			

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	act <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.238 <sup>a</sup>	.057	.049	13.820	.057	7.616	1

a. Predictors: (Constant), act

b. Dependent Variable: fnlexam

# Model Summary<sup>b</sup>

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	127	.007	1.715

#### b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>								
1	Vodel	Sum of Squares	Df	Mean Square	F	Sig.			
	1 Regression	1454.539	1	1454.539	7.616	.007 <sup>a</sup>			
	Residual	24255.601	127	190.989					
	Total	25710.140	128						

a. Predictors: (Constant), act

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>							
Mode	el			Standardized				
		Unstandardized Coefficients		Coefficients				
		В	Std. Error	Beta	t	Sig.		
1	(Constant)	28.430	13.081		2.173	.032		
	act	2.465	.893	.238	2.760	.007		

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	2.546	54.314			
	act	.697	4.232	.238	.238	.238

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model		Collinearity	Statistics	
		Tolerance	VIF	
1	(Constant)			
	act	1.000	1.000	

a. Dependent Variable: fnlexam

#### **Coefficient Correlations**<sup>a</sup>

Model			act
1	Correlations	act	1.000
	Covariances	act	.798

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	Act
1	1	1.996	1.000	.00	.00
	<sup>-</sup> 2	.004	21.454	1.00	1.00

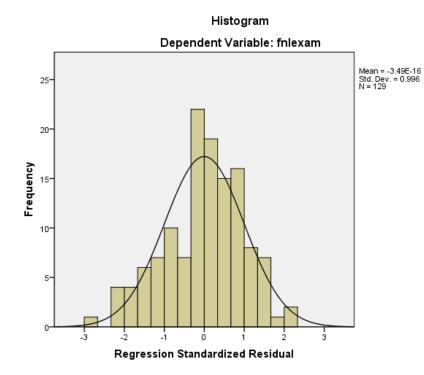
a. Dependent Variable: fnlexam

## Residuals Statistics<sup>a</sup>

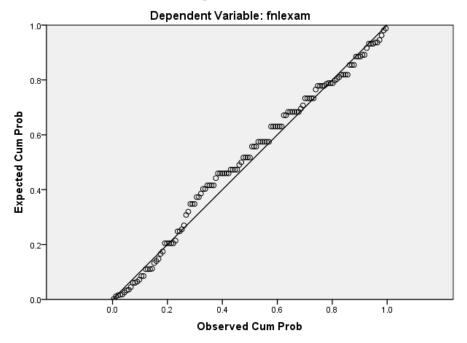
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	53.08	80.19	64.37	3.371	129
Residual	-39.404	31.061	.000	13.766	129
Std. Predicted Value	-3.350	4.693	.000	1.000	129
Std. Residual	-2.851	2.248	.000	.996	129

a. Dependent Variable: fnlexam

# Charts







REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER sat /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

## Regression

	Notes	
Output Created		02-Oct-2011 12:17:01
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION
- ,		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER sat
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
_		
Resources	Processor Time	00:00:01.139
	Elapsed Time	00:00:01.111
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet2] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

## **Descriptive Statistics**

	Mean	Std. Deviation	N
fnlexam	64.20	15.332	10
sat	348.00	38.528	10

Correlations					
		fnlexam	sat		
Pearson Correlation	fnlexam	1.000	067		
	sat	067	1.000		
Sig. (1-tailed)	fnlexam		.427		
	sat	.427			
Ν	fnlexam	10	10		
	sat	10	10		

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	sat <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model Summary <sup>t</sup>
----------------------------

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.067 <sup>a</sup>	.004	120	16.225	.004	.036	1

a. Predictors: (Constant), sat

b. Dependent Variable: fnlexam

Model	Summary <sup>b</sup>

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	8	.854	2.012

b. Dependent Variable: fnlexam

ANOVA <sup>b</sup>	
--------------------	--

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	9.486	1	9.486	.036	.854 <sup>a</sup>
	Residual	2106.114	8	263.264		
	Total	2115.600	9			

a. Predictors: (Constant), sat

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>					
Model				Standardized		
		Unstandardized Coefficients		Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	73.473	49.120		1.496	.173
	sat	027	.140	067	190	.854

a. Dependent Variable: fnlexam

## **Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	-39.797	186.743			
	sat	350	.297	067	067	067

a. Dependent Variable: fnlexam

## **Coefficients**<sup>a</sup>

Mod	lel	Collinearity	Statistics
		Tolerance	VIF
1	(Constant)		
	sat	1.000	1.000

a. Dependent Variable: fnlexam

#### **Coefficient Correlations**<sup>a</sup>

Model			sat
1	Correlations	sat	1.000
	Covariances	sat	.020

a. Dependent Variable: fnlexam

Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	sat
1	1	1.995	1.000	.00	.00
	2	.005	19.094	1.00	1.00

#### **Collinearity Diagnostics**<sup>a</sup>

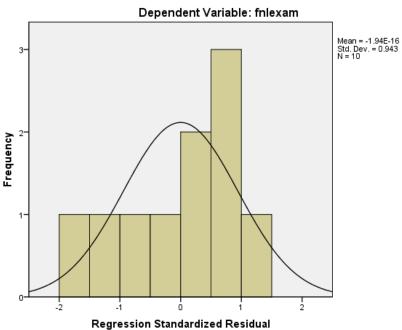
a. Dependent Variable: fnlexam

Residuals Statistics	sa
----------------------	----

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	63.35	65.75	64.20	1.027	10
Residual	-25.347	18.653	.000	15.297	10
Std. Predicted Value	831	1.505	.000	1.000	10
Std. Residual	-1.562	1.150	.000	.943	10

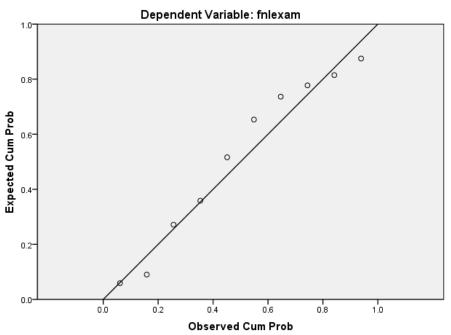
a. Dependent Variable: fnlexam

# Charts



Histogram Dependent Variable: fnlexam





#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER comcol /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3). Regression

	Notes	
Output Created		02-Oct-2011 12:19:40
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	I I

Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER comcol
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:01.170
	Elapsed Time	00:00:01.188
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet2] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Descriptive Statistics						
	Mean	Std. Deviation	N			
fnlexam	65.14	13.994	169			
comcol	.54	.500	169			

#### Correlations

		fnlexam	comcol
Pearson Correlation	fnlexam	1.000	174
	comcol	174	1.000

Sig. (1-tailed)	fnlexam		.012
	comcol	.012	
Ν	fnlexam	169	169
	comcol	169	169

#### Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	comcol <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

	Model Summary <sup>b</sup>								
Model			Change Statistics						
			Adjusted R	Std. Error of the	R Square				
	R	R Square	Square	Estimate	Change	F Change	df1		
1	.174 <sup>a</sup>	.030	.025	13.822	.030	5.223	1		

a. Predictors: (Constant), comcol

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	167	.024	1.728

#### b. Dependent Variable: fnlexam

**ANOVA<sup>b</sup>** 

_	ANOVA					
	Model	Sum of Squares	df	Mean Square	F	Sig.
	1 Regression	997.890	1	997.890	5.223	.024 <sup>a</sup>
I	Residual	31903.980	167	191.042		
	Total	32901.870	168			

a. Predictors: (Constant), comcol

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>						
Mod	el			Standardized			
		Unstandardized Coefficients		Coefficients			
		В	Std. Error	Beta	t	Sig.	
1	(Constant)	67.792	1.575		43.039	.000	
	comcol	-4.879	2.135	174	-2.285	.024	

a. Dependent Variable: fnlexam

## **Coefficients**<sup>a</sup>

Mode	1	95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	64.682	70.902			
	comcol	-9.094	664	174	174	174

a. Dependent Variable: fnlexam

## **Coefficients**<sup>a</sup>

Mod	el	Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	comcol	1.000	1.000	

a. Dependent Variable: fnlexam

## Coefficient Correlations<sup>a</sup>

Model			comcol
1	Correlations	comcol	1.000
	Covariances	comcol	4.558

a. Dependent Variable: fnlexam

Model	Dimension			Variance Proportions			
		Eigenvalue	Condition Index	(Constant)	comcol		
1	1	1.738	1.000	.13	.13		
	2	.262	2.575	.87	.87		

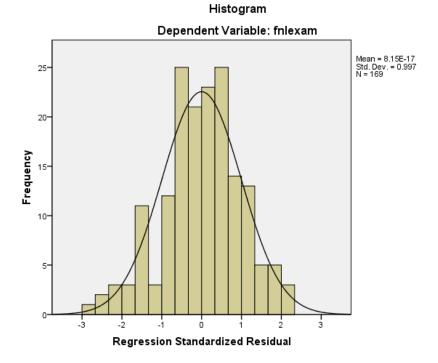
#### **Collinearity Diagnostics**<sup>a</sup>

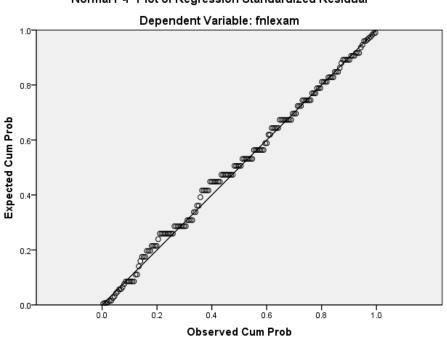
a. Dependent Variable: fnlexam

Residuals Statistics <sup>a</sup>					
	Minimum	Maximum	Mean	Std. Deviation	N
Predicted Value	62.91	67.79	65.14	2.437	169
Residual	-36.913	32.208	.000	13.781	169
Std. Predicted Value	912	1.090	.000	1.000	169
Std. Residual	-2.671	2.330	.000	.997	169

a. Dependent Variable: fnlexam

## Charts





#### Normal P-P Plot of Regression Standardized Residual

REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE

/STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP

/CRITERIA=PIN(.05) POUT(.10)

/NOORIGIN

/DEPENDENT fnlexam

/METHOD=ENTER pretest

/RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID)

/CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes					
Output Created		02-Oct-2011 12:21:24			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet2			
	Filter	<none></none>			
	Weight	<none></none>			
	Split File	<none></none>			

	N of Rows in Working Data	252	
	File		
Missing Value Handling	Definition of Missing	User-defined missing values are	
		treated as missing.	
	Cases Used	Statistics are based on cases with no	
		missing values for any variable used.	
Syntax		REGRESSION	
		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER pretest	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:01.092	
	Elapsed Time	00:00:01.163	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots	A fall 2001 with alagerooms ago	

Descriptive Statistics						
	Mean	Std. Deviation	Ν			
fnlexam	65.14	13.994	169			
pretest	41.60	16.845	169			

fnlexam	pretest

Pearson Correlation	fnlexam	1.000	.370
	pretest	.370	1.000
Sig. (1-tailed)	fnlexam		.000
	pretest	.000	
Ν	fnlexam	169	169
	pretest	169	169

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables		
	Entered	Removed	Method	
1	pretest <sup>a</sup>		Enter	

a. All requested variables entered.

b. Dependent Variable: fnlexam

# Model Summary<sup>b</sup>

Model					Change Statistics		
	R	R Square	Adjusted R Square	Std. Error of the Estimate	R Square Change	F Change	df1
1	.370 <sup>a</sup>	.137	.132	13.040	.137	26.489	1

a. Predictors: (Constant), pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2 Sig. F Change		Durbin-Watson				
1	167	.000	1.864				

ANOVA <sup>b</sup>						
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	4504.384	1	4504.384	26.489	.000 <sup>a</sup>
	Residual	28397.486	167	170.045		

Total 32901.870	68
-----------------	----

a. Predictors: (Constant), pretest

b. Dependent Variable: fnlexam

_	Coefficients <sup>a</sup>							
ſ	Model			Standardized				
		Unstandardized Coefficients		Coefficients				
		В	Std. Error	Beta	t	Sig.		
	1 (Constant)	52.349	2.679		19.538	.000		
	pretest	.307	.060	.370	5.147	.000		

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

_	obericients						
Model		95.0% Confidence Interval for B		Correlations			
		Lower Bound Upper Bound		ound Zero-order Partial		Part	
1	(Constant)	47.059	57.639				
	pretest	.189	.425	.370	.370	.370	

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		Collinearity	Statistics
		Tolerance	VIF
1	(Constant)		
	pretest	1.000	1.000

a. Dependent Variable: fnlexam

### Coefficient Correlations<sup>a</sup>

Model			pretest
1	Correlations	pretest	1.000
	Covariances	pretest	.004

Somilarity Braghootico					
Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	pretest
1	1	1.927	1.000	.04	.04
	2	.073	5.148	.96	.96

**Collinearity Diagnostics**<sup>a</sup>

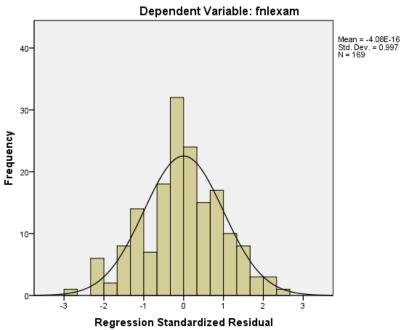
a. Dependent Variable: fnlexam

# Residuals Statistics<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	53.89	80.01	65.14	5.178	169
Residual	-37.108	32.429	.000	13.001	169
Std. Predicted Value	-2.173	2.873	.000	1.000	169
Std. Residual	-2.846	2.487	.000	.997	169

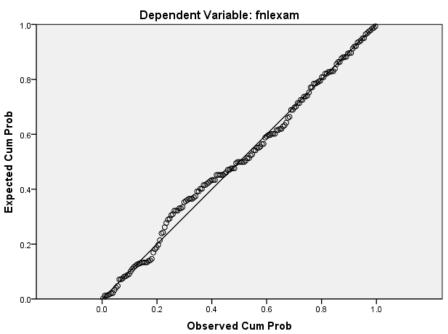
a. Dependent Variable: fnlexam

# Charts



Histogram )ependent Variable: folexam





#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER ascgr /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3). Regression

	Notes	
Output Created		02-Oct-2011 12:24:01
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	I I

Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER ascgr
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:00.687
	Elapsed Time	00:00:01.135
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

Descriptive Statistics					
	Mean	Std. Deviation	Ν		
fnlexam	65.14	13.994	169		
ascgr	.88	.324	169		

Correlations				
		fnlexam	ascgr	
Pearson Correlation	fnlexam	1.000	036	
	ascgr	036	1.000	
Sig. (1-tailed)	fnlexam		.322	
	ascgr	.322		
Ν	fnlexam	169	169	
	ascgr	169	169	

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	ascgr <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Chai	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.036 <sup>a</sup>	.001	005	14.027	.001	.214	1

a. Predictors: (Constant), ascgr

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	167	.644	1.724			

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	42.199	1	42.199	.214	.644 <sup>a</sup>		
	Residual	32859.671	167	196.764				
	Total	32901.870	168					

a. Predictors: (Constant), ascgr

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
		Unstandardize	ed Coefficients	Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	66.500	3.137		21.201	.000			
	ascgr	-1.547	3.340	036	463	.644			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	60.308	72.692			
	ascgr	-8.142	5.048	036	036	036

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model		Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	ascgr	1.000	1.000	

#### Coefficient Correlations<sup>a</sup>

Model			ascgr
1	Correlations	ascgr	1.000
	Covariances	ascgr	11.159

a. Dependent Variable: fnlexam

### Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	ascgr
1	1	1.939	1.000	.03	.03
	2	.061	5.636	.97	.97

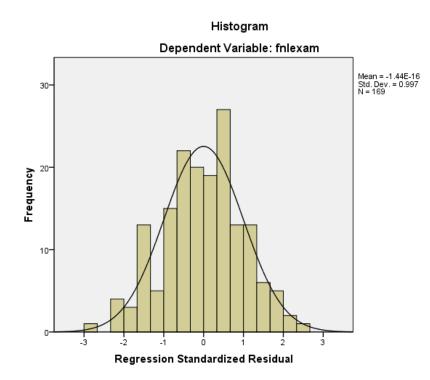
a. Dependent Variable: fnlexam

#### **Residuals Statistics**<sup>a</sup>

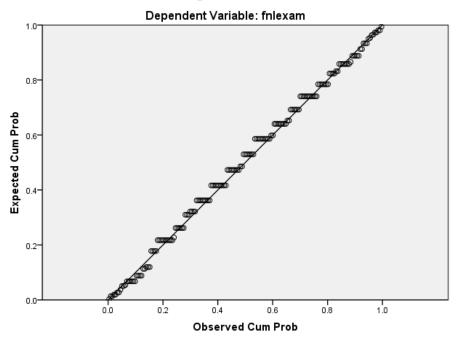
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	64.95	66.50	65.14	.501	169
Residual	-38.953	35.047	.000	13.985	169
Std. Predicted Value	365	2.721	.000	1.000	169
Std. Residual	-2.777	2.498	.000	.997	169

a. Dependent Variable: fnlexam

## Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER techsex /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 12:27:23
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION	
o jinax		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER techsex	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:00.982	
	Elapsed Time	00:00:01.121	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

### **Descriptive Statistics**

	Mean	Std. Deviation	N
fnlexam	65.14	13.994	169
techsex	.36	.480	169

Correlations					
		fnlexam	techsex		
Pearson Correlation	fnlexam	1.000	.147		
	techsex	.147	1.000		
Sig. (1-tailed)	fnlexam		.028		
	techsex	.028			
Ν	fnlexam	169	169		
	techsex	169	169		

#### Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	techsex <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model Summary	b
---------------	---

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.147 <sup>a</sup>	.022	.016	13.884	.022	3.687	1

a. Predictors: (Constant), techsex

b. Dependent Variable: fnlexam

Model	Summary <sup>b</sup>
-------	----------------------

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	167	.057	1.760

#### b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
Mode	el	Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	710.653	1	710.653	3.687	.057 <sup>a</sup>		
	Residual	32191.217	167	192.762				
	Total	32901.870	168					

a. Predictors: (Constant), techsex

	Coefficients <sup>a</sup>					
Mode	el			Standardized		
		Unstandardized Coefficients		Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	63.615	1.330		47.837	.000
	techsex	4.285	2.232	.147	1.920	.057

a. Dependent Variable: fnlexam

#### Coefficients<sup>a</sup>

Mode	el	95.0% Confidence Interval for B Correlations				
		Lower Bound Upper Bound Ze		Zero-order	Partial	Part
1	(Constant)	60.989	66.240			
	techsex	121	8.692	.147	.147	.147

a. Dependent Variable: fnlexam

### **Coefficients**<sup>a</sup>

Mod	lel	Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	techsex	1.000	1.000	

a. Dependent Variable: fnlexam

# Coefficient Correlations<sup>a</sup>

Model			techsex
1	Correlations	techsex	1.000
	Covariances	techsex	4.981

Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	techsex
1	1	1.596	1.000	.20	.20
	- 2	.404	1.987	.80	.80

#### **Collinearity Diagnostics**<sup>a</sup>

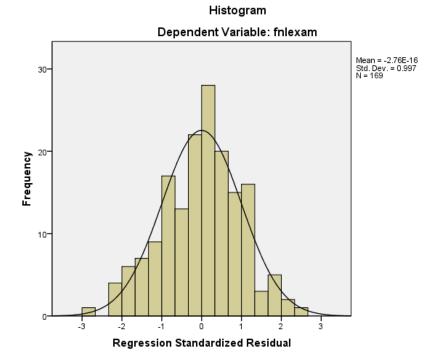
a. Dependent Variable: fnlexam

#### Residuals Statistics<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	63.61	67.90	65.14	2.057	169
Residual	-37.615	36.385	.000	13.842	169
Std. Predicted Value	740	1.344	.000	1.000	169
Std. Residual	-2.709	2.621	.000	.997	169

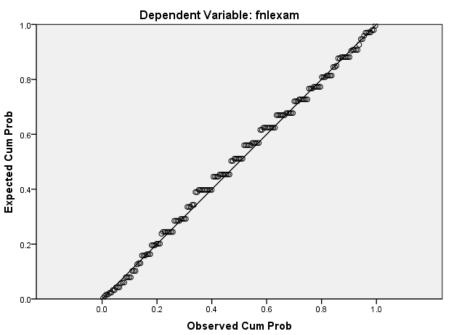
a. Dependent Variable: fnlexam

# Charts



215





#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER adj096 /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3). Regression

	Notes	
Output Created		02-Oct-2011 12:30:05
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	I I

Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER adj096
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:00.889
	Elapsed Time	00:00:01.102
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

Descriptive Statistics						
	Mean	Std. Deviation	Ν			
fnlexam	65.14	13.994	169			
adj096	.46	.500	169			

Correlations						
		fnlexam	adj096			
Pearson Correlation	fnlexam	1.000	.192			
	adj096	.192	1.000			
Sig. (1-tailed)	fnlexam		.006			
	adj096	.006				
Ν	fnlexam	169	169			
	adj096	169	169			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	adj096 <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.192 <sup>a</sup>	.037	.031	13.776	.037	6.373	1

a. Predictors: (Constant), adj096

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	167	.013	1.749			

	ANOVA <sup>b</sup>						
Model		Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	1209.482	1	1209.482	6.373	.013 <sup>a</sup>	
	Residual	31692.388	167	189.775			
	Total	32901.870	168				

a. Predictors: (Constant), adj096

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model		Standardized							
		Unstandardized Coefficients		Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	62.659	1.444		43.390	.000			
	adj096	5.366	2.126	.192	2.525	.013			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	59.808	65.510			
	adj096	1.170	9.563	.192	.192	.192

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity Statistics			
		Tolerance	VIF		
1	(Constant)				
	adj096	1.000	1.000		

#### Coefficient Correlations<sup>a</sup>

Model			adj096
1	Correlations	adj096	1.000
	Covariances	adj096	4.518

a. Dependent Variable: fnlexam

### Collinearity Diagnostics<sup>a</sup>

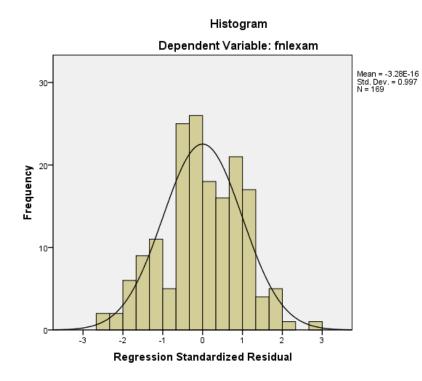
Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	adj096
1	1	1.679	1.000	.16	.16
	2	.321	2.289	.84	.84

a. Dependent Variable: fnlexam

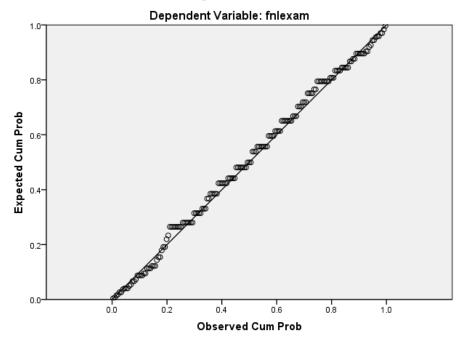
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	62.66	68.03	65.14	2.683	169
Residual	-36.659	37.341	.000	13.735	169
Std. Predicted Value	923	1.077	.000	1.000	169
Std. Residual	-2.661	2.711	.000	.997	169

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER mozartuse /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes					
Output Created		02-Oct-2011 12:32:01			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet2			
	Filter	<none></none>			
	Weight	<none></none>			
	Split File	<none></none>			
	N of Rows in Working Data	252			
	File				
Missing Value Handling	Definition of Missing	User-defined missing values are			
		treated as missing.			
	Cases Used	Statistics are based on cases with no			
		missing values for any variable used.			

Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER mozartuse /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID)
Dessuress		00-00-00 052
Resources	Processor Time	00:00:00.952
	Elapsed Time	00:00:01.171
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

Descriptive Statistics						
Maan Std Deviation N						
	Mean	Std. Deviation	N			
fnlexam	65.14	13.994	169			
mozartuse	.18	.383	169			

Correlations						
		fnlexam	mozartuse			
Pearson Correlation	fnlexam	1.000	.042			
	mozartuse	.042	1.000			
Sig. (1-tailed)	fnlexam		.293			
	mozartuse	.293				
Ν	fnlexam	169	169			

Correlations							
		fnlexam	mozartuse				
Pearson Correlation	fnlexam	1.000	.042				
	mozartuse	.042	1.000				
Sig. (1-tailed)	fnlexam		.293				
	mozartuse	.293	-				
Ν	fnlexam	169	169				
	mozartuse	169	169				

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	mozartuse <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.042 <sup>a</sup>	.002	004	14.024	.002	.296	1

a. Predictors: (Constant), mozartuse

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2	Sig. F Change	Durbin-Watson				
1	167	.587	1.707				

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	58.267	1	58.267	.296	.587 <sup>a</sup>		
	Residual	32843.603	167	196.668				
	Total	32901.870	168					

a. Predictors: (Constant), mozartuse

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
		Unstandardize	ed Coefficients	Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	64.863	1.189		54.530	.000			
	mozartuse	1.537	2.823	.042	.544	.587			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B Correlations					
Lower Bound		Lower Bound	Upper Bound	Zero-order	Partial	Part	
1	1 (Constant)	62.515	67.212				
	mozartuse	-4.037	7.110	.042	.042	.042	

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity Statistics			
		Tolerance VIF			
1	(Constant)				
	mozartuse	1.000	1.000		

### Coefficient Correlations<sup>a</sup>

Model			mozartuse
1	Correlations	mozartuse	1.000
	Covariances	mozartuse	7.970

a. Dependent Variable: fnlexam

### Collinearity Diagnostics<sup>a</sup>

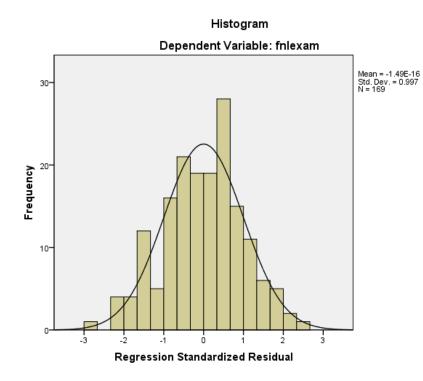
Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	mozartuse
1	1	1.421	1.000	.29	.29
	<sup>-</sup> 2	.579	1.567	.71	.71

a. Dependent Variable: fnlexam

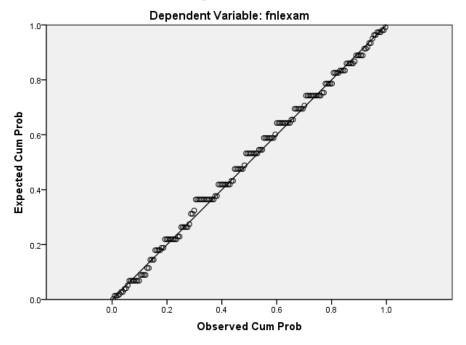
### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	64.86	66.40	65.14	.589	169
Residual	-40.400	33.600	.000	13.982	169
Std. Predicted Value	463	2.146	.000	1.000	169
Std. Residual	-2.881	2.396	.000	.997	169

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER amisone /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 12:33:49
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER amisone /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:01.092
	Elapsed Time	00:00:01.173
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
1	for Residual Plots	

Descriptive Statistics					
Mean Std. Deviation N					
fnlexam	65.24	13.973	168		
amisone	.36	.481	168		

Correlations				
		fnlexam	amisone	
Pearson Correlation	fnlexam	1.000	.018	
	amisone	.018	1.000	
Sig. (1-tailed)	fnlexam		.411	
	amisone	.411		
Ν	fnlexam	168	168	

Correlations					
		fnlexam	amisone		
Pearson Correlation	fnlexam	1.000	.018		
	amisone	.018	1.000		
Sig. (1-tailed)	fnlexam		.411		
	amisone	.411	-		
Ν	fnlexam	168	168		
	amisone	168	168		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	amisone <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.018 <sup>a</sup>	.000	006	14.013	.000	.051	1

a. Predictors: (Constant), amisone

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>				
Model	Chang	ge Statistics		
	df2	Durbin-Watson		
1	166	.821	1.705	

	ANOVA <sup>b</sup>						
Mode	9	Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	10.076	1	10.076	.051	.821 <sup>a</sup>	
	Residual	32596.400	166	196.364			
	Total	32606.476	167				

a. Predictors: (Constant), amisone

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>							
Model		Unstandardize	ed Coefficients	Standardized Coefficients				
		В	Std. Error	Beta	t	Sig.		
1	(Constant)	65.056	1.348		48.246	.000		
	amisone	.511	2.256	.018	.227	.821		

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
	1 (Constant)	62.393	67.718			
	amisone	-3.944	4.966	.018	.018	.018

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model	Model Collinearity Statistics			
Tole		Tolerance	VIF	
1	(Constant)			
	amisone	1.000	1.000	

#### **Coefficient Correlations**<sup>a</sup>

Model			amisone
1	Correlations	amisone	1.000
	Covariances	amisone	5.091

a. Dependent Variable: fnlexam

### Collinearity Diagnostics<sup>a</sup>

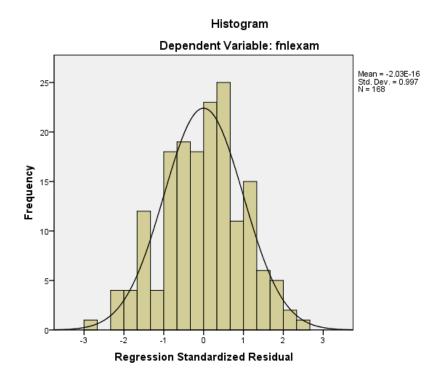
Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	amisone
1	1	1.598	1.000	.20	.20
	- 2	.402	1.993	.80	.80

a. Dependent Variable: fnlexam

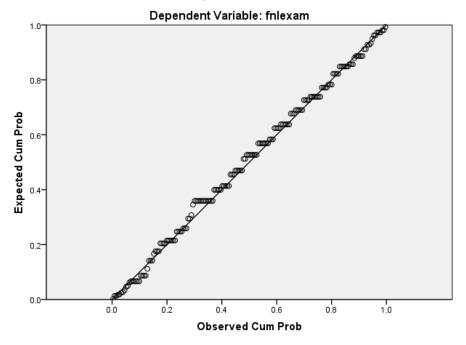
# **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	65.06	65.57	65.24	.246	168
Residual	-39.567	34.433	.000	13.971	168
Std. Predicted Value	743	1.338	.000	1.000	168
Std. Residual	-2.824	2.457	.000	.997	168

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER numbmeet /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 12:35:48
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet2
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER numbmeet /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID)
		/CASEWISE PLOT(ZRESID) OUTLIERS(3).
Dessuress	Processor Time	00.00.00.020
Resources		00:00:00.920
	Elapsed Time	00:00:01.108
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

Descriptive Statistics					
	M				
	Mean	Std. Deviation	N		
fnlexam	65.14	13.994	169		
numbmeet	3.56	1.326	169		

	Correlations	6	
		fnlexam	numbmeet
Pearson Correlation	fnlexam	1.000	172
	numbmeet	172	1.000
Sig. (1-tailed)	fnlexam		.012
	numbmeet	.012	
Ν	fnlexam	169	169

#### Correlation

Correlations						
		fnlexam	numbmeet			
Pearson Correlation	fnlexam	1.000	172			
	numbmeet	172	1.000			
Sig. (1-tailed)	fnlexam		.012			
	numbmeet	.012				
Ν	fnlexam	169	169			
	numbmeet	169	169			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	numbmeet <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.172 <sup>a</sup>	.030	.024	13.826	.030	5.121	1

a. Predictors: (Constant), numbmeet

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	167	.025	1.755			

	ANOVA <sup>b</sup>						
Model		Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	978.924	1	978.924	5.121	.025 <sup>a</sup>	
	Residual	31922.946	167	191.155			
	Total	32901.870	168				

a. Predictors: (Constant), numbmeet

b. Dependent Variable: fnlexam

Coefficients <sup>a</sup>	
---------------------------	--

Ν	lodel			Standardized		
		Unstandardize	ed Coefficients	Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	71.618	3.056		23.439	.000
	numbmeet	-1.820	.804	172	-2.263	.025

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

M	odel	95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	65.586	77.651			
	numbmeet	-3.407	232	172	172	172

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>						
Model		Collinearity Statistics				
		Tolerance	VIF			
1	(Constant)					
	numbmeet	1.000	1.000			

Coefficient Correlations <sup>a</sup>						
Model	Model numbmeet					
1	Correlations	numbmeet	1.000			
	Covariances	numbmeet	.647			

a. Dependent Variable: fnlexam

Collinearity	/ Diagnostics <sup>a</sup>
Commeanity	Diagnostics

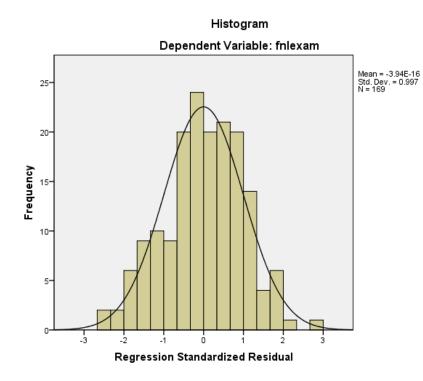
-					
Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	numbmeet
1	1	1.937	1.000	.03	.03
	2	.063	5.566	.97	.97

a. Dependent Variable: fnlexam

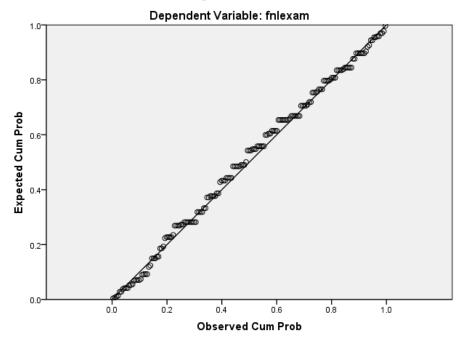
**Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	62.52	67.98	65.14	2.414	169
Residual	-36.519	37.481	.000	13.785	169
Std. Predicted Value	-1.084	1.178	.000	1.000	169
Std. Residual	-2.641	2.711	.000	.997	169

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER classize /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes					
Output Created		02-Oct-2011 12:37:35			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet2			
	Filter	<none></none>			
	Weight	<none></none>			
	Split File	<none></none>			
	N of Rows in Working Data	252			
	File				
Missing Value Handling	Definition of Missing	User-defined missing values are			
		treated as missing.			
	Cases Used	Statistics are based on cases with no			
		missing values for any variable used.			

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Syntax		REGRESSION	
5		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER classize	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:01.185	
	Elapsed Time	00:00:01.066	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

[DataSet2] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

## **Descriptive Statistics**

	Mean	Std. Deviation	Ν
fnlexam	65.14	13.994	169
classize	34.07	12.132	169

Correlations						
		fnlexam	classize			
Pearson Correlation	fnlexam	1.000	.106			
	classize	.106	1.000			
Sig. (1-tailed)	fnlexam		.085			
	classize	.085				
Ν	fnlexam	169	169			
	classize	169	169			

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	classize <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.106 <sup>a</sup>	.011	.005	13.957	.011	1.896	1

a. Predictors: (Constant), classize

b. Dependent Variable: fnlexam

Model \$	Summary <sup>b</sup>
----------	----------------------

Model	Chang		
	df2 Sig. F Change		Durbin-Watson
1	167	.170	1.736

### b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
Mode	el	Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	369.390	1	369.390	1.896	.170 <sup>a</sup>		
	Residual	32532.480	167	194.805				
	Total	32901.870	168					

a. Predictors: (Constant), classize

	Coefficients <sup>a</sup>							
Mod	el			Standardized				
		Unstandardized Coefficients		Coefficients				
		В	Std. Error	Beta	t	Sig.		
1	(Constant)	60.972	3.209		19.001	.000		
	classize	.122	.089	.106	1.377	.170		

a. Dependent Variable: fnlexam

#### Coefficients<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	54.637	67.307			
	classize	053	.297	.106	.106	.106

a. Dependent Variable: fnlexam

## **Coefficients**<sup>a</sup>

Mod	el	Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	classize	1.000	1.000	

a. Dependent Variable: fnlexam

# Coefficient Correlations<sup>a</sup>

Model			classize
1	Correlations	classize	1.000
	Covariances	classize	.008

Model	Dimension			Variance Proportions				
		Eigenvalue	Condition Index	(Constant)	classize			
1	1	1.942	1.000	.03	.03			
	<sup>-</sup> 2	.058	5.805	.97	.97			

### **Collinearity Diagnostics**<sup>a</sup>

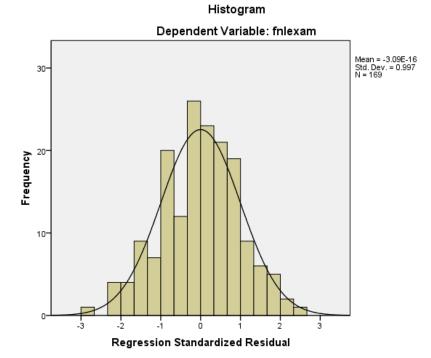
a. Dependent Variable: fnlexam

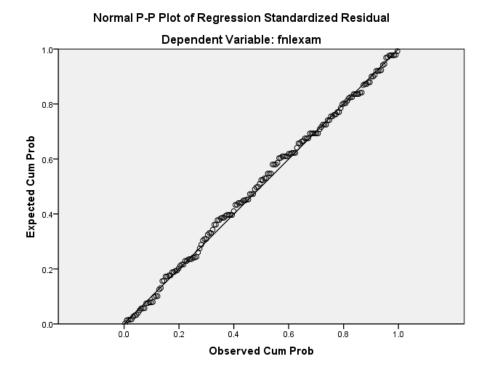
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	62.32	66.96	65.14	1.483	169
Residual	-40.350	33.650	.000	13.916	169
Std. Predicted Value	-1.902	1.230	.000	1.000	169
Std. Residual	-2.891	2.411	.000	.997	169

a. Dependent Variable: fnlexam

# Charts





## GET

FILE='C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav'.

DATASET NAME DataSet1 WINDOW=FRONT.

REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N

/MISSING LISTWISE

/STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP

/CRITERIA=PIN(.05) POUT(.10)

/NOORIGIN

/DEPENDENT fnlexam

/METHOD=ENTER gender

/RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes						
Output Created		02-Oct-2011 13:34:52				
Comments						
Input	Data	C:\Users\Lin\Documents\math 097 fall				
		2001 with classrooms no names.sav				
	Active Dataset	DataSet1				

	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER gender
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:02.152
	Elapsed Time	00:00:01.872
	Memory Required	2260 bytes
	Additional Memory Required for Residual Plots	656 bytes

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics						
	Mean	Std. Deviation	Ν			
fnlexam 68.55 14.175						

Descriptive Statistics						
	Mean Std. Deviation					
fnlexam	68.55	14.175	508			
gender	508					

Correlations						
		fnlexam	gender			
Pearson Correlation	fnlexam	1.000	149			
	gender	149	1.000			
Sig. (1-tailed)	fnlexam		.000			
	gender	.000				
Ν	fnlexam	508	508			
	gender	508	508			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	gender <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.149 <sup>a</sup>	.022	.020	14.031	.022	11.425	1

a. Predictors: (Constant), gender

b. Dependent Variable: fnlexam

# Model Summary<sup>b</sup>

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	506	.001	1.659

Model Summary <sup>b</sup>					
Model	Chang	ge Statistics			
	df2	Sig. F Change	Durbin-Watson		
1	506	.001	1.659		

#### b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>						
Model		Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	2249.389	1	2249.389	11.425	.001 <sup>a</sup>	
	Residual	99622.176	506	196.882			
	Total	101871.565	507				

### a. Predictors: (Constant), gender

b. Dependent Variable: fnlexam

### **Coefficients**<sup>a</sup>

Mode	el	Unstandardize	ed Coefficients	Standardized Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	70.214	.793		88.531	.000
	gender	-4.327	1.280	149	-3.380	.001

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Mode	el	95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	68.656	71.772			
	gender	-6.842	-1.812	149	149	149

Coefficients <sup>a</sup>				
Model		Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	gender	1.000	1.000	

a. Dependent Variable: fnlexam

## **Coefficient Correlations**<sup>a</sup>

Model			gender
1	Correlations	gender	1.000
	Covariances	gender	1.639

a. Dependent Variable: fnlexam

# Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	gender
1	1	1.620	1.000	.19	.19
	<sup>-</sup> 2	.380	2.063	.81	.81

a. Dependent Variable: fnlexam

## Casewise Diagnostics<sup>a</sup>

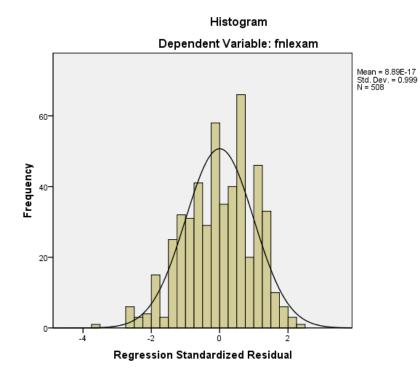
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
_ 551	-3.579	20	70.21	-50.214

a. Dependent Variable: fnlexam

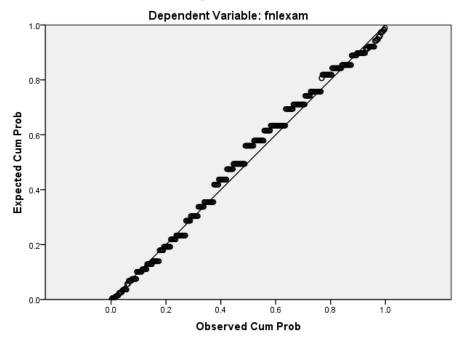
### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	65.89	70.21	68.55	2.106	508
Residual	-50.214	32.113	.000	14.018	508
Std. Predicted Value	-1.266	.789	.000	1.000	508
Std. Residual	-3.579	2.289	.000	.999	508

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER act /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:36:59
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER act
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
_		
Resources	Processor Time	00:00:01.014
	Elapsed Time	00:00:01.256
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics
------------------------

	Mean	Std. Deviation	Ν
fnlexam	68.39	14.223	470
act	16.83	1.156	470

Correlations						
		fnlexam	act			
Pearson Correlation	fnlexam	1.000	.240			
	act	.240	1.000			
Sig. (1-tailed)	fnlexam		.000			
	act	.000				
Ν	fnlexam	470	470			
	act	470	470			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	act <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.240 <sup>a</sup>	.058	.056	13.820	.058	28.724	1

a. Predictors: (Constant), act

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang						
	df2	Sig. F Change	Durbin-Watson				
1	468	.000	1.694				

	ANOVA <sup>b</sup>						
Model		Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	5486.429	1	5486.429	28.724	.000 <sup>a</sup>	
	Residual	89390.866	468	191.006			
	Total	94877.296	469				

a. Predictors: (Constant), act

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model			Standardized						
		Unstandardize	ed Coefficients	Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	18.577	9.315		1.994	.047			
	act	2.960	.552	.240	5.359	.000			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	.271	36.882			
	act	1.874	4.045	.240	.240	.240

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model Collinearity Statistics					
		Tolerance	VIF		
1	(Constant)				
	act	1.000	1.000		

#### **Coefficient Correlations**<sup>a</sup>

Model			act
1	Correlations	act	1.000
	Covariances	act	.305

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Pr	roportions
		Eigenvalue	Condition Index	(Constant)	act
1	1	1.998	1.000	.00	.00
	- 2	.002	29.191	1.00	1.00

a. Dependent Variable: fnlexam

### Casewise Diagnostics<sup>a</sup>

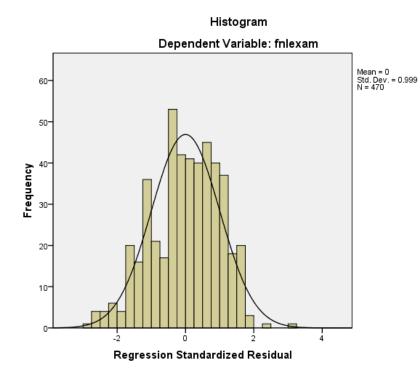
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.108	100	57.05	42.949

a. Dependent Variable: fnlexam

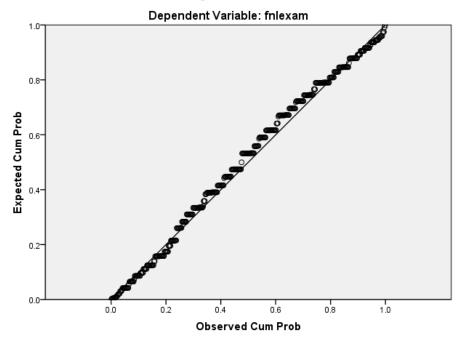
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	54.09	83.69	68.39	3.420	470
Residual	-40.889	42.949	.000	13.806	470
Std. Predicted Value	-4.179	4.474	.000	1.000	470
Std. Residual	-2.959	3.108	.000	.999	470

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER sat /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:38:30
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION	
- ,		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER sat	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:01.123	
	Elapsed Time	00:00:01.087	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics
------------------------

	Mean	Std. Deviation	N
fnlexam	68.73	15.163	64
sat	412.19	46.886	64

Correlations						
		fnlexam	sat			
Pearson Correlation	fnlexam	1.000	.285			
	sat	.285	1.000			
Sig. (1-tailed)	fnlexam		.011			
	sat	.011				
Ν	fnlexam	64	64			
	sat	64	64			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	sat <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Chai	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.285 <sup>a</sup>	.081	.067	14.649	.081	5.493	1

a. Predictors: (Constant), sat

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2	Sig. F Change	Durbin-Watson				
1	62	.022	1.947				

			ANOVA <sup>b</sup>			
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	1178.801	1	1178.801	5.493	.022 <sup>a</sup>
	Residual	13305.684	62	214.608		
	Total	14484.484	63			

a. Predictors: (Constant), sat

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
		Unstandardize	ed Coefficients	Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	30.707	16.329		1.881	.065			
	sat	.092	.039	.285	2.344	.022			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Мос	del	95.0% Confidence Interval for B		Correlations		
		Lower Bound Upper Bound		Zero-order Partial		Part
1	(Constant)	-1.934	63.347			
	sat	.014	.171	.285	.285	.285

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model Collinearity Statistics			Statistics	
		Tolerance	VIF	
1	(Constant)			
	sat	1.000	1.000	

#### **Coefficient Correlations**<sup>a</sup>

Model			sat
1	Correlations	sat	1.000
	Covariances	sat	.002

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

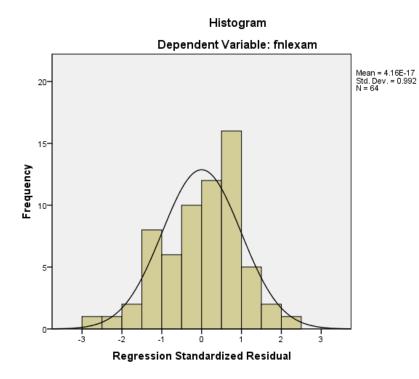
Model	Dimension		Variance Proportic		oportions
		Eigenvalue	Condition Index	(Constant)	sat
1	1	1.994	1.000	.00	.00
	<sup></sup> 2	.006	17.778	1.00	1.00

a. Dependent Variable: fnlexam

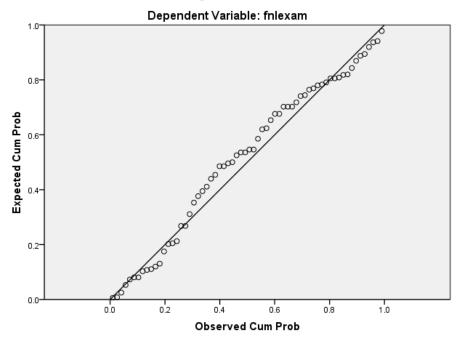
## **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	N
Predicted Value	54.69	79.60	68.73	4.326	64
Residual	-36.687	29.467	.000	14.533	64
Std. Predicted Value	-3.246	2.513	.000	1.000	64
Std. Residual	-2.504	2.011	.000	.992	64

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /<u>DEPENDENT fnlexam</u> /<u>METHOD=ENTER comcol</u> /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:41:00
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER comcol /RESIDUALS DURBIN HISTOGRAM(ZRESID)
		NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:01.217
	Elapsed Time	00:00:01.208
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics				
	Mean	Std. Deviation	Ν	
fnlexam	68.52	14.194	517	
comcol	.18	.384	517	

Correlations				
		fnlexam	comcol	
Pearson Correlation	fnlexam	1.000	088	
	comcol	088	1.000	
Sig. (1-tailed)	fnlexam		.023	
	comcol	.023	-	
Ν	fnlexam	517	517	
	comcol	517	517	

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	comcol <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.088 <sup>a</sup>	.008	.006	14.152	.008	4.018	1

a. Predictors: (Constant), comcol

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>					
Model	Chang	ge Statistics			
	df2	Sig. F Change	Durbin-Watson		
1	515	.046	1.678		

			ANOVA <sup>b</sup>			
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	804.756	1	804.756	4.018	.046 <sup>a</sup>
	Residual	103150.192	515	200.292		
	Total	103954.948	516			

a. Predictors: (Constant), comcol

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>							
Model				Standardized				
		Unstandardized Coefficients		Coefficients				
		В	Std. Error	Beta	t	Sig.		
1	(Constant)	69.108	.687		100.550	.000		
	comcol	-3.248	1.621	088	-2.004	.046		

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	nfidence Interval for B Correlations		_	
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	67.758	70.459			
	comcol	-6.432	065	088	088	088

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model	Model Collinearity Statistics			
		Tolerance	VIF	
1	(Constant)			
	comcol	1.000	1.000	

### **Coefficient Correlations**<sup>a</sup>

Model			comcol
1	Correlations	comcol	1.000
	Covariances	comcol	2.626

a. Dependent Variable: fnlexam

## **Collinearity Diagnostics**<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	comcol
1	1	1.424	1.000	.29	.29
	<sup></sup> 2	.576	1.573	.71	.71

a. Dependent Variable: fnlexam

### Casewise Diagnostics<sup>a</sup>

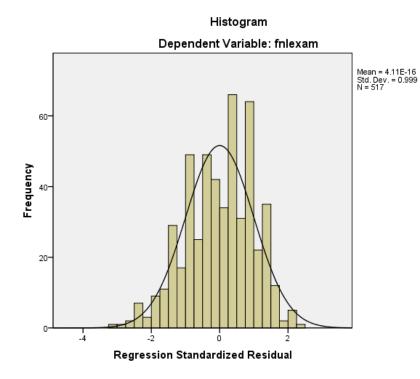
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
_ 551	-3.240	20	65.86	-45.860

a. Dependent Variable: fnlexam

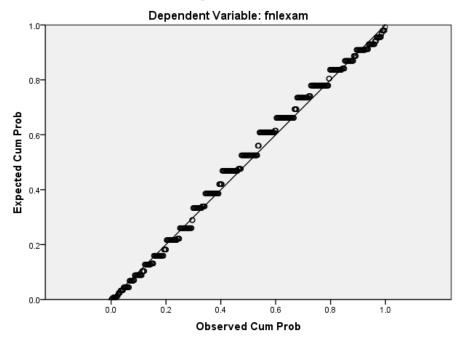
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	65.86	69.11	68.52	1.249	517
Residual	-45.860	34.140	.000	14.139	517
Std. Predicted Value	-2.133	.468	.000	1.000	517
Std. Residual	-3.240	2.412	.000	.999	517

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /<u>DEPENDENT fnlexam</u> /METHOD=ENTER pretest /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:42:12
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION	
		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER pretest	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
5			
Resources	Processor Time	00:00:01.092	
	Elapsed Time	00:00:01.257	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

### **Descriptive Statistics**

	Mean	Std. Deviation	Ν
fnlexam	68.87	14.561	434
pretest	47.03	14.554	434

Correlations					
		fnlexam	pretest		
Pearson Correlation	fnlexam	1.000	.294		
	pretest	.294	1.000		
Sig. (1-tailed)	fnlexam		.000		
	pretest	.000			
Ν	fnlexam	434	434		
	pretest	434	434		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables		
	Entered	Removed	Method	
1	pretest <sup>a</sup>		Enter	

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.294 <sup>a</sup>	.087	.085	13.933	.087	40.966	1

a. Predictors: (Constant), pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang						
	df2	Sig. F Change	Durbin-Watson				
1	432	.000	1.603				

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	7952.318	1	7952.318	40.966	.000 <sup>a</sup>		
	Residual	83859.712	432	194.120				
	Total	91812.030	433					

a. Predictors: (Constant), pretest

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
		Unstandardized Coefficients		Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	55.026	2.265		24.299	.000			
	pretest	.294	.046	.294	6.400	.000			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Мо	odel	95.0% Confidence Interval for B		Correlations		_
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	50.575	59.476			
	pretest	.204	.385	.294	.294	.294

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity Statistics			
		Tolerance	VIF		
1	(Constant)				
	pretest	1.000	1.000		

### Coefficient Correlations<sup>a</sup>

Model			pretest
1	Correlations	pretest	1.000
	Covariances	pretest	.002

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	pretest
1	1	1.955	1.000	.02	.02
	- 2	.045	6.621	.98	.98

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

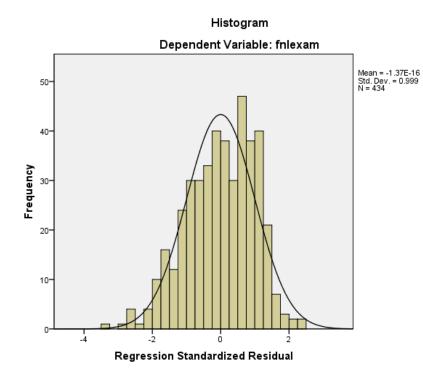
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.254	20	65.33	-45.332

a. Dependent Variable: fnlexam

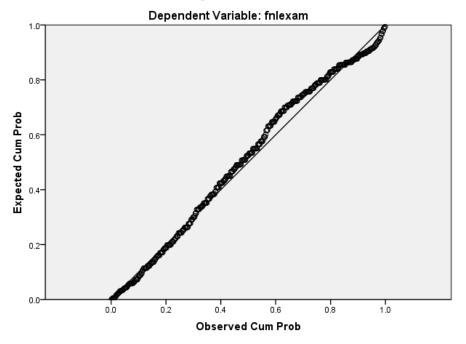
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	56.50	80.05	68.87	4.286	434
Residual	-45.332	34.668	.000	13.917	434
Std. Predicted Value	-2.888	2.609	.000	1.000	434
Std. Residual	-3.254	2.488	.000	.999	434

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER ascgr /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:44:57
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER ascgr
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
Resources	Processor Time	00:00:01.076
	Elapsed Time	00:00:01.853
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics
------------------------

	Mean	Std. Deviation	N
fnlexam	68.55	14.212	514
ascgr	.86	.347	514

Correlations					
		fnlexam	ascgr		
Pearson Correlation	fnlexam	1.000	.170		
	ascgr	.170	1.000		
Sig. (1-tailed)	fnlexam		.000		
	ascgr	.000			
Ν	fnlexam	514	514		
	ascgr	514	514		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	ascgr <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Chai	nge Statistics		
			Adjusted R	Std. Error of the	R Square			
	R	R Square	Square	Estimate	Change	F Change	df1	
1	.170 <sup>a</sup>	.029	.027	14.019	.029	15.179	1	1

a. Predictors: (Constant), ascgr

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2 Sig. F Change		Durbin-Watson			
1	512	.000	1.699			

## ANOVA<sup>b</sup>

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	2983.355	1	2983.355	15.179	.000 <sup>a</sup>
	Residual	100627.726	512	196.539		
	Total	103611.082	513			

a. Predictors: (Constant), ascgr

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>					
Model				Standardized		
		Unstandardized Coefficients		Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	62.583	1.652		37.879	.000
	ascgr	6.942	1.782	.170	3.896	.000

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
	1 (Constant)	59.337	65.829			
	ascgr	3.441	10.442	.170	.170	.170

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity	Statistics		
		Tolerance	VIF		
1	(Constant)				
	ascgr	1.000	1.000		

### **Coefficient Correlations**<sup>a</sup>

Model			ascgr
1	Correlations	ascgr	1.000
	Covariances	ascgr	3.174

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	ascgr
1	1	1.927	1.000	.04	.04
	<sup>-</sup> 2	.073	5.150	.96	.96

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

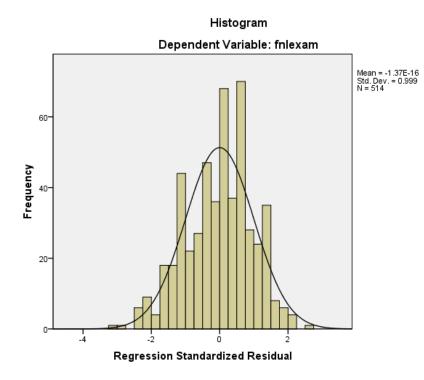
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
_ 551	-3.037	20	62.58	-42.583

a. Dependent Variable: fnlexam

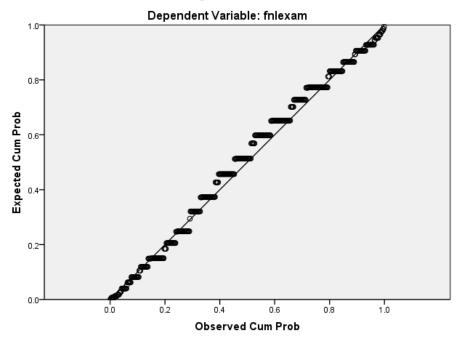
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	62.58	69.52	68.55	2.412	514
Residual	-42.583	35.417	.000	14.006	514
Std. Predicted Value	-2.475	.403	.000	1.000	514
Std. Residual	-3.037	2.526	.000	.999	514

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER techsex /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:46:52
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER techsex
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
_		
Resources	Processor Time	00:00:00.983
	Elapsed Time	00:00:01.163
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

<b>Descriptive Statistics</b>	
-------------------------------	--

	Mean	Std. Deviation	Ν
fnlexam	68.50	14.192	518
techsex	.31	.462	518

Correlations					
		fnlexam	techsex		
Pearson Correlation	fnlexam	1.000	098		
	techsex	098	1.000		
Sig. (1-tailed)	fnlexam		.013		
	techsex	.013	-		
Ν	fnlexam	518	518		
	techsex	518	518		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	techsex <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.098 <sup>a</sup>	.010	.008	14.138	.010	4.980	1

a. Predictors: (Constant), techsex

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Durbin-Watson				
1	516	.026	1.678			

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	995.407	1	995.407	4.980	.026 <sup>a</sup>		
	Residual	103142.091	516	199.888				
	Total	104137.498	517					

a. Predictors: (Constant), techsex

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
Unstandardized Coefficients		Coefficients							
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	69.421	.746		93.034	.000			
	techsex	-3.006	1.347	098	-2.232	.026			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	67.955	70.887			
	techsex	-5.651	360	098	098	098

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model Collinearity Statistics			Statistics	
		Tolerance	VIF	
1	(Constant)			
	techsex	1.000	1.000	

### **Coefficient Correlations**<sup>a</sup>

Model			techsex
1	Correlations	techsex	1.000
	Covariances	techsex	1.814

a. Dependent Variable: fnlexam

# Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	techsex
1	1	1.554	1.000	.22	.22
	<sup></sup> 2	.446	1.867	.78	.78

a. Dependent Variable: fnlexam

\

## Casewise Diagnostics<sup>a</sup>

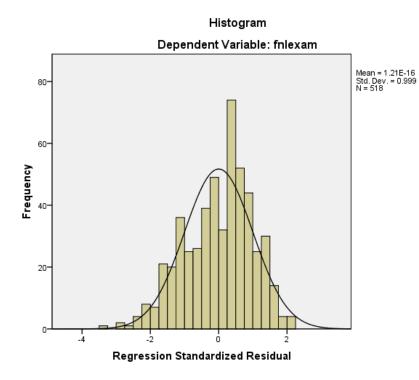
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.496	20	69.42	-49.421

a. Dependent Variable: fnlexam

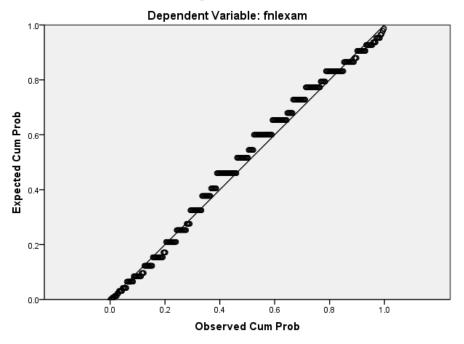
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	66.42	69.42	68.50	1.388	518
Residual	-49.421	31.585	.000	14.124	518
Std. Predicted Value	-1.501	.665	.000	1.000	518
Std. Residual	-3.496	2.234	.000	.999	518

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER adj097 /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes				
Output Created		02-Oct-2011 13:48:35		
Comments				
Input	Data	C:\Users\Lin\Documents\math 097 fall		
		2001 with classrooms no names.sav		
	Active Dataset	DataSet1		
	Filter	<none></none>		
	Weight	<none></none>		
	Split File	<none></none>		
	N of Rows in Working Data	675		
	File			
Missing Value Handling	Definition of Missing	User-defined missing values are		
		treated as missing.		
	Cases Used	Statistics are based on cases with no		
		missing values for any variable used.		

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Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER adj097 /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID)
		NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).
Resources	Processor Time	00:00:01.372
	Elapsed Time	00:00:01.280
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics						
	Mean	Std. Deviation	N			
	INICALL	Stu. Deviation	IN			
fnlexam	68.50	14.192	518			
adj097	.49	.500	518			

Correlations					
		fnlexam	adj097		
Pearson Correlation	fnlexam	1.000	017		
	adj097	017	1.000		
Sig. (1-tailed)	fnlexam		.351		
	adj097	.351			
Ν	fnlexam	518	518		
	adj097	518	518		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	adj097 <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.017 <sup>a</sup>	.000	002	14.204	.000	.147	1

a. Predictors: (Constant), adj097

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang					
	df2	Sig. F Change	Durbin-Watson			
1	516	.701	1.662			

	ANOVA <sup>b</sup>						
Mode	el	Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	29.710	1	29.710	.147	.701 <sup>a</sup>	
	Residual	104107.788	516	201.759			
	Total	104137.498	517				

a. Predictors: (Constant), adj097

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>					
Mode	I	Unstandardized Coefficients		Standardized Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	68.732	.873		78.771	.000
	adj097	479	1.249	017	384	.701

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confiden	ce Interval for B	Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	67.018	70.446			
	adj097	-2.932	1.974	017	017	017

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model		Collinearity Statistics		
		Tolerance	VIF	
1	(Constant)			
	adj097	1.000	1.000	

### Coefficient Correlations<sup>a</sup>

Model			adj097
1	Correlations	adj097	1.000
	Covariances	adj097	1.559

a. Dependent Variable: fnlexam

## **Collinearity Diagnostics**<sup>a</sup>

Model	Dimension			Variance Pr	oportions
		Eigenvalue	Condition Index	(Constant)	adj097
1	1	1.699	1.000	.15	.15
	<sup>-</sup> 2	.301	2.375	.85	.85

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

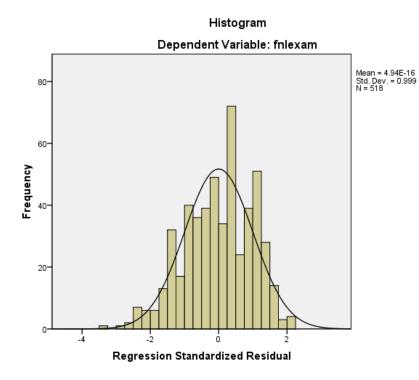
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
_ 551	-3.397	20	68.25	-48.253

a. Dependent Variable: fnlexam

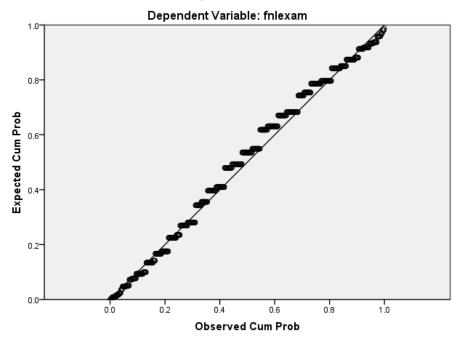
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	68.25	68.73	68.50	.240	518
Residual	-48.253	31.268	.000	14.190	518
Std. Predicted Value	-1.022	.976	.000	1.000	518
Std. Residual	-3.397	2.201	.000	.999	518

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER mozartuse /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:50:09
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER mozartuse
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3).
_		
Resources	Processor Time	00:00:00.967
	Elapsed Time	00:00:01.217
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

<b>Descriptive Statistics</b>	;
-------------------------------	---

	Mean	Std. Deviation	Ν
fnlexam	68.50	14.192	518
mozartuse	.11	.316	518

Correlations					
		fnlexam	mozartuse		
Pearson Correlation	fnlexam	1.000	.124		
	mozartuse	.124	1.000		
Sig. (1-tailed)	fnlexam		.002		
	mozartuse	.002			
Ν	fnlexam	518	518		
	mozartuse	518	518		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	mozartuse <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Chai	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.124 <sup>a</sup>	.015	.014	14.096	.015	8.111	1

a. Predictors: (Constant), mozartuse

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	516	.005	1.689			

	ANOVA <sup>b</sup>					
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	1611.634	1	1611.634	8.111	.005 <sup>a</sup>
	Residual	102525.864	516	198.694		
	Total	104137.498	517			

a. Predictors: (Constant), mozartuse

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>							
Model		Unstandardized Coefficients		Standardized Coefficients				
		Unstanuaruize		COEfficients				
		В	Std. Error	Beta	t	Sig.		
1	(Constant)	67.872	.657		103.270	.000		
	mozartuse	5.594	1.964	.124	2.848	.005		

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
ſ	1 (Constant)	66.581	69.163			
L	mozartuse	1.735	9.452	.124	.124	.124

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity	Statistics		
		Tolerance	VIF		
1	(Constant)				
	mozartuse	1.000	1.000		

### Coefficient Correlations<sup>a</sup>

Model			mozartuse
1	Correlations	mozartuse	1.000
	Covariances	mozartuse	3.858

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance F	Proportions
		Eigenvalue	Condition Index	(Constant)	mozartuse
1	1	1.335	1.000	.33	.33
	- 2	.665	1.416	.67	.67

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

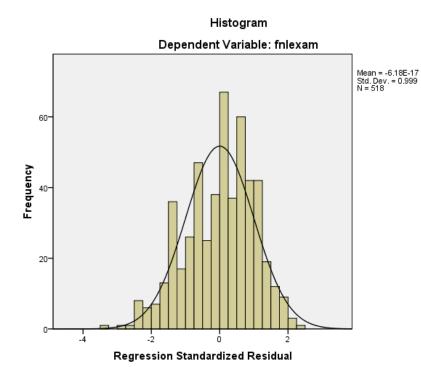
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.396	20	67.87	-47.872

a. Dependent Variable: fnlexam

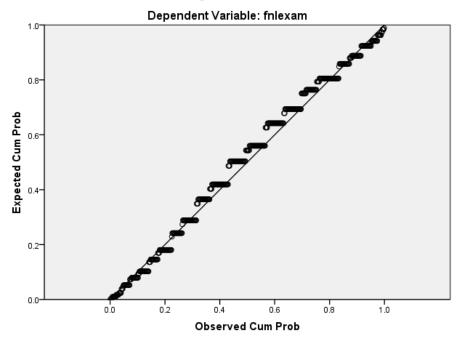
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	67.87	73.47	68.50	1.766	518
Residual	-47.872	32.128	.000	14.082	518
Std. Predicted Value	355	2.813	.000	1.000	518
Std. Residual	-3.396	2.279	.000	.999	518

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN <u>/DEPENDENT fnlexam</u> /METHOD=ENTER ALEKSuse /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

Notes					
Output Created		02-Oct-2011 13:52:25			
Comments					
Input	Data	C:\Users\Lin\Documents\math 097 fall			
		2001 with classrooms no names.sav			
	Active Dataset	DataSet1			
	Filter	<none></none>			
	Weight	<none></none>			
	Split File	<none></none>			
	N of Rows in Working Data	675			
	File				
Missing Value Handling	Definition of Missing	User-defined missing values are			
		treated as missing.			
	Cases Used	Statistics are based on cases with no			
		missing values for any variable used.			

300

Syntax		REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER ALEKSuse /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID)
Resources	Processor Time	00:00:01.451
	Elapsed Time	00:00:01.298
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics						
	Mean	Std. Deviation	Ν			
fnlexam	68.50	14.192	518			
ALEKSuse	.04	.202	518			

Correlations					
		fnlexam	ALEKSuse		
Pearson Correlation	fnlexam	1.000	036		
	ALEKSuse	036	1.000		
Sig. (1-tailed)	fnlexam		.204		
	ALEKSuse	.204			
Ν	fnlexam	518	518		
	ALEKSuse	518	518		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	ALEKSuse <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.036 <sup>a</sup>	.001	001	14.197	.001	.686	1

a. Predictors: (Constant), ALEKSuse

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>								
Model	Chang	ge Statistics						
	df2	Sig. F Change	Durbin-Watson					
1	516	.408	1.663					

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	138.207	1	138.207	.686	.408 <sup>a</sup>		
	Residual	103999.291	516	201.549				
	Total	104137.498	517					

a. Predictors: (Constant), ALEKSuse

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>									
Model				Standardized						
		Unstandardize	ed Coefficients	Coefficients						
		В	Std. Error	Beta	t	Sig.				
1	(Constant)	68.607	.637		107.626	.000				
	ALEKSuse	-2.561	3.093	036	828	.408				

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1 (Constant)		67.355	69.859			
	ALEKSuse	-8.638	3.515	036	036	036

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>						
Model		Collinearity Statistics				
		Tolerance	VIF			
1	(Constant)					
	ALEKSuse	1.000	1.000			

### **Coefficient Correlations**<sup>a</sup>

Model			ALEKSuse
1	Correlations	ALEKSuse	1.000
	Covariances	ALEKSuse	9.568

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Nodel Dimension			Variance F	Proportions
		Eigenvalue	Condition Index	(Constant)	ALEKSuse
1	1	1.206	1.000	.40	.40
	- 2	.794	1.233	.60	.60

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

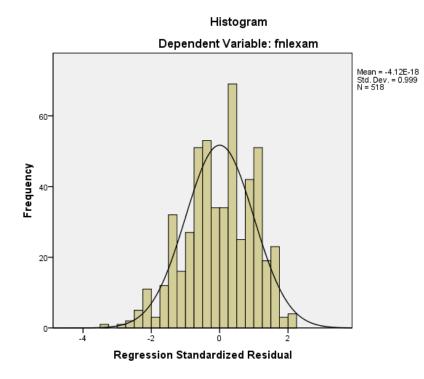
Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.424	20	68.61	-48.607

a. Dependent Variable: fnlexam

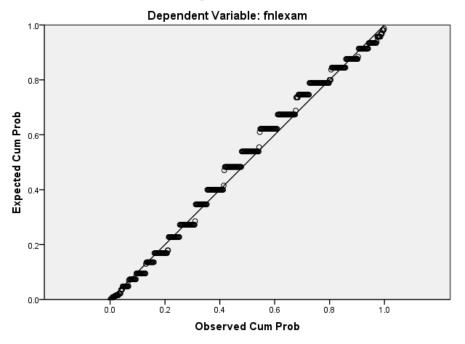
#### **Residuals Statistics**<sup>a</sup>

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	66.05	68.61	68.50	.517	518
Residual	-48.607	31.393	.000	14.183	518
Std. Predicted Value	-4.744	.210	.000	1.000	518
Std. Residual	-3.424	2.211	.000	.999	518

# Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER amisone /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

# Regression

	Notes	
Output Created		02-Oct-2011 13:54:32
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax	/DESCRIPTIVE CORR SIG N /MISSING LIST /STATISTICS O BCOV R ANOV/ CHANGE ZPP /CRITERIA=PII /NOORIGIN /DEPENDENT /METHOD=EN /RESIDUALS D HISTOGRAM(ZI NORMPROB(ZF)	
		HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).
Resources	Processor Time	00:00:01.014
	Elapsed Time	00:00:01.139
	Memory Required	2260 bytes
	Additional Memory Required	656 bytes
	for Residual Plots	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics						
	Mean	Ν				
fnlexam	68.63	14.173	510			
amisone	.36	.481	510			

Correlations				
		fnlexam	amisone	
Pearson Correlation	fnlexam	1.000	.020	
	amisone	.020	1.000	
Sig. (1-tailed)	fnlexam		.329	
	amisone	.329	-	
Ν	fnlexam	510	510	
	amisone	510	510	

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	amisone <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Char	nge Statistics		
			Adjusted R	Std. Error of the	R Square			
	R	R Square	Square	Estimate	Change	F Change	df1	
1	.020 <sup>a</sup>	.000	002	14.184	.000	.196		1

a. Predictors: (Constant), amisone

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>					
Model	Chang				
	df2	Sig. F Change	Durbin-Watson		
1	508	.658	1.648		

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>						
Mode	el	Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	39.449	1	39.449	.196	.658 <sup>a</sup>	
	Residual	102201.249	508	201.184			
	Total	102240.698	509				

a. Predictors: (Constant), amisone

b. Dependent Variable: fnlexam

#### **Coefficients**<sup>a</sup> Model Standardized Unstandardized Coefficients Coefficients В Std. Error Beta Sig. t 68.422 1 (Constant) .787 86.964 .000 amisone .578 1.306 .020 .443 .658

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Ν	lodel	95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	66.876	69.967			
	amisone	-1.988	3.145	.020	.020	.020

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model		Collinearity	Statistics	
		Tolerance	VIF	
1	(Constant)			
	amisone	1.000	1.000	

a. Dependent Variable: fnlexam

### **Coefficient Correlations**<sup>a</sup>

Model			amisone
1	Correlations	amisone	1.000
	Covariances	amisone	1.707

a. Dependent Variable: fnlexam

### Collinearity Diagnostics<sup>a</sup>

Mod	el Dimension			Variance Pr	roportions
		Eigenvalue	Condition Index	(Constant)	amisone
1	1	1.602	1.000	.20	.20
	<sup>-</sup> 2	.398	2.007	.80	.80

a. Dependent Variable: fnlexam

### Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.414	20	68.42	-48.422

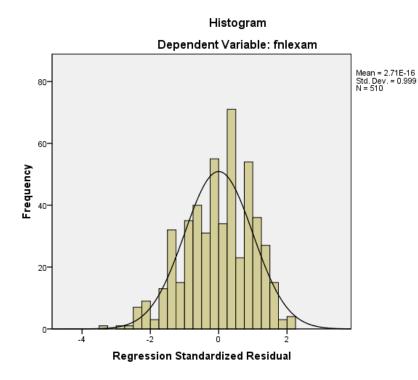
a. Dependent Variable: fnlexam

#### **Residuals Statistics**<sup>a</sup>

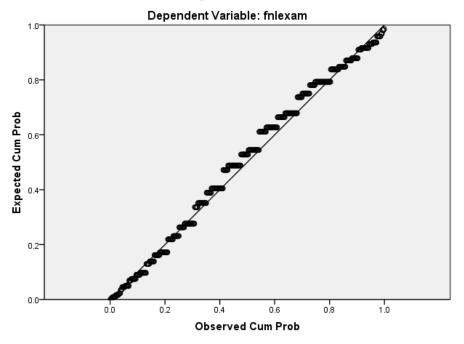
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	68.42	69.00	68.63	.278	510
Residual	-48.422	31.000	.000	14.170	510
Std. Predicted Value	754	1.324	.000	1.000	510
Std. Residual	-3.414	2.186	.000	.999	510

a. Dependent Variable: fnlexam

## Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER numbmeet /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

## Regression

	Notes	
Output Created		02-Oct-2011 13:55:43
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION	
		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER numbmeet	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:01.139	
	Elapsed Time	00:00:01.182	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

Descriptive Statistics					
	Mean	Std. Deviation			

	Mean	Std. Deviation	Ν
fnlexam	68.63	14.173	510
numbmeet	3.24	1.275	510

Correlations					
		fnlexam	numbmeet		
Pearson Correlation	fnlexam	1.000	.068		
	numbmeet	.068	1.000		
Sig. (1-tailed)	fnlexam		.062		
	numbmeet	.062			
Ν	fnlexam	510	510		
	numbmeet	510	510		

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	numbmeet <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.068 <sup>a</sup>	.005	.003	14.154	.005	2.380	1

a. Predictors: (Constant), numbmeet

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	508	.124	1.658			

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>					
Mod	lel	Sum of Squares	df	Mean Square	F	Sig.
1	Regression	476.802	1	476.802	2.380	.124 <sup>a</sup>
	Residual	101763.896	508	200.323		
	Total	102240.698	509			

a. Predictors: (Constant), numbmeet

b. Dependent Variable: fnlexam

Coefficients <sup>a</sup>
---------------------------

N	lodel			Standardized		
		Unstandardize	ed Coefficients	Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	66.176	1.711		38.688	.000
	numbmeet	.759	.492	.068	1.543	.124

a. Dependent Variable: fnlexam

#### **Coefficients**<sup>a</sup>

Ν	lodel 95.0% Confidence		95.0% Confidence Interval for B		Correlations	
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	62.815	69.536			
	numbmeet	208	1.725	.068	.068	.068

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>				
Model Collinearity Statistic			Statistics	
		Tolerance	VIF	
1	(Constant)			
	numbmeet	1.000	1.000	

a. Dependent Variable: fnlexam

### Coefficient Correlations<sup>a</sup>

Model			numbmeet
1	Correlations	numbmeet	1.000
	Covariances	numbmeet	.242

a. Dependent Variable: fnlexam

### **Collinearity Diagnostics**<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	numbmeet
1	1	1.930	1.000	.03	.03
	- 2	.070	5.269	.97	.97

a. Dependent Variable: fnlexam

#### Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.370	20	67.69	-47.694

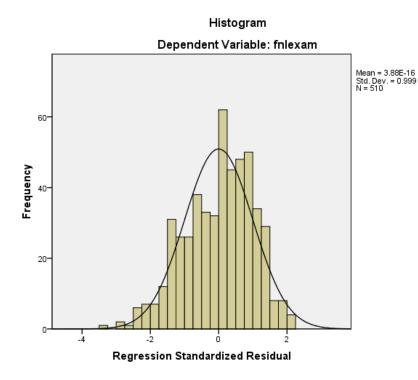
a. Dependent Variable: fnlexam

#### **Residuals Statistics**<sup>a</sup>

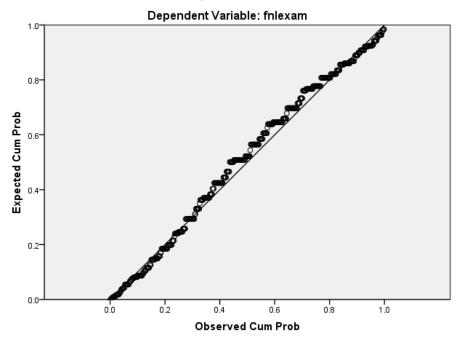
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	67.69	69.97	68.63	.968	510
Residual	-47.694	30.306	.000	14.140	510
Std. Predicted Value	969	1.384	.000	1.000	510
Std. Residual	-3.370	2.141	.000	.999	510

a. Dependent Variable: fnlexam

## Charts



Normal P-P Plot of Regression Standardized Residual



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /<u>DEPENDENT fnlexam</u> /METHOD=ENTER classize /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3).

## Regression

	Notes	
Output Created		02-Oct-2011 13:56:52
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION	
,		/DESCRIPTIVES MEAN STDDEV	
		CORR SIG N	
		/MISSING LISTWISE	
		/STATISTICS COEFF OUTS CI(95)	
		BCOV R ANOVA COLLIN TOL	
		CHANGE ZPP	
		/CRITERIA=PIN(.05) POUT(.10)	
		/NOORIGIN	
		/DEPENDENT fnlexam	
		/METHOD=ENTER classize	
		/RESIDUALS DURBIN	
		HISTOGRAM(ZRESID)	
		NORMPROB(ZRESID)	
		/CASEWISE PLOT(ZRESID)	
		OUTLIERS(3).	
Resources	Processor Time	00:00:01.123	
	Elapsed Time	00:00:01.182	
	Memory Required	2260 bytes	
	Additional Memory Required	656 bytes	
	for Residual Plots		

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no

names.sav

### **Descriptive Statistics**

	Mean	Std. Deviation	N
fnlexam	68.50	14.192	518
classize	25.72	6.002	518

Correlations				
		fnlexam	classize	
Pearson Correlation	fnlexam	1.000	.085	
	classize	.085	1.000	
Sig. (1-tailed)	fnlexam		.026	
	classize	.026		
Ν	fnlexam	518	518	
	classize	518	518	

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	classize <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
			Adjusted R	Std. Error of the	R Square		
	R	R Square	Square	Estimate	Change	F Change	df1
1	.085 <sup>a</sup>	.007	.005	14.155	.007	3.776	1

a. Predictors: (Constant), classize

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	516	.053	1.669			

b. Dependent Variable: fnlexam

			ANOVA <sup>b</sup>			
Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	756.575	1	756.575	3.776	.053 <sup>a</sup>
	Residual	103380.923	516	200.351		
	Total	104137.498	517			

a. Predictors: (Constant), classize

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
		Unstandardized Coefficients		Coefficients					
		В	Std. Error	Beta	t	Sig.			
1	(Constant)	63.314	2.739		23.112	.000			
	classize	.202	.104	.085	1.943	.053			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order Partial F		Part
1	(Constant)	57.932	68.695			
	classize	002	.405	.085	.085	.085

a. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>				
Model Collinearity Statistics					
		Tolerance	VIF		
1	(Constant)				
	classize	1.000	1.000		

a. Dependent Variable: fnlexam

### **Coefficient Correlations**<sup>a</sup>

Model			classize
1	Correlations	classize	1.000
	Covariances	classize	.011

a. Dependent Variable: fnlexam

### **Collinearity Diagnostics**<sup>a</sup>

Model	Dimension			Variance Proportions	
		Eigenvalue	Condition Index	(Constant)	classize
1	1	1.974	1.000	.01	.01
	- 2	.026	8.695	.99	.99

a. Dependent Variable: fnlexam

### Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
551	-3.331	20	67.14	-47.143

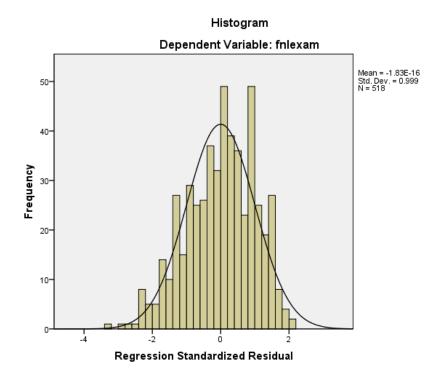
a. Dependent Variable: fnlexam

#### **Residuals Statistics**<sup>a</sup>

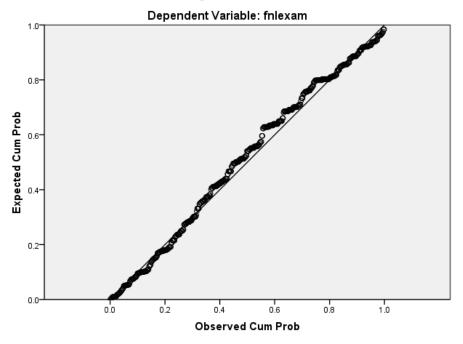
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	67.14	71.17	68.50	1.210	518
Residual	-47.143	30.438	.000	14.141	518
Std. Predicted Value	-1.120	2.212	.000	1.000	518
Std. Residual	-3.331	2.150	.000	.999	518

a. Dependent Variable: fnlexam

## Charts



Normal P-P Plot of Regression Standardized Residual



# Appendix T: SPSS Simple Binary Logistic Regression Output for

Elementary Algebra and Intermediate Algebra

GET

<u>FILE='C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav'.</u> DATASET NAME DataSet1 WINDOW=FRONT. <u>LOGISTIC REGRESSION VARIABLES fnlgrd</u> /<u>METHOD=ENTER gender</u> /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:48:42
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER gender
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.015
	Elapsed Time	00:00:00.032

## [DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	234	92.9		
	Missing Cases	18	7.1		
	Total	252	100.0		
Unselected Cases	;	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

## **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant	
Step 0	1	320.003	.274	
	2	320.003	.275	
	3	320.003	.275	

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 320.003

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed		d			
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	101	.0		
		1	0	133	100.0		
	Overall	Percentage			56.8		

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.275	.132	4.349	1	.037	1.317

#### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables gender	.000	1	.990
	Overall Statistics	.000	1	.990

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>							
Iteration			Coeffi	cients			
		-2 Log likelihood	Constant	gender			
Step 1	1	320.003	.272	.003			
	2	320.003	.274	.003			
	3	320.003	.274	.003			

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 320.003

Iteration History<sup>a,b,c,d</sup>

Iteration			Coeffic	cients
		-2 Log likelihood	Constant	gender
Step 1	1	320.003	.272	.003
	2	320.003	.274	.003
	3	320.003	.274	.003

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 320.003

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	.000	1	.990
	Block	.000	1	.990
	Model	.000	1	.990

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	320.003 <sup>a</sup>	.000	.000

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	.000	0	

#### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	54	54.000	71	71.000	125
	2	47	47.000	62	62.000	109

	Classification Table <sup>a</sup>						
Observed		Predicted					
		fnlgrd					
				Percentage			
		0	1	Correct			
Step 1	fnlgrd 0	0	101	.0			
	1	0	133	100.0			
	Overall Percentage			56.8			

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	gender	.003	.265	.000	1	.990	1.003
	Constant	.274	.181	2.298	1	.130	1.315

a. Variable(s) entered on step 1: gender.

### Variables in the Equation

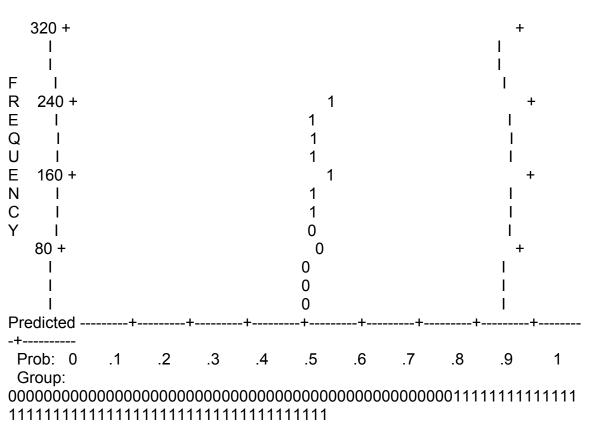
		95% C.I.for EXP(B)	
		Lower	Upper
Step 1 <sup>a</sup>	gender	.597	1.685
	Constant		

a. Variable(s) entered on step 1: gender.

**Correlation Matrix** 

		Constant	gender
Step 1	Constant	1.000	682
	gender	682	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 20 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER act /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## Logistic Regression

Notes				
Output Created		02-Oct-2011 14:50:00		
Comments				
Input	Data	C:\Users\Lin\Documents\math 096 fall		
		2001 with classrooms.sav		
	Active Dataset	DataSet1		
	Filter	<none></none>		
	Weight	<none></none>		
	Split File	<none></none>		
	N of Rows in Working Data	252		
	File			
Missing Value Handling	Definition of Missing	User-defined missing values are		
		treated as missing		
Syntax		LOGISTIC REGRESSION VARIABLES		
		fnlgrd		
		/METHOD=ENTER act		
		/CLASSPLOT		
		/CASEWISE OUTLIER(2)		
		/PRINT=GOODFIT CORR ITER(1)		
		CI(95)		
		/CRITERIA=PIN(0.05) POUT(0.10)		
		ITERATE(20) CUT(0.5).		
Resources	Processor Time	00:00:00.000		
	Elapsed Time	00:00:00.034		

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	Ν	Percent		
Selected Cases	Included in Analysis	177	70.2		
	Missing Cases	75	29.8		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

## **Block 0: Beginning Block**

Iteration	History <sup>a,b,c</sup>
-----------	--------------------------

Iteration		Coefficients
	-2 Log likelihood	Constant
Step 0 1	244.101	.169
2	244.101	.170

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 244.101

c. Estimation terminated at iteration number 2

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
	Observe	ed	Predicted			
			fnlı	grd		
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	81	.0	
		1	0	96	100.0	
	Overall	Percentage			54.2	

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		в	S.E.	Wald	df	Sig.	Exp(B)
Step 0 C	onstant	.170	.151	1.268	1	.260	1.185

#### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables act	4.990	1	.025
	Overall Statistics	4.990	1	.025

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant	act	
Step 1	1	238.905	-4.101	.295	
	2	238.859	-4.551	.326	
	3	238.859	-4.560	.327	
	4	238.859	-4.560	.327	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 244.101

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	5.242	1	.022
	Block	5.242	1	.022
	Model	5.242	1	.022

#### **Model Summary**

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	238.859 <sup>a</sup>	.029	.039

a. Estimation terminated at iteration number 4 because parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	1.756	2	.416

		fnlgrd = 0		fnlgro		
		Observed	Expected	Observed	Expected	Total
Step 1	1	13	13.738	9	8.262	22
	2	28	24.332	21	24.668	49
	3	37	40.323	60	56.677	97
	4	3	2.607	6	6.393	9

Contingency Table for Hosmer and Lemeshow Test

### Classification Table<sup>a</sup>

Observed		Predicted			
			fnlgrd		Percentage
			0	1	Correct
Step 1	fnlgrd	0	13	68	16.0
		1	9	87	90.6
	Overall	Percentage			56.5

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	act	.327	.152	4.638	1	.031	1.386
	Constant	-4.560	2.201	4.292	1	.038	.010

a. Variable(s) entered on step 1: act.

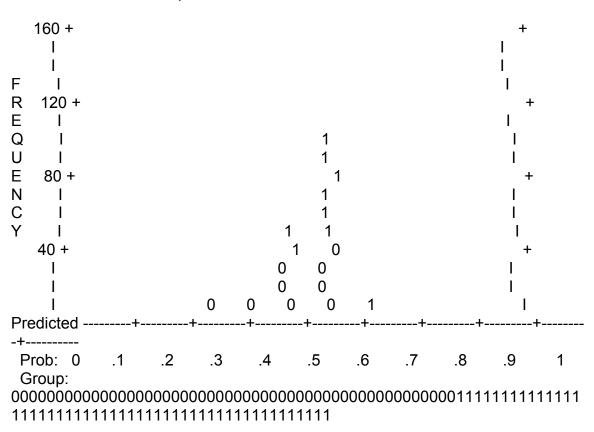
### Variables in the Equation

		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	act	1.030	1.866	
	Constant			

a. Variable(s) entered on step 1: act.

Correlation Matrix					
		Constant	act		
Step 1	Constant	1.000	998		
	act	998	1.000		

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER sat /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:51:19
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER sat
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.016
	Elapsed Time	00:00:00.035

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	Ν	Percent			
Selected Cases	Included in Analysis	17	6.7			
	Missing Cases	235	93.3			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

## **Block 0: Beginning Block**

### Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	23.035	.353
	2	23.035	.357
	3	23.035	.357

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 23.035

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>							
	Observe	ed		Predicte	d			
			fnl					
					Percentage			
			0	1	Correct			
Step 0	fnlgrd	0	0	7	.0			
		1	0	10	100.0			
	Overall	Percentage			58.8			

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.357	.493	.524	1	.469	1.429

#### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables sat	.278	1	.598
Overall Statistics	.278	1	.598

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant	sat	
Step 1	1	22.760	-1.408	.005	
	2	22.759	-1.440	.005	
	3	22.759	-1.440	.005	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 23.035

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	sat
Step 1	1	22.760	-1.408	.005
	2	22.759	-1.440	.005
	3	22.759	-1.440	.005

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 23.035

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	.276	1	.600
	Block	.276	1	.600
	Model	.276	1	.600

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	22.759 <sup>a</sup>	.016	.022

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

Step	Chi-square	df	Sig.
1	5.157	4	.272

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	0	.595	1	.405	1
	2	2	1.434	1	1.566	3
	3	2	.865	0	1.135	2
	4	1	1.163	2	1.837	3
	5	1	1.126	2	1.874	3
	6	1	1.815	4	3.185	5

**Contingency Table for Hosmer and Lemeshow Test** 

Classification Table<sup>a</sup>

Observed	Predicted			
	fnl	grd 1	Percentage Correct	
Step 1 fnlgrd 0	0	7	.0	
1	1	9	90.0	
Overall Percentage			52.9	

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	sat	.005	.010	.272	1	.602	1.005
	Constant	-1.440	3.481	.171	1	.679	.237

a. Variable(s) entered on step 1: sat.

#### Variables in the Equation

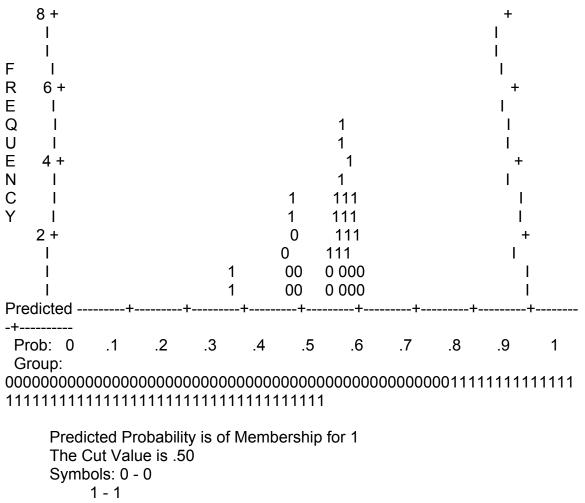
		95% C.I.fe	or EXP(B)
		Lower	Upper
Step 1 <sup>a</sup>	sat	.986	1.025
	Constant		

a. Variable(s) entered on step 1: sat.

Correlation Matrix				
		Constant	sat	
Step 1	Constant	1.000	990	
	sat	990	1.000	

Step number: 1

Observed Groups and Predicted Probabilities



Each Symbol Represents .5 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER comcol /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

	Notes	
Output Created		02-Oct-2011 14:52:54
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER comcol
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.026

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	235	93.3		
	Missing Cases	17	6.7		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
	Observe	ed	Predicted			
			fnlgrd			
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	101	.0	
		1	0	134	100.0	
	Overall	Percentage			57.0	

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

		Score	df	Sig.
Step 0	Variables comcol	1.553	1	.213
	Overall Statistics	1.553	1	.213

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant comcol		
Step 1	1	319.573	.462	324	
	2	319.572	.470	332	
	3	319.572	.470	332	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	comcol
Step 1	1	319.573	.462	324
	2	319.572	.470	332
	3	319.572	.470	332

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

		Chi-square	df	Sig.
Step 1	Step	1.558	1	.212
	Block	1.558	1	.212
	Model	1.558	1	.212

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	319.572 <sup>a</sup>	.007	.009

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	.000	0	

## **Contingency Table for Hosmer and Lemeshow Test**

		fnlgro	d = 0	= 0 fnlgro		
		Observed	Expected	Observed	Expected	Total
Step 1	1	61	61.000	70	70.000	131
	2	40	40.000	64	64.000	104

Classification Table <sup>a</sup>						
Observed	Predicted					
	fnlgrd					
			Percentage			
	0	1	Correct			
Step 1 fnlgrd 0	0	101	.0			
1	0	134	100.0			
Overall Percentage			57.0			

a. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	comcol	332	.267	1.549	1	.213	.717
	Constant	.470	.202	5.438	1	.020	1.600

a. Variable(s) entered on step 1: comcol.

## Variables in the Equation

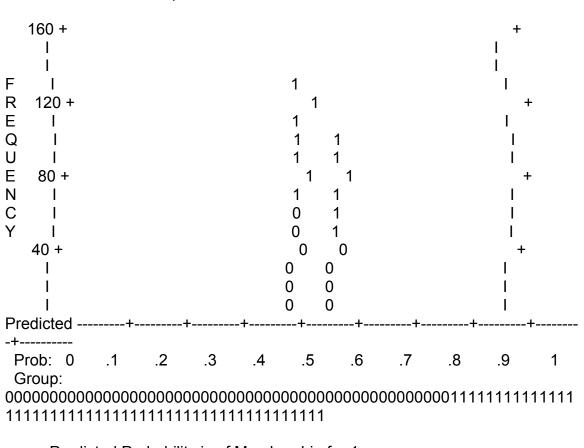
		95% C.I.for EXP(B)		
		Lower	Upper	
Step 1 <sup>a</sup>	comcol	.425	1.210	
	Constant			

a. Variable(s) entered on step 1: comcol.

**Correlation Matrix** 

		Constant	comcol
Step 1	Constant	1.000	755
	comcol	755	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:54:14
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.024

# [DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	N	Percent			
Selected Cases	Included in Analysis	235	93.3			
	Missing Cases	17	6.7			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
Observed				Predicte	d		
			fnle	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	101	.0		
		1	0	134	100.0		
	Overall	Percentage			57.0		

## a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

	В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Pretest	15.826	1	.000
Overall Statistics	15.826	1	.000

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>									
Iteration			Coefficients						
		-2 Log likelihood	Constant pretest						
Step 1	1	304.724	-1.043	.032					
	2	304.531	-1.163	.036					
	3	304.531	-1.166	.036					
	4	304.531	-1.166	.036					

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration			Coeffi	cients
		-2 Log likelihood	Constant	pretest
Step 1	1	304.724	-1.043	.032
	2	304.531	-1.163	.036
	3	304.531	-1.166	.036
	4	304.531	-1.166	.036

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	16.599	1	.000
	Block	16.599	1	.000
	Model	16.599	1	.000

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	304.531 <sup>a</sup>	.068	.092

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

Hosmer a	nd Lemes	how Test
----------	----------	----------

Step	Chi-square	Df	Sig.	
1	6.195	7	.517	

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	20	18.690	10	11.310	30
	2	10	9.617	7	7.383	17
	3	14	18.755	22	17.245	36
	4	15	11.421	9	12.579	24
	5	13	13.367	18	17.633	31
	6	7	6.977	11	11.023	18
	7	7	8.644	18	16.356	25
	8	11	8.936	20	22.064	31
	9	4	4.593	19	18.407	23

Contingency Table for Hosmer and Lemeshow Test

## Classification Table<sup>a</sup>

Observed			Predicted			
			fnlgrd		Dereentere	
			0	1	Percentage Correct	
			0	1	Conect	
Step 1	fnlgrd	0	44	57	43.6	
		1	39	95	70.9	
	Overall	Percentage			59.1	

a. The cut value is .500

Variables	in	the	Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.036	.009	14.851	1	.000	1.037
	Constant	-1.166	.393	8.822	1	.003	.311

a. Variable(s) entered on step 1: pretest.

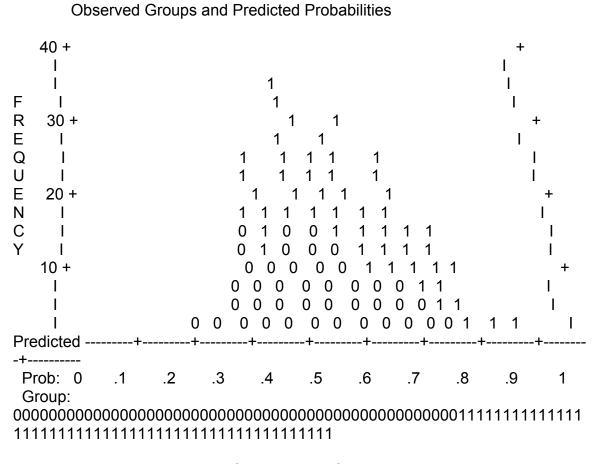
		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>ª</sup>	pretest Constant	1.018	1.056	

a. Variable(s) entered on step 1: pretest.

## **Correlation Matrix**

		Constant	pretest
Step 1	Constant	1.000	938
	pretest	938	1.000

Step number: 1



Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 2.5 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER ascgr /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

Notes				
Output Created		02-Oct-2011 14:56:59		
Comments				
Input	Data	C:\Users\Lin\Documents\math 096 fall		
		2001 with classrooms.sav		
	Active Dataset	DataSet1		
	Filter	<none></none>		
	Weight	<none></none>		
	Split File	<none></none>		
	N of Rows in Working Data	252		
	File			
Missing Value Handling	Definition of Missing	User-defined missing values are		
		treated as missing		
Syntax		LOGISTIC REGRESSION VARIABLES		
		fnlgrd		
		/METHOD=ENTER ascgr		
		/CLASSPLOT		
		/CASEWISE OUTLIER(2)		
		/PRINT=GOODFIT CORR ITER(1)		
		CI(95)		
		/CRITERIA=PIN(0.05) POUT(0.10)		
		ITERATE(20) CUT(0.5).		
Resources	Processor Time	00:00:00.015		
	Elapsed Time	00:00:00.034		

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	235	93.3		
	Missing Cases	17	6.7		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Classification Table <sup>a,b</sup>						
Observed Predicted			d			
		fnlgrd				
				Percentage		
		0	1	Correct		
Step 0	fnlgrd 0	0	101	.0		
	1	0	134	100.0		
	Overall Percentage			57.0		

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Ascgr	49.372	1	.000
Overall Statistics	49.372	1	.000

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant ascgr		
Step 1	1	270.783	-1.099	1.977	
	2	270.411	-1.230	2.171	
	3	270.411	-1.235	2.177	
	4	270.411	-1.235	2.177	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coeffic	cients	
		-2 Log likelihood	Constant	ascgr	
Step 1	1	270.783	-1.099	1.977	
	2	270.411	-1.230	2.171	

270.411

270.411

-1.235

-1.235

2.177

2.177

a. Method: Enter

3

4

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

## **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	50.719	1	.000
	Block	50.719	1	.000
	Model	50.719	1	.000

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	270.411 <sup>a</sup>	.194	.261

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

#### **Hosmer and Lemeshow Test**

Step	Chi-square	Df	Sig.
1	.000	0	

## **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	55	55.000	16	16.000	71
	2	46	46.000	118	118.000	164

	Classification Table <sup>a</sup>				
Observed			Predicte	d	
		fnl	grd		
				Percentage	
		0	1	Correct	
Step 1	fnlgrd 0	55	46	54.5	
	1	16	118	88.1	
	Overall Percentage			73.6	

a. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	ascgr	2.177	.333	42.729	1	.000	8.818
	Constant	-1.235	.284	18.896	1	.000	.291

a. Variable(s) entered on step 1: ascgr.

## Variables in the Equation

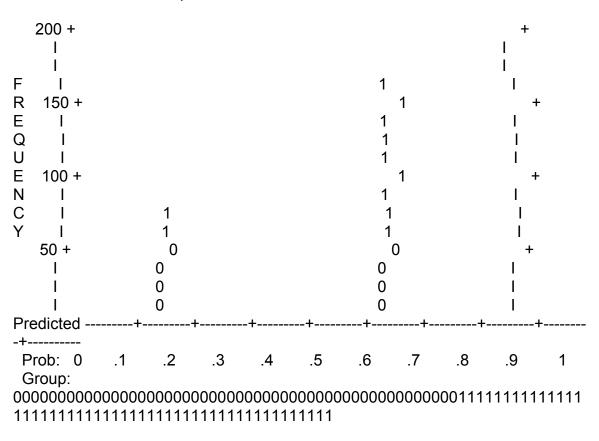
		95% C.I.for EXP(B)	
		Lower Upper	
Step 1 <sup>a</sup>	ascgr Constant	4.591	16.937

a. Variable(s) entered on step 1: ascgr.

**Correlation Matrix** 

		Constant	ascgr
Step 1	Constant	1.000	853
	ascgr	853	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 12.5 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER techsex /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# **Logistic Regression**

	Notes	
Output Created		02-Oct-2011 14:58:15
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER techsex
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.000
	Elapsed Time	00:00:00.022

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	235	93.3		
	Missing Cases	17	6.7		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed	Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	101	.0		
		1	0	134	100.0		
	Overall	Percentage			57.0		

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Techsex	6.833	1	.009
Overall Statistics	6.833	1	.009

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>							
Iteration			Coeffi	cients			
		-2 Log likelihood	Constant	techsex			
Step 1	1	314.205	.027	.703			
	2	314.182	.027	.738			
	3	314.182	.027	.738			

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	techsex
Step 1	1	314.205	.027	.703
	2	314.182	.027	.738
	3	314.182	.027	.738

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	6.948	1	.008
	Block	6.948	1	.008
	Model	6.948	1	.008

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	314.182 <sup>a</sup>	.029	.039

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

## **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	74	74.000	76	76.000	150
	2	27	27.000	58	58.000	85

	Classification Table <sup>a</sup>						
	Observed	Predicted					
		fnlgrd					
				Percentage			
		0	1	Correct			
Step 1	fnlgrd 0	0	101	.0			
	1	0	134	100.0			
	Overall Percentage			57.0			

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	techsex	.738	.285	6.727	1	.009	2.092
	Constant	.027	.163	.027	1	.870	1.027

a. Variable(s) entered on step 1: techsex.

## Variables in the Equation

		95% C.I.for EXP(B)		
		Lower Upp		
Step 1 <sup>a</sup>	techsex	1.198	3.653	
	Constant			

a. Variable(s) entered on step 1: techsex.

**Correlation Matrix** 

		Constant	techsex
Step 1	Constant	1.000	574
	techsex	574	1.000

Step number: 1

**Observed Groups and Predicted Probabilities** 160 ++ 1 T 1 Т F 1 I R 120 +1 + Е Т 1 Q Ι 1 U 1 1 L Е 80 + 1 1 Ν L 0 1 С 0 L 1 Y L 0 1 40 + 0 1 0 0 I 0 0 I 0 0 Predicted -Prob: 0 .2 .3 .4 .5 .6 .7 .8 .9 1 .1 Group:

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER adj096 /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# **Logistic Regression**

	Notes	
Output Created		02-Oct-2011 14:59:57
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER adj096
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.024

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	235	93.3		
	Missing Cases	17	6.7		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>				
Observed				Predicte	d
			fnl	grd	
					Percentage
			0	1	Correct
Step 0	fnlgrd	0	0	101	.0
		1	0	134	100.0
	Overall	Percentage			57.0

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

		Score	df	Sig.
Step 0	Variables adj096	11.809	1	.001
	Overall Statistics	11.809	1	.001

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant	adj096	
Step 1	1	309.199	125	.891	
	2	309.160	125	.932	
	3	309.160	125	.933	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	adj096
Step 1	1	309.199	125	.891
	2	309.160	125	.932
	3	309.160	125	.933

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	11.970	1	.001
	Block	11.970	1	.001
	Model	11.970	1	.001

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	309.160 <sup>a</sup>	.050	.067

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	.000	0	

## **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	68	68.000	60	60.000	128
	2	33	33.000	74	74.000	107

	Classification Table <sup>a</sup>							
	Observed	erved Predicted						
		fnlgrd						
				Percentage				
		0	1	Correct				
Step 1	fnlgrd 0	68	33	67.3				
	1	60	74	55.2				
	Overall Percentage			60.4				

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	adj096	.933	.274	11.570	1	.001	2.541
	Constant	125	.177	.499	1	.480	.882

a. Variable(s) entered on step 1: adj096.

## Variables in the Equation

		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	adj096	1.485	4.350	
	Constant			

a. Variable(s) entered on step 1: adj096.

**Correlation Matrix** 

		Constant	adj096
Step 1	Constant	1.000	646
	adj096	646	1.000

Step number: 1

160 ++ T I Т I F 1 T R 120 +1 + Е 1 Т 1 Q Ι 1 1 U 1 1 L Е 80 + 1 1 Ν 0 L 1 С 0 L 1 Y L 0 1 40 + 0 1 0 0 I 0 I 0 0 0 I Predicted Prob: 0 .2 .3 .4 .5 .6 .7 .8 .9 1 .1 Group: 

**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER mozartuse /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 15:01:37
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER mozartuse
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.047
	Elapsed Time	00:00:00.027

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	N	Percent				
Selected Cases	Included in Analysis	235	93.3			
	Missing Cases	17	6.7			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	321.130	.281
	2	321.130	.283
	3	321.130	.283

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed		Predicte	d		
			fnl	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	101	.0		
		1	0	134	100.0		
	Overall	Percentage			57.0		

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

		Score	df	Sig.
Step 0	Variables Mozartuse	9.474	1	.002
	Overall Statistics	9.474	1	.002

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>						
Iteration			Coefficients			
		-2 Log likelihood	d Constant mozartuse			
Step 1	1	311.719	.466	-1.038		
	2	311.713	.475	-1.063		
	3	311.713	.475	-1.063		

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration	ı		Coefficients	
		-2 Log likelihood	Constant	mozartuse
Step 1	1	311.719	.466	-1.038
	2	311.713	.475	-1.063
	3	311.713	.475	-1.063

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	9.417	1	.002
	Block	9.417	1	.002
	Model	9.417	1	.002

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	311.713 <sup>a</sup>	.039	.053

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

## **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0 fnlgrd =		d = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	27	27.000	15	15.000	42
	2	74	74.000	119	119.000	193

	Classification Table <sup>a</sup>						
	Observed	Predicted					
		fnl					
				Percentage			
		0	1	Correct			
Step 1	fnlgrd 0	27	74	26.7			
	1	15	119	88.8			
	Overall Percentage			62.1			

a. The cut value is .500

Variables	in	the	Fa	uation
variables		uie	ЦЧ	uation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	mozartuse	-1.063	.354	8.992	1	.003	.345
	Constant	.475	.148	10.297	1	.001	1.608

a. Variable(s) entered on step 1: mozartuse.

## Variables in the Equation

95% C.I.for EXP(B)		
Lower Upper		
.172	.692	
	Lower	

a. Variable(s) entered on step 1: mozartuse.

**Correlation Matrix** 

		Constant	mozartuse
Step 1	Constant	1.000	418
	mozartuse	418	1.000

Step number: 1

200 + + 1 I 1 I F 1 R 150 +1 + Е 1 Т Q T 1 U 1 T Е 100 +1 Ν 1 I С 0 L Y L 0 50 + 0 1 0 I 0 0 I 0 0 I Predicted Prob: 0 .2 .3 .4 .5 .6 .7 .8 .9 1 .1 Group: 

**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 12.5 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER amisone /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 15:03:37
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER amisone
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.000
	Elapsed Time	00:00:00.033

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	N	Percent			
Selected Cases	Included in Analysis	223	88.5		
	Missing Cases	29	11.5		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

## **Block 0: Beginning Block**

### Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	300.801	.386
	2	300.800	.391
	3	300.800	.391

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 300.800

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
Observed Predicted			d			
			fnlgrd			
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	90	.0	
		1	0	133	100.0	
	Overall	Percentage			59.6	

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.391	.136	8.187	1	.004	1.478

#### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Amisone	5.763	1	.016
Overall Statistics	5.763	1	.016

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	d Constant amisone		
Step 1	1	295.081	.629	653	
	2	295.066	.651	675	
	3	295.066	.651	675	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 300.800

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	amisone
Step 1	1	295.081	.629	653
	2	295.066	.651	675
	3	295.066	.651	675

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 300.800

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

		Chi-square	df	Sig.
Step 1	Step	5.734	1	.017
	Block	5.734	1	.017
	Model	5.734	1	.017

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	295.066 <sup>a</sup>	.025	.034

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

#### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	42	42.000	41	41.000	83
	2	48	48.000	92	92.000	140

	Classification Table <sup>a</sup>							
	Observed	Predicted						
		fnlgrd						
				Percentage				
		0	1	Correct				
Step 1	fnlgrd 0	42	48	46.7				
	1	41	92	69.2				
	Overall Percentage			60.1				

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	amisone	675	.283	5.697	1	.017	.509
	Constant	.651	.178	13.351	1	.000	1.917

a. Variable(s) entered on step 1: amisone.

### Variables in the Equation

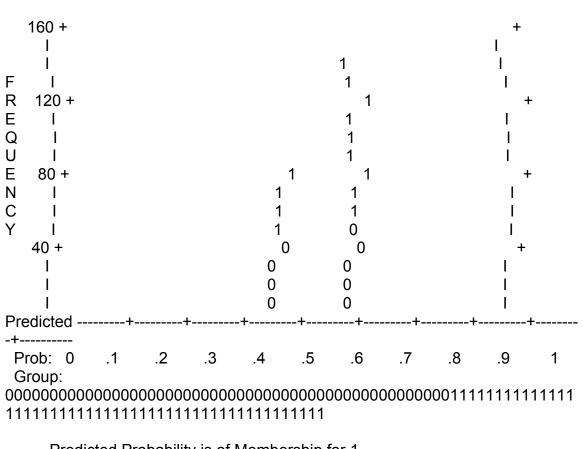
		95% C.I.for EXP(B) Lower Upper		
Step 1 <sup>a</sup>	amisone	.293	.886	
	Constant			

a. Variable(s) entered on step 1: amisone.

**Correlation Matrix** 

		Ormatant	
		Constant	amisone
Step 1	Constant	1.000	630
	amisone	630	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER numbmeet /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 15:05:10
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER numbmeet
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.026

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	Ν	Percent			
Selected Cases	Included in Analysis	226	89.7			
	Missing Cases	26	10.3			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

## **Block 0: Beginning Block**

### Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	305.453	.372
	2	305.452	.376
	3	305.452	.376

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed		Predicte	d		
			fnl	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	92	.0		
		1	0	134	100.0		
	Overall	Percentage			59.3		

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		в	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.376	.135	7.714	1	.005	1.457

#### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables Numbmeet	9.660	1	.002
	Overall Statistics	9.660	1	.002

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>								
Iteration			Coefficients					
		-2 Log likelihood	Constant	numbmeet				
Step 1	1	295.686	1.484	309				
	2	295.623	1.584	330				
	3	295.623	1.585	330				
	4	295.623	1.585	330				

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

Iteration History <sup>a,b,c,d</sup>								
Iteration			Coefficients					
		-2 Log likelihood	Constant	numbmeet				
Step 1	1	295.686	1.484	309				
	2	295.623	1.584	330				
	3	295.623	1.585	330				
	4	295.623	1.585	330				

a. Method: Enter

Т

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	9.828	1	.002
	Block	9.828	1	.002
	Model	9.828	1	.002

Мо	del Summary

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	295.623 <sup>a</sup>	.043	.057

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

Hosmer	and	Lemeshow	Test
--------	-----	----------	------

Step	Chi-square	Df	Sig.
1	.157	1	.692

#### Contingency Table for Hosmer and Lemeshow Test

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	42	42.839	41	40.161	83
	2	26	24.742	31	32.258	57

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	42	42.839	41	40.161	83
	2	26	24.742	31	32.258	57
	3	24	24.419	62	61.581	86

Contingency Table for Hosmer and Lemeshow Test

#### Classification Table<sup>a</sup>

	Observed			Predicted				
		fnlgrd		Percentage				
			0	1	Correct			
Step 1	fnlgrd 0		42	50	45.7			
	1		41	93	69.4			
	Overall Percentage				59.7			

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)		
Step 1 <sup>a</sup>	numbmeet	330	.107	9.442	1	.002	.719		
	Constant	1.585	.423	14.011	1	.000	4.878		

a. Variable(s) entered on step 1: numbmeet.

### Variables in the Equation

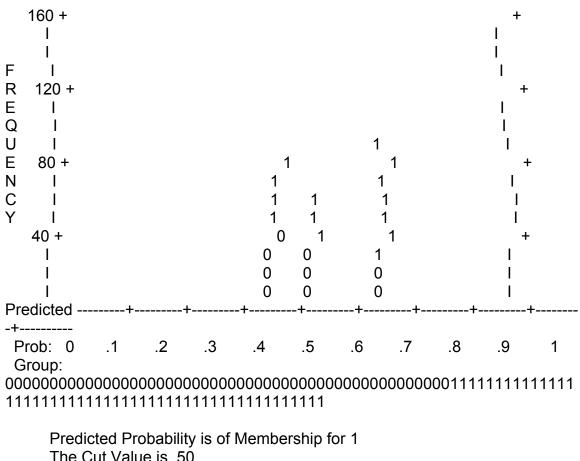
95% C.I.for EXP(B)		
Lower Upper		
.583	.887	
	Lower	

a. Variable(s) entered on step 1: numbmeet.

Correlation Matrix					
		Constant	numbmeet		
Step 1	Constant	1.000	945		
	numbmeet	945	1.000		

Step number: 1





The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER classize /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 15:07:00
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER classize
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.000
	Elapsed Time	00:00:00.032

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	235	93.3		
	Missing Cases	17	6.7		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	321.130	.281		
	2	321.130	.283		
	3	321.130	.283		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
	Observed	Predicted				
		fnl	grd			
				Percentage		
		0	1	Correct		
Step 0	fnlgrd 0	0	101	.0		
	1	0	134	100.0		
	Overall Percentage			57.0		

## .

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

#### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables Classize	.015	1	.902
	Overall Statistics	.015	1	.902

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant classize		
Step 1	1	321.115	.325	001	
	2	321.114	.327	001	
	3	321.114	.327	001	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients		
		-2 Log likelihood	Constant	classize	
Step 1	1	321.115	.325	001	
	2	321.114	.327	001	
	3	321.114	.327	001	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	.015	1	.902
	Block	.015	1	.902
	Model	.015	1	.902

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	321.114 <sup>a</sup>	.000	.000

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

Step	Chi-square	Df	Sig.
1	19.779	5	.001

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	19	19.556	26	25.444	45
	2	27	18.184	15	23.816	42
	3	8	17.293	32	22.707	40
	4	10	8.964	11	12.036	21
	5	12	8.524	8	11.476	20
	6	15	17.036	25	22.964	40
	7	10	11.442	17	15.558	27

Contingency Table for Hosmer and Lemeshow Test

	Classification Table <sup>a</sup>				
	Observed		Predicte	d	
		fnl	grd		
				Percentage	
		0	1	Correct	
Step 1	fnlgrd 0	0	101	.0	
	1	0	134	100.0	
	Overall Percentage			57.0	

a. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	classize	001	.011	.015	1	.902	.999
	Constant	.327	.385	.724	1	.395	1.387

a. Variable(s) entered on step 1: classize.

#### Variables in the Equation

		95% C.I.for EXP(B)	
		Lower	Upper
Step 1 <sup>a</sup>	classize Constant	.978	1.020
	Constant		

a. Variable(s) entered on step 1: classize.

**Correlation Matrix** 

		Constant	classize
Step 1	Constant	1.000	940
	classize	940	1.000

Step number: 1

160 ++ T I Т I F 1 T R 120 +1 + Е 11 Т Q Т 11 U 11 Т Е 80 + 11 Ν 11 I С 11 T Y L 00 40 + 00 00 I 00 I 00 I Predicted --+ .2 .3 .5 .6 .7 .8 .9 1 Prob: 0 .1 .4 Group: 

**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 10 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER gender /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:02:10
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER gender
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.072

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

#### Case Processing Summary

Unweighted Cases	N	Percent	
Selected Cases	Included in Analysis	659	97.6
	Missing Cases	16	2.4
	Total	675	100.0
Unselected Cases		0	.0
Total		675	100.0

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# Block 0: Beginning Block

Iteration	History <sup>a,b,c</sup>
-----------	--------------------------

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	836.709	.677
	2	836.596	.704
	3	836.596	.705

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 836.596

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>							
	Observed			Predicte	d			
			fnl	grd				
					Percentage			
			0	1	Correct			
Step 0	fnlgrd	0	0	218	.0			
		1	0	441	100.0			
	Overall	Percentage			66.9			

#### a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 Co	nstant	.705	.083	72.416	1	.000	2.023

#### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Gender	19.981	1	.000
Overall Statistics	19.981	1	.000

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>							
Iteration			Coefficients				
		-2 Log likelihood	Constant	gender			
Step 1	1	817.244	.960	662			
	2	816.700	1.045	744			
	3	816.699	1.046	746			
	4	816.699	1.046	746			

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 836.596

Iteration	History <sup>a,b,c,d</sup>
-----------	----------------------------

Iteration			Coefficients	
		-2 Log likelihood	Constant gender	
Step 1	1	817.244	.960	662
	2	816.700	1.045	744
	3	816.699	1.046	746
	4	816.699	1.046	746

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 836.596

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	19.897	1	.000
	Block	19.897	1	.000
	Model	19.897	1	.000

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	816.699 <sup>a</sup>	.030	.041

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

#### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	120	120.000	162	162.000	282
	2	98	98.000	279	279.000	377

Classification Table <sup>a</sup>					
Observed		Predicte	d		
	fnl	grd			
			Percentage		
	0	1	Correct		
Step 1 fnlgrd 0	0	218	.0		
1	0	441	100.0		
Overall Percentage			66.9		

a. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	gender	746	.168	19.676	1	.000	.474
	Constant	1.046	.117	79.388	1	.000	2.847

a. Variable(s) entered on step 1: gender.

### Variables in the Equation

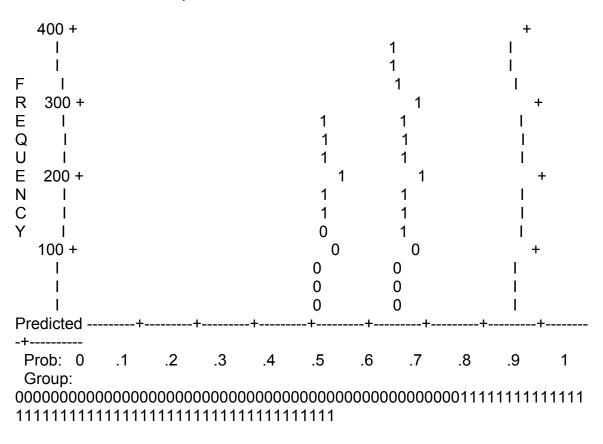
	95% C.I.for EXP(B)		
	Lower Upper		
gender	.341	.659	
Constant			
	0	Lower gender .341	

a. Variable(s) entered on step 1: gender.

**Correlation Matrix** 

		Constant	gender
Step 1	Constant	1.000	698
	gender	698	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 25 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

### LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER act /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

	Notes	
Output Created		02-Oct-2011 14:12:13
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER act
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.037

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	Ν	Percent			
Selected Cases	Included in Analysis	608	90.1			
	Missing Cases	67	9.9			
	Total	675	100.0			
Unselected Cases	3	0	.0			
Total		675	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

## **Block 0: Beginning Block**

### Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	775.924	.658
	2	775.836	.683
	3	775.836	.683

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 775.836

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
Observed Predicted			d			
			fnl			
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	204	.0	
		1	0	404	100.0	
	Overall I	Percentage			66.4	

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.683	.086	63.288	1	.000	1.980

#### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Act	36.450	1	.000
Overall Statistics	36.450	1	.000

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant	act	
Step 1	1	739.794	-5.516	.370	
	2	738.454	-6.713	.445	
	3	738.452	-6.767	.448	
	4	738.452	-6.767	.448	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 775.836

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	act
Step 1	1	739.794	-5.516	.370
	2	738.454	-6.713	.445
	3	738.452	-6.767	.448
	4	738.452	-6.767	.448

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 775.836

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	37.384	1	.000
	Block	37.384	1	.000
	Model	37.384	1	.000

Model	Summary
	• annan y

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	738.452 <sup>a</sup>	.060	.083

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

Hosmer a	nd Lemes	how Test
----------	----------	----------

Step	Chi-square	Df	Sig.
1	5.224	2	.073

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	37	39.555	31	28.445	68
	2	71	68.136	99	101.864	170
	3	70	61.375	135	143.625	205
	4	26	34.935	139	130.065	165

### Contingency Table for Hosmer and Lemeshow Test

#### **Classification Table**<sup>a</sup>

Observed		Predicted			
			fnlgrd		
					Percentage
			0	1	Correct
Step 1	fnlgrd	0	37	167	18.1
		1	31	373	92.3
	Overall	Percentage			67.4

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	act	.448	.078	32.678	1	.000	1.565
	Constant	-6.767	1.303	26.981	1	.000	.001

a. Variable(s) entered on step 1: act.

#### Variables in the Equation

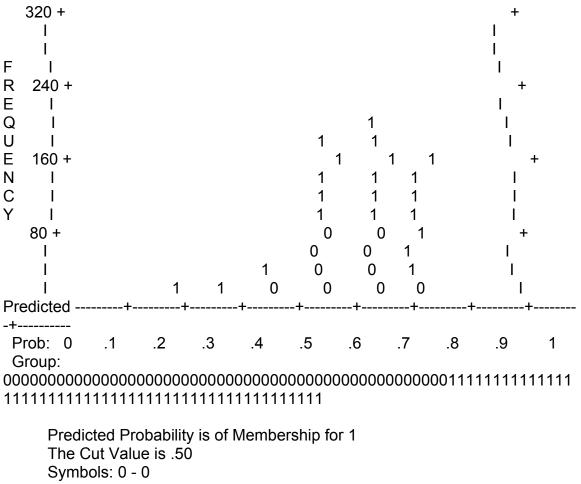
		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	act	1.342	1.825	
	Constant			

a. Variable(s) entered on step 1: act.

Correlation Matrix				
		Constant	act	
Step 1	Constant	1.000	998	
act998 1.000				

Step number: 1

### **Observed Groups and Predicted Probabilities**



1 - 1

Each Symbol Represents 20 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

### LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER sat /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5). Logistic Regression

#### Notes **Output Created** 02-Oct-2011 14:13:39 Comments Input Data C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav Active Dataset DataSet1 Filter <none> Weight <none> Split File <none> N of Rows in Working Data 675 File Missing Value Handling Definition of Missing User-defined missing values are treated as missing Syntax LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER sat /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5). Resources Processor Time 00:00:00.031 00:00:00.028 Elapsed Time

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Unweighted Cases <sup>a</sup>		N	Percent	
Selected Cases	Included in Analysis	80	11.9	
	Missing Cases	595	88.1	
	Total	675	100.0	
Unselected Cases		0	.0	
Total		675	100.0	

Case Processing Summary

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant	
Step 0	1	100.910	.700	
	2	100.893	.731	
	3	100.893	.731	

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 100.893

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
Observed				Predicte	d		
			fnlg	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	26	.0		
		1	0	54	100.0		
	Overall	Percentage			67.5		

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.731	.239	9.375	1	.002	2.077

#### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Sat	.345	1	.557
Overall Statistics	.345	1	.557

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant sat		
Step 1	1	100.577	255	.002	
	2	100.554	328	.003	
	3	100.554	329	.003	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 100.893

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	sat
Step 1	1	100.577	255	.002
	2	100.554	328	.003
	3	100.554	329	.003

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 100.893

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	.339	1	.561
	Block	.339	1	.561
	Model	.339	1	.561

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	100.554 <sup>a</sup>	.004	.006

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

Step	Chi-square	Df	Sig.	
1	7.578	8	.476	

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	2	2.800	5	4.200	7
	2	4	2.844	4	5.156	8
	3	2	3.341	8	6.659	10
	4	1	.657	1	1.343	2
	5	5	4.516	9	9.484	14
	6	4	2.218	3	4.782	7
	7	3	2.179	4	4.821	7
	8	2	2.140	5	4.860	7
	9	1	3.602	11	8.398	12
	10	2	1.705	4	4.295	6

Contingency Table for Hosmer and Lemeshow Test

#### Classification Table<sup>a</sup>

	Observed	Predicted							
		fnlgrd							
				Percentage					
		0	1	Correct					
Step 1	fnlgrd 0	0	26	.0					
	1	0	54	100.0					
	Overall Percentage			67.5					

a. The cut value is .500

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	sat	.003	.004	.342	1	.559	1.003
	Constant	329	1.824	.033	1	.857	.720

a. Variable(s) entered on step 1: sat.

## Variables in the Equation

		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	sat	.994	1.011	
	Constant			

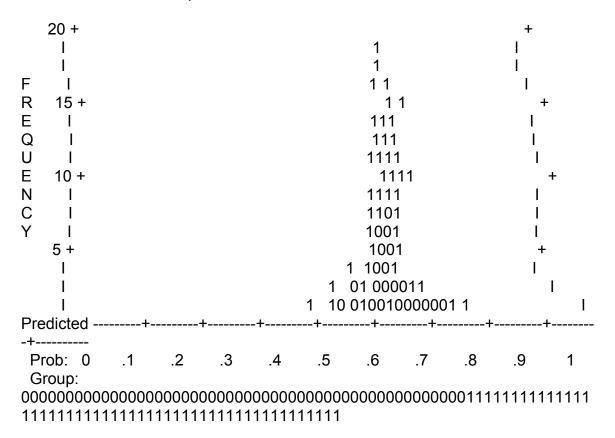
a. Variable(s) entered on step 1: sat.

**Correlation Matrix** 

		Constant	sat
Step 1	Constant	1.000	991
	sat	991	1.000

Step number: 1

Observed Groups and Predicted Probabilities



Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1.25 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER comcol /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:16:39
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER comcol
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.015
	Elapsed Time	00:00:00.038

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	N	Percent			
Selected Cases	Included in Analysis	670	99.3			
	Missing Cases	5	.7			
	Total	675	100.0			
Unselected Cases	3	0	.0			
Total		675	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	851.181	.675
	2	851.069	.702
	3	851.069	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 851.069

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed	Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	222	.0		
		1	0	448	100.0		
	Overall	Percentage			66.9		

### a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

	В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 Constant	.702	.082	73.177	1	.000	2.018

### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Comcol	20.127	1	.000
Overall Statistics	20.127	1	.000

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>						
Iteration			Coefficients			
		-2 Log likelihood	Constant	comcol		
Step 1	1	832.089	.843	801		
	2	831.744	.899	856		
	3	831.744	.899	857		

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 851.069

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	19.324	1	.000
	Block	19.324	1	.000
	Model	19.324	1	.000

#### **Model Summary**

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	831.744 <sup>a</sup>	.028	.040

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### Contingency Table for Hosmer and Lemeshow Test

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	69	69.000	72	72.000	141
	2	153	153.000	376	376.000	529

Classification	Table <sup>a</sup>
----------------	--------------------

Observed		Predicted			
		fnlgrd		Percentage	
		0	1	Correct	
Step 1 fnlgrd 0		0	222	.0	
1		0	448	100.0	
Overall Percentage				66.9	

a. The cut value is .500

Variables in the Equation						
	В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup> comcol	857	.194	19.526	1	.000	.425
Constant	.899	.096	87.920	1	.000	2.458

## Variables in the Equation

a. Variable(s) entered on step 1: comcol.

## Variables in the Equation

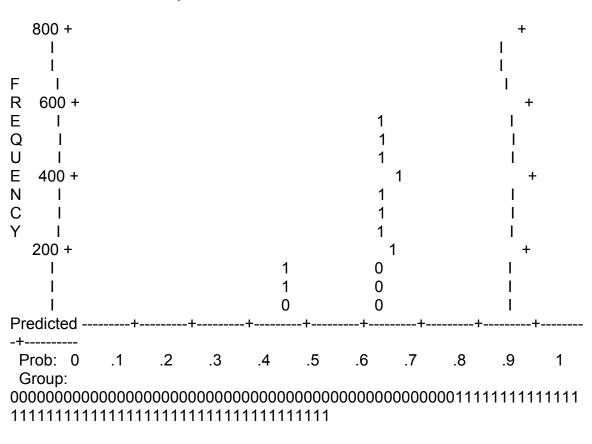
		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	comcol	.290	.621	
	Constant			

a. Variable(s) entered on step 1: comcol.

### **Correlation Matrix**

		Constant	comcol
Step 1	Constant	1.000	495
	comcol	495	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:19:02
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.000
	Elapsed Time	00:00:00.033

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	Ν	Percent			
Selected Cases	Included in Analysis	542	80.3			
	Missing Cases	133	19.7			
	Total	675	100.0			
Unselected Cases	3	0	.0			
Total		675	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	674.407	.745
	2	674.240	.783
	3	674.240	.783

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 674.240

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed	Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	170	.0		
		1	0	372	100.0		
	Overall	Percentage			68.6		

### a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

	В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 Constan	nt .783	.093	71.552	1	.000	2.188

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Pretest	24.733	1	.000
Overall Statistics	24.733	1	.000

# Block 1: Method = Enter

	Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients			
		-2 Log likelihood	Constant	pretest		
Step 1	1	650.113	509	.027		
	2	648.792	701	.033		
	3	648.789	711	.034		
	4	648.789	711	.034		

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 674.240

Iteration	History <sup>a,b,c,d</sup>
-----------	----------------------------

Iteration			Coefficients	
		-2 Log likelihood	Constant	pretest
Step 1	1	650.113	509	.027
	2	648.792	701	.033
	3	648.789	711	.034
	4	648.789	711	.034

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 674.240

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	25.451	1	.000
	Block	25.451	1	.000
	Model	25.451	1	.000

Model	Summary
Model	Summary

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	648.789 <sup>a</sup>	.046	.064

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

Hosmer a	nd Lemes	how Test
----------	----------	----------

Step	Chi-square	Df	Sig.
1	5.576	7	.590

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	25	24.850	25	25.150	50
	2	17	22.613	36	30.387	53
	3	28	27.808	44	44.192	72
	4	25	20.491	34	38.509	59
	5	22	21.101	46	46.899	68
	6	20	16.536	40	43.464	60
	7	13	13.387	42	41.613	55
	8	9	10.479	40	38.521	49
	9	11	12.735	65	63.265	76

Contingency Table for Hosmer and Lemeshow Test

### Classification Table<sup>a</sup>

Observed			Predicted		
			fnlgrd		
					Percentage
			0	1	Correct
Step 1	fnlgrd	0	14	156	8.2
		1	8	364	97.8
	Overall	Percentage			69.7

a. The cut value is .500

Variables	in	the	Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.034	.007	23.706	1	.000	1.034
	Constant	711	.313	5.156	1	.023	.491

a. Variable(s) entered on step 1: pretest.

## Variables in the Equation

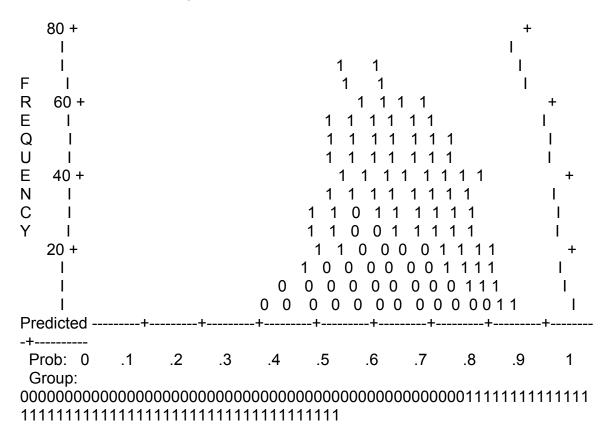
		95% C.I.for EXP(B)		
		Lower	Upper	
Step 1 <sup>ª</sup>	pretest Constant	1.020	1.048	

a. Variable(s) entered on step 1: pretest.

**Correlation Matrix** 

		Constant	pretest
Step 1	Constant	1.000	953
	pretest	953	1.000

Step number: 1 Observed Groups and Predicted Probabilities



Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 5 Cases.

Casewise L	ist <sup>b</sup>
------------	------------------

Case		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
460	S	0**	.878	1	878	-2.682

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER ascgr /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:20:23
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER ascgr
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.016
	Elapsed Time	00:00:00.044

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary							
Unweighted Case	N	Percent					
Selected Cases	Included in Analysis	665	98.5				
	Missing Cases	10	1.5				
	Total	675	100.0				
Unselected Cases	;	0	.0				
Total		675	100.0				

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
- 1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	842.939	.683
	2	842.819	.711
	3	842.819	.711

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 842.819

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed		Predicte	d		
			fnl	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	219	.0		
		1	0	446	100.0		
	Overall	Percentage			67.1		

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.711	.083	74.302	1	.000	2.037

### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables Ascgr	197.749	1	.000
	Overall Statistics	197.749	1	.000

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>						
Iteration			Coefficients			
		-2 Log likelihood	Constant	ascgr		
Step 1	1	654.670	-1.000	2.307		
	2	649.555	-1.096	2.641		
	3	649.531	-1.099	2.662		
	4	649.531	-1.099	2.662		

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 842.819

Iteration History <sup>a,b,c,d</sup>							
Iteration		Coefficients					
	-2 Log likelihood	Constant	ascgr				
Step 1 1	654.670	-1.000	2.307				
2	649.555	-1.096	2.641				
3	649.531	-1.099	2.662				
4	649.531	-1.099	2.662				

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 842.819

d. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	193.288	1	.000
	Block	193.288	1	.000
	Model	193.288	1	.000

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	649.531 <sup>a</sup>	.252	.351

a. Estimation terminated at iteration number 4 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	135	135.000	45	45.000	180
	2	84	84.000	401	401.000	485

Classification Table <sup>a</sup>						
Observed	Predicted					
	fnlgrd					
			Percentage			
	0	1	Correct			
Step 1 fnlgrd 0	135	84	61.6			
1	45	401	89.9			
Overall Percentage			80.6			

a. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	ascgr	2.662	.210	160.919	1	.000	14.321
	Constant	-1.099	.172	40.735	1	.000	.333

a. Variable(s) entered on step 1: ascgr.

## Variables in the Equation

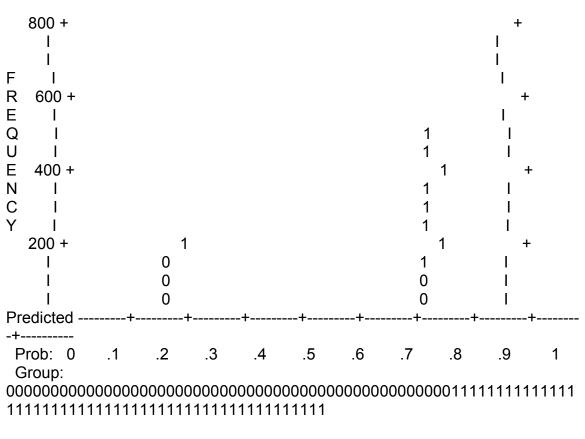
	95% C.I.for EXP(B)	
	Lower	Upper
ascgr	9.492	21.607
Constant		
	U	Lower ascgr 9.492

a. Variable(s) entered on step 1: ascgr.

**Correlation Matrix** 

		Constant	ascgr
Step 1	Constant	1.000	820
	ascgr	820	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER techsex /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:22:49
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER techsex
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.041

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary				
Unweighted Case	S <sup>a</sup>	N	Percent	
Selected Cases	Included in Analysis	673	99.7	
	Missing Cases	2	.3	
	Total	675	100.0	
Unselected Cases	;	0	.0	
Total		675	100.0	

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	855.001	.675
	2	854.888	.702
	3	854.888	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 854.888

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>				
Observed			Predicte	d	
			fnl	grd	
					Percentage
			0	1	Correct
Step 0	fnlgrd	0	0	223	.0
		1	0	450	100.0
	Overall	Percentage			66.9

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.702	.082	73.497	1	.000	2.018

### Variables not in the Equation

		Score	df	Sig.
Step 0	Variables Techsex	1.060	1	.303
	Overall Statistics	1.060	1	.303

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant techsex	
Step 1	1	853.957	.623	.159
	2	853.820	.644	.182
	3	853.820	.644	.182

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	techsex
Step 1	1	853.957	.623	.159
	2	853.820	.644	.182
	3	853.820	.644	.182

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	1.068	1	.301
	Block	1.068	1	.301
	Model	1.068	1	.301

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	853.820 <sup>a</sup>	.002	.002

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgro		
		Observed	Expected	Observed	Expected	Total
Step 1	1	156	156.000	297	297.000	453
	2	67	67.000	153	153.000	220

Classification Table <sup>a</sup>							
Observed Predicted			d				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 1	fnlgrd 0		0	223	.0		
	1		0	450	100.0		
	Overall Percentage				66.9		

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	techsex	.182	.177	1.059	1	.303	1.199
	Constant	.644	.099	42.402	1	.000	1.904

a. Variable(s) entered on step 1: techsex.

## Variables in the Equation

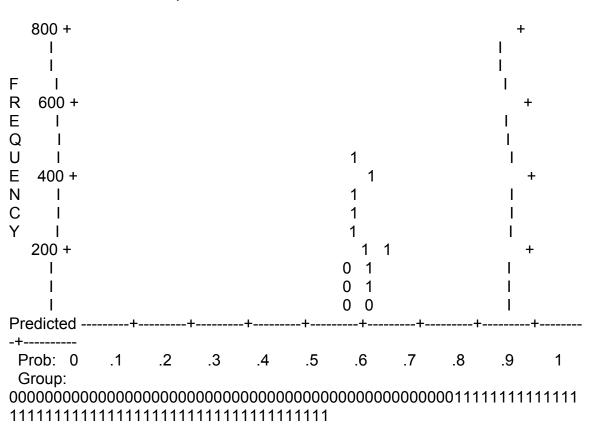
		95% C.I.fe	or EXP(B)	
		Lower Upper		
Step 1 <sup>a</sup>	techsex	.848	1.696	
	Constant			

a. Variable(s) entered on step 1: techsex.

**Correlation Matrix** 

		Constant	techsex
Step 1	Constant	1.000	559
	techsex	559	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER adj097 /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:25:11
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER adj097
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.016
	Elapsed Time	00:00:00.034

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary							
Unweighted Case	S <sup>a</sup>	N	Percent				
Selected Cases	Included in Analysis	673	99.7				
	Missing Cases	2	.3				
	Total	675	100.0				
Unselected Cases	;	0	.0				
Total		675	100.0				

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	855.001	.675
	2	854.888	.702
	3	854.888	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 854.888

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
Observed			Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	223	.0		
		1	0	450	100.0		
	Overall	Percentage			66.9		

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.702	.082	73.497	1	.000	2.018

### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables adj097	1.572	1	.210
Overall Statistics	1.572	1	.210

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coeffi	cients	
		-2 Log likelihood	Constant	adj097	
Step 1	1	853.461	.764	182	
	2	853.315	.804	205	
	3	853.315	.805	206	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

Iteration History<sup>a,b,c,d</sup>

		1		
Iteration			Coefficients	
		-2 Log likelihood	Constant	adj097
Step 1	1	853.461	.764	182
	2	853.315	.804	205
	3	853.315	.805	206

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	1.572	1	.210
	Block	1.572	1	.210
	Model	1.572	1	.210

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	853.315 <sup>a</sup>	.002	.003

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	117	117.000	213	213.000	330
	2	106	106.000	237	237.000	343

	Classification Table <sup>a</sup>						
	Observed		Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 1	fnlgrd 0		0	223	.0		
	1		0	450	100.0		
	Overall Percentage				66.9		

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	adj097	206	.164	1.570	1	.210	.814
	Constant	.805	.117	47.418	1	.000	2.236

a. Variable(s) entered on step 1: adj097.

## Variables in the Equation

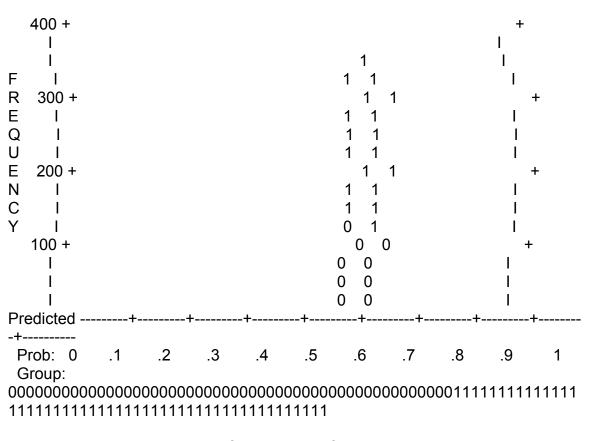
		95% C.I.for EXP(B)	
		Lower	Upper
Step 1 <sup>a</sup>	adj097	.590	1.123
	Constant		

a. Variable(s) entered on step 1: adj097.

**Correlation Matrix** 

		Constant	adj097
Step 1	Constant	1.000	712
	adj097	712	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 25 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

## LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER mozartuse /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:26:23
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER mozartuse
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.035

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	673	99.7		
	Missing Cases	2	.3		
	Total	675	100.0		
Unselected Cases	;	0	.0		
Total		675	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

## Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	855.001	.675
	2	854.888	.702
	3	854.888	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 854.888

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
	Observe	Predicted			d	
			fnl			
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	223	.0	
		1	0	450	100.0	
	Overall	Percentage			66.9	

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.702	.082	73.497	1	.000	2.018

### Variables not in the Equation

	Score	df	Sig.
Step 0 Variables Mozartuse	2.877	1	.090
Overall Statistics	2.877	1	.090

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	d Constant mozartuse		
Step 1	1	852.249	.718	394	
	2	852.102	.751	424	
	3	852.102	.751	424	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

Iteration History<sup>a,b,c,d</sup>

Iteration	l		Coefficients	
		-2 Log likelihood	Constant	mozartuse
Step 1	1	852.249	.718	394
	2	852.102	.751	424
	3	852.102	.751	424

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	2.785	1	.095
	Block	2.785	1	.095
	Model	2.785	1	.095

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	852.102 <sup>a</sup>	.004	.006

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	31	31.000	43	43.000	74
	2	192	192.000	407	407.000	599

	Classification Table <sup>a</sup>					
	Observed	Predicted		d		
		fnlgrd				
				Percentage		
		0	1	Correct		
Step 1	fnlgrd 0	0	223	.0		
	1	0	450	100.0		
	Overall Percentage			66.9		

a. The cut value is .500

Variables	in	the	Fa	uation
V al labico		unc.	-4	aation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	mozartuse	424	.251	2.847	1	.092	.654
	Constant	.751	.088	73.640	1	.000	2.120

a. Variable(s) entered on step 1: mozartuse.

# Variables in the Equation

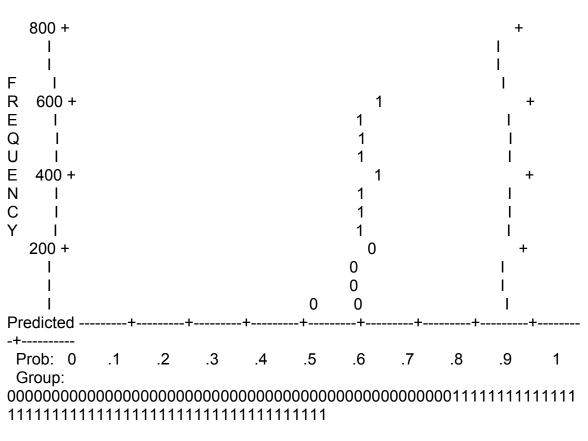
95% C.I.fc	or EXP(B)	
Lower Upper		
.400	1.071	
	Lower	

a. Variable(s) entered on step 1: mozartuse.

**Correlation Matrix** 

		Constant	mozartuse
Step 1	Constant	1.000	348
	mozartuse	348	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER ALEKSuse /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:30:21
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER ALEKSuse
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.047
	Elapsed Time	00:00:00.036

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary				
Unweighted Case	S <sup>a</sup>	N	Percent	
Selected Cases	Included in Analysis	673	99.7	
	Missing Cases	2	.3	
	Total	675	100.0	
Unselected Cases	;	0	.0	
Total		675	100.0	

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

# Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	855.001	.675
	2	854.888	.702
	3	854.888	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 854.888

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
Observed Predicted			d			
			fnl	grd		
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	223	.0	
		1	0	450	100.0	
	Overall	Percentage			66.9	

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.702	.082	73.497	1	.000	2.018

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables ALEKSuse	2.145	1	.143
Overall Statistics	2.145	1	.143

# Block 1: Method = Enter

	Iteration History <sup>a,b,c,d</sup>				
Iteration			Coefficients		
		-2 Log likelihood	Constant	ALEKSuse	
Step 1	1	852.960	.703	430	
	2	852.825	.734	459	
	3	852.825	.734	460	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

Iteration	History <sup>a,b,c,d</sup>
-----------	----------------------------

Iteration			Coefficients	
		-2 Log likelihood	Constant	ALEKSuse
Step 1	1	852.960	.703	430
	2	852.825	.734	459
	3	852.825	.734	460

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

<b>Omnibus Tests o</b>	f Model	Coefficients
------------------------	---------	--------------

		Chi-square	df	Sig.
Step 1	Step	2.063	1	.151
	Block	2.063	1	.151
	Model	2.063	1	.151

Model Summary

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	852.825 <sup>a</sup>	.003	.004

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0 fnlgrd		d = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	19	19.000	25	25.000	44
	2	204	204.000	425	425.000	629

	Classification Table <sup>a</sup>							
	Observed	Predicted						
		fnlgrd						
				Percentage				
		0	1	Correct				
Step 1	fnlgrd 0	0	223	.0				
	1	0	450	100.0				
	Overall Percentage			66.9				

a. The cut value is .500

Variables	in	the	Ea	uation
Variabico		unc.	- 4	aation

		В	S.E.	Wald	df	Sig.	Exp(B)	
Step 1 <sup>a</sup>	ALEKSuse	460	.316	2.114	1	.146	.632	
	Constant	.734	.085	74.255	1	.000	2.083	

a. Variable(s) entered on step 1: ALEKSuse.

# Variables in the Equation

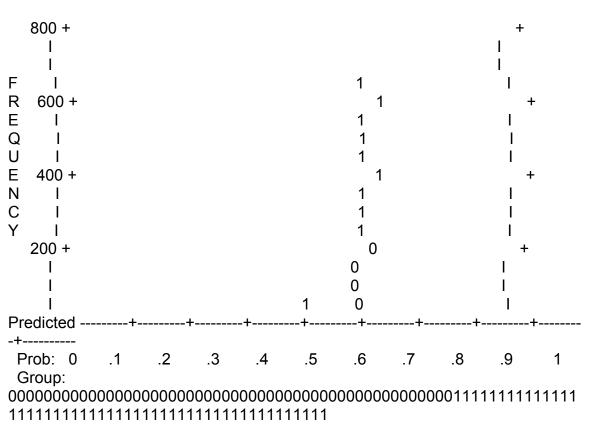
		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	ALEKSuse	.340	1.173	
	Constant			

a. Variable(s) entered on step 1: ALEKSuse.

**Correlation Matrix** 

		Constant	ALEKSuse
Step 1	Constant	1.000	270
	ALEKSuse	270	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER amisone /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:35:07
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER amisone
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.015
	Elapsed Time	00:00:00.031

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Case	S <sup>a</sup>	N	Percent			
Selected Cases	Included in Analysis	652	96.6			
	Missing Cases	23	3.4			
	Total	675	100.0			
Unselected Cases	3	0	.0			
Total		675	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

# Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	822.560	.699
	2	822.424	.730
	3	822.424	.730

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 822.424

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed		Predicte	d		
			fnl	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	212	.0		
		1	0	440	100.0		
	Overall	Percentage			67.5		

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.730	.084	76.280	1	.000	2.075

## Variables not in the Equation

		Score	df	Sig.
Step 0 Variables Ami	sone	.000	1	.991
Overall Statistics	.000	1	.991	

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>				
Iteration			Coeffi	cients
		-2 Log likelihood	Constant	amisone
Step 1	1	822.560	.699	.002
	2	822.424	.729	.002
	3	822.424	.730	.002

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 822.424

Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients	
		-2 Log likelihood	Constant	amisone
Step 1	1	822.560	.699	.002
	2	822.424	.729	.002
	3	822.424	.730	.002

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 822.424

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	.000	1	.991
	Block	.000	1	.991
	Model	.000	1	.991

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	822.424 <sup>a</sup>	.000	.000

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	.000	0	

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	135	135.000	280	280.000	415
	2	77	77.000	160	160.000	237

	Classification Table <sup>a</sup>			
	Observed	Predicted		d
		fnl	grd	
				Percentage
		0	1	Correct
Step 1	fnlgrd 0	0	212	.0
	1	0	440	100.0
	Overall Percentage			67.5

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	amisone	.002	.174	.000	1	.991	1.002
	Constant	.730	.105	48.474	1	.000	2.074

a. Variable(s) entered on step 1: amisone.

# Variables in the Equation

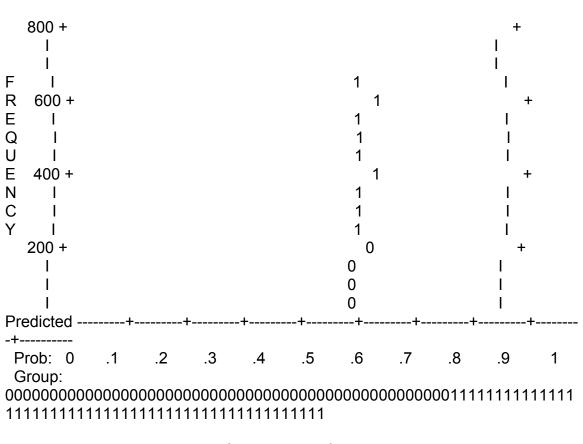
		95% C.I.fe	or EXP(B)
		Lower	Upper
Step 1 <sup>a</sup>	amisone	.713	1.409
	Constant		

a. Variable(s) entered on step 1: amisone.

**Correlation Matrix** 

		Constant	amisone
Step 1	Constant	1.000	603
	amisone	603	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 50 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER numbmeet /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:37:05
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER numbmeet
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.031
	Elapsed Time	00:00:00.038

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary					
Unweighted Case	S <sup>a</sup>	N	Percent		
Selected Cases	Included in Analysis	652	96.6		
	Missing Cases	23	3.4		
	Total	675	100.0		
Unselected Cases	3	0	.0		
Total		675	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
- 1	1

# **Block 0: Beginning Block**

# Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	822.560	.699
	2	822.424	.730
	3	822.424	.730

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 822.424

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>					
Observed				Predicte	d	
			fnl	grd		
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	212	.0	
		1	0	440	100.0	
	Overall	Percentage			67.5	

a. Constant is included in the model.

b. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.730	.084	76.280	1	.000	2.075

## Variables not in the Equation

		Score	df	Sig.
Step 0	Variables numbmeet	.891	1	.345
	Overall Statistics	.891	1	.345

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant numbmeet		
Step 1	1	821.688	.523	.055	
	2	821.530	.530	.063	
	3	821.530	.530	.063	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 822.424

	Iteration History <sup>a,b,c,d</sup>						
Iteration			Coefficients				
		-2 Log likelihood	Constant numbmeet				
Step 1	1	821.688	.523	.055			
	2	821.530	.530	.063			
	3	821.530	.530	.063			

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 822.424

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

		Chi-square	df	Sig.
Step 1	Step	.894	1	.344
	Block	.894	1	.344
	Model	.894	1	.344

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	821.530 <sup>a</sup>	.001	.002

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	.958	2	.619

### **Contingency Table for Hosmer and Lemeshow Test**

		fnlgrd = 0		fnlgro		
		Observed	Expected	Observed	Expected	Total
Step 1	1	111	109.351	209	210.649	320
	2	8	8.193	17	16.807	25
	3	46	50.560	115	110.440	161

		fnlgrd = 0		fnlgro		
		Observed	Expected	Observed	Expected	Total
Step 1	1	111	109.351	209	210.649	320
	2	8	8.193	17	16.807	25
	3	46	50.560	115	110.440	161
	4	47	43.896	99	102.104	146

**Contingency Table for Hosmer and Lemeshow Test** 

## Classification Table<sup>a</sup>

Observed		Predicted			
			fnlgrd		Percentage
			0	1	Correct
Step 1	fnlgrd	0	0	212	.0
		1	0	440	100.0
	Overall	Percentage			67.5

a. The cut value is .500

## Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	numbmeet	.063	.067	.890	1	.345	1.065
	Constant	.530	.227	5.448	1	.020	1.699

a. Variable(s) entered on step 1: numbmeet.

## Variables in the Equation

	95% C.I.for EXP(B)		
	Lower Upp		
numbmeet	.935	1.213	
Constant			
		Lower numbmeet .935	

a. Variable(s) entered on step 1: numbmeet.

Correlation Matrix							
		Constant	numbmeet				
Step 1	Constant	1.000	930				
	numbmeet	930	1.000				

Step number: 1 Observed Groups and Predicted Probabilities

320 + I F R 240 + E I Q I U I E 160 + N I C I Y I 80 + I I Predicted	+.			-+	+	$ \begin{array}{c} 1\\ 1\\ 1\\ 1\\ 1\\ 1\\ 1\\ 1\\ 1\\ 0\\ 11\\ 0\\ 11\\ 0\\ 11\\ 0\\ 0\\ 100\\+\\ +$	1 1	+-	+                   	
-+ Prob: 0 Group:	.1	.2	-		.5	.6	.7	.8	.9	1
0000000000 11111111111 Predic	11111	111111	111111	11111	000000 11111 ership f		000000			
The C Symb	ut Valu ols: 0 - 1 - 1	ie is .5								

Each Symbol Represents 20 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

# LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER classize /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT CORR ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		02-Oct-2011 14:38:25
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER classize
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT CORR ITER(1)
		CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.047
	Elapsed Time	00:00:00.038

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Case	N	Percent				
Selected Cases	Included in Analysis	673	99.7			
	Missing Cases	2	.3			
	Total	675	100.0			
Unselected Cases	;	0	.0			
Total		675	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

# **Block 0: Beginning Block**

# Iteration History<sup>a,b,c</sup>

Iteration			Coefficients
		-2 Log likelihood	Constant
Step 0	1	855.001	.675
	2	854.888	.702
	3	854.888	.702

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 854.888

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
	Observe	ed	Predicted				
			fnl				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	223	.0		
		1	0	450	100.0		
	Overall	Percentage			66.9		

### a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

	В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 Constant	.702	.082	73.497	1	.000	2.018

## Variables not in the Equation

	Score	df	Sig.
Step 0 Variables classize	.671	1	.413
Overall Statistics	.671	1	.413

# Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>					
Iteration			Coeffi	cients	
		-2 Log likelihood	Constant	classize	
Step 1	1	854.339	.421	.010	
	2	854.210	.412	.011	
	3	854.210	.412	.011	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 854.888

d. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	.678	1	.410
	Block	.678	1	.410
	Model	.678	1	.410

#### **Model Summary**

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	854.210 <sup>a</sup>	.001	.001

a. Estimation terminated at iteration number 3 because

parameter estimates changed by less than .001.

## Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	13.061	7	.071

			d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	32	26.690	45	50.310	77
	2	33	28.794	51	55.206	84
	3	12	14.970	32	29.030	44
	4	50	45.923	86	90.077	136
	5	21	31.837	74	63.163	95
	6	18	16.630	32	33.370	50
	7	19	24.737	57	51.263	76
	8	10	10.158	23	22.842	33
	9	28	23.261	50	54.739	78

## Contingency Table for Hosmer and Lemeshow Test

	Classification Table <sup>a</sup>					
Observed			Predicte	d		
			fnl	grd		
					Percentage	
			0	1	Correct	
Step 1	fnlgrd 0		0	223	.0	
	1		0	450	100.0	
	Overall Percentage				66.9	

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	classize	.011	.014	.670	1	.413	1.011
	Constant	.412	.363	1.288	1	.256	1.510

a. Variable(s) entered on step 1: classize.

# Variables in the Equation

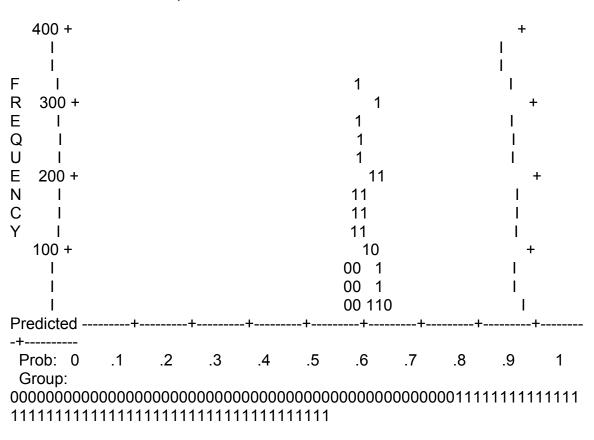
		95% C.I.fe	or EXP(B)
		Lower	Upper
Step 1 <sup>a</sup>	classize	.984	1.039
	Constant		

a. Variable(s) entered on step 1: classize.

**Correlation Matrix** 

		Constant	classize
Step 1	Constant	1.000	974
	classize	974	1.000

Step number: 1



**Observed Groups and Predicted Probabilities** 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 25 Cases.

Casewise List<sup>a</sup>

a. The casewise plot is not produced because no outliers were found.

Appendix U: SPSS Multiple Regression Output for Elementary Algebra

# GET

FILE='C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav'. DATASET NAME DataSet1 WINDOW=FRONT. REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

# Assumptions are discussed in bold.

	Notes	
Output Created		15-Oct-2011 14:09:50
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS BCOV R
		ANOVA COLLIN TOL CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:02.761
	Elapsed Time	00:00:02.861
	Memory Required	2524 bytes
	Additional Memory Required	904 bytes
	for Residual Plots	
Variables Created or	MAH_1	Mahalanobis Distance
Modified	COO_1	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Descriptive Statistics						
	Mean	Std. Deviation	Ν			
fnlexam	64.37	14.173	129			
pretest	40.43	15.996	129			
Act	14.58	1.368	129			

N = 129 with k = 2 predictors

N > 50 + 8k

Therefore, the sample size is appropriate.

Correlations						
		fnlexam	pretest	act		
Pearson Correlation	fnlexam	1.000	.334	.238		
	pretest	.334	1.000	.201		
	act	.238	.201	1.000		
Sig. (1-tailed)	fnlexam		.000	.003		
	pretest	.000		.011		
	act	.003	.011			
Ν	fnlexam	129	129	129		
	pretest	129	129	129		
	act	129	129	129		

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	act, pretest <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model	Summary <sup>b</sup>

Model					Change Statistics		
	_		Adjusted R	Std. Error of the	R Square		
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1
1	.377 <sup>a</sup>	<mark>.142</mark>	.128	13.233	.142	10.415	2

a. Predictors: (Constant), act, pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2	Sig. F Change	Durbin-Watson				
1	126	.000	1.821				

b. Dependent Variable: fnlexam

## ANOVA<sup>b</sup>

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	3647.484	2	1823.742	<mark>10.415</mark>	.000 <sup>a</sup>
	Residual	22062.655	126	175.100		
	Total	25710.140	128			

a. Predictors: (Constant), act, pretest

b. Dependent Variable: fnlexam

Coefficients <sup>a</sup>							
Model				Standardized			
		Unstandardize	ed Coefficients	Coefficients			
		B	Std. Error	Beta	t	Sig.	
1	(Constant)	<mark>26.810</mark>	12.533		2.139	<mark>.034</mark>	
	pretest	.264	.075	.298	3.539	.001	
	act	<mark>1.844</mark>	.873	.178	2.112	<mark>.037</mark>	

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>							
Model		(	Correlations			Collinearity Statistics	
		Zero-order Partial Part Tolerance			VIF		
1	(Constant)						
	pretest	.334	.301	.292	.960	<mark>1.042</mark>	
	act	.238	.185	.174	<mark>.960</mark>	<mark>1.042</mark>	

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

Coefficient Correlations <sup>a</sup>						
Model			act	pretest		
1	Correlations	act	1.000	201		
		pretest	201	1.000		
	Covariances	act	.762	013		
		pretest	013	.006		

а

Coefficient Correlations <sup>a</sup>						
Model			act	pretest		
1	Correlations	act	1.000	201		
		pretest	201	1.000		
	Covariances	act	.762	013		
		pretest	013	.006		

a. Dependent Variable: fnlexam

## Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions		
		Eigenvalue	Condition Index	(Constant)	pretest	act
1	1	2.906	1.000	.00	.01	.00
	_ 2	.089	5.706	.02	.98	.01
	3	.004	25.988	.98	.01	.99

a. Dependent Variable: fnlexam

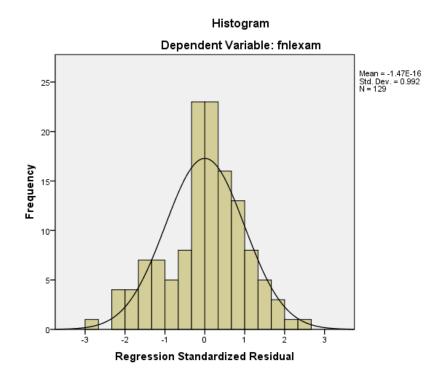
Residuals Statistics <sup>a</sup>					
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	50.53	82.70	64.37	5.338	129
Std. Predicted Value	-2.593	3.433	.000	1.000	129
Standard Error of Predicted	1.223	6.057	1.863	.778	129
Value					
Adjusted Predicted Value	51.24	80.62	64.40	5.358	129
Residual	-37.710	33.454	.000	13.129	129
Std. Residual	-2.850	2.528	.000	.992	129
Stud. Residual	-2.864	2.543	001	1.002	129
Deleted Residual	-38.083	33.853	024	13.391	129
Stud. Deleted Residual	-2.950	2.601	002	1.010	129
<mark>Mahal. Distance</mark>	<mark>.102</mark>	<mark>25.822</mark>	<mark>1.984</mark>	<mark>3.446</mark>	<mark>129</mark>
Cook's Distance	.000	<mark>.094</mark>	.007	.012	<mark>129</mark>
Centered Leverage Value	.001	.202	.016	.027	129

Residuals Statistics<sup>a</sup>

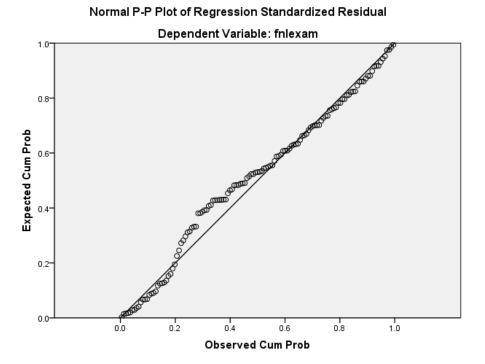
a. Dependent Variable: fnlexam

A maximum Mahalanobis distance 25.822 greater than the critical chisquare value of 13.816 for df = 2 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .094 (which is less than one) means outliers should not be a concern.

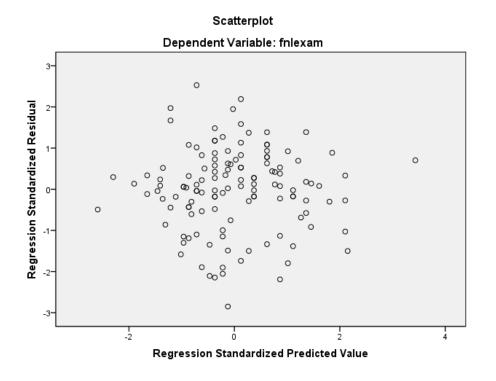
# Charts



# The points are clustered fairly close along the line indicating that the residuals are normally distributed.



# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



# REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act adj096 /SCATTERPLOT=(\*ZRESID,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

Notes					
Output Created		15-Oct-2011 14:13:48			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet1			

	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS BCOV R
		ANOVA COLLIN TOL CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act adj096
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Decourses		00.00.04 057
Resources	Processor Time	00:00:01.357
	Elapsed Time	00:00:01.701
	Memory Required	2860 bytes
	Additional Memory Required	896 bytes
Variables Created at	for Residual Plots	Mahalanobis Distance
Variables Created or	MAH_2	
Modified	COO_2	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Descriptive Statistics					
	Mean	Std. Deviation	Ν		
fnlexam	64.37	14.173	129		
pretest	40.43	15.996	129		
act	14.58	1.368	129		
adj096	.42	.495	129		

# N = 129 with k = 3 predictors

N > 50 + 8k

# Therefore, the sample size is appropriate.

Correlations						
		fnlexam	pretest	act	adj096	
Pearson Correlation	Fnlexam	1.000	.334	.238	.214	
	Pretest	.334	1.000	.201	.130	
	Act	.238	.201	1.000	155	
	adj096	.214	.130	155	1.000	
Sig. (1-tailed)	Fnlexam		.000	.003	.008	
	Pretest	.000		.011	.071	
	Act	.003	.011		.040	
	adj096	.008	.071	.040		
Ν	Fnlexam	129	129	129	129	
	Pretest	129	129	129	129	
	Act	129	129	129	129	
	adj096	129	129	129	129	

### Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	adj096, pretest,		Enter

a. All requested variables entered.

Model Summary <sup>b</sup>									
Model					Char	nge Statistics			
	_		Adjusted R	Std. Error of the	R Square				
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1		
1	.430 <sup>a</sup>	.185	.165	12.947	.185	9.457	3		

a. Predictors: (Constant), adj096, pretest, act

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang						
	df2	Sig. F Change	Durbin-Watson				
1	125	.000	1.876				

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
Mode	9	Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	4756.040	3	1585.347	<mark>9.457</mark>	.000 <sup>a</sup>		
	Residual	20954.099	125	167.633				
	Total	25710.140	128					

a. Predictors: (Constant), adj096, pretest, act

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>							
Model		Unstandardize	ed Coefficients	Standardized Coefficients				
		B	Std. Error	Beta	t	Sig.		
<mark>1</mark>	(Constant)	<mark>19.474</mark>	12.590		1.547	<mark>.124</mark>		
	pretest	<mark>.232</mark>	.074	.262	3.138	.002		
	act	<mark>2.260</mark>	.869	.218	2.599	<mark>.010</mark>		
	adj096	<mark>6.100</mark>	2.372	.213	2.572	.011		

Coefficients <sup>a</sup>								
Model		(	Correlations		Collinearity	Collinearity Statistics		
		Zero-order	Partial	Part	Tolerance	VIF		
1	(Constant)							
	pretest	.334	.270	.253	<mark>.933</mark>	<mark>1.072</mark>		
	act	.238	.226	.210	.926	<mark>1.079</mark>		
	adj096	.214	.224	.208	<mark>.949</mark>	<mark>1.054</mark>		

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations"						
Model			adj096	pretest	act		
1	Correlations	adj096	1.000	167	.186		
		pretest	167	1.000	226		
		act	.186	226	1.000		
	Covariances	adj096	5.626	029	.384		
		pretest	029	.005	015		
		act	.384	015	.756		

Coefficient Correlations<sup>a</sup>

a. Dependent Variable: fnlexam

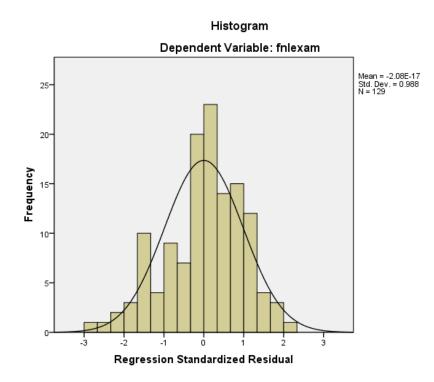
Model	Dimension			Variance Proportions			
		Eigenvalue	Condition Index	(Constant)	pretest	act	adj096
1	1	3.415	1.000	.00	.01	.00	.03
	2	.492	2.634	.00	.01	.00	.93
	- 3	.089	6.193	.02	.96	.01	.00
	4	.004	28.804	.98	.01	.99	.04

Collinearity Diagnostics<sup>a</sup>

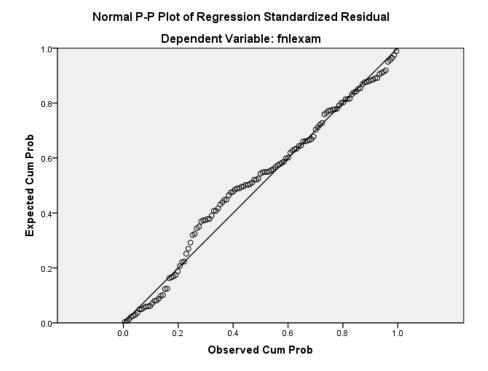
Residuals Statistics <sup>a</sup>								
	Minimum	Maximum	Mean	Std. Deviation	N			
Predicted Value	46.72	82.03	64.37	6.096	129			
Std. Predicted Value	-2.896	2.897	.000	1.000	129			
Standard Error of Predicted	1.510	5.931	2.175	.687	129			
Value								
Adjusted Predicted Value	47.06	79.80	64.40	6.123	129			
Residual	-35.503	29.819	.000	12.795	129			
Std. Residual	-2.742	2.303	.000	.988	129			
Stud. Residual	-2.762	2.331	001	1.002	129			
Deleted Residual	-36.014	30.542	024	13.160	129			
Stud. Deleted Residual	-2.839	2.374	003	1.010	129			
<mark>Mahal. Distance</mark>	<mark>.748</mark>	<mark>25.865</mark>	<mark>2.977</mark>	<mark>3.359</mark>	<mark>129</mark>			
Cook's Distance	<mark>.000</mark>	<mark>.081</mark>	<mark>.007</mark>	.011	<mark>129</mark>			
Centered Leverage Value	.006	.202	.023	.026	129			

A maximum Mahalanobis distance 25.865 greater than the critical chisquare value of 16.266 for df = 3 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .081 (which is less than one) means outliers should not be a concern.

## Charts

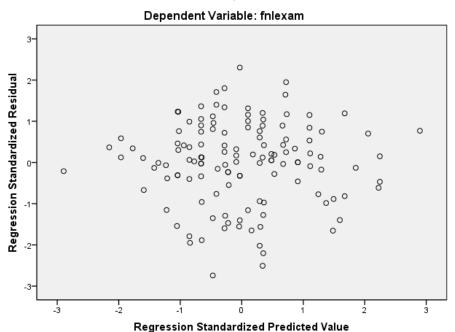


The points are clustered fairly close along the line indicating that the residuals are normally distributed.



# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.

Scatterplot



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act adj096 comcol /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

Notes							
Output Created		15-Oct-2011 14:14:51					
Comments							
Input	Data	C:\Users\Lin\Documents\math 096 fall					
		2001 with classrooms.sav					
	Active Dataset	DataSet1					
	Filter	<none></none>					

	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data File	252
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
-,		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS BCOV R
		ANOVA COLLIN TOL CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act adj096
		comcol
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Basauraaa	Dragonar Time	00.00.04 425
Resources	Processor Time	00:00:01.435 00:00:01.743
	Elapsed Time	
	Memory Required	3228 bytes
	Additional Memory Required	888 bytes
Variables Created ar	for Residual Plots	Mahalanobis Distance
Variables Created or	MAH_3	
Modified	COO_3	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Descriptive Statistics							
	Mean	Std. Deviation	N				
fnlexam	64.37	14.173	129				
pretest	40.43	15.996	129				
act	14.58	1.368	129				
adj096	.42	.495	129				
comcol	.53	.501	129				

# N = 129 with k = 4 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations									
		fnlexam	pretest	act	adj096	comcol			
Pearson Correlation	fnlexam	1.000	.334	.238	.214	142			
	pretest	.334	1.000	.201	.130	189			
	act	.238	.201	1.000	155	211			
	adj096	.214	.130	155	1.000	.017			
	comcol	142	189	211	.017	1.000			
Sig. (1-tailed)	fnlexam		.000	.003	.008	.054			
	pretest	.000		.011	.071	.016			
	act	.003	.011		.040	.008			
	adj096	.008	.071	.040		.425			
	comcol	.054	.016	.008	.425				
Ν	fnlexam	129	129	129	129	129			
	pretest	129	129	129	129	129			
	act	129	129	129	129	129			
	adj096	129	129	129	129	129			
	comcol	129	129	129	129	129			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	comcol, adj096, pretest, act <sup>a</sup>		Enter

a. All requested variables entered.

#### Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables						
	Entered	Removed	Method					
1	comcol, adj096,		Enter					
	pretest, act <sup>a</sup>							

a. All requested variables entered.

b. Dependent Variable: fnlexam

## Model Summary<sup>b</sup>

Model					Change Statistics		
			Adjusted R	Std. Error of the	R Square	5.01	
	R R	R Square	Square	Estimate	Change	F Change	df1
1	.433 <sup>a</sup>	<mark>.188</mark>	.161	12.978	.188	7.162	4

a. Predictors: (Constant), comcol, adj096, pretest, act

b. Dependent Variable: fnlexam

### Model Summary<sup>b</sup>

Model	Chang		
	df2	Sig. F Change	Durbin-Watson
1	124	.000	1.878

#### b. Dependent Variable: fnlexam

	ANOVA <sup>D</sup>								
Mode		Sum of Squares	df	Mean Square	F	Sig.			
1	Regression	4825.341	4	1206.335	<mark>7.162</mark>	.000 <sup>a</sup>			
	Residual	20884.798	124	168.426					
	Total	25710.140	128		i.				

a. Predictors: (Constant), comcol, adj096, pretest, act

_	Coefficients <sup>a</sup>								
Model		Unstandardize	ed Coefficients	Standardized Coefficients					
		B	Std. Error	Beta	t	Sig.			
1	(Constant)	<mark>22.019</mark>	13.229		1.664	<mark>.099</mark>			
	pretest	.225	.075	.254	2.996	.003			
	act	<mark>2.160</mark>	.885	.208	2.440	<mark>.016</mark>			
	adj096	<mark>6.114</mark>	2.378	.214	2.571	<mark>.011</mark>			
	comcol	<mark>-1.520</mark>	2.370	054	641	<mark>.522</mark>			

<b>Coefficients</b> <sup>a</sup>
----------------------------------

Model		Correlations			Collinearity Statistics	
		Zero-order	Partial	Part	Tolerance	
1	(Constant)					
	pretest	.334	.260	.243	<mark>.911</mark>	<mark>1.097</mark>
	act	.238	.214	.197	<mark>.898</mark>	<mark>1.114</mark>
	adj096	.214	.225	.208	<mark>.949</mark>	<mark>1.054</mark>
	comcol	142	058	052	<mark>.933</mark>	<mark>1.072</mark>

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations"							
Model			comcol	adj096	pretest	act		
1	Correlations	Comcol	1.000	009	.152	.175		
		adj096	009	1.000	166	.181		
		Pretest	.152	166	1.000	193		
		Act	.175	.181	193	1.000		
	Covariances	Comcol	5.615	053	.027	.368		
		adj096	053	5.653	030	.382		
		Pretest	.027	030	.006	013		
		Act	.368	.382	013	.784		

Coefficient Correlations<sup>a</sup>

Model	Dimension			Variance Proportions		ons
		Eigenvalue	Condition Index	(Constant)	pretest	act
1	1	3.969	1.000	.00	.01	.00
	2	.534	2.727	.00	.00	.00
	_ 3	.413	3.100	.00	.05	.00
	4	.080	7.039	.02	.94	.02
	5	.004	31.999	.98	.01	.98

**Collinearity Diagnostics**<sup>a</sup>

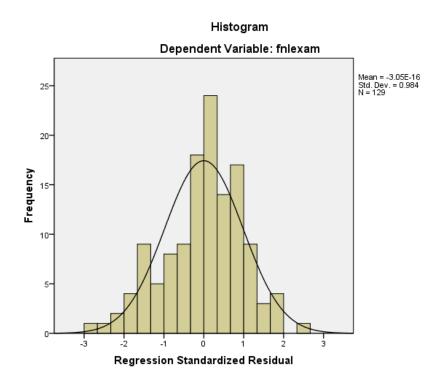
Model	Dimension	Variance Proportions				
		adj096	comcol			
1	1	.02	.02			
	2	.72	.23			
	_ 3	.22	.58			
	4	.00	.11			
	5	.04	.06			

#### Collinearity Diagnostics<sup>a</sup>

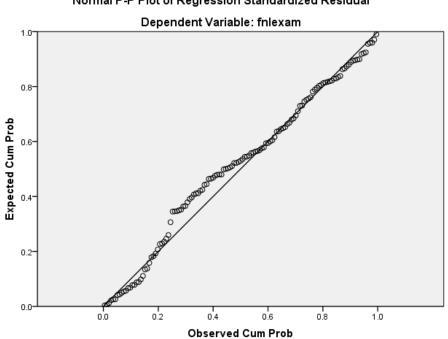
Residuals Statistics <sup>a</sup>									
	Minimum Maximum Mean Std. Deviation N								
Predicted Value	46.60	80.49	64.37	6.140	129				
Std. Predicted Value	-2.894	2.625	.000	1.000	129				
Standard Error of Predicted	1.916	6.053	2.475	.638	129				
Value									
Adjusted Predicted Value	46.93	78.84	64.40	6.164	129				
Residual	-34.777	30.395	.000	12.774	129				
Std. Residual	-2.680	2.342	.000	.984	129				
Stud. Residual	-2.709	2.376	001	1.002	129				
Deleted Residual	-35.552	31.286	025	13.256	129				
Stud. Deleted Residual	-2.782	2.422	003	<u>1.011</u>	129				
<mark>Mahal. Distance</mark>	<mark>1.798</mark>	<mark>26.851</mark>	<mark>3.969</mark>	<mark>3.441</mark>	<mark>129</mark>				
Cook's Distance	.000	.068	<mark>.008</mark>	.012	<mark>129</mark>				
Centered Leverage Value	.014	.210	.031	.027	129				

A maximum Mahalanobis distance 26.851 greater than the critical chisquare value of 18.467 for df = 4 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .068 (which is less than one) means outliers should not be a concern.

## Charts

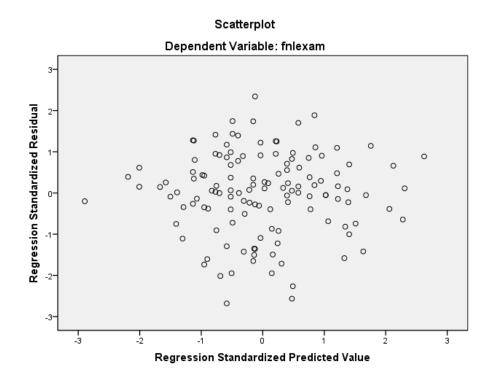


# The points are clustered fairly close along the line indicating that the residuals are normally distributed.



Normal P-P Plot of Regression Standardized Residual

# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act adj096 numbmeet /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

Notes						
Output Created		15-Oct-2011 14:16:37				
Comments						
Input	Data	C:\Users\Lin\Documents\math 096 fall				
		2001 with classrooms.sav				
	Active Dataset	DataSet1				

	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS BCOV R
		ANOVA COLLIN TOL CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act adj096
		numbmeet
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Basauraaa	Processor Time	00:00:01.404
Resources	Elapsed Time	00:00:01.404
	Memory Required	3268 bytes
	Additional Memory Required for Residual Plots	888 bytes
Variables Created or		Mahalanobis Distance
	MAH_4	
Modified	COO_4	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Descriptive Statistics							
	Mean	Std. Deviation	N				
fnlexam	64.37	14.173	129				
pretest	40.43	15.996	129				
act	14.58	1.368	129				
adj096	.42	.495	129				
numbmeet	3.61	1.342	129				

# N = 129 with k = 4 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations								
		fnlexam	pretest	act	adj096	numbmeet		
Pearson Correlation	fnlexam	1.000	.334	.238	.214	180		
	pretest	.334	1.000	.201	.130	074		
	act	.238	.201	1.000	155	.141		
	adj096	.214	.130	155	1.000	588		
	numbmeet	180	074	.141	588	1.000		
Sig. (1-tailed)	fnlexam		.000	.003	.008	.020		
	pretest	.000		.011	.071	.202		
	Act	.003	.011		.040	.056		
	adj096	.008	.071	.040	-	.000		
	numbmeet	.020	.202	.056	.000			
Ν	Fnlexam	129	129	129	129	129		
	Pretest	129	129	129	129	129		
	Act	129	129	129	129	129		
	adj096	129	129	129	129	129		
	numbmeet	129	129	129	129	129		

Variables Entered/Removed <sup>b</sup>							
Model	Variables	Variables					
	Entered	Removed	Method				
1	numbmeet,		Enter				
	pretest, act,						
	pretest, act, adj096 <sup>a</sup>						

a. All requested variables entered.

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
	D		Adjusted R	Std. Error of the	R Square	C Change	454
	R R	R Square	Square	Estimate	Change	F Change	df1
1	.438 <sup>a</sup>	<mark>.192</mark>	.166	12.946	.192	7.354	4

a. Predictors: (Constant), numbmeet, pretest, act, adj096

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model	Chang	ge Statistics		
	df2	Sig. F Change	Durbin-Watson	
1	124	.000	1.891	

b. Dependent Variable: fnlexam

ANOVA <sup>b</sup>	
--------------------	--

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	4929.457	4	1232.364	<mark>7.354</mark>	.000 <sup>a</sup>
	Residual	20780.682	124	167.586		
	Total	25710.140	128			

a. Predictors: (Constant), numbmeet, pretest, act, adj096

Coefficients <sup>a</sup>										
Model		Unstandardiz	ed Coefficients	Standardized Coefficients						
		B	Std. Error	Beta	t	Sig.				
1	(Constant)	<mark>23.279</mark>	13.132		1.773	.079				
	pretest	<mark>.232</mark>	.074	.261	3.127	.002				
	act	<mark>2.316</mark>	.871	.223	2.659	<mark>.009</mark>				
	adj096	<mark>4.413</mark>	2.894	.154	1.525	<mark>.130</mark>				
	numbmeet	<mark>-1.075</mark>	1.056	102	-1.017	. <mark>311</mark>				

Coefficients <sup>a</sup>								
Model		(	Correlations		Collinearity	Statistics		
		Zero-order	Partial	Part	Tolerance	VIF		
1	(Constant)							
	pretest	.334	.270	.252	<mark>.933</mark>	<mark>1.072</mark>		
	act	.238	.232	.215	<mark>.923</mark>	<mark>1.084</mark>		
	adj096	.214	.136	.123	.638	<mark>1.569</mark>		
	numbmeet	180	091	082	.651	<mark>1.536</mark>		

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations <sup>a</sup>								
Model			numbmeet	pretest	act	adj096			
1	Correlations	numbmeet	1.000	.011	063	.573			
		pretest	.011	1.000	226	130			
		act	063	226	1.000	.116			
		adj096	.573	130	.116	1.000			
	Covariances	numbmeet	1.116	.001	058	1.751			
		pretest	.001	.005	015	028			
		act	058	015	.759	.292			
		adj096	1.751	028	.292	8.373			

Model	Dimension			Variance Proportions		ons
		Eigenvalue	Condition Index	(Constant)	pretest	act
1	1	4.236	1.000	.00	.01	.00
	2	.607	2.641	.00	.00	.00
	_ 3	.111	6.182	.00	.83	.00
	4	.042	10.068	.04	.15	.05
	5	.004	32.320	.96	.01	.95

Collinearity Diagnostics<sup>a</sup>

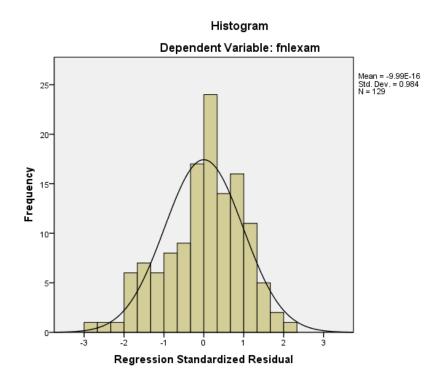
Commeanly Diagnostics						
Model	Dimension	Variance Proportions				
		adj096	numbmeet			
1	1	.01	.00			
	2	.49	.02			
	_ 3	.12	.14			
	4	.32	.82			
	5	.06	.02			

#### Collinearity Diagnostics<sup>a</sup>

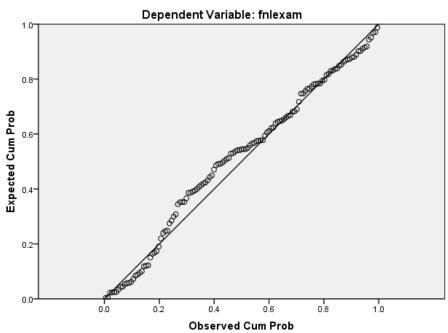
Residuals Statistics <sup>a</sup>							
	Minimum	Maximum	Mean	Std. Deviation	N		
Predicted Value	48.92	82.66	64.37	6.206	129		
Std. Predicted Value	-2.490	2.947	.000	1.000	129		
Standard Error of Predicted	1.573	5.965	2.442	.733	129		
Value							
Adjusted Predicted Value	49.71	80.54	64.41	6.227	129		
Residual	-34.746	29.090	.000	12.742	129		
Std. Residual	-2.684	2.247	.000	.984	129		
Stud. Residual	-2.708	2.278	002	1.002	129		
Deleted Residual	-35.365	29.890	043	13.199	129		
Stud. Deleted Residual	-2.780	2.318	003	1.009	129		
<mark>Mahal. Distance</mark>	<mark>.898</mark>	<mark>26.189</mark>	<mark>3.969</mark>	<mark>3.606</mark>	<mark>129</mark>		
Cook's Distance	<mark>.000</mark>	<mark>.073</mark>	.007	<mark>.011</mark>	<mark>129</mark>		
Centered Leverage Value	.007	.205	.031	.028	129		

A maximum Mahalanobis distance 26.189 greater than the critical chisquare value of 18.467 for df = 4 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .073 (which is less than one) means outliers should not be a concern.

# Charts

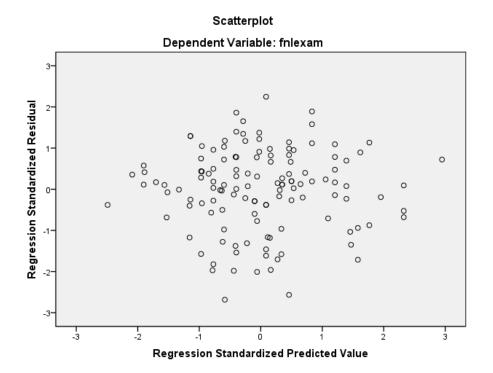


# The points are clustered fairly close along the line indicating that the residuals are normally distributed.



Normal P-P Plot of Regression Standardized Residual

# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act adj096 /METHOD=ENTER comcol numbmeet /SCATTERPLOT=(\*ZRESID,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

 Notes

 Output Created
 15-Oct-2011 14:20:30

 Comments
 15-Oct-2011 14:20:30

Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	202
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS BCOV R
		ANOVA COLLIN TOL CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act adj096
		/METHOD=ENTER comcol
		numbmeet
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.420
	Elapsed Time	00:00:01.709
	Memory Required	3724 bytes
	Additional Memory Required	880 bytes
	for Residual Plots	
Variables Created or	MAH_5	Mahalanobis Distance
Modified	_ COO_5	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

	Descriptive Statistics							
	Mean	Std. Deviation	Ν					
fnlexam	64.37	14.173	129					
pretest	40.43	15.996	129					
act	14.58	1.368	129					
adj096	.42	.495	129					
comcol	.53	.501	129					
numbmeet	3.61	1.342	129					

#### ecriptivo Statietic

# N = 129 with k = 5 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations							
		fnlexam	pretest	act	adj096		
Pearson Correlation	fnlexam	1.000	.334	.238	.214		
	pretest	.334	1.000	.201	.130		
	act	.238	.201	1.000	155		
	adj096	.214	.130	155	1.000		
	comcol	142	189	211	.017		
	numbmeet	180	074	.141	588		
Sig. (1-tailed)	fnlexam		.000	.003	.008		
	pretest	.000		.011	.071		
	act	.003	.011		.040		
	adj096	.008	.071	.040			
	comcol	.054	.016	.008	.425		
	numbmeet	.020	.202	.056	.000		
Ν	fnlexam	129	129	129	129		
	pretest	129	129	129	129		
	act	129	129	129	129		
	adj096	129	129	129	129		
	comcol	129	129	129	129		
	numbmeet	129	129	129	129		

Corrolations

Correlations						
		comcol	numbmeet			
Pearson Correlation	fnlexam	142	180			
	pretest	189	074			
	act	211	.141			
	adj096	.017	588			
	comcol	1.000	.027			
	numbmeet	.027	1.000			
Sig. (1-tailed)	fnlexam	.054	.020			
	pretest	.016	.202			
	act	.008	.056			
	adj096	.425	.000			
	comcol		.379			
	numbmeet	.379				
Ν	fnlexam	129	129			
	pretest	129	129			
	act	129	129			
	adj096	129	129			
	comcol	129	129			
	numbmeet	129	129			

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	adj096, pretest, act <sup>a</sup>		Enter
2	comcol, numbmeet <sup>a</sup>		Enter

a. All requested variables entered.

	Model Summary <sup>c</sup>									
Model					Change Statistics					
	_		Adjusted R	Std. Error of the	R Square					
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1			
1	.430 <sup>a</sup>	<mark>.185</mark>	.165	12.947	.185	9.457	3			
2	.440 <sup>b</sup>	<mark>.194</mark>	.161	12.980	.009	.684	2			

a. Predictors: (Constant), adj096, pretest, act

b. Predictors: (Constant), adj096, pretest, act, comcol, numbmeet

c. Dependent Variable: fnlexam

Model Summary <sup>c</sup>							
Model	Chang	ge Statistics					
	df2	Sig. F Change	Durbin-Watson				
1	125	.000					
2	123	.507	1.892				

#### c. Dependent Variable: fnlexam

	ANOVA <sup>c</sup>						
Mode	I	Sum of Squares	df	Mean Square	F	Sig.	
1	Regression	4756.040	3	1585.347	<mark>9.457</mark>	.000 <sup>a</sup>	
	Residual	20954.099	125	167.633			
	Total	25710.140	128				
2	Regression	4986.482	5	997.296	<mark>5.919</mark>	.000 <sup>b</sup>	
	Residual	20723.657	123	168.485			
	Total	25710.140	128				

a. Predictors: (Constant), adj096, pretest, act

b. Predictors: (Constant), adj096, pretest, act, comcol, numbmeet

			Coefficients <sup>a</sup>			
Model		Unstandardize	ed Coefficients	Standardized Coefficients		
		В	Std. Error	Beta	t	Sig.
1	(Constant)	19.474	12.590		1.547	.124
	pretest	.232	.074	.262	3.138	.002
	act	2.260	.869	.218	2.599	.010
	adj096	6.100	2.372	.213	2.572	.011
2	(Constant)	25.461	13.691		1.860	.065
	pretest	.225	.075	.254	2.994	.003
	act	2.223	.888	.215	2.504	.014
	adj096	4.484	2.904	.157	1.544	.125
	comcol	-1.381	2.374	049	582	.562
	numbmeet	-1.038	1.061	098	978	.330

			Coefficients			
Model		(	Correlations	Collinearity Statistics		
		Zero-order	Partial	Part	Tolerance	VIF
1	(Constant)					
	pretest	.334	.270	.253	<mark>.933</mark>	<mark>1.072</mark>
	act	.238	.226	.210	<mark>.926</mark>	<mark>1.079</mark>
	adj096	.214	.224	.208	<mark>.949</mark>	<mark>1.054</mark>
2	(Constant)					
	pretest	.334	.261	.242	<mark>.911</mark>	<mark>1.097</mark>
	act	.238	.220	.203	<mark>.893</mark>	<mark>1.120</mark>
	adj096	.214	.138	.125	<mark>.636</mark>	<mark>1.571</mark>
	comcol	142	052	047	<mark>.930</mark>	<mark>1.076</mark>
	numbmeet	180	088	079	<mark>.649</mark>	<mark>1.541</mark>

Coefficients<sup>a</sup>

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Excluded Variables <sup>b</sup>						
Model					Partial		
		Beta In	t	Sig.	Correlation		
1	comcol	054 <sup>a</sup>	641	.522	058		
	numbmeet	102 <sup>a</sup>	-1.017	.311	091		

a. Predictors in the Model: (Constant), adj096, pretest, act

b. Dependent Variable: fnlexam

	Excluded Variables <sup>b</sup>						
Mode	el	C	Collinearity Statistics				
				Minimum			
		Tolerance	VIF	Tolerance			
1	comcol	.933	1.072	.898			
	numbmeet	.651	1.536	.638			

Model			adj096	pretest	act	comcol	Numbmeet
1	Correlations	adj096	1.000	167	.186		
		pretest	167	1.000	226		
		act	.186	226	1.000		
	Covariances	adj096	5.626	029	.384		
		pretest	029	.005	015		
		act	.384	015	.756		
2	Correlations	adj096	1.000	135	.106	042	.574
		pretest	135	1.000	193	.152	.002
		act	.106	193	1.000	.179	073
		comcol	042	.152	.179	1.000	060
		numbmeet	.574	.002	073	060	1.000
	Covariances	adj096	8.432	029	.274	289	1.768
		pretest	029	.006	013	.027	.000
		act	.274	013	.788	.377	069
		comcol	289	.027	.377	5.637	151
		numbmeet	1.768	.000	069	151	1.126

**Coefficient Correlations**<sup>a</sup>

Model	Dimension				Variance Pr	oportions	
		Eigenvalue	Condition Index	(Constant)	pretest	act	adj096
1	1	3.415	1.000	.00	.01	.00	.03
	2	.492	2.634	.00	.01	.00	.93
	3	.089	6.193	.02	.96	.01	.00
	4	.004	28.804	.98	.01	.99	.04
2	1	4.795	1.000	.00	.00	.00	.01
	2	.614	2.795	.00	.00	.00	.49
	3	.443	3.289	.00	.03	.00	.01
	- 4	.103	6.840	.00	.79	.00	.15
	5	.041	10.758	.04	.17	.05	.30
	6	.004	35.368	.96	.01	.95	.05

Collinearity Diagnostics<sup>a</sup>

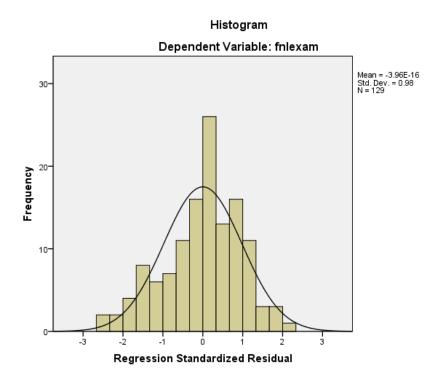
Model	Dimension	Variance	Proportions
		comcol	numbmeet
1	1		
	2		
	- 3		
	4		
2	1	.01	.00
	2	.02	.02
	3	.81	.01
	- 4	.09	.18
	5	.01	.78
	6	.06	.01

# Collinearity Diagnostics<sup>a</sup>

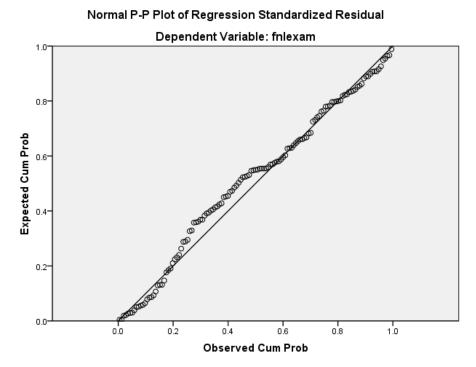
Residuals Statistics <sup>a</sup>						
	Minimum	Maximum	Mean	Std. Deviation	N	
Predicted Value	48.74	81.24	64.37	6.242	129	
Std. Predicted Value	-2.505	2.702	.000	1.000	129	
Standard Error of Predicted	1.954	6.102	2.715	.686	129	
Value						
Adjusted Predicted Value	49.51	79.73	64.42	6.262	129	
Residual	-34.112	29.638	.000	12.724	129	
Std. Residual	-2.628	2.283	.000	.980	129	
Stud. Residual	-2.672	2.321	002	1.002	129	
Deleted Residual	-35.403	30.619	046	13.301	129	
Stud. Deleted Residual	-2.741	2.364	004	1.010	129	
<mark>Mahal. Distance</mark>	<mark>1.909</mark>	<mark>27.295</mark>	<mark>4.961</mark>	<mark>3.684</mark>	<mark>129</mark>	
Cook's Distance	.000	<mark>.063</mark>	.008	.012	<mark>129</mark>	
Centered Leverage Value	.015	.213	.039	.029	129	

A maximum Mahalanobis distance 27.295 greater than the critical chisquare value of 20.515 for df = 5 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .063 (which is less than one) means outliers should not be a concern.

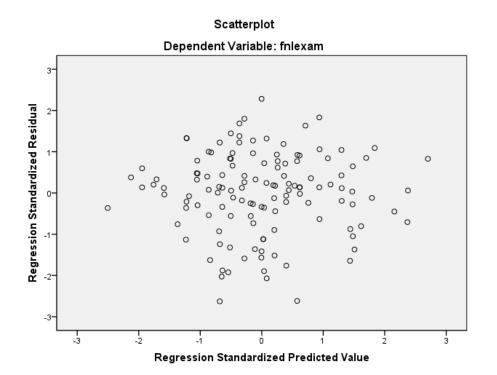
## Charts



# The points are clustered fairly close along the line indicating that the residuals are normally distributed.



# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



Appendix V: SPSS Multiple Regression Output for Intermediate Algebra

### GET

FILE='C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav'. DATASET NAME DataSet1 WINDOW=FRONT. REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:30:20
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.810
	Elapsed Time	00:00:02.857
	Memory Required	2524 bytes
	Additional Memory Required	904 bytes
	for Residual Plots	
Variables Created or	MAH_1	Mahalanobis Distance
Modified	COO_1	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no
names.sav

Descriptive Statistics					
	Mean	Std. Deviation	N		
fnlexam	68.84	14.532	394		
pretest	47.14	14.380	394		
act	16.86	1.147	394		

N = 394 with k = 2 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations								
		fnlexam	pretest	act				
		mexam	protost	401				
Pearson Correlation	fnlexam	1.000	.299	.219				
	pretest	.299	1.000	.188				
	act	.219	.188	1.000				
Sig. (1-tailed)	fnlexam		.000	.000				
	pretest	.000		.000				
	act	.000	.000					
Ν	fnlexam	394	394	394				
	pretest	394	394	394				
	act	394	394	394				

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	act, pretest <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

Model	Summary <sup>b</sup>

Model					Change Statistics		
	_		Adjusted R	Std. Error of the	R Square		
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1
1	.342 <sup>a</sup>	<mark>.117</mark>	.112	13.690	.117	25.902	2

a. Predictors: (Constant), act, pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2 Sig. F Change		Durbin-Watson				
1	391	.000	1.644				

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	9708.947	2	4854.473	<mark>25.902</mark>	.000 <sup>a</sup>		
	Residual	73279.979	391	187.417				
	Total	82988.926	393					

a. Predictors: (Constant), act, pretest

b. Dependent Variable: fnlexam

	Coefficients <sup>a</sup>								
Model				Standardized					
Unstandardized Coe		d Coefficients	Coefficients						
		B	Std. Error	Beta	t	Sig.			
1	(Constant)	<mark>20.067</mark>	10.187		1.970	<mark>.050</mark>			
	pretest	<mark>.271</mark>	.049	.268	5.533	.000			
	<mark>Act</mark>	<mark>2.136</mark>	.613	.169	3.483	.001			

a. Dependent Variable: fnlexam

**Coefficients**<sup>a</sup>

N	lodel	95.0% Confiden	ce Interval for B	B Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	.039	40.095			
	pretest	.174	.367	.299	.269	.263
	Act	.930	3.342	.219	.173	.166

	Coefficients <sup>a</sup>							
Model		Collinearity Statistics						
		Tolerance VIF						
1	(Constant)							
	pretest		<mark>.965</mark>	<mark>1.0</mark> ;	<mark>36</mark>			
	Act		<mark>.965</mark>	<mark>1.0</mark> 3	<mark>36</mark>			

Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

Coefficient Correlations <sup>a</sup>							
Model			act	pretest			
1	Correlations	Act	1.000	188			
		Pretest	188	1.000			
	Covariances	Act	.376	006			
		Pretest	006	.002			

#### Collinearity Diagnostics<sup>a</sup>

Model	Dimension			Variance Proportions		ons
		Eigenvalue	Condition Index	(Constant)	pretest	act
1	1	2.942	1.000	.00	.01	.00
	_ 2	.056	7.249	.01	.98	.01
	3	.002	35.895	.99	.01	.99

a. Dependent Variable: fnlexam

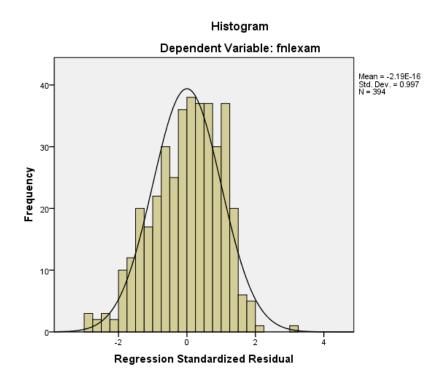
#### Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.119	100	57.30	42.697

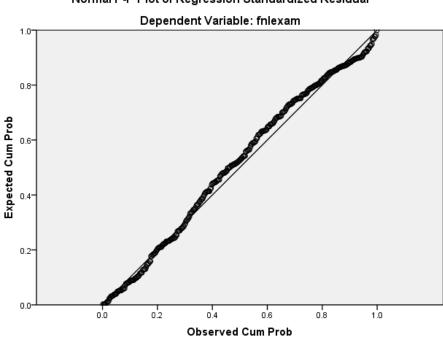
Residuals Statistics <sup>a</sup>							
	Minimum	Maximum	Mean	Std. Deviation	Ν		
Predicted Value	55.38	87.35	68.84	4.970	394		
Std. Predicted Value	-2.708	3.724	.000	1.000	394		
Standard Error of Predicted	.705	3.262	1.137	.368	394		
Value							
Adjusted Predicted Value	55.27	87.01	68.84	4.970	394		
Residual	-38.963	42.697	.000	13.655	394		
Std. Residual	-2.846	3.119	.000	.997	394		
Stud. Residual	-2.855	3.169	.000	1.002	394		
Deleted Residual	-39.218	44.087	.001	13.767	394		
Stud. Deleted Residual	-2.882	3.207	.000	1.004	394		
<mark>Mahal. Distance</mark>	<mark>.045</mark>	<mark>21.321</mark>	<mark>1.995</mark>	<mark>2.374</mark>	<mark>394</mark>		
Cook's Distance	.000	<mark>.109</mark>	.003	<mark>.008</mark>	<mark>394</mark>		
Centered Leverage Value	.000	.054	.005	.006	394		

A maximum Mahalanobis distance 21.321 greater than the critical chisquare value of 13.816 for df = 2 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .109 (which is less than one) means outliers should not be a concern.

### Charts



# The points clustered fairly close to the line indicate that the residuals are normally distributed.



Normal P-P Plot of Regression Standardized Residual

# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.

Scatterplot Dependent Variable: fnlexam 0 Regression Standardized Residual 8 2 0 0 0 0 0 8 0-0 8 0 0 -2 0 0 00 -2 2 4 ó

Regression Standardized Predicted Value

#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

## Regression

	Notes	
Output Created		20-Oct-2011 14:37:44
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav

	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.232
	Elapsed Time	00:00:01.932
	Memory Required	2860 bytes
	Additional Memory Required	896 bytes
	for Residual Plots	000 5/103
Variables Created or	MAH_2	Mahalanobis Distance
Modified	_ COO_2	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics							
	Mean	Std. Deviation	N				
fnlexam	68.89	14.550	391				
pretest	47.11	14.392	391				
act	16.87	1.143	391				
ascgr	.87	.340	391				

## N = 391 with k = 3 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations						
					_	
		fnlexam	pretest	act	Ascgr	
Pearson Correlation	fnlexam	1.000	.306	.217	.163	
	pretest	.306	1.000	.189	.031	
	Act	.217	.189	1.000	.029	
	Ascgr	.163	.031	.029	1.000	
Sig. (1-tailed)	fnlexam		.000	.000	.001	
	pretest	.000	-	.000	.269	
	Act	.000	.000		.286	
	Ascgr	.001	.269	.286		
Ν	fnlexam	391	391	391	391	
	pretest	391	391	391	391	
	Act	391	391	391	391	
	Ascgr	391	391	391	391	

### Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	ascgr, act, pretest <sup>a</sup>		Enter

a. All requested variables entered.

	Model Summary <sup>b</sup>								
Model					Char	nge Statistics			
	_		Adjusted R	Std. Error of the	R Square				
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1		
1	.377 <sup>a</sup>	<mark>.142</mark>	.136	13.528	.142	21.389	3		

a. Predictors: (Constant), ascgr, act, pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang	ge Statistics					
	df2	Sig. F Change	Durbin-Watson				
1	387	.000	1.688				

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
М	lodel	Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	11742.375	3	3914.125	<mark>21.389</mark>	.000 <sup>a</sup>		
	Residual	70818.674	387	182.994				
	Total	82561.049	390					

a. Predictors: (Constant), ascgr, act, pretest

b. Dependent Variable: fnlexam

_			Coefficients			
Mode	el	Unstandardize	ed Coefficients	Standardized Coefficients		
		B	Std. Error	Beta	t	Sig.
1	(Constant)	<mark>15.716</mark>	10.242		1.534	<mark>.126</mark>
	pretest	<mark>.273</mark>	.048	.270	5.639	.000
	act	<mark>2.059</mark>	.610	.162	3.373	<mark>.001</mark>
	ascgr	<mark>6.407</mark>	2.016	.150	3.178	.002

	Coefficients <sup>a</sup>							
Model		95.0% Confiden	ce Interval for B	(	Correlations			
		Lower Bound	Lower Bound Upper Bound Zero-orde		Partial	Part		
1	(Constant)	-4.421	35.853					
	pretest	.178	.369	.306	.276	.265		
	act	.859	3.259	.217	.169	.159		
	ascgr	2.443	10.371	.163	.159	.150		

Coefficients <sup>a</sup>							
Model		Collinearity Statistics					
		Tolerance VIF					
1	(Constant)						
	pretest		.96	3		<mark>1.038</mark>	
	act		<mark>.96</mark>	4		<mark>1.038</mark>	
	ascgr		.99	8		<mark>1.002</mark>	

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations <sup>®</sup>							
Model			ascgr	act	pretest			
1	Correlations	ascgr	1.000	023	026			
		act	023	1.000	189			
		pretest	026	189	1.000			
	Covariances	ascgr	4.065	029	003			
		act	029	.373	006			
		pretest	003	006	.002			

Coefficient Correlations<sup>a</sup>

Model	Dimension				Variance Pr	oportions	
		Eigenvalue	Condition Index	(Constant)	pretest	act	ascgr
1	1	3.834	1.000	.00	.01	.00	.01
	2	.112	5.842	.00	.19	.00	.82
	- 3	.051	8.664	.02	.80	.02	.16
	4	.002	41.206	.98	.01	.98	.00

**Collinearity Diagnostics**<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.071	100	58.46	41.543

a. Dependent Variable: fnlexam

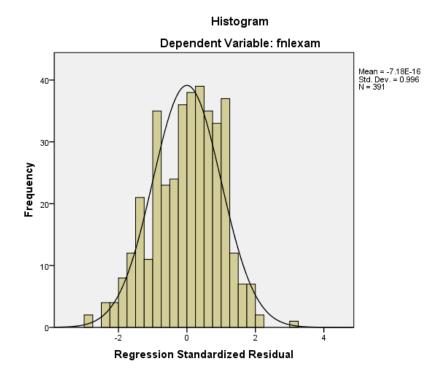
Residuals Statistics <sup>a</sup>						
	Minimum	Maximum	Mean	Std. Deviation	Ν	
Predicted Value	51.36	87.92	68.89	5.487	391	
Std. Predicted Value	-3.195	3.469	.000	1.000	391	
Standard Error of Predicted	.747	3.439	1.284	.473	391	
Value						
Adjusted Predicted Value	51.45	87.61	68.89	5.483	391	
Residual	-39.894	41.543	.000	13.475	391	
Std. Residual	-2.949	3.071	.000	.996	391	
Stud. Residual	-2.959	3.123	.000	1.002	391	
Deleted Residual	-40.171	42.948	.002	13.632	391	
Stud. Deleted Residual	-2.990	3.159	.000	1.004	391	
<mark>Mahal. Distance</mark>	<mark>.193</mark>	<mark>24.212</mark>	<mark>2.992</mark>	<mark>3.289</mark>	<mark>391</mark>	
Cook's Distance	<mark>.000</mark>	.082	.003	.007	<mark>391</mark>	
Centered Leverage Value	.000	.062	.008	.008	391	

Residuals Statistics<sup>a</sup>

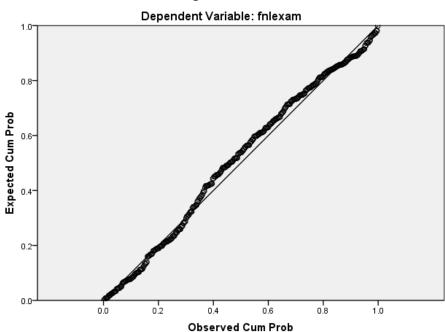
a. Dependent Variable: fnlexam

A maximum Mahalanobis distance 24.212 greater than the critical chisquare value of 16.266 for df = 3 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .082 (which is less than one) means outliers should not be a concern.

# Charts

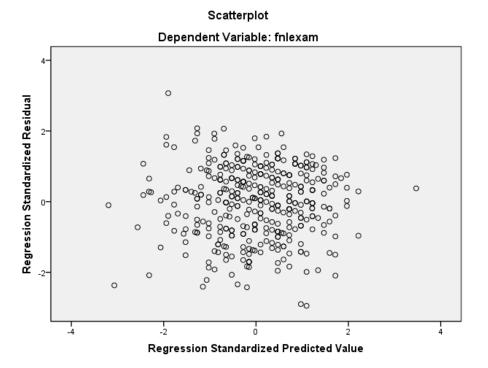


# The points clustering farily close to the line indicate that the residuals are normally distributed.



Normal P-P Plot of Regression Standardized Residual

The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N

/MISSING LISTWISE

/STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP

/CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr gender /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:43:43
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav

	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	013
Missing Value Handling	Definition of Missing	User-defined missing values are
	Deminion of Missing	treated as missing.
	Cases Used	Statistics are based on cases with no
	00303 0300	missing values for any variable used.
Syntax		REGRESSION
Syntax		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		gender
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
December		
Resources	Processor Time	00:00:01.903
	Elapsed Time	00:00:01.889
	Memory Required	3228 bytes
	Additional Memory Required	888 bytes
	for Residual Plots	
Variables Created or	MAH_3	Mahalanobis Distance
Modified	<u>COO_3</u>	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics					
	Mean	Std. Deviation	N		
fnlexam	68.99	14.508	383		
pretest	47.02	14.486	383		
act	16.86	1.152	383		
ascgr	.87	.340	383		
gender	.38	.486	383		

## N = 383 with k = 4 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations						
		fnlexam	pretest	act	ascgr	gender
Pearson Correlation	fnlexam	1.000	.318	.222	.169	173
	pretest	.318	1.000	.191	.023	119
	act	.222	.191	1.000	.034	015
	ascgr	.169	.023	.034	1.000	104
	gender	173	119	015	104	1.000
Sig. (1-tailed)	fnlexam		.000	.000	.000	.000
	pretest	.000	-	.000	.327	.010
	act	.000	.000		.254	.386
	ascgr	.000	.327	.254	-	.021
	gender	.000	.010	.386	.021	
Ν	fnlexam	383	383	383	383	383
	pretest	383	383	383	383	383
	act	383	383	383	383	383
	ascgr	383	383	383	383	383
	gender	383	383	383	383	383

Correlations

#### Variables Entered/Removed<sup>b</sup>

Model	Variables Entered	Variables Removed	Method
1	gender, act, ascgr, pretest <sup>a</sup>		Enter

a. All requested variables entered.

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
	_		Adjusted R	Std. Error of the	R Square		
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1
1	.409 <sup>a</sup>	<mark>.168</mark>	.159	13.307	.168	19.029	4

a. Predictors: (Constant), gender, act, ascgr, pretest

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model	Chang			
	df2	Sig. F Change	Durbin-Watson	
1	378	.000	1.697	

b. Dependent Variable: fnlexam

#### ANOVA<sup>b</sup>

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	13477.658	4	3369.415	<mark>19.029</mark>	.000 <sup>a</sup>
	Residual	66930.300	378	177.064		
	Total	80407.958	382			

a. Predictors: (Constant), gender, act, ascgr, pretest

	Coefficients <sup>a</sup>						
Model		Unstandardize	ed Coefficients	Standardized Coefficients			
		B	Std. Error	Beta	t	Sig.	
1	(Constant)	<mark>17.532</mark>	10.129		1.731	<mark>.084</mark>	
	pretest	<mark>.269</mark>	.048	.269	5.579	.000	
	act	<mark>2.068</mark>	.602	.164	3.432	<mark>.001</mark>	
	ascgr	<mark>6.158</mark>	2.013	.144	3.059	<mark>.002</mark>	
	gender	<mark>-3.678</mark>	1.417	123	-2.595	<mark>.010</mark>	

<b>Coefficients</b> <sup>a</sup>
----------------------------------

Model		95.0% Confidence Interval for B		Correlations		
		Lower Bound	Upper Bound	Zero-order	Partial	Part
1	(Constant)	-2.384	37.447			
	pretest	.174	.364	.318	.276	.262
	act	.883	3.252	.222	.174	.161
	ascgr	2.199	10.116	.169	.155	.144
	gender	-6.465	891	173	132	122

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>					
Model		Collinearity Statistics			
		Tolerance			
1	(Constant)				
	pretest	<mark>.950</mark>	<mark>1.053</mark>		
	act	<mark>.962</mark>	<mark>1.039</mark>		
	ascgr	<mark>.988</mark>	<mark>1.012</mark>		
	gender	.975	<mark>1.025</mark>		

a. Dependent Variable: fnlexam

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations					
Model		gender	act	ascgr	pretest	
1	Correlations	Gender	1.000	011	.102	.117
		Act	011	1.000	031	191
		Ascgr	.102	031	1.000	005
		Pretest	.117	191	005	1.000
	Covariances	Gender	2.009	010	.291	.008
		Act	010	.363	038	006
		Ascgr	.291	038	4.053	.000
		Pretest	.008	006	.000	.002

Coefficient Correlations<sup>a</sup>

Model	Dimension			Variance Proportions		ons
		Eigenvalue	Condition Index	(Constant)	pretest	act
1	1	4.254	1.000	.00	.00	.00
	2	.582	2.703	.00	.01	.00
	_ 3	.113	6.142	.00	.21	.00
	4	.048	9.380	.02	.77	.02
	5	.002	43.075	.98	.01	.98

**Collinearity Diagnostics**<sup>a</sup>

Commeanly Diagnostics					
Model	Dimension	Variance Proportions			
		Ascgr	gender		
1	1	.01	.02		
	2	.01	.91		
	3	.78	.01		
	4	.20	.06		
	5	.00	.00		

#### **Collinearity Diagnostics**<sup>a</sup>

a. Dependent Variable: fnlexam

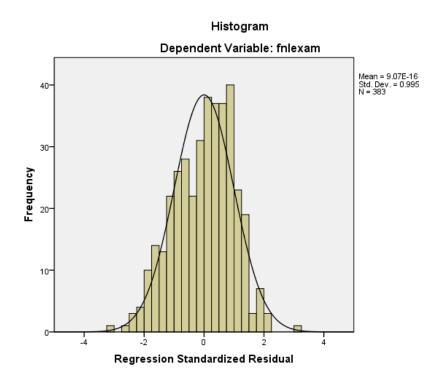
#### **Casewise Diagnostics**<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.007	100	59.99	40.013
<sup></sup> 582	-3.059	35	75.71	-40.706

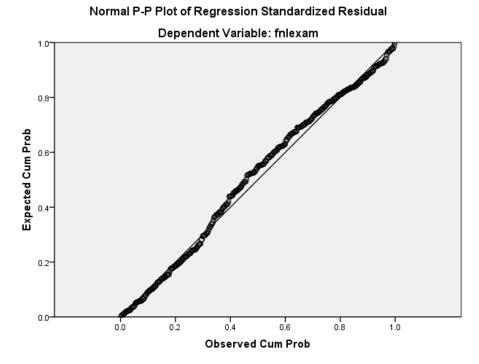
Residuals Statistics <sup>a</sup>					
	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	49.43	85.68	68.99	5.940	383
Std. Predicted Value	-3.293	2.810	.000	1.000	383
Standard Error of Predicted	.895	3.457	1.457	.436	383
Value					
Adjusted Predicted Value	49.39	85.18	68.99	5.937	383
Residual	-40.706	40.013	.000	13.237	383
Std. Residual	-3.059	3.007	.000	.995	383
Stud. Residual	-3.070	3.060	.000	1.002	383
Deleted Residual	-40.998	41.447	.001	13.432	383
Stud. Deleted Residual	-3.105	3.095	.000	1.004	383
<mark>Mahal. Distance</mark>	<mark>.730</mark>	<mark>24.790</mark>	<mark>3.990</mark>	<mark>3.337</mark>	<mark>383</mark>
Cook's Distance	<mark>.000</mark>	<mark>.073</mark>	.003	.007	<mark>383</mark>
Centered Leverage Value	.002	.065	.010	.009	383

A maximum Mahalanobis distance 24.790 greater than the critical chisquare value of 18.467 for df = 4 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .073 (which is less than one) means outliers should not be a concern.

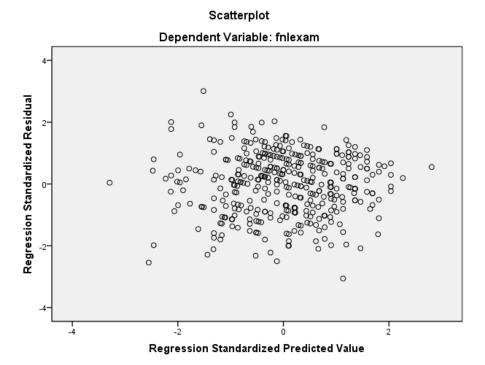
### Charts







The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



#### REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N

/MISSING LISTWISE

/STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP

/CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr gender mozartuse /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:51:37
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav

	Active Dataset	DataSet1
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	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	013
Missing Value Handling	Definition of Missing	User-defined missing values are
initiality for the first state of the state	2 cm licer of meening	treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
Official		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		gender mozartuse
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.279
	Elapsed Time	00:00:01.827
	Memory Required	3628 bytes
	Additional Memory Required	880 bytes
	for Residual Plots	
Variables Created or	MAH_4	Mahalanobis Distance
Modified	COO 4	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics						
	Mean	Std. Deviation	N			
fnlexam	68.99	14.508	383			
pretest	47.02	14.486	383			
act	16.86	1.152	383			
ascgr	.87	.340	383			
gender	.38	.486	383			
mozartuse	.12	.322	383			

### N = 383 with k = 5 predictors N > 50 + 8k Therefore, the sample size is appropriate.

Correlations							
		fnlexam	pretest	act	ascgr	gender	mozartuse
Pearson Correlation	fnlexam	1.000	.318	.222	.169	173	.117
Pearson Conelation							
	pretest	.318	1.000	.191	.023	119	.016
	act	.222	.191	1.000	.034	015	020
	ascgr	.169	.023	.034	1.000	104	.024
	gender	173	119	015	104	1.000	036
	mozartuse	.117	.016	020	.024	036	1.000
Sig. (1-tailed)	fnlexam		.000	.000	.000	.000	.011
	pretest	.000		.000	.327	.010	.376
	act	.000	.000		.254	.386	.346
	ascgr	.000	.327	.254		.021	.322
	gender	.000	.010	.386	.021		.241
	mozartuse	.011	.376	.346	.322	.241	
Ν	fnlexam	383	383	383	383	383	383
	pretest	383	383	383	383	383	383
	act	383	383	383	383	383	383
	ascgr	383	383	383	383	383	383
	gender	383	383	383	383	383	383
	mozartuse	383	383	383	383	383	383

	Variables Entered/Removed <sup>b</sup>							
Model	Variables	Variables						
	Entered	Removed	Method					
1	mozartuse, pretest, ascgr, gender, act <sup>a</sup>		Enter					

a. All requested variables entered.

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model					Char	nge Statistics	
	B	R Square	Adjusted R Square	Std. Error of the Estimate	R Square Change	F Change	df1
		R Square	Square	LStimate	Change	r Change	un
1	.423 <sup>a</sup>	<mark>.179</mark>	.168	13.231	.179	16.469	5

a. Predictors: (Constant), mozartuse, pretest, ascgr, gender, act

b. Dependent Variable: fnlexam

#### Model Summary<sup>b</sup>

Model	Chang			
	df2	Sig. F Change	Durbin-Watson	
1	377	.000	1.721	

b. Dependent Variable: fnlexam

Model		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	14414.268	5	2882.854	<mark>16.469</mark>	.000 <sup>a</sup>
	Residual	65993.690	377	175.050		
	Total	80407.958	382			

a. Predictors: (Constant), mozartuse, pretest, ascgr, gender, act

	Coefficients <sup>a</sup>						
Model				Standardized			
		Unstandardize	ed Coefficients	Coefficients			
		B	Std. Error	Beta	t	Sig.	
1	(Constant)	<mark>16.521</mark>	10.080		1.639	<mark>.102</mark>	
	pretest	<mark>.267</mark>	.048	.267	5.573	.000	
	act	<mark>2.101</mark>	.599	.167	3.507	<mark>.001</mark>	
	ascgr	<mark>6.062</mark>	2.002	.142	3.028	<mark>.003</mark>	
	gender	- <mark>3.574</mark>	1.410	120	-2.535	<mark>.012</mark>	
	<mark>mozartuse</mark>	<mark>4.862</mark>	2.102	.108	2.313	.021	

	Coefficients <sup>a</sup>							
Model		95.0% Confiden	ce Interval for B	Correlations				
		Lower Bound	Upper Bound	Zero-order	Partial	Part		
1	(Constant)	-3.300	36.342					
	pretest	.173	.362	.318	.276	.260		
	act	.923	3.280	.222	.178	.164		
	ascgr	2.125	9.999	.169	.154	.141		
	gender	-6.347	802	173	129	118		
	mozartuse	.729	8.995	.117	.118	.108		

a. Dependent Variable: fnlexam

Coefficients <sup>a</sup>							
Model		Collin	earity	Statis	stics		
		Tolera	Tolerance		<mark>/IF</mark>		
1	(Constant)						
	pretest		<mark>.950</mark>		<mark>1.053</mark>		
	act		<mark>.962</mark>		<mark>1.040</mark>		
	ascgr		<mark>.988</mark>		<mark>1.012</mark>		
	gender		<mark>.974</mark>		<mark>1.026</mark>		
	mozartuse		<mark>.998</mark>		<mark>1.002</mark>		

a. Dependent Variable: fnlexam

Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

Model			mozartuse	pretest	ascgr	gender	Act
1	Correlations	Mozartuse	1.000	016	021	.032	.024
		Pretest	016	1.000	004	.117	191
		Ascgr	021	004	1.000	.101	032
		Gender	.032	.117	.101	1.000	010
		Act	.024	191	032	010	1.000
	Covariances	Mozartuse	4.419	002	087	.094	.031
		Pretest	002	.002	.000	.008	005
		Ascgr	087	.000	4.009	.286	038
		Gender	.094	.008	.286	1.988	009
		Act	.031	005	038	009	.359

#### **Coefficient Correlations**<sup>a</sup>

a. Dependent Variable: fnlexam

#### Model Dimension Variance Proportions Eigenvalue **Condition Index** (Constant) pretest act ascgr 1 .00 .01 1 4.396 1.000 .00 .00 2 2.251 .00 .00 .00 .867 .00 3 .573 2.770 .00 .01 .00 .01 4 .113 6.244 .00 .21 .00 .78 5 .048 9.537 .02 .77 .02 .20 6 .002 43.814 .98 .01 .98 .00

#### **Collinearity Diagnostics**<sup>a</sup>

Collinearity Diagnostics <sup>a</sup>							
Model	Dimension	Variance	Proportions				
		gender	mozartuse				
1	1	.01	.01				
	2	.04	.94				
	3	.87	.05				
	- 4	.01	.00				
	5	.06	.00				
	6	.00	.00				

#### Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual			
192	3.080	100	59.25	40.745			
<sup></sup> 582	-3.031	35	75.11	-40.106			

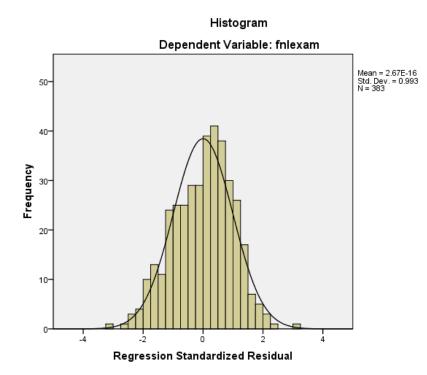
a. Dependent Variable: fnlexam

	Minimum	Maximum	Mean	Std. Deviation	N
Predicted Value	48.85	85.31	68.99	6.143	383
Std. Predicted Value	-3.278	2.657	.000	1.000	383
Standard Error of Predicted	.930	3.447	1.580	.496	383
Value					
Adjusted Predicted Value	48.77	85.50	68.99	6.138	383
Residual	-40.106	40.745	.000	13.144	383
Std. Residual	-3.031	3.080	.000	.993	383
Stud. Residual	-3.043	3.135	.000	1.002	383
Deleted Residual	-40.409	42.230	.002	13.362	383
Stud. Deleted Residual	-3.077	3.173	.000	1.004	383
<mark>Mahal. Distance</mark>	.889	<mark>24.925</mark>	<mark>4.987</mark>	<mark>3.993</mark>	<mark>383</mark>
Cook's Distance	.000	.060	.003	.006	<mark>383</mark>
Centered Leverage Value	.002	.065	.013	.010	383

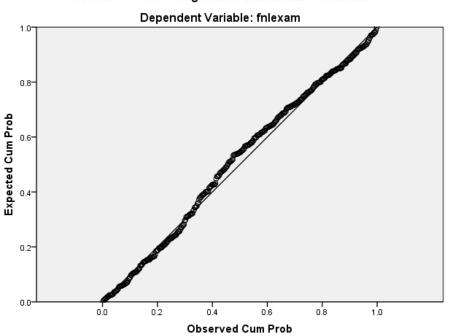
#### **Residuals Statistics**<sup>a</sup>

A maximum Mahalanobis distance 24.925 greater than the critical chisquare value of 20.515 for df = 5 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .060 (which is less than one) means outliers should not be a concern.

#### Charts

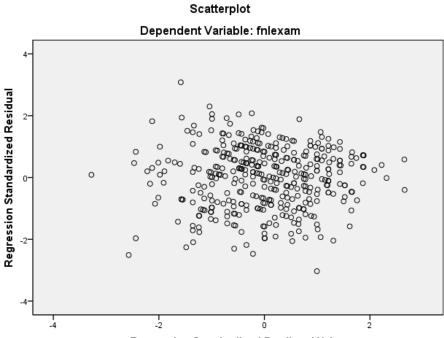


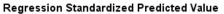
# The points clustering fairly close to the line indicate that the residuals are normally distribute



Normal P-P Plot of Regression Standardized Residual

# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.





REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr gender mozartuse sat /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:57:16
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

555

Syntax		REGRESSION
,		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		gender mozartuse sat
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.373
	Elapsed Time	00:00:01.765
	Memory Required	4060 bytes
	Additional Memory Required	872 bytes
	for Residual Plots	
Variables Created or	MAH_5	Mahalanobis Distance
Modified	COO_5	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics						
	Mean Old Deviation					
	Mean	Std. Deviation	N			
fnlexam	69.93	16.955	27			
pretest	50.56	16.428	27			
Act	17.07	1.708	27			
ascgr	.96	.192	27			
gender	.26	.447	27			
mozartuse	.11	.320	27			
Sat	402.96	54.900	27			

# N = 27 with k = 6 N is not greater than 50 + 8k. Therefore, the sample size is not appropriate.

Correlations						
		fnlexam	pretest	act	ascgr	gender
Pearson Correlation	Fnlexam	1.000	.699	.622	.235	307
	Pretest	.699	1.000	.451	.128	335
	Act	.622	.451	1.000	.594	127
	Ascgr	.235	.128	.594	1.000	331
	Gender	307	335	127	331	1.000
	Mozartuse	.306	.317	.055	.069	209
	Sat	.492	.538	.449	.229	080
Sig. (1-tailed)	Fnlexam		.000	.000	.119	.060
	Pretest	.000		.009	.262	.044
	Act	.000	.009		.001	.264
	Ascgr	.119	.262	.001		.046
	Gender	.060	.044	.264	.046	
	Mozartuse	.060	.054	.393	.366	.148
	Sat	.005	.002	.009	.125	.347
Ν	Fnlexam	27	27	27	27	27
	Pretest	27	27	27	27	27
	Act	27	27	27	27	27
	Ascgr	27	27	27	27	27
	Gender	27	27	27	27	27
	Mozartuse	27	27	27	27	27

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Correlations							
		fnlexam	pretest	act	ascgr	gender	
Pearson Correlation	Fnlexam	1.000	.699	.622	.235	307	
	Pretest	.699	1.000	.451	.128	335	
	Act	.622	.451	1.000	.594	127	
	Ascgr	.235	.128	.594	1.000	331	
	Gender	307	335	127	331	1.000	
	Mozartuse	.306	.317	.055	.069	209	
	Sat	.492	.538	.449	.229	080	
Sig. (1-tailed)	Fnlexam		.000	.000	.119	.060	
	Pretest	.000		.009	.262	.044	
	Act	.000	.009		.001	.264	
	Ascgr	.119	.262	.001		.046	
	Gender	.060	.044	.264	.046		
	Mozartuse	.060	.054	.393	.366	.148	
	Sat	.005	.002	.009	.125	.347	
Ν	Fnlexam	27	27	27	27	27	
	Pretest	27	27	27	27	27	
	Act	27	27	27	27	27	
	Ascgr	27	27	27	27	27	
	Gender	27	27	27	27	27	
	Mozartuse	27	27	27	27	27	
	Sat	27	27	27	27	27	

Correlations						
		mozartuse	sat			
Pearson Correlation	Fnlexam	.306	.492			
	Pretest	.317	.538			
	Act	.055	.449			
	Ascgr	.069	.229			
	Gender	209	080			
	Mozartuse	1.000	.199			
	Sat	.199	1.000			
Sig. (1-tailed)	Fnlexam	.060	.005			
	Pretest	.054	.002			
	Act	.393	.009			
	Ascgr	.366	.125			
	Gender	.148	.347			
	Mozartuse		.159			
	Sat	.159				
Ν	Fnlexam	27	27			
	Pretest	27	27			
	Act	27	27			
	Ascgr	27	27			
	Gender	27	27			
	Mozartuse	27	27			
	Sat	27	27			

# Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	sat, gender,		Enter
	mozartuse,		
	ascgr, pretest,		
	act <sup>a</sup>		

a. All requested variables entered.

	Model Summary <sup>b</sup>							
Model					Char	nge Statistics		
	_		Adjusted R	Std. Error of the	R Square			
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1	
1	.806 <sup>a</sup>	.649	.544	11.450	.649	6.167	6	

a. Predictors: (Constant), sat, gender, mozartuse, ascgr, pretest, act

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>							
Model	Chang						
	df2	Sig. F Change	Durbin-Watson				
1	20	.001	2.205				

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	4851.618	6	808.603	<mark>6.167</mark>	.001 <sup>a</sup>		
	Residual	2622.234	20	131.112				
	Total	7473.852	26					

a. Predictors: (Constant), sat, gender, mozartuse, ascgr, pretest, act

	Coefficients <sup>a</sup>						
Model		Unstandardized Coefficients		Standardized Coefficients			
		B	Std. Error	Beta	t	<mark>Sig.</mark>	
1	(Constant)	<mark>-28.150</mark>	25.050		-1.124	<mark>.274</mark>	
	pretest	<mark>.371</mark>	.195	.359	1.898	<mark>.072</mark>	
	act	<mark>5.113</mark>	1.973	.515	2.591	.017	
	ascgr	<mark>-17.052</mark>	16.316	194	-1.045	<mark>.308</mark>	
	gender	<mark>-5.784</mark>	5.940	152	974	<mark>.342</mark>	
	mozartuse	<mark>6.935</mark>	7.526	.131	.921	<mark>.368</mark>	
	<mark>sat</mark>	.023	.051	.074	.445	<mark>.661</mark>	

	Coefficients <sup>a</sup>							
Model		95.0% Confiden	ce Interval for B	(	Correlations			
		Lower Bound	Upper Bound	Zero-order	Partial	Part		
1	(Constant)	-80.403	24.104					
	pretest	037	.779	.699	.391	.251		
	act	.997	9.228	.622	.501	.343		
	ascgr	-51.087	16.983	.235	228	138		
	gender	-18.175	6.607	307	213	129		
	mozartuse	-8.764	22.635	.306	.202	.122		
	sat	084	.130	.492	.099	.059		

Coefficients <sup>a</sup>						
Model		Collinearity Statistics				
		Tolerance	VIF			
1	(Constant)					
	pretest	<mark>.489</mark>	<mark>2.044</mark>			
	act	<mark>.444</mark>	<mark>2.252</mark>			
	ascgr	<mark>.511</mark>	<mark>1.955</mark>			
	gender	<mark>.717</mark>	<mark>1.395</mark>			
	mozartuse	.868	<mark>1.152</mark>			
	sat	.638	<mark>1.567</mark>			

Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

Model			sat		mozartuse	ascar
wouer			ક્તા	gender	mozantuse	ascgr
1	Correlations	Sat	1.000	156	076	087
		Gender	156	1.000	.082	.409
		Mozartuse	076	.082	1.000	072
		Ascgr	087	.409	072	1.000
		Pretest	407	.393	236	.340
		Act	164	240	.139	623
	Covariances	Sat	.003	047	029	072
		Gender	047	35.285	3.671	39.625
		Mozartuse	029	3.671	56.646	-8.848
		Ascgr	072	39.625	-8.848	266.215
		Pretest	004	.456	347	1.086
		Act	017	-2.810	2.060	-20.064

Coefficient Correlations<sup>a</sup>

	Coefficient Correlations <sup>a</sup>					
Model			pretest	act		
1	Correlations	Sat	407	164		
		Gender	.393	240		
		Mozartuse	236	.139		
		Ascgr	.340	623		
		Pretest	1.000	417		
		Act	417	1.000		
	Covariances	Sat	004	017		
		Gender	.456	-2.810		
		Mozartuse	347	2.060		
		Ascgr	1.086	-20.064		
		Pretest	.038	161		
		Act	161	3.893		

Model	Dimension				Variance Pr	oportions	
		Eigenvalue	Condition Index	(Constant)	pretest	act	ascgr
1	1	5.302	1.000	.00	.00	.00	.00
	2	1.011	2.290	.00	.00	.00	.00
	3	.601	2.970	.00	.00	.00	.00
	_ 4	.059	9.474	.01	.53	.00	.08
	5	.017	17.830	.12	.21	.00	.54
	6	.008	25.844	.26	.07	.05	.04
	7	.003	41.624	.62	.20	.95	.34

Collinearity Diagnostics<sup>a</sup>

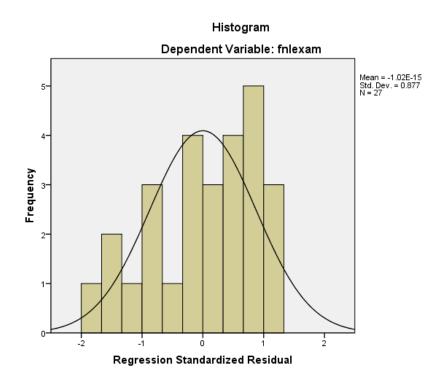
	Collinearity Diagnostics"					
Model	Dimension	Variance Proportions				
		Gender	mozartuse	sat		
1	1	.01	.00	.00		
	2	.19	.48	.00		
	3	.50	.44	.00		
	4	.03	.06	.00		
	5	.22	.00	.12		
	6	.00	.01	.88		
	7	.05	.01	.00		

Collinearity Diagnostics<sup>a</sup>

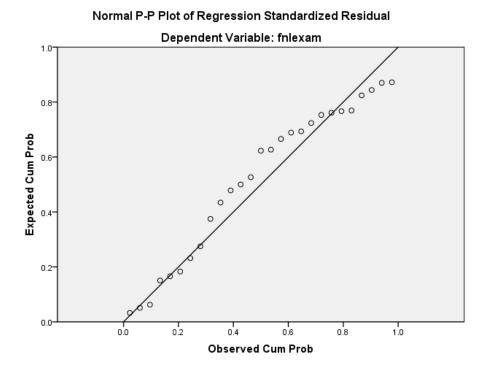
Residuals Statistics <sup>a</sup>					
	Minimum	Maximum	Mean	Std. Deviation	N
Predicted Value	47.57	101.38	69.93	13.660	27
Std. Predicted Value	-1.637	2.303	.000	1.000	27
Standard Error of Predicted	3.102	11.450	5.506	1.955	27
Value					
Adjusted Predicted Value	45.55	115.97	71.13	15.302	26
Residual	-21.067	13.015	.000	10.043	27
Std. Residual	-1.840	1.137	.000	.877	27
Stud. Residual	-1.984	1.310	014	1.025	26
Deleted Residual	-25.579	19.450	437	13.765	26
Stud. Deleted Residual	-2.157	1.335	032	1.059	26
<mark>Mahal. Distance</mark>	<mark>.945</mark>	<mark>25.037</mark>	<mark>5.778</mark>	<mark>5.233</mark>	<mark>27</mark>
Cook's Distance	<mark>.000</mark>	<mark>.365</mark>	<mark>.054</mark>	.085	<mark>26</mark>
Centered Leverage Value	.036	.963	.222	.201	27

A maximum Mahalanobis distance 25.037 greater than the critical chisquare value of 22.458 for df = 6 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .365 (which is less than one) means outliers should not be a concern.

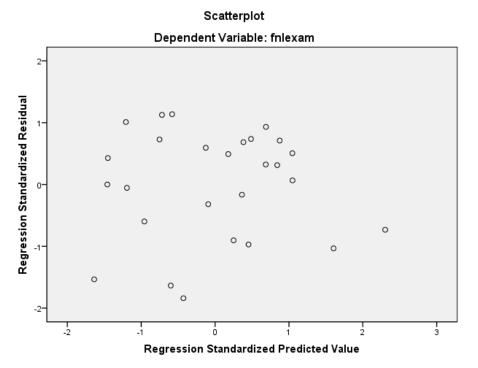
# Charts



# The points clustering fairly close to the line indicate that the residuals are normally distributed.



# The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



REGRESSION

/DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN

/NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr gender mozartuse techsex /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:58:14
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.
Syntax		REGRESSION
		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		gender mozartuse techsex
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.389
	Elapsed Time	00:00:01.821
	Memory Required	4100 bytes
	······	•

		Additional Memory Required		872 bytes
		for Residual Plots		
Va	ariables Created or	MAH_6	Mahalanobis Distance	
Mo	odified	COO_6	Cook's Distance	

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics						
	Mean	Std. Deviation	N			
Fnlexam	68.99	14.508	383			
Pretest	47.02	14.486	383			
Act	16.86	1.152	383			
Ascgr	.87	.340	383			
Gender	.38	.486	383			
Mozartuse	.12	.322	383			
Techsex	.25	.431	383			

# N = 383 with k = 6 predictors N > 50 + 8k

Therefore, the sample size is appropriate.

Correlations						
		fnlexam	pretest	act	ascgr	gender
Pearson Correlation	fnlexam	1.000	.318	.222	.169	173
	pretest	.318	1.000	.191	.023	119
	act	.222	.191	1.000	.034	015
	ascgr	.169	.023	.034	1.000	104
	gender	173	119	015	104	1.000
	mozartuse	.117	.016	020	.024	036
	techsex	075	059	.004	009	060
Sig. (1-tailed)	fnlexam		.000	.000	.000	.000
	pretest	.000		.000	.327	.010
	act	.000	.000		.254	.386
	ascgr	.000	.327	.254		.021
	gender	.000	.010	.386	.021	
	mozartuse	.011	.376	.346	.322	.241
	techsex	.070	.125	.469	.433	.119
Ν	fnlexam	383	383	383	383	383
	pretest	383	383	383	383	383
	act	383	383	383	383	383
	ascgr	383	383	383	383	383
	gender	383	383	383	383	383
	mozartuse	383	383	383	383	383
	techsex	383	383	383	383	383

Correlations						
		mozartuse	techsex			
Pearson Correlation	fnlexam	.117	075			
	pretest	.016	059			
	act	020	.004			
	ascgr	.024	009			
	gender	036	060			
	mozartuse	1.000	208			
	techsex	208	1.000			
Sig. (1-tailed)	fnlexam	.011	.070			
	pretest	.376	.125			
	act	.346	.469			
	ascgr	.322	.433			
	gender	.241	.119			
	mozartuse		.000			
	techsex	.000	-			
Ν	fnlexam	383	383			
	pretest	383	383			
	act	383	383			
	ascgr	383	383			
	gender	383	383			
	mozartuse	383	383			
	techsex	383	383			

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	techsex, act,		Enter
	ascgr, gender,		
	mozartuse,		
	pretest <sup>a</sup>		

a. All requested variables entered.

Model Summary <sup>b</sup>								
Model					Change Statistics			
	_		Adjusted R	Std. Error of the	R Square			
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1	
1	.426 <sup>a</sup>	<mark>.181</mark>	.168	13.232	.181	13.877	6	

a. Predictors: (Constant), techsex, act, ascgr, gender, mozartuse, pretest

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>						
Model	Chang	ge Statistics				
	df2	Sig. F Change	Durbin-Watson			
1	376	.000	1.727			

b. Dependent Variable: fnlexam

	ANOVA <sup>b</sup>							
Model		Sum of Squares	df	Mean Square	F	Sig.		
1	Regression	14577.638	6	2429.606	<mark>13.877</mark>	.000 <sup>a</sup>		
	Residual	65830.320	376	175.081				
	Total	80407.958	382					

a. Predictors: (Constant), techsex, act, ascgr, gender, mozartuse, pretest

	Coefficients <sup>a</sup>						
Mod	el						
		Unstandardize	ed Coefficients	Coefficients			
		B	Std. Error	Beta	t	Sig.	
1	(Constant)	<mark>17.043</mark>	10.096		1.688	<mark>.092</mark>	
	pretest	<mark>.264</mark>	.048	.264	5.496	.000	
	act	<mark>2.108</mark>	.599	.167	3.518	<mark>.000</mark> .	
	ascgr	<mark>6.041</mark>	2.002	.142	3.017	.003	
	gender	<mark>-3.680</mark>	1.414	123	-2.602	<mark>.010</mark>	
	mozartuse	<mark>4.426</mark>	2.150	.098	2.059	<mark>.040</mark>	
	techsex	<mark>-1.559</mark>	1.614	046	966	<mark>.335</mark>	

### **Coefficients**<sup>a</sup> Model 95.0% Confidence Interval for B Correlations Lower Bound Upper Bound Zero-order Partial Part 1 (Constant) -2.808 36.894 pretest .170 .359 .318 .273 .256 .930 3.287 .222 act .179 .164 2.104 9.979 .169 .141 ascgr .154 -6.462 -.899 -.173 gender -.133 -.121 mozartuse .198 8.654 .096 .117 .106 techsex -4.732 1.614 -.075 -.050 -.045

Coefficients <sup>a</sup>						
Model		Collinearity	Statistics			
		Tolerance	VIF			
1	(Constant)					
		_				
	pretest	<mark>.945</mark>	<mark>1.058</mark>			
	act	<mark>.962</mark>	<mark>1.040</mark>			
	ascgr	.988	<mark>1.013</mark>			
	gender	.969	<mark>1.032</mark>			
	mozartuse	<mark>.954</mark>	<mark>1.049</mark>			
	techsex	<mark>.948</mark>	<mark>1.055</mark>			

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations <sup>®</sup>							
Model			techsex	act	ascgr	gender		
1	Correlations	techsex	1.000	012	.011	.078		
		act	012	1.000	032	011		
		ascgr	.011	032	1.000	.102		
		gender	.078	011	.102	1.000		
		mozartuse	.210	.021	018	.047		
		pretest	.066	191	004	.121		
	Covariances	techsex	2.604	011	.035	.177		
		act	011	.359	038	010		
		ascgr	.035	038	4.010	.289		
		gender	.177	010	.289	2.001		
		mozartuse	.728	.027	077	.144		
		pretest	.005	006	.000	.008		

Coefficient Correlations<sup>a</sup>

Coefficient Correlations"						
Model			mozartuse	pretest		
1	Correlations	techsex	.210	.066		
		act	.021	191		
		ascgr	018	004		
		gender	.047	.121		
		mozartuse	1.000	002		
		pretest	002	1.000		
	Covariances	techsex	.728	.005		
		act	.027	006		
		ascgr	077	.000		
		gender	.144	.008		
		mozartuse	4.623	.000		
		pretest	.000	.002		

Coefficient Correlations<sup>a</sup>

a. Dependent Variable: fnlexam

Model	Dimension			Variance Proportions			
		Eigenvalue	Condition Index	(Constant)	pretest	act	ascgr
1	1	4.653	1.000	.00	.00	.00	.01
	2	1.000	2.157	.00	.00	.00	.00
	3	.678	2.620	.00	.00	.00	.00
	_ 4	.507	3.031	.00	.01	.00	.02
	5	.113	6.425	.00	.21	.00	.78
	6	.047	9.919	.02	.77	.02	.20
	7	.002	45.086	.98	.01	.98	.00

	Collinearity Diagnostics <sup>a</sup>							
Model	Dimension	Var	Variance Proportions					
		gender	gender mozartuse techsex					
1	1	.01	.01	.01				
	2	.00	.56	.24				
	3	.44	.20	.36				
	_ 4	.47	.22	.37				
	5	.01	.00	.00				
	6	.07	.00	.02				
	7	.00	.00	.00				

Casewise Diagnostics<sup>a</sup>

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.043	100	59.74	40.263
- 582	-3.065	35	75.56	-40.561

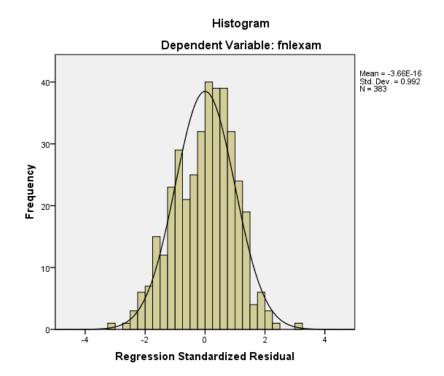
a. Dependent Variable: fnlexam

Residuals Statistics								
	Minimum	Maximum	Mean	Std. Deviation	Ν			
		05.00	00.00	0.477	000			
Predicted Value	49.23	85.60	68.99	6.177	383			
Std. Predicted Value	-3.199	2.688	.000	1.000	383			
Standard Error of Predicted	1.046	3.469	1.727	.467	383			
Value								
Adjusted Predicted Value	49.17	85.45	68.99	6.172	383			
Residual	-40.561	40.263	.000	13.127	383			
Std. Residual	-3.065	3.043	.000	.992	383			
Stud. Residual	-3.079	3.100	.000	1.002	383			
Deleted Residual	-40.920	41.792	.002	13.379	383			
Stud. Deleted Residual	-3.114	3.136	.000	1.004	383			
<mark>Mahal. Distance</mark>	<mark>1.390</mark>	<mark>25.254</mark>	<mark>5.984</mark>	<mark>3.947</mark>	<mark>383</mark>			
Cook's Distance	<mark>.000</mark>	<mark>.055</mark>	.003	.005	<mark>383</mark>			
Centered Leverage Value	.004	.066	.016	.010	383			

### **Residuals Statistics**<sup>a</sup>

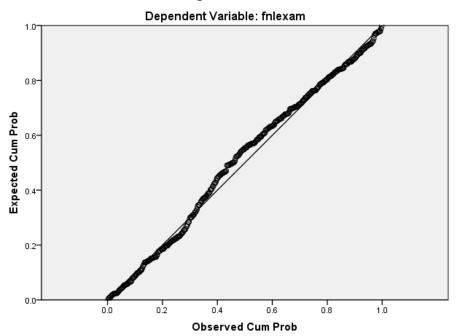
A maximum Mahalanobis distance of 25.254 greater than the critical chisquare value of 22.458 for df = 6 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .055 (which is less than one) means outliers should not be a concern.

# Charts

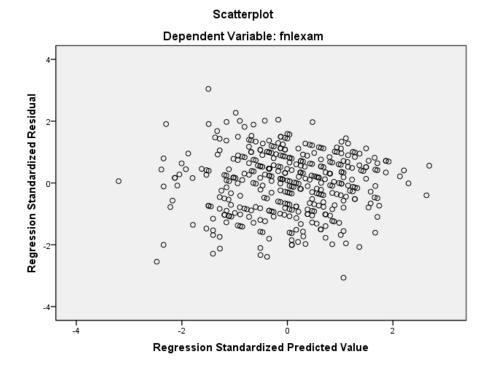


The points clustered fairly close to the line indicate that the residuals are normally distributed.





The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.



REGRESSION /DESCRIPTIVES MEAN STDDEV CORR SIG N /MISSING LISTWISE /STATISTICS COEFF OUTS CI(95) BCOV R ANOVA COLLIN TOL CHANGE ZPP /CRITERIA=PIN(.05) POUT(.10) /NOORIGIN /DEPENDENT fnlexam /METHOD=ENTER pretest act ascgr gender mozartuse comcol /SCATTERPLOT=(\*ZRESID ,\*ZPRED) /RESIDUALS DURBIN HISTOGRAM(ZRESID) NORMPROB(ZRESID) /CASEWISE PLOT(ZRESID) OUTLIERS(3) /SAVE MAHAL COOK.

# Regression

	Notes	
Output Created		20-Oct-2011 14:59:02
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing.
	Cases Used	Statistics are based on cases with no
		missing values for any variable used.

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Syntax		REGRESSION
,		/DESCRIPTIVES MEAN STDDEV
		CORR SIG N
		/MISSING LISTWISE
		/STATISTICS COEFF OUTS CI(95)
		BCOV R ANOVA COLLIN TOL
		CHANGE ZPP
		/CRITERIA=PIN(.05) POUT(.10)
		/NOORIGIN
		/DEPENDENT fnlexam
		/METHOD=ENTER pretest act ascgr
		gender mozartuse comcol
		/SCATTERPLOT=(*ZRESID
		,*ZPRED)
		/RESIDUALS DURBIN
		HISTOGRAM(ZRESID)
		NORMPROB(ZRESID)
		/CASEWISE PLOT(ZRESID)
		OUTLIERS(3)
		/SAVE MAHAL COOK.
Resources	Processor Time	00:00:01.700
	Elapsed Time	00:00:01.809
	Memory Required	4140 bytes
	Additional Memory Required	872 bytes
	for Residual Plots	
Variables Created or	MAH_7	Mahalanobis Distance
Modified	COO_7	Cook's Distance

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Descriptive Statistics							
	Mean	Std. Deviation	Ν				
fnlexam	68.99	14.508	383				
pretest	47.02	14.486	383				
Act	16.86	1.152	383				
ascgr	.87	.340	383				
gender	.38	.486	383				
mozartuse	.12	.322	383				
comcol	.19	.389	383				

N = 383 with k = 6 predictors

N > 50 + 8k

Therefore, the sample size is appropriate.

Correlations							
		fnlexam	pretest	act	ascgr	gender	
Pearson Correlation	fnlexam	1.000	.318	.222	.169	173	
	pretest	.318	1.000	.191	.023	119	
	act	.222	.191	1.000	.034	015	
	ascgr	.169	.023	.034	1.000	104	
	gender	173	119	015	104	1.000	
	mozartuse	.117	.016	020	.024	036	
	comcol	104	083	265	.029	.096	
Sig. (1-tailed)	fnlexam		.000	.000	.000	.000	
	pretest	.000		.000	.327	.010	
	act	.000	.000		.254	.386	
	ascgr	.000	.327	.254		.021	
	gender	.000	.010	.386	.021		
	mozartuse	.011	.376	.346	.322	.241	
	comcol	.021	.052	.000	.287	.030	
Ν	fnlexam	383	383	383	383	383	
	pretest	383	383	383	383	383	
	act	383	383	383	383	383	
	ascgr	383	383	383	383	383	
	gender	383	383	383	383	383	
	mozartuse	383	383	383	383	383	
	comcol	383	383	383	383	383	

Correlations							
		mozartuse	comcol				
Pearson Correlation	fnlexam	.117	104				
	pretest	.016	083				
	act	020	265				
	ascgr	.024	.029				
	gender	036	.096				
	mozartuse	1.000	.014				
	comcol	.014	1.000				
Sig. (1-tailed)	fnlexam	.011	.021				
	pretest	.376	.052				
	act	.346	.000				
	ascgr	.322	.287				
	gender	.241	.030				
	mozartuse		.394				
	comcol	.394	-				
Ν	fnlexam	383	383				
	pretest	383	383				
	act	383	383				
	ascgr	383	383				
	gender	383	383				
	mozartuse	383	383				
	comcol	383	383				

## Variables Entered/Removed<sup>b</sup>

Model	Variables	Variables	
	Entered	Removed	Method
1	comcol,		Enter
	mozartuse,		
	ascgr, pretest,		
	gender, act <sup>a</sup>		

a. All requested variables entered.

	Model Summary <sup>b</sup>								
Model			Change Statistics						
	_		Adjusted R	Std. Error of the	R Square				
	R	<mark>R Square</mark>	Square	Estimate	Change	F Change	df1		
1	.425 <sup>a</sup>	<mark>.180</mark>	.167	13.240	.180	13.787	6		

a. Predictors: (Constant), comcol, mozartuse, ascgr, pretest, gender, act

b. Dependent Variable: fnlexam

Model Summary <sup>b</sup>					
Model	Chang	ge Statistics			
	df2	Sig. F Change	Durbin-Watson		
1	376	.000	1.730		

b. Dependent Variable: fnlexam

			ANOVA <sup>b</sup>			
Mode		Sum of Squares	df	Mean Square	F	Sig.
1	Regression	14499.826	6	2416.638	<mark>13.787</mark>	.000 <sup>a</sup>
	Residual	65908.132	376	175.288		
	Total	80407.958	382			

a. Predictors: (Constant), comcol, mozartuse, ascgr, pretest, gender, act

-	Coefficients <sup>a</sup>						
Mod	el			Standardized			
		Unstandardize	ed Coefficients	Coefficients			
		B	Std. Error	Beta	t	Sig.	
1	(Constant)	<mark>18.577</mark>	10.508		1.768	<mark>.078</mark>	
	pretest	. <mark>266</mark>	.048	.266	5.551	.000	
	act	<mark>1.990</mark>	.621	.158	3.207	<mark>.001</mark>	
	ascgr	<mark>6.131</mark>	2.006	.144	3.056	<mark>.002</mark>	
	gender	- <mark>3.478</mark>	1.418	117	-2.453	.015	
	mozartuse	<mark>4.879</mark>	2.104	.108	2.319	.021	
	comcol	<mark>-1.269</mark>	1.816	034	699	<mark>.485</mark>	

### **Coefficients**<sup>a</sup> Model 95.0% Confidence Interval for B Correlations Lower Bound Upper Bound Zero-order Partial Part 39.239 1 (Constant) -2.084 .172 .361 .318 .275 .259 pretest .222 .770 3.210 .163 .150 act 2.187 10.075 .169 .143 ascgr .156 gender -6.266 -.691 -.173 -.126 -.115 9.016 mozartuse .743 .117 .119 .108 -4.841 2.303 -.104 -.036 -.033 comcol

Coefficients <sup>a</sup>					
Model		Collinearity	Statistics		
		Tolerance	VIF		
1	(Constant)				
	pretest	<mark>.949</mark>	<mark>1.054</mark>		
	act	.898	<mark>1.114</mark>		
	ascgr	.985	<mark>1.015</mark>		
	gender	<mark>.965</mark>	<mark>1.036</mark>		
	mozartuse	<mark>.997</mark>	<mark>1.003</mark>		
	comcol	. <mark>918</mark>	<mark>1.089</mark>		

# Tolerance > 0.2 and VIF < 5 indicate that multicollinearity is not a problem.

	Coefficient Correlations*					
Model			comcol	mozartuse	ascgr	pretest
1	Correlations	comcol	1.000	012	049	.024
		mozartuse	012	1.000	020	017
		ascgr	049	020	1.000	005
		pretest	.024	017	005	1.000
		gender	097	.033	.106	.114
		act	.258	.020	043	179
	Covariances	comcol	3.300	044	180	.002
		mozartuse	044	4.425	085	002
		ascgr	180	085	4.024	001
		pretest	.002	002	001	.002
		gender	250	.098	.300	.008
		act	.290	.027	054	005

**Coefficient Correlations**<sup>a</sup>

	Coefficient Correlations <sup>a</sup>					
Model			gender	act		
1	Correlations	comcol	097	.258		
		mozartuse	.033	.020		
		ascgr	.106	043		
		pretest	.114	179		
		gender	1.000	035		
		act	035	1.000		
	Covariances	comcol	250	.290		
		mozartuse	.098	.027		
		ascgr	.300	054		
		pretest	.008	005		
		gender	2.010	031		
		act	031	.385		

Model	Dimension				Variance Pr	oportions	
		Eigenvalue	Condition Index	(Constant)	pretest	act	ascgr
1	1	4.620	1.000	.00	.00	.00	.01
	2	.875	2.297	.00	.00	.00	.00
	3	.774	2.443	.00	.00	.00	.00
	_ 4	.568	2.853	.00	.01	.00	.01
	5	.112	6.415	.00	.21	.00	.79
	6	.048	9.790	.02	.78	.02	.19
	7	.002	46.676	.98	.00	.98	.00

	Collinearity Diagnostics <sup>a</sup>					
Model	Dimension	Variance Proportions				
		gender	mozartuse	comcol		
1	1	.01	.01	.01		
	2	.04	.88	.06		
	3	.00	.07	.82		
	_ 4	.88	.04	.03		
	5	.01	.00	.00		
	6	.06	.00	.00		
	7	.00	.00	.07		

Casewise Diagnostics<sup>a</sup>

\_

Case Number	Std. Residual	fnlexam	Predicted Value	Residual
192	3.125	100	58.63	41.368
- 582	-3.035	35	75.18	-40.179

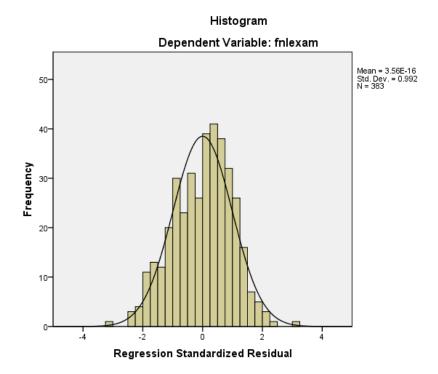
a. Dependent Variable: fnlexam

	Minimum	Maximum	Mean	Std. Deviation	Ν
Predicted Value	48.37	85.39	68.99	6.161	383
Std. Predicted Value	-3.348	2.662	.000	1.000	383
Stu. Fledicted Value	-3.340	2.002	.000	1.000	303
Standard Error of Predicted	.969	4.029	1.710	.530	383
Value					
Adjusted Predicted Value	48.24	85.57	68.99	6.156	383
Residual	-40.179	41.368	.000	13.135	383
Std. Residual	-3.035	3.125	.000	.992	383
Stud. Residual	-3.046	3.188	.000	1.002	383
Deleted Residual	-40.485	43.077	.003	13.397	383
Stud. Deleted Residual	-3.080	3.228	.000	1.004	383
<mark>Mahal. Distance</mark>	<mark>1.047</mark>	<mark>34.376</mark>	<mark>5.984</mark>	<mark>4.499</mark>	<mark>383</mark>
Cook's Distance	<mark>.000</mark>	<mark>.060</mark>	.003	.006	<mark>383</mark>
Centered Leverage Value	.003	.090	.016	.012	383

### **Residuals Statistics**<sup>a</sup>

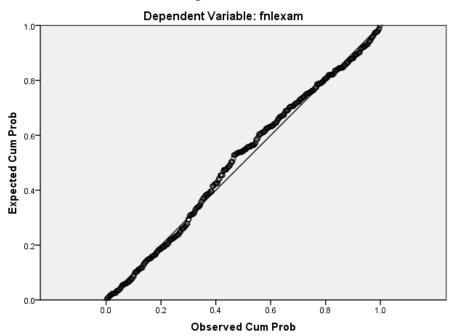
A maximum Mahalanobis distance of 34.376 greater than the critical chisquare value of 22.458 for df = 6 at  $\alpha$  = .001 indicates the presence of one or more multivariate outliers, but the maximum Cook's distance of .060 (which is less than one) means outliers should not be a concern.

### Charts

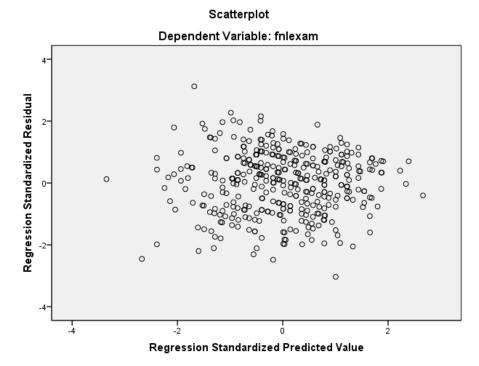


The points clustered fairly close to the line indicate that the residuals are normally distributed.





The absence of a pattern in the scatterplot indicates normality, linearity, and homoscedasticity of residuals.





# Appendix W: SPSS Multiple Binary Logistic Regression Output for

Elementary Algebra

GET

FILE='C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav'. DATASET NAME DataSet1 WINDOW=FRONT. LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr /CONTRAST (ascgr)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

**Logistic Regression** 

	Notes	
Output Created		20-Oct-2011 18:10:15
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		/CONTRAST (ascgr)=Indicator(1)
		/SAVE=PRED PGROUP COOK
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.016
	Elapsed Time	00:00:00.051
Variables Created or	PRE_1	Predicted probability
Modified	PGR_1	Predicted group
	COO_1	Analog of Cook's influence statistics
	LEV_1	Leverage value
	ZRE_1	Normalized residual
	DFB0_1	DFBETA for constant
	DFB1_1	DFBETA for pretest
	DFB2_1	DFBETA for ascgr(1)

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	N	Percent				
Selected Cases	Included in Analysis	235	93.3			
	Missing Cases	17	6.7			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

### Dependent Variable Encoding

Original Value	Internal Value		
0	0		
<sup>—</sup> 1	1		

### **Categorical Variables Codings**

			Parameter coding
		Frequency	(1)
ascgr	0	71	.000
	1	164	1.000

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	321.130	.281		
	2	321.130	.283		
	3	321.130	.283		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Classification Table <sup>a,b</sup>						
Observed			Predicted				
		fnlgrd					
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	101	.0		
		1	0	134	100.0		
	Overall	Percentage			57.0		

### a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0 C	onstant	.283	.132	4.603	1	.032	1.327

### Variables not in the Equation

			Score	df	Sig.
Step 0	Variables	pretest	15.826	1	.000
		ascgr(1)	49.372	1	.000
Overall Statistics		65.643	2	.000	

# Block 1: Method = Enter

			- i		
Iteration			Coefficients		
		-2 Log likelihood	Constant	pretest	ascgr(1)
Step 1	1	251.458	-2.448	.033	1.986
	2	248.607	-3.201	.046	2.382
	3	248.561	-3.312	.048	2.438
	4	248.561	-3.315	.048	2.439
	5	248.561	-3.315	.048	2.439

Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 5 because parameter

estimates changed by less than .001.

### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	72.569	2	.000
	Block	72.569	2	.000
	Model	72.569	2	.000

#### **Model Summary**

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	248.561 <sup>a</sup>	.266	.357

a. Estimation terminated at iteration number 5 because

parameter estimates changed by less than .001.

#### **Hosmer and Lemeshow Test**

Step	Chi-square	Df	Sig.
1	5.952	7	.545

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	19	19.546	3	2.454	22
	2	21	21.154	5	4.846	26
	3	16	14.886	7	8.114	23
	4	13	10.784	9	11.216	22
	5	6	5.062	6	6.938	12
	6	5	9.493	21	16.507	26
	7	11	9.544	23	24.456	34
	8	5	6.525	28	26.475	33
	9	5	4.006	32	32.994	37

Contingency Table for Hosmer and Lemeshow Test

### Classification Table<sup>a</sup>

	Observed		Predicted			
			fnlgrd		Percentage	
			0	1	Correct	
Step 1	fnlgrd	0	58	43	57.4	
		1	16	118	88.1	
	Overall	Percentage			74.9	

a. The cut value is .500

Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.048	.011	18.690	1	.000	1.049
	ascgr(1)	2.439	.366	44.339	1	.000	11.464
	Constant	-3.315	.586	32.016	1	.000	.036

a. Variable(s) entered on step 1: pretest, ascgr.

Variables in the Equation				
		95% C.I.for EXP(B)		
		Lower Upper		
Step 1 <sup>a</sup>	pretest	1.026	1.072	
	ascgr(1)	5.591	23.503	
	Constant			

a. Variable(s) entered on step 1: pretest, ascgr.

Step number: 1

**Observed Groups and Predicted Probabilities** 32 + + T T I F L 1 R 24 +1 Е 1 Т 1 Q Τ 1 1 1 U Τ 1 1 1 1 Е 16 + 1 1 1 1 Ν 1 1 1 L 1 1 L С Т 1 1 1 1 1 1 1 1 Y I 001 1 1 11 T 0 1 1 1 1 8 + 1 0 0 0 0 1 1 1 1 1 1 1 1 1 01000 0 1 0 0 0 0 1 1 1 0 0 1 1 000000 0 0 0 11 0 0 Т 0 0 0 0 1 1 1 1 1 1 I 00 0000000 01 00 00 0 0 0 10 0 0 0 0 0 L 11011 I Predicted -----\_+\_\_ -+-----Prob: 0 .1 .2 .3 .4 .5 .6 .7 .8 .9 1 Group: 

+

I

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 2 Cases.

	Oucowice En	
Observed		

Casewise List<sup>b</sup>

0400		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	Fnlgrd	Predicted	Predicted Group	Resid	ZResid
51	S	0**	.879	1	879	-2.694
67	S	1**	.086	0	.914	3.258
77	S	1**	.107	0	.893	2.892
145	S	0**	.950	1	950	-4.337
168	S	0**	.902	1	902	-3.035
177	S	1**	.132	0	.868	2.568

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

Case

	Notes	
Output Created		20-Oct-2011 18:12:24
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096
		/CONTRAST (ascgr)=Indicator(1)
		/CONTRAST (adj096)=Indicator(1) /SAVE=PRED PGROUP COOK
		LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.032
	Elapsed Time	00:00:00.059
Variables Created or	PRE_2	Predicted probability
Modified	PGR_2	Predicted group
	COO_2	Analog of Cook's influence statistics
	LEV_2	Leverage value
	ZRE_2	Normalized residual
	DFB0_2	DFBETA for constant
	DFB1_2	DFBETA for pretest
	DFB2_2	DFBETA for ascgr(1)
	DFB3_2	DFBETA for adj096(1)

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary						
Unweighted Case	N	Percent				
Selected Cases	Included in Analysis	235	93.3			
	Missing Cases	17	6.7			
	Total	252	100.0			
Unselected Cases	;	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

## Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

### Categorical Variables Codings

			Parameter coding
		Frequency	(1)
adj096	0	128	.000
	1	107	1.000
ascgr	0	71	.000
	1	164	1.000

# Block 0: Beginning Block

Iteration History <sup>a,b,c</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant	
Step 0	1	321.130	.281	
	2	321.130	.283	
	3	321.130	.283	

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

Iteration History <sup>a,b,c</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant	
Step 0	1	321.130	.281	
	2	321.130	.283	
	3	321.130	.283	

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 321.130

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Classification Table						
	Predicte	d				
fnl	grd					
		Percentage				
0	1	Correct				
0	101	.0				
0	134	100.0				
		57.0				
	fnl 0 0	Predicte fnlgrd 0 1 0 101				

## Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.283	.132	4.603	1	.032	1.327

## Variables not in the Equation

			Score	df	Sig.
Step 0	Variables	Pretest	15.826	1	.000
		ascgr(1)	49.372	1	.000
		adj096(1)	11.809	1	.001
	Overall Stat	istics	80.498	3	.000

## Block 1: Method = Enter

Iteration History <sup>a,b,c,d</sup>							
Iteration				Coeffi	cients		
		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)	
Step 1	1	233.558	-2.838	.029	2.121	1.016	
	2	226.843	-3.977	.042	2.791	1.504	
	3	226.527	-4.279	.045	2.981	1.649	
	4	226.526	-4.298	.045	2.993	1.659	
	5	226.526	-4.298	.045	2.993	1.659	

a,b,c,d

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 321.130

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	94.604	3	.000
	Block	94.604	3	.000
	Model	94.604	3	.000

#### Model Summary

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	226.526 <sup>a</sup>	.331	.445

a. Estimation terminated at iteration number 5 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.
1	5.120	7	.645

		fnlgrd = 0		fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	22	21.626	1	1.374	23
	2	19	21.294	7	4.706	26
	3	19	15.972	6	9.028	25
	4	13	14.975	16	14.025	29
	5	12	10.569	14	15.431	26
	6	6	7.771	21	19.229	27
	7	6	5.318	24	24.682	30
	8	3	2.111	18	18.889	21
	9	1	1.364	27	26.636	28

Contingency Table for Hosmer and Lemeshow Test

### Classification Table<sup>a</sup>

Observed		Predicted			
		fnlı	grd	Percentage	
			0	1	Correct
Step 1	fnlgrd	0	69	32	68.3
		1	20	114	85.1
	Overall	Percentage			77.9

a. The cut value is .500

Variables	in	the	Fountion
variables		uie	Lyuanon

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.045	.011	15.496	1	.000	1.046
	ascgr(1)	2.993	.436	47.066	1	.000	19.952
	adj096(1)	1.659	.389	18.229	1	.000	5.255
	Constant	-4.298	.679	40.068	1	.000	.014

a. Variable(s) entered on step 1: pretest, ascgr, adj096.

### Variables in the Equation

		95% C.I.for EXP(B)	
		Lower	Upper
Step 1 <sup>a</sup>	pretest	1.023	1.070
	ascgr(1)	8.484	46.922
	adj096(1)	2.454	11.256
	Constant		

a. Variable(s) entered on step 1: pretest, ascgr, adj096.

Step number: 1

**Observed Groups and Predicted Probabilities** 

16 + + 1 1 T T 1 1 L F Т 1 1 1 R 12 +1 1 1 1 Е 1 Т 1 1 1 Q T 1 1 1 1 1 1 U 1 1 0 1 1 1 1 1 I Е 8 + 0 0 1 1 1 1 111 1 1 Ν 0 1 1 1 1 1 1111 T Т 0 1 1 С 00 1 0 1 1 1 0 1 1 11 11 1111 L Y 00 1 0 01 0 0 0 1 1 11 Т 1 11 1111 I 001 0 1 1 1 0 1 0 00 0 0 1 11 1 11 4 + 0 0 1 11 111 + I 00 00 0 0 1 0 0 1 0 00 0 01 0 0 1 1 11 1 11 11 1 111 111 I 00 00 0 0 0 0 0 0 0 0 0 0 0 0 0 01 0 0 1 00 11 11 1 1 1 1111111 11 000 0111111 -+-----Prob: 0 .1 .2 .3 .4 .5 .6 .7 .8 .9 1 Group: Predicted Probability is of Membership for 1

The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1 Case.

Casewise L	ist <sup>b</sup>
------------	------------------

Case	Selected	Observed			Temporar	y Variable
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.076	0	.924	3.481
69	S	0**	.916	1	916	-3.292
145	S	0**	.909	1	909	-3.160
238	S	0**	.931	1	931	-3.685
251	S	0**	.896	1	896	-2.941

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 numbmeet /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# **Logistic Regression**

	Notes	
Output Created		20-Oct-2011 18:13:31
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	I I

Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096 numbmeet
		/CONTRAST (ascgr)=Indicator(1)
		/CONTRAST (adj096)=Indicator(1)
		/SAVE=PRED PGROUP COOK
		/CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.047
	Elapsed Time	00:00:00.050
Variables Created or	PRE_3	Predicted probability
Modified	PGR_3	Predicted group
	COO_3	Analog of Cook's influence statistics
	LEV_3	Leverage value
	ZRE_3	Normalized residual
	DFB0_3	DFBETA for constant
	DFB1_3	DFBETA for pretest
	DFB2_3	DFBETA for ascgr(1)
	DFB3_3	DFBETA for adj096(1)
	DFB4_3	DFBETA for numbmeet

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Cases	N	Percent			
Selected Cases	Included in Analysis	226	89.7		
	Missing Cases	26	10.3		
	Total	252	100.0		
Unselected Cases	i de la companya de l	0	.0		
Total		252	100.0		

## **Case Processing Summary**

Case Processing Summary						
Unweighted Case	Unweighted Cases <sup>a</sup>					
Selected Cases	Included in Analysis	226	89.7			
	Missing Cases	26	10.3			
	Total	252	100.0			
Unselected Cases	3	0	.0			
Total		252	100.0			

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

#### **Categorical Variables Codings**

			Parameter coding
		Frequency	(1)
adj096	0	119	.000
	1	107	1.000
ascgr	0	65	.000
	1	161	1.000

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	305.453	.372		
	2	305.452	.376		
	3	305.452	.376		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	305.453	.372		
	2	305.452	.376		
	3	305.452	.376		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Predicted						
fnle	grd					
		Percentage				
0	1	Correct				
0	92	.0				
0	134	100.0				
		59.3				
	fnlı 0 0	Predicte fnlgrd 0 1 0 92				

## Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.376	.135	7.714	1	.005	1.457

## Variables not in the Equation

			Score	df	Sig.
Step 0	Variables	pretest	15.997	1	.000
		ascgr(1)	45.456	1	.000
		adj096(1)	8.196	1	.004
	-	numbmeet	9.660	1	.002

Valiables not in the Equation						
			Score	df	Sig.	
Step 0	Variables	pretest	15.997	1	.000	
		ascgr(1)	45.456	1	.000	
		adj096(1)	8.196	1	.004	
		numbmeet	9.660	1	.002	
	Overall Stat	istics	78.216	4	.000	

Variables not in the Equation

# Block 1: Method = Enter

## Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients				
-2 Log likeliho		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)	numbmeet
Step 1	1	220.828	-1.729	.030	2.116	.588	251
	2	213.026	-2.375	.044	2.827	.899	385
	3	212.562	-2.562	.049	3.059	1.006	429
	4	212.559	-2.577	.049	3.079	1.015	433
	5	212.559	-2.577	.049	3.079	1.015	433

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

### Omnibus Tests of Model Coefficients

		Chi-square	df	Sig.
Step 1	Step	92.893	4	.000
	Block	92.893	4	.000
	Model	92.893	4	.000

### Model Summary

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	212.559 <sup>a</sup>	.337	.455

Model Summary						
Step	Cox & Snell R Nagelkerke R					
	-2 Log likelihood	Square	Square			
1	212.559 <sup>a</sup>	.337	.455			

a. Estimation terminated at iteration number 5 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.	
1	6.372	8	.606	

	fn		d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	23	22.300	1	1.700	24
	2	18	18.209	5	4.791	23
	3	15	14.256	8	8.744	23
	4	9	11.548	14	11.452	23
	5	12	8.762	10	13.238	22
	6	6	6.973	17	16.027	23
	7	2	4.529	20	17.471	22
	8	4	2.645	15	16.355	19
	9	2	1.974	21	21.026	23
	10	1	.804	23	23.196	24

## Contingency Table for Hosmer and Lemeshow Test

Classification	Table <sup>a</sup>
----------------	--------------------

Observed		Predicted			
			fnlgrd		Percentage
			0	1	Correct
Step 1	fnlgrd	0	62	30	67.4
		1	20	114	85.1
	Overall	Percentage			77.9

a. The cut value is .500

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.049	.012	16.454	1	.000	1.050
	ascgr(1)	3.079	.464	44.021	1	.000	21.747
	adj096(1)	1.015	.431	5.554	1	.018	2.760
	numbmeet	433	.160	7.284	1	.007	.649
	Constant	-2.577	.864	8.891	1	.003	.076

## Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr, adj096, numbmeet.

Variables in the Equation				
		95% C.I.fo	or EXP(B)	
		Lower	Upper	
Step 1 <sup>a</sup>	pretest	1.026	1.075	
	ascgr(1)	8.756	54.009	
	adj096(1)	1.186	6.423	
	numbmeet	.474	.888	
	Constant			

a. Variable(s) entered on step 1: pretest, ascgr, adj096, numbmeet.

Step number: 1

**Observed Groups and Predicted Probabilities** 

16 + + T T I F Т Т R 12 +Е Q T U Τ 1 1 Е 8+ 1 + 1 Ν 1 1 1 Т 1 1111 С Т 1 11 1 1 1 1111 Υ L 1 1 1 1 11 1 11 1 1 11 1 1 1 111 111 4 + 01 0 1 11 11 11 11 1 11 1 11 11 11 11 111 11 + I 000 01 01 1 00 11 00 011 10 011 11 1111 1111 1 1 1 111 111 I 000000 000 00 0 00 00 11 000 010 00 010 0111 0110 111 11111 1 111 111 1 10 1110 010000111011111 т. -+-----Prob: 0 .1 .2 .3 .4 .5 .6 .7 .8 .9 1 Group: Predicted Probability is of Membership for 1

The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1 Case.

Case						
		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.058	0	.942	4.015
44	S	0**	.890	1	890	-2.840
69	S	0**	.946	1	946	-4.174
80	S	0**	.893	1	893	-2.890
145	S	0**	.905	1	905	-3.095
147	S	1**	.142	0	.858	2.460
168	S	0**	.876	1	876	-2.661
177	S	1**	.139	0	.861	2.487
238	S	0**	.859	1	859	-2.465

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 numbmeet mozartuse /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /CONTRAST (mozartuse)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

Notes					
Output Created		20-Oct-2011 18:14:33			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet1			
	Filter	<none></none>			
	Weight	<none></none>			

	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096 numbmeet mozartuse
		/CONTRAST (ascgr)=Indicator(1)
		/CONTRAST (adj096)=Indicator(1) /CONTRAST
		(mozartuse)=Indicator(1)
		/SAVE=PRED PGROUP COOK
		LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.062
	Elapsed Time	00:00:00.073
Variables Created or	PRE_4	Predicted probability
Modified	PGR_4	Predicted group
	COO_4	Analog of Cook's influence statistics
	LEV_4	Leverage value
	ZRE_4	Normalized residual
	DFB0_4	DFBETA for constant
	DFB1_4	DFBETA for pretest
	DFB2_4	DFBETA for ascgr(1)
	DFB3_4	DFBETA for adj096(1)
	DFB4_4	DFBETA for numbmeet
	DFB5_4	DFBETA for mozartuse(1)

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	Unweighted Cases <sup>a</sup>				
Selected Cases	Included in Analysis	226	89.7		
	Missing Cases	26	10.3		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

## Categorical Variables Codings

			Parameter coding
		Frequency	(1)
mozartuse	0	184	.000
	1	42	1.000
adj096	0	119	.000
	1	107	1.000
ascgr	0	65	.000
	1	161	1.000

# Block 0: Beginning Block

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	305.453	.372		
	2	305.452	.376		
	3	305.452	.376		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

Iteration History <sup>a,b,c</sup>				
Iteration			Coefficients	
		-2 Log likelihood	Constant	
Step 0	1	305.453	.372	
	2	305.452	.376	
	3	305.452	.376	

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

		Classificat			
Observed			Predicte	d	
			fnl	grd	
					Percentage
			0	1	Correct
Step 0	fnlgrd	0	0	92	.0
		1	0	134	100.0
	Overall	Percentage			59.3

## Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.376	.135	7.714	1	.005	1.457

Variables not in the Equation								
			Score	df	Sig.			
Step 0	Variables	Pretest	15.997	1	.000			
		ascgr(1)	45.456	1	.000			
		adj096(1)	8.196	1	.004			
		Numbmeet	9.660	1	.002			
		mozartuse(1)	11.881	1	.001			
	Overall Stat	istics	79.920	5	.000			

# Variables not in the Equation

# Block 1: Method = Enter

Iteration			Coefficients					
		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)		
Step 1	1	219.005	-1.757	.029	2.093	.481		
	2	211.155	-2.457	.044	2.808	.767		
	3	210.701	-2.660	.049	3.035	.870		
	4	210.699	-2.675	.049	3.054	.879		
	5	210.699	-2.676	.049	3.054	.879		

Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

# Iteration History<sup>a,b,c,d</sup>

Iteration		Coefficients		
		numbmeet	mozartuse(1)	
Step 1	1	193	527	
	2	306	658	
	3	346	674	
	4	349	674	
	5	349	674	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 5

because parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	94.753	5	.000
	Block	94.753	5	.000
	Model	94.753	5	.000

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	210.699 <sup>a</sup>	.342	.462

a. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	2.966	8	.936

		fnlgro	d = 0	fnlgrd = 1		
		Observed	Expected	Observed	Expected	Total
Step 1	1	21	21.414	2	1.586	23
	2	19	18.495	4	4.505	23
	3	16	14.927	7	8.073	23
	4	9	11.073	13	10.927	22
	5	9	9.391	14	13.609	23
	6	8	6.881	15	16.119	23
	7	4	4.405	19	18.595	23
	8	3	2.940	20	20.060	23
	9	3	1.842	20	21.158	23
	10	0	.632	20	19.368	20

Contingency Table for Hosmer and Lemeshow Test

#### Classification Table<sup>a</sup>

Observed				Predicte	d
		fnlgrd			
					Percentage
			0	1	Correct
Step 1	fnlgrd	0	64	28	69.6
		1	20	114	85.1
	Overall	Percentage			78.8

a. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.049	.012	16.171	1	.000	1.050
	ascgr(1)	3.054	.460	44.058	1	.000	21.199
	adj096(1)	.879	.444	3.910	1	.048	2.408
	numbmeet	349	.171	4.143	1	.042	.705
	mozartuse(1)	674	.497	1.840	1	.175	.510
	Constant	-2.676	.876	9.323	1	.002	.069

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.049	.012	16.171	1	.000	1.050
	ascgr(1)	3.054	.460	44.058	1	.000	21.199
	adj096(1)	.879	.444	3.910	1	.048	2.408
	numbmeet	349	.171	4.143	1	.042	.705
	mozartuse(1)	674	.497	1.840	1	.175	.510
	Constant	-2.676	.876	9.323	1	.002	.069

## Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr, adj096, numbmeet, mozartuse.

		95% C.I.for EXP(B)				
		Lower	Upper			
Step 1 <sup>a</sup>	pretest	1.025	1.075			
	ascgr(1)	8.604	52.234			
	adj096(1)	1.008	5.754			
	numbmeet	.504	.987			
	mozartuse(1)	.193	1.350			
	Constant					

#### Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr,

adj096, numbmeet, mozartuse.

Step number: 1

**Observed Groups and Predicted Probabilities** 

16 + + T Т F Т R 12 +Е 1 Q T 1 U T 1 Е 8+ 1 1 1 1 + Ν 1 1 1 1 11 Т 1 1 С Т 1 1 1 1 1 1 11111 1 Y L 1 1 1 1 11 1 1 11111 1 1 1 100 1 0 0111 10 4 + 1 1 1 1 1 1 1 1 1 11 11 + 000 0 0 0011011100111111 10 1 0 0 11111 1 11 11 1 110 111110111 111111 1101 11 011001110000111 01111 | -+-----Prob: 0 1 .2 .3 .5 .7 .8 .9 .1 .4 .6 Group: 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1 Case.

Case		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.078	0	.922	3.430
44	S	0**	.893	1	893	-2.894
69	S	0**	.940	1	940	-3.974
80	S	0**	.884	1	884	-2.754
145	S	0**	.867	1	867	-2.549
147	S	1**	.103	0	.897	2.946
168	S	0**	.897	1	897	-2.945
238	S	0**	.876	1	876	-2.660

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 numbmeet techsex /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /CONTRAST (techsex)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

Notes							
Output Created		20-Oct-2011 18:17:31					
Comments							
Input	Data	C:\Users\Lin\Documents\math 096 fall					
		2001 with classrooms.sav					
	Active Dataset	DataSet1					
	Filter	<none></none>					

	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data File	252
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096 numbmeet techsex
		/CONTRAST (ascgr)=Indicator(1)
		/CONTRAST (adj096)=Indicator(1)
		/CONTRAST (techsex)=Indicator(1)
		/SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.032
	Elapsed Time	00:00:00.056
Variables Created or	PRE_5	Predicted probability
Modified	PGR_5	Predicted group
	COO_5	Analog of Cook's influence statistics
	LEV_5	Leverage value
	ZRE_5	Normalized residual
	DFB0_5	DFBETA for constant
	DFB1_5	DFBETA for pretest
	DFB2_5	DFBETA for ascgr(1)
	DFB3_5	DFBETA for adj096(1)
	DFB4_5	DFBETA for numbmeet
	DFB5_5	DFBETA for techsex(1)

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary							
Unweighted Case	N	Percent					
Selected Cases	Included in Analysis	226	89.7				
	Missing Cases	26	10.3				
	Total	252	100.0				
Unselected Cases	3	0	.0				
Total		252	100.0				

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

#### Categorical Variables Codings

			Parameter coding
		Frequency	(1)
techsex	0	149	.000
	1	77	1.000
adj096	0	119	.000
	1	107	1.000
ascgr	0	65	.000
	1	161	1.000

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	305.453	.372		
	2	305.452	.376		
	3	305.452	.376		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	305.453	.372		
	2	305.452	.376		
	3	305.452	.376		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 305.452

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Observe	ed		Predicte	d	
			fnl	grd		
					Percentage	
			0	1	Correct	
Step 0	fnlgrd	0	0	92	.0	
		1	0	134	100.0	
	Overall	Percentage			59.3	

## Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.376	.135	7.714	1	.005	1.457

Valiables not in the Equation						
			Score	df	Sig.	
Step 0	Variables	Pretest	15.997	1	.000	
		ascgr(1)	45.456	1	.000	
		adj096(1)	8.196	1	.004	
		Numbmeet	9.660	1	.002	
		techsex(1)	12.438	1	.000	
	Overall Statistics			5	.000	

# Variables not in the Equation

# Block 1: Method = Enter

Iteration			Coefficients			
		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)
Step 1	1	216.964	-2.287	.028	2.069	.707
	2	207.925	-3.207	.043	2.799	1.072
	3	207.288	-3.503	.048	3.062	1.200
	4	207.283	-3.530	.048	3.090	1.213
	5	207.283	-3.530	.048	3.090	1.213

Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

## Iteration History<sup>a,b,c,d</sup>

Iteration		Coefficients		
		numbmeet	techsex(1)	
Step 1	1	139	.563	
	2	231	.846	
	3	267	.941	
	4	271	.949	
	5	271	.949	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 305.452

d. Estimation terminated at iteration number 5

because parameter estimates changed by less than .001.

**Omnibus Tests of Model Coefficients** 

		Chi-square	df	Sig.
Step 1	Step	98.168	5	.000
	Block	98.168	5	.000
	Model	98.168	5	.000

**Model Summary** 

Step		Cox & Snell R	Nagelkerke R	
	-2 Log likelihood	Square	Square	
1	207.283 <sup>a</sup>	.352	.475	

a. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	7.905	8	.443

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	22	22.399	2	1.601	24
	2	18	18.504	5	4.496	23
	3	18	14.416	5	8.584	23
	4	9	11.674	14	11.326	23
	5	9	9.042	13	12.958	22
	6	5	6.903	18	16.097	23
	7	5	4.718	20	20.282	25
	8	5	2.512	17	19.488	22
	9	1	1.443	22	21.557	23
	10	0	.390	18	17.610	18

Contingency Table for Hosmer and Lemeshow Test

### Classification Table<sup>a</sup>

Observed		Predicted			
		fnlgrd			
				Percentage	
		0	1	Correct	
Step 1 fnlgrd 0		63	29	68.5	
1		20	114	85.1	
Overall Percentag	je			78.3	

a. The cut value is .500

Variables	in	the	Equation
-----------	----	-----	----------

		В	S.E.	Wald	df	Sig	Evp(P)
		D	3.E.	vvalu	u	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.048	.012	15.304	1	.000	1.050
	ascgr(1)	3.090	.471	43.015	1	.000	21.974
	adj096(1)	1.213	.445	7.442	1	.006	3.365
	numbmeet	271	.178	2.336	1	.126	.762
	techsex(1)	.949	.423	5.049	1	.025	2.584
	Constant	-3.530	.990	12.714	1	.000	.029

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.048	.012	15.304	1	.000	1.050
	ascgr(1)	3.090	.471	43.015	1	.000	21.974
	adj096(1)	1.213	.445	7.442	1	.006	3.365
	numbmeet	271	.178	2.336	1	.126	.762
	techsex(1)	.949	.423	5.049	1	.025	2.584
	Constant	-3.530	.990	12.714	1	.000	.029

Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr, adj096, numbmeet, techsex.

		95% C.I.fo	or EXP(B)
		Lower	Upper
Step 1 <sup>a</sup>	pretest	1.024	1.075
	ascgr(1)	8.727	55.324
	adj096(1)	1.407	8.046
	numbmeet	.538	1.080
	techsex(1)	1.129	5.916
	Constant		

#### Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr,

adj096, numbmeet, techsex.

Step number: 1

**Observed Groups and Predicted Probabilities** 

16 + + T T I F Т T R 12 ++ Е Q T U T 1 Е 8+ 1 11 1 +Ν 1 1 Т 1 1111 С 10 1 1 1 11111 1 1 Y 10 11 1 1111111 1 1 1 1 4 + 01 11 11 1 11 11 11 1 01 1 01 1 11 1 1 11 + 0 10 0 001 1 00 1 11 11 11 11 11 11 1 1 00 01 11 111 1 111 11 11111101101111 11 | 0001 10 01 100001011011111111 -+-----Prob: 0 1 .1 .2 .3 .5 .6 .7 .8 .9 .4 Group: 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1 Case.

Case						
		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.119	0	.881	2.720
44	S	0**	.916	1	916	-3.300
51	S	0**	.887	1	887	-2.797
59	S	0**	.860	1	860	-2.478
67	S	1**	.131	0	.869	2.574
69	S	0**	.918	1	918	-3.336
145	S	0**	.889	1	889	-2.824
147	S	1**	.121	0	.879	2.694
177	S	1**	.125	0	.875	2.651
238	S	0**	.863	1	863	-2.506

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 techsex amisone /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /CONTRAST (techsex)=Indicator(1) /CONTRAST (amisone)=Indicator(1) /CONTRAST (amisone)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

### **Logistic Regression**

Notes					
Output Created		20-Oct-2011 18:19:28			
Comments					
Input	Data	C:\Users\Lin\Documents\math 096 fall			
		2001 with classrooms.sav			
	Active Dataset	DataSet1			

	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096 techsex amisone
		/CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1)
		/CONTRAST (techsex)=Indicator(1)
		/CONTRAST (amisone)=Indicator(1)
		/SAVE=PRED PGROUP COOK
		LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.032
	Elapsed Time	00:00:00.048
Variables Created or	PRE_6	Predicted probability
Modified	PGR_6	Predicted group
	COO_6	Analog of Cook's influence statistics
	LEV_6	Leverage value
	ZRE_6	Normalized residual
	DFB0_6	DFBETA for constant
	DFB1_6	DFBETA for pretest
	DFB2_6	DFBETA for ascgr(1)
	DFB3_6	DFBETA for adj096(1)
	DFB4_6	DFBETA for techsex(1)
	DFB5_6	DFBETA for amisone(1)

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary					
Unweighted Case	s <sup>a</sup>	Ν	Percent		
Selected Cases	Included in Analysis	223	88.5		
	Missing Cases	29	11.5		
	Total	252	100.0		
Unselected Cases	3	0	.0		
Total		252	100.0		

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

### Categorical Variables Codings

			Parameter coding
		Frequency	(1)
amisone	0	140	.000
	1	83	1.000
adj096	0	119	.000
	1	104	1.000
techsex	0	148	.000
	1	75	1.000
ascgr	0	64	.000
	1	159	1.000

# Block 0: Beginning Block

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	300.801	.386		
	2	300.800	.391		
	3	300.800	.391		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 300.800

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Observed	Predicted					
	fnl	grd				
			Percentage			
	0	1	Correct			
Step 0 fnlgrd 0	0	90	.0			
1	0	133	100.0			
Overall Percentage			59.6			

#### Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.391	.136	8.187	1	.004	1.478

Valiables not in the Equation					
			Score	df	Sig.
Step 0	Variables	pretest	16.870	1	.000
		ascgr(1)	44.750	1	.000
		adj096(1)	9.014	1	.003
		techsex(1)	12.564	1	.000
		amisone(1)	5.763	1	.016
	Overall Stat	istics	83.477	5	.000

Variables not in the Equation

## **Block 1: Method = Enter**

Iteration			Coefficients			
		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)
Step 1 1		209.635	-2.696	.029	2.062	.788
2	2	198.718	-4.015	.046	2.848	1.250
3	3	197.729	-4.552	.052	3.178	1.445
4	Ļ	197.716	-4.622	.053	3.222	1.470
5	5	197.716	-4.623	.053	3.223	1.471
6	6	197.716	-4.623	.053	3.223	1.471

Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 300.800

d. Estimation terminated at iteration number 6 because parameter estimates changed by less than .001.

# Iteration History<sup>a,b,c,d</sup>

Iteration		Coefficients	
		techsex(1)	amisone(1)
Step 1	1	.665	452
	2	1.051	736
	3	1.209	856
	4	1.229	871
	5	1.229	871
	6	1.229	871

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 300.800

d. Estimation terminated at iteration number 6

because parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	103.084	5	.000
	Block	103.084	5	.000
	Model	103.084	5	.000

wodel Summary	Model	Summary
---------------	-------	---------

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	197.716 <sup>a</sup>	.370	.500

a. Estimation terminated at iteration number 6 because parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	8.734	8	.365

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	19	20.760	3	1.240	22
	2	19	18.261	3	3.739	22
	3	17	14.460	5	7.540	22
	4	12	11.494	10	10.506	22
	5	8	9.082	14	12.918	22
	6	6	7.262	17	15.738	23
	7	3	4.517	21	19.483	24
	8	5	2.461	17	19.539	22
	9	1	1.366	23	22.634	24
	10	0	.336	20	19.664	20

Contingency Table for Hosmer and Lemeshow Test

#### Classification Table<sup>a</sup>

Observed		Predicted			
			fnlgrd		
				-	Percentage
			0	1	Correct
Step 1	fnlgrd	0	66	24	73.3
		1	19	114	85.7
	Overall	Percentage			80.7

a. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.053	.013	16.667	1	.000	1.055
	ascgr(1)	3.223	.493	42.791	1	.000	25.095
	adj096(1)	1.471	.427	11.857	1	.001	4.352
	techsex(1)	1.229	.405	9.211	1	.002	3.417
	amisone(1)	871	.389	5.030	1	.025	.418
	Constant	-4.623	.780	35.121	1	.000	.010

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.053	.013	16.667	1	.000	1.055
	ascgr(1)	3.223	.493	42.791	1	.000	25.095
	adj096(1)	1.471	.427	11.857	1	.001	4.352
	techsex(1)	1.229	.405	9.211	1	.002	3.417
	amisone(1)	871	.389	5.030	1	.025	.418
	Constant	-4.623	.780	35.121	1	.000	.010

Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr, adj096, techsex, amisone.

		95% C.I.for EXP(B)			
		Lower	Upper		
Step 1 <sup>a</sup>	pretest	1.028	1.082		
	ascgr(1)	9.555	65.905		
	adj096(1)	1.884	10.052		
	techsex(1)	1.545	7.556		
	amisone(1)	.195	.896		
	Constant				

#### Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr,

adj096, techsex, amisone.

Step number: 1

**Observed Groups and Predicted Probabilities** 

16 + + T T T F Т T R 12 ++ Е T Q T T U 1 1 111 Е 8+ 1 11 +Ν Τ 11 1111 С 1 11 1 1 1 1 1 1 Y 10 1 1 10 1 11 1 11 1 111 1 1 4 + 0 0 0 0 10 1101 1 1 11 1 1 111 1 1 11+ 100 100 0 100 0 00 1 1 1 0 1 1 1 111 11 1111111 111 111 10000001000000001001110011011111 111 111 01011111111 111 I 000000 0010000 001 0 00 000 00 0 000 0 0 0 0 1 0100011 00110 01001 1 00 101 110 01001101111111 -+-----Prob: 0 1 .1 .2 .3 .5 .6 .7 .8 .9 .4 Group: 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 1 Case.

Case						
		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.106	0	.894	2.904
44	S	0**	.924	1	924	-3.481
51	S	0**	.896	1	896	-2.940
59	S	0**	.869	1	869	-2.573
67	S	1**	.110	0	.890	2.837
77	S	1**	.140	0	.860	2.483
145	S	0**	.880	1	880	-2.711
147	S	1**	.092	0	.908	3.150
168	S	0**	.888	1	888	-2.809
238	S	0**	.866	1	866	-2.541

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER pretest ascgr adj096 techsex amisone act /CONTRAST (ascgr)=Indicator(1) /CONTRAST (adj096)=Indicator(1) /CONTRAST (techsex)=Indicator(1) /CONTRAST (amisone)=Indicator(1) /CONTRAST (amisone)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

	Notes	
Output Created		20-Oct-2011 18:20:21
Comments		
Input	Data	C:\Users\Lin\Documents\math 096 fall
		2001 with classrooms.sav
	Active Dataset	DataSet1

	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	252
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER pretest ascgr
		adj096 techsex amisone act
		/CONTRAST (ascgr)=Indicator(1)
		/CONTRAST (adj096)=Indicator(1)
		/CONTRAST (techsex)=Indicator(1) /CONTRAST (amisone)=Indicator(1)
		/SAVE=PRED PGROUP COOK
		LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.046
	Elapsed Time	00:00:00.081
Variables Created or	PRE_7	Predicted probability
Modified	PGR_7	Predicted group
	COO_7	Analog of Cook's influence statistics
	LEV_7	Leverage value
	ZRE_7	Normalized residual
	DFB0_7	DFBETA for constant
	DFB1_7	DFBETA for pretest
	DFB2_7	DFBETA for ascgr(1)
	DFB3_7	DFBETA for adj096(1)
	DFB4_7	DFBETA for techsex(1)
	DFB5_7	DFBETA for amisone(1)
	DFB6_7	DFBETA for act

[DataSet1] C:\Users\Lin\Documents\math 096 fall 2001 with classrooms.sav

Case Processing Summary							
Unweighted Case	N	Percent					
Selected Cases	Included in Analysis	168	66.7				
	Missing Cases	84	33.3				
	Total	252	100.0				
Unselected Cases	3	0	.0				
Total		252	100.0				

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value
0	0
<sup>—</sup> 1	1

#### Categorical Variables Codings

			Parameter coding
		Frequency	(1)
amisone	0	99	.000
	1	69	1.000
adj096	0	97	.000
	1	71	1.000
techsex	0	109	.000
	1	59	1.000
ascgr	0	47	.000
	1	121	1.000

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>						
Iteration			Coefficients			
		-2 Log likelihood	Constant			
Step 0	1	230.008	.262			
	2	230.008	.263			
	3	230.008	.263			

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 230.008

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Observed			Predicted				
			fnl	grd			
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	73	.0		
		1	0	95	100.0		
	Overall	Percentage			56.5		
		-					

#### Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.263	.156	2.864	1	.091	1.301

		variables not in tr	le Equation		
			Score	df	Sig.
Step 0	Variables	Pretest	10.388	1	.001
		ascgr(1)	29.173	1	.000
		adj096(1)	7.778	1	.005
		techsex(1)	14.397	1	.000
		amisone(1)	4.930	1	.026
		Act	4.981	1	.026
	Overall Stat	istics	61.968	6	.000

### Variables not in the Equation

## Block 1: Method = Enter

			Iteration	metery				
Iteration	ration			Coefficients				
		-2 Log likelihood	Constant	pretest	ascgr(1)	adj096(1)	techsex(1)	
Step 1	1	161.240	-5.689	.019	1.935	.946	.796	
	2	153.535	-8.402	.030	2.694	1.455	1.165	
	3	152.883	-9.455	.034	3.001	1.668	1.296	
	4	152.876	-9.579	.035	3.039	1.696	1.310	
	5	152.876	-9.580	.035	3.040	1.696	1.311	
	6	152.876	-9.580	.035	3.040	1.696	1.311	

### Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 230.008

d. Estimation terminated at iteration number 6 because parameter estimates changed by less than .001.

# Iteration History<sup>a,b,c,d</sup>

Iteration		Coefficients		
		amisone(1)	act	
Step 1	1	468	.227	
	2	761	.341	
	3	883	.385	
	4	898	.391	
	5	898	.391	
	6	898	.391	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 230.008

d. Estimation terminated at iteration number 6

because parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	77.132	6	.000
	Block	77.132	6	.000
	Model	77.132	6	.000

Model Summary	N	lo	de	I S	um	ma	ry
---------------	---	----	----	-----	----	----	----

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	152.876 <sup>a</sup>	.368	.494

a. Estimation terminated at iteration number 6 because parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	3.044	8	.932

		fnlgro	d = 0	= 0 fnlgrd =		
		Observed	Expected	Observed	Expected	Total
Step 1	1	16	15.939	1	1.061	17
	2	14	14.234	3	2.766	17
	3	13	11.940	4	5.060	17
	4	10	9.745	7	7.255	17
	5	5	7.690	12	9.310	17
	6	7	6.464	11	11.536	18
	7	5	3.645	12	13.355	17
	8	2	2.212	15	14.788	17
	9	1	.911	16	16.089	17
	10	0	.220	14	13.780	14

Contingency Table for Hosmer and Lemeshow Test

#### Classification Table<sup>a</sup>

Observed			Predicted			
					~	
			fnl	grd		
					Percentage	
			0	1	Correct	
Step 1	fnlgrd	0	54	19	74.0	
		1	17	78	82.1	
	Overall	Percentage			78.6	

a. The cut value is .500

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	pretest	.035	.015	5.351	1	.021	1.035
	ascgr(1)	3.040	.571	28.334	1	.000	20.904
	adj096(1)	1.696	.516	10.786	1	.001	5.452
	techsex(1)	1.311	.450	8.493	1	.004	3.708
	amisone(1)	898	.441	4.146	1	.042	.407
	act	.391	.205	3.629	1	.057	1.478
	Constant	-9.580	2.996	10.224	1	.001	.000

#### Variables in the Equation

a. Variable(s) entered on step 1: pretest, ascgr, adj096, techsex, amisone, act.

	variables in t		
		95% C.I.fc	or EXP(B)
		Lower	Upper
Step 1 <sup>a</sup>	pretest	1.005	1.066
	ascgr(1)	6.825	64.024
	adj096(1)	1.981	15.002
	techsex(1)	1.536	8.953
	amisone(1)	.172	.967
	act	.989	2.209
	Constant		

Variables	in the	Equation
V GI I GI I GI I GI GI GI GI GI GI GI GI		Equation

a. Variable(s) entered on step 1: pretest, ascgr,

adj096, techsex, amisone, act.

Step number: 1

Observed Groups and Predicted Probabilities

16 + + T T I F L T R 12 ++ Е I Q T T U T 11 Е 8+ 1+ Ν 11 Т 1 С 1 1 1 Т 1 Y L 1 1 1 1 1 4 + 01 0 1 1 1 111 +1 100 1 10 0 111 1 1 1 1 11 1 11 1 11 11 I 100110 111 1 1 111110111 I 00 0 0 0 0 0 1 0 1011 111 1111 11 11111 011 10011 0 011 00111111011111 -+-----Prob: 0 .1 .2 .3 .4 .5 .6 .7 .8 .9 1 Group: Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1

Each Symbol Represents 1 Case.

Case	Selected	Observed			Temporar	y Variable
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
15	S	1**	.128	0	.872	2.612
44	S	0**	.878	1	878	-2.683
51	S	0**	.952	1	952	-4.450
80	S	0**	.887	1	887	-2.796
147	S	1**	.073	0	.927	3.555

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

# Appendix X: SPSS Multiple Binary Logistic Regression Output for

Intermediate Algebra

### GET

FILE='C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav'.

DATASET NAME DataSet1 WINDOW=FRONT. LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER gender act comcol pretest ascgr /CONTRAST (gender)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

## **Logistic Regression**

Notes						
Output Created		20-Oct-2011 19:01:41				
Comments						
Input	Data	C:\Users\Lin\Documents\math 097 fall				
		2001 with classrooms no names.sav				
	Active Dataset	DataSet1				
	Filter	<none></none>				
	Weight	<none></none>				
	Split File	<none></none>				
	N of Rows in Working Data	675				
	File					
Missing Value Handling	Definition of Missing	User-defined missing values are				
		treated as missing				

Syntax		LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER gender act comcol pretest ascgr /CONTRAST (gender)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (comcol)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /CLASSPLOT /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.063
	Elapsed Time	00:00:00.053
Variables Created or	PRE_1	Predicted probability
Modified	PGR_1	Predicted group
	COO_1	Analog of Cook's influence statistics
	LEV_1	Leverage value
	ZRE_1	Normalized residual
	DFB0_1	DFBETA for constant
	DFB1_1	DFBETA for gender(1)
	DFB2_1	DFBETA for act
	DFB3_1	DFBETA for comcol(1)
	DFB4_1	DFBETA for pretest
	DFB5_1	DFBETA for ascgr(1)

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary						
Unweighted Cases	a S	N	Percent			
Selected Cases	Included in Analysis	474	70.2			
	Missing Cases	201	29.8			
	Total	675	100.0			
Unselected Cases		0	.0			

Total

675 100.0

a. If weight is in effect, see classification table for the total number of cases.

### Dependent Variable Encoding

Original Value	Internal Value
0	0
1	1

Categorical Variables Codings					
			Parameter coding		
		Frequency	(1)		
Ascgr	0	120	.000		
	1	354	1.000		
Comcol	0	374	.000		
	1	100	1.000		
Gender	0	279	.000		
	1	195	1.000		

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	591.845	.734		
	2	591.712	.770		
	3	591.712	.770		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 591.712

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

	Observed		Predicte	Ч
		fnlgrd		u
		0	1	Percentage Correct
Step 0	fnlgrd 0	0	150	.(
	1	0	324	100.0
	Overall Percentage			68.4

a. Constant is included in the model.

b. The cut value is .500

Variables in the Equation							
			0 -			0.	
		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.770	.099	60.808	1	.000	2.160

Variables not in the Equation						
			Score	df	Sig.	
Step 0	Variables	gender(1)	10.690	1	.001	
		Act	20.457	1	.000	
		comcol(1)	10.449	1	.001	
		Pretest	22.948	1	.000	
		ascgr(1)	129.090	1	.000	
	Overall Sta	tistics	153.149	5	.000	

## **Block 1: Method = Enter**

Iteration History								
Iteration Coefficients			cients					
		-2 Log likelihood	Constant	gender(1)	act	comcol(1)	pretest	ascgr(1)
Step 1	1	444.500	-4.300	245	.177	403	.015	2.064
	2	434.615	-6.244	385	.269	572	.024	2.453
	3	434.338	-6.652	418	.288	603	.027	2.524
	4	434.338	-6.666	419	.288	604	.027	2.527
	5	434.338	-6.666	419	.288	604	.027	2.527

#### Iteration History<sup>a,b,c,d</sup>

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 591.712

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

#### Omnibus Tests of Model Coefficients

		Chi-square	df	Sig.
Step 1	Step	157.374	5	.000
	Block	157.374	5	.000

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	157.374	5	.000
	Block	157.374	5	.000
	Model	157.374	5	.000

#### **Model Summary**

Step		Cox & Snell R	Nagelkerke R
	-2 Log likelihood	Square	Square
1	434.338 <sup>a</sup>	.283	.396

a. Estimation terminated at iteration number 5 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	df	Sig.
1	10.358	8	.241

Contingency rable for Hostiler and Lemeshow Test							
		fnlgro	d = 0	fnlgrd = 1			
		Observed	Expected	Observed	Expected	Total	
Step 1	1	37	40.664	11	7.336	48	
	2	35	34.185	13	13.815	48	
	3	21	23.310	26	23.690	47	
	4	19	13.743	28	33.257	47	
	5	12	10.363	35	36.637	47	
	6	10	8.209	37	38.791	47	
	7	5	6.721	42	40.279	47	
	8	7	5.549	41	42.451	48	
	9	4	4.397	45	44.603	49	
	10	0	2.858	46	43.142	46	

#### Contingency Table for Hosmer and Lemeshow Test

	Classification Table <sup>a</sup>							
	Observe	ed	Predicted					
			fnle	grd				
					Percentage			
			0	1	Correct			
Step 1	fnlgrd	0	87	63	58.0			
		1	33	291	89.8			
	Overall	Percentage			79.7			

a. The cut value is .500

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	gender(1)	419	.243	2.971	1	.085	.658
	act	.288	.112	6.682	1	.010	1.334
	comcol(1)	604	.291	4.313	1	.038	.547
	pretest	.027	.009	8.608	1	.003	1.027
	ascgr(1)	2.527	.263	92.191	1	.000	12.512
	Constant	-6.666	1.874	12.653	1	.000	.001

#### Variables in the Equation

a. Variable(s) entered on step 1: gender, act, comcol, pretest, ascgr.

variables in the Equation							
		95% C.I.for EXP(B)					
		Lower	Upper				
Step 1 <sup>a</sup>	gender(1)	.409	1.059				
	act	1.072	1.660				
	comcol(1)	.309	.967				
	pretest	1.009	1.045				
	ascgr(1)	7.470	20.956				
	Constant						

#### Variables in the Equation

a. Variable(s) entered on step 1: gender, act,

comcol, pretest, ascgr.

Step number: 1

Observed Groups and Predicted Probabilities

F R E Q U	32 +     24 +   								1	1 11 11 1 1 11 1 1 1 11 1	+     +   11	1
Е	16 +									11111	1111	1 +
Ν	I								1	1 1 11 1	111111	I
С	I			1					1 1	1 1 111	1 1111	11 I
Y				0					1 1	1 1 111	1 1111 <sup>-</sup>	11 I
	8 +	1	1 1		0				1 11	1 111 1	111111	111111
+ 111	I 11111111	101 11 I	000	)	0				1111	10 111 <sup>-</sup>	1	
	(	00 00	0100	001	1	1 10	01	1	1 11	1 11 0	111100	11110
100	10101111	111 I										
	I 0 000	000000	00000	00000	001	0000	000 C	00 00	00 01	10 001	10	
001	110000 00	000000	00000	01000	000	00000	0111	1				
Pre	dicted	+	+-		+	+		+	+-	+		+
-+ Pr	rob: 0	.1	.2	.3	.4		.5	.6	.7	.8	.9	1
G	roup:											
	00000000000	000000	00000	00000	000	00000	0000	0000	00000	000111	111111	11111
111	11111111	111111	11111	1111	111	11111	111					

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 2 Cases.

Case		Observed			Temporar	y Variable
	Selected					
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
32	s	0**	.877	1	877	-2.666
40	S	0**	.877	1	877	-2.666
204	S	0**	.903	1	903	-3.050
219	S	0**	.922	1	922	-3.441
232	S	1**	.130	0	.870	2.584
241	S	1**	.139	0	.861	2.493
292	S	0**	.914	1	914	-3.255
545	S	0**	.890	1	890	-2.850
595	S	0**	.905	1	905	-3.080
606	S	0**	.888.	1	888	-2.822
614	S	0**	.890	1	890	-2.850
619	S	0**	.888.	1	888	-2.822
635	S	0**	.886	1	886	-2.787

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

LOGISTIC REGRESSION VARIABLES fnlgrd /METHOD=ENTER act comcol pretest ascgr /CONTRAST (comcol)=Indicator(1) /CONTRAST (ascgr)=Indicator(1) /SAVE=PRED PGROUP COOK LEVER DFBETA ZRESID /CLASSPLOT /CASEWISE OUTLIER(2) /PRINT=GOODFIT ITER(1) CI(95) /CRITERIA=PIN(0.05) POUT(0.10) ITERATE(20) CUT(0.5).

# Logistic Regression

	Notes	
Output Created		20-Oct-2011 19:02:47
Comments		
Input	Data	C:\Users\Lin\Documents\math 097 fall
		2001 with classrooms no names.sav
	Active Dataset	DataSet1
	Filter	<none></none>
	Weight	<none></none>
	Split File	<none></none>
	N of Rows in Working Data	675
	File	
Missing Value Handling	Definition of Missing	User-defined missing values are
		treated as missing
Syntax		LOGISTIC REGRESSION VARIABLES
		fnlgrd
		/METHOD=ENTER act comcol
		pretest ascgr
		/CONTRAST (comcol)=Indicator(1)
		/CONTRAST (ascgr)=Indicator(1) /SAVE=PRED PGROUP COOK
		LEVER DFBETA ZRESID
		/CLASSPLOT
		/CASEWISE OUTLIER(2)
		/PRINT=GOODFIT ITER(1) CI(95)
		/CRITERIA=PIN(0.05) POUT(0.10)
		ITERATE(20) CUT(0.5).
Resources	Processor Time	00:00:00.078
	Elapsed Time	00:00:00.051
Variables Created or	PRE_2	Predicted probability
Modified	PGR_2	Predicted group
	COO_2	Analog of Cook's influence statistics
	LEV_2	Leverage value
	_ ZRE_2	Normalized residual
	DFB0 2	DFBETA for constant
	DFB1_2	DFBETA for act
	—	
	DFB2_2	DFBETA for comcol(1)

DFB3_2	DFBETA for pretest
DFB4_2	DFBETA for ascgr(1)

[DataSet1] C:\Users\Lin\Documents\math 097 fall 2001 with classrooms no names.sav

Case Processing Summary							
Unweighted Cases	N	Percent					
Selected Cases	Included in Analysis	485	71.9				
	Missing Cases	190	28.1				
	Total	675	100.0				
Unselected Cases	i	0	.0				
Total		675	100.0				

a. If weight is in effect, see classification table for the total number of cases.

#### Dependent Variable Encoding

Original Value	Internal Value	
0	0	
<sup>—</sup> 1	1	

#### **Categorical Variables Codings**

			Parameter coding	
		Frequency	(1)	
ascgr	0	124	.000	
	1	361	1.000	
comcol	0	383	.000	
	1	102	1.000	

# **Block 0: Beginning Block**

Iteration History <sup>a,b,c</sup>						
Iteration			Coefficients			
		-2 Log likelihood	Constant			
Step 0	1	604.841	.738			
	2	604.701	.774			
	3	604.701	.775			

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 604.701

Iteration History <sup>a,b,c</sup>					
Iteration			Coefficients		
		-2 Log likelihood	Constant		
Step 0	1	604.841	.738		
	2	604.701	.774		
	3	604.701	.775		

a. Constant is included in the model.

b. Initial -2 Log Likelihood: 604.701

c. Estimation terminated at iteration number 3

because parameter estimates changed by less than .001.

Observed			Predicted				
			fnlgrd				
					Percentage		
			0	1	Correct		
Step 0	fnlgrd	0	0	153	.0		
		1	0	332	100.0		
	Overall	Percentage			68.5		

#### Classification Table<sup>a,b</sup>

a. Constant is included in the model.

b. The cut value is .500

#### Variables in the Equation

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 0	Constant	.775	.098	62.857	1	.000	2.170

#### Variables not in the Equation

			Score	df	Sig.
Step 0	Variables	Act	22.940	1	.000
		comcol(1)	12.631	1	.000
		Pretest	24.676	1	.000
	-	ascgr(1)	135.054	1	.000

			Score	df	Sig.
Step 0	Variables	Act	22.940	1	.000
		comcol(1)	12.631	1	.000
		Pretest	24.676	1	.000
		ascgr(1)	135.054	1	.000
	Overall Stat	istics	158.707	4	.000

Variables not in the Equation

# Block 1: Method = Enter

# Iteration History<sup>a,b,c,d</sup>

Iteration			Coefficients					
-2 Log likeliho		-2 Log likelihood	Constant	act	comcol(1)	pretest	ascgr(1)	
Step 1	1	452.505	-4.455	.179	458	.016	2.099	
	2	442.516	-6.496	.272	647	.025	2.504	
	3	442.251	-6.920	.291	680	.027	2.576	
	4	442.251	-6.935	.292	681	.027	2.579	
	5	442.251	-6.935	.292	681	.027	2.579	

a. Method: Enter

b. Constant is included in the model.

c. Initial -2 Log Likelihood: 604.701

d. Estimation terminated at iteration number 5 because parameter estimates changed by less than .001.

#### **Omnibus Tests of Model Coefficients**

		Chi-square	df	Sig.
Step 1	Step	162.450	4	.000
	Block	162.450	4	.000
	Model	162.450	4	.000

Model Summary						
Step		Cox & Snell R	Nagelkerke R			
	-2 Log likelihood	Square	Square			
1	442.251 <sup>a</sup>	.285	.399			

Model Summary						
Step Cox & Snell R Nagelke						
	-2 Log likelihood	Square	Square			
1	442.251 <sup>a</sup>	.285	.399			

a. Estimation terminated at iteration number 5 because

parameter estimates changed by less than .001.

#### Hosmer and Lemeshow Test

Step	Chi-square	Df	Sig.	
1	8.302	8	.405	

		fnlgro	d = 0	fnlgro	d = 1	
		Observed	Expected	Observed	Expected	Total
Step 1	1	41	40.661	7	7.339	48
	2	35	35.243	14	13.757	49
	3	21	24.993	28	24.007	49
	4	17	14.355	33	35.645	50
	5	15	9.754	32	37.246	47
	6	6	8.338	44	41.662	50
	7	5	5.989	38	37.011	43
	8	8	6.432	48	49.568	56
	9	4	4.384	45	44.616	49
	10	1	2.851	43	41.149	44

#### Contingency Table for Hosmer and Lemeshow Test

Classification	Table <sup>a</sup>
----------------	--------------------

	Observed		Predicted			
			fnlgrd		Percentage	
			0	1	Correct	
Step 1	fnlgrd	0	91	62	59.5	
		1	35	297	89.5	
	Overall	Percentage			80.0	

a. The cut value is .500

		В	S.E.	Wald	df	Sig.	Exp(B)
Step 1 <sup>a</sup>	act	.292	.111	6.899	1	.009	1.339
	comcol(1)	681	.288	5.610	1	.018	.506
	pretest	.027	.009	9.222	1	.002	1.028
	ascgr(1)	2.579	.259	98.981	1	.000	13.183
	Constant	-6.935	1.872	13.729	1	.000	.001

Variables in the Equation

a. Variable(s) entered on step 1: act, comcol, pretest, ascgr.

#### Variables in the Equation

		95% C.I.for EXP(B)		
		Lower	Upper	
Step 1 <sup>a</sup>	act	1.077	1.664	
	comcol(1)	.288	.889	
	pretest	1.010	1.046	
	ascgr(1)	7.932	21.911	
	Constant			

a. Variable(s) entered on step 1: act, comcol,

pretest, ascgr.

Step number: 1

**Observed Groups and Predicted Probabilities** 

40 + + T Т I F Т 1 L R 30 +1 1 + Е 1111 Т Q T 11 111 Т U T 111 1111 Т Е 20 + 1 1 1 11 111 + Ν 1 1 1 11 1111 Т Т С 1 Т 1 1 1 1 11 11111 T Y L 1 Т 10 +1 11 1 1 1 11 111111 0 + 1 01 1 1 1 1 1 1 1 1 1 1 0 1 1 Τ 101111111 T Ι 01101111111 0100 0 0101001010111 I -+-----Prob: 0 .9 1 .1 .2 .3 .4 .5 .6 .7 .8 Group: 

Predicted Probability is of Membership for 1 The Cut Value is .50 Symbols: 0 - 0 1 - 1 Each Symbol Represents 2.5 Cases.

Case		Observed			Temporar	y Variable
	Selected	Obscived			Гстрога	y variable
	Status <sup>a</sup>	fnlgrd	Predicted	Predicted Group	Resid	ZResid
20	S	0**	.875	1	875	-2.645
37	S	0**	.877	1	877	-2.670
204	S	0**	.926	1	926	-3.541
219	S	0**	.913	1	913	-3.246
292	S	0**	.904	1	904	-3.061
318	S	0**	.893	1	893	-2.886
493	S	0**	.877	1	877	-2.670
538	S	0**	.875	1	875	-2.645
545	S	0**	.877	1	877	-2.670
594	S	0**	.893	1	893	-2.886
595	S	0**	.893	1	893	-2.886
606	S	0**	.915	1	915	-3.277
614	S	0**	.877	1	877	-2.670
619	S	0**	.915	1	915	-3.277
635	S	0**	.873	1	873	-2.621
658	S	0**	.877	1	877	-2.670

Casewise List<sup>b</sup>

a. S = Selected, U = Unselected cases, and \*\* = Misclassified cases.

b. Cases with studentized residuals greater than 2.000 are listed.

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## Presentations

Alge-blocks Spring 2007 Kentucky Mathematical Association of Two-Year Colleges Joint presentation with Jame McCumbee

Get Your Hands On Developmental Mathematics Fall 2006 Kentucky Association for Developmental Education Joint presentation with Ralfred Hall

Move Over, Math Anxiety Spring 2003 West Virginia Council of Teachers of Mathematics conference Joint presentation with Tom Klein

Math Anxiety: Definitions, Assessments, & Research 2002 Ashland Teaching and Learning Conference Joint presentation with Mildred Battle and Kay Thompson

### Professional Development

New Horizons Conference on Teaching and Learning May 2007, May 2008, May 2010, May 2011

Annual Teaching/Learning Conference, Ashland, KY 1998, 2000, 2001, 2004, 2005, 2006, 2007, 2008, 2009, 2010

## ACCLAIM Conferences

June 19-21, 2002, August 16-17, 2002, & September 17, 2004

Advanced Kellogg Institute, Appalachian State University August 3-8, 2002

NADE National Conference

1997 - 2011

Memberships

2004 – present

Delta Kappa Gamma Kentucky Association for Developmental Education 2006 -Kentucky Math Association of Two-Year Colleges 2007 –