

# What You Do in High School Matters: The Effects of High School GPA on Educational Attainment and Labor Market Earnings in Adulthood

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# **What You Do in High School Matters: The Effects of High School GPA on Educational Attainment and Labor Market Earnings in Adulthood**

## **Abstract**

Using abstracted grades and other data from Add Health, we investigate the effects of cumulative high school GPA on educational attainment and labor market earnings among a sample of young adults (ages 24-34). We estimate several models with an extensive list of control variables and high school fixed effects. Results consistently show that high school GPA is a positive and statistically significant predictor of educational attainment and earnings in adulthood. Moreover, the effects are large and economically important for each gender. Interesting and somewhat unexpected findings emerge for race. Various sensitivity tests support the stability of the core findings.

**JEL Classification:** I2, J24, J31

*Keywords:* High school grades; Educational attainment; Earnings; Panel data

## I. Introduction

Teenagers face numerous and often dramatic changes in physical appearance, emotional status, character, personality, and human capital over the course of high school. From an academic perspective, one needs to do well in high school to gain entry into a reputable college or university or even land a lucrative job. Most people assume, however, that academic performance in high school is less predictive of overall educational attainment and only weakly related to labor market earnings in adulthood. If these assumptions are false, however, then one's academic standing in high school could be an important boost or drag on one's educational attainment and labor market success in adulthood.

Numerous studies have found that higher educational attainment is associated with greater earnings (Mincer, 1974; Card, 1999; Crissey, 2009). Two general mechanisms help to explain this relationship. According to human capital theory, education enhances an individual's skills and ability (Wise, 1975). Beyond skills acquisition, prospective employers use grades and academic performance to differentiate among job candidates (Spence, 1973; Lazear, 1977; Jones and Jackson, 1990). An extensive review by Card (1999) concludes that certain observable characteristics such as race, school quality, family background, and cognitive ability are related to the returns to education.

In contrast to the broad availability of literature on educational *attainment* and earnings, little is known about the association between educational *performance* and earnings. The vast majority of studies in this area focus on whether college grades, college major, and college selectivity affect future earnings (Thomas, 2003; Loury and Garman, 1995; Hamermesh and Donald, 2008; Jones and Jackson 1990; Monks, 2000; Hoekstra, 2009; Zhang, 2008). Although this literature indicates that more demanding majors and higher quality institutions contribute to higher earnings (Zhang, 2008), some studies that control for ability and/or selection have found smaller effects (Dale and Krueger, 2002; Hamermesh and Donald, 2008). Only a small number of studies control for high school

grades (which are usually self reported) when evaluating the relationship between academic factors in college (e.g., GPA, major, or graduation status) and labor market outcomes. Moreover, high school achievement does little to mediate these relationships (Wise, 1975; Grogger and Eide, 1995; Hamermesh and Donald, 2008).

Compared to the research focusing on college performance and related characteristics, very few studies have investigated the impact of high school performance, as measured by cumulative GPA, on future academic and labor market outcomes. Crawford and colleagues (1997) and Bishop and colleagues (1985) examine whether high school GPA influences short-term (three years or less) labor market outcomes for individuals who began working immediately after completing high school. Both find positive effects. Specifically, Crawford and colleagues (1997) estimate that a one-point increase in high school GPA adds \$800-\$1000 to annual earnings after high school. Meyer and Wise (1982) find that high school rank is positively related to weeks worked and wages for men four years after high school graduation. The authors suggest that class rank may reflect work ethic and that those with a strong work ethic in high school may continue to work hard in the labor force. Wolfie and Smith's (1956) descriptive analysis of high school class rank and earnings twenty years after high school graduation indicates that, among individuals with college degrees, those at the top of their high school classes earn higher incomes than those with lower class ranks, but they find no significant differences for individuals who do not attend college.

Most of the other studies examining longer-term labor market outcomes and high school academic performance focus primarily on high school curriculum (Altonji, 1995; Levine and Zimmerman, 1995; Rose and Betts, 2004) or achievement tests (Lleras, 2008) rather than GPA per se. For example, Rose and Betts (2004) investigate whether specific high school math classes (e.g., pre-algebra, algebra/geometry) are associated with earnings ten years after high school graduation. They control for math GPA and school fixed effects and use an instrumental variables approach to

account for unobserved ability and motivation. Math courses in high school exert a large and significant impact on earnings, with greater effects for more advanced courses.

One would expect educational performance in high school to affect ultimate educational attainment in addition to future labor market outcomes. Although high school achievement is a predictor of academic performance in college (Betts and Morell, 1999; Cohn et al., 2004), the relationship between high school GPA and educational attainment for adults has not been fully examined. Studies in which educational attainment is the dependent variable frequently include young samples or college students with incomplete academic careers (Betts and Morell, 1999; Ou and Reynolds, 2008; Melby et al., 2008). Other correlates of total educational attainment include personal characteristics (e.g., race, history of delinquency), previous academic achievement (e.g., standardized test scores, educational expectations, grade retention, school absences), and family background (e.g., maternal education, parental involvement in education, income) (Betts and Morell, 1999; Ou and Reynolds, 2008; Melby et al., 2008; Cameron and Heckman, 2001).

In this paper, we use data from Waves 1, 3, and 4 of the National Longitudinal Survey of Adolescent Health (Add Health) to examine whether cumulative GPA in high school is significantly related to educational attainment and labor market earnings during early adulthood. Our investigation overcomes several important limitations of the existing literature in this area and offers new insights into the complex relationship between the academic performance of teenagers and various outcomes in young adulthood.

First, using GPA data abstracted from high school records rather than subjective measures of student competence or self-reported grades allows us to avoid the potential measurement error problems that plague many previous studies. Only a few studies in this area (notably Crawford et al., 1997; Rose and Betts, 2004) do not use individual self reports of high school performance. Second, Add Health respondents were between the ages of 24 and 34 at Wave 4, when educational

attainment and wages are likely to be well established. Third, some of the studies noted above are limited to men (Loury and Garman, 1995; Hoekstra, 2009; Wise, 1975), Whites (Hoekstra, 2009; Wise, 1975), or individuals with a common level of education (e.g., those who did not attend college or college graduates) (Thomas, 2003; Monks, 2000; Mueller, 1988). Analyzing diverse samples is essential since notable disparities in educational attainment as well as different returns to education based on level of educational attainment exist for different racial, ethnic, and gender groups (U.S. Census Bureau, 2010). Our analysis accounts for seven categories of educational attainment ranging from those who did not finish high school to those with advanced graduate degrees. Finally, we study males and females separately (Levine and Zimmerman, 1995) and control for numerous personal and family background characteristics that could influence the outcomes of interest (Cameron and Heckman, 2001; Loury and Garman, 1995; Lleras, 2008).

## **II. Conceptual Framework**

Adolescence is a time of rapid intellectual, mental, and emotional developments with corresponding physical and social changes (Maggs et al., 1997; Kroger, 2006). These formidable challenges sidetrack many high school students in their academic pursuits, often leading to poor grades and other educational difficulties. If such scholastic lapses are atypical and transitory, then high school achievement plays a role in college admission but not thereafter. Alternatively, high school grades may accurately characterize underlying potential and behavior, thereby serving as a reliable signal of future success in the classroom and workplace.

Identity formation and decisions about whether to conform to or challenge adult conventions also occur during adolescence (Erikson, 1968; Maggs et al., 1997; Kroger, 2006). Nonconformist behaviors and academic achievement are both avenues through which adolescents identify with their peers and rebel against or abide by adult conventions. Based partly on studies in the educational psychology and economics literature (Anderson and Keith, 1997; Heckman, 2008;

Lounsbury et al., 2003; Neisser et al., 1996; Rivkin et al., 2005), we assume that each student is endowed with a certain level of intelligence or ability (proxied by the Peabody Picture Vocabulary Test [PVT] score in our data) as well as an innate degree of social capital. Students then make decisions about combining these initial endowments with a host of variable resources (e.g., time spent studying, selection of friends, participation in extracurricular activities) to achieve certain goals related to their identities. For example, students who desire to attend college are likely to invest a greater amount of their time in studying and developing their human capital. If high school GPA is positively correlated with adult earnings, this suggests that academic achievement early in life is an accurate predictor of future outcomes. Alternatively, a negative association between adult earnings and high school GPA could reflect decisions by students to allocate time to other objectives such as socializing rather than to schoolwork. The analysis that follows cannot definitively resolve these possible mechanisms, but the results offer new insight into a topic that heretofore has been focused almost exclusively on contemporaneous outcomes and predictors.

Another distinct advantage of this paper vis-à-vis most of the published literature is our ability to control for intellectual potential (via the PVT score), adolescent living conditions, parental education, and other important endowment factors. If these variables are missing from the models and correlated with high school GPA, then the effects of high school GPA on years of education and earnings may be overstated. Access to abstracted GPA data as well as these key background characteristics allows us to estimate more fully specified models and generate more precise estimates. Thus, while omitted variable bias remains a distinct possibility, our inclusion of a large list of important control variables should lessen this problem considerably.

Another advantage of Add Health's extensive data is that we can decompose the effects of our key and secondary variables to examine extensions and masked relationships. For example, does the estimated coefficient for high school GPA in the earnings equations drop considerably when we

add PVT score to the model? Do racial and ethnic minorities complete fewer years of education relative to Whites? Does allowing for school fixed effects mediate these relationships? We discuss these and other analyses in the Results section.

Although high school grades are almost certainly endogenous to other outcomes such as grooming, anti-social behavior, and athletic success among teenagers, we don't believe that high school grades are endogenous to educational attainment and labor market earnings among adults. Temporally, high school performance is measured from age 14 to 18 for the vast majority of individuals whereas formal education ceases well into the 20s for most students. Our annual earnings measure pertains to adults who are 24 to 34 years of age. Thus, reverse causality is clearly not a concern, and endogeneity bias is unlikely.

The final conceptual point regarding sample formation concerns possible gender differences. Some studies of academic achievement and labor market outcomes among adults have estimated separate models for men and women (e.g., Mueller, 1988; Levine and Zimmerman, 1995). Other studies (mainly in the educational and psychology literature) have also identified gender differences in academic achievement, with females earning higher grades than their male counterparts (Dwyer and Johnson, 1997; Kleinfeld, 1998). In the Add Health data, overall high school GPA is significantly higher among females whereas males have significantly higher annual earnings at Wave 4. For these reasons, it seems appropriate to analyze males and females separately.

### **III. Methods**

The primary variables of interest in this study are high school GPA (the key explanatory variable), highest level of education attained (the first outcome variable), and personal earnings (the second outcome variable). Cumulative high school GPA, the average of all classes and years during an individual's high school career, is reported on a 4-point scale.<sup>1</sup> Highest level of education at

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<sup>1</sup> The cumulative GPA is calculated based on the number of years a student has course data.



Wave 4 contains seven categories ranging from less than or some high school to an advanced professional degree (e.g., JD, MD, PhD). Personal earnings include all income derived from employment before taxes during the calendar year prior to the Wave 4 interview.<sup>2</sup> Because earnings are highly skewed, we follow the literature by analyzing the natural logarithm of earnings.<sup>3</sup> We estimate the categorical measure of educational attainment using an ordered probit model, and we estimate the log of personal earnings using robust regression, a hybrid form of OLS that down-weights outlier observations (StataCorp, 2009).<sup>4</sup>

Although cumulative high school GPA is the explanatory variable of interest, we estimate five distinct specifications for each dependent variable. The first specification, A, starts with a set of demographic control variables and familial background characteristics, which provides a benchmark set of estimates as all later specifications include this principal set of variables. Specification B adds cumulative high school GPA to determine the full effect of this variable without any other controls for ability/intelligence or current school attendance. Specification C adds the PVT score as a proxy measure of ability/intelligence. We add a dummy variable for currently attending school to Specification D, and the final specification, E, includes school fixed effects.

We can formally write the fully-specified model, E, as:

$$Y_i = \beta_0 + \beta_x X_i + \beta_{gpa} GPA_i + \beta_{pvt} PVT_i + \beta_{as} AS_i + \beta_s S_j + \varepsilon_i \quad (1)$$

where  $Y_i$  is either highest level of education attained (the categorical value) or the logarithm of annual personal earnings for individual  $i$ ,  $X_i$  is a vector of demographic and familial variables,  $GPA_i$  is cumulative grade point average in high school,  $PVT_i$  is the score on the Peabody Picture Vocabulary Test,  $AS_i$  is an indicator variable for currently attending school,  $S_j$  is a vector of indicator

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<sup>2</sup> Respondents were asked to report how much income they received from personal earnings before taxes (i.e., wages or salaries, including tips, bonuses, and overtime pay, and income from self employment). The question refers to personal income earned in the calendar year prior to the interview year.

<sup>3</sup> For non-earners, we set earnings to \$1, so the natural logarithm of earnings is zero in this case.

<sup>4</sup> As a sensitivity test, we estimate the log of personal earnings using OLS. We present these results later in the paper.

variables for the schools that participated in the Add Health project, and  $\varepsilon_i$  is a random error term. We estimate separate models for males and females. The Add Health survey provides sampling weights for a limited number of respondents, but we do not use them, choosing instead to control directly for a number of variables related to the sampling distribution.<sup>5</sup>

After presenting results from our basic models, we perform numerous sensitivity tests to examine the stability of the core findings. We re-estimate Equation (1) using an alternative education measure, different exclusion criteria, different estimation techniques, and additional control variables (some available only for a smaller sample). We also estimate several additional models with a much more limited set of control variables to better understand the potential mechanisms associated with variation in education and earnings. We discuss these specifications in greater detail in the Results section.

#### IV. Data

The analysis uses several waves of data from Add Health, a school-based, longitudinal study of adolescent health-related behaviors and their consequences in young adulthood. Wave 1 was administered during 1994-1995 and included in-home interviews with 20,745 adolescents sampled from 80 high schools and 52 middle schools. The study design ensures that the sample is representative of U.S. schools based on region, school type, size, and ethnicity. In-home interviews took one to two hours to complete and were administered as Computer-Assisted Personal Interviews (CAPI)/Audio Computer-Assisted Self Interviews (CASI). In 2001-2002, 15,170 respondents were re-interviewed in Wave 3 when they were 18 to 27 years old. Wave 2 (administered approximately one year after Wave 1) only included adolescents from Wave 1 who were still attending high school while Wave 3 conducted follow-up interviews with all Wave 1 respondents who could be contacted. Wave 4 is the most recently available data in Add Health as it

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<sup>5</sup> Although the sample is smaller, the results are qualitatively similar when we use the sampling weights. Significance levels change somewhat for a few of the variables (results available on request from the authors).

was completed in 2009. At the time of the Wave 4 data collection, subjects were between the ages of 24 and 34, with a mean of 28 years. Thus, the average respondent was approximately 10 years removed from his or her high school experience.

High school transcripts were requested and abstracted for approximately 80 percent of Wave 3 respondents. The most common reason for missing GPA data was difficulty in obtaining records from various high schools. A careful investigation of the missing cases reveals that they have characteristics (e.g., lower income and PVT scores, less parental education) that are typically associated with lower GPAs. We control for these characteristics in our empirical models that explain educational achievement and earnings. In addition, we perform a sensitivity test using imputed GPA data for about 19 percent of the 13,034 respondents who were interviewed at Waves 1, 3, and 4. We discuss these results later in the paper.

The Add Health data have many desirable features pertinent to our study, the most notable being an extensive list of background characteristics and official records of high school grades. However, the education and earnings measures have some limitations. Specifically, highest level of education is reported in categories rather than years, so we must employ a categorical estimation technique (e.g., ordered probit model) instead of OLS. In addition, the timing of the question about annual personal earnings does not coincide with the questions about specific job characteristics or numbers of hours worked. Therefore, we are unable to construct an hourly wage measure or control for job conditions using the information about personal earnings.

We present descriptive statistics for all variables used in the analysis in Table 1. The cumulative high school GPA for this sample was slightly above 2.5, and the average GPA for females (2.717) was significantly higher than that for males (2.436). Mean personal earnings for the full sample was \$36,022, with nearly a \$13,000 gap between males and females. Among the seven educational categories in Add Health, some college was the modal category for both genders, and

advanced professional degree was the least common category. On average, PVT scores for male students were significantly higher than those for female students. In addition, several socioeconomic, demographic, and familial variables displayed significant gender differences.

## V. Results

### A. Educational Attainment

We present the estimation results for educational attainment in Tables 2A for males and 2B for females. For males, high school GPA is positive and significant in all the specifications, with a magnitude of approximately 0.90, implying that an additional point in high school GPA is associated with close to a one-category increase in educational attainment.<sup>6</sup> Interestingly, this estimate is robust and stable when we sequentially add PVT score, currently attending school, and school fixed effects to the model. As expected, both PVT score and a dummy variable for currently attending school are positive and highly significant.

Some intriguing results emerge for the other explanatory variables in Table 2A, particularly when we examine changes across specifications. Compared to being White, being African American is associated with lower educational attainment among males in the benchmark specification (Model A). However, this estimate turns positive and statistically significant when we add high school GPA to the model, and it becomes even larger in magnitude when we enter PVT score and currently attending school. Thus, failing to control for innate ability and academic performance in high school would lead one to incorrectly conclude that African American males complete fewer years of education than their White counterparts. However, controlling for other personal and familial factors indicates that the opposite is true: African American males complete about one quarter of an educational category more than Whites.

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<sup>6</sup> See Table 1 for a description of the seven educational categories.

The pattern for Hispanic males is similar to that for African American males in that the estimate changes from negative and significant in the benchmark specification to positive and significant in the augmented models, but it loses significance in the fully-specified model with school fixed effects (Model E). The same is true for being born in the U.S., which is negative and significant until it also loses significance in Model E. The implication here is that high school-specific factors (e.g., teacher credentials, honors courses, remedial courses, extracurricular activities, class sizes) are significantly correlated with Hispanic ethnicity and place of birth, thereby creating bias in these estimates when school fixed effects are omitted from the model.

Living in a single-parent (negative) or other non-intact household (negative) during high school, having a residential mother (positive), having a father in a white-collar occupation (positive), and having a residential parent who attended college (positive) have the expected signs and all are statistically significant in Model E.

Many of the estimates for females (Table 2B) are similar to those for males. High school GPA is positive and significant in all specifications, with a magnitude of 0.96 (Model E). PVT score and currently attending school are also positive and significant. Only three differences vis-à-vis males appear with the control variables. First, the coefficient estimate for being African American is *positive* and significant in the benchmark specification (Model A) and becomes much larger in successive models. Second, the number of children under age 18 in the household at Wave 1 is negative and significant in Model E, suggesting that teenage girls may help to care for younger siblings, thus impacting their educational progress. Third, receiving welfare in Wave 1 is negative and significant in the fully specified model, indicating that household financial hardship impedes academic achievement for girls but not for boys.

To further understand the effect of high school GPA on educational achievement for both genders, we calculate the marginal effects for each of the educational categories (Table 3). For boys,

an additional point in cumulative high school GPA has a positive and significant effect on the probability of attending college and completing all types of subsequent degrees. The turning point for girls occurs at one category higher—completing a college degree—which is also the category with the largest effects for both genders. Although the estimates are statistically significant, high school GPA has a relatively small effect on the probability of completing a terminal degree. In other words, academic performance in high school has a large and significant effect on educational milestones shortly thereafter but not on the most exclusive degrees.

### B. Annual Personal Earnings

We use robust regression to estimate the log of annual personal earnings for males (Table 4A) and females (Table 4B). The model specifications for earnings are similar to those used for educational achievement, with a benchmark specification (Model A) followed by sequentially-added variables to arrive at a fully-specified model with school fixed effects (Model E). For men, cumulative high school GPA has a positive and significant effect on annual earnings as an adult. Using the estimate in Column E to examine the magnitude of the effect, a one-unit increase in high school GPA leads to a 12.2 percent ( $e^{.115} - 1$ ) increase in annual earnings. With mean annual earnings for males equaling \$42,930, this translates into \$5,237 in extra earnings, *ceteris paribus*. Moreover, the effect is stable and robust across specifications. The PVT score is small and not statistically significant, but currently attending school has a negative and statistically significant effect on annual earnings.

Among the control variables, we include dummy variables for the educational categories described earlier. Relative to being a high school dropout, all of the educational categories are

positive and significant, with effects increasing as one completes more education. Age is also positive and significant in all specifications.<sup>7</sup>

The African-American effect for earnings is qualitatively different than it is for educational attainment. The estimate is initially negative and significant in the benchmark specification (Model A) and remains essentially unchanged through Model E. Relative to White males, African-American males have 21.7 percent lower earnings at Wave 4. The finding contradicts the African-American effect on educational attainment and might illustrate racial discrimination in labor market compensation. We make this statement cautiously, however, because important and unobserved individual-specific factors could significantly mediate the estimated effects.

Most of the other variables in Table 4A are not statistically significant at the 5% level or lower. The exceptions are having a residential mother at Wave 1 (negative), a mother working in a white-collar occupation (positive), and an employed father (positive).

Table 4B shows that the effect for high school GPA is larger for females. With a point estimate of 0.132, this implies that girls who raise their high school GPAs by one unit can expect to receive 14.1 percent ( $e^{.132} - 1$ ) higher earnings as adults. Because mean earnings for women (\$30,153) are lower than the mean for men (\$42,930) at Wave 4, the estimated effect converted to dollars is not as great (\$4,252). As with men, PVT score is not significant, and currently attending school has a negative and significant effect on current earnings. Similarly, age and education are positive and significant in all specifications.

In stark contrast to the results for men, African-American women do not have significantly different earnings than their White counterparts for all models other than the baseline specification. A combined group of other races as well as individuals of Hispanic ethnicity show a positive and

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<sup>7</sup> Traditional earnings models typically include age-squared as well as age to allow for eventual depreciation of human capital. However, the oldest person in our sample is 34 and unlikely to be experiencing depreciating human capital.

significant effect on earnings except in the model with school fixed effects. Here, both estimates are still positive, but neither is statistically significant. The only other variable that is statistically significant for women in Model E is a dummy variable for living in a single-parent household at Wave 1.

### C. Extensions and Sensitivity Analyses

The core findings clearly demonstrate that high school GPA has a positive, statistically significant, and economically meaningful effect on educational attainment and future earnings. But what are the predicted probabilities of meeting various educational thresholds given a particular high school GPA? Appendix A presents these estimates. For both genders, having a median GPA in high school leads to a nearly 0.50 probability of completing some college coursework, which is the modal category. Advancing to the 75<sup>th</sup> percentile (3.058 GPA for boys and 3.333 for girls) results in higher predicted probabilities of graduating from college and attending graduate school. This trend accelerates through the 99<sup>th</sup> percentile of high school GPAs, as the majority of these high-achieving students are likely to graduate from college and/or pursue an advanced degree (74.7 percent for boys and 79.0 percent for girls). Interestingly, based on high school GPA, the top one percent of girls have a higher predicted probability of attending graduate school and earning a terminal degree than the top one percent of boys.

Recall that the African-American coefficient was negative and significant for men in the educational attainment model with no controls for high school GPA, PVT score, or currently attending school (Model A, Table 2A). The estimate then became positive and significant in the fully specified model (Model E, Table 2A). Although the African-American coefficient was positive and significant for women in the parsimonious model (Model A, Table 2B), it became much larger and more significant after we included other controls and school fixed effects (Model E, Table 2B).



Does this imply that we may misestimate the effect of being African American on educational attainment if we look at simple correlations or include a limited number of control variables? To explore the mechanism underlying the African-American effect for educational attainment and earnings, we estimate a simple ordered probit model of educational attainment, age, race, ethnicity, and being born in the U.S. The coefficient estimates are negative and significant for both genders and much larger in absolute value for men relative to Model A in Table 2A. We find similar qualitative results when including the same explanatory variables in a model for earnings. Again, these results underscore the importance of estimating fully-specified models when investigating the effects of race on educational attainment and labor market earnings (Rose and Betts, 2004; Lleras, 2008).<sup>8</sup>

We conduct a number of sensitivity tests to evaluate the robustness of the core results. Appendix B reports on two of these sensitivity tests. First, we replace the categorical dummies for educational attainment with a continuous measure for years of education in the earnings regressions (A). We construct the continuous measure by either using the typical number of years for a completed degree or taking the midpoint for a multi-year category.<sup>9</sup> After we make this change in the specifications for men and women, the coefficient estimates for high school GPA remain positive and statistically significant, with slightly higher values (see Tables 4A and 4B). As expected, years of education is positive and highly significant. All other estimates are virtually identical to those reported earlier.

Next, rather than estimate the categorical measure of educational attainment with ordered probit, we estimate the newly constructed continuous measure of years of education with OLS

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<sup>8</sup> Lleras (2008) evaluated whether high school achievement tests affect educational attainment and earnings ten years later. After controlling for cognitive and non-cognitive abilities, being African American and Hispanic were positively associated with educational attainment but negatively associated with earnings.

<sup>9</sup>We assign the value of 10 to those individuals with less than a high school degree, 12 for a high school degree, 13 for vocational/technical training after high school, 14 for some college, 16 for a college degree, 18 for graduate school, and 20 for an advanced professional degree.

regression (B). Cumulative high school GPA is positive and significant once again. We estimate that raising high school GPA by one point increases educational attainment by 1.322 years for men and 1.427 years for women. The slightly higher estimate for women is consistent with the predicted probabilities reported in Appendix A. Again, the estimates for the other explanatory variables are consistent in sign and significance with those in Tables 2A and 2B.

An additional sensitivity test addresses the absence of GPA data for 19 percent of the sample interviewed at Waves 1, 3, and 4. The approach we adopt for estimation of the core models is to drop these observations. As noted earlier, however, individuals with missing GPA data are, on average, demographically dissimilar from the analysis sample. Thus, we impute GPA data for these individuals using the multiple imputation routine (*impute* command) in Stata with all the Wave 1 control variables listed in Table 1 and some school and regional characteristics (e.g., whether the high school was in an urban area). We then re-estimate the educational attainment and earnings models for each gender with the augmented sample. The coefficient estimates for GPA, which now includes both actual and imputed values, are slightly smaller than our core results but identical in sign and statistical significance.<sup>10</sup> This test at least partially confirms that missing GPA data is not seriously biasing our core results.

We conduct three additional robustness checks for the personal earnings equation. First, we re-estimate the personal earnings equations with OLS regression instead of robust regression, the technique used in the core models to reduce bias from outliers. The effects of cumulative overall high school GPA on personal earnings is larger in magnitude and highly significant ( $p < 0.01$ ) when we use OLS. Second, we control for the age at which respondents first began working full time to evaluate whether labor market experience alters the relationship between high school GPA and personal earnings. The effects of high school GPA on personal earnings are similar in sign,

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<sup>10</sup> Results available on request.

magnitude, and significance for men and women after we control for labor market experience. Finally, we control for three post-secondary educational institution characteristics (selectivity based on median SAT score, private institution, and public institution).<sup>11</sup> Because these measures are only available for 940 men and 1,297 women, we are cautious about drawing any conclusions based on the results. The effect of cumulative overall high school GPA is no longer significant for men<sup>12</sup> but remains significant for women ( $p < 0.05$ ). Unfortunately, the Add Health dataset does not provide better information about academic performance beyond high school for a larger sample.

## VI. Discussion and Conclusion

Academic performance in high school usually plays a major role in college selection and admission. However, few studies have investigated whether high school grades are significantly related to overall academic attainment and personal earnings in adulthood. Using multiple waves of data from Add Health, we estimate the effects of cumulative high school GPA on the highest level of education attained and annual personal earnings when respondents are 24 to 34 years of age, or approximately 10 years removed from high school. We estimate each outcome variable with five distinct specifications for each gender. We use the ordered probit model to estimate the seven-category measure of educational attainment, and we estimate the natural log of earnings using robust regression. The analyses include a long list of control variables including PVT score as a proxy for intellectual ability, a dummy variable for currently attending school, socio-demographic variables, household characteristics, and school fixed effects. Finally, we perform several sensitivity tests to examine the robustness of the core findings.

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<sup>11</sup> These measures were collected by Add Health for some of the respondents attending postsecondary institutions at the time of the Wave 3 interview.

<sup>12</sup> However, GPA is statistically significant if college selectivity based on median SAT scores is not included in the model.

All the main results are consistent and stable in direction and statistical significance. In addition, the effects sizes are relatively large and economically meaningful. In quantitative terms, we estimate that a one-unit increase in GPA leads to nearly a full category jump in educational attainment for boys and girls. Similarly, an equivalent increase in high school GPA raises annual earnings in adulthood by an estimated 12.2 percent for males and 14.1 percent for females. Replacing the categorical measure of educational attainment with a continuous measure for years of education, using multiple imputation for all missing GPA observations, and estimating all models with a small set of unambiguously exogenous variables have little influence on the core findings.

One of the interesting ancillary findings from this research is the changing sign and significance for African Americans in the educational attainment specifications. When we exclude high school GPA, PVT score, currently attending school, and school fixed effects from the models, it appears that being African American has a negative and significant effect on educational attainment for males and a significant yet small positive effect for females. However, the estimates are positive, significant, and considerably larger for both genders when we enter the above variables in the fully specified model. This implies that, given the same high school GPA, PVT score, and school characteristics, African Americans advance further in the formal educational system than their White counterparts. It is beyond the scope of this paper to determine whether this estimated disparity is due largely to affirmative action programs, unobserved personal characteristics and differences in individual motivation, or some other phenomenon. Nevertheless, the findings should encourage educators, school administrators, and policy makers who are interested in promoting educational advancement programs for racial minorities. It is important to note, however, that African American men continue to earn less than White men, even after controlling for these characteristics.

In summary, this research quantifies and highlights the long-term importance of academic performance in high school on two socially desirable outcomes in adulthood. Perhaps these estimates will surprise and inspire some, particularly adolescent students, who often require incentives and other motivations to invest more time and effort in their high school coursework. Regardless, parents, teachers, and guidance counselors now have some tangible evidence to rouse those high school students who may be spending too little time with their books.

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**Table 1. Descriptive Statistics**

Variable	Full sample (N=10,216)			Men (N=4,692)	Women (5,524)
	Mean (SD)	Min	Max	Mean (SD)	Mean (SD)
Cumulative overall high school GPA	2.588 (0.829)	0	4	2.436 (0.838)	2.717 (0.798)
Personal earnings (\$)	36,022 (45,253)	0	999,995	42,930 (49,450)	30,153 (40,443)
Natural logarithm of personal earnings <sup>1</sup>	9.594 (2.583)	0	13.816	10.101 (1.886)	9.164 (2.986)
Highest level of education					
Did not complete H.S.	0.053	0	1	0.065	0.042
H.S. graduate	0.149	0	1	0.181	0.122
Voc/tech training after H.S. <sup>2</sup>	0.098	0	1	0.101	0.095
Some college	0.344	0	1	0.343	0.345
Completed college (bachelor's degree)	0.217	0	1	0.206	0.226
Graduate school <sup>3</sup>	0.109	0	1	0.079	0.135
Advanced professional degree <sup>4</sup>	0.030	0	1	0.026	0.034
Years of education <sup>5</sup>	14.444 (2.246)	10	20	14.159 (2.202)	14.686 (2.254)
Add Health Picture Vocabulary Test (PVT)	101.801 (13.663)	13	146	102.763 (13.602)	100.983 (13.662)
Missing Add Health Picture Vocabulary Test (PVT)	0.046	0	1	0.053	0.040
Currently attending school	0.170	0	1	0.139	0.197
Age	28.910 (1.737)	24.405	34.346	29.033 (1.748)	28.806 (1.722)
White	0.644	0	1	0.655	0.634
African American	0.198	0	1	0.177	0.216

Other race	0.158	0	1	0.168	0.150
Hispanic ethnicity	0.149	0	1	0.156	0.144
Born in the U.S.	0.928	0	1	0.922	0.933
Children under age 18 in household	1.254	0	9	1.210	1.292
	(1.192)			(1.149)	(1.227)
Oldest child	0.302	0	1	0.304	0.301
Live in other non-intact household	0.180	0	1	0.181	0.179
Live in single-parent household	0.216	0	1	0.209	0.223
Residential mother	0.952	0	1	0.950	0.953
Residential father	0.738	0	1	0.757	0.721
Residential mother employed	0.822	0	1	0.827	0.818
Residential mother white collar	0.508	0	1	0.520	0.498
Residential father employed	0.699	0	1	0.717	0.684
Residential father white collar	0.271	0	1	0.288	0.257
Residential parent attended college	0.514	0	1	0.535	0.495
Missing residential parent's education	0.016	0	1	0.019	0.014
Residential parent received welfare	0.091	0	1	0.082	0.098

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at Wave 1.

<sup>1</sup>When personal earnings is reported as zero, the natural logarithm of personal earnings is set to zero.

<sup>2</sup>Voc/tech training after H.S. includes respondents who had only some or who completed vocational/technical training after high school.

<sup>3</sup>Graduate school includes respondents with some graduate school, a master's degree, and some graduate training beyond a master's degree.

<sup>4</sup>Advanced professional degree includes respondents who have completed a doctoral degree and who have some or who completed post baccalaureate professional education (e.g., law school, medical school).

<sup>5</sup>Years of education ranges from 10 (less than or some high school) to 20 (doctoral degree or post baccalaureate professional education), depending on respondent's highest level of education.

**Table 2A. Coefficient Estimates for Highest Level of Education Using Ordered Probit (Men, N=4,692)<sup>1</sup>**

Model	A	B	C	D	E
Variables	b/se	b/se	b/se	b/se	b/se
Cumulative overall HS		0.901***	0.852***	0.853***	0.915***
GPA		(0.024)	(0.024)	(0.024)	(0.026)
PVT score			0.015***	0.015***	0.012***
			(0.001)	(0.001)	(0.001)
Missing PVT score			-0.023	-0.024	-0.031
			(0.070)	(0.070)	(0.071)
Currently attending school				0.455***	0.440***
				(0.046)	(0.047)
Age	-0.026***	-0.004	0.0002	0.007	-0.011
	(0.009)	(0.009)	(0.009)	(0.009)	(0.013)
African American	-0.114***	0.182***	0.301***	0.297***	0.240***
	(0.044)	(0.045)	(0.046)	(0.046)	(0.060)
Other race	0.056	-0.018	0.020	0.018	0.011
	(0.048)	(0.048)	(0.048)	(0.048)	(0.058)
Hispanic ethnicity	-0.095**	0.079	0.132***	0.126**	0.058
	(0.047)	(0.049)	(0.049)	(0.050)	(0.058)
Born in the U.S.	-0.188***	-0.115*	-0.235***	-0.214***	-0.109
	(0.061)	(0.064)	(0.065)	(0.065)	(0.069)
Children under age 18 in household	-0.047***	-0.040***	-0.027*	-0.027*	-0.016
	(0.014)	(0.014)	(0.015)	(0.015)	(0.015)
Oldest child	0.043	0.003	-0.024	-0.024	-0.023
	(0.034)	(0.035)	(0.035)	(0.035)	(0.036)
Live in other non-intact household	-0.217***	-0.121***	-0.153***	-0.156***	-0.161***
	(0.043)	(0.044)	(0.044)	(0.044)	(0.046)
Live in single-parent household	-0.179***	-0.098	-0.100	-0.109	-0.150**
	(0.064)	(0.066)	(0.067)	(0.067)	(0.069)
Residential mother	0.122	-0.098	-0.083	-0.087	-0.104
	(0.087)	(0.093)	(0.093)	(0.093)	(0.095)
Residential father	-0.015	-0.049	-0.026	-0.052	-0.013
	(0.092)	(0.101)	(0.100)	(0.100)	(0.105)
Residential mother employed	0.082	0.117**	0.109**	0.107**	0.109*
	(0.052)	(0.054)	(0.054)	(0.054)	(0.057)
Residential mother white collar	0.169***	0.145***	0.124***	0.115***	0.114***
	(0.037)	(0.037)	(0.037)	(0.037)	(0.038)
Residential father employed	-0.051	-0.045	-0.062	-0.034	-0.061
	(0.079)	(0.087)	(0.086)	(0.086)	(0.091)
Residential father white collar	0.458***	0.322***	0.300***	0.311***	0.233***
	(0.038)	(0.039)	(0.039)	(0.039)	(0.040)

Residential parent attended college	0.583*** (0.036)	0.455*** (0.036)	0.418*** (0.036)	0.406*** (0.036)	0.350*** (0.037)
Missing residential parent's education	-0.509*** (0.115)	-0.357*** (0.111)	-0.302*** (0.107)	-0.325*** (0.108)	-0.322*** (0.109)
Residential parent received welfare	-0.191*** (0.060)	-0.107* (0.062)	-0.099 (0.061)	-0.102* (0.062)	-0.106 (0.064)
cut1 _cons	-2.149*** (0.302)	0.168 (0.317)	1.613*** (0.342)	1.812*** (0.342)	1.134** (0.440)
cut2 _cons	-1.222*** (0.300)	1.321*** (0.316)	2.784*** (0.343)	2.999*** (0.342)	2.361*** (0.440)
cut3 _cons	-0.889*** (0.300)	1.715*** (0.316)	3.186*** (0.343)	3.411*** (0.343)	2.789*** (0.440)
cut4 _cons	0.125 (0.299)	2.950*** (0.317)	4.441*** (0.345)	4.677*** (0.345)	4.120*** (0.441)
cut5 _cons	0.980*** (0.299)	3.986*** (0.319)	5.492*** (0.347)	5.738*** (0.346)	5.248*** (0.442)
cut6 _cons	1.735*** (0.301)	4.844*** (0.323)	6.365*** (0.352)	6.628*** (0.352)	6.195*** (0.448)
School fixed effects	No	No	No	No	Yes

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at Wave 1.

<sup>1</sup> Categorical dependent variable is highest level of education, which is constructed using the education variable in Table 1. The variable ranges from 0 (less than or some high school) to 6 (doctoral degree or post baccalaureate professional education). Highest level of education was measured at Wave 4 when respondents were 24 to 34 years of ages.

\* p<.10, \*\* p<.05, \*\*\* p<.01.

**Table 2B. Coefficient Estimates for Highest Level of Education using Ordered Probit(Women, N=5,524)<sup>1</sup>**

Model	A	B	C	D	E
Variables	b/se	b/se	b/se	b/se	b/se
Cumulative overall high school GPA		0.904*** (0.025)	0.842*** (0.025)	0.852*** (0.025)	0.955*** (0.027)
PVT score			0.014*** (0.001)	0.014*** (0.001)	0.010*** (0.001)
Missing PVT score			-0.002 (0.069)	-0.004 (0.069)	-0.006 (0.068)
Currently attending school				0.343*** (0.035)	0.327*** (0.036)
Age	-0.002 (0.008)	0.029*** (0.008)	0.027*** (0.008)	0.033*** (0.008)	0.012 (0.012)
African American	0.102*** (0.037)	0.406*** (0.039)	0.498*** (0.039)	0.466*** (0.040)	0.376*** (0.051)
Other race	0.025 (0.049)	0.018 (0.049)	0.048 (0.049)	0.041 (0.049)	0.045 (0.055)
Hispanic ethnicity	0.008 (0.047)	0.234*** (0.049)	0.277*** (0.050)	0.263*** (0.050)	0.095* (0.055)
Born in the U.S.	-0.225*** (0.062)	-0.116* (0.065)	-0.223*** (0.066)	-0.212*** (0.066)	-0.059 (0.069)
Children under age 18 in household	-0.065*** (0.012)	-0.053*** (0.013)	-0.041*** (0.013)	-0.041*** (0.013)	-0.027** (0.013)
Oldest child	0.128*** (0.032)	0.065** (0.032)	0.035 (0.032)	0.032 (0.032)	0.017 (0.033)
Live in other non-intact household	-0.308*** (0.040)	-0.191*** (0.040)	-0.191*** (0.040)	-0.197*** (0.040)	-0.183*** (0.041)
Live in single-parent household	-0.030 (0.058)	0.038 (0.059)	0.042 (0.058)	0.052 (0.058)	0.055 (0.060)
Residential mother	-0.003 (0.086)	-0.119 (0.085)	-0.119 (0.084)	-0.112 (0.084)	-0.115 (0.086)
Residential father	-0.140 (0.094)	-0.121 (0.094)	-0.095 (0.092)	-0.067 (0.093)	0.005 (0.094)
Residential mother employed	0.008 (0.048)	-0.009 (0.048)	-0.010 (0.048)	-0.009 (0.048)	-0.038 (0.050)
Residential mother white collar	0.228*** (0.034)	0.184*** (0.034)	0.174*** (0.035)	0.174*** (0.035)	0.156*** (0.035)
Residential father employed	0.156* (0.084)	0.085 (0.084)	0.079 (0.082)	0.059 (0.083)	-0.006 (0.084)
Residential father white collar	0.419*** (0.037)	0.293*** (0.037)	0.265*** (0.037)	0.264*** (0.037)	0.168*** (0.038)

Residential parent attended college	0.568*** (0.033)	0.428*** (0.033)	0.382*** (0.034)	0.374*** (0.034)	0.315*** (0.035)
Missing residential parent's education	-0.346*** (0.116)	-0.245** (0.114)	-0.208* (0.113)	-0.240** (0.112)	-0.226** (0.112)
Residential parent received welfare	-0.389*** (0.051)	-0.275*** (0.054)	-0.253*** (0.054)	-0.254*** (0.054)	-0.226*** (0.055)
cut1 _cons	-1.818*** (0.280)	1.096*** (0.294)	2.210*** (0.315)	2.407*** (0.313)	1.841*** (0.415)
cut2 _cons	-0.978*** (0.277)	2.133*** (0.292)	3.259*** (0.314)	3.474*** (0.313)	2.948*** (0.415)
cut3 _cons	-0.605** (0.277)	2.567*** (0.293)	3.701*** (0.314)	3.927*** (0.313)	3.421*** (0.415)
cut4 _cons	0.438 (0.277)	3.825*** (0.294)	4.975*** (0.316)	5.210*** (0.315)	4.780*** (0.417)
cut5 _cons	1.225*** (0.276)	4.755*** (0.295)	5.918*** (0.318)	6.156*** (0.317)	5.788*** (0.419)
cut6 _cons	2.169*** (0.278)	5.802*** (0.298)	6.980*** (0.322)	7.226*** (0.320)	6.914*** (0.423)
School fixed effects	No	No	No	No	Yes

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at Wave 1.

<sup>1</sup> Categorical dependent variable is highest level of education, which is constructed using the education variable in Table 1. The variable ranges from 0 (less than or some high school) to 6 (doctoral degree or post baccalaureate professional education). Highest level of education was measured at Wave 4 when respondents were 24 to 35 years of ages.

\* p<.10, \*\* p<.05, \*\*\* p<.01.

**Table 3. Marginal Effects of GPA on Highest Level of Education**

Pr(Highest level of education)	Men	Women
	dy/dx (SE)	dy/dx (SE)
Less than or some H.S.	-0.030*** (0.003) [-0.557]	-0.016*** (0.002) [-0.529]
H.S. graduate	-0.189*** (0.007) [-2.068]	-0.124*** (0.005) [-2.255]
Voc/tech training after H.S.	-0.089*** (0.005) [-1.926]	-0.104*** (0.005) [-2.535]
Some college	0.030*** (0.007) [0.210]	-0.105*** (0.007) [-0.782]
Completed college	0.214*** (0.008) [3.128]	0.211*** (0.009) [2.971]
Graduate school	0.056*** (0.004) [2.273]	0.123*** (0.005) [2.989]
Advanced professional degree	0.007*** (0.001) [0.923]	0.015*** (0.002) [1.519]

Notes: Marginal effects, standard errors (in parentheses), and elasticities[in brackets] are reported. Estimates are based on Models E for men and women (see Tables 2A and 2B), which use ordered probit and include school fixed effects. The elasticities,  $E(i)$ , are calculated as follows:

$E(i)=[dy/dx]*[xbar_i/ybar_i]$ , where  $dy/dx$  are the marginal effects above,  $xbar_i$  is the mean GPA in education category  $i$ , and  $ybar_i$  is the proportion of the sample in category  $i$ .

\*  $p<.10$ , \*\*  $p<.05$ , \*\*\*  $p<.01$ .



**Table 4A. Robust Regression Estimates for Natural Logarithm of Personal Earnings in Wave 4(Men, N=4,692)**

Model	A	B	C	D	E
Variables	b/se	b/se	b/se	b/se	b/se
Cumulative overall high school GPA		0.106*** (0.014)	0.104*** (0.014)	0.103*** (0.014)	0.115*** (0.015)
PVT score			0.0003 (0.0008)	0.0003 (0.0008)	-0.00003 (0.0008)
Missing PVT score			-0.034 (0.039)	-0.031 (0.039)	-0.040 (0.040)
Currently attending school				-0.218*** (0.027)	-0.206*** (0.027)
Highest level of education <sup>1</sup>					
H.S. graduate	0.186*** (0.041)	0.096** (0.041)	0.097** (0.041)	0.100** (0.042)	0.084** (0.042)
Voc/tech training after H.S.	0.330*** (0.045)	0.217*** (0.046)	0.217*** (0.046)	0.228*** (0.046)	0.195*** (0.047)
Some college	0.332*** (0.039)	0.201*** (0.041)	0.200*** (0.041)	0.240*** (0.042)	0.211*** (0.042)
Completed college	0.589*** (0.042)	0.398*** (0.047)	0.397*** (0.048)	0.408*** (0.048)	0.361*** (0.049)
Graduate school	0.614*** (0.049)	0.405*** (0.055)	0.404*** (0.055)	0.472*** (0.056)	0.410*** (0.057)
Advanced professional degree	0.643*** (0.067)	0.421*** (0.072)	0.418*** (0.072)	0.450*** (0.072)	0.354*** (0.074)
Age	0.054*** (0.005)	0.055*** (0.005)	0.056*** (0.005)	0.052*** (0.005)	0.050*** (0.007)
African American	-0.209*** (0.025)	-0.180*** (0.025)	-0.177*** (0.026)	-0.177*** (0.026)	-0.196*** (0.032)
Other race	0.016 (0.027)	0.009 (0.027)	0.011 (0.027)	0.007 (0.027)	-0.054* (0.031)
Hispanic ethnicity	0.028 (0.028)	0.047* (0.027)	0.047* (0.028)	0.051* (0.028)	0.009 (0.032)
Born in the U.S.	-0.042 (0.036)	-0.034 (0.036)	-0.037 (0.036)	-0.045 (0.036)	-0.016 (0.039)
Children under age 18 in household	0.009 (0.008)	0.010 (0.008)	0.010 (0.008)	0.011 (0.008)	0.016* (0.008)
Oldest child	-0.009 (0.020)	-0.013 (0.020)	-0.014 (0.020)	-0.015 (0.020)	-0.018 (0.020)
Live in other non-intact household	-0.044* (0.025)	-0.037 (0.025)	-0.038 (0.025)	-0.038 (0.025)	-0.044* (0.025)
Live in single-parent	-0.052	-0.048	-0.048	-0.042	-0.057

household	(0.037)	(0.037)	(0.037)	(0.037)	(0.037)
Residential mother	-0.080	-0.105**	-0.105**	-0.099*	-0.102**
	(0.051)	(0.051)	(0.051)	(0.051)	(0.052)
Residential father	-0.076	-0.082	-0.081	-0.069	-0.087
	(0.055)	(0.054)	(0.054)	(0.054)	(0.055)
Residential mother	0.043	0.050	0.050*	0.050*	0.056*
employed	(0.030)	(0.030)	(0.030)	(0.030)	(0.031)
Residential mother white	0.049**	0.049**	0.048**	0.050**	0.044**
collar	(0.021)	(0.021)	(0.021)	(0.021)	(0.022)
Residential father	0.110**	0.105**	0.104**	0.091*	0.103**
employed	(0.048)	(0.047)	(0.047)	(0.047)	(0.048)
Residential father white	0.019	0.014	0.014	0.010	-0.010
collar	(0.023)	(0.023)	(0.023)	(0.023)	(0.023)
Residential parent	-0.028	-0.030	-0.030	-0.029	-0.034
attended college	(0.021)	(0.021)	(0.021)	(0.021)	(0.021)
Missing residential parents'	-0.024	-0.017	-0.017	-0.008	-0.020
education	(0.065)	(0.064)	(0.064)	(0.065)	(0.065)
Residential parent received	-0.104***	-0.093***	-0.092***	-0.094***	-0.067*
welfare	(0.035)	(0.034)	(0.034)	(0.034)	(0.035)
Constant	8.664***	8.501***	8.469***	8.572***	8.788***
	(0.179)	(0.179)	(0.193)	(0.194)	(0.250)
R <sup>2</sup>	0.141	0.150	0.150	0.158	0.208
School fixed effects	No	No	No	No	Yes

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at wave 1.

<sup>1</sup>Those with less than or some high school are the index group.

\* p<.10, \*\* p<.05, \*\*\* p<.01.

**Table 4B. Robust Regression Estimates for Natural Logarithm of Personal Earnings in Wave 4 (Women, N=5,524)**

Model	A	B	C	D	E
Variables	b/se	b/se	b/se	b/se	b/se
Cumulative overall high school GPA		0.105*** (0.015)	0.103*** (0.016)	0.098*** (0.016)	0.132*** (0.016)
PVT score			0.0003 (0.0008)	0.0004 (0.0008)	0.00008 (0.0008)
Missing PVT score			0.022 (0.047)	0.023 (0.047)	0.038 (0.047)
Currently attending school				-0.135*** (0.024)	-0.147*** (0.024)
Highest level of education <sup>1</sup>					
H.S. graduate	0.295*** (0.052)	0.222*** (0.053)	0.222*** (0.053)	0.222*** (0.053)	0.166*** (0.053)
Voc/tech training after H.S.	0.554*** (0.054)	0.465*** (0.055)	0.465*** (0.055)	0.477*** (0.055)	0.414*** (0.056)
Some college	0.625*** (0.048)	0.520*** (0.051)	0.520*** (0.051)	0.556*** (0.051)	0.474*** (0.051)
Completed college	1.007*** (0.051)	0.847*** (0.056)	0.846*** (0.056)	0.857*** (0.056)	0.742*** (0.057)
Graduate school	1.093*** (0.053)	0.922*** (0.060)	0.921*** (0.060)	0.962*** (0.060)	0.855*** (0.061)
Advanced professional degree	1.244*** (0.069)	1.051*** (0.074)	1.049*** (0.075)	1.063*** (0.075)	0.929*** (0.076)
Age	0.047*** (0.005)	0.050*** (0.005)	0.050*** (0.005)	0.048*** (0.005)	0.038*** (0.007)
African American	-0.074*** (0.024)	-0.038 (0.025)	-0.036 (0.025)	-0.023 (0.025)	-0.021 (0.032)
Other race	0.074** (0.031)	0.074** (0.031)	0.075** (0.031)	0.077** (0.031)	0.023 (0.034)
Hispanic ethnicity	0.133*** (0.030)	0.160*** (0.030)	0.161*** (0.030)	0.165*** (0.030)	0.048 (0.035)
Born in the U.S.	-0.111*** (0.040)	-0.107*** (0.040)	-0.110*** (0.040)	-0.114*** (0.040)	-0.044 (0.042)
Children under age 18 in household	-0.004 (0.008)	-0.003 (0.008)	-0.003 (0.008)	-0.003 (0.008)	0.003 (0.008)
Oldest child	0.009 (0.021)	0.005 (0.021)	0.004 (0.021)	0.004 (0.021)	0.008 (0.021)
Live in other non-intact household	-0.060** (0.026)	-0.054** (0.026)	-0.054** (0.026)	-0.051* (0.026)	-0.030 (0.026)
Live in single-parent	-0.076**	-0.068*	-0.068*	-0.073**	-0.088**

household	(0.036)	(0.036)	(0.036)	(0.035)	(0.035)
Residential mother	0.013	0.001	0.002	-0.001	-0.007
	(0.051)	(0.051)	(0.051)	(0.051)	(0.051)
Residential father	-0.079	-0.075	-0.074	-0.082	-0.088
	(0.056)	(0.056)	(0.056)	(0.056)	(0.056)
Residential mother	0.040	0.040	0.040	0.042	0.024
employed	(0.031)	(0.030)	(0.030)	(0.030)	(0.031)
Residential mother white	0.035	0.034	0.034	0.034	0.016
collar	(0.022)	(0.022)	(0.022)	(0.022)	(0.022)
Residential father employed	0.019	0.008	0.008	0.015	-0.002
	(0.050)	(0.050)	(0.050)	(0.050)	(0.050)
Residential father white	0.036	0.031	0.031	0.030	0.016
collar	(0.025)	(0.024)	(0.025)	(0.024)	(0.025)
Residential parent attended	0.045**	0.036*	0.035*	0.037*	0.016
college	(0.021)	(0.021)	(0.022)	(0.021)	(0.022)
Missing residential parents'	-0.077	-0.070	-0.071	-0.060	-0.037
education	(0.079)	(0.079)	(0.079)	(0.079)	(0.079)
Residential parent received	-0.075**	-0.067**	-0.067**	-0.068**	-0.060*
welfare	(0.034)	(0.034)	(0.034)	(0.034)	(0.034)
Constant	8.189***	7.922***	7.896***	7.970***	8.344***
	(0.186)	(0.189)	(0.199)	(0.199)	(0.255)
R <sup>2</sup>	0.222	0.230	0.230	0.233	0.301
School fixed effects	No	No	No	No	Yes

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at wave 1.

<sup>1</sup>Those with less than or some high school are the index group.

\* p<.10, \*\* p<.05, \*\*\* p<.01.

### Appendix A. Predicted Probabilities for Highest Level of Education

Pr(Highest level of education)	Men			Women		
	Median GPA (=2.475)	75th Percentile (=3.058)	99th Percentile (=3.972)	Median GPA (=2.8)	75th Percentile (=3.333)	99th Percentile (=4.0)
Less than or some high school (H.S.)	0.011 [0.009, 0.014]	0.003 [0.002, 0.003]	0.0001 [0.001, 0.001]	0.0046 [0.004, 0.006]	0.0009 [0.001, 0.001]	0.0001 [0.001, 0.001]
H.S. graduate	0.136 [0.119, 0.153]	0.054 [0.046, 0.063]	0.008 [0.006, 0.009]	0.063 [0.053, 0.073]	0.0216 [0.018, 0.026]	0.0041 [0.003, 0.005]
Voc/tech training after H.S.	0.120 [0.108, 0.132]	0.067 [0.060, 0.075]	0.015 [0.013, 0.017]	0.086 [0.077, 0.095]	0.0404 [0.036, 0.045]	0.011 [0.009, 0.012]
Some college	0.494 [0.476, 0.511]	0.446 [0.423, 0.468]	0.231 [0.213, 0.249]	0.4788 [0.459, 0.499]	0.3692 [0.348, 0.390]	0.1945 [0.179, 0.210]
Completed college	0.206 [0.202, 0.211]	0.335 [0.324, 0.346]	0.426 [0.423, 0.445]	0.2783 [0.270, 0.286]	0.3664 [0.352, 0.380]	0.3694 [0.350, 0.389]
Graduate school	0.030 [0.029, 0.031]	0.084 [0.081, 0.087]	0.242 [0.228, 0.255]	0.0825 [0.081, 0.084]	0.1766 [0.172, 0.181]	0.3286 [0.316, 0.342]
Advanced professional degree	0.003 [0.002, 0.004]	0.012 [0.009, 0.016]	0.079 [0.062, 0.096]	0.0067 [0.005, 0.008]	0.0248 [0.020, 0.030]	0.0924 [0.077, 0.108]

Predicted probabilities are reported with 95% confidence intervals in brackets. Other than cumulative overall H.S. GPA, all independent variables are set to their mean values.

## Appendix B. Sensitivity Tests

Model	Natural logarithm of personal earnings (A)		Years of Education (B)	
	Men	Women	Men	Women
Variables	b/se	b/se	b/se	b/se
Cumulative overall high school GPA	0.122*** (0.014)	0.145*** (0.016)	1.322*** (0.033)	1.427*** (0.034)
PVT score	0.000 (0.001)	0.000 (0.001)	0.016*** (0.002)	0.015*** (0.002)
Missing PVT score	-0.042 (0.040)	0.035 (0.047)	-0.051 (0.107)	-0.016 (0.116)
Currently attending school	-0.215*** (0.026)	-0.147*** (0.023)	0.593*** (0.069)	0.415*** (0.057)
Years of education (w4)	0.051*** (0.006)	0.100*** (0.006)		
Age	0.048*** (0.007)	0.036*** (0.007)	-0.009 (0.018)	0.023 (0.018)
African American	-0.198*** (0.032)	-0.013 (0.032)	0.321*** (0.085)	0.566*** (0.077)
Other race	-0.055* (0.031)	0.027 (0.034)	0.017 (0.083)	0.068 (0.084)
Hispanic ethnicity	0.007 (0.032)	0.046 (0.035)	0.078 (0.086)	0.148* (0.086)
Born in the U.S.	-0.022 (0.039)	-0.047 (0.042)	-0.145 (0.102)	-0.085 (0.103)
Children under age 18 in household	0.015* (0.008)	0.001 (0.008)	-0.014 (0.022)	-0.042** (0.020)
Oldest child	-0.016 (0.020)	0.006 (0.021)	-0.051 (0.054)	0.022 (0.051)
Live in other non-intact household	-0.047* (0.025)	-0.030 (0.026)	-0.254*** (0.067)	-0.295*** (0.064)
Live in single-parent household	-0.058 (0.037)	-0.088** (0.035)	-0.229** (0.098)	0.050 (0.087)
Residential mother	-0.102** (0.052)	-0.003 (0.051)	-0.135 (0.137)	-0.171 (0.126)
Residential father	-0.087 (0.055)	-0.108* (0.056)	-0.030 (0.146)	0.036 (0.139)
Residential mother employed	0.057* (0.031)	0.021 (0.031)	0.145* (0.082)	-0.080 (0.075)
Residential mother white collar	0.044** (0.022)	0.018 (0.022)	0.152*** (0.057)	0.239*** (0.054)
Residential father employed	0.102**	0.022	-0.098	-0.052

	(0.048)	(0.050)	(0.127)	(0.124)
Residential father white collar	-0.009	0.015	0.368***	0.288***
	(0.023)	(0.025)	(0.061)	(0.061)
Residential parent attended college	-0.029	0.020	0.484***	0.466***
	(0.021)	(0.022)	(0.056)	(0.054)
Missing residential parents' education	-0.028	-0.040	-0.399**	-0.381*
	(0.065)	(0.079)	(0.172)	(0.195)
Residential parent received welfare	-0.065*	-0.062*	-0.142	-0.332***
	(0.035)	(0.034)	(0.093)	(0.083)
Constant	8.325***	7.421***	9.173***	8.024***
	(0.253)	(0.256)	(0.659)	(0.622)
N	4,692	5,524	4,692	5,524
R <sup>2</sup>	0.207	0.293	0.504	0.488

Notes: GPA, personal earnings, highest level of education, years of education, currently attending school, and age are measured at Wave 4. All other variables are measured at wave 1.

Model A uses years of education (ranging from 10 to 20) as a right hand side variable instead of separate dummy variables for the different categories of educational attainment found in the core models.

Model B presents OLS regression results with years of education as the dependent variable.

<sup>1</sup> Those with less than or some high school are the index group.

\* p<.10, \*\* p<.05, \*\*\* p<.01