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Improving Measurement of Ambiguity-Tolerance Among Teacher Candidates

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ABSTRACT

The process of learning often requires dealing with the ambiguity of uncertain interpretations. A learner's tolerance for ambiguity involves the degree of willingness to accept or adapt to unfamiliar, unpredictable, or uncertain situations and ideas. This study examined the measurement of ambiguity tolerance (AT) among teacher candidates. Pre-service teachers (n = 114) attending a medium size university in the southeastern United States were administered McLain's (1993) Multiple Stimulus Types Ambiguity Tolerance (MSTAT-I) scale. Analyses were first conducted on item responses from the MSTAT-I, then on item responses from the MSTAT-II (McLain, 2009), an instrument comprised of a subset of 13 particular items from the original 22-item instrument. Rasch-model measures and diagnostic analyses were compared and illustrated graphically for the two versions of the AT instrument and then for two shorter versions. Findings indicated validity support for the MSTAT-II instrument, measurement improvement to reducing the number of Likert scale categories to 5, and further measurement benefits of an alternative 9-item AT instrument. The distribution of measured AT among participants was discussed with regard to individual differences among teacher candidates and the potential application of AT measurement as an intellectual disposition among educators.

Uncertainty, Teaching, and Learning. Throughout the learning process, when interpretations of meaning are uncertain students can experience ambiguity. Both teaching and self-directed learning typically involve the reduction of ambiguity relative to the information being learned. Yet, for a student to reduce ambiguity, or for a teacher to help students in that regard, the student must ultimately confront the reality that complete elimination of all possible uncertainty is not likely to occur. Thus, both teacher and student may find ways to tolerate some degree of ambiguity while reducing other ambiguities during the teachinglearning process. Hence, one of the personal characteristics that are of particular relevance to teachers and pupils is their *ambiguity tolerance* (AT).

AT involves a person's willingness to adapt to encounters with ambiguous situations or ideas (Jonassen & Grabowski, 1993). The importance of AT to teaching relates to the earliest uses of the construct when Frenkel-Brunswik (1949) suggested that people intolerant of ambiguity tend to arrive at premature closure, tend to think concretely, and tend to seek simplistic solutions. Though teachers typically simplify material to help students understand concepts at their

current cognitive levels, tendencies in teachers to oversimplify issues and solutions to problems would arguably not serve students well in our increasingly complex world. Supporting this concern Peters and Amburgey (1982) found that those teachers who had a higher AT tended to use higher cognitive levels of verbal responses in their teaching. Also, Ream (1984) showed that group discussions with students involving ambiguous situations or personal experience examples increased the students' tolerance for ambiguity. Furthermore, investigations on AT among educators and learners has supported learning that involves complex problems, novel transfer using new examples, divergent learning tasks, and brainstorming (Jonassen & Grabowski, 1993). With these findings in mind, it should not be a surprise that AT has correlated positively (r = .59) with a constructivist teaching orientation (Gottleib, 2006).

Educators encounter ambiguous situations every day in dealing with student needs and new information. The ways that educators deal with these ambiguous situations have many implications for both the teacher and the students. By understanding their own AT, teachers can consider their many possible reactions to unpredictable situations, and how such reactions and associated choices might differentially affect their pupils. The relevance of a teacher's AT to teaching includes the ways in which associated thought and behavior are influenced by a teachers' existing schema for the classroom environment.

Through much effort by researchers, teachers, and students, instruments used to measure AT have evolved and improved over the course of the past few decades (Budner, 1962; McLain, 2009; Ward, 1994). One early instrument was an attempt to measure AT based upon the idea that perceived ambiguity arises from stimuli that are complex, unfamiliar, and insoluble (Budner, 1962). A different and improved instrument, Multiple Stimulus Types Ambiguity Tolerance (MSTAT-I), was developed three decades later and included items reflecting these same three stimulus types as well as uncertain stimuli and ambiguous stimuli in general (McLain, 1993). The MSTAT-I was further refined as the MSTAT-II using the psychometrically strongest 13 of the 22 items based on empirical evidence supporting the theoretical definition of the AT construct (McLain, 2009). The reduction of items to 13 was also intended to further reduce the cognitive burden on respondents. Items on both the MSTAT-I and MSTST-II were designed to be broadly understandable, value neutral, and context independent to help allow the instruments to be usable in many settings. Additionally, McClain (2009) provided evidence that the MSTAT-II did not encourage socially desirable responding responses, as opposed to truthful responses. Together these instrument characteristics may help improve the measurement validity in future studies, in comparison with characteristics of past instruments used. The current study investigates the quality of both the MSTAT-I and MSTAT-II toward use in future studies involving educators. Although factor analytic research (Lauriola, Foschi, Mosca, Weller, 2016) has identified multiple AT related attitude factors among instruments that may represent meaningful AT dimensions, here we use a Rasch modeling approach in order to employ modern item response theory (IRT) procedures and corresponding diagnostics to address

our research questions using our modest sized sample of participants.

Research Objectives and Hypotheses. To determine whether versions of a selected instrument are well suited to investigating ambiguity tolerance as a general dimension among pre-service educators the following objectives were pursued.

- 1. Examine and compare measurement characteristics of the MSTAT-I and MSTAT-I instruments with teacher candidates.
- 2. Examine whether empirical data supports the use of the MSTAT-I and MSTAT-II with teacher candidates.
- 3. Determine whether Rasch model measures and diagnostics identify any need for revisions within either instrument version.
- 4. Explore and examine possible improved versions of the instrument for teacher candidates.

Method

Participants. Pre-service teachers (n = 114) attending a medium size university in the southeastern United States volunteered for the study. Participants were primarily female (85%) between the ages of 19 to 22 years, though a few participants were older, and age data was not collected.

Instruments and Items. Individual differences were measured using a 7-step Likert scale of the MSTAT-I instrument. The MSTAT-I is McLain's (1993) 22item AT instrument that yielded an alpha reliability of .86 within his discriminant validity investigation. In addition to reliability considerations, the MSTAT-I was chosen for this because items within this instrument were developed to specifically avoid being highly suggestive of socially desirable responses. Figure 1 identifies items that make up the MSTAT-I instrument.

Procedure: Within a classroom environment, participants were asked by an experimenter to complete a paper and pencil MSTAT-I survey as follows: Participants were asked to rate their agreement level on the scale. The survey required approximately 15 minutes to complete. Participants were instructed to answer questions honestly, and to skip items that they did not want to respond to.

Results

Rasch rating-scale model (Andrich, 1978) analyses were conducted across six total calibrations. Four instrument versions stemming from the original 22 item MSTAT-I scale were analyzed within the six calibrations to construct measures and diagnostics from the ordinal raw scores (Linacre, 2016a).

Within the rating-scale model equation below, Likert scale steps, or categories, often numbering between four and seven on instruments, are represented by k.

$$\ln\left(\frac{P_{nik}}{P_{ni(k-1)}}\right) = B_n - D_i - F_k \qquad \text{Rasch Rating Scale Equation}$$

The threshold difficulty level in the rating scale formula is represented as F_k , which can be calculated across a set of items. Thresholds are the points at which the probability of opting for one Likert category is equal to that of the prior adjacent category, or the 0.5 probability level. Also within the rating scale equation, In represents the normal log, making it a logistic formula. P_{nik} represents the probability that a person n on an item i is observed in a rating category k, while $P_{ni(k-1)}$ represents the probability that this same person is observed in the category just prior to k. Thus, the rating scale formula represents the log of the odds of responding with respect to the adjacent categories of the scale (Wright and Mok, 2004). Per the rating scale equation, the rating scale model describes how the probability of a person responding to an item category is a logistic function of the relative distance on a linear scale between the respondent (person) measure location (B_n) , the item difficulty measure location (D_i) , and the 0.5 probability point threshold difficulty (F_k) for choosing between adjacent rating categories of the item.

This formula allows a linear transformation of the ordinal raw scores to derive person measures (B_n) , item difficulties (D_i) , and point threshold difficulties (F_k) in log-odds units referred to as logits. These logits

are the units of a Rasch ruler. The common logit scale for item measures and person measures allows items and persons to be directly compared in a valid and meaningful way. Graphic depictions of Rasch rulers show the distributions of items and persons together, and are commonly referred to as variable maps, which are depicted in figures 8 through 13 and discussed below.

Calibrations and Diagnostics. Table 1 describes the eight categories of diagnostics examined within this study and the questions addressed with each diagnostic tool. Two of these diagnostic categories, item and person model fit, were examined using standardized (Zstd) and Means Square (Mnsq) infit and outfit Zstd fit values within the -2.0 to 2.0 indices. thresholds were considered as fitting the measurement model. Infit is a weighted index that is most sensitive to typical values while outfit is an unweighted index that is sensitive to extreme scores. Diagnostic categories including reliability, separation, sample targeting, person fit, item fit, dimensionality, item polarity, and category functioning were graphically represented in figures 2 through 33 and tables 2 and 3, including summary statistics output, item statistics output, item-person maps, item pathway plots, and category probability curves and output. These findings correspond with each of the six calibrations and the eight categories of measurement diagnostics used for this analysis (Table 1).

The six calibrations used are briefly described, then overall findings and interpretations across these calibrations are summarized. Calibration 1: The MSTAT-I, 22 item instrument was analyzed. Calibration 2: The MSTAT-II, 13 item instrument was analyzed to determine whether measurement characteristics improved from the prior version of the instrument. Calibration 3: Misfitting participants from calibration 2 were deleted for this additional calibration of the 13 item MSTAT-II including persons 49, 32, 74, 75, 2, 78, 99, and 103 based on the second calibration fit values. Calibration 4: A 9 item version was analyzed using all seven categories of the scale. Items eliminated from the 13 item scale included those with low polarity (< .50 point measure

correlation) or high underfit (> 2.0 Zstd). Those eliminated included items 2, 5, 10, and 12 from the MSTAT-II. *Calibration 5:* The 9 item version was analyzed using only 5 categories that were collapsed from the 7 categories used after examining overlapping categories. *Calibration 6:* A five item version was analyzed using 5 categories that were collapsed from the 7 categories used. These five items were the *general ambiguity* items only (items 1, 3, 7, 11, and 13 from MSTAT-II) to provide a comparison of measurement data with only this stimulus type included.

Figures 2 through 7 and table 2 show reliability and separation data indicating levels supporting the differentiation of two groupings of person data across calibrations of instrument versions. Two or more groupings is favorable for differentiating people on the measure. Reliability/Separation levels declined, though very little, with reduction to 13 and 9 item instruments. The expected cost of reducing items was relatively minimal with respect to person reliability.

Tables 2 through 7 also provide summary statistics that include totaled raw scores (column 1) and Rasch measures (columns 3). Averaging the MSTAT-I raw score person total (calibration 1) provided a mean of 4.12 (SD = .71) which serves as a reference relative to the 7 levels of the Likert scale, though such average raw scores should not to be confused with true measures. However these raw score data can show that as a group, participants tended to be almost evenly divided on ambiguity tolerance. The MSTAT-I distribution was relatively normal. Cronbach alpha reliability for MSTAT-I = .86. On the Rasch person measures constructed from raw scores the normal distribution on the MSTAT-I had a mean measure of .07 (SD = .51) range of 3.43 logits. Though the item measure mean is calibrated to 0.0 (SD = .41), the items range was 1.99logits on endorsement difficulty, making it a narrower distribution than that of items.

Sample targeting across all calibrations (figures 8-13) indicates a suitable instrument match to this group of educators. The difficulty level of the instrument's items corresponded with the participant measure locations, which helps minimize measurement error when compared with a poorly targeted sample. Sample targeting redundancy declined with 13 item MSTAT-IIversion but distribution coverage decline was minimal so overall the reduction in items with MSTAT-II did not sacrifice targeting very much. However, by reducing to and instrument with 9 items and then 5 items, a reduced range in distribution coverage resulted, as expected by the reduced diversity of items. Sample targeting overlap was diminished somewhat by the further reduction in items.

Person fit analysis of MSTAT-I (Figure 14) and MSTAT-II supported removal of eight extremely underfitting persons for calibration 3, which yielded improved item fit overall (Figures 15 and 16), but under-fit of one item (#5) persisted. The 9 item version resulted in improved item fit (Figure 17) compared with the 13 item MSTAT-II though two items were under-fitting slightly (#8 and #10). The five item calibration yielded the strongest item fit, as expected. In general, by eliminating items that did not support the dimension well, the measurement became more concentrated on the dimension and corresponding cohesion of the items. Collapsing to 5 categories the original 7 categories further reduced the level item underfit (Figures 18 and 19).

Principal components of residuals was used to examine dimensionality (Table 3). The analyses of 13 and 9 item instrument versions indicated a dominant dimension but also some evidence for secondary dimensions. With 39.0% and 48.3% of the variance accounted for by measures for the 13 item and 9 item instruments respectively, these shorter instrument versions allowed more cohesive AT measurement than the original 22 item MSTAT-I wherein 35.5% of the variance was accounted for by measures. First contrasts revealed relatively smaller unexplained variance though further examination of this unexplained variance is needed in subsequent studies.

Polarity of items improved in strength with the MSTAT-II and removal of misfitting persons (Figures 20-22). Further improvement was seen with reduced item versions that better represented the dominant dimension (figures 23-25), though the relative size of a secondary dimension increased with the five item instrument calibration.

Collapsing to 5 categories from the original 7 categories of the Likert scale was supported by category function analysis (Figures 26-33). Measure and threshold ordering was maintained, and overlap among categories was reduced, though not entirely, as illustrated on figures 31-33.

Limitations. Within this analysis the six calibrations were conducted using one modest sized sample of college level participants (n = 114) who all had career goals within teaching fields. Calibrations with data from other and larger samples of educator candidates and other groups could be used to corroborate findings and resulting interpretations across a more diverse population. Although fit analysis was used to help identify instrument mis-use such as careless responding, the elimination of all instrument misuse effects may not be possible because some instances of misuse may be undetected. Similarly, although McLain (2009) presented data supporting a low and nonsignificant relationship between MSTAT-II and a measure of social desirability, as with many self-rating approaches it remains possible that some form of socially desirable responding affected ratings and measurement of AT using these items despite the intent to create items that would not encourage socially desirable responding responding.

This investigation described only some of the crucial, fundamental Rasch diagnostic tools available for examining the measurement process. Further analyses of fit, dimensionality, and category functioning were not discussed here. Analyses such as differential item functioning (DIF), and factor sensitivity were also not covered within this paper.

Application of Findings. The 13 item MSTAT–II was supported by Rasch measurement findings toward valid assessment of AT among pre-service educators. Reliability and separation levels were appropriate, strong, and only modestly smaller than those of the longer 22 item version (Rasch person reliability was .84 for MSTAT-II versus .87 for MSTST-I). Targeting showed a favorable range of items to people and was only partially affected by the reduction in items due to the redundancy in locations of many of the items removed. That is, redundancy permitted some items to be expendable without changing the overlap of items and persons greatly. Polarity, item fit, and dimensionality were all improved with the MSTAT-II by comparison, particularly following elimination of under-fitting persons whose measures were extremely inconsistent and thereby difficult to interpret. Category functioning was characterized by measures and thresholds within the appropriate increasing direction for each level but categories overlapped excessively with seven levels used. These Rasch analytic findings build upon McLain's (2009) factor analysis and regression analysis investigation of business students and emergency medical technicians.

Examination of a shorter 9 item version of the AT instrument was also promising in that evidence for a single dimension was stronger than that of the 13 or 22 item instruments, while other measurement characteristics were also improved by the elimination of additional items. The instrument was more focused in its targeting of items to people, while retaining items across three of four stimulus types including general ambiguity stimuli, complex stimuli, and insoluble stimuli. Item 12 was the remaining item that was explicitly an uncertain stimuli item, though uncertainty can also be interpreted as overlapping the meaning within general ambiguity items without explicit use of the term 'uncertain' within items. Through a second calibration of the 9 item instrument, the reduction in rating categories from 7 to 5 or possibly fewer was supported for future administrations and empirical testing with a modified instrument. Related to this reduction in categories, using and even number such as 4 rating categories would eliminate the often overused middle category which perhaps requires respondents to use a lesser degree of thought in some cases. As Wolfe and Smith (2007) have noted in favor of an even number of categories "...the middle category is often used as a 'dumping ground' for participants that are compelled to provide a response but would not do so otherwise (pp. 231-232)."

Although the five item version yielded favorable fit and dimensionality because of its emphasis on general ambiguity, the range of items was considerably limited. While calibration of these five *general ambiguity* items was conducted primarily for a relative comparison to those of the other five calibrations, this five item version of general AT could be of use in studies where *complexity* and *insolubility* were not crucial, and the need to minimize items rated existed.

The range of pre-service teacher participant measures was relatively wide and normally distributed on AT. Considering this diversity found among the sample participants, as an intellectual disposition construct within studies of teacher candidate characteristics, AT has interesting potential to help provide meaningful insights on individual differences that have consequences for student learning. In light of prior AT research to date on teaching related tendencies regarding cognition, problem solving, and instructional approaches used, a measurement instrument such as the MSTAT-II and the shorter 9 item version appear to be valuable research tools toward better understanding differences among preservice teachers that could shed light on means to addressing their needs within the teacher education process. That is, by carefully measuring whether an educator tends toward an aversion to ambiguity or to an attraction to ambiguity, for instance, meaningful investigations regarding the many possible implications of these educator tendencies for the thinking, problem solving, and learning among their pupils become more feasible.

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Table 1

Rasch Model Diagnostic Tools for Improving Rating Scale Measurement

Diagnostic Tool	General and Specific Questions and Criteria Regarding Unidimensional Measurement
Reliability	<i>General:</i> Do person measures indicate a broad enough ability range and a sufficient number of items to be reproducible and do item measures represent a broad enough difficulty range and a sufficient number of students to be reproducible? <i>Specific:</i> Is the person reliability level .8 or above and is the item reliability .9 or above?
Separation	<i>General:</i> Do person measures allow discrimination between at least two different levels of the scale (high and low ratings) and do item measures allow discrimination of at least three different levels of difficulty (high, medium, low)? <i>Specific:</i> Are separation index levels greater than 2.0 for persons and greater than 3.0 for items?
Sample Targeting	<i>General:</i> Do item measures and person measures correspond to one another? <i>Specific:</i> Are the distributions of item and person measures located within a shared range on the logit scale, and thereby matched on the variable map?
Person Fit	<i>General:</i> Do person measures function together as an overall measure of the modeled dimension? <i>Specific:</i> Are fit values for each person within the expected range (less than Zstd = 2.0) to avoid underfitting the model?
Item Fit	<i>General:</i> Do item measures function together as an overall measure of the modeled dimension? <i>Specific:</i> Are fit values for each item within the expected range (less than Zstd = 2.0) to avoid underfitting the model?
Dimensionality	<i>General:</i> Do the items of the instrument as a whole measure a primary, dominant dimension? <i>Specific:</i> Are the variances (percentages) accounted for relatively small for any non-primary dimensions and largest for the primary dimension (ideally above 50%)?
Item Polarity	<i>General:</i> Do items function in unison? <i>Specific:</i> Are point measure correlations positive and strong (ideally above .50)?
Category Functioning	<i>General:</i> Do constructed measures function in the expected manner relative to "more" and "less"? <i>Specific:</i> Are average measures and thresholds for each subsequent Likert category in advancing order, from smaller to larger?

- 1. I don't tolerate ambiguous situations very well. (1. General Ambiguity)
- 2. I find it difficult to respond when faced with an unexpected event. (MSTAT-I Only)
- 3. I don't think new situations are any more threatening than familiar situations. (MSTAT-I Only)
- 4. I'm drawn to situations that can be interpreted in more than one way. (MSTAT-I Only)
- 5. I would rather avoid solving a problem that must be viewed from several different perspectives. (2. Insoluble)
- 6. I try to avoid situations that are ambiguous. (3. General Ambiguity)
- 7. I am good at managing unpredictable situations. (MSTAT-I Only)
- 8. I prefer familiar situations to new ones. (4 Unfamiliar)
- 9. Problems that cannot be considered from just one point of view are a little threatening. (5. Insoluble)
- 10. I avoid situations that are too complicated for me to easily understand. (6. Complex)
- 11. I am tolerant of ambiguous situations. (7. General Ambiguity)
- 12. I enjoy tackling problems that are complex enough to be ambiguous. (8. Complex)
- 13. I try to avoid problems that don't seem to have only one "best" solution. (9. Insoluble)
- 14. I often find myself looking for something new, rather than trying to hold things constant in my life. (MSTAT-I Only)
- 15. I generally prefer novelty over familiarity. (10 Unfamiliar)
- 16. I dislike ambiguous situations. (11. General Ambiguity)
- 17. Some problems are so complex that just trying to understand them is fun. (MSTAT-I Only)
- 18. I have little trouble coping with unexpected events. (MSTAT-I Only)
- 19. I pursue problem situations that are so complex some people call them "mind boggling". (MSTAT-I Only)
- 20. I find it hard to make a choice when the outcome is uncertain. (12. Uncertain)
- 21. I enjoy an occasional surprise. (MSTAT-I Only)
- 22. I prefer a situation in which there is some ambiguity. (13. General Ambiguity)

Figure 1. MSTAT-I (McLain, 1993) items are numbered 1 through 22. MSTAT-II (McLain, 2009) item numbers and stimulus type are shown to the right of the corresponding items included in this more recent version.

Rasch Summary Statistics.

SUMMARY OF 114 MEASURED Person

6	TOTAL				MODEL		INF	TIT	OUTF	IT
	SCORE	COUNT	MEAS	URE	S.E.	М	NSQ	ZSTD	MNSQ	ZSTD
MEAN	90.7	22.0		.07	.18	1	.03	4	1.05	4
P.SD	15.5	.2		.51	.01		.82	2.4	.86	2.4
S.SD	15.6	.2		.51	.01		.82	2.4	.86	2.4
MAX.	139.0	22.0	2	.02	.27	5	.44	6.6	6.02	6.9
MIN.	47.0	21.0	-1	.41	.17		.14	-5.1	.15	-5.0
REAL	RMSE .21	TRUE SD	.46	SEPA	RATION	2.22	Pers	son REL	IABILITY	.83
MODEL	RMSE .18	TRUE SD	.47	SEPA	RATION	2.65	Pers	son REL	IABILITY	.87
S.E.	OF Person M	EAN = .05								

Person RAW SCORE-TO-MEASURE CORRELATION = 1.00 CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .86 SEM = 5.79

SUMMARY OF 22 MEASURED Item

	TOTAL				MODEL		INFI	T	OUTF	T
	SCORE	COUNT	MEAS	URE	S.E.	N	INSQ	ZSTD	MNSQ	ZSTD
MEAN	470.0	113.8		.00	.08	1	.01	1	1.05	.1
P.SD	61.7	.4		.40	.00		.31	2.4	.37	2.7
S.SD	63.1	.4		.41	.00		.31	2.4	.38	2.8
MAX.	666.0	114.0		.63	.10	1	.77	5.1	2.11	6.9
MIN.	367.0	113.0	-1	.36	.08		.55	-4.4	.54	-4.5
REAL	RMSE .08	TRUE SD	. 39	SEPA	RATION	4.68	Item	REL	IABILITY	.96
MODEL	RMSE .08	TRUE SD	.39	SEPA	RATION	4.96	Item	REL	IABILITY	.96
S.E.	OF Item MEAN	V = .09								

Item RAW SCORE-TO-MEASURE CORRELATION = -1.00
Global statistics: please see Table 44.
UMEAN=.0000 USCALE=1.0000

Figure 2. Calibration 1. MSTAT-I.

SUMMARY OF 114 MEASURED Person

	TOTAL				MODEL		IN	TIT	OUTF	IT
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	51.8	13.0		.03	.24	1	.02	3	1.03	3
P.SD	10.0	.1		.60	.01		.90	1.9	.96	2.0
S.SD	10.0	.1		.60	.01		.91	2.0	.97	2.0
MAX.	79.0	13.0	1	.77	.31	6	.54	6.6	7.67	7.3
MIN.	27.0	12.0	-1	.57	.24		.06	-4.7	.06	-4.7
REAL I	RMSE .28	TRUE SD	.52	SEPA	RATION	1.86	Pers	son REL	IABILITY	.78
MODEL I	RMSE .24	TRUE SD	.54	SEPA	RATION	2.22	Pers	son REL	IABILITY	.83
S.E. (OF Person M	1EAN = .06								

CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .82 SEM = 4.27

SUMMARY OF 13 MEASURED Item

	TOTAL				MODEL		INFI	T	OUTF	T
	SCORE	COUNT	MEAS	URE	S.E.	Μ	INSQ	ZSTD	MNSQ	ZSTD
MEAN	454.4	113.8		.00	.08	1	.00	3	1.03	1
P.SD	46.5	.4		.31	.00		.34	2.5	.41	2.8
S.SD	48.4	.4		.32	.00		.36	2.6	.42	2.9
MAX.	528.0	114.0		.59	.08	1	.96	6.0	2.21	7.2
MIN.	367.0	113.0	87	.49	.08		.66	-3.0	.66	-3.0
REAL F	RMSE .09	TRUE SD	.30	SEPA	RATION	3.43	Item	REL	IABILITY	.92
MODEL F	RMSE .08	TRUE SD	.30	SEPA	RATION	3.65	Item	REL	IABILITY	.93
S.E. (OF Item MEAN	1 = .09								

DELETED: 9 Item Item RAW SCORE-TO-MEASURE CORRELATION = -1.00 Global statistics: please see Table 44. UMEAN=.0000 USCALE=1.0000

Figure 3. Calibration 2. MSTAT-II.

SUMMARY OF 106 MEASURED Person

	TOTAL				MODEL		INF	TIT	OUTFI	IΤ
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	51.7	13.0		.08	.27	1	.00	2	1.00	2
P.SD	9.4	.1		.67	.01		.55	1.6	. 55	1.6
S.SD	9.4	.1		.67	.01		.56	1.6	.55	1.6
MAX.	75.0	13.0	1	.73	.32	2	.20	2.6	2.20	2.6
MIN.	31.0	12.0	-1	.56	.26		.08	-4.4	.08	-4.4
REAL R	MSE .30	TRUE SD	.60	SEPA	RATION	2.02	Pers	son REL	IABILITY	.80
MODEL R	MSE .27	TRUE SD	.62	SEPA	RATION	2.30	Pers	on REL	IABILITY	.84
S.E. 0	F Person M	EAN = .07								

DELETED: 8 Person

Person RAW SCORE-TO-MEASURE CORRELATION = 1.00 CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .83 SEM = 3.90

SUMMARY OF 13 MEASURED Item

	TOTAL				MODEL		INF]	T	OUTFI	ET .
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	421.2	105.8		.00	.09	1	.00	2	1.00	2
P.SD	43.2	.4		.37	.00		.33	2.3	.33	2.4
S.SD	45.0	.4		.39	.00		.34	2.4	.35	2.5
MAX.	489.0	106.0		.70	.10	1	.85	5.2	1.86	5.2
MIN.	341.0	105.0	0 	.59	.09		.64	-3.1	.64	-3.1
REAL	RMSE .10	TRUE SD	.36	SEPA	RATION	3.62	Item	REL	IABILITY	.93
MODEL	RMSE .09	TRUE SD	.36	SEPA	RATION	3.86	Item	REL	IABILITY	.94
S.E.	OF Item MEA	N = .11								

DELETED: 9 Item Item RAW SCORE-TO-MEASURE CORRELATION = -1.00 Global statistics: please see Table 44. UMEAN=.0000 USCALE=1.0000

Figure 4. Calibration 3. MSTAT-II, underfitting persons removed.

SUMMARY OF 106 MEASURED Person

	TOTAL				MODEL		INF	TIT	OUTF:	IT	
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD	
MEAN	36.0	9.0		.18	.36		.99	2	.99	2	
P.SD	7.2	.0		.95	.02		.64	1.5	.64	1.5	
S.SD	7.2	.0		.96	.03		.64	1.5	.64	1.5	
MAX.	53.0	9.0	2	.42	.46	2	.87	2.9	2.92	2.9	
MIN.	18.0	9.0	-2	.45	.34		.07	-3.6	.07	-3.6	
REAL F	RMSE .4	1 TRUE SD	.86	SEPA	RATION	2.12	Pers	son REL	IABILITY	.82	
MODEL F	RMSE .3	6 TRUE SD	.88	SEPA	RATION	2.42	Pers	son REL	IABILITY	.85	
S.E. (OF Person	MEAN = .09									

DELETED: 8 Person

Person RAW SCORE-TO-MEASURE CORRELATION = 1.00 CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .83 SEM = 2.93

SUMMARY OF 9 MEASURED Item

	TOTAL				MODEL		INF1	T	OUTFI	LΤ
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	423.8	106.0		.00	.11		.99	- <u>.</u> 1	.99	2
P.SD	40.7	.0		.44	.00		.25	1.8	.24	1.8
S.SD	43.2	.0		.47	.00		.26	1.9	.26	1.9
MAX.	489.0	106.0		.89	.11	1	.42	2.7	1.40	2.7
MIN.	341.0	106.0	<u> </u>	.74	.10		.66	-2.9	.67	-2.8
REAL	RMSE .11	TRUE SD	.43	SEPA	RATION	3.89	Item	REL	IABILITY	.94
MODEL	RMSE .11	TRUE SD	.43	SEPA	RATION	4.10	Item	REL	IABILITY	.94
S.E.	OF Item MEAN	N = .16								

DELETED: 13 Item

Item RAW SCORE-TO-MEASURE CORRELATION = -1.00
Global statistics: please see Table 44.
UMEAN=.0000 USCALE=1.0000

Figure 5. Calibration 4. A 9-Item subset of MSTAT-II.

SUMMARY OF 106 MEASURED Person

	TOTAL				MODEL		INF	IT	OUTF	IT
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	27.1	9.0		.01	.40	1	.01	2	1.02	1
P.SD	6.8	.0	1	.07	.07		.64	1.5	.66	1.5
S.SD	6.8	.0	1	.07	.07		.65	1.5	.66	1.5
MAX.	43.0	9.0	3	.21	.75	3	.41	3.1	3.30	2.8
MIN.	12.0	9.0	-2	.58	.36		.08	-3.8	.08	-3.7
REAL	RMSE .46	TRUE SD	.96	SEPA	RATION	2.10	Pers	on REL	IABILITY	.82
MODEL	RMSE .40	TRUE SD	.99	SEPA	RATION	2.44	Pers	on REL	IABILITY	.86
S.E. (OF Person M	EAN = .10								

DELETED: 8 Person Person RAW SCORE-TO-MEASURE CORRELATION = .99 CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .84 SEM = 2.73

SUMMARY OF 9 MEASURED Item

	TOTAL				MODEL		INF	Т	OUTF	IT
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	319.3	106.0		.00	.11	1	.00	1	1.02	.1
P.SD	36.4	.0		.46	.00		.22	1.6	.22	1.6
S.SD	38.6	.0		.49	.00		.23	1.7	.23	1.7
MAX.	380.0	106.0		.89	.12	1	.40	2.7	1.39	2.5
MIN.	249.0	106.0	-	.78	.11		.68	-2.8	.67	-2.8
REAL	RMSE .12	TRUE SD	.45	SEPA	RATION	3.77	Item	REL	IABILITY	.93
MODEL	RMSE .11	TRUE SD	.45	SEPA	RATION	3.96	Item	REL	IABILITY	.94
S.E.	OF Item MEA	N = .16								
	DE	LETED:	13 Ite							
tem RA	W SCORE-TO-	MEASURE COP	RELATI	ON =	-1.00					
lobal	statistics:	please see	a Table	44.						
MEAN=.	0000 USCALE	=1.0000								

Figure 6. Calibration 5. A 9-item subset of MSTAT-II with 5 categories collapsed from the 7 categories used.

SUMMARY OF 104 MEASURED (NON-EXTREME) Person

	TOTAL					MODEL		INF	TI	OUTF	IT
	SCORE	CO	UNT	MEAS	URE	S.E.	P	INSQ	ZSTD	MNSQ	ZSTD
MEAN	14.6		5.0	-	.17	. 59		.99	3	1.00	3
P.SD	3.9		.0	1	.33	.09		.87	1.6	.88	1.6
S.SD	3.9		.0	1	.34	.09		.88	1.6	.88	1.6
MAX.	24.0		5.0	3	.76	1.06	3	.38	2.7	3.37	2.7
MIN.	6.0		5.0	-3	.47	.54		.07	-3.2	.07	-3.1
REAL	RMSE .	9 TRUE	SD	1.14	SEP	ARATION	1.65	Pers	son REL	IABILITY	.73
MODEL	RMSE .	Ø TRUE	SD	1.19	SEP/	ARATION	1.99	Pers	son REL	IABILITY	.80
S.E.	OF Person	MEAN =	.13								

MAXIMUM EXTREME SCORE: 2 Person 1.9% DELETED: 8 Person

SUMMARY OF 106 MEASURED (EXTREME AND NON-EXTREME) Person

	Т	OTAL					MODEL		INF	IT		OUTF	TI
	s	CORE	COL	INT	MEAS	URE	S.E.	M	INSQ	ZS	TD	MNSQ	ZST
MEAN		14.8	5	5.0	-	.07	.62						
P.SD		4.1		.0	1	.50	.19						
S.SD		4.1		.0	1	.50	.19						
MAX.		25.0	5	5.0	5	.05	1.85						
MIN.		6.0	5	5.0	-3	.47	.54						
REAL	RMSE	.73	TRUE	SD	1.31	SEP	ARATION	1.80	Pers	on	RELIA	BILITY	.76
MODEL	RMSE	.65	TRUE	SD	1.35	SEP.	ARATION	2.09	Pers	on	RELIA	BILITY	.81
S.E.	OF Pe	rson M	EAN =	.15									

Person RAW SCORE-TO-MEASURE CORRELATION = .98 CRONBACH ALPHA (KR-20) Person RAW SCORE "TEST" RELIABILITY = .78 SEM = 1.91

SUMMARY OF 5 MEASURED (NON-EXTREME) Item

	TOTAL				MODEL		INFI	Т	OUTF	IT
	SCORE	COUNT	MEAS	URE	S.E.	M	INSQ	ZSTD	MNSQ	ZSTD
MEAN	314.6	106.0		.00	.13		.99	1	1.00	.0
P.SD	21.6	.0		.35	.00		.15	1.1	.17	1.3
S.SD	24.1	.0		.39	.00		.16	1.3	.19	1.4
MAX.	340.0	106.0		.54	.13	1	.18	1.3	1.21	1.5
MIN.	281.0	106.0	<u>1</u> Norska kole na n	.41	.13		.74	-2.1	.72	-2.2
REAL	RMSE .1	3 TRUE SD	.32	SEPA	ARATION	2.47	Item	REL	IABILITY	.86
MODEL	RMSE .1	3 TRUE SD	.32	SEPA	RATION	2.54	Item	REL	IABILITY	.87
S.E.	OF Item ME	AN = .17								
	D	ELETED:	17 Ite	m						
tem RA	W SCORE-TO	-MEASURE COR	RELATI	ON =	-1.00					
lobal	statistics	: please see	e Table	44.						
MEAN=	.0000 USCAL	E=1.0000								

Figure 7. Calibration 6. A 5-item subset of MSTAT-II using only general ambiguity stimuli items and 5 categories.

Table 2

Rasch Person (Test) Reliability and Separation Indices

Calibration/Items	Reliability	Separation
1) 22	.87	2.65
2) 13	.83	2.22
3) 13	.84	2.30
4) 9	.85	2.42
5) 9	.86	2.44
6) 5	.81	2.09

Summary of person reliability and separation across six calibrations.

Rasch Sample Targeting Variable Maps.



Figure 8. Item-Person Variable Maps using MSTAT-I and MSTAT-II compared.

9 Item Instrument



Figure 9. Item-Person Variable Map using 9 item version derived from MSTAT-II.



Figure 10. Item-Person Variable Map using 5 item version derived from MSTAT-II.



Figure 11. Histogram Variable Map using MSTAT-II, 13 Item Instrument for sample targeting graphic visualization.



Person

Figure 12. Histogram variable map using 9 item instrument subset of MSTAT-II for sample targeting graphic visualization.



Person

Figure 13. Histogram variable map using 5 Item instrument subset (general ambiguity stimuli) from MSTAT-II for sample targeting graphic visualization.



Figure 14. Person Outfit using MSTAT-I



Figure 15. Item Outfit using MSTAT-I



Figure 16. Item Outfit using MSTAT-II



Figure 17. Item Outfit using 9 item, 7 category version derived from MSTAT-II



Figure 18. Item Outfit using 9 item, 5 category version derived from MSTAT-II



Figure 19. Item Outfit using 5 item, 5 category version derived from MSTAT-II

 Table 3

 Dimensionality Analysis with Principal Components of Residuals across Calibrations.

tion / Items	Measures	<u>First Contrast</u>
22	35.5%	9.3%
13	37.1%	9.9%
13	39.0%	9.6%
9	48.1%	9.6%
9	48.3%	8.9%
5	52.6%	15.6%
	ion / Items 22 13 13 9 9 5	ion / Items Measures 22 35.5% 13 37.1% 13 39.0% 9 48.1% 9 48.3% 5 52.6%

Note. Percentage of variance accounted for by measures and by the first contrast within principal components analysis (PCA) of residuals. Higher percentages for measures provide stronger support for unidimensionality (e.g. calibrations 4 through 6). First contrasts of residuals indicate strength of an additional dimension. Percentages of variance calculated from PCA eigenvalues.

Polarity Analysis of Items and corresponding Item Statistics

Item STATISTICS: CORRELATION ORDER

ENTRY	TOTAL	TOTAL		MODEL	IN	FIT	OUT	FIT	PTMEAS	UR-AL	EXACT	MATCH	
NUMBER	SCORE	COUNT	MEASURE	S.E.	MNSQ	ZSTD	MNSQ	ZSTD	CORR.	EXP.	OBS%	EXP%	Item
5	505	114	19	.08	1.77	5.1	2.11	6.9	.31	.50	24.6	31.4	AMBIG5
21	666	114	-1.36	.10	1.20	1.4	1.36	2.3	.32	.41	37.7	38.0	AMBIG2
9	483	113	09	.08	1.21	1.6	1.51	3.6	.33	.51	36.3	31.0	AMBIG9
14	426	114	.28	.08	1.45	3.3	1.47	3.4	.34	.52	30.7	30.6	AMBIG14
18	465	114	.05	.08	.99	.0	1.00	.0	.41	.51	31.6	30.7	AMBIG1
17	471	114	.01	.08	1.61	4.3	1.63	4.3	.44	.51	21.9	30.8	AMBIG1
13	505	114	19	.08	1.00	.1	1.04	.4	.46	.50	37.7	31.4	AMBIG1
15	413	113	.33	.08	.88	-1.0	.88	-1.0	.46	.52	37.2	30.5	AMBIG1
20	383	114	.53	.08	1.12	1.0	1.12	1.0	.50	.52	38.6	30.7	AMBIG2
7	508	114	21	.08	.83	-1.4	.82	-1.5	.51	.50	42.1	31.4	AMBIG7
3	475	113	04	.08	.97	2	.97	2	.53	.51	31.9	30.9	AMBIG3
2	516	114	26	.08	1.11	.9	1.09	.8	.55	.50	32.5	31.6	AMBIG2
10	528	114	34	.08	.93	5	.90	7	.55	.50	34.2	32.1	AMBIG1
22	473	114	.00	.08	.59	-3.9	.59	-3.8	.55	.51	42.1	30.8	AMBIG2
19	391	113	.47	.08	1.14	1.1	1.13	1.1	.56	.52	29.2	30.6	AMBIG19
1	419	114	.32	.08	1.02	.2	1.02	.2	.57	.52	31.6	30.5	AMBIG1
11	477	114	03	.08	.65	-3.2	.66	-3.1	.59	.51	45.6	30.8	AMBIG1:
6	444	114	.17	.08	.73	-2.4	.73	-2.4	.63	.52	41.2	30.6	AMBIG6
4	516	114	26	.08	.82	-1.5	.81	-1.6	.63	.50	43.0	31.6	AMBIG4
8	367	114	.63	.08	.92	6	.93	5	.64	.51	35.1	31.0	AMBIG8
12	463	114	.06	.08	.55	-4.4	.54	-4.5	.68	.51	52.6	30.7	AMBIG1
16	447	114	.15	.08	.70	-2.7	.70	-2.7	.68	.51	43.9	30.6	AMBIG1
MEAN	470.0	113.8	.00	.08	1.01	1	1.05	.1			36.4	31.3	
P.SD	61.7	.4	.40	.00	.31	2.4	.37	2.7			7.0	1.5	

Figure 20. Calibration 1, MSTAT-I instrument. Point measure correlation (column 10, 'PTMEASUR') indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

Item STATISTICS: CORRELATION ORDER

Item STATISTICS: CORRELATION ORDER

	MATCH	EXACT	UR-AL	PTMEAS	FIT	FIT OUT	L IN	1	TOTAL	TOTAL	NTRY
Item	EXP%	OBS%	EXP.	CORR.	ZSTD	ZSTD	MNSQ	EASURE	COUNT	SCORE	UMBER
AT5	33.7	24.6	.55	.34	7.2	6.0 2.21	8 1.96	33	114	505	5
AT15	33.1	40.7	.56	.39	1.0	.8 1.13	8 1.10	.25	113	413	15
AT9	33.6	39.8	.56	.42	3.0	1.9 1.43	8 1.26	21	113	483	9
AT22	33.5	40.4	.56	.53	-2.5	-2.6 .71	8 .70	12	114	473	22
AT13	33.7	37.7	.55	.53	.3	.2 1.03	8 1.02	33	114	505	13
AT20	32.8	37.7	.56	.56	1.2	1.3 1.15	8 1.16	.48	114	383	20
AT10	33.8	32.5	.55	.61	5	5 .93	8 .94	49	114	528	10
AT1	33.1	33.3	.56	.62	.4	.4 1.04	8 1.04	.24	114	419	1
AT12	33.4	49.1	.56	.64	-2.8	-2.9 .68	8 .67	05	114	463	12
AT8	32.7	36.0	.56	.64	.0	.1 1.00	8 1.01	.59	114	367	8
AT11	33.5	42.1	.56	.65	-3.0	-3.0 .66	8 .66	15	114	477	11
AT6	33.3	50.0	.56	.69	-2.4	-2.4 .72	8 .72	.07	114	444	6
AT16	33.3	51.8	.56	.74	-2.8	-2.7 .69	8 .69	.06	114	447	16
	33.3	39.7	+		1	3 1.03	8 1.00	.00	113.8	454.4	MEAN
	.3	7.3	i	1	2.8	2.5 .41	0 .34	.31	.4	46.5	P.SD

Figure 21. Calibration 2, MSTAT-II instrument. Point measure correlation (column 10, 'PTMEASUR') indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

ENTRY	TOTAL	TOTAL		MODEL I	VFIT OUT	FIT	PTMEAS	UR-AL	EXACT	MATCH	
NUMBER	SCORE	COUNT	MEASURE	S.E. MNSQ	ZSTDMNSQ	ZSTD	CORR.	EXP.	OBS%	EXP%	Item
15	383	105	.29	.09 1.24	1.7 1.28	2.0	.36	.57	41.9	35.9	AT15
5	479	106	50	.09 1.85	5.2 1.86	5.2	.47	.56	30.2	37.0	AT5
20	357	106	.56	.09 1.25	1.8 1.25	1.8	.50	.58	38.7	36.0	AT20
22	442	106	17	.09 .65	-3.0 .65	-3.0	.52	.57	44.3	36.4	AT22
9	442	105	21	.09 1.14	1.0 1.16	1.2	.53	.57	41.9	36.5	AT9
10	489	106	59	.10 1.06	.5 1.05	.4	.58	.56	34.9	37.2	AT10
11	447	106	22	.09 .79	-1.7 .78	-1.7	.60	.57	46.2	36.5	AT11
13	455	106	29	.09 .87	-1.0 .87	-1.0	.61	.57	39.6	36.7	AT13
8	341	106	.70	.09 1.14	1.0 1.13	1.0	.62	.58	38.7	36.0	AT8
1	382	106	.34	.09 1.01	.1 1.01	.1	.62	.58	32.1	35.9	AT1
12	429	106	06	.09 .69	-2.6 .69	-2.6	.63	.57	52.8	36.2	AT12
6	416	106	.05	.09 .69	-2.6 .70	-2.6	.70	.57	52.8	36.2	AT6
16	<mark>413</mark>	106	.08	.09 .64	-3.1 .64	-3.1	.77	.57	54.7	36.2	AT16
MEAN	421.2	105.8	.00	.09 1.00	2 1.00	2		+	42.2	36.4	
P.SD	43.2	.4	.37	.00 .33	2.3 .33	2.4		- i	7.5	.4	

Figure 22. Calibration 3, MSTAT-II instrument. Point measure correlation (column 10, 'PTMEASUR') indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

Item STATISTICS: CORRELATION ORDER

ENTRY	TOTAL	TOTAL		MODEL	IN	FIT	OUT	FIT	PTMEAS	UR-AL	EXACT	MATCH	
NUMBER	SCORE	COUNT	MEASURE	S.E.	MNSQ	ZSTD	MNSQ	ZSTD	CORR.	EXP.	OBS%	EXP%	Item
22	442	106	19	.11	.80	-1.5	.80	-1.6	.56	.65	45.3	41.6	AT22
10	489	106	74	.11	1.42	2.7	1.40	2.7	.57	.63	35.8	43.1	AT10
8	341	106	.89	.10	1.35	2.5	1.33	2.4	.61	.67	39.6	39.3	AT8
13	455	106	34	.11	1.01	.1	1.03	.3	.65	.64	45.3	42.1	AT13
1	382	106	.46	.10	1.14	1.0	1.15	1.1	.66	.66	38.7	39.2	AT1
6	416	106	.09	.10	.92	5	.93	5	.67	.65	52.8	40.8	AT6
11	447	106	25	.11	.88	9	.85	-1.1	.67	.64	47.2	41.7	AT11
12	429	106	05	.10	.77	-1.8	.75	-2.0	.68	.65	51.9	41.3	AT12
16	413	106	.13	.10	.66	-2.9	.67	-2.8	.81	.65	55.7	40.8	AT16
MEAN	423.8	106.0	.00	.11	.99	1	.99	2			45.8	41.1	
P.SD	40.7	.0	.44	.00	.25	1.8	.24	1.8		í	6.4	1.2	

Figure 23. Calibration 4, Nine item instrument. Point measure correlation (column 10, 'PTMEASUR') indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

ENTRY	TOTAL	TOTAL		MODEL	IN	FIT	OUT	FIT	PTMEAS	UR-AL	EXACT	MATCH	
NUMBER	SCORE	COUNT	MEASURE	S.E.	MNSQ	ZSTD	MNSQ	ZSTD	CORR.	EXP.	OBS%	EXP%	Item
22	336	106	21	.11	.88	9	.92	5	.56	.65	44.3	43.3	AT22
10	380	106	78	.12	1.40	2.7	1.39	2.5	.56	.62	37.7	45.6	AT10
8	249	106	.89	.12	1.31	2.2	1.28	2.0	.61	.67	39.6	41.8	AT8
13	348	106	36	.11	1.00	.0	1.17	1.3	.65	.64	45.3	44.4	AT13
11	340	106	26	.11	.90	8	.86	-1.1	.66	.64	48.1	43.7	AT11
1	281	106	.48	.11	1.12	.9	1.12	.9	.66	.67	42.5	41.7	AT1
12	324	106	06	.11	.85	-1.2	.91	7	.67	.65	51.9	42.7	AT12
6	310	106	.12	.11	.89	8	.89	8	.69	.66	56.6	42.8	AT6
16	306	106	.17	.11	.68	-2.8	.67	-2.8	.81	.66	56.6	42.8	AT16
MEAN	319.3	106.0	.00	.11	1.00	1	1.02	.1			47.0	43.2	
P.SD	36.4	.0	.46	.00	.22	1.6	.22	1.6	1	1	6.5	1.1	

Figure 24. Calibration 5, Nine item, five category instrument. Point measure correlation (column 10) indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

Item STATISTICS: CORRELATION ORDER

ENTRY	TOTAL	TOTAL		MODEL	IN	IFIT	OUT	FIT	PTMEAS	UR-AL	EXACT	MATCH		1
NUMBER	SCORE	COUNT	MEASURE	S.E.	MNSQ	ZSTD	MNSQ	ZSTD	CORR.	EXP.	OBS%	EXP%	Item	1
22	336	106	34	.13	1.00	.1	1.07	.5	.62	.71	53.8	48.5	AT22	i
6	310	106	.08	.13	1.07	.5	1.07	.5	.71	.72	59.6	48.9	AT6	Ì
11	340	106	41	.13	.95	3	.91	6	.71	.70	59.6	48.3	AT11	Ì
1	281	106	.54	.13	1.18	1.3	1.21	1.5	.72	.74	52.9	47.3	AT1	j
16	306	106	.14	.13	.74	-2.1	.72	-2.2	.82	.73	64.4	48.8	AT16	1
MEAN	314.6	106.0	.00	.13	.99	1	1.00	.0			58.1	48.4		1
P.SD	21.6	.0	.35	.00	.15	1.1	.17	1.3		i	4.2	.6		İ

Figure 25. Calibration 6, Five item, five category instrument. Point measure correlation (column 10, 'PTMEASUR') indicates item polarity. Item statistics are sequenced in order of correlation. Item fit statistics are shown in columns 6 through 9 as standardsized (Zstd) and means square (Mnsq) indices of weighted infit and unweighted outfit.

LABEL	SCOP	RE COUNT	%	AVRGE	E E)	MPLE .	MNSQ	MNSQ	THRESHOLD	MEAS	JRE			
1	1	105	4	5	4	65	1.27	1.49	NONE	(-2.8	88)	1		
2	2	263	11	4	1	43	1.07	1.17	-1.46	-1.5	50 1 3	2		
3	3	489	20	2	26	22	.85	.85	94	(59 :	3		
4	4	606	24	0	2	.00	.76	.73	33	(84 j 4	4		
5	5	567	23	.2	25	.24	.91	.92	.18		53 !	5		
6	6	334	13	. 5	57	.55	.96	.96	.92	1.	53 (6		
7	7	140	6	.9	9	.99	1.09	1.08	1.63	(3.0	01) i	7		
			+			+		+	+	+				
MISSI	NG ED AN	4 /ERAGE	0 .s n	.1 Iean c	of n	neasur	es in c	ategory	y. It is no	ot a pa	arame	ter est	im <mark>at</mark> e.	
MISSI	NG ED AN	4 /ERAGE	0 Ls m	.1 iean d	1 of n <u>s</u>	neasuro SCORE-	es in c TO-MEAS	ategory	 y. It is no 50% CUM.	ot a pa	arame ¹	ter est	imate.	20-400
MISSI BSERV CATEGO LABEI	NG ED AV ORY L	4 /ERAGE 3 STRUC MEASURE	0 Ls m TUR	.1 Iean d RE 5.E.	0f n 	neasur SCORE- F CAT.	es in c TO-MEAS ZO	ategory SURE	y. It is no 50% CUM. PROBABLTY	 coner M->C	arame RENCE C->M	ter est: 	imate. ESTIM DISCR	and the second second
MISSI BSERV CATEG LABEI	NG ED AV ORY L	4 /ERAGE 3 STRUC MEASURE NONE		.1 nean c RE 5.E.	1 of n <u>s</u> A1	neasuro SCORE- T CAT. -2.88)	es in c TO-MEAS ZO -INF	ategory URE NE -2.20	y. It is no 50% CUM. PROBABLTY 	ot a pa COHEN M->C	RENCE C->M	ter est RMSR 2.4744	imate. ESTIM DISCR +	and the state of t
MISSI BSERVI CATEG LABEI 1 2	NG ED AV ORY L	4 /ERAGE 3 STRUC MEASURE NONE -1.46	0 Ls m TUR	.1 nean c RE 5.E. .11	1 of n <u>s</u> A1	neasur SCORE- T CAT. -2.88) -1.50	es in c TO-MEAS ZO -INF -2.20	URE -2.20 -1.05	y. It is no 50% CUM. PROBABLTY 	ot a pa COHEI M->C 0% 38%	RENCE C->M 0% 11%	ter est RMSR 2.4744 1.6291	imate. ESTIM DISCR +	
MISSI BSERVI CATEG LABEI 1 2 3	NG ED AV ORY L	4 /ERAGE 3 STRUC MEASURE NONE -1.46 94	Ø Ls m TUR	.1 nean c RE 5.E. .11 .06	1 of n 	neasuro 5CORE- T CAT. -2.88) -1.50 69	es in c TO-MEAS ZO -INF -2.20 -1.05	URE -2.20 -1.05 36	y. It is no 50% CUM. PROBABLTY -1.86 99	ot a pa COHEI M->C 0% 38% 36%	 arame RENCE C->M 0% 11% 39%	RMSR 2.4744 1.6291 .9212	imate. ESTIM DISCR + .85 .87	
MISSI BSERVI CATEG LABEI 1 2 3 4	NG ED AV ORY L	4 /ERAGE 3 STRUC MEASURE -1.46 94 33		.1 nean c S.E. .11 .06 .05	1 of n AT (-	neasur 5CORE- 7 CAT. -2.88) -1.50 69 04	es in c TO-MEAS ZO -INF -2.20 -1.05 36	URE -2.20 -1.05 -36 .28	y. It is no 50% CUM. PROBABLTY - 1.86 99 35	ot a pa COHEF M->C 0% 38% 36% 34%	 arame RENCE C->M 0% 11% 39% 62%	RMSR 2.4744 1.6291 .9212 .6070	imate. ESTIM DISCR + .85 .87 1.05	
MISSII BSERVI CATEG LABEI 1 2 3 4 5	NG ED AV ORY L	4 VERAGE 3 STRUC MEASURE NONE -1.46 94 33 .18	0 Ls n TUR	.1 nean c RE 5.E. .11 .06 .05 .05	1 of n A1 (-	neasuro 5CORE- 7 CAT. -2.88) -1.50 69 04 .63	-INF -2.20 -1.05 36 .28	URE -2.20 -1.05 .28 1.03	y. It is no 50% CUM. PROBABLTY -1.86 99 35 .25	ot a pa COHEI M->C 0% 38% 36% 34% 40%	 arame RENCE C->M 11% 39% 62% 43%	RMSR 2.4744 1.6291 .9212 .6070 .9056	imate. ESTIM DISCR + .85 .87 1.05 1.14	
MISSII BSERVI CATEG LABEI 1 2 3 4 5 6	NG ED AV ORY L	4 STRUC MEASURE -1.46 94 33 .18 .92	0 is n TUF	.11 .06 .05 .06	11 of n A1	neasuro 5CORE- 7 CAT. -2.88) -1.50 69 04 .63 1.53		URE -2.20 -1.05 36 .28 1.03 2.29	y. It is no 50% CUM. PROBABLTY -1.86 99 35 .25 .97	cohei COHei M->C 0% 38% 36% 34% 40% 39%	I RENCE C->M 0% 11% 39% 62% 43% 22%	RMSR 2.4744 1.6291 .9212 .6070 .9056 1.3938	imate. ESTIM DISCR + .85 .85 .87 1.05 1.14 1.04	

C->M = Does Category imply Measure?





LABEI	SCOF	RE COUN	T %//	AVRGE E	XPECT	MNSQ	MNSQ	THRESHOLD	MEASURE	
1	1	67	5	74	85	1.28	1.53	NONE	(-3.01)	1
2	2	159	11	60	58	1.03	1.05	-1.58	-1.63	2
3	3	308	21	35	32	.86	.86	-1.11	77	3
4	4	410	28	05	05	.72	.69	47	04	4
5	5	316	21	.24	.24	.93	.92	.35	.72	5
6	6	162	11	.59	. 55	.97	.97	1.06	1.65	6
7	7	58	4	.89	.90	1.12	1.13	1.75	(3.13)	7
MISS	ENG	2	0	.37						

OBSERVED AVERAGE is mean of measures in category. It is not a parameter estimate.

CATEGORY	STRUCT	URE	SCORE-	TO-MEAS	URE	50% CUM.	COHE	RENCE		ESTIM	
LABEL	MEASURE	S.E.	AT CAT.	Z0	NE	PROBABLTY	M->C	C->M	RMSR	DISCR	
1	NONE		(-3.01)	-INF	-2.33		0%	0%	2.3052		1
2	-1.58	.13	-1.63	-2.33	-1.16	-1.99	46%	17%	1.4595	.85	2
3	-1.11	.08	77	-1.16	40	-1.12	36%	42%	.8576	.95	3
4	47	.06	04	40	.33	42	41%	65%	.5864	1.07	4
5	.35	.06	.72	.33	1.14	.33	39%	42%	.9220	1.07	5
6	1.06	.08	1.65	1.14	2.41	1.10	39%	17%	1.4143	1.06	6
7	1.75	.14	(3.13)	2.41	+INF	2.10	0%	0%	2.0821	1.00	7

M->C = Does Measure imply Category?

C->M = Does Category imply Measure?







Figure 28. Calibration 2. MSTAT-II Category probability curves for 7 rating categories showing excessive overlap.

LABEI	SCOF	E COUN	T %	AVRGE	EXPECT	MNSQ	MNSQ	THRESHOLD		IEASURE	
1	1	42	3	-1.11	-1.09	1.01	1.02	NONE	1	-3.52)	1
2	2	146	11	73	77	1.13	1.17	-2.17	1	-1.97	2
3	3	303	22	45	44	.95	.97	-1.33	1	95	3
4	4	399	29	09	09	.76	.74	54	1	08	4
5	5	310	23	.26	.28	1.03	1.03	.34	1	.85	5
6	6	146	11	.73	.68	1.00	1.00	1.23	1	2.02	6
7	7	30	2	1.08	1.10	1.04	1.03	2.47	1(3.75)	7
MISS	ENG	2	0	.40	1				1		

OBSERVED AVERAGE is mean of measures in category. It is not a parameter estimate.

CATEGORY	STRUCT	URE	SCORE -	TO-MEAS	URE	50% CUM. COHE			HERENCE		
LABEL	MEASURE	S.E.	AT CAT.	ZC	NE	PROBABLTY	M->C	C->M	RMSR	DISCR	
1	NONE		(-3.52)	-INF	-2.78		0%	0%	1.9770		1
2	-2.17	.16	-1.97	-2.78	-1.42	-2.48	50%	16%	1.4618	1.02	2
3	-1.33	.09	95	-1.42	51	-1.38	39%	44%	.8549	.96	3
4	54	.07	08	51	.37	52	43%	65%	.5716	.99	4
5	.34	.07	.85	.37	1.38	.36	42%	43%	.9239	1.01	5
6	1.23	.09	2.02	1.38	2.95	1.32	55%	20%	1.3716	1.07	6
7	2.47	.19	(3.75)	2.95	+INF	2.70	0%	0%	1.9405	1.00	7

M->C = Does Measure imply Category?

C->M = Does Category imply Measure?

CATEGORY PROBABILITIES: MODES - Andrich thresholds at intersections Ρ ·+····+ R 1.0 + 0 В Α 111 B .8 + 11 77+ Ι 11 77 L 1 7 Ι 1 1 77 Т .6 + 7 1 Y 11 7 1 7 .5 + 0 1 66666667 1 F 2*2222 555 6 766 .4 + 222 1 2*3333344444*55 **5 77 66 1 33 2 4433 5 44 6 5 7 3* ** *5 ** 5* 33 1 4 2 5 33 6 4 7 55 66 R 22 Е T 22 66 .2 + 22 S 666+ 3 ** ** *6 4477 P 222 55 44 1155 2266 333 7744 0 333 55 33333 4444 555111666222 77**3 4444 55555 Ν S E --+----+----<mark>+</mark>----<mark>+</mark>----+---+--4 -3 -2 -1 0 1 2 3 4 Person [MINUS] Item MEASURE

Figure 29. Calibration 3. MSTAT-II using seven categories and underfitting persons removed.

	CATEGORY MEASURE	ANDRICH THRESHOLD	UTFIT MNSQ	NFIT O MNSQ	AMPLE I	OBSVD S	VED 0	OBSER	ORY SCOR	CATEC
			++		+-		+			
1	(-4.03)	NONE	1.10	1.06	-1.69	-1.65	3	21	1	1
2	-2.41	-2.72	.95	.94	-1.22	-1.28	10	96	2	2
3	-1.30	-1.67	.99	.98	76	76	20	190	3	3
4	25	98	.77	.77	25	23	32	305	4	4
5	.99	.35	1.10	1.10	.35	.33	23	224	5	5
6	2.61	1.52	.97	.99	1.09	1.19	10	99	6	6
7	(4.70)	3.50	1.30	1.42	1.86	1.38	1	13	7	7

OBSERVED AVERAGE is mean of measures in category. It is not a parameter estimate.

1	ESTIM		RENCE	COHE	50% CUM.	SCORE-TO-MEASURE				STRUCT	CATEGORY
1	DISCR	RMSR	C->M	M->C	PROBABLTY	NE	Z0	AT CAT.	S.E.	MEASURE	LABEL
1	1	1.8261	0%	0%		-3.26	-INF	-4.03)	i	NONE	1
2	.98	1.2161	22%	43%	-2.99	-1.82	-3.26	-2.41	.21	-2.72	2
3	1.08	.8077	47%	38%	-1.76	80	-1.82	-1.30	.11	-1.67	3
4	1.05	.5621	64%	49%	85	.35	80	25	.08	98	4
5	.97	.8839	46%	44%	.34	1.73	.35	.99	.08	.35	5
6	1.04	1.1650	30%	65%	1.64	3.79	1.73	2.61	.12	1.52	6
17	.86	1.8640	0%	0%	3.62	+INF	3.79	4.70)	.30	3.50	7

M->C = Does Measure imply Category?

C->M = Does Category imply Measure?



Figure 30. Calibration 4 on 9 selected items from MSTAT-II using seven categories.

CATEG	DRY (OBSERV	ED	OBSVD	SAMPLE	INFIT	OUTFIT	ANDRICH	C	TEGORY	
LABEL	SCORE	COUNT	%	AVRGE	EXPECT	MNSQ	MNSQ	THRESHOLD	1 1	IEASURE	
1 1	1	123	13	-1.27	-1.24	1.00	1.02	NONE	+	-2.78)	1
2	2	190	20	66	66	1.00	1.02	-1.38	i ì	-1.27	3
3	3	305	32	06	10	.76	.82	86	İ.	09	4
4	4	224	23	.53	.56	1.11	1.16	.52	ĺ.	1.24	5
5	5	112	12	1.57	1.57	1.05	1.09	1.71	(3.00)	6

OBSERVED AVERAGE is mean of measures in category. It is not a parameter estimate.

CATEGORY	CATEGORY STRUCTURE			CORE -	TO-MEAS	URE	50% CUM.	COHE	RENCE		ESTIM	
LABEL	MEASURE	S.E.	AT	CAT.	ZC	NE	PROBABLTY	M->C	C->M	RMSR	DISCR	
1	NONE		1(-	2.78)	-INF	-2.07		80%	16%	1.2253		1
2	-1.38	.11	-	1.27	-2.07	67	-1.75	38%	51%	.7875	1.07	3
3	86	.08	1	09	67	.53	71	49%	65%	.5272	1.05	4
4	.52	.08	1	1.24	.53	2.19	.51	44%	46%	.8639	.92	5
5	1.71	.12	1	3.00)	2.19	+INF	1.94	73%	29%	1.1380	1.01	6

M->C = Does Measure imply Category?

C->M = Does Category imply Measure?



Figure 31. Calibration 5 on 9 selected items from MSTAT-II using five categories.

	ATEGORY	C/	ANDRICH	UTFIT	NFIT O	SAMPLE :	OBSVD S	VED	OBSER	SORY	ATEC
	MEASURE		THRESHOLD	MNSQ	MNSQ	XPECT	AVRGE E	т %[/	E COUN	SCOP	ABEL
	(-3.19)	1(NONE	1.07	1.05	-1.73	-1.72	12	65	1	1
	-1.60	1	-1.80	.96	.91	95	-1.05	19	103	2	2
4	14	ĺ.	-1.24	.62	.65	20	17	38	200	3	3
5	1.58	İ	.82	1.12	1.07	.66	.78	20	108	4	4
E	3.47)	10	2.22	1.23	1.28	2.08	1.85	10	54	5	5

OBSERVED AVERAGE is mean of measures in category. It is not a parameter estimate.

CATEGORY	STRUCTURE			SCORE -	TO-MEAS	URE	50% CUM.	COHER	RENCE	ESTIM		
LABEL	MEASURE	S.E.	14	AT CAT.	Z0	NE	PROBABLTY	M->C	C->M	RMSR	DISCR	
1	NONE		10	-3.19)	-INF	-2.46		89%	25%	1.1338		1
2	-1.80	.16	1	-1.60	-2.46	90	-2.15	47%	64%	.6911	.98	3
3	-1.24	.12	1	14	90	.71	-1.01	62%	78%	.4395	1.13	4
4	.82	.12	i	1.58	.71	2.63	.75	56%	44%	.8304	1.04	5
5	2.22	.20	10	3.47)	2.63	+INF	2.41	67%	36%	1.1621	.79	6

M->C = Does Measure imply Category?

C->M = Does Category imply Measure?







Figure 33. Calibration 6. Category probability curves for 5-rating categories on 5-Item subset of MSTAT-II (General Ambiguity) showing overlap.

Augusta, Georgia