A LONGITUDINAL STUDY OF THE DOMAIN-SPECIFICITY OF ABILITY AND EFFORT ATTRIBUTIONS IN AFRICAN AMERICAN STUDENTS

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ABSTRACT

Heidi A. Vuletich: A Longitudinal Study of the Domain-Specificity of Ability and Effort Attributions in African American Students
(Under the direction Beth E. Kurtz-Costes)

Students’ causal attributions about the reasons underlying their academic successes and failures influence their academic motivation and subsequent achievement. We investigated whether students’ attributions vary across academic subjects, and whether such domain-generality or specificity changes with development. African American students (N = 565) reported their causal attributions for math, science, and English successes longitudinally from elementary to high school. Structural equation modeling showed that individual differences in students’ tendencies to attribute successes to ability and effort were domain-general, not differing across academic content areas. The lack of domain-specificity in attributions suggests that African American students may view academic outcomes as a single achievement domain rather than differentiating among school subjects. Students may cope with race related stresses, such as discrimination and negative stereotypes, by uniformly making adaptive attributions about their successes.
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CHAPTER 1: INTRODUCTION

The way students respond to their academic successes such as getting a good grade on a test often depends on why they believe they succeeded. A student may believe that the exam was easy and that anyone could have done well. Another student may attribute success to studying hard for the exam. Alternatively, a student could believe that he or she is particularly talented in that domain. These causal attributions are important because they can shape students’ expectations about their future performance, inform their academic self-concept, and influence their motivation and classroom engagement (Kurtz-Costes & Schneider, 1994; Swinton, Kurtz-Costes, Rowley, & Okeke-Adeyanju, 2011; Wolters, Fan, & Daugherty, 2013). For instance, a student who believes that she did well in a math exam because she is particularly talented in that domain is more likely to expect future success in other math examinations, have a high math self-concept, and be more motivated and engaged in the classroom.

The antecedents and consequences of attributions have been investigated in detail, but less is known about their cross-domain generality versus specificity. Knowing whether students develop an attributional style within academic subjects is important for three reasons. First, given the nontrivial consequences of causal attributions, interventions aimed at attributional retraining are important, and a key question that may inform the scope of efficient, successful interventions is whether or not causal attributions generalize across academic domains. Second, having a better understanding of the generalizability of school-related causal attributions is important for understanding achievement motivation in general. For example, there is evidence that
attributions to causes that generalize across success situations predict higher levels of academic performance, whereas those that do not are related to poorer performance (Houston, 2016). Third, analyzing the stability or change in the extent to which attributions generalize across academic domains is informative for understanding children’s developing beliefs about intelligence and effort.

In this study, we sought to test whether ability and effort success attributions are domain-general or domain-specific. Other studies have explored similar questions (Boekaerts, Otten, & Voeten, 2003; Bong, 2004; Vispoel & Austin, 1995), but none with African American students. Findings based on other samples may not generalize to African American students due to the unique racial experiences that often characterize the educational environment in which these youth form attributions.

Our longitudinal design also allowed us to test the possibility that students’ attributions are more general only when students are younger, but become more specific over time. Thus, we tested three models—one specifying cross-domain generality, another specifying cross-domain specificity, and the third specifying developmental change. As discussed below, these models have different theoretical implications that could inform our understanding of achievement motivation in African American students.

**Attribution Theory**

According to attribution theory (Weiner, 1979, 1985, 2012), students attempt to explain the causes of their academic successes and failures, and these causal attributions influence their subsequent motivation. Causes can vary by locus (whether it is external or internal to the student), stability (whether it can change over time), and control (whether a student has power to influence it). Each causal dimension has different consequences for students’ expectancy of
future success or failure (Weiner, 1979). For example, if a student attributes his good score on an essay to an external cause, such as an easy test, he may expect that he will do less well in the future. However, if he attributes his good score to an internal cause, such as having high ability, he is more likely to expect future successes. The theory predicts that students who attribute their success to internal causes, such as ability and effort, have better academic outcomes than students who attribute their success to external causes, such as ease of the task (Sandra Graham & Williams, 2009; Weiner, 1985, 2012).

The relevance of causal attributions for achievement motivation has been well-supported by empirical evidence (for a review, see Graham & Williams, 2009). The findings are especially clear for situations of success. Students’ beliefs about the role of their academic abilities and effort in shaping their successes are important determinants of achievement striving and educational performance (Sandra Graham, 2014; Sandra Graham & Williams, 2009; Swinton et al., 2011). More specifically, attributing successful academic outcomes to ability and effort is positively associated with better academic skills and greater persistence in the face of failure (Salanova, Martinez, & Llorens, 2012; Swinton et al., 2011). These internal attributions also result in more positive self-esteem and pride compared to attributing success to external causes such as the teacher or luck (Weiner, 2012).

In this study, we were interested in investigating the generality or domain-specificity of adaptive attributions for success outcomes. Therefore, we focused on analyzing ability and effort attributions. In addition to being adaptive, ability and effort are among the most common perceived causes of success (Sandra Graham & Williams, 2009).
**Are Youth’s Attributions Specific to Academic Domains?**

A key question that researchers have asked is whether attributions about success generalize across domains. In Frieze and Snyder’s classical study (1980), children’s attributions for performing well on a school test differed from their causal attributions for successfully catching frogs. However, when asked only about academic domains, children might be more likely to show consistent attributional styles. If a child does well on a math test and attributes it to studying hard, does she attribute, to the same extent, her good grade in an English essay to how much effort she put forth in her writing? Or does her tendency to attribute success to effort differ across academic subjects?

Findings from several studies seem to favor the hypothesis that ability attributions are domain-specific. Yet, methodological and analytical limitations in these studies preclude strong conclusions and leave open the question of whether we should expect attributions to be domain-general or domain-specific, especially in African American students. For example, in one study, students in their first year of secondary education in the Netherlands (12- to 13-year-olds) attributed their success in history primarily to effort, whereas they attributed their success in language and math primarily to easiness of the task (Boekaerts et al., 2003). These results were based on students’ attributions of their performance in examinations and were assessed using open-ended questions. However, the data were analyzed using aggregate trends, focusing on school subjects rather than individual differences. Therefore, it is unclear whether students who rated ability as most important to their history performance were also more likely than students with other history attributions (e.g., effort) to rate ability as important for math and language successes. An advantage of our data was that our sample was sufficiently large to compare the fit
of various structural equation models, and we were not limited to comparing mean group differences.

Subject-area differences in ability success attributions were also found in a sample of U.S. middle school students. Students were more likely to attribute their success in English and general music to their high ability than they were to make the same attribution for their success in math and physical education (Vispoel & Austin, 1995). In that study, students were asked to think of a situation in which they had done particularly well or poorly, and then they rated how important several causes were to their success, including internal causes such as ability and effort, and external causes such as task difficulty and teacher influence. The authors assessed the dimensionality of attributions using factor analysis. They found that internal attributions clustered in factors that represented specific subject areas. On the other hand, external attributions clustered in factors by attribution (i.e., teacher influence versus task difficulty), regardless of the academic area. Vispoel and Austin (1995) concluded that internal attributions were domain-specific, whereas external attributions were domain-general. In other words, internal attributions such as ability and effort were similar to each other within academic domains (English or Math), but different across domains (English versus Math), whereas external attributions were similar between academic domains (task difficulty attributions were similar for Math and English), but differed across attribution category (task difficulty versus teacher influence). However, these findings should be interpreted with caution. Attributions regarding critical events in which participants performed particularly well or poorly may not reflect the attributions that students would commonly make about their performance. Our study addressed this particular concern by asking students to think about general situations when they
do well or get a good grade, without invoking specific past instances that may have been uncharacteristic of the student’s performance.

Other evidence for the domain-specificity of attributions comes from a study of 9th grade Korean students (Bong, 2004). Bong (2004) found that ability attributions in three different academic subjects (Korean, English, and mathematics) were mostly independent from each other, whereas effort attributions were somewhat more strongly correlated across domains. For instance, students’ tendencies to attribute outcomes to ability were not significantly correlated for English and Korean classes, or between mathematics and Korean, and only a small, significant correlation was found between English and mathematics ability attributions. In contrast, the correlations between all three pairs of academic domains were significant for effort attributions, albeit the correlations were fairly small. That is, a student who attributed an outcome in one of those classes to effort was likely to attribute outcomes in the other classes to effort. These findings can be interpreted as meaning that ability attributions are domain-specific, whereas effort attributions are somewhat more domain-general. One important limitation of this study, however, was that both success and failure attribution items were used as indicators of a single attribution factor (i.e., ability or effort). However, success and failure outcomes have been shown to elicit different patterns of attributions and should therefore be analyzed separately (e.g., Houston, 2016; Swinton et al., 2011). It is possible that results would have differed had success and failure attributions been analyzed individually.

In sum, although previous research has shown some support for the hypothesis that success attributions are domain-specific, none of the studies provide a strong test of the domain-specific versus domain-general question. Moreover, the findings from previous studies may not generalize to African American students, who differ in social background and history from the
samples analyzed in previous research. In the next section, we discuss the social factors that may play a role in the attributions that African American students make regarding their successes in school, and how these may lead to domain-specific or domain-general attributions.

**African American Students’ Causal Attributions for Academic Outcomes**

Some theoretical frameworks suggest that attributions can be informed by contextual information such as cultural norms and stereotypes (Brandt & Reyna, 2010; Reyna, 2000). Thus, though it is possible that the findings regarding the cross-domain specificity of causal attributions generalize to African American students, it is also possible that race-related factors may limit the generalizability of those findings. For example, a pervasive cultural norm in the United States is that effort is a viable means for success. Cross-cultural research has shown that cultural norms such as these can influence the attributions and beliefs of teachers and students. Compared to German teachers, U.S. teachers are more likely to show effort-related attributions, and U.S. children are more likely than German children to believe that high effort indicates intelligence (Kurtz-Costes, McCall, Kinlaw, Wiesen, & Holland Joyner, 2005; Kurtz, Schneider, Carr, Borkowski, & Rellinger, 1990). In turn, some Asian cultures emphasize effort more than the mainstream U.S. culture, resulting in even greater endorsement of effort attributions. For instance, Chinese-American mothers endorse effort more highly than European-American mothers (Kinlaw, Kurtz-Costes, & Goldman-Fraser, 2001).

In a similar manner, the attributions that African American students make may be informed by the social context in which they form those attributions. In particular, perceptions of discrimination in the classroom or at school could impact the way that African American youth ascribe causality for their performance. Race-related experiences could lead youth to believe that, in order to succeed and overcome the barriers imposed by discrimination, they must exert
more effort and be more intelligent than the average person in all domains of their life (Rowley, Helaire, & Banerjee, 2010). Consequently, African American youth may be more inclined to form domain-general attributions for success outcomes because their social context emphasizes system-level biases that they must overcome in order to succeed, as opposed to specific academic content.

In line with this view, Boekaerts (2003) suggested that students can make causal attributions at different levels of situational description, from superordinate, defined as the most abstract level and encompassing all situations of success or failure, to momentary, defined as the immediate realistic situation. Between these two levels is the middle level, which she theorized represents the school subject. Boekaerts (2003) proposed that, after repeatedly encountering similar learning situations, students form an internal model from which they quasi-automatically infer causality when they encounter another equivalent learning situation. However, the situational level at which students form this activated internal model is not determined. What we propose here is that different social contexts may highlight different situational levels. For most students, the most salient level may be the level of the school subject because schools are organized such that they emphasize subject differences, especially in secondary school. For example, as Boekaerts (2003) points out, different school subjects are taught by different teachers, using different methods, and requiring different examination procedures. However, this middle level may not necessarily be the most salient for all groups. African American youth perceive increasing levels of discrimination with age (Greene, Way, & Pahl, 2006). These perceptions may lead African American youth to form internal models that are focused on the superordinate level, in which their tendency to attribute academic success to ability or effort is similar across academic domains.
Racial stereotypes could also contribute to the generality or specificity of African American students’ attributions. The attributional model of stereotypes, for example, posits that cultural stereotypes serve as readily available sources of information about the causes of outcomes, thus informing causal attributions (Brandt & Reyna, 2010; Reyna, 2008). Evidence supporting this model has shown that members of negatively stereotyped groups who experience more stereotype vulnerability underestimate their ability (Aronson & Inzlicht, 2004), may sometimes be reluctant to take credit for their successes (Crocker, Major, & Steele, 1998), or may experience self-doubt and loss of confidence in their environment (Graham & Taylor, 2002). If students use stereotypes as a source of information about the causes of their academic outcomes, their attributions may be more domain-general, as race-related academic stereotypes tend to be cross-domain. Stereotypes may also play a role in the domain-generality of students’ attributions if students reject the negative stereotypes about their group and uniformly make adaptive attributions about their school successes. Indeed, despite evidencing lower performance than Whites in large national data sets, African American students have similar or higher self-esteem (Gray-Little & Hafdahl, 2000). Van Laar (2000) proposed that African American students maintain positive self-esteem despite low achievement because they attribute low achievement to external causes rooted in discrimination, but maintain internal attributions for success. Support for this model was found in a comparative study of White and Black college students (van Laar, 2000). Compared to White students, Black students’ expectations of future economic success declined after one year in college, but their self-esteem remained high. Black students who made external attributions for failure, but internal attributions for success, were most likely to evidence high academic motivation. Thus, the awareness of stereotypes alone,
regardless of whether students are susceptible to them or reject them, could potentially lead to domain-general attributions.

Analyzing whether the attributions of African American students are domain-general or domain-specific would contribute to our understanding of academic motivation by elucidating the situational level informing students’ internal models and the possible role of social context in shaping attributions of success. If African American students’ attributions are domain-general, it would suggest that they are making attributions at the superordinate level and that students may cope with social factors such as discrimination and stereotypes by making generalized attributions about their academic successes. However, if attributions are domain-specific, it would suggest that these students’ attributions are organized at the middle level, or school subject.

One caveat to the literature analyzed above is that it does not account for the possibility of developmental changes in students’ attributions. Children’s developing theories of intelligence and their increasing self-complexity may lead to changes in their attributional styles. In this study, we seek to address this gap in the literature by testing a developmental model of attributions in addition to the domain-general and domain-specific models.

**Developmental Changes in Attributions**

As children age, their understanding of the nature of ability and effort changes. Early evidence of this developmental change was documented by Nicholls (1984) and has more recently been replicated by Folmer and colleagues (2008). Their basic finding was that younger children do not differentiate between effort and ability, believing that to work hard means to be smart. Older children, on the other hand, see a reciprocal relation between ability and effort such that having to expend greater effort on a task signals lower ability. More specifically, children
ages 5-6 conflate the meaning of ability and effort; children 7- to 9-years-old attribute outcome primarily to effort; children ages 10-11 inconsistently attribute outcome to one or the other; and youth 12 and older see ability as a factor that limits the effects of effort (Folmer et al., 2008). Although these developmental changes are not directly related to the domain-generality versus specificity of attributions, they indicate that children’s understanding about the role that ability and effort play in their successes changes across development. As children develop their theories of intelligence through concept acquisition (Kinlaw & Kurtz-Costes, 2003), their organizational frameworks about ability and effort may change while simultaneously guiding their explanations of new learning situations. Therefore, youth’s more complex understanding of intelligence and effort may lead to more complex causal inferences, resulting in more specialized attributions that differ across academic subjects. For example, a student might perceive that she receives high math grades because she is talented in math, whereas she must work hard to receive good grades in English. This high degree of specificity would be expected only in adolescence, when youth have a more stable and complex understanding of effort and ability. Evidence of this developmental pattern can be found in children’s motivational beliefs. Motivational beliefs (i.e., self-efficacy, task value, and achievement goals) are more generalized in middle-school, becoming more differentiated by academic subject in high school (Bong, 1997, 2001).

Not only do students develop a more complex understanding of ability and effort with age, but they are also increasingly exposed to highly specialized classes as compared to the more domain-integrated classrooms in the early years of education. These structural changes in their learning environments could also prompt domain-specificity in attributions as youth advance to secondary education.
Developmental changes indicating that high school is a time when students’ attributions become more specialized would have important implications for achievement motivation, as high school is a time when students have more freedom to forge their own career path through their course selections and choice of extracurricular activities. Moreover, from middle school to high school, youth undergo other global changes in motivation. For example, from Grades 7 to 11, youth show declines in school participation, sense of belonging to school, and self-regulated learning (Wang & Eccles, 2011). Engagement also declines, and these changes are similar across race and gender (Upadyaya & Salmela-Aro, 2013). As attributions are important for all these motivational factors, understanding how attributions change across development is critical. One particular strength of our study is that it assesses students’ attributions starting in elementary school, which other studies investigating domain-specificity have not done.

Current Study

In this longitudinal study, we investigated the nature of African American students’ attributions regarding their successes in math, science and English to ability and effort. More specifically, we used youth’s attribution responses as given in Grades 5, 7, 10, and 12 to test three models. These models represented students’ attributions for their successes in math, English, and science as domain-general, domain-specific, or changing in specificity over time. Although we did not have a specific hypothesis favoring specificity over generality, we hypothesized that if African American students’ attributions changed in specificity over time, the direction of change would be from more general in elementary school and middle school, to more specific in high school. The hypothesis is based on the idea that as students develop more complex theories regarding ability and effort, develop a more complex self-concept, and are exposed to more specialized course work, their attributions, too, become more specialized.
CHAPTER 2: METHOD

Participants

Participants were 565 African American children and adolescents (314 girls; 251 boys) in the Youth Identity Project, a longitudinal study of academic motivation and achievement. Youth were originally recruited in three cohorts of fifth graders. They were drawn from seven elementary schools in an urban school district in the southeastern region of the United States. The schools were selected based on their racial composition, which was predominantly Black, with 61% or more of the student body being African American in each school. Cohort 1 fifth grade data were collected in 2002-2003, Cohort 2 in 2003-2004, and Cohort 3 in 2004-2005. During original recruitment, 78% of African American 5th graders who were invited to participate returned signed consent forms. Of those, 97% (N = 382) agreed to participate. Attribution data were available for 375 of these students. The average age in Grade 5 was 11.1 years (SD = 0.73).

Subsequent waves of data collection for the original sample occurred when youth were in Grades 7, 10, and 12. Students had enrolled in 17 different middle schools and 22 different high schools. When youth were in Grade 10, additional participants were recruited from the three high schools with the largest number of participants from the original three cohorts. These new participants were in the same English and mathematics classes as students who had been recruited as fifth graders. The new sample included youth from other races, but only data from African American students were analyzed here (N = 176). The new recruits also participated in
Grade 12. Data for the present study were drawn from the four time points (Grades 5 and 7 if available, and Grades 10 and 12).

The sample retention rate for youth with attribution data was 80% from Wave 1 to Wave 2 (N = 301), 86% from Wave 2 to Wave 3 (N = 259; plus the new recruits, N = 434), and 75% from Wave 3 to Wave 4 (N = 325; including the new recruits added at Wave 3). Youth who were lost to attrition came from households with lower household income and lower parent education than that of youth who remained in the study, $F(1, 336) = 14.1$ and $19.1, p \text{'s} = .001$. Participants who remained in the study across the seven years did not differ from those who were lost on Grade 5 attribution measures, all $F$'s < 1.0.

The sample was largely low income, with approximately 39% of the children’s families reporting yearly incomes less than $30,000 during the third wave of data collection, 35% reporting yearly incomes between $30,000 and $60,000, and 26% reporting incomes greater than $60,000 per year. About 8% of parents or caregivers had not completed high school, 57% had earned a high school diploma, 11% had earned an associate’s degree, and 24% had completed a bachelor’s degree or higher.

**Procedure**

Letters requesting informed consent from parents and assent from children were distributed at the selected schools. Children with consent completed a questionnaire in small groups at school and received a small gift as compensation. All questionnaires included attribution items and other measures that are not included in this report. Re-consent of children and their parents and similar data collection procedures were used at each wave.
Measures

**Student Attribution Scale.** Youth completed a 24-item measure of causal attributions for success and failure in four academic domains: English, math, writing, and science. For each domain, students rated how likely it was that their success or failure was due to ability, effort, or the teacher. Only the 4 items attributing success to ability and the 4 items attributing success to effort were analyzed here. The items were (ability/effort): “When I do exceptionally well in English, it is because *I am very smart in verbal areas/ I worked really hard*”; “When I do well in math, it is because *I am really good at math/ I studied hard for the test.*”; “When I get an excellent grade on a science project, it is because *I am talented in science/ I worked very hard at it*”; and, “When I do very well on a writing assignment, it is because *I write well/ I have worked very hard on the assignment*”. The English and writing items were averaged to make a single verbal attribution score, and the math and science items were used as individual item indicators of math success attributions and science success attributions, respectively, to ability or effort.

In early years of the study (i.e., 2002 through 2010), children responded to items using a 4-point Likert scale (*1* = not at all likely; *4* = extremely likely). All responses on this 4-point scale were later converted to values on a 7-point scale (i.e., *1* = 1, *1.5* = 2, *2* = 3, *2.5* = 4, *3* = 5, *3.5* = 6, *4* = 7). This conversion was done because some students used the spaces between numbers to indicate their response or they circled two adjacent numbers. After 2010, students responded to each item using a 7-point Likert scale (*1* = not at all likely; *7* = extremely likely).

**Demographic Information.** Parental educational attainment was measured on a 10-point scale with responses ranging from “less than high school” to “doctoral or professional degree.” Parents reported their household income before taxes on an 11-point scale ranging from “under $10,000 yearly/under $200 weekly” to “over $100,000 yearly/over $2000 weekly.”
Data Analysis

We planned to compare the fit of three structural equation models (SEM’s), each assessing a different theoretical perspective on how African American students make success attributions to ability or effort. The first model was a Domain-General Attribution Model, whereby success attributions for math, science, and English are indicators of a single latent variable (domain-general attribution style for ability or effort) at times 1, 2, 3, and 4 (see Figure 1). The ovals in the diagram represent the latent variables and the rectangles are the observed indicator variables. Each latent variable represents the student’s domain-general attribution style at each time point, and each time point influences the subsequent one. The second model was a simultaneous equation model with no latent variables. This model was the Domain-Specific Attributions Model, whereby attributions of success to ability or effort are entered separately for each academic subject and each time point is regressed on the previous one for each subject (see Figure 2). Cross-lagged paths allowing each academic subject to influence the other academic subjects were included in order to test the least constrained model. The third model was a Developmental Attribution Model, whereby students’ attributions about their success in math, science, and English were indicators of a latent variable (i.e., domain-general attribution style) in the 5th and 7th grade, but then they were domain-specific in 10th and 12th grade (see Figure 3).

In the event that the theorized models did not fit the data, we planned to correlate the errors of the indicator variables by academic subject in an attempt to improve model fit. These correlations would represent relations within each academic subject across time that were not captured by the domain-general attribution latent variables. We planned on modifying Models 1 and 3 with correlated errors (see Figures 4 & 5, respectively), but not Model 2, as Model 2 already includes relations within academic subjects across time. The models with correlated
errors would suggest that there is some degree of specificity by academic subject once the influence of the latent variables is accounted. For this reason, we also planned on modifying Models 1 and 3 to include a unique factor for each academic subject and comparing these models to the models with correlated errors (see Figures 6 & 7, respectively).

Finally, as a more robust test of the domain-specific hypothesis, we planned to fit a simultaneous equations quasi-simplex model, with one quasi-simplex model for each academic subject (see Figure 8). Quasi-simplex models are longitudinal, latent variable models with single indicators for each time point (Heise, 1969; Wiley & Wiley, 1970). These models require the assumption of constant error variance over time in order to be identified. The model we fit is conceptually similar to Model 2, but unlike Model 2, it allows for measurement error to be accounted. Cross-lagged paths allowing each academic subject to influence the other academic subjects were included in order to test the least constrained model.

Other modifications made in order for the models to converge are addressed in the Results section.

Missing data. Due to attrition and to the addition of new cohorts in Grade 10, we had a significant amount of missing data. We used Full Information Maximum Likelihood (FIML) to address this issue. FIML provides good estimates under conditions of data missing completely at random (MCAR) and data missing at random (MAR). A substantial amount of our missing data was by design (MCAR). The rest was assumed to be MAR because students of low SES were more likely than higher SES students to have missing data by attrition. Given that there are many reasons contributing to absenteeism for low SES students, we did not have a strong reason to suspect data missing not at random (MNAR).
**Identification.** The models were identified under the 2-step rule of identification, which requires a transformation of the model into a measurement model in the first step in order to establish that all variances and covariances of the latent variables and factor loadings are identified (Bollen, 1989b). The second step is used to show that the latent variable model would be identified if all the latent variables were perfectly measured. Model 1 is identified under the three-indicator rule (Step 1) and the fully recursive rule with no correlated errors (Step 2). Model 2 is identified under the fully recursive rule with no correlated errors. Model 3 is identified under the three-indicator rule (Step 1) and the fully recursive rule with no correlated errors (Step 2). The remaining models are identified under the rank rule (Step 1) and the fully recursive rule with no correlated errors (Step 2).
CHAPTER 3: RESULTS

Means and standard deviations for all variables appear in Table 1. Correlations for ability attributions appear in Table 2 and for effort attributions in Table 3. To test the domain-generality, domain-specificity and developmental change of attributions, we estimated all the models using a distributionally robust maximum likelihood estimator (MLR) that accounts for possible non-normality of the error distributions. We conducted all analyses in RStudio using lavaan 0.5-20. All the variables were treated as continuous (Rhemtulla, Brosseau-Liard, & Savalei, 2012). To assess model fit, we used the Chi-square test, the Incremental Fit Index (IFI) (Bollen, 1989a), a modified root mean square error of approximation (1-RMSEA) (Steiger & Lind, 1980), the Tucker-Lewis Index (TLI) (Tucker & Lewis, 1973), and the new Bayesian Information Criterion (BIC) (Raftery, 1995; Schwarz, 1978). The Chi-square test allows us to reject or retain the null hypothesis that the implied moment matrices derived from the parameters in our model equal those in the population. When a model is specified correctly, the Chi-square should not be significant. Values approximating 1 are indicative of excellent model fit by IFI, 1-RMSEA and TLI. BIC approximates whether the saturated model is better than the hypothesized model. Scores less than 0 favor the hypothesized model.

In the following sections, we report the results for success ability and success effort attribution models. First, we present the fit indices for all the models that were tested. Then we interpret the factor loadings and coefficient estimates for the model that provided the best fit.
Success Ability Attributions

**Fit indices.** Table 4 displays the fit statistics for each success ability attribution model. The theorized models (Models 1-3) did not provide a good fit. Therefore, the Domain-General and Developmental Attribution Models were modified to include correlated errors by academic subject (Models 4 and 5; Figures 4 and 5, respectively) and to include unique factors by academic subject (Models 6 and 7; Figures 6 and 7, respectively). The Domain-Specific Attribution Model was also modified such that it was transformed into a simultaneous quasi-simplex model with single-indicator latent variables (Model 8; Figure 8). However, the quasi-simplex model did not converge, even after constraining the autoregressive paths and the cross-lagged paths to be equal. Therefore, we simplified the model by only fitting two academic subjects at a time (Math and English, Science and English, Math and Science) and comparing these models to the corresponding two-subject domain-general model. A quasi-simplex model accounts for measurement error, and is therefore a more robust test of the domain-specific hypothesis, providing further evidence of whether a domain-specific model describes these data. The resulting fit indices for these modified, two-subject models are displayed in Table 5.

All the fit indices favored the Domain-General Attribution Model with Correlated Errors (Model 4). The Domain-General Attribution Model with Unique Factors (Model 6) also provided an adequate fit, especially evidenced by the highly negative BIC statistic. However, this model had a significant Chi-square test and lower IFI, TLI and 1-RMSEA fit indices than the Domain-General Model. Based on these results, we selected Model 4 as the best model to describe these data.

The model suggests that students make similar ability attributions about their success in math, science, and English. In other words, individual differences in students’ tendency to view
ability as contributing to their success are stable across academic subjects, and this domain-generality is stable across time. However, the correlated errors also indicate that there is a relation between attributions within each academic subject across time that is not accounted for by the latent variable, signifying a small degree of specificity above and beyond the domain-generality. We analyze and interpret the coefficients for the Domain-General Attribution Model with Correlated Errors in the next section.

**Measurement model.** The coefficient estimates for the measurement model were all significant and suggest that attributions about success in math, English and science to ability are all indicators of a single latent variable: a domain-general attribution style. Figure 9 displays the conceptual model with coefficients and standard errors. Though the model predicts a single latent variable at each time point, a student’s ability attribution style is expected to have a greater effect on success ability attributions for science than for math or English (except at Time 1), with math having the weakest relation with the latent variable. More specifically, for every unit increase in a student’s ability attribution style at any of the four time points, our model predicts that success ability attributions for science will increase by more than half a standard deviation, whereas success ability attributions for math and English will be consistently lower. We should note that the coefficient for success ability attributions for science at Time 4 (12th grade) has a much larger standard error than that of any of the other coefficients, warning us to interpret the estimate with caution. One possibility for the high measurement uncertainty in this coefficient is that some students no longer take science courses later in high school, yet they may have still provided responses because the measurement items are phrased hypothetically (e.g., “*When I do well in a science test, it’s because…*”). This procedure may have introduced more measurement error to our model at the last time point, thus increasing the standard errors at Time 4.
**SEM model.** The regression coefficients for the latent variables suggest that a student’s ability attribution style at an earlier time point will have a significant positive effect on that student’s ability attribution style at a subsequent time point, with these effects getting stronger over time. More specifically, a standard deviation increase in general ability attribution style at Time 1 predicts a .32 standard deviation increase at Time 2, but a standard deviation increase at Time 2 predicts a .50 standard deviation increase at Time 3 (see Figure 9). Finally, a standard deviation increase at Time 3 predicts a .58 standard deviation increase at Time 4. All these coefficients are statistically significant. Each of our latent variables accounts for a good portion of the variability in our data (ranging from 10.5% to 33.5%).

**Success Effort Attributions**

**Fit indices.** Table 6 displays the fit statistics for each success effort attribution model. Overall, the fit indices suggest that the Domain-General Attribution Model corresponds best to the data (Model 1). The fit indices approximated their optimal value of 1 under the robust estimator. The Chi-square test was non-significant and the BIC was highly negative. The Developmental Attributions Model had a moderately good fit according to the supplemental fit indices, but the Chi-square test was significant. Therefore, we selected Model 1 as the best model to describe these data and did not fit any additional modified models. The model suggests that students make similar effort attributions about their success in math, science, and English. In other words, students view effort as contributing equally to their success across academic subjects, across time. We analyze and interpret the coefficients for the Domain-General Attribution Model in the next section.

**Measurement model.** The coefficient estimates for the measurement model were all significant and suggest that attributing success in math, English and science to effort are all
indicators of a single latent variable: a domain-general effort attribution style. Figure 10 displays the conceptual model with coefficients and standard errors. Success effort attributions for science were not as strongly associated with the latent variable in 5th and 7th grade compared to math and English effort attributions. However, they became more closely related to the latent variable in 10th grade.

**SEM model.** The regression coefficients for the latent variables suggest that a student’s general effort attribution style at an earlier time point will have a significant positive effect on his or her effort attribution style at a subsequent time point, with these effects getting stronger over time. More specifically, a standard deviation increase in effort attribution style at Time 1 predicts a .45 standard deviation increase at Time 2, whereas a standard deviation increase at Time 2 predicts a .71 standard deviation increase at Time 3. Finally, a standard deviation increase at Time 3 predicts a .98 standard deviation increase at Time 4. All these coefficients are statistically significant. Each of our latent variables accounts for a modest to good portion of the variability in our data (ranging from 10.5% to 33.5%).
CHAPTER 4: DISCUSSION

The goal of this study was to examine whether ability and effort success attributions are domain-general, domain-specific, or change in specificity over time among African American students. Our results supported the domain-general model of attributions, whereby students endorse ability and effort attributions to the same extent for math, English, and science. In the following sections we discuss the implications of our results.

Ability Attributions

The model that best described students’ success ability attributions was the domain-general model. This model suggests that students view ability as contributing equally to their success across academic subjects. Our results differ from those of other research studies, which have found evidence in European, White American, and Asian samples that success ability attributions are domain-specific (Boekaerts et al., 2003; Bong, 2004; Vispoel & Austin, 1995). Certain methodological and analytical differences between those studies and the current project could be responsible for the differences in findings. However, another explanation is that students’ attributions are informed by the social context in which they form those attributions, and that African American students experience a unique social context. Boekaerts (2003) proposed that, after repeatedly encountering learning situations, students form an internal model from which they quasi-automatically infer causality when they encounter another similar learning situation. Importantly, the “situation” may be understood at different levels: superordinate (outcome-based), middle (school-subject-based), or momentary (examination-
task-based). For most students, especially those in secondary school, the most salient situational level may be the level of the academic subject. Students’ focus on the middle level in secondary school may be a reflection of their more complex understanding of ability and intelligence. It could also reflect the increasing specialization of school subjects. However, other levels may be more significant under specific social circumstances. For African American students, the more salient situational level may be the superordinate level if they experience discrimination at school, which emphasizes system-level biases. Students who perceive system-level biases may be inclined to believe that their successes are in spite of social barriers, and thus attributable to their ability equally across domains.

Stereotype awareness may also contribute to the domain-generality of students’ ability attributions. For example, a Black boy entering high school may be aware that Black boys are stereotyped as being particularly low-achieving compared to other racial groups. When he receives a good grade in math and attributes it to his high ability, he may view this as evidence that the stereotype does not apply to him. Rejecting the broader stereotype may then lead him to attribute his success in other classes to his high ability as well. On the other hand, a student who is more susceptible to internalizing stereotypes may be reluctant to attribute success to her ability across all subjects.

We want to note, however, that the domain-general model only provided a good fit for the data when measurement errors were correlated within academic domains across time. These correlated errors represent a degree of specificity within attributions for each academic subject that was not accounted for by the latent variable. The degree of domain-specificity is weak, however. When we incorporated a unique factor representing each school subject to replace the error correlations, the model did not fit the data as well as the model with correlated errors.
Other findings from this model were that science is a stronger indicator of students’ ability attribution style than math or English. Although the mechanism driving this effect is unclear, it could imply that students’ attributions regarding science should be targeted in attributional retraining interventions with African American youth.

**Effort Attributions**

In contrast to ability attributions, effort attributions were best characterized by a domain-general model without the need to specify correlated errors. The implications of the model are similar to those outlined for ability attributions. Yet, the lack of correlated errors is consistent with the idea that effort attributions are more strongly associated with situational descriptions at the superordinate level, whereas ability attributions draw from both the superordinate and middle levels. In practical terms, this could mean that experiences of discrimination lead students to believe that, regardless of the academic subject, they must exert greater effort than other youth in order to succeed (Sanders, 1997). On the other hand, their beliefs about ability, though still following the same guiding principle, also depend to some extent on the academic subject.

Similarly, experiences of discrimination may lead to low effort attributions across academic subjects if students believe in a reciprocal relation between ability and effort, whereby overcoming barriers signals high ability, and thus low effort. Although the mechanisms driving high-effort and low-effort attributions may be different depending on students’ interpretations of their environments, the results lead to the same outcome: domain-generality.

**Educational Implications**

Our results suggest that, for success outcomes, African American students have a general attribution style whereby their tendency to endorse ability and effort attributions is similar across academic domains. The implications for attributional retraining interventions are that teachers
and educators may choose to focus on retraining students’ attributions in a particular subject-area as their resources allow, and expect to see changes in ability and effort attributions in other academic domains.

**Limitations**

Although we found evidence of domain-generality for effort and ability success attributions, it is important to note that we only analyzed attributions regarding closely related traditional academic subjects. It is possible that attributional responses become more content-specific as academic domains become increasingly different. For example, English, math and science may be viewed as core academic classes and thus lumped into one category characterized by similar ability and effort requirements. In contrast, art, music and physical education classes may be categorized differently. Indeed, Vispoel and Austin (1995) found that the only subject area differences in attributions in their study distinguished physical education and general music from the other subject areas as well as from each other.

Another limitation to our data was that we only had a single indicator for math and science attributions. Having multiple indicators would have allowed us to test a simpler domain-specific model with latent variables that took into account measurement error.

Despite these limitations, our study highlights the importance of investigating attributions across different social groups. It also contributes to our understanding of achievement motivation in African American students. Having domain-general attributions suggests that African American students may be more attuned to system-level factors rather than situations at the level of the school subject, and that they may cope with race related stresses at the system-level, such as discrimination and negative stereotypes, by uniformly making adaptive attributions about their successes.
## APPENDIX A: TABLES

Table 1

*Means and Standard Deviations for Success Ability Attributions and Success Effort Attributions at Times 1-4*

<table>
<thead>
<tr>
<th>Attribution</th>
<th>Time Point</th>
<th>Math</th>
<th>N</th>
<th>Science</th>
<th>N</th>
<th>English</th>
<th>N</th>
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<td>Ability</td>
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<td>M (SD)</td>
<td>N</td>
<td>M (SD)</td>
<td>N</td>
<td>M (SD)</td>
<td>N</td>
</tr>
<tr>
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<td>Time 2</td>
<td>5.04 (1.71)</td>
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<td>4.24 (1.84)</td>
<td>299</td>
<td>4.82 (1.53)</td>
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<tr>
<td>Ability</td>
<td>Time 3</td>
<td>4.79 (1.59)</td>
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<td>4.28 (1.80)</td>
<td>431</td>
<td>4.81 (1.33)</td>
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<td>Ability</td>
<td>Time 4</td>
<td>4.72 (1.57)</td>
<td>325</td>
<td>4.46 (1.68)</td>
<td>323</td>
<td>5.07 (1.25)</td>
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<tr>
<td>Effort</td>
<td>Time 1</td>
<td>5.49 (1.81)</td>
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<td>6.40 (1.35)</td>
<td>374</td>
<td>6.14 (1.28)</td>
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<tr>
<td>Effort</td>
<td>Time 2</td>
<td>4.67 (1.92)</td>
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<td>300</td>
<td>5.64 (1.34)</td>
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<tr>
<td>Effort</td>
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<td>4.56 (1.78)</td>
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<td>Effort</td>
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<td>4.92 (1.63)</td>
<td>324</td>
<td>5.63 (1.48)</td>
<td>322</td>
<td>5.47 (1.19)</td>
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Table 2

Pearson Correlations for Ability Attributions in Math, Science and English Across Four Time Points

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<tr>
<td>6. Time 2 Sci Ability</td>
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<tr>
<td>7. Time 3 Sci Ability</td>
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<td>8. Time 4 Sci Ability</td>
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<td>9. Time 1 Eng Ability</td>
<td>.302** .104 -.001 .019 .354** .163** .060 .124</td>
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Table 3

Pearson Correlations for Effort Attributions in Math, Science and English Across Four Time Points

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<tr>
<td>3. Time 3 Math Effort</td>
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<tr>
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<td>.116 .281** .424**</td>
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<td>5. Time 1 Sci Effort</td>
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<tr>
<td>6. Time 2 Sci Effort</td>
<td>.148* .294** .126 .218** .117*</td>
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<tr>
<td>8. Time 4 Sci Effort</td>
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<tr>
<td>9. Time 1 Eng Effort</td>
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<tr>
<td>10. Time 2 Eng Effort</td>
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<tr>
<td>11. Time 3 Eng Effort</td>
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<tr>
<td>12. Time 4 Eng Effort</td>
<td>.156* .131 .227** .332** -.073 .271** .219** .513** .179* .176* .312**</td>
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### Table 4

*Fit Indices for Success Ability Attribution Models; Model 4 Shows the Best Fit for These Data*

<table>
<thead>
<tr>
<th>Model</th>
<th>Chi-sq</th>
<th>df</th>
<th>IFI</th>
<th>TLI</th>
<th>1-RMSEA</th>
<th>BIC</th>
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</thead>
<tbody>
<tr>
<td>1- General Ability Attributions</td>
<td>263.94***</td>
<td>51</td>
<td>.667</td>
<td>.559</td>
<td>.913 (.903, .924)</td>
<td>-58.509</td>
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<tr>
<td>2- Specific Ability Attributions</td>
<td>207.09***</td>
<td>45</td>
<td>.705</td>
<td>.573</td>
<td>.920 (.908, .930)</td>
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<tr>
<td>3- Developmental Attributions</td>
<td>162.52***</td>
<td>47</td>
<td>.821</td>
<td>.741</td>
<td>.934 (.922, .946)</td>
<td>-134.645</td>
</tr>
<tr>
<td>4- General Ability Att. w/ Correlated Errors</td>
<td><strong>35.21</strong></td>
<td><strong>33</strong></td>
<td><strong>.997</strong></td>
<td><strong>.993</strong></td>
<td><strong>.989 (.996, 1)</strong></td>
<td><strong>-173.430</strong></td>
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<tr>
<td>5- Developmental Attributions with Correlated Errors</td>
<td>144.34***</td>
<td>44</td>
<td>.845</td>
<td>.759</td>
<td>.936 (.924, .947)</td>
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<tr>
<td>6- General Ability Att. w/ Unique Factors</td>
<td>82.99***</td>
<td>42</td>
<td>.937</td>
<td>.897</td>
<td>.958 (.945, .972)</td>
<td>-182.561</td>
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<tr>
<td>7- Developmental Attributions with Unique Factors</td>
<td>125.60***</td>
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<td>.867</td>
<td>.757</td>
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<td>8- Specific Ability Attributions Quasi Simplex Model</td>
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<td>N/A</td>
<td>N/A</td>
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</tbody>
</table>

*Note. When a model is specified correctly, the Chi-square should not be significant. Values approximating 1 are indicative of excellent model fit by the Incremental Fit Index (IFI), the modified root mean square error of approximation (1-RMSEA) and the Tucker-Lewis Index (TLI). New Bayesian Information Criterion (BIC) scores less than 0 favor the hypothesized model. N/A- fit indices not available because the model did not converge.***p < .001*
Table 5

*Fit Indices for Two-Subject, Quasi-Simplex Attribution Models for Success Ability Attributions; The Domain-General Models (4 & 6) Show the Best Fit for These Data*

<table>
<thead>
<tr>
<th>Model</th>
<th>Chi-sq</th>
<th>df</th>
<th>IFI</th>
<th>TLI</th>
<th>1-RMSEA</th>
<th>BIC</th>
</tr>
</thead>
<tbody>
<tr>
<td>1- Math and English Quasi-Simplex Model</td>
<td>79.455***</td>
<td>22</td>
<td>.828</td>
<td>.776</td>
<td>.932 (.915, .947)</td>
<td>-59.642</td>
</tr>
<tr>
<td>2- Math and English Domain-General w/Correlated Errors Model</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
</tr>
<tr>
<td>3- Math and Science Quasi-Simplex Model</td>
<td>84.991***</td>
<td>22</td>
<td>.827</td>
<td>.776</td>
<td>.928 (.912, .944)</td>
<td>-54.105</td>
</tr>
<tr>
<td>4- Math and Science Domain-General Model w/Correlated Errors Model</td>
<td>2.130</td>
<td>5</td>
<td>1.00</td>
<td>1.04</td>
<td>1.00 (.965, 1.00)</td>
<td>-29.483</td>
</tr>
<tr>
<td>5- English and Science Quasi-Simplex Model</td>
<td>95.638***</td>
<td>22</td>
<td>.787</td>
<td>.724</td>
<td>.922 (.906, .938)</td>
<td>-43.459</td>
</tr>
<tr>
<td>6- English and Science Domain-General Model w/Correlated Errors Model</td>
<td>6.736</td>
<td>5</td>
<td>.995</td>
<td>.971</td>
<td>.975 (.932, 1.00)</td>
<td>-24.877</td>
</tr>
</tbody>
</table>

Note. When a model is specified correctly, the Chi-square should not be significant. Values approximating 1 are indicative of excellent model fit by the Incremental Fit Index (IFI), the modified root mean square error of approximation (1-RMSEA) and the Tucker-Lewis Index (TLI). New Bayesian Information Criterion (BIC) scores less than 0 favor the hypothesized model. N/A—fit indices not available because the model did not converge.

***p < .001
Table 6

*Fit Indices for Success Effort Attribution Models; Model 1 shows the best fit for these data*

<table>
<thead>
<tr>
<th>Model</th>
<th>Chi-sq</th>
<th>df</th>
<th>IFI</th>
<th>TLI</th>
<th>1-RMSEA</th>
<th>BIC</th>
</tr>
</thead>
<tbody>
<tr>
<td>1- General Effort Attributions</td>
<td>58.58</td>
<td>51</td>
<td>.988</td>
<td>.984</td>
<td>.984 (.967, 1.00)</td>
<td>-236.240</td>
</tr>
<tr>
<td>2- Specific Effort Attributions</td>
<td>236.35***</td>
<td>45</td>
<td>.692</td>
<td>.556</td>
<td>.913 (.902, .923)</td>
<td>-96.126</td>
</tr>
<tr>
<td>3- Developmental Effort Attributions</td>
<td>66.993*</td>
<td>47</td>
<td>.975</td>
<td>.965</td>
<td>.972 (.958, .991)</td>
<td>-230.167</td>
</tr>
</tbody>
</table>

Note. When a model is specified correctly, the Chi-square should not be significant. Values approximating 1 are indicative of excellent model fit by the Incremental Fit Index (IFI), the modified root mean square error of approximation (1-RMSEA) and the Tucker-Lewis Index (TLI). New Bayesian Information Criterion (BIC) scores less than 0 favor the hypothesized model.

***p < .001  
*p < .05
Figure 1. Domain-general model for success ability attributions and success effort attributions.

Figure 2. Domain-specific model for success ability attributions and success effort attributions with cross-lagged paths. The cross-lagged paths were included in order to test the least constrained model. All exogenous latent variables are correlated.
Figure 3. Developmental model for success ability attributions and success effort attributions.
Figure 4. Domain-general model with correlated errors for success ability attributions and success effort attributions.
Figure 5. Developmental model with correlated errors for success ability attributions and success effort attributions.
Figure 6. Domain-general model with a unique factor for each academic subject: success ability attributions and success effort attributions.
Figure 7. Developmental model with a unique factor for each academic subject: success ability attributions and success effort attributions.
Figure 8. Simultaneous equations with a quasi-simplex model representing each academic subject. This model assumes equality of error variance across time. The cross-lagged paths were included in order to test the least constrained model. All exogenous latent variables are correlated.
Figure 9. Domain-general attributions with correlated errors model: theoretical model with standardized coefficients. **p < .01, ***p < .001.

Figure 10. Domain-general effort attributions model: theoretical model with standardized coefficients. ***p < .001.
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