Peering into the Black Box of Grit:
How Does Grit Influence the Engagement of Undergraduates?

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In the search to build a successful student body, admissions leaders at postsecondary institutions have long been focused on identifying skills and traits beyond grades and standardized test scores to recognize students with the potential to succeed at their institutions. One non-cognitive factor with this potential is grit. The increasing popularity of grit, coined by Angela Duckworth, has encouraged admissions officers to review non-cognitive factors to build a diverse class and create an engaging campus community (Powell, 2013; Sedlacek, 2017; Wick, 2015). Grit has the potential to assist colleges in selecting a more well-rounded student body as it may help identify students from historically marginalized backgrounds with a strong likelihood of success. Due to grit’s focus on stamina and commitment to long-term goals, characteristics pertinent to college retention and completion, grit can theoretically identify students with a strong probability of succeeding despite some deficiency in the criteria traditionally used in admissions decisions.

The focus on grit has expanded beyond the admissions and research communities. The United States Department of Education promoted using grit as a tool to prepare future generations for college and beyond (US Department of Education, 2013). Furthermore, grit has been the focus of a TED talk with over 12 million views (Duckworth, 2013), numerous articles in the popular press (e.g., Del Giudice, 2014; Engber, 2016), and earned Duckworth the MacArthur “Genius” grant (MacArthur Foundation, 2017). Despite the growth in the popularization of grit, the concept has not been exposed to the empirical scrutiny it deserves if it is going to be a factor in high stakes decisions within the educational research community.

Duckworth, Peterson, Matthews, and Kelly’s (2007) initial research on grit has examined this
concept with high-achieving populations such as West Point cadets and Scripps Spelling Bee finalists, populations that are not representative of the typical student. Furthermore, much of the current research examines smaller populations ranging in the hundreds. Consequently, in this study, we utilized the National Survey of Student Engagement’s ability to collect data on over 11,000 undergraduates to test the construct validity of grit and its concurrent validity for measures of engagement, self-perceived gains, time use, and GPA using a sample that represents a more typical college student population.

Literature Review

Grit is the “perseverance and passion for long-term goals” (Duckworth et al., 2007, p. 1087). Grit is a non-cognitive personality trait that is operationalized as a high-order construct with two lower order features, perseverance of effort and consistency of interest (Duckworth et al., 2007; Duckworth & Quinn, 2009). Perseverance of effort refers to an individual’s tendency to work hard in the face of setbacks or obstacles while the latter, consistency of interest, is the tendency not to change goals and interest frequently (Duckworth et al., 2007; Duckworth & Quinn, 2009). These facets theoretically work together to influence an individual’s attitude and behavior towards long-term goals. One of the attractive features of grit is the lack of correlation with other measures of intelligence and that it is a trait that can be potentially changed (Duckworth, 2016; Duckworth et al., 2007). Consequently, grit can theoretically help all students succeed. Previous research has correlated grit with outcomes like persistence in higher education and success in long-term, difficult tasks (Duckworth et al., 2007, Duckworth & Quinn, 2009). However, the grit concept has been criticized from a variety of perspectives (Credé et al., 2016; Ris, 2015). Below, we summarize grit and its two sub-constructs, perseverance of effort and
consistency of interest, review previous research of how grit influences outcomes, and discuss multiple criticisms of grit.

**The Grit Constructs**

According to Duckworth et al. (2007; 2009), grit is operationalized as a construct consisting of two sub-constructs: perseverance of effort and consistency of interest. Working collaboratively, perseverance of effort is viewed as the tendency to overcome initial failures to achieve long-term goals, while consistency of interest focuses on an individual’s tendency to pursue the same goals across time.

**Grit and Academics**

Studies focusing on grit among college students have shown mixed outcomes. Studies focusing on grit among college students have shown that grittier students more frequently persist and succeed academically (Bowman, Hill, Denson, and Bronkema, 2015; Cross, 2014; Duckworth, 2007; Strayhorn, 2014). Duckworth and colleagues’ (2007) study on grit and various high-achieving populations such as the West Point cadets in training and Scripps Spelling Bee finalist found that grit predicted completion of their training program for the cadets and the number of hours the spelling bee students practiced after holding constant other factors. Bowman et al. (2015) found that grit was positively correlated with academic adjustment, GPA, satisfaction, sense of belonging, and more frequent faculty-student interaction and co-curricular engagement. However, they found that these effects were attributable to the perseverance of effort dimension, not consistency of interest. Strayhorn’s (2014) study examined grit using a sample of African American males and found that grittier black males had higher college grade point average than their counterparts. Cross’ (2014) study examined the relationship between grit scores, GPA, and gender. The results showed that there was a significant and positive
relationship between grit and GPA for females but not for males. He also discovered a positive relationship between grit and doctoral degree attainment (Cross, 2014).

Wolter and Hussain (2014) investigated grit and its relations to college students’ self-regulated learning and academic achievement. Self-regulated learning is the process in which “students take an active, purposeful role in managing motivational, cognitive, and behavioral aspects of their learning” (Wolter & Hussain, 2014, p. 295). This management consisted of students engaging in different sub-processes such as goal setting, the activation of prior knowledge, progress monitoring, engagement and regulation of learning strategies, and reflection to learn. Their results showed that grittier students were less likely to procrastinate and had reduced levels of delay in beginning and completing academic tasks. They concluded self-regulated learning appears to mediate the relationship between grit and academic performance.

In contrast, other studies have not demonstrated a correlation between grit and academic outcomes (Bazelais, David, & Tenzin, 2016; Cross, 2013; Stewart, 2015). Stewart (2015) found that high school GPA and test scores were predictors of college academic performance but not grit. This finding was replicated by Bazelais, Lemay, and Doleck (2016) who examined the relationship between grit and GPA and final exam scores for students taking a gateway physics course. Among graduate students, Cross (2013) found no relationship between grit and doctoral students’ dissertation completion.

Recent studies on grit have examined its cross-cultural applicability, an important facet if admissions staff seek to use grit as a way to boost enrollment of students from historically marginalized communities. Datu, Valdez, and King (2015) examined grit’s validity for a sample of Filipino undergraduate and high school students from a collectivist culture. Their results showed that grittier students were more likely to achieve higher satisfaction in life and emotional
well-being. Additionally, these effects comport with Bowman and colleague’s results in that the effects of grit appear to be isolated to the perseverance of effort dimension. O’Neal and colleagues (2014) examined the relationships between grit, stress, and depression and GPA using a sample of documented and undocumented Latinx first-generation students. They found high levels of grit among both documented and undocumented Latinx students. Additionally, the relationship between grit and depression was negative as undocumented Latinx students with lower levels of grit were more likely to report higher levels of depression than documented Latinx students.

**Grit and Engagement**

To date, limited research has focused on grit’s influence on engagement in educationally purposeful activities (Hodge, Wright & Bennet, in press; Robinson, 2015). As mentioned above, Bowman and colleagues (2015) found that the perseverance of effort dimension of grit was positively correlated with both faculty-student interaction and co-curricular engagement. Holbein and colleagues (2016) studied the relationship between grit, middle and high school students’ school and civic engagement. Their findings show that grittier students performed noticeably better on a standardized test, had higher levels of school attendance, and a stronger belief of their future engagement in the political process. Hodge, Wright, and Bennet’s (in press) study about the engagement of Australian university students found a positive relationship between grit and engagement, which positively correlated with better academic productivity. Moreover, Robinson (2015) found a strong association between grit and engagement in coursework for nursing students.
Critiques of Grit

The grit construct has been hailed as the next new thing in the research literature and popular press (e.g., Duckworth et al., 2007; Duckworth et al., 2009; Del Giudice, 2014; Engber, 2016; Powell, 2013; Sedlacek, 2017; Wick, 2015). However, these claims about grit have not always stood up to scrutiny. Credé and colleagues (2016) conducted a meta-analysis of 88 studies focusing on grit and critiqued grit’s validity. They claim that the relationship between grit and success has been overstated by Duckworth, as the correlation of grit with academic success is less than .20 in their meta-analysis.

Credé and colleagues (2016) also contest the uniqueness of grit due to its correlation (.84) with the conscientiousness dimension of the Big Five personality traits and suggest that grit is an old, but repackaged concept. Individuals that are conscientious are “thorough, careful, reliable, organized, industrious and self-controlled” (Duckworth et al., 2007, p. 1089). Individuals that are conscientiousness are highly achievement-oriented, a hallmark of grit (Cross, 2014). However, grit’s proponents argue that grit includes self-control traits that makeup conscientiousness, but also focuses on the long-term stamina rather than short-term intensity (Cross, 2014; Duckworth et al., 2007).

Others have contested the construct validity of grit. Credé and colleagues (2016) point out that Duckworth et al.’s confirmatory factor analysis (CFA) of the Short Grit Scale was invalid. Their CFA model used a higher-order factor structure with two first-order factors which produces an unidentified model without the imposition of unusual constraints¹. Muenks,

¹ Higher-order factors require three subscales for a CFA model to be identified, without the imposition of unusual constraints. When Duckworth and Quinn’s (2009) model is replicated in AMOS 24 (Duckwork and Quinn used version 6.0), the following error message is produced along with the results: “Minimization was unsuccessful. The results that follow are therefore incorrect. The model is probably unidentified. In order to achieve identifiability, it will probably be necessary to impose 2 additional constraints.” Consequently, the results associated with the published higher-order factor grit solution are highly suspect, especially given that a standardized loading is greater than 1.
Wigfield, Yang, and O’Neal (2017) further examined the factor structure of grit, its relation to other constructs, and the best factor structure model of grit. In their study, high school and college students’ grit, conscientiousness, self-control, cognitive regulations, effort regulation, and behavioral engagement and disaffection were measured via a survey. Results showed that for high school students the two correlated-factor model was the best fit for measuring grit while the bi-factor model was the best fit for college students. Consequently, their findings indicate that the grit scale is configurally variant between different populations condition, suggesting that Duckworth’s grit scales lack construct validity. Moreover, Muenks and colleagues. (2017) found that grit and its subscales significantly overlapped with personality, self-regulation and engagement literature.

Grit has alternately been critiqued from a class-reproduction standpoint as it helps legitimate existing inequalities throughout society (Gonzales-Stokas, 2015; Ris, 2015; Socal, 2014). This legitimization of existing qualities can be reproduced through the fundamental attribution error -the tendency to overvalue personality-based explanations for behaviors and situations while ignoring the institutional and systemic constructs that act as barriers to an individual’s aspirations (Gonzalez-Stokas, 2015). An example of fundamental attribution error in the education system is the unethical action of telling “children who face a society of entrenched economic inequality, that achievement is the result of individual effort and is disconnected from systemic privilege” (Gonzalez-Stokas, 2015, p. 516).

Despite these critiques, grit has captured the imagination of college administrators, policymakers, and the popular press. Consequently, it is essential to investigate the efficacy of grit’s relationship with college outcomes. The study of grit has examined various types of populations and circumstances, yet these analyses have primarily focused on a narrow population
of interest (i.e., high achieving students at a military academy or African American males) or students attending a single institution. Additionally, there is limited research about grit’s influence on undergraduate student engagement (Wang & Degol, 2016). This lack of research is concerning, as it limits the potential to understand how grit influences student outcomes. As grit is not correlated with intelligence, grit must be related to behavioral outcomes if it ultimately influences academic success. Consequently, we choose to fill in these research gaps by investigating the construct validity of the most popular version of grit scale and its concurrent validity by investigating its association with students’ engagement in effective educational practices, perceived gains, time usage, and GPA for a diverse, multi-institutional sample of undergraduates attending bachelor’s degree-granting institutions.

**Conceptual Framework**

At its core, this study is an investigation into the validity of grit. Our validity inquiries were guided by Messick’s (1989) unified validity framework. In the framework, Messick moved beyond Cureton (1951) and Cronbach’s (1971) notions of validity which focus on the characteristics of a measure. While traditional criterion validity is an important aspect of Messick’s framework, he also emphasized the need to focus on how a measure is interpreted and used in practice. Messick (1995) offers a progressive matrix of validity that ranges from (1) construct validity, (2) construct validity and relevance/utility, (3) construct validity and value implications, (4) construct validity, relevance/utility, value implications, and social consequences. Messick’s notion of validity has gone on to inform the current jointly created *Standards for Educational and Psychological Testing* (American Educational Research Association, American Psychological Association & National Council on Measurement in Education, 2014).
From this unified view of validity, it is important to assess multiple facets of validity in relationship to grit. The questions to inquire about include:

- Whether the Short Grit Scale measures grit (the latent construct)?
- Is the theoretical basis of grit sound?
- Does grit have concurrent, discriminant, and predictive validity?
- Is grit generalizable across groups? Does grit discriminate against certain populations?
- What are the risks associated with using grit in a practical setting?

Until these questions are fully answered, using grit in high stakes decisions may lead to unintended consequences.

**Purpose**

In this study, we sought to examine the construct and concurrent validity of the Short Grit Scale for a large, diverse, multi-institutional sample of college students. To analyze the construct validity of grit, we utilized confirmatory factor analysis (CFA) to test if the scale meets generally accepted thresholds for construct validity using the largest and most diverse data set on grit collected (Credé et al., 2017). Additionally, we conducted multi-group CFAs to test if the relationships differ between subpopulations of students to investigate the measurement invariance of the Short Grit Scale. Next, we examine the concurrent validity of the grit scale by examining its relationship with process indicators of student engagement, self-perceived gains, time spent studying, and GPA. As the hallmarks of grit are perseverance of effort and consistency of interest, we hypothesize that grit is a factor leading to student effort and time dedicated to study, which ultimately leads to learning and development. However, to date, this relationship has not been extensively tested. Consequently, we examined the association between the psychological concept of grit and the behavioral dimensions of student engagement in
effective educational practices. Establishing this link between the psychological realm and actual behaviors is important to understanding how students learn. Additionally, if this association does not exist, it calls into question the validity of grit as a standalone concept.

**Methods**

**Data**

We utilized data from the National Survey of Student Engagement (NSSE) administered in the winter and spring of 2016. NSSE is a multi-institutional study of first-year and senior students attending bachelor’s-granting institutions that examines how often students engage in educationally beneficial activities, students’ time-use, and their perceptions of the campus environment. Due to our focus on grit, we focused our analyses on students who attended one of 38 institutions that received a supplemental set of items that included the Short Grit Scale (Duckworth & Quinn, 2009). These institutions were randomly selected from a pool of institutions that did not elect to append two additional item sets (topical modules and/or consortium items) to the core NSSE instrument. A total of 4,668 first-year and 7,082 senior students responded to the item set. The response rate for first-year and senior students was 21% and 24%, respectively.

The students in our sample attended a diverse set of institutions. Roughly a quarter attended institutions that awarded doctoral degrees, over half attended master’s colleges and universities, 15% were enrolled at baccalaureate colleges, and 5% attended special focus institutions. Just over 40% of the sample was enrolled at a public college. A plurality of students attended institutions with undergraduate enrollments between 5,000 and 9,999 students. Half of the students attended institutions with a Barron’s rating of competitive. A third of the sample attended a minority serving institution.
Slightly less than two-thirds of the sample were female. White students comprised 57% of the respondents. Asian, Black, Latinx, and multi-racial students represented 6%, 11%, 12%, and 7% of the sample respectively. Five percent of the sample were international. Slightly over a quarter of the sample did not have a parent who enrolled in college. Twelve percent of the students had parents with some college education. One in ten students had a parent who received an associate’s degree. Over a quarter of the sample had a parent who earned a bachelor’s degree. Slightly less than one in four students had a parent with a master’s degree or higher.

We utilized two sets of key variables in our analyses. First, grit was represented by the 8 items in Duckworth and Quinn’s (2009) Short Grit Scale. These items were lightly edited from the original scale to conform to NSSE’s style and appended to the end of NSSE (see Appendix A for the items wording). We created subscale scores by taking the means of the component items for the items included in the consistency of interest and perseverance of effort subscales. Additionally, we reverse coded the items in the consistency of interest subscale, so that they indicate higher levels of this trait.

Our second set of key variables were process indicators of student learning and development. These variables were represented by 9 of the 10 NSSE Engagement Indicators (we excluded Effective Teaching Practices, as this scale focuses on instructors’ efforts), a perceived gains scale, time spent preparing for class, and self-reported grades. We chose to focus on measures of engagement as they have been found to predict students learning and development and retention (Kuh, Kinzie, Cruce, Shoup & Gonyea, 2006; National Survey of Student Engagement, n.d.; Passarela, Seifert, & Blaich, 2010). Additionally, students’ grades have long been a proxy for student learning and development. Information on reliability and validity of the Engagement Indicators is available in NSSE’s (2017) Psychometric Portfolio. The items in the
perceived gains scale asked about how much their college experience contributed to students’ knowledge, skills, and personal development in 10 areas. The Cronbach’s α for the perceived gains scale was .91 for first-year students and .90 for seniors. Students’ time spent preparing for class was captured in ranges (0, 1-5, 6-10, 11-15, 16-20, 21-25, 26-30, more than 30 hours per week) and recoded to the midpoint (the top category was set to 33 hours per week). Students were asked to report their typical grades in the following categories: A, A-, B+, B, B-, C+, C, C- or lower. We recoded these values to reflect the typical 4.0 GPA scale. To aid in the interpretation of the results, we standardized both grit subscales, the Engagement Indicators, perceived gains scale, time spent preparing for class, and grades to have a mean of 0 and standard deviation of 1.

In addition to these key variables, we also used data on a variety of control variables: race/ethnicity, sex, standardized test score (SAT/ACT), parental education, academic major, greek-life participation, age, athletics participation, transfer status, part-time status, educational aspirations, and on-campus residency.

Analyses

**Construct Validity.** We began our analyses by assessing the construct validity of the Short Grit Scale using confirmatory factor analysis. Duckworth and Quinn (2009) propose the Short Grit Scale as a higher order factor with two subscales: consistency of interest and perseverance of effort. However, this model is unidentified without the imposition of multiple unusual constraints (see the critiques of grit section for more details). Therefore, we conducted a CFA using MPLUS where the subscales were correlated and not subsumed under a higher-order factor. The model was estimated using full information maximum likelihood. We assessed model fit using the following standards (Hu & Bentler, 1999). The model was judged to have good fit if
the comparative fit index (CFI) and Tucker-Lewis index (TLI) were greater than or equal to .95 and the root mean square error of approximation (RMSEA) was less than or equal to .05. The thresholds for adequate fit used were CLI and TLI were greater than or equal to .90 and the RMSEA was less than or equal to .06. Also, we report the $\chi^2$ results for the models; however, due to our large sample size and the sensitivity of the $\chi^2$ statistic to sample sizes, we focused our interpretation on the CFI, TLI, and RMSEA.

Next, we examined the measurement invariance of the Short Grit Scale by replicating the confirmatory factor analysis by subgroup. If grit has applicability for high-stakes decisions, like admissions, it is important to assess whether the scale varies by subgroups, as the use of grit in such decisions could unfairly advantage or disadvantage certain groups of students. Measurement invariance was assessed for students by class level (first-year vs. senior), parental education (first-generation vs. non-first-generation), sex (male vs. female), and race/ethnicity (White vs. non-White). We selected these groups to examine if the scale adequately performs longitudinally and does not discriminate against historically marginalized populations. To assess measurement invariance, we performed the following steps for each group. First, we tested the fit of the model for each group separately. Next, we tested for configural, metric, and scalar invariance by estimating a series of models and comparing their fit indices. The model for configural invariance constrained the factor structure across groups. The model for metric invariance constrained the factor structure and loadings to be equal across groups. The model for scalar invariance constrained the factor structure, loadings, and item intercepts across groups. We used the following thresholds for assessing measurement invariance as suggested by Chen (2007). For metric (loading) invariance, we used the criteria of a change less than or equal to -.010 in CFI and less than or equal to .015 in RMSEA compared to the configural model. The
criteria for scalar (loading and intercept) invariance was a change of less than or equal to -.010 in CFI and less than or equal to .015 in RMSEA compared to the metric model. We also report the \( \Delta \chi^2 \); however, due to the large sample size, we relied on the other fit indices to assess measurement invariance.

**Concurrent Validity.** To assess the concurrent validity of grit, we estimated a series of OLS regression models that predicted NSSE’s Engagement Indicators, a perceived gains scale, GPA, and time spent preparing for class using the grit subscales and the control variables described in the data section. Additionally, due to the multi-level structure of our data, we included institution-specific fixed effects in the models. The fixed effects are essentially dummy variables that control for differences in institution attended and encompass both structural differences in institution type and unobservable qualitative factors. Additionally, we adjusted the standard errors to account for the clustering of students within institutions. All of the concurrent validity analyses were performed separately for first-year and senior students, following NSSE’s standard practices. As we standardized both the outcome variables and grit subscales, the results reported represent the expected standard deviation change in the outcome for a standard deviation change in one of the grit subscales.

**Results**

We present the results from the validity tests in two sections: construct validity and concurrent validity.

**Construct Validity**

We began by assessing the model fit of the Short Grit Scale by performing a confirmatory factor analysis. The analysis revealed that the model fit our data marginally, \( \chi^2(19)=971.839 \ p < .001, \ CFI = .954, \ TLI = .932, \ RMSEA = .065 \ (90\% \ CI .062 - .069). \) The CFI
and TLI values were above our threshold for adequate fit (.90), but the confidence interval for the RMSEA did not include ≤.06. Additionally, the standardized factor loading for one item (Setbacks don’t discourage me) was extremely low at .10. All other standardized loadings were .59 or greater. The correlation between the two grit subscales was -.51 (the consistency of interest items are reverse coded). The results for the model are visually displayed in Figure 1.

As the fit indices indicated that the fit could be improved and the low loading of the “setbacks don’t discourage me” item (part of the perseverance of effort subscale), we estimated a second CFA without this item. The fit statistics for this analysis were $\chi^2(13) = 630.690 \ p < .001$, CFI = .969, TLI = .951, RMSEA = .064 (90% CI .059 - .068). Removing the low loading item improved the overall model fit and all of the indices met our thresholds for adequate fit. The standardized loadings for this model were all .59 or greater. The correlation between the subscales was -.52. The standardized item loadings for the revised model are displayed in Figure 2.

Next, we assessed the measurement invariance of the short grit scale using the procedures previously described to examine the consistency of the scale across subgroups. Due to the improved fit of the scale without the “setbacks don’t discourage me” item, our analyses were based off the modified model. Table 1 displays the results for baseline models for each group and the combined models testing configural, metric, and scalar invariance.

The baseline model for first-year students indicated adequate fit, while the senior model indicated marginal fit (RMSEA >.06). The configural model for class level just met our threshold for adequate fit (RMSEA lower 90% CI=.060), indicating that the factor structure is adequately or marginally equivalent between first-year and senior students. The change in CFI and RMSEA between the configural model and the metric and scalar models were lower than Chen’s (2007)
suggested threshold indicating that the modified, two-factor grit model has equivalent loadings and intercepts for both the first-year and senior undergraduate population.

Next, we examined the models by parental education status. The baseline model for first-generation students indicated marginal fit (RMSEA > .06) for first-generation students and adequate fit for non-first-generation students. The configural model for parental education barely met our threshold for adequate fit (RMSEA lower 90% CI=.059), indicating that the factor structure is adequately or marginally equivalent between first-generation and non-first-generation students. However, the change in CFI and RMSEA from the configural to the metric and scalar models was relatively unchanged.

The baseline models for males and females indicated adequate or marginal fit for both groups as both RMSEAs were > .06, but the outer edge of the confidence intervals included .06. The configural model also had adequate to marginal fit due to a RMSEA of .064, with a confidence interval of .060 to .068. The fit indices change from the configural to the metric and scalar models were modest.

Finally, we assessed invariance by race/ethnicity. The baseline model for non-White students had adequate fit. However, the model for White students had marginal fit, RMSEA=.069 (90% CI .064 - .075). The configural model also had a marginal fit due to the RMSEA of .065 (90% CI .061 - .069). The indices for the metric and scalar models did not substantially vary from the configural model.

**Concurrent Validity**

Due to the findings from the construct validity analyses, we utilized a modified version of the perseverance of effort subscale in our concurrent validity analyses, as the modified subscale was a better fit to the data. Table 2 contains the results from our multivariate analyses. The
coefficients represent the expected standard deviation change in the Engagement Indicators, the perceived gains scale, time spent preparing for class, and students grades for a standard deviation change in the two grit subscales consistency of interest and perseverance of effort, holding constant other factors.

Our results indicate that consistency of interest has a negligible or relatively weak association with engagement, perceived gains, time spent preparing for class, and grades, after we controlled for other variables. The absolute values of the relationships ranged from .01 for Supportive Environment to .11 for Reflective and Integrative Learning for first-year students. Furthermore, consistency of interest was consistently negatively related to the Engagement Indicators. However, the relationship was positive for time spent preparing for class and grades. The results were relatively similar for seniors as the absolute value of the coefficients ranged from .02 to .10. However, we did not observe any significant and positive relationships between consistency of interest and our dependent variables for seniors.

In contrast to the consistency of interest results, we found a pattern of positive relationships of a larger magnitude for the perseverance of effort subscale. For first-year students, the perseverance of effort estimates ranged from .15 for Quality of Interactions to .32 for Learning Strategies after holding constant other characteristics. For seniors, the estimates ranged from .11 for Quality of Interactions to .24 for Learning Strategies and Grades. All of the relationships were statistically significant at $p < .001$ for both the first-year and senior subsamples.

**Discussion**

Admissions leaders are increasingly focusing on using non-cognitive traits, like grit, as a factor in the admissions process (Powell, 2013; Sedlacek, 2017; Wick, 2015). Grit has the
theoretical potential to help admissions professionals identify talented students who would thrive in college, yet may lack the traditional academic criteria needed for admission to a highly selective college. Despite the push for using grit as a factor in holistic admissions review, the concept has not been fully empirically validated and is the subject of debate within research circles (Credé et al. 2017; Gonzales-Stokas, 2015; Muenks et al., 2017; Ris, 2015; Socal, 2014).

Furthermore, Duckworth’s initial validation studies of grit focused on narrow samples atypical of common educational settings (e.g., West Point cadets, Scripps Spelling Bee finalists), indicating the need to reassess the validity of the scale in more representative population (Duckworth et al., 2007; Duckworth & Quinn, 2009). Therefore, we sought to investigate the validity of grit and, in particular, the Short Grit Scale, using Messick’s (1989) unified framework of validity.

Based on data from nearly 12,000 undergraduates attending a diverse group of U.S. colleges and universities, our study investigated both the construct and concurrent validity of the Short Grit Scale. Though the factor structure of grit has been investigated by others before, we are the first to apply these techniques to a large data set comprised of students attending numerous postsecondary institutions to confirm its factor structure and stability across different student groups. Additionally, prior investigations into grit’s relationship to student engagement in educationally beneficial practices has not been extensively studied. Overall, our results demonstrated that 1) a modified, two correlated factor model for grit’s consistency of interest and perseverance of effort subscales adequately fits the underlying data; 2) the sub-scale scores generally have the same meaning across different student groups (class level, first-generation status, sex, and racial/ethnic minority status); and 3) the perseverance of effort scale is a more powerful predictor of NSSE measures than consistency of interest, although perseverance of
effort’s effects are for the most part small in size and vary between class level and across outcomes.

These findings when integrated with prior research allow us to come to a number of conclusions about grit. First, the Short Grit Scale does not meet generally accepted criteria for use in high-stakes situations. As previously noted by Credé et al. (2017), the proposed factor structure for the Short Grit Scale by Duckworth and Quinn (2009) is invalid (see the Critiques of Grit section for more details). Our confirmatory factor analysis found a poor fit to the data, largely because one item was relatively unrelated to the perseverance of effort subscale. After removing the item, we achieved adequate fit. However, the scale still did not display good fit, which would be needed to utilize the scale in high stakes decisions like admission to highly selective institutions. However, it is appropriate for use in research applications, according to this standard.

Second, we found that one of the grit subscales, perseverance of effort, was significantly and positively related to engagement in educationally purposeful activities, perceived gains, and GPA when we held constant a basket of student characteristics and fixed institutional effects. However, the second subscale was weakly and sometimes negatively related to our dependent variables. Across all outcomes, the average standardized coefficient for perseverance of effort was about .22 and .18 for first-year and senior students, respectively; for consistency of interest the mean coefficients were .07 and .05, respectively. The grit subscales accounted for a significant proportion of the total explained variance for many of the dependent variables. For instance, for first-year students, the subscales accounted for over 10%, 8%, and 7% of the total variation in learning strategies, perceived gains scales, and GPA, respectively.
Our finding that perseverance of effort, but not consistency of interest, is positively related to a host of outcomes comports with numerous prior studies (e.g., Bowman et al, 2015; Credé et al., 2017; Datu et al., 2015; Muenks et al. 2017). Their relationship to engagement is important due to engagement’s association with persistence and student learning (Kuh et al., 2006; National Survey of Student Engagement, n.d.; Passcarella et al., 2010). Furthermore, our measures of engagement were process indicators of student learning, which we would expect students to participate in if grit has positive impacts on college student outcomes. Our concurrent validity results indicates that admissions leaders may want to include students’ perseverance of effort when making admissions decisions in a holistic framework. However, our results do not suggest that perseverance of effort should be a dominant factor in admissions decisions as the partial correlation with our outcomes was not overwhelmingly strong.

Third, our results largely suggest that grit, when measured by the Short Grit Scale, is largely invariant across populations. This feature is critical as it suggests that grit is not biased against important subgroups like historically underrepresented racial and ethnic groups, first-generation college students, and women. As grit has been previously critiqued as a vehicle to legitimize class reproduction, the invariance of grit across these subscales indicates that the critique is not substantiated by the quantitative evidence. Furthermore, it suggests that the use of grit in high stakes decisions, like college admission, should not have a deleterious societal impact and important consideration in Messick’s (1989) validity framework. However, we must be cognizant that grit, as operationalized by the Short Grit Scale, does not meet the standards for use in high stakes decisions; therefore, an invariance analysis should be repeated if an improved scale becomes available.
Given the limited existing research exploring grit’s relationship to college student engagement processes, two potential avenues to explore in the future should be considered. First, we should better understand the potential moderating or mediating effects that perseverance of effort has on various student background/demographic characteristics when explaining student engagement behavior and other student outcomes (i.e., academic major, sex, first-generation status, first generation status). An enticing aspect of grit is its possibility for compensatory effects, where high levels of grit could result in outsized gains for those with lower academic credentials. Second, while Duckworth (2016) has indicated that grit may be malleable, there is limited information about efforts by colleges to increase student grit. Systematic research exploring who is attempting to change students’ levels of grit and how they are going about it can be beneficial for understanding the grit landscape among colleges. Once relevant institutions are identified large scale data collection efforts can be implemented to track students to see if they are positively impacted by school efforts. Other types of academic success programs can be tracked as well to see how grit-related programs compare to them. Credé and colleagues (2017) suggest that other programs that focus on study skills might be a wiser investment in terms of time and funding. Third, our findings in combination with others indicate that the operationalization of grit could be improved (Credé et al., 2017; Muenks et al., 2017). Consequently, future efforts should be made to create an improved grit scale that can conform to strict psychometric scrutiny, given the interest of using non-cognitive factors like grit in high-stakes decisions. Furthermore, much research has demonstrated that the predictive power in grit largely resides in the perseverance of effort component (Bowman et al., 2015; Credé et al., 2017; Muenks et al., 2017), therefore, such future efforts may want to focus on this component of grit.
Conclusion

Despite the desire of admissions professionals to use grit in their decision making processes, the current operationalized version of grit does not appear to be a valid measure for high-stakes decisions. However, one dimension of grit, perseverance of effort, shows some promise in its ability to predict important postsecondary outcomes. Additionally, grit appears to be relatively invariant, suggesting that that grit maybe a pathway to identify historically underrepresented students that show great promise for postsecondary success. Consequently, more future research should be devoted to creating an empirically valid grit scale.
References


Table 1
Fit indices for class level, parental education, sex, and race invariance analyses

<table>
<thead>
<tr>
<th>Model</th>
<th>Overall fit indices</th>
<th>Comparative fit indices</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>( \chi^2 )</td>
<td>df</td>
</tr>
<tr>
<td>Class level</td>
<td></td>
<td></td>
</tr>
<tr>
<td>First-year</td>
<td>237.734***</td>
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</tr>
<tr>
<td>Senior</td>
<td>414.245***</td>
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</tr>
<tr>
<td>1. Configural</td>
<td>651.979***</td>
<td>26</td>
</tr>
<tr>
<td>2. Metric</td>
<td>664.530***</td>
<td>31</td>
</tr>
<tr>
<td>3. Scalar</td>
<td>753.147***</td>
<td>36</td>
</tr>
<tr>
<td>Parental education</td>
<td></td>
<td></td>
</tr>
<tr>
<td>First-generation</td>
<td>352.859***</td>
<td>13</td>
</tr>
<tr>
<td>Non-first-generation</td>
<td>279.646***</td>
<td>13</td>
</tr>
<tr>
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<td>26</td>
</tr>
<tr>
<td>2. Metric</td>
<td>646.436***</td>
<td>31</td>
</tr>
<tr>
<td>3. Scalar</td>
<td>749.836***</td>
<td>36</td>
</tr>
<tr>
<td>Sex</td>
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<td></td>
</tr>
<tr>
<td>Female</td>
<td>408.30***</td>
<td>13</td>
</tr>
<tr>
<td>Male</td>
<td>246.177***</td>
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</tr>
<tr>
<td>1. Configural</td>
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<td>26</td>
</tr>
<tr>
<td>2. Metric</td>
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<td>31</td>
</tr>
<tr>
<td>3. Scalar</td>
<td>829.624***</td>
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<tr>
<td>Race/ethnicity</td>
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<tr>
<td>Non-white</td>
<td>231.419***</td>
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<tr>
<td>White</td>
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</tr>
<tr>
<td>1. Configural</td>
<td>660.079***</td>
<td>26</td>
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<tr>
<td>2. Metric</td>
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</tr>
<tr>
<td>3. Scalar</td>
<td>793.149***</td>
<td>36</td>
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</table>
Note: $\chi^2$ = Chi-square; $df$ = degrees of freedom; RMSEA = root mean square error of approximation; CFI = comparative fit index; TLI = Tucker-Lewis index; $\Delta \chi^2$ and $\Delta df$ = change in $\chi^2$ and degrees of freedom between models; $\Delta CFI$ = change in CFI score between models; $\Delta RMSEA$ = change in RMSEA score between models.

* $p < .05$, ** $p < .01$, *** $p < .001$
### Table 2

*Fixed Effect Estimates of the Relationship between Grit and Student Engagement and Perceived Gains for First-Year and Senior Students*

<table>
<thead>
<tr>
<th></th>
<th>Consistency of interest</th>
<th>Perseverance of effort</th>
<th></th>
<th></th>
<th></th>
<th></th>
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<tr>
<td></td>
<td>Est.</td>
<td>Sig.</td>
<td>SE</td>
<td>Est.</td>
<td>Sig.</td>
<td>SE</td>
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<td><strong>First-year students</strong></td>
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<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Higher-Order Learning</td>
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<td>**</td>
<td>0.02</td>
<td>0.26</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Reflective &amp; Integrative Learning</td>
<td>-0.11</td>
<td>***</td>
<td>0.02</td>
<td>0.25</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Quantitative Reasoning</td>
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<td>***</td>
<td>0.02</td>
<td>0.22</td>
<td>***</td>
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<tr>
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<td>0.02</td>
<td>0.32</td>
<td>***</td>
<td>0.02</td>
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<tr>
<td>Collaborative Learning</td>
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<td>***</td>
<td>0.02</td>
<td>0.19</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Discussions w/ Diverse Others</td>
<td>-0.07</td>
<td>***</td>
<td>0.02</td>
<td>0.16</td>
<td>***</td>
<td>0.02</td>
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<tr>
<td>Student-Faculty Interaction</td>
<td>-0.05</td>
<td>**</td>
<td>0.02</td>
<td>0.20</td>
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<tr>
<td>Quality of Interactions</td>
<td>-0.07</td>
<td>***</td>
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<td>Supportive Environment</td>
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<td></td>
<td>0.02</td>
<td>0.21</td>
<td>***</td>
<td>0.02</td>
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<tr>
<td>Perceived gains</td>
<td>-0.06</td>
<td>***</td>
<td>0.02</td>
<td>0.30</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Time spent: preparing for class</td>
<td>0.08</td>
<td>***</td>
<td>0.02</td>
<td>0.16</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>GPA</td>
<td>0.09</td>
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<td>0.02</td>
<td>0.26</td>
<td>***</td>
<td>0.02</td>
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<tr>
<td><strong>Seniors</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Higher-Order Learning</td>
<td>-0.05</td>
<td>*</td>
<td>0.02</td>
<td>0.17</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Reflective &amp; Integrative Learning</td>
<td>-0.06</td>
<td>**</td>
<td>0.02</td>
<td>0.17</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Quantitative Reasoning</td>
<td>-0.07</td>
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<td>0.02</td>
<td>0.17</td>
<td>***</td>
<td>0.02</td>
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<td>Learning Strategies</td>
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<td>0.24</td>
<td>***</td>
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<tr>
<td>Collaborative Learning</td>
<td>-0.10</td>
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<td>0.02</td>
<td>0.17</td>
<td>***</td>
<td>0.02</td>
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<tr>
<td>Discussions w/ Diverse Others</td>
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<td></td>
<td>0.02</td>
<td>0.14</td>
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<tr>
<td>Student-Faculty Interaction</td>
<td>-0.08</td>
<td>***</td>
<td>0.02</td>
<td>0.19</td>
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<td>0.02</td>
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<tr>
<td>Quality of Interactions</td>
<td>0.00</td>
<td></td>
<td>0.02</td>
<td>0.11</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Supportive Environment</td>
<td>-0.05</td>
<td>*</td>
<td>0.02</td>
<td>0.16</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Perceived gains</td>
<td>-0.08</td>
<td>***</td>
<td>0.02</td>
<td>0.20</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>Time spent: preparing for class</td>
<td>0.03</td>
<td></td>
<td>0.02</td>
<td>0.18</td>
<td>***</td>
<td>0.02</td>
</tr>
<tr>
<td>GPA</td>
<td>0.03</td>
<td></td>
<td>0.02</td>
<td>0.24</td>
<td>***</td>
<td>0.02</td>
</tr>
</tbody>
</table>

*Note: All dependent variables, consistency of interest, and perseverance of effort were standardized with a mean of 0 and standard deviation of 1. ΔR² = change in explained variance after the grit subscales were added to the model. Control variables included race/ethnicity, sex,*
standardized test score (SAT/ACT), parental education, major, greek-life participation, age, athletics participation, transfer status, part-time status, educational aspirations, and on-campus residency. Models included fixed institutional effects.

* $p < .05$, ** $p < .01$, *** $p < .001$

1Modified from Duckworth’s proposed subscale. Excludes “setbacks don’t discourage me” due to improved model fit without the item (see construct validity section for more details).
Figure 1

Confirmatory Factor Analysis of Duckworth and Quinn’s (2009) Short Grit Scale

Note: CI = consistency of interest; PE = Perseverance of effort; See Appendix A for item wordings
Figure 2

Confirmatory Factor Analysis of the Modified Short Grit Scale

\[\text{Note: CI} = \text{consistency of interest; PE} = \text{Perseverance of effort; See Appendix A for item wordings}\]
Appendix A

Items in the Short Grit Scale

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable Label</th>
<th>Values and labels</th>
</tr>
</thead>
<tbody>
<tr>
<td>GRM1601a</td>
<td>New ideas and projects distract me from previous ones^1</td>
<td></td>
</tr>
<tr>
<td>GRM1601b</td>
<td>Setbacks don’t discourage me.</td>
<td></td>
</tr>
<tr>
<td>GRM1601c</td>
<td>I have been obsessed with a certain idea or project for a short time but later lost interest^1</td>
<td></td>
</tr>
<tr>
<td>GRM1601d</td>
<td>I am a hard worker.</td>
<td></td>
</tr>
<tr>
<td>GRM1601e</td>
<td>I set goals but later choose to pursue different ones^1</td>
<td></td>
</tr>
<tr>
<td>GRM1601f</td>
<td>I have difficulty maintaining my focus on projects that take more than a few months to complete^1</td>
<td></td>
</tr>
<tr>
<td>GRM1601g</td>
<td>I finish whatever I begin.</td>
<td></td>
</tr>
<tr>
<td>GRM1601h</td>
<td>I am diligent.</td>
<td></td>
</tr>
</tbody>
</table>

^1Reverse coded in the concurrent validity analyses