FEATURE ARTICLE

Active Learning, Students Who Are Academically At-Risk, and Institutional Classification

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ABSTRACT

In this study, self-reported survey results from the National Survey of Student Engagement (NSSE) 2017 and 2018 are examined to understand the extent to which students who were academically at-risk and academically prepared engaged in active learning versus traditional learning methods across bachelor's, master's, and doctoral degree-granting institutions. The NSSE Report Builder Public (2018) was utilized to create a data set from first year student responses selecting for teaching methodologies, Carnegie Institutional Categories, and student academic level as determined by course grades. Researchers used chi-square analyses to establish associations between the variables; all chi-square results were statistically significant except for one; there was no association found between students who were academically at-risk and coursework that emphasized evaluative learning activities. Next, researchers analyzed the frequencies of types of learning activities reported by students. Students who were were academically at-risk reported lower frequencies of using active learning techniques and tended to engage in study for fewer hours across all institution types. From this analysis, suggestions for improving the instruction for students who are academically at-risk include increased use of active learning teaching strategies for the various types of degree-granting institutions.

Keywords: students who are academically at-risk, active learning, Carnegie Institutional Categories

n higher education, institutions at all levels are challenged to meet students' academic, social, and personal needs. Whether the institution grants bachelor's, master's, or doctoral degrees, more and more students enroll with academic needs, Brothen and Wambach (2012) stated, "Policy makers can argue over which institutions should provide access to nontraditional students, but the reality is that most institutions will serve at least some students who are underprepared relative to their peers" (p. 38). Brothen and Wambach further suggested it is not important to worry about whether an institution will serve students who are underprepared but rather how are they going to reach the students and bring them up to college readiness. With students who are underprepared needing assistance to reach college-level readiness, teaching students active learning strategies may be a way to reduce academic barriers and to increase overall student success. Indeed, when compared to traditional teaching methods, active learning techniques were found to enhance student performance (McCarthy & Anderson, 2020).

With the nation's focus on preparing a better workforce, redesigning first year and developmental coursework by embedding active learning strategies will help students get a better start on their college educations (Lumnia Foundation, n.d.). The purpose of this study is to consider what types of instruction and learning activities are reported by students who are academically successful and academically at-risk in the first year of studies at degree-granting higher education institutions in the United States, specifically at bachelor's-, master's-, or doctoral- degree granting institutions. By considering types of instruction, namely active learning and traditional modes of learning, the

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researchers provide implications for instruction for students who are academically at-risk during the first year of higher education. Namely, if the appropriate instructional methods are utilized, then students who are academically at-risk may have a better chance of success in completing their freshman level coursework and be better equipped for their next academic steps.

Operational Definitions

The following are definitions of terms and phases specific to the current study.

Students who are academically at-risk are those who reported final grades of C or lower in first year courses. This grade benchmark corresponds to the National Survey of Student Engagement (NSSE) division of grades into "A & B" or "C or lower" (NSSE Report Build-Different

er-Public, 2018, Grades section). Students who are academi-

cally prepared are students who reported A or B in first year courses. This grade benchmark corresponds to the NSSE division of grades into "A & B" or "C or lower" (NSSE Report Builder-Public, 2018, Grades section).

Classifications Carneaie is a modified version of the Basic Classification of the Carnegie Classification of Higher Education Institutions (CCHEI) used by the researchers for this study. Rather than the nine types of institutions of higher education in the Basic Classification, this study uses three institutional types: bachelor's, master's, and doctoral degree-granting institutions.

Active learning is an ambiguous concept (Lombardi & Shipley, 2021) that may be defined as any instructional tool that involves students during

their educational development (Prince, 2004). As Lombardi and Shipley (2021) indicated active learning may be viewed "as a generalized instructional process (a) for constructing knowledge and (b) for deepening engagement" or "as (a) antithetical to passive learning and (b) antithetical to lecture" (p. 10). For this study, "Applying facts, theories, or methods to practical problems or new situations... Analyzing an idea, experience, or line of reasoning in depth by examining its parts...[and] Evaluating a point of view, decision, or information source" (NSSE, 2015, p. 1) are considered active learning strategies because students are actively engaging with the content.

Also, the number of hours spent in learning activities is taken to be a measure of active engagement with the class. While some fields reguire more out-of-class activities, weekly self-reported study time that is less than the same number of enrolled credit hours may not be enough course engagement to be successful. Therefore, hours of study is viewed as an indicator of engagement with the coursework.

In traditional learning, the teacher is the primary giver of all information learned in these types of courses. Lessons are usually taught through lectures and memorization. Lecture-based teaching does not promote active learning (Lombardi & Shipley, 2021). For this study, "memorizing course material" is considered traditional learning (NSSE,

2015, p. 1).

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Literature Review

In this literature review, we will consider active learning teaching strategies; active learning and its relationships to Bloom's Revised Taxonomy; and how institutional type is defined by the Carnegie Classification. This relevant literature frames our study and identifies current research gaps.

Active Learning Teaching Strategies

Learning strategies are procedures and practices used by faculty to increase student learning (Rachal et al., 2007). Traditional lectures and memorization are still a leading form of undergraduate instruction. The traditional method of teaching, sometimes called lecture-based teaching, is viewed as a classroom where students are listening to the teacher lecture, copying notes, memorizing facts,

and working independently (Shi et al., 2018). This traditional approach for teaching involves memorization and limited student classroom engagement. Lecture-based teaching (traditional) is also when the teacher is the primary giver of all information learned (Shi et al., 2018).

Different learning strategies may have differing success with various student populations. For example, one study indicated that high course withdrawals and failure rates may be the result of student boredom with classes in which skill and drill activities have little to do with their college-level courses (Grubb & Associates, 1999). Students tend to be more involved and have self-efficacy when they are in an engaging learning environment (Churach & Fisher,

2001). Student engagement is defined as having students actively involved and contributing to the lesson by using resources and persistence in learning (Rachal et al., 2007). One approach to an engaging learning environment is utilizing active teaching strategies in order to engage students in active learning. Meyers and Jones (1993) stated that active learning originated from the two basic expectations that "learning is by nature an active endeavor and that different people learn in different ways" (Meyers & Jones, 1993, p. xi). Taken together, it is imperative to have multiple strategies to use in the classroom to keep students engaged and motivated to learn.

There are a variety of active learning teaching strategies. Some examples include learning communities (Boylan et al., 2005), cooperative groups (Opdecam et al., 2014), the flipped classroom model (Zamora-Polo et al., 2019), High-impact Practices (HIPs) (Kuh et al., 2017), and gamification (Fulks & Lord, 2016). Boylan et al. (2005) proposed using learning communities, also called cohorts, in developmental education, and further suggested that learning communities, which involve students taking coursework together or living as a group, allowed students to feel more comfortable and encouraged to participate in the classroom. Cooperative groups are also another way to involve students actively during the course and are one of the most commonly used methods. In cooperative learning, students are arranged into small clusters to finish an assignment, find solutions, or examine a situation (Opdecam et al., 2014). A flipped classroom (Zamora-Polo et al., 2019) involves students preparing for work outside of class and finishing practicing inside the classroom. This strategy encourages students to be self-reliant. High-impact practices use a variety of tools, such as internship and global learning (study abroad), to engage the learner in the academic process, and many are measurable (Kuh et al., 2017). Gamification is using gaming strategies to teach content or to evaluate mastery (Fulks & Lord, 2016). Numerous other forms of active learning teaching strategies exist.

Nolting reported in an interview with Boylan (2011) that students in developmental courses, and students in developmental mathematics in particular, "need a multimodality instructional approach which means integrating the lecture with manipulatives, math study skills, and group work" (p. 22). Therefore, students who are academically at-risk and students in developmental programs may especially benefit from active learning teaching strategies. A classroom with active learning allows further student involvement with richer knowledge and better skills to explain problems and think analytically (Smart & Csapo, 2007). Furthermore, use of active learning teaching strategies may result in an increase in student motivation and positive self-efficacy (Wadsworth et al., 2007).

Active Learning and Bloom's Revised Taxonomy

Bloom's revised taxonomy is a model that provides a theoretical construct for defining and better classifying active learning. The cognitive domain of the Bloom's revised taxonomy ranges from the lowest level, remembering, to the highest level, creating (Anderson & Krathwohl, 2001). Tabrizi and Rideout (2017) define active learning as a system for engaging students in Bloom's revised taxonomy's higher-order thinking skills through the use of various activities (Tabrizi & Rideout, 2017). It is hoped that as students are exposed to more active learning teaching strategies, they will begin to progress through Bloom's revised taxonomy and be able to apply those skills learned in other courses. Most educators develop questions for the higher levels of Bloom's revised taxonomy to improve and to gauge students' critical thinking skills (Fowler, 2006). Interactive teaching strategies like the ones used to engage students in active learning are believed to involve higher-level thinking skills, while traditional techniques are thought to elicit lower-level thinking skills. For example, students taking a large enrollment STEM course who were identified as under-represented or not under-represented were equally successful when active learning teaching strategies that targeted higher-order thinking skills were employed (Kressler & Kressler, 2020).

Institutional Type and the Carnegie Classifications

Institutional type has been used in a variety of research studies. From studies on moral reasoning (Mayhew, 2012) to capstone courses (Grahe & Hauhart, 2013) to teaching undergraduate economics (Becker & Watts, 1996), institutional type has been considered as a possible influencing factor. Numerous systems exist to categorize institutional type, and one commonly used system is the Carnegie Classification of Higher Education Institutions (CCHEI). Developed by the Carnegie Commission in 1970, the Basic Classification was first available for everyone to use in 1973. Under the Basic Classification there are nine categories of different universities and colleges (CCHEI, n.d.). This study considers institutional type using a modified version of the CCHEI. Specifically, this study only considers students in their first year at bachelor's degree-, master's degree-, and doctoral degree-granting institutions.

Based on the literature reviewed, this study attempts to close a gap in the relevant literature. We were not able to identify any studies that simultaneously addressed institutional type, active learning and traditional teaching methodologies, and classification of students who are academically at-risk and academically prepared.

Methodology National Survey of Student Engagement (NSSE)

According to Price and Baker (2012), the "NSSE is predicated on the assumption that certain student behaviors are indicators of students' engagement in the learning process" (p. 21). The NSSE Report Builder-Public 2017 and 2018 data

was chosen for this study as it is from an international survey (NSSE, 2018) that included students at a variety of types of institutions, including both public and private institutions (Gonyea & Kinzie, 2015) and numerous students across the country. Specifically, the NSSE Report Builder-Public 2017 and 2018 data set was utilized as it was the most current data set available. According to the National Survey of Student Engagement (2018), even though institutions must pay for participation, nearly 325,485 first-year students from the United States and Canada participated in the survey; around 1,020 institutions in total participated, with 8% of the institutions being Canadian. However, only results from 4-year or higher institutions in the United States were considered in this study due to possible differences in academic systems. Results from

the survey give an approximation about the time undergraduates spent on schoolwork and what they got out of attending a higher education institution (NSSE, 2019).

Researchers used chi-square analyses to establish associations between the variables. The first chi-square test of proportions was performed to determine if the proportions differed by Carnegie Classification for the populations who were academically prepared and who were academically at-risk. Next, chi-square tests of independence were performed to determine if there were statistically significant associations between student academic levels (academically prepared or academically at-risk), course emphasis on learning activities (traditional or active), and Carnagie Classifications (doctoral-, master, or bachelor-degree awarding institutions).

NSSE Variables and Characteristics

A report was generated from the NSSE Report Builder-Public 2017 and 2018 (2018) considering first-year students and institution type (Carnegie Classification), academically prepared versus academically at-risk status (self-reported grades), and traditional-learning versus active learning classroom type (types of activities required in the courses and hours spent studying). Grouped descriptive statistics were employed as individual responses to survey questions, and student background characteristics were not available in the NSSE Report Builder-Public 2017 and 2018 data set.

First Year

Interactive

teaching strategies

like the ones

used to engage

students in active

learning are

believed to involve

higher-level

thinking skills,

while traditional

techniques are

thought to elicit

lower-level

thinking skills.

For the purposes of this study, only students who were "First Year" in the United States were considered (NSSE, 2015, p. 3). Responses from students who were considered "First Year" were utilized because they represent all beginning students, those who were academically prepared and those who were academically at-risk (NSSE, 2015, p. 3). Using responses from those who have progressed further in their academic careers, such as a student classified as a "Senior," would self-select for students who were successful enough to progress (NSSE, 2015, p. 3). **Carnegie System**

This study only considers first-year students at bachelor's degree-, master's degree-, and doctoral degree-granting institutions. Therefore, to better compare overall institutional types, the Car-

negie Classification categories were coalesced in the NSSE Report Builder-Public 2017 and 2018 data to three institutional categories: bachelor's, master's, and doctoral degree-granting institutions.

Academically Prepared and Academically At-Risk

For the purposes of this study, students in the first year who in the NSSE Report Builder-Public 2017 and 2018 data had a self-reported average of "A" or "B" were considered academically prepared, and students in the first year who had a self-reported average of "C" or lower were considered academically underprepared and were believed to be academically atrisk. The NSSE Report Builder-Public 2017 and 2018 data does not record student background characteristics, such as high school grade point average, ACT scores, or SAT scores. Therefore, the NSSE Report Builder-Public 2017 and 2018 data was then grouped by the grade categories of "A and B" and "C" or lower to consider academic preparedness (academically prepared or academically at-risk).

Traditional Learning or Active Learning

In order to capture data on instructional methodologies, responses for two questions from the NSSE survey were collected. The first question, "During the current school year, how much has your coursework emphasized the following?" (NSSE, 2015, p. 1), asked students to indicate the frequency that their coursework emphasized various types of learning strategies. For this study, the first type, "Memorizing course materials" (NSSE, 2015, p. 1) is considered a traditional learning technique. The three types of learning strategies considered as active learning techniques are: "Applying facts, theories, or methods to practical problems or new situations... Analyzing an idea, experience, or line of reasoning in depth by examining its parts...[and] Evaluating a point of view, decision, or information source" (NSSE, 2015, p. 1). "About how many hours do you spend in a typical 7-day week doing the following?" (NSSE, 2015, p. 1) was the second survey question with responses collected from the NSSE survey data. For the analylsis, researchers assumed that a greater number of hours spent in learning activities to be equated with meaningful, active engagement with the content.

Limitations

The first limitation is if the questions in the NSSE Report Builder-Public 2017 and 2018 data actually reflect the variables being studied: (a) Are self-reported grades a true measure of being academically at-risk? and (b) Do the self-reported learning activities correspond to the actual teaching methods utilized in the courses the students are taking? The second limitation is due to the nature of the survey instrument, as it "is a self-selected and voluntary survey" (Rabourn et al., 2018, p. 29) at both the individual response and institutional levels. Next, even though the NSSE Report Builder-Public displayed responses from students across the United States (NSSE, 2018), the responses will not be generalizable to a particular institution. Further, student background characteristics that were not assessed by the NSSE Report Builder-Public 2017 and 2018 data could compound or impact these results. Various causalities and correlational factors may influence self-reported grades other than academic preparedness, such as motivation, student age status, or hours of employment.

Results

Out of the 188,836 first year students surveyed in the National Survey of Student Engagement 2017 & 2018, 173,579 were classified as academically prepared and 15,257 as academically at-risk (NSSE Report Builder-Public, 2018). See Table 1 for descriptive statistics regarding the numbers of students that were academically prepared and academically at-risk at the bachelor, master, and doctoral levels.

Table 1

Number and Percentage of Students by Academic Level and Carnegie Classification

				-					
Academic Level	Doctoral		Mas	ster	Bach	elor	Total		
	n	%	n	%	n	%	n	%	
Academically prepared	72,007	92.34	75,047	91.39	26,525	92.16	173,579	91.92	
Academically at-risk	5,934	7.66	7,066	8.61	2,257	7.84	15,257	8.08	

Chi-Square Analyses

Researchers used numerical data available in the Appendix to perform all chi-square testing. First, a chi-squared test of proporations was used to determine if the distributions of students by academic level differed significantly across the Carnegie Classifications. There was a statistically significant association between student academic level and enrollment across bachelor's, master's, and doctoral degree-granting institutions, $X^2(2, N=188, 836) = 55.543, p < 0.001)$.

Researchers also conducted chi-square tests of independence for each type of learning activity reported by students as well as student study hours to observe any differences in survey responses from students who were academically prepared and those who were academically at-risk. Table 2 displays these chi-square results. The resulting p values for students who were acadmecially prepared were statistically significant (p < .001). For students who reported grades of C or lower, the null hypothesis was rejected for all of the questions at the .001 or .05 significance level except for the "Evaluating" a point of view, decision, or information source" (NSSE, 2015, p. 1), which had a p value of 0.223. This indicated that no association could be established between students who were academically at-risk and their survey responses for learning activities involving evaluation.

Table 2 Chi-Square Results for Learning Activities Reported by Student Academic Levels

Les avia - A stillite Tense		Acade	mically Prepared		Academically At-Risk					
Learning Activity Type	X ²	df	п	p	X ²	df	п	p		
Question 1: During the current school year, how much has your coursework emphasized the following?										
Memorizing course material	577.18	6	174,227	<.001	49.01	6	15,313	<.001		
Applying facts, theories, or methods to practical problems or new situations	254.64	6	174,127	<.001	36.08	6	15,300	<.001		
Analyzing an idea, experience, or line of reasoning in depth by examining its parts	168.88	6	173,987	<.001	17.99	6	15,290	.006		
Evaluating a point of view, decision, or information source	357.65	6	173,995	<.001	8.21	6	15,276	.223		
Question 2: About how many hours do you spend in a typical 7-day week doing the following?										
Preparing for class (studying, reading, writing, doing homework or lab work, analyzing data, rehearsing, and other academic activities)	1055.13	14	173,579	<.001	65.87	14	15,257	<.001		

Table 3

Learning Activity Responses Per Academic Level Per Carnegie Classification (N = 188,836)

		Acadmically	/ Prenared		Academically At-Risk				
Learning Activity Type	Responses	DOC ^a	MAS ^b	BAC ^c	Responses	DOC ^a	MAS ^b	BAC	
Question 1: During the current school year, how					•	200	111/10	Brite	
	Very little	3.1	3.1	4.3	Very little	4.6	5.1	5.4	
	Some	24.8	24.9	29.7	Some	27.5	30.4	29.9	
Memorizing course material	Quite a bit	46.2	47.0	45.6	Quite a bit	42.5	43.5	44.8	
	Very much	25.9	25.0	20.3	Very much	25.3	21.0	19.9	
	Very little	2.7	3.2	2.7	Very little	6.3	6.8	5.4	
Applying facts, theories, or methods to	Some	22.9	24.8	22.5	Some	30.6	34.3	32.3	
practical problems or new situations	Quite a bit	47.7	48.5	49.4	Quite a bit	43.7	42.4	44.8	
	Very much	26.6	23.5	25.3	Very much	19.3	16.5	17.4	
	Very little	3.2	3.2	2.5	Very little	6.1	6.6	5.0	
Analyzing an idea, experience, or line of	Some	24.8	25.8	23.1	Some	32.2	33.9	32.9	
reasoning in depth by examining its parts	Quite a bit	45.7	46.1	46.5	Quite a bit	42.3	42.2	43.0	
	Very much	26.1	24.7	27.7	Very much	19.3	17.2	18.7	
	Very little	4.2	3.2	2.7	Very little	5.8	5.6	4.5	
Evaluating a point of view, decision, or	Some	26.4	25.1	23.0	Some	31.1	31.6	31.1	
information source	Quite a bit	45.1	47.3	47.4	Quite a bit	43.9	44.6	45.0	
	Very much	24.1	24.3	26.8	Very much	19.1	18.1	19.0	
Question 2: About how many hours do you spen	,							2010	
Question 2. About now many nours up you spen									
	0 hrs	0.3	0.3	0.2	0 hrs	1.2	1.4	1.3	
	1-5 hrs	9.3	12.1	8.5	1-5 hrs	20.7	23.5	18.8	
Preparing for class (studying, reading, writing,	6-10 hrs	20.7	23.1	18.9	6-10 hrs	25.5	25.4	24.2	
doing homework or lab work, analyzing	11-15 hrs	22.3	22.2	21.3	11-15 hrs	21.0	21.5	21.0	
data, rehearsing, and other academic	16-20 hrs	20.6	19.5	21.4	16-20 hrs	15.1	14.7	16.5	
activities)	21-25 hrs	12.9	11.3	14.5	21-25 hrs	8.4	6.9	8.0	
	26-30 hrs	6.4	5.5	7.7	26-30 hrs	3.4	3.0	4.7	
	> 30 hrs	6.5	5.2	7.0	> 30 hrs	4.3	3.5	5.1	

Note. Values are expressed in percentages. For numerical values, see the Appendix.

^a Doctoral degree-awarding institution. ^b Master degree-awarding institution. ^c Bachelor degree-awarding institution.

Analysis of Student Groups and Carnegie Classification

Because the chi-square test results confirmed that there were statistically significant association between all but one interaction, further examination of the NSSE data was warranted. Table 3 shows a comparison by Carnegie Classification of traditional learning versus active learning for students who were academically prepared as compared to students who were academically at-risk. For ease of comparison, percentages are displayed for the number of first-year students responding for each of the composite Carnegie Classifications. The results from Table 3 are discussed from three main aspects: traditional learning, active learning, and hours spent studying.

Traditional Learning (Memorization)

Students who were academically at-risk at doctoral degree-granting institutions reported memorizing course content very much (NSSE, 2015, p. 1) more frequently (25%) than at either master's (21%) or bachelor's degree-granting (20%) institutions. Students who were academically prepared at doctoral degree-granting institutions reported memorizing course content very much (NSSE, 2015, p. 1) more frequently (26%) than at bachelor's degree-granting (20%) institutions but similarly to master's degree-granting (25%) institutions (NSSE Report Builder-Public, 2018).

Students who were academically prepared and academically atrisk reported similar frequencies for the three active learning categories of applying, analyzing, and evaluating across all institution types. However, students who were aca-

demically at-risk reported lower *very much* use of active learning techniques across all instructional types. For example, at all institutional types, 24% to 27% of students who were academically prepared reported *very much* as their frequency of "Applying facts, theories, or methods to practical problems or new situations" (NSSE, 2015, p. 1), whereas only 17% to 19% of students who were academically at-risk reported *very much* as their frequency of "Applying facts, theories, or methods to practical problems or new situations" (NSSE, 2015, p. 1).

Hours Spent Studying

In terms of self-reported hours spent studying, at all institutional types, there is a considerable shift from the more limited hours spent studying by students who were academically at-

Active learning teaching strategies do not imply that there is only one way to get students actively involved, but the use of a variety of teaching methods can enhance learning.

risk to the increased hours spent studying by the students who were academically prepared. For example, at all institutional types, 20% to 22% of students who were academically prepared reported spending 16–20 hours studying, whereas only 15% to 17% of students who were academically at-risk reported spending 16-20 hours studying (NSSE Report Builder-Public, 2018). When stepping away from the Carnegie Classifications and only looking at the hours spent studying for the academically prepared versus the academically at-risk groups, the percentage of students spending 0-10 hours was 32% for the academically prepared and 48% for the academically at-risk. However, when looking at more than 10 hours per week studying, the percentages for students who were academically

> prepared was 68%, whereas the percentage for students who were academically at-risk was 52%.

Discussion

As discussed previously, active learning teaching strategies engage the student, encourage greater involvement in the learning process, and encourage increased use of higher-level thinking skills. While our results indicated that students who were academically at-risk reported using traditional methods very *much* at similar rates to their peers who were academically prepared, students who were academically atrisk consistently reported very much using active learning strategies at much lower rates than their peers who were academically prepared. In addition, a lower percentage of students who were academically at-risk reported studying 16 or more hours

a week than their peers who were academically prepared. Therefore, active learning strategies should be targeted for students who are academically at-risk to increase their engagement, study time, and performance. This is a general suggestion, as outside responsibilities, such as work and family, will vary for students and is not captured in this study; students who work more hours will likely have fewer hours to devote to study.

People learn in different ways, and learning is an active venture (Meyers & Jones, 1993). Active learning teaching strategies do not imply that there is only one way to get students actively involved, but the use of a variety of teaching methods can enhance learning. Using multiple active learning strategies increases the student's ability to comprehend the course material and move on to their college-level courses (Fowler & Boylan, 2010). Possible suggestions and tips to help students be successful in those first-year courses are below:

1. Include at least one active learning activity in your course to increase student participation.

2. Have students complete a collaborative project (Kuh & AAC&U, 2008).

3. Have students make connections from your course to other courses.

4. Use real world and/or problem-based teaching strategies.

5. Use gaming strategies to engage students in the classroom (Fulks & Lord, 2016).

6. Implement High-impact Practices (Kuh et al., 2017).

7. Utilize a flipped classroom (Zamora-Polo et al., 2019).

Further, many students who are not college-ready require assistance in mathematics. Although the NSSE Report Builder-Public 2017 and 2018 does not allow for the segregation of data based on initial academic testing or specific content needs, one can assume that students earning grades of "C" or lower in college may have some academic needs in mathematics. Mathematics is a course that many students need in college, and successful completion is a major goal for many students to graduate. Indeed, the enrollment in developmental mathematics courses at institutions of higher education is steadily increasing; therefore, educators are becoming accustomed to instructing students who are unprepared (Mireles et al., 2011). Students who take developmental mathematics courses are less likely to graduate, and most students are likely to repeat the course (Mireles et al., 2011). With the focus on getting students who are underprepared up to college level, mathematics will be a factor that will persist. This study suggests that students who are academically at-risk need to increase their study time and that instructors need to incorporate active learning teaching strategies to boost collegiate success rates.

Implications

Creating and providing quality instruction has become a major goal for many colleges and universities (Hall & Ponton, 2005); and many institutions have devoted resources to create and to enhance programs targeted at students who are academically underprepared (Hall & Ponton, 2005). One example of a potential enhancement, for institutions that provide admission to students who are academically atrisk, could be within departments that teach any freshman-level courses. The results of this study may help educators of all freshman-level courses more effectively prepare students who are academically at-risk by utilizing differentiated teaching strategies and helping faculty members to compare teaching methods. Additionally, the study may help educators of students in their first year to consider what teaching methods (i.e., active learning and/or traditional learning) may increase their students' performance, retention, and passing rates. In turn, students' academic performance may then be enhanced based on changes that faculty members may make to their instruction.

Conclusions

Results from the National Survey of Student Engagement 2017 and 2018 indicate that students who were academically at-risk report spending less time using the higher-order thinking skills associated with active learning (applying, analyzing, and evaluating) than their peers who were academically prepared. Not surprisingly, students who were academically at-risk reported fewer hours spent studying than their peers who were academically prepared (NSSE Report Builder-Public, 2018). These results could indicate that students who were academically at-risk had-poorer study habits, did not come to the institution with quality study strategies, or presented with more interferences to outside studying. Students who were academically at-risk may have reported fewer active learning strategies because they were not fully completing assignments, or they may have had instructors who were not utilizing active learning strategies.

More research is needed to consider whether instructors at all types of institutions are utilizing active learning or if students who are academically at-risk are not participating in the required class activities. Further, other studies controlling for student background characteristics, especially outside responsibilities such as employment and hours spent caring for family, are needed to delineate the effects of active learning on students who are academically at-risk.

Future Research

Generally, more research on active learning teaching strategies versus traditional instructional methods needs to be conducted to address additional teaching strategies, such as collaborative learning, high-impact practices, and gamification. Future research should include using data from the most current NSSE as it becomes available, especially as the COVID-19 pandemic could have created differences in institutional types, instructional strategies, and student characteristics. Future studies should also control for student background characteristics that were not assessed by the NSSE Report Builder-Public 2017 and 2018. Finally, other causalities and correlational factors that could influence self-reported grades, such as motivation, student age status, employment hours, and family responsibilities, should also be considered in additional research.

Disclosure Statement

No potential conflict of interest was reported by the authors.

About the Authors

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by Academic Level and Carriegie Classification											
Question and		cademically Prepared (A and B)				Academically At-Risk (C or Lower)					
Learning Activity	Response	DOCª	MAS ^b	BAC ^c	Total	Response	DOCª	MAS ^b	BAC ^c	Total	
During the current school year, how much has your coursework emphasized the following?											
Memorizing course material	Very little	2,244	2,313	1,143	5,700	Very little	276	364	122	762	
	Some	17,900	18,737	7,933	44,570	Some	1,641	2,154	678	4,473	
	Quite a bit	33,394	35,404	12,171	80,969	Quite a bit	2,533	3,083	1,016	6,632	
	Very much	18,751	18,815	5,422	42,988	Very much	1,508	1,486	452	3,466	
	Total	72,289	75,269	26,669	174,227	Total	5,958	7,087	2,268	15,313	
Applying facts,	Very little	1,987	2,387	725	5,099	Very little	373	480	122	975	
theories, or meth-	Some	16,536	18,634	5,995	41,165	Some	1,824	2,428	732	4,984	
ods to practical problems or new	Quite a bit	34,500	36,520	13,186	84,206	Quite a bit	2,604	3,006	1,015	6,625	
situations	Very much	19,228	17,689	6,740	43,657	Very much	1,152	1,170	394	2,716	
	Total	72,251	75,230	26,646	174,127	Total	5,953	7,084	2,263	15,300	
Analyzing an idea,	Very little	2,306	2,434	677	5,417	Very little	365	465	114	944	
experience, or line	Some	17,943	19,443	6,164	43,550	Some	1,917	2,403	746	5,066	
of reasoning in depth by examin-	Quite a bit	33,055	34,687	12,393	80,135	Quite a bit	2,520	2,993	975	6,488	
ing its parts	Very much	18,890	18,612	7,383	44,885	Very much	1,149	1,218	425	2,792	
	Total	72,194	75,176	26,617	173,987	Total	5,951	7,079	2,260	15,290	
Evaluating a point	Very little	3,013	2,444	715	6,172	Very little	343	395	101	839	
of view, decision,	Some	19,112	18,925	6,121	44,158	Some	1,853	2,236	705	4,794	
or information source	Quite a bit	32,615	35,569	12,638	80,822	Quite a bit	2,614	3,161	1,020	6,795	
Jource	Very much	17,434	18,253	7,156	42,843	Very much	1,136	1,280	432	2,848	
	Total	72,174	75,191	26,630	173,995	Total	5,946	7,072	2,258	15,276	
About how many h	ours do you spend	d in a typica	al 7-day we	eek doing t	the followi	ng?					
Preparing for class (studying, reading, writing, doing homework or lab work, analyzing data, rehearsing, and other academic activities)	0 hrs	209	216	61	486	0 hrs	72	96	29	197	
	1–5 hrs	7,181	9,089	2,259	18,529	1–5 hrs	1,231	1,662	426	3,319	
	6–10 hrs	14,967	17,392	5,024	37,383	6–10 hrs	1,517	1,801	548	3,866	
	11–15 hrs	16,094	17,098	5,678	38,870	11–15 hrs	1,253	1,523	477	3,253	
	16–20 hrs	14,880	14,665	5,712	35,257	16–20 hrs	898	1,039	374	2,311	
	21–25 hrs	9,351	8,539	3,857	21,747	21–25 hrs	500	486	181	1,167	
	26–30 hrs	4,647	4,123	2,056	10,826	26–30 hrs	205	209	107	521	
	> 30 hrs	4,678	3,925	1,878	10,481	> 30 hrs	258	250	115	623	
	Total	72,007	75,047	26,525	173,579	Total	5,934	7,066	2,257	15,257	
·				1		1			1		

Appendix NSSE 2017 & 2018 First Year Students Frequency Data for Learning Activity by Academic Level and Carnegie Classification