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What is This?
A Multisite Study of Learning in Introductory Psychology Courses

Regan A. R. Gurung¹, David B. Daniel², and R. Eric Landrum³

Abstract
Four hundred and fifty-four students enrolled in introductory psychology at different institutions across the nation participated in a study of factors related to learning. Key measures included an instructor rating, ratings of textbook quality and helpfulness, study time, student self-evaluations of study behaviors, approach to learning, self-report of learning, and a measure of quiz performance using biopsychology and learning chapter questions from a College Board Advanced Placement exam. The authors found significant predictors of both self-reported learning (deep approach, less surface approach, instructor ratings, student self-evaluations, and study behaviors) and quiz performance (grade point average, study time, metacognitive activity, and less use of a surface approach to learning). These results are discussed in the context of optimizing student learning and providing a foundation for future research.

Keywords
student learning, instructor, textbooks, study behavior

Any examination of how students learn necessitates a focus on at least three major components: student behaviors (e.g., study techniques), instructor behaviors (how is learning facilitated?), and the means by which the content is delivered (e.g., textbooks). In a multisite study, we assessed how student study behaviors, textbook quality, and instructor quality influence learning.

Many researchers have attempted to classify study techniques and to identify the techniques that are optimal (e.g., for reviews, see Entwistle & McCune, 2004; Gurung & Schwartz, 2009; Hattie, 2009). Study behaviors are broadly defined as behaviors functioning to acquire, organize, synthesize, evaluate, remember, and use information (Crede, & Kuncel, 2008; Gettinger & Seibert, 2002)—specific examples include time management, goal setting, selecting what to study (how and where), taking good notes, reading, and self-testing. Although clearly establishing the link between study behaviors and performance, large meta-analytical studies (Crede & Kuncel, 2008; Hattie, 2009) do not provide specific prescriptions of how exactly students should study.

There are also many ways the instructor can influence learning. As summarized by Buskist (2011), volumes of information are available about excellent teaching, and a number of factors such as instructor characteristics (ideal/master teacher factors) may all influence student learning (Buskist & Benassi, 2012; Fink, 2003; Gurung & Vespia, 2007; Keeley, Smith, & Buskist, 2006). In addition, there are suggestions of effective ways to teach. Bain (2004) culled a set of characteristics identifying the best college teachers. Similarly, Buskist (2004) identified a set of features held by master teachers, such as passion, organization, and inspiration. However, few of these efforts have addressed instructor variables as they impact student performance.

The modality by which information is transmitted is also very important, and the primary sources of content delivery tend to be the instructor and the textbook, both of which are highly variable. The vast research literature on textbooks show that textbooks have little similarity in content and vary in length, writing style, number and usefulness of pedagogical aids used, applied or research focus, tone, and comprehensiveness (Christopher, Griggs, & Hagens, 2000; Griggs, Bujak-Johnson, & Proctor, 2004; Gurung, 2004; Landrum & Hormel, 2002; Nevid & Lampmann, 2003; Sappington, Kinsey, & Mun-sayac, 2002; Weiten, Duguay, Rehmke, & Sewell, 1999).

We designed this study to improve on three major shortcomings with previous research. First, few studies focus on more than one influence on learning. Learning is a complex process influenced by many factors (Daniel & Poole, 2009). Hattie (2009) used meta-analytic strategies and examined 15,000 studies to tease out variance in learning due to teachers, students, and other factors, but research outcomes that examine student behaviors, teacher characteristics, and textbook quality

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in one study are not present in the literature. Second, many relevant studies primarily use self-reported learning as the sole outcome measure. Student perceptions of their own learning are not good proxies for learning (Hacker, Dunlosky, & Hacker, 1998; Kennedy, Lawton, & Plumlee, 2002; Wesp & Miele, 2008; Zinn, Magnotti, Marchuk, Schultz, Luther, & Varfolomeeva, 2011; see Dunning, Heath, & Suls, 2004, for a review of issues associated with self-assessment). In this study, we assessed student perceptions of their own learning, but we also utilized a standard measure of objective learning (e.g., quiz) for comparison across sites and classes taken from items vetted for use in the Advance Placement (AP) Psychology exam (College Board, 2004). The latter parallels a common form of assessment of learning in introductory psychology courses and offers a more rigorously developed and standardized assessment than has been used previously. Finally, most studies are conducted with students from one institution (e.g., Lammers & Smith, 2008), or one course (Chase & Houmanfar, 2009), or using one textbook (e.g., Gurung, 2004). We assessed students from different parts of the United States using different textbooks.

Our primary research goal was to contrast the different factors that influence student learning in the introductory psychology class. Introductory psychology is the highest enrollment class in our discipline and students may be particularly likely to overestimate their learning, given that the content of psychology seems to be common sense. Psychology instructors are also particularly well positioned to address students’ study habits and metacognitive abilities, given that areas of the field focus on learning and behavior. We reviewed the literature to identify key variables and included them here to assess and account for correlations between factors and the consequent relationship to learning. For example, we measured the extent to which students take a deep approach to learning. Students adopting a deep approach set out to understand what they have read or heard for themselves, which makes it more likely they will grasp the authors’ meaning and learn the material better (Entwistle, 2009). Researchers from cognitive psychology suggest that study behaviors such as repeated testing and metacognitive strategies (e.g., assessing what is known) are key to learning (Björk, 1994; Winne & Nesbit, 2010). Consequently, we examined the extent of students’ awareness of their knowledge. We also asked students how they rated their textbook (Gurung & Martin, 2011). Finally, we assessed student perceptions of instructor characteristics. In short, we measured four major contributors to learning: (a) student approach to learning (e.g., deep/surface), (b) study behaviors, (c) perceived textbook quality, and (d) instructor ratings by students. We also tested the extent to which each predicted self-reported learning and scores on a quiz.

**Method**

**Participants**

Four hundred and fifty-four students from different universities participated in this study (59% women; 65 students did not indicate their sex). Two schools were large public universities (one in the Midwest, \( N = 252 \) from three sections of introductory psychology, one in the South, \( N = 151 \) from two sections), and the last school was a small liberal arts college in the South (\( N = 28 \)). The remainder of the sample did not identify a school. The average age was 19.26 (SD = 2.58). The majority (58%) were first-year students. The remaining students were sophomores (20%), juniors (5%), and seniors (5%). Of those reporting a grade point average (GPA; \( N = 400 \)), the average GPA was 2.97 (SD = 0.58). Ethnicity data were not collected.

**Materials**

Participants completed an online questionnaire designed to capture various aspects of student learning behavior. We modified existing scales to measure ratings of the instructor, study strategies, textbook quality, attitude toward learning, and wrote items to capture student behavior. A complete copy of the 95-item questionnaire is available upon request.

**Instructor ratings.** We used items derived from the Teacher Behaviors Checklist (Keeley, Smith, & Buskist, 2006). We asked students to indicate how each of the 20 different characteristics described their instructor: communicates well, uses relevant examples, humorous, likeable, passionate/enthusiastic, approachable, organized, effectively uses visual aids, conscientiousness, encouraging/caring, entertaining, knowledgeable, easy to understand, engaging, uses a good mix of teaching approaches/styles, professional, challenging, respectful, is a fair grader, and flexible. Students responded on a Likert-type scale from 1 (strongly disagree) to 9 (strongly agree). We summed responses on all 20 items to create a total score (Cronbach’s \( \alpha = .96 \)).

**Textbook assessment.** We used a short version of the Textbook Assessment and Usage scale (Gurung & Martin, 2011), utilizing nine questions designed to assess the main components of a textbook (study aids, figures, visual appeal, boxed information, tables, research studies, writing, examples, and photographs). Students first rated their textbook on the quality of each of the items (with regard to placement, relevance, and clarity). Students responded on a Likert-type scale from 1 (poor) to 9 (excellent). Next, students indicated the extent to which each component of the textbook helped them to understand the material. Students responded on a Likert-type scale from 1 (not helpful at all) to 9 (very helpful). We created two total scores, summing items for quality and helpfulness. Both scales showed high reliability (\( \alpha = .90 \) for text quality and \( \alpha = .91 \) for text helpfulness).

**Study time.** We used items modified from the Study Behavior Checklist (Gurung, Weidert, & Jeske, 2010) and created a composite score from the 11 items to represent total time studied. We asked students to report the amount of time in minutes, spent on each of the 11 study behaviors in preparation for their most recent exam: read the text, make up examples to
understand material, read notes, rewrite notes and/or skim notes, review information highlighted from text, take notes from the book, use concept checks and other textbook pedagogical aids, study with a friend, memorize definitions through repetition, think of mnemonic/memory devices, and review figures and tables in text.

**Self-evaluation related study activity (SERSA).** Because student perceptions of their own learning tend to guide their study behavior (Gurung & McCann, in press), we assessed student perceptions of their learning and study-related activity with two sets of questions. One set of three questions (SERSA-metacognition, \( \alpha = .93 \)) asked how much students remembered the material for the most recent exam, how much students’ thought they understood the material, and how confident students’ were that they knew the material. A second set of four questions assessed quizzing behavior, given that quizzes are a major way students’ can test their knowledge (SERSA-quizzing, \( \alpha = .84 \)). We asked students general questions about their use of quizzing results (online or in-class) as feedback to study more for the exam, how often they reviewed quiz results to see what they got wrong, how often they reread parts of the book associated with missed quiz questions, and how often they closely read the book before taking the quiz.

**Student attitude toward learning.** We used the approaches to learning and studying items from the Shortened Experiences of Teaching and Learning Questionnaire (SETLQ; Tait, Entwistle, & McCune, 1997). The subscale of the SETLQ measured three different approaches to studying and learning: deep approach (9 items, \( \alpha = .76 \), e.g., I’ve looked at evidence carefully to reach my own conclusions about what I’m studying), surface approach (4 items, \( \alpha = .76 \), e.g., I’ve tended to take what we’ve been taught at face value without questioning it much), and organized effort (4 items, \( \alpha = .87 \), e.g., I’ve organized my study time carefully to make the best use of it). We created composite scores for each approach.

**Student performance.** We measured performance or learning in two ways. Students indicated how much they were learning in the class (self-reported learning) by responding to a singular item on a 9-point scale ranging from 1 = nothing to 9 = a lot. Additionally, all students also took a 15-item quiz that covered biology (eight questions) and learning (seven questions). We utilized items from the released 2007 AP exam with permission from the College Board. We created a composite quiz score (\( \alpha = .86 \)).

**Procedure**

We created the survey with Qualtrics software. A link to the survey was sent to students in the first and third authors’ classes. In addition, we sent the survey link to three colleagues teaching different sections of the class at different institutions. Some sections already completed quizzes online; however, all participants were asked to complete our quiz questions online using online survey software (Qualtrics). All classes included some evaluative component to gauge student comprehension of the course material; our online quizzes were asked in addition to the normal course and testing formats utilized by the respective instructors. We timed the release of the survey participation request to students in the introductory psychology classes for after they had covered the sections/chapters on biology and learning in class (approximately the sixth week of the semester). We told students to complete the survey in response to their introductory psychology class and the most recent exam. Students completed the quiz out of class (online) and some received research credit for participation. We could not measure how many students the e-mail was sent to and correspondingly do not have a return rate measure. Student participants had a chance to win one of the three music players.

**Results**

Our primary goal was to compare the relative contributions of factors such as the instructor, textbook, student attitude toward learning, and student study behavior, on learning and performance to provide a detailed picture of student learning behavior. To that end, we first provide descriptive statistics reporting a variety of behaviors and then predict self-reported learning and scores on our quiz. Table 1 provides the mean values and standard deviations for the major variables and the extent each correlates with the two main outcome variables, self-reported learning and quiz performance. All four categories of measures significantly correlated with quiz scores and all categories except study time correlated with self-reported learning. SERSA correlated with both learning measures, whereas online quizzing activity correlated with quiz scores. Five study behaviors correlated with quiz scores, whereas only two correlated with self-reported learning. Quiz scores and self-reported learning were correlated, \( r(444) = .20, p = .005 \).

**Predicting Learning**

We next conducted analyses predicting self-reported learning (how much the students said they learned in class) and their score on our quiz. We conducted two multiple regression analyses. In each regression, we entered all four major factors in as a block: student factors that could influence the outcome (GPA, amount of time studied, and approaches to learning; SERSA, instructor ratings, and textbook factors [quality and perceived helpfulness]). Table 2 shows a summary of the two regression analyses.

The predictor variables accounted for 26% of the variance in quiz scores. Five variables were significant: students’ GPA, surface processing (negatively related), total study time, SERSA, and online quizzing activity.

The predictor variables accounted for 36% of the variance in self-reported learning. Four variables were significant. The extent students held a deep approach, a surface approach (negatively related), ratings of the instructor, and SERSA all predicted self-reported learning.
A variety of factors influence student learning, and the results of our study illustrate the relative importance of categories of key factors and highlight differences between predicting learning using self-report measures and an objective test. Our results imply that factors that may guide student study behavior, such as their approach to learning and SERSA, are significantly related to their learning. This finding reinforces the emphasis recently placed on student responsibility in their own learning (Quality Principle 1, American Psychological Association, 2011) and suggests that efforts to guide students in productive study and monitoring behaviors may positively impact learning.

A finding of particular interest involves the "approach" that students reported to guide their own learning (Entwistle, 2009). The more a student used a surface approach, the lower their quiz scores and their self-reported learning. Likewise, taking more of a deep processing approach was associated with higher self-reported learning (though not quiz scores). Whereas the exact source of these differences is difficult to ascertain in a broad study such as this, the latter difference may indicate a disconnect between what many introductory psychology students think they are learning, and what their quiz performance indicates. In the absence of objective measures of learning (e.g., formative quizzing), students who are studying deeply may think they are learning (self-reported learning) and inaccurately gauge their actual progress (quiz score). This possibility is supported by the finding that the measure of SERSA, which included questions about how students utilized the results of quizzing, significantly correlated with quiz scores. Another potential explanation is that of the testing effect; students exposed to multiple quizzes or tests can achieve beneficial test performance as testing is an effective method of studying (Roediger & Karpicke, 2006). It could be that utilizing the testing effect is a concrete metacognitive strategy that can positively influence student performance—continued research in this area is needed to help disentangle these possible explanations for our data.

We conducted two separate multiple regression analyses (self-reported learning and quiz learning) to answer the question "what predicts learning?" Students at all GPA levels tend to overestimate their performance in general psychology (Prohaska, 1994) and our results show how misleading using only self-reported learning measures can be; thus, the predictors of the two types of learning are not the same. Teacher behavior predicted self-reported learning but not an objective measure. Total time studied and online quizzing behavior predicted an objective measure but not self-reported learning. We believe that the latter discrepancy indicates that students may not fully integrate their study behaviors into their own sense of learning, although said behaviors do indeed influence their actual learning as measured by an objective quiz.

Our results provide new challenges for pedagogical research attempting to identify optimal study skills. Not surprisingly, study behaviors continue to play a critical role in determining student success. These behavioral outcomes also suggest that students' inaccurate perceptions of their learning may encourage less effective study behaviors. However, there is evidence that

### Table 1. Descriptive Data for Major Variables and Correlations With Learning

<table>
<thead>
<tr>
<th>Variable</th>
<th>M</th>
<th>SD</th>
<th>Range</th>
<th>Quiz Score</th>
<th>Self-Reported Learning</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age</td>
<td>19.26</td>
<td>2.58</td>
<td>18–47</td>
<td>.14**</td>
<td>−.02</td>
</tr>
<tr>
<td>Grade point average (GPA)</td>
<td>2.97</td>
<td>.86</td>
<td>3.6–4.0</td>
<td>.29**</td>
<td>.13*</td>
</tr>
<tr>
<td>Textbook quality</td>
<td>6.93</td>
<td>1.29</td>
<td>2.33–9.0</td>
<td>.10**</td>
<td>.39**</td>
</tr>
<tr>
<td>Textbook usefulness</td>
<td>6.85</td>
<td>1.32</td>
<td>2–9.00</td>
<td>.09</td>
<td>.40**</td>
</tr>
<tr>
<td>Teacher behavior</td>
<td>161.82</td>
<td>21.81</td>
<td>20–180</td>
<td>.15**</td>
<td>.48**</td>
</tr>
<tr>
<td>Deep processing</td>
<td>5.41</td>
<td>1.32</td>
<td>1.33–8.33</td>
<td>.08</td>
<td>.30**</td>
</tr>
<tr>
<td>Surface processing</td>
<td>3.95</td>
<td>1.52</td>
<td>1.00–8.70</td>
<td>−.34**</td>
<td>−.26**</td>
</tr>
<tr>
<td>Organizational effectiveness</td>
<td>5.62</td>
<td>1.63</td>
<td>1.00–8.75</td>
<td>.21**</td>
<td>.33**</td>
</tr>
<tr>
<td>SERSA-quizzing</td>
<td>5.73</td>
<td>2.39</td>
<td>1.00–9.00</td>
<td>.21**</td>
<td>.26**</td>
</tr>
<tr>
<td>SERSA-metacognitive</td>
<td>6.64</td>
<td>1.52</td>
<td>1.00–9.00</td>
<td>.31**</td>
<td>.47**</td>
</tr>
<tr>
<td>Total study time</td>
<td>342.93</td>
<td>348.65</td>
<td>0–2650</td>
<td>.19**</td>
<td>.09</td>
</tr>
<tr>
<td>Review highlighted material</td>
<td>26.23</td>
<td>42.53</td>
<td>0–420</td>
<td>.14**</td>
<td>.08</td>
</tr>
<tr>
<td>Think of mnemonics</td>
<td>18.84</td>
<td>83.77</td>
<td>0–1600</td>
<td>.01</td>
<td>.02</td>
</tr>
<tr>
<td>Make up examples</td>
<td>18.05</td>
<td>25.86</td>
<td>0–250</td>
<td>.07</td>
<td>.05</td>
</tr>
<tr>
<td>Review figures and tables</td>
<td>18.13</td>
<td>27.38</td>
<td>0–300</td>
<td>.03</td>
<td>.00</td>
</tr>
<tr>
<td>Memorize definitions</td>
<td>50.46</td>
<td>68.97</td>
<td>0–500</td>
<td>.15**</td>
<td>.05</td>
</tr>
<tr>
<td>Use pedagogical aids to test knowledge</td>
<td>23.43</td>
<td>39.29</td>
<td>0–500</td>
<td>.21**</td>
<td>.16**</td>
</tr>
<tr>
<td>Read notes</td>
<td>50.56</td>
<td>58.87</td>
<td>0–500</td>
<td>.08</td>
<td>.02</td>
</tr>
<tr>
<td>Study with a friend</td>
<td>29.78</td>
<td>100.19</td>
<td>0–1080</td>
<td>.08</td>
<td>.01</td>
</tr>
<tr>
<td>Rewrite notes</td>
<td>33.74</td>
<td>49.81</td>
<td>0–434</td>
<td>.06</td>
<td>.01</td>
</tr>
<tr>
<td>Take notes from book</td>
<td>25.54</td>
<td>52.01</td>
<td>0–480</td>
<td>.17**</td>
<td>.07</td>
</tr>
<tr>
<td>Read the text</td>
<td>49.12</td>
<td>74.99</td>
<td>0–600</td>
<td>.19**</td>
<td>.13**</td>
</tr>
<tr>
<td>Self-reported Learning</td>
<td>7.05</td>
<td>1.67</td>
<td>1–9</td>
<td>.20**</td>
<td>—</td>
</tr>
<tr>
<td>Quiz score</td>
<td>6.03</td>
<td>3.97</td>
<td>0–14</td>
<td>—</td>
<td>.20**</td>
</tr>
</tbody>
</table>

Note. SERSA = self-evaluation related study activity. Total study time to read the text measured in minutes.

*p < .05. **p < .01.
interventions to improve student awareness of their learning and develop study skills can be effective (Winne & Nesbitt, 2010). For example, Hattie (2009) found that interventions to improve study skills were significant in improving learning with effect sizes ranging from .59 (for teaching study skills) to .71 (for spaced vs. mass practice) and .69 (metacognitive strategies), significantly above the average effect sizes found (.40). Indeed, our data also indicate a positive correlation between a student’s use of organizationally effective behaviors and quiz scores. We recommend introductory psychology teachers consider designing exercises that get students to improve awareness of learning.

We were surprised by some lack of associations. For example, instructor rating did not predict quiz score, although it was highly related to student self-perceptions of learning or vice versa. The literature is clear that instructor characteristics are significantly related to student motivation, enthusiasm for the subject, and enjoyment of the course (Buskist, 2004), in addition to perceptions of learning. It is noteworthy that once we included in the analysis other factors more closely related to student effort and behavior, the role of the instructor was diminished. Similarly, ratings of textbook quality and helpfulness did not predict quiz scores or self-reported learning. Whereas both textbooks and instructors are more closely related to student effort and behavior, the role of learning or vice versa. The literature is clear that instructor characteristics are significantly related to student motivation, enthusiasm for the subject, and enjoyment of the course (Buskist, 2004), in addition to perceptions of learning. It is noteworthy that once we included in the analysis other factors more closely related to student effort and behavior, the role of the instructor was diminished. Similarly, ratings of textbook quality and helpfulness did not predict quiz scores or self-reported learning. Whereas both textbooks and instructors are more closely related to student effort and behavior, the role

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