## Perceived Grading and Student Evaluation of Instruction

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Abstract

We analyzed student evaluations for 3,585 classes collected over 20 years to determine 13 stability and evaluate the relationship of perceived grading to global evaluations, perceived 14 fairness, and appropriateness of assignments. Using class as the unit of analysis, we found 15 small evaluation reliability when professors taught the same course in the same semester, 16 with much weaker correlations for differing courses. Expected grade and grading related 17 questions correlated with overall evaluations of courses. Differences in course evaluations on 18 expected grades, grading questions, and overall grades were found between full-time faculty 19 and other types of instructors. These findings are expanded to a model of grading type questions mediating the relationship between expected grade and overall course evaluations 21 with a moderating effect of type of instructor.

23 Keywords: Student evaluation, teacher evaluation, perceived grading, reliability

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#### Perceived Grading and Student Evaluation of Instruction

Student evaluations of professors have been disputed over time with regard to validity 25 and reliability. The impact of student evaluations on professor advancement can be great 26 and often acts as a deciding factor in professor promotion or demotion, along with the access 27 to certain funding opportunities, coursework choice, and tenureship. There are certain variables researched that result in improving evaluations, such as giving higher grades (???; 29 ???; ???). Student evaluations have also been found to be influenced by likability, 30 attractiveness, and dress (???; ???; ???). Further, 20 years ago, (Neath 1996) suggests 20 31 tips in which professors may bolster their evaluations from students that have no relationship with proven instructional methods or further learning retention among the 33 student body, such as being a male professor and only teaching only male students. In more recent research, (???) confirms that student evaluations of teaching are biased against female instructors, and the authors conclude student evaluations are more representative of the students' grading expectations and biases rather than an evaluation of objective instructional methods. All together, these findings elicit the argument that student evaluations are not necessarily measuring whether the instructional methods of professors are sound, rather student evaluations of instruction are measuring whether or not the professor is likeable among the students with regard to their expectation of their performance in the classroom, in addition to the instructor meeting pre-existing biases. 42 However, some authors (???; ???; ???) have discovered professors are not able to 43 increase their positive evaluations by only providing their students with higher grades. We believe this is due to what we consider the effect of "perceived grading". We operationally define perceived grading as the students' perceptions of assignment appropriateness, grading fairness, and the expected course grade at the time the evaluations are being completed. We believe social psychology theory would support that students with low perceived grading may reduce cognitive dissonance and engage in ego defense by giving low evaluations of professors who give them lower grades (???), resulting in decreased validity and reliability of

the proposed construct, professor instruction. We argue that both social psychology theory
and the evidence from student evaluations supports that higher perceived grading can lead
to better student evaluations of instruction. This theory and evidence from student
evaluation leads us to further posit student evaluations of professors as biased methods of
data collection and irrelevant to the quality of the instructor and the instructional methods
used over the course of a semester.

Much of the literature on student evaluations involves diverse and complex analyses

(e.g., (???)) and lacks social-psychological theoretical guidance on human judgment. To

expect that student evaluations would not be influenced by expected grade would contradict

a long-standing history of social psychology research on cognitive dissonance, attribution,

and ego threat. As we know, failure threatens the ego [(???); Snyder, Stephan & Rosenfield,

1978] and motivates us to find rationales to defend the ego. Failing students, or those

performing below personal expectations, would be expected to defend their ego by

attributing low grades to poor teaching or unfair evaluation practices (???). One common

strategy involves diminishing the value of the activity (???), which would result in lowered

perceived value of a course.

Similarly, Cognitive Dissonance Theory (???) predicts that people who experience poor performance but perceive themselves as competent will experience dissonance, of which they can reduce through negative evaluations of the instruction (???). Attribution research (???) also supports the argument that among low achievement motivation students, failure is associated with external attributions for cause, and the most plausible external attribution is the quality of instruction and grading practices. Although arguments regarding degree of influence are reasonable, the position that they are not affected is inconsistent with existing and established theory. Thus, it is not surprising that the majority of faculty perceive student evaluations to be biased by perceived grading and course choice (???).

Considerable research has been conducted in support of widely distributed evaluation systems. (???) reported that in a study of 9,194 class averages using the Student

Instructional Support, the relationship between expected grades and global ratings was only
20. He further argued that when variance due to perceived learning outcomes was regressed
from the global evaluation, the effect of expected grades was eliminated. However, a
student's best assessment of "perceived learning outcome" is their expected grade, and thus,
these should be highly correlated. When perceived learning is regressed from the global
evaluations, it is not surprising that suppression effects would eliminate or could even reverse
the correlation between expected grade and global evaluation. In general, there are several
reasons why the relationship of expected grade to global evaluations is suppressed. For
example, faculty ratings are generally very high on average (i.e. quality instructors are hired),
which restricts variation; thus, weakening their reliability as a measure of professor
attributes. This restriction in range suppresses correlation.

However, (???) provided causal evidence of lowered student evaluations due to
expected grades. In her study of 444 students completing faculty evaluations at two separate
points in a semester, students who expected to get Fs significantly lowered their evaluations
while students who expected to receive As and Bs significantly raised their evaluations.

(???) argued that the individual is also not the proper unit of analysis because such 93 analyses could suggest false findings related to individual differences in students. Therefore, he argued the use of class as the suggested unit of analysis. We agree, both for his reasoning and because analyses with individual ratings can mask significant relationships as well (do we have a source for this claim???). Individual differences in expectancy will attenuate the 97 correlation less when class average is used as the unit of analysis. To the extent that the same class average would be expected across all courses, an assumption we will challenge, the class average for expected grade is a good measure of perceived grading as an instructor attribute. Course quality, not individual attributes of students, is what we are attempting to 101 assess when we are using student evaluations of courses. Several studies provide support that 102 when class is the unit of analysis, expected grade is a more significant biasing factor in 103 student evaluations (???; ???; ???).

Additionally, (???) analyzed 167 psychology classes in a multiple regression analysis 105 with class as the unit of analysis and found that the two most significant predictors of 106 instructor ratings were average grade given by the instructor and instructor status (TA or 107 rank of faculty). Because of the limited number of classes, the power of the analysis was 108 limited. However, in addition to the concern regarding the relationship between grades and 100 global course evaluations, it was found that TAs were rated more highly than ranked faculty. 110 This finding raises additional questions on validity student evaluation of instructional quality. 111 We must either accept that the least trained and qualified instructors are actually better 112 teachers, or we must believe this result suggests that student evaluations have given us false 113 information on the quality of instruction via their perceptions of grading. 114

(???) provided further evidence that using course as a unit of analysis increased the 115 correlation between expected grade and other course ratings. Within specific groupings of 116 classes, these correlations ranged from .23 to .53. Two factors limited the level of their 117 relationships. First, the classes used were all upper division courses and graduate courses. 118 Secondly, over 90% of the students in these classes expected an A or a B. Consequently, the 119 correlations between expected grade and global course ratings would be reduced due to the 120 absence of variation in expected grades. Similarly, (???) found a correlation of .35 between average course grade and average rating of the instructor in 165 classes during a two-year period. However, these studies did not consider the predictive relationship for instructors 123 across different courses and semesters, which was one aim of the current study. 124

It is pertinent to note that different disciplines and subject areas have diverse GPA standards, and students have differing grade and workload expectations in different courses, as well. For example, an instructor in Anatomy giving a 3.00 GPA might be considered lenient while an Education instructor giving a 3.25 GPA might be considered hard (examples for illustration only). To have a valid measure of workload and leniency factors, correlations should be conducted with varied teachers of the same course. Further, different populations take courses in different disciplines, resulting in potential population differences between

anatomy classes and education classes, which could create or mask findings. Hence, analysis
of these correlations within the same discipline and course would be expected to strengthen
the relationship between expected grades and quality measures, offering more valid results.

Further, in most studies of student evaluations, reliability is established through 135 internal consistency reliability. However, this form of reliability is confounded with halo 136 effects (i.e. a cognitive bias that influences ratings based on an overall perception of the 137 person teaching, rather than the individual components of the course), and tells only 138 whether the individual responding to the questions is consistent and reliable (???? do we have 139 a source for halo effects from internal consistency reliability????). By having many different 140 classes for the same instructor, we can establish the reliability of ratings across the same and 141 different courses during the same and different semesters. As a result, we should be able to 142 deduce if student ratings can be considered a valid measure of an instructor's teaching skills 143 if they are or are not able to reliably differentiate instructors within the same course across 144 different semesters.

If ratings are, in fact, valid measures of instructor attributes, it should be expected
that ratings would have some stability across semester and specific course taught. If
variation were due to instructor attributes and not the course they are assigned, we would
expect ratings to be most stable across two different courses during the same semester. We
would expect these correlations to decline somewhat for the same course in a different
semester, since faculty members may improve or decline with experience. But if they are
reliable and stable enough to use in making choices about retention, their stability should be
demonstrated across different semesters, as well. Therefore, in the current study, we first
sought to establish reliability of ratings for the instructors across courses and semesters.

The current study used data collected over a 20-year period to allow for more powerful analyses, with such analyses occurring within many sections of the same course at the same university. After examining reliability, we sought to show that items on instructor evaluations were positively correlated for undergraduate and graduate students (???didn't we

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want to eliminate graduate students from the analyses because they're a special 159 population???), demonstrating that overall course evaluations are related to the perceived 160 grading of the students. We also expected correlations to be substantially higher than those 161 obtained by previous researchers who used individual students as their unit of analysis, since 162 we used the course as the unit of analysis. Next, we examined if rating differences across 163 these questions were found between types of instructors compared to full-time faculty, such 164 as teaching-assistants and per-course faculty. The presumption of university hiring 165 requirements that include a terminal degree for regular faculty is that better-trained faculty 166 will be more effective teachers. Therefore, if student evaluations are a valid measure, 167 better-trained, full-time faculty should receive higher ratings than per-course instructors and 168 teaching assistants. However, existing literature appears to contradict this expectation (???). 169 Given these differences, we proposed and examined a moderated mediation analysis to portray the expected relationship of the variables across instructor type. 171

Method 172

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The archival study was conducted using data from the psychology department at a 173 large Midwestern public university. We used data from 4313 undergraduate, 397 mixed-level 174 undergraduate, and 687 graduate psychology classes taught from 1987 to 2016 that were 175 evaluated by students using the same 15-item instrument. The graduate courses were 176 excluded from analyses due to the ceiling effects on expected grades. Faculty followed set 177 procedures in distributing scan forms no more than two weeks before the conclusion of the 178 semester. A student was assigned to collect the forms and deliver them to the departmental 179 secretary. The instructor was required to leave the room while students completed the forms. 180 We focused upon the five items, which seemed most pertinent to the issues of perceived 181 grading and evaluation. We were most interested in how grades related to global course 182 evaluation and grading/assignment evaluations. These items were presented with a five-point 183 scale from 1 (strongly disagree) to 5 (strongly agree):

- 185 1. The overall quality of this course was among the top 20% of those I have taken.
- 2. The examinations were representative of the material covered in the assigned readings
- 3. The instructor used fair and appropriate methods in the determination of grades.
- 188 4. The assignments and required activities in this class were appropriate.
- 5. What grade do you expect to receive in this course? (A = 5, B, C, D, F = 1).

190 Results

All data were checked for course coding errors, and type of instructor was coded as 191 graduate assistant, per-course faculty, instructors, and tenure-track faculty. This data was 192 considered structured by instructor; therefore, all analyses below were coded in R using the 193 nlme package (???) to control for correlated error of instructor as a random intercept in a 194 multilevel model. The overall dataset was screened for normality, linearity, homogeneity, and 195 homoscedasticity using procedures from (???). Data generally met assumptions with a slight 196 skew and some heterogeneity. This data was not screened for outliers because it was 197 assumed that each score was entered correctly from student evaluations. The complete set of 198 all statistics can be found online at http://osf.io/jdpfs. This page also includes the 199 manuscript written inline with the statistical analysis with the papaja package (???) for interested researchers/reviewers. 201

#### Reliability of Instructor Scores DONE

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Reliability of ratings of instructors can be inferred by the consistency of ratings across courses and semester, assuming that we infer there is a stable good/poor instructor attribute and that these multiple administrations of the same question are multiple assessments of that attribute. A file was created with all possible course pairings for every instructor, semester, and course combination. Therefore, this created eight possible combinations of matching v. no match for instructor by semester by course. Multilevel models were used to calculate correlations on each fo the eight combinations controlling for response size for both courses (i.e., course 1 number of ratings and course 2 number of ratings) and random

intercepts for instructor(s). Correlations were calculated separately for each question, 211 however, the overall pattern of the data was the same for each of the eight combinations, and 212 these were averaged for Table @ref:(tab:rel-table). The complete set of all correlations can be 213 found online. Because the large sample size can bias statistical significance, we focused on 214 the size of the correlations. The correlations were largest for the same instructor in the same 215 semester and course, followed by the same instructor in the same semester with a different 216 course and the same instructor in a different semester with the same course. The first shows 217 that scores are somewhat reliable (i.e.,  $rs \sim .45$ ) for instructors teaching two or more of the 218 same class at the same time. The correlations within instructor then drop to  $rs \sim .09$  for the 219 same semester or same course. All other correlations are nearly zero, with the same semester, 220 same course, and different instructor as the next largest at  $rs \sim .05$ . Given these values are 221 still low for traditional reliability standards, this results may indicate that student demand characteristics or course changes impact instructor ratings. 223

#### 224 Correlations of Evaluation Questions DONE

Table 2 presents the inter-correlations for the five relevant evaluation questions using 225 instructor as a random intercept in a multilevel model with evaluation sample size as an 226 adjustor variable. The partial correlation (pr) is the standardized coefficient from the 227 multilevel model analysis between items while adjusting for sample size and random effects 228 of instructor. The raw coefficient b, standard error, and significance statistics are also provided. We found class expected grade was related to class overall rating, exams reflecting the material, grading fairness, and appropriateness of assignments; however, these partial correlations were approximately half of all other pairwise correlations. The correlations 232 between grading related items were high, representing some consistency in evaluation, as well 233 as the overall course evaluation to grading questions. 234

#### Instructor Status and Ratings DO WE WANT THIS

We compared teaching assistants, per-course faculty, instructors, and ranked faculty in 236 undergraduate courses that included evaluations for all four types of teacher, usually general 237 education classes (i.e., Introductory Psychology), required major courses (i.e., Statistics, 238 Research Methods), and popular electives (i.e., Abnormal Psychology). This analysis 239 included 179 teachers: 49 teaching assistants, 54 per-course instructors, 17 instructors, and 240 59 full-time faculty who taught 2744 courses: 266 teaching assistants, 400 per-course 241 instructors, 354 instructors, and 1724 full-time faculty. 242 . All comparisons were made against full-time faculty to control for Type 1 error using 243 a multilevel model with a dummy coded instructor variable, and dummy coded t values were 244 used to determine which comparison groups were different from full-time faculty. Overall means and standard deviations are presented in Table 3, and the complete set of t value comparisons for these analyses can be found online. As shown in the Table 3, the ratings of 247 all groups were fairly high, hovering around 4.00 on a 5.00 point scale, and the expected 248 grade for courses was approximately a B. 249 For overall ratings, faculty were found to be rated less highly than teaching assistants, 250 p = .027, but not significantly different than per-course faculty (p = .181) or instructors (p 251 = .814). When rating if exams were representative of course material, full-time faculty were 252 rated lower than both teaching assistants (p < .001) and per-course faculty (p = .047), but 253 were not significantly different than instructors (p = .740). Full-time faculty were rated as 254 less fair and appropriate in their grades than teaching assistants (p = .003), while per-course 255 faculty (p = .128) and instructors (p = .657) had similar scores to faculty. Teaching assistants were designated to have more appropriate assignments than faculty (p < .001), while per-course (p = .060) and instructors (p = .073) had the same ratings as faculty on 258 assignments. Finally, faculty showed significantly lower expected grades than teaching 259 assistants (p < .001) and per-course faculty (p = .044), while having similar grades to 260 instructors (p = .705). 261

#### Moderated Mediation

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Given the correlations between items and differences between items and ranked faculty, 263 we proposed a mediation relationship between expected grade, perceived grading, and overall 264 course grades that varies by instructor type. Figure 1 demonstrates the predicted relationship 265 between these variables. We hypothesized that expected course grade would impact the 266 overall course rating, but this relationship would be mediated by the perceived grading in 267 the course, which was calculated by averaging questions about exams, fairness of grading, 268 and assignments. Therefore, as students expected to earned higher grades (leniency), their 269 perception and ratings of the grading would increase, thus, leading to higher overall course 270 scores. This relationship was tested using traditional and newer approaches to mediation (???; Baron & Kenny, 1986). All categorical interactions were compared to ranked faculty. Each step of the model is described below. Because significant interactions were found, we 273 calculated each group separately (Figure 1) to portray these differences in path coefficients. 274 Tables of t values for the overall and separated analyses are available at http://osf.io/jdpfs. 275 First, expected grade was used to predict the overall rating of the course, 276 along with the interaction of type of instructor and expected grade. The expected grade positively predicted overall course rating, p < .001, wherein higher expected grades was 278 related to higher overall ratings for the course (b = 0.39). A significant interaction between type and expected grade rating was found for instructors versus faculty. In looking at Figure 280 1, we find that instructors (b = 0.56) have a stronger relationship between expected grade 281 and overall course rating than faculty (b = 0.39, interaction p = .020), while per-course (b = 0.39) 282 0.41, interaction p = .621) and teaching assistants (b = 0.71, interaction p = .068) were not 283 significantly different than faculty on the c path coefficient. 284 A Path. Expected grade was then used to predict the average of the grading related 285 questions, along with the interaction of type of instructor. Higher expected grades were 286 related to higher ratings of appropriating grading (b = 0.21, p < 0.01), and a significant 287

interaction of faculty and all three other instructor types emerged: teaching assistants (p =

.001), per-course faculty (p = .001), and instructors (p < .001). As seen in Figure 1, faculty 289 (b = 0.21) have a much weaker relationship between expected grade and average ratings of 290 grading than teaching assistants (b = 0.55), per-course (b = 0.41), and instructors (b = 291 0.45). B and C' Paths. In the final model, expected grade, average ratings of grading, and 292 the two-way interactions of these two variables with type were used to predict overall course 293 evaluation. Average rating of grading was a strong significant predictor of overall course 294 rating (b = 1.10, p < .001), indicating that a perception of fair grading was related 295 positively to overall course ratings. An interaction between per-course faculty and fair 296 grading emerged, p < .001, wherein faculty (b = 1.10) had a less positive relationship than 297 per-course (b = 1.28), while teaching assistants (b = 1.37, interaction p = .071) and 298 instructors (b = 1.16, interaction p = .187) were not significantly different coefficients. The 299 relationship between expected grade and overall course rating decreased from the original model (b = 0.16, p < 0.001). However, the interaction between this path and per-course (p < .001) and instructors (p = .041) versus faculty was significant, while faculty versus teaching 302 assistants' paths were not significantly different (p = .133). Faculty relationship between 303 expected grade and overall course scoring, while accounting for ratings of grading was 304 stronger (b = 0.16) than instructors (b = 0.04) and per-course (b = -0.10), but not that of 305 teaching assistants (b = -0.04). 306

## ###Mediation Strength

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We then analyzed the indirect effects (i.e. the amount of mediation) for each type of 308 instructor separately, using both the Aroian version of the Sobel test (Baron & Kenny, 1986), 309 as well as bootstrapped samples to determine the 95% confidence interval of the mediation 310 (Preacher & Hayes, 2008; ???) because of the criticisms on Sobel. For confidence interval 311 testing, we ran 5,000 bootstrapped samples examining the mediation effect and interpreted 312 that the mediation was different from zero if the confidence interval did not include zero. For 313 teaching assistants, we found mediation significantly greater than zero, indirect = 0.74 (SE 314 = 0.14), Z = 5.15, p < .001, 95% CI[0.48, 1.02]. Per-course faculty showed mediation 315

between expected grade and overall course rating, indirect = 0.52 (SE = 0.09), Z = 6.06, p < .001, 95% CI[0.36, 0.73]. Instructors showed a similar indirect mediation effect, indirect = 0.53 (SE = 0.07), Z = 7.31, p < .001, 95% CI[0.40, 0.66]. Last, faculty showed the smallest mediation effect, indirect = 0.23 (SE = 0.02), Z = 8.71, p < .001, 95% CI[0.19, 0.28], wherein the confidence interval did not include zero, but also did not overlap with any other instructor group.

322 Discussion

The findings support the model that student evaluations of Psychology faculty are 323 related to what one might consider leniency (i.e., overall average scores of B) in grading 324 through perceptions of assignment appropriateness, grading fairness, and the expected course 325 grade. This position is supported both in the strong relationships between expected grade 326 and global ratings by the evidence that greater training and experience is related to poorer 327 evaluations, lower expected grades, and lower relationships between grading and evaluations. 328 Faculty received lower scores than teaching assistants in every category and often lower scores than per-course faculty, but not instructors. Mediation analyses showed that expected grade is positively related to overall course ratings, although this relationship is mediated by 331 the perceived grading in the course. Therefore, as students have higher expected grades, the 332 perceived grading scores increase, and the overall course score also increases. Moderation of 333 this mediation effect indicated differences in the strength of the relationships between 334 expected grade, grading questions, and overall course rating, wherein faculty generally had 335 weaker relationships between these variables. 336

Because the study was not experimental, causal conclusions from this study alone need to be limited. However, (???) provides some evidence of the causal direction of student ratings of instructors and expected grades. She had 444 students complete faculty evaluations after 3-4 weeks of classes, and again after 13 weeks. Students who expected to get Fs significantly lowered their evaluations while students who expected to receive As and Bs significantly raised their evaluations.

It is compelling that the correlations suggest that we can do a better job of 343 understanding global ratings, perception of exams, fairness, and appropriateness of 344 assignments based upon what grade students expected as compared to relating these ratings 345 using ratings for the same course in a different semester or ratings for a different course in 346 the same semester for instructor (i.e., correlations between items in the same semester are 347 higher than reliability estimates across the board). It is very likely that these correlations 348 with expected grade are suppressed by the loading of scores at the high end of the scale for 349 course ratings and expected grade. Generally, evaluation items reflect scores at the high end 350 of the 1-5 scale (see Table 3) even when items are intentionally constructed to move 351 evaluators from the ends. The item, "The overall quality of this course was among the top 352 20% of those I have taken," is conspicuously designed to move subjects away from the top 353 rating. Yet average global ratings remain about a 4.00. The grade expectation average was 354 approximately 4.00, which relates to a B average or 3.00 GPA. 355

One way of establishing convergent validity would be a finding that better trained and 356 more experienced teachers get higher ratings than less well trained instructors. If the 357 measure were valid, we would expect that regular faculty and full time instructors would get 358 higher ratings than per course faculty and teaching assistants. To argue otherwise is to 359 challenge the merits of higher education units with a faculty of professors with doctoral 360 status. If the university were a researcher powerhouse where faculty research is emphasized 361 over teaching and graduate assistants are admitted from the highest ranks of undergraduates, 362 the finding that teaching assistants and per course faculty get higher ratings might be less of a challenge to the validity of these ratings. However, the university at which the data were collected is a non-doctoral program with greater emphasis on teaching and moderate emphasis on research, and teaching assistants are master's candidates with less substantial admission expectations than doctoral programs. Hence, these findings challenge the 367 convergent validity of the teaching evaluations. 368

Like most studies in this area, a major limitation is the absence of an independent 369 measure of learning. Of course, this limitation is based upon the belief that the goal is to 370 create educated persons, not just satisfied consumers. Even when common tests are used, 371 these are invalid if the instructors are aware of the course content. Teachers seeking high 372 evaluations are able to improve their ratings and scores by directly addressing the content of 373 the specific test items. ETS now allows faculty who administer Major Field Tests to access 374 the specific items which thereby invalidates it as a measure for these purposes. Ultimately, 375 answering questions about the validity of student evaluations is a daunting task without such 376 measures. 377

Evidence suggests that student evaluations are influenced by likability, attractiveness, 378 and dress (???; ???; ???) in addition to leniency and low demands (???). One must 379 question whether a factor like instructor warmth, which relates to student evaluation (???), 380 is really fitting to the ultimate purposes of a college education. In a unique setting where 381 student assignments to courses were random and common tests were used, (???) 382 demonstrated that teaching strategies that enhanced student evaluations led to poorer 383 performance in subsequent classes. With the sum of invalid variance from numerous factors 384 being potentially high, establishment of a high positive relationship to independent measures of achievement is essential to the acceptance of student evaluations as a measure of teaching quality. 387

Perception of the influence of leniency on teacher evaluations is far more detrimental to
the quality of education than the biased evaluations themselves. It is unlikely that good
teachers, even if more challenging, will get bad evaluations (i.e. evaluations where the
majority of students rate the course poorly). Good teachers are rarely losing their positions
due to low quality evaluations. But (???) found that faculty perceives evaluations to be
biased based upon course difficulty (72%), expected grade (68%), and course workload (60%).
If one's goal is high merit ratings and teaching awards, and the most significant factor is
student evaluations of teaching, then putting easier and low-level questions on the test,

adding more extra credit, cutting the project expectations, letting students off the hook for missing deadlines, and boosting borderline grades would all be likely strategies for boosting evaluations.

Effective teachers will get positive student ratings even when they have high 399 expectations and do not inflate grades. But, many excellent teachers will score below 400 average. It is maladaptive to try to increase a 3.90 global rating to a 4.10, because it often 401 requires that the instructor try to emphasize avoidance of the lowest rating (1.00) because 402 these low ratings in a skewed distribution have in inordinate influence on the mean. This 403 effort of competing against the norms is likely to lead to grade inflation and permissiveness 404 for the least motivated and most negligent students. Some researchers (???; ???) argue that 405 student evaluations of instruction should be adjusted on the basis of grades assigned. 406 However, there are problems with such an approach. The regression Betas are likely to differ 407 based upon course and many other factors. In our research and in research by DuCettte and 408 Kenney (1982), substantial variation in correlations was found across different course sets. 400 Establishing valid adjustments would be problematic at best. Further, such an approach 410 would punish instructors when they happen to get an unusually intelligent and motivated 411 class (or teach an honors class) and give students the grades they deserve. Student evaluations are not a proper motivational factor for instructors in grade assignment, whether 413 it is to inflate or deflate grades.

It would seem nearly impossible to eliminate invalid bias in student ratings of
instruction. Yet, they may tell us a teacher is ineffective when the majority give poor ratings.

It is the normative, competitive use that makes student evaluations of teaching subject to
problematic interpretation. This finding is especially critical in light of recent research that
portrays that student evaluations are largely biased against female teachers, and that
student bias in evaluation is related to course discipline and student gender (???). (???)
also examine the difficulty in adjusting faculty evaluation for bias and determined that the
complex nature of ratings makes unbiased evaluation nearly impossible. (???) further

explain that evaluations are often negatively related to more objective measures of teaching 423 effectiveness, and biased additionally by perceived attractiveness and ethnicity. In line with 424 the current paper, he suggests dropping overall teaching effectiveness or value of the course 425 type questions because they are influenced by many variables unrelated to actual teaching. 426 Last, they suggest the distribution and response rate of the data are critical information, and 427 this point becomes particularly important when recent research shows that online 428 evaluations of teaching experience a large drop in response rates (???). Our study 429 contributes to the literature of how student evaluations are a misleading and unsuccessful 430 measure of teaching effectiveness, especially focusing on reliability and the impact of grading 431 on overall questions. We conclude that it may be possible to manipulate these values by 432 lowering teaching standards, which implies that high stakes hiring and tenure decisions 433 should probably follow the advice of (???) or (???) in implementing teaching portfolios and syllabus review, particularly because a recent meta-analysis of student evaluations showed 435 they are unrelated to student learning (???).

# References

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 $\label{thm:constructor} \begin{tabular}{ll} Table 1 \\ Correlations for Instructor, Semester, and Course Combinations \\ \end{tabular}$ 

Instructor	Semester	Course	b	SE	t	df	p
Different Instructor	Different Semester	Different Course	001	.000	10144295	-3.58	.013
Different Instructor	Same Semester	Different Course	.006	.002	152801	2.91	.048
Different Instructor	Different Semester	Same Course	.008	.001	517353	6.24	.027
Different Instructor	Same Semester	Same Course	.054	.010	6265	5.40	< .001
Same Instructor	Different Semester	Different Course	038	.003	108849	-13.13	< .001
Same Instructor	Same Semester	Different Course	.095	.020	1872	4.66	< .001
Same Instructor	Different Semester	Same Course	.090	.004	55057	21.77	< .001
Same Instructor	Same Semester	Same Course	.446	.023	1401	19.63	< .001

 $\begin{tabular}{ll} Table 2 \\ t \ Statistics \ for \ Undergraduate \ Correlations \\ \end{tabular}$ 

Coefficient	*pr*	*b*	*SE*	*df*	*t*	*p*
Overall to Exams	.637	.828	.014	4447	60.813	< .001
Overall to Fair	.606	.903	.016	4447	57.837	< .001
Overall to Assignments	.675	.999	.016	4447	63.251	< .001
Overall to Grade	.344	.597	.022	4447	27.167	< .001
Exams to Fair	.655	.751	.012	4447	61.387	< .001
Exams to Assignments	.615	.700	.014	4447	50.425	< .001
Exams to Grade	.311	.416	.018	4447	23.066	< .001
Fair to Assignments	.720	.715	.011	4447	63.912	< .001
Fair to Grade	.375	.438	.016	4447	27.865	< .001
Assignments to Grade	.344	.404	.015	4447	26.913	< .001