Creating birds of similar feathers

Leveraging similarity to improve teacher-student relationships and academic achievement

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Abstract

When people perceive themselves as similar to others, greater liking and closer relationships typically result. In the first randomized field experiment that leverages actual similarities to improve real-world relationships, we examined the affiliations between 315 ninth grade students and their 25 teachers. Students in the treatment condition received feedback on five similarities that they shared with their teachers; each teacher received parallel feedback regarding about half of his/her ninth grade students. Five weeks after our intervention, those in the treatment conditions perceived greater similarity with their counterparts. Furthermore, when teachers received feedback about their similarities with specific students, they perceived better relationships with those students, and those students earned higher course grades. Exploratory analyses suggest that these effects are concentrated within relationships between teachers and their “underserved” students. This brief intervention appears to close the achievement gap at this school by over 60%.

Humans foster social connections with others as a fundamental, intrinsic social motivation – we are hard-wired to be social animals (Lieberman, 2013; Ryan & Deci, 2000). Those who more successfully relate to others experience a broad constellation of positive outcomes ranging from greater happiness (Gilbert, 2006) to superior health (Taylor et al., 2004). Children who thrive typically cultivate positive relationships with parents, peers, and teachers (Wentzel, 1998). Even for adolescents, achieving positive teacher-student relationships (TSRs) is an important outcome in its own right and may catalyze important downstream benefits (Eccles et al., 1993).

Thus, for those who study positive youth development, schooling, and social motivation (e.g., Bronk, 2012; Pintrich, 2003) the topic of improving TSRs sparks tremendous interest. One promising approach might leverage individuals’ perceptions of similarity as a means to promote a sense of relatedness. Numerous basic social psychological texts underscore some version of the basic message that “likeness begets liking” (Myers, 2015, p. 330). Similarity along various dimensions (style of dress, background, interests, personality traits, hobbies, attitudes, etc.) connects to a wide array of relationship-related outcomes (such as attraction, liking, compliance, and prosocial behavior) in scores of studies (Cialdini, 2009; Montoya, Horton, & Kirchner, 2008).

The theory behind the promise of this approach is that interacting with similar others supports one’s sense of self, one’s values, and one’s core identity (Montoya et al., 2008; Myers, 2015). In other words, as an individual interacts with similar others, she reaps positive reinforcement in the form of validation. For instance, imagine a 9th grade student enrolling in high school in a new town. As she encounters peers who also value religion, enjoy sports, participate in math club, and aspire to attend college, she learns that her values and beliefs are socially acceptable within her new community. Continuing to affiliate with these individuals will reinforce a perception that her values and beliefs have merit. Conversely, her peers who eschew religion, think sports are silly, ridicule math club, and see no point in college will cast doubt on the values and beliefs that lie at the core of her identity. Spending time with these students will not be reinforcing. In this way, similarity acts as a powerfully self-affirming motivator (Brady et al., this issue) in the context of friendships and close relationships.

Unfortunately, a fundamental problem arises in using similarity to improve relationships: people either share something in common or they do not. Thus, scholars can develop experimental manipulations of similarities but these interventions typically rely upon fictitious similarities (e.g., Burger, Messian, Patel, del Prado, & Anderson, 2004; Galinsky & Moskowitz, 2000). While these studies enable causal inferences to be made, the fictitious nature of the similarities minimizes their utility for real-world interventions. On the other hand, numerous correlational studies have identified real similarities between individuals in real relationships and have shown that these similarities correspond with improved relationship outcomes (e.g., Chen, Luo, Yue, Xu, & Zhaoyang, 2009; Gonzaga, Campos, & Bradbury, 2007; Ireland et al., 2011). However, the correlational nature of these studies precludes causal inferences from being made. Thus, how scholars might successfully leverage real similarities to improve real-world relationships, such as TSRs, remains a vexing challenge.

In this study, we test the effects of an intervention that potentially mitigates these trade-offs. Specifically, we experimentally manipulate perceptions of veridical similarities as a means to try and improve TSRs between ninth graders and their teachers. In addition to examining TSRs as a key outcome, we note that these relationships have shown robust associations with consequential student outcomes (McLaughlin & Clarke, 2010). Thus, we also test whether the intervention affects students’ classroom grades. To our knowledge, this is the first experimental
study to use actual similarities as a means to improving real, ongoing relationships.

Similarity and Relationships

Of the research connecting similarity and interpersonal relationships, two main types of studies proliferate: those that have fabricated similarities for the sake of experimental manipulations and those that have investigated actual similarities. Both types of studies have enhanced scientific understanding of the importance and potency of similarity in relationships. Across both the experimental and correlational approaches, two notable themes emerge.

First, the content of the similarities associated with improved relationship outcomes covers an impressively disparate array of topics. For example, scholars have experimentally manipulated the similarity of names to boost liking and compliance. One researcher bolstered return rates on a questionnaire by using names on a cover letter that were similar to respondents’ own names (Garner, 2005). In a series of primarily correlational studies, Mackinnon, Jordan, and Wilson (2011) found that students who are physically similar to one another (e.g., both wearing glasses) will tend to sit next to one another in class. Using both experimental and correlational approaches, Boer et al. (2011) found that shared music preferences helped foster closer social bonds between people.

Although few scholars have explored the idea of using similarities to improve relationships in education, some have examined whether students perform better academically when their teacher shares their ethnicity. For instance, Dee (2004) found significant positive effects on test score outcomes for black students who were assigned to black teachers and for white students who were assigned to white teachers. Although he does not examine TSRs, he does hypothesize that trust and role-modeling may be crucial mechanisms in explaining his findings.

Second, even the most trivial similarities can lead to positive sentiments toward another person. Laboratory experiments informing participants that they and another participant share: a preference for Klee versus Kandinsky paintings (Ames, 2004), the tendency to over- or under-estimate the number of dots on a computer screen (Galinsky & Moskowitz, 2000), or purported similarity in fingerprint patterns (Burger et al., 2004), have all enhanced relationship-related outcomes. Correlational studies show comparably surprising findings. For example, people who have similar initials are disproportionately likely to get married (Jones, Pelham, Carvallo, & Mirenberg, 2004).

Despite their contributions, these two approaches to studying the connections between similarity and relationships leave two important gaps in our knowledge. First, this work leaves open the crucial scientific question of whether real similarities cause improved outcomes in real relationships. Certainly, the preponderance of this experimental and correlational evidence, generalized across so many types of similarities – including ones that seem especially unimportant – suggests that this causal association should exist. However, without direct experimental evidence, some doubt remains.

A second gap in our knowledge is particularly salient for educational practitioners. Without some way to leverage real similarities between individuals within a classroom, the associations between similarity and relationship outcomes have limited practical applications. Car salespeople may be well-served by suggesting that they too enjoy camping, golf, or tennis if they notice tents, clubs, or rackets in the trunk of your car (Cialdini, 2009). However, teachers who lie about what they share in common with individual students will likely be found out over the course of an ongoing relationship (to say nothing of the ethically dubious
nature of this tactic). One could argue that teachers might leverage similarity by learning what students have in common with each other and assigning them to collaborative groups with like-minded classmates. However, it seems important for schools to socialize students to work effectively with those from different backgrounds. In sum, as compelling and robust as the similarity-relationship research is, important scientific and applied gaps plague our understanding of these associations.

**Teacher-Student Relationships and Student Outcomes**

In addition to healthy relationships as an important outcome in their own right (Leary, 2010), TSRs matter because they are associated with a broad array of valued student outcomes including: academic achievement, affect, behavior, and motivation. As McCombs (2014) concludes from a series of studies she conducted, “What counts and what leads to positive growth and development from pre-kindergarten to Grade 12 and beyond is caring relationships and supportive learning rigour” (p. 264).

Many studies have shown that students with better TSRs tend to achieve more highly in school (Cornelius-White, 2007; Roorda, Koomen, Split, & Oort, 2011). For example, Wentzel (2002) found that middle-school students’ perceptions of their teachers on relational dimensions such as fairness and holding high expectations predicted students’ end-of-year grades. Estimated effect sizes of TSRs on achievement range from $r = .13$ to $r = .28$ for positive relationships at the secondary level (Roorda et al., 2011).

With respect to students’ affect towards school, students in classes with more supportive middle school teachers have more positive attitudes toward school (Roeser, Midgley, & Urdan, 1996; Ryan, Stiller, & Lynch, 1994) and their subject matter (Midgley, Feldlaufer, & Eccles, 1989). Conversely, middle school students who lack a bond with their teacher are more likely to disengage or feel alienated from school (Murdock, 1999). Cornelius-White’s (2007) meta-analysis showed that TSRs were correlated with students’ satisfaction with school ($r = .44$).

Associations between TSRs and students’ behavior include findings that middle school students more willingly pay attention in class when they think their teacher cares more (Wentzel, 1997). On the other hand, adolescents’ who perceived more disinterest and/or criticism from their teachers were more likely to cause discipline problems (Murdock, 1999). Cornelius-White’s (2007) findings show that more positive student perceptions of their TSRs corresponded with increased student participation ($r = .55$) and attendance ($r = .25$), and decreased disruptive behavior ($r = .25$).

Studies of TSRs and student motivation follow similar patterns. Adolescents’ perceptions of teacher support and caring predict student effort as reported by both teachers (Goodenow, 1993; Murdock & Miller, 2003) and students (Sakiz, Pape, & Hoy, 2012; Wentzel, 1997). Meta-analyses (Cornelius-White, 2007; Roorda et al., 2011) show that TSRs are associated with motivation ($r = .32$) and secondary school engagement ($r = .30$ to .45).

Of this array of important outcomes, we chose to focus on students’ classroom grades. Among the associations between TSRs and these outcomes, we felt grades were (arguably) the most consequential for students’ futures – potentially affecting advancement/retention decisions, tracking, graduation, college placement, and additional, important outcomes.

**Scientific Context of the Study**

In striving to contribute to the scientific theories linking similarity and relationships, we structured...
the study to learn whether the causal associations between similarity and relationships found in laboratory studies generalized to real, ongoing relationships. Furthermore, if successful, our intervention would have important applications for classrooms. Specifically, it would offer a tangible example of how similarities might be leveraged to actually improve relationships in the classroom. Simultaneously, we hoped to evaluate the effects of our intervention as rigorously as possible in a naturalistic setting and to err on the side of being conservative in the inferences we made from our data.

We evaluated our intervention using a 2 X 2 design and focusing on a single class period. Through this design, each individual within every teacher-student dyad was randomly assigned to receive feedback (or not) from a "get-to-know-you" survey. Specifically, students were randomly assigned to either learn what they had in common with one of their teachers (i.e., students in the "Student Treatment" group), or not learn about similarities with their teacher (i.e., students in the "Student Control" group). Teachers found out what they had in common with about half of their students in the focal class (i.e., students in the "Teacher Treatment" group) but not with the other half (i.e., students in the "Teacher Control" group). Thus, all randomization occurred at the student level.

In the spirit of recent recommendations (Cumming, 2014; Simmons, Nelson, & Simonsohn, 2011), we identified six "pre-specified hypotheses" prior to analyzing our data. Specifically, we expected that students in the Student Treatment group would (1) perceive themselves as more similar to their teachers and (2) report a more positive TSR as compared to those in the Student Control group. For students in the Teacher Treatment, we expected that, (3) their teacher would perceive these students as more similar, (4) their teacher would rate their TSR more positively, and the students’ (5) mid-quarter grade, and (6) final quarter grade would be higher than students in the Teacher Control group. As described in the Statement of Transparency in our supplemental online materials we also collected additional variables and conducted further analyses that we treat as exploratory.

These main hypotheses reflect an underlying logic that by focusing teachers’ and students’ attention on what they have in common, we will change their perceptions of how similar they are to one another. Congruent with the aforementioned research on similarity, we expect these changed perceptions will lead to more positive relationships between teachers and students. In other words, the core social psychological theory that we are reinforced by our social interactions with similar others (Montoya et al., 2008), will generalize to the educational setting we studied. These more positive relationships, in turn, will cause other downstream benefits for students.

Two explanatory notes about these hypotheses are in order. First, we hypothesized that students’ grades would be affected by the Teacher Treatment (but not the Student Treatment) based on previous correlational work. Brinkworth, McIntyre, Harris, and Gehlbach (manuscript under review) showed that when accounting for both teachers’ and students’ perceptions of their TSR, the teachers’ perceptions (but not students’ perceptions) of the TSR are associated with students’ grades. Second, similar studies of brief interventions that have impacted students’ grades have found that the effect of the intervention was concentrated within a sub-population of students, such as African-American students (Cohen, Garcia, Apfel, & Master, 2006; Walton & Cohen, 2011), Latino students (Sherman et al., 2013), or low self-efficacy students (Hulleman & Harackiewicz, 2009). However, in the absence of information about which sub-groups might react most positively to the intervention, we made no predictions about potential sub-group effects of the intervention.
Methods

Participants
We conducted the study at a large, suburban high school in the southwestern United States. We focused on ninth graders because they were just transitioning to high school and might particularly benefit from connecting with an adult in a school where they did not know any authority figures. The students in our final sample (N = 315) were 60% female, 51% White, 19% Latino, 11% Asian, 6% Black, and 10% reporting multiple categories or “other.” These proportions of different races/ethnicities are similar to the school as a whole (54% White, 20% Latino, 13% Asian, and 10% Black). These students were mostly native English speakers (81%) and came from families where college graduation represented the median educational level of the mothers and fathers (though the range included mothers and fathers who had not attended elementary school to those who completed graduate school).

The teachers in our sample (N = 25) were 52% male, 80% White, and 92% native English speakers. These 25 teachers were part of a faculty of 170, 41 of whom taught 9th graders. The mean age of the teachers was 47.5 years old (sd = 10.42), and the mean years of experience was 18.0 (sd = 9.5). Most teachers (72%) had completed a graduate degree and came from families where 1 year of college represented the median educational level for both their mothers and fathers (though the range extended from those completing fourth grade to those who completed graduate school). Both teachers and students were blind to the purpose of the study.

Measures
Our main measures were borrowed from Gehlbach, Brinkworth, and Harris (2012). Students’ perceptions of their degree of similarity to their teachers were assessed through a six-item scale (α = .88), which included items such as “How similar do you think your personality is compared to your teacher’s?” Students’ perceptions of their TSR were measured with a nine-item scale (α = .90) that asked students to evaluate their overall relationship with their teachers, e.g., “How much do you enjoy learning from <teacher’s name>?” To minimize the burden on teachers, we asked them a single item to assess their perceptions of similarity to each student, “Overall, how similar do you think you and <student’s name> are?” However, they did complete the full parallel nine-item teacher-form of the TSR scale (α = .86 for teachers; see the online appendix for a complete listing of these scales).

We collected mid-quarter and final quarter grades from student records. Because teachers at this high school have autonomy to decide on the most appropriate way to grade students, this measure represents a combination of homework, quizzes, and other assessments depending upon teachers’ individual approaches and the subject matter they teach.

Our exploratory analyses employed additional measures. Teachers rated the amount that they interacted with their students by answering, “Compared to your average student, how much have you interacted with <student’s name> this marking period?” We also collected attendance and tardiness data and (eventually) end-of-semester grades from school records. These measures are listed in the supplementary online materials.

Procedure
The study unfolded over the course of the first marking period at the school. Just prior to the beginning of the school year, the principal helped our research team recruit as many ninth grade teachers as were interested in participating. In turn, during the first week of school these 27 consenting teachers helped us collect consent forms from their students. Throughout the following week of school, these students and teachers visited their computer lab and completed the initial get-to-know-you survey. We mailed our
feedback forms to the school by the middle of the third week of classes. Students (N = 315) and 24 teachers then completed these forms over the course of the next two weeks. An additional teacher submitted her feedback sheet late (though her students completed their sheets on time); this teacher and her students were retained in the sample. Two teachers and their classes never completed the feedback forms, thereby reducing the final sample size to 315 students and 25 teachers. Mid-quarter progress grades were finalized at the end of the fifth week of classes. During the eighth and ninth weeks of classes, students and teachers took the follow-up survey. (Because teachers were allowed to take the survey on their own time, some teachers completed the follow up survey up to one month later). The quarter concluded at the end of the tenth week of classes.

Students and teachers took the 28 item get-to-know-you survey during their first period class. The survey asked teachers and students what they thought the most important quality in a friend was, which class format is best for student learning, what they would do if the principal announced that they had a day off, which foreign languages they spoke, and so on (See Figure 1). From these surveys we composed the feedback sheets that comprised the core of the intervention.

On these feedback sheets, we listed either five things students had in common with their teacher (in the Student Treatment group) or five commonalities the students shared with students at a school in another state (in the Student Control group). Each teacher received five items that they had in common with each student who was among those randomly selected into the Teacher Treatment group (i.e., half of the participating students from the teacher’s first period class). Teachers were informed that in the interest of providing prompt feedback, we could not provide reports on their remaining first-period students (the Teacher Control group). The five similarities were chosen based on an approximate rank ordering of the similarities that had seemed to be most important for generating perceptions of similarity from the pilot test in the previous year (see the Statement of Transparency for more on the pilot

Figure 1: Screen shot of the get-to-know-you survey.
test). Students and teachers responded to a series of brief questions on their feedback sheets such as, "Looking over the five things you have in common, please circle the one that is most surprising to you." Our hope was that by completing these questions on their feedback sheets, students and teachers would more deeply consider and better remember their points of commonality with one another. Current copies of the measures and materials are available from the first author upon request.

Results and Discussion

Pre-specified Hypotheses

As detailed in our “Statement of Transparency” (see the supplemental online materials), we pre-specified six hypotheses (Cumming, 2014). Specifically, we anticipated that (as compared to those in the Student Control group) students in the Student Treatment group (1) would perceive more similarities and (2) a more positive TSR with their teacher. As compared to those in the Teacher Control group, we hypothesized that teachers would perceive students in the Teacher Treatment group as (3) being more similar, and (4) teachers would develop a more positive TSR with these students. Finally, we expected that the students in the Teacher Treatment group would earn (5) higher mid-quarter and (6) higher end-of-quarter grades than their counterparts in the Teacher Control group.

As described in the Statement of Transparency, we expected to test these hypotheses through a combination of multi-level modeling (i.e., hypotheses 3, 5, and 6 when the outcome was a single item) and multi-level structural equation modeling (i.e., hypotheses 1, 2, and 4 when the outcome was a latent variable). However, our statistical consultant advised us that the number of teachers (i.e., level 2 clusters) was inadequate for Mplus to provide trustworthy estimates for the models using latent variables. Our models for latent variables had more parameters to be estimated than clusters, making multilevel SEM impossible. Due to this nested structure of our data, we relied on mean- and variance-adjusted weighted least squares for complex survey data (WLSMV-complex) estimation, using the CLUSTER option in Mplus. WLSMV-complex, which uses a variance correction procedure to account for clustered data, provides corrected standard errors, confidence intervals, and coverage (Asparouhov, 2005). We used full information maximum likelihood (FIML) to address missing data. The maximum proportion of missing data for any variable was .012. However, we used Mplus’ robust standard error approach when our outcomes were latent. To evaluate each hypothesis, we regressed the outcome on the condition as described above. Because random assignment produced equivalent groups between both treatment groups and their respective control groups on key demographic characteristics (specifically gender, race, English language status, and parents’ educational level), no covariates were used in these analyses. Consistent with Cumming’s (2014) recommendation, we evaluated our hypotheses using 95% confidence intervals to emphasize the range of plausible values for the treatment effect rather than relying on p-values. In
addition, we report standardized $\beta$ to provide an estimate of effect size (except for grade-related outcomes where the original 0 to 4.0 scale provides meaningful equivalents of an F through an A). We present descriptive statistics in Table 1.

Our results are congruent with the similarity hypotheses (i.e., hypotheses 1 and 3). Each treatment made students and teachers feel more similar to one another by the end of the marking period ($\beta = 0.33$, SE = 0.12, CI: 0.10, 0.56 for students; and $\beta = 0.33$, SE = 0.11, CI: 0.11, 0.55 for teachers). In other words, we retain the null-hypothesis that the true standardized treatment effect fell within the range from .11 and .55 (and between .10 and .56 for students), while bearing in mind that the most plausible values are those closest to .33.

By contrast, the students perceived their TSRs to be relatively similar regardless of the condition to which they were assigned ($\beta = 0.09$, SE = 0.14, CI: -0.18, 0.36). In other words, we found minimal support for hypothesis 2. Within the Teacher Treatment, teachers perceived a more positive relationship with these students ($\beta = 0.21$, SE = 0.11, CI: 0.00, 0.42). For students in the Teacher Condition, we found no compelling support for an

### Table 1: Descriptive statistics for key variables in the study (unadjusted mean, sd, and Pearson ($r$) correlations).

<table>
<thead>
<tr>
<th>Variable Name</th>
<th>Mean</th>
<th>$sd$</th>
<th>Min.</th>
<th>Max.</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
</tr>
</thead>
<tbody>
<tr>
<td>1) Students' similarity</td>
<td>2.68</td>
<td>0.73</td>
<td>1.00</td>
<td>4.17</td>
<td>--</td>
<td></td>
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<td></td>
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<tr>
<td>2) Teachers' similarity</td>
<td>2.90</td>
<td>0.91</td>
<td>1.00</td>
<td>5.00</td>
<td>.13</td>
<td>--</td>
<td></td>
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<td></td>
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<td></td>
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<tr>
<td>3) Students' TSR</td>
<td>3.68</td>
<td>0.68</td>
<td>1.00</td>
<td>5.00</td>
<td>.69</td>
<td>.18</td>
<td>--</td>
<td></td>
<td></td>
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<tr>
<td>4) Teachers' TSR</td>
<td>3.85</td>
<td>0.55</td>
<td>2.22</td>
<td>5.00</td>
<td>.29</td>
<td>.63</td>
<td>.32</td>
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<tr>
<td>5) Mid-quarter Grade</td>
<td>3.26</td>
<td>0.99</td>
<td>4.00</td>
<td>.34</td>
<td>.23</td>
<td>.35</td>
<td>.41</td>
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<tr>
<td>6) End-of-quarter Grade</td>
<td>3.16</td>
<td>1.10</td>
<td>0.00</td>
<td>4.00</td>
<td>.30</td>
<td>.18</td>
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<td>.43</td>
<td>.76</td>
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<tr>
<td>7) Semester grade</td>
<td>2.79</td>
<td>1.11</td>
<td>0.00</td>
<td>4.00</td>
<td>.24</td>
<td>.31</td>
<td>.28</td>
<td>.47</td>
<td>.67</td>
<td>.79</td>
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<td>8) Tardies</td>
<td>0.26</td>
<td>0.66</td>
<td>0.00</td>
<td>9.00</td>
<td>-.13</td>
<td>-.01</td>
<td>-.08</td>
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<td>-.20</td>
<td>-.22</td>
<td>-.13</td>
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<tr>
<td>9) Absences</td>
<td>1.29</td>
<td>1.61</td>
<td>0.00</td>
<td>5.00</td>
<td>-.15</td>
<td>-.08</td>
<td>-.06</td>
<td>-.16</td>
<td>-.20</td>
<td>-.15</td>
<td>-.10</td>
<td>.15</td>
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</tr>
<tr>
<td>10) Teacher reported interactions</td>
<td>4.74</td>
<td>1.10</td>
<td>2.00</td>
<td>7.00</td>
<td>.21</td>
<td>.37</td>
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<td>.16</td>
<td>.22</td>
<td>.21</td>
<td>-.11</td>
<td>-.10</td>
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</table>

Notes:
1) Ns ranged from 275-362.
2) Correlations are unadjusted for the nesting of students within classrooms.
3) Approximate significance levels are as follows: for $|rs|$ ranging from 0 to .12, $p = ns$; for $|rs|$ ranging from .13 to .16, $p < .05$; for $|rs|$ ranging from .17 to .20, $p < .01$; for $|rs|$ ≥.21 and greater, $p < .001$. 9
effect on mid-quarter grades ($\beta = 0.04$, SE = 0.10, CI: -0.15, 0.23). Although the confidence interval does include 0, our point estimate and the range of plausible responses suggests that students in the Teacher Condition probably earned higher end-of-quarter grades ($\beta = 0.21$, SE = 0.11, CI: 0.00, 0.43). Figures 1-4 in the supplementary online materials show how the unadjusted means are distributed when the Teacher and Student Conditions are separated into their four unique groupings of the 2 X 2 design.

The first pair of findings shows that the intervention successfully enhanced teachers’ and students’ perceptions of similarity. On the one hand, the effects do not seem particularly potent – perhaps reflecting only a mildly-to-moderately strong intervention. On the other hand, students processed their feedback sheets for approximately fifteen minutes before handing them back in, and yet, still perceived themselves as being more similar to their teacher over a month later. Teachers presumably spent even less time on each feedback sheet given that most teachers had several to complete. Thus, while one might argue that the effects of the intervention were weak, this interpretation should be calibrated against the brevity of the intervention and the amount of time that elapsed before the outcomes were collected (Cumming, 2014).

Although the intervention appeared to improve teachers’ perceptions of their relationships with students, we do not find compelling evidence that the intervention improved TSRs from students’ perspectives. To the extent that this result reflects a genuine difference in the effect of the intervention, one plausible explanation is that teachers view part of their role as needing to foster positive relationships with students. Thus, they are motivated to perceive students whom they view as similar in a positive light. By contrast, students may not feel any particular obligation to form a positive relationship with their teachers. Learning that they share common ground with their teacher may not change their perception of their teacher because 9th grade students have no particular motivation to cultivate this social relationship.

Our findings for students’ academic achievement seem paradoxical: the intervention appears to show positive effects at the end of the quarter after finding no effects half-way through the marking period. However, we think this apparent paradox results from a logistical issue rather than a finding of substantive interest. In an unfortunate oversight, we finalized our pre-specified hypotheses prior to reviewing the timing of each key aspect of the study. Although the direction of the estimate for students’ mid-quarter grades is the same as the end-of-quarter grades, we suspect that the intervention occurred too close to teachers’ grade-submission deadline to have a meaningful effect in most classes. In other words, students may not have had a sufficient opportunity to do enough graded work between the time that they (and their teachers) completed their feedback sheets and the date that mid-quarter grades were due. As a result, we do not discuss this outcome further. Students’ performance on their final quarter grades, by contrast, suggests that the intervention probably caused students’ grades to increase. Our point estimate of this increase corresponds to a little less than a fifth of a letter grade.

To better understand our initial pattern of results, we examined whether our intervention might have had differential effects on different sub-populations of students. By fitting a series of multi-level models (for observed outcomes) and models with robust standard errors (for latent outcomes) in MPlus, we conducted a series of exploratory analyses on different student subgroups.

**Exploratory Analyses**

A number of previous studies that employ relatively brief, social psychological interventions (Cohen et
suggest that certain subgroups of students often benefit disproportionately from the interventions. Specifically, we thought that the school might serve some students better than others, or that there might be a dominant culture at the school that was more inclusive of some students than others. After speaking with the principal about this possibility, he suggested that the White and Asian students were typically well-served by the school, while Black and Latino students typically faced more challenging circumstances at home, at school, and throughout their community. Thus, we re-examined our data by analyzing the White and Asian students as a separate group from the remaining “underserved” students. Because these are exploratory analyses, we do not retain the same level of confidence in these findings as our pre-specified hypotheses. However, we argue that these results are likely to be instructive for generating future hypotheses (Cumming, 2014).

When fitting our models, we found little evidence for any effects of the intervention on the White and Asian students. We find no particularly compelling evidence that White and Asian students in the Student Treatment group perceived different levels of similarity with their teachers ($\beta = 0.17$, SE = 0.15, CI: -0.13, 0.46) or felt their relationships to be different ($\beta = -0.12$, SE = 0.17, CI: -0.46, 0.21) as compared to those in the Student Control group. We find a comparable lack of evidence that the intervention affected teachers’ perceptions of their similarity to their White and Asian students ($\beta = 0.11$, SE = 0.16, CI: -0.20, 0.41) and teachers’ perceptions of their relationships with these students ($\beta = 0.00$, SE = 0.15, CI: -0.29, 0.29).

Finally, we find no evidence that the intervention affected White and Asian students’ end-of-quarter grades ($\beta = -0.01$, SE = 0.15, CI: -0.29, 0.27).

For the underserved students, the story differed. Underserved students who received feedback about commonalities with their teachers felt much more similar to their teachers ($\beta = 0.56$, SE = 0.20, CI: 0.18, 0.96) than their counterparts who did not receive this feedback. It was less clear whether these students felt more positive about their relationships with their teachers ($\beta = 0.39$, SE = 0.24, CI: -0.08, 0.86), though the estimated effect size was moderate and in the expected direction. When teachers received feedback about similarities with their underserved students, they perceived greater levels of similarity with those students as compared to their control counterparts ($\beta = 0.56$, SE = 0.24, CI: 0.08, 1.04). Similar to the underserved students, it was unclear whether teachers in the treatment group felt more positive about their TSRs with these students ($\beta = 0.43$, SE = 0.27, CI: -0.11, 0.96).

Finally, we found some evidence that underserved students’ end-of-quarter grades ($\beta = 0.36$, SE = 0.20, CI: -0.04, 0.75) were most likely higher when their teacher received feedback about their commonalities as compared to students in the Teacher Control condition, although the confidence interval does include 0. As depicted in Figure 2, the point estimate for this difference translates into about .4 of a letter grade on a 4.0 scale and corresponds to the difference between a C+/B- versus a B.

Assuming the point estimate approximates the true value of the treatment effect, these effects on grades are substantial. If we compare the White and Asian students with the underserved students in Figure 2, we can estimate the achievement gap between well-served and underserved ninth graders at this school to be approximately .6 of a letter grade. When teachers learned about the similarities that they shared with their underserved students, the achievement gap was reduced by two-thirds to only .2 of a letter grade. This reduction is in line with other relatively brief interventions that have closed the achievement gap. For example, Cohen et al. (2006) report a 40% closure with an even briefer intervention; Walton and Cohen (2011) report a 52% to 79%
Given the potential importance of these differences, we carried out two final sets of analyses. First, in order to see the extent to which these results persisted over time, we obtained students’ grades in their focal class for the full semester. These analyses showed that the effects of the intervention on the underserved students trended in the same direction as the results for students’ end-of-quarter grades ($\beta = 0.33$, SE = 0.22, CI: -0.11, 0.77).
Second, in anticipation of trying to understand more about the effect of the intervention, we tested whether the intervention appeared to affect other variables we had collected. In particular, we examined attendance and tardiness data from school records and how much teachers reported interacting with each student as compared to the average student. The results from these analyses suggest that the intervention did not affect students’ attendance in their focal class (see Figures 4a and 4b in the supplemental online materials). However, the previously noted subgroup differences emerged in how much teachers reported interacting with their students. Specifically, we found no differences by condition in how much teachers interacted with their White and Asian students ($\beta = -0.13$, SE = 0.16, CI: -0.43, 0.17), but they interacted more with those underserved students who were in the Teacher Treatment Condition ($\beta = 0.43$, SE = 0.16, CI: 0.12, 0.74).

Conclusion

Our study builds on the robust social psychological research showing that similarity fosters liking and more positive relationships. By experimentally manipulating teachers’ and students’ perceptions of actual similarities, our study allows for causal inferences to be made about the effects of similarity on real-world, ongoing relationships. Results from our pre-specified hypotheses suggest that the intervention alters students’ and teachers’ perceptions of how much they have in common, benefits TSRs (at least from the teacher’s perspective), and likely bolsters students’ classroom grades.

A primary theoretical contribution of this work is the demonstration that the causal association between similarity and relationship outcomes found in numerous laboratory studies can generalize to real-life relationships. However, the potential of this intervention to generate broad impact in classrooms is every bit as important. If this approach of connecting students and teachers fosters more positive TSRs (even if the effects are primarily teachers’ perceptions of their relationships with certain students), it represents a relatively quick and easy way to improve an important outcome. In addition, if future studies replicate the narrowing of the achievement gap found in this sample, this intervention would be a particularly “scaleable” from a policy perspective.

Like any study, ours includes a number of limitations that warrant readers’ attention. First, the implementation of the various steps of the intervention was imperfect (e.g., a teacher failing to complete the feedback sheets on time, two other teachers responding to the final survey later, etc.). We hope that future studies can remedy these problems and design systems to administer the intervention consistently. However, we also note that implementation of all manner of interventions (new curricula, disciplinary systems, web portals for parents, and so on) in schools tend to be imperfect. The fact that our intervention was largely effective despite the flaws in execution is an important footnote for practitioners.

Second, our analyses (particularly the exploratory analyses) lacked the statistical power we desired. This caused us to shift to a different statistical approach than the one we had originally planned in our statement of transparency. Our statistical consultant also noted that the multi-level model and clustered standard error approaches we employed, may still result in too many Type-I errors when the number of clusters is small, i.e., fewer than 50 (see, for example, Bertrand, Duflo, & Mullainathan, 2004; Donald & Lang, 2007). To address this potential limitation, we employed a wild cluster bootstrap-t (Cameron, Gelbach, & Miller, 2008). As shown in Table 1 in the appendix, our findings using that approach were generally consistent with those we obtained from our multi-level model and robust standard error models. Particularly given the emerging hypothesis that the effectiveness of the intervention may be localized
to underserved students, future replications should try to obtain substantially larger samples with more clusters across a variety of schools to better evaluate this possibility.

Third, our exploratory findings suggested differences between well-served (White and Asian) and underserved (primarily Black and Latino) students. However, this division of students may mask a more accurate understanding of what moderates the effects of the intervention. For example, we lacked a reasonable measure of socio-economic status in our data set. Given the correspondence between race and socio-economic status in this country, we may have actually detected a moderating effect of socio-economic status that our data masked as a race-based effect. Thus, future studies that can collect a wider array of more precise demographic measures would also be particularly beneficial.

Fourth, the underlying logic of our study describes a story of mediation. Specifically, the effect of our similarity intervention on students’ grades may be mediated by teachers’ perceptions of their relationships with students. However, recent work has sharpened our understanding of mediation. Proving mediation is a difficult and ongoing journey rather than a succinct set of equations (Bullock, Green, & Ha, 2010) that establish a particular variable as a mediator. Thus, we can only say that our data largely cohere with this mediation story; we do not (and cannot) establish mediation per se within a single study. In the same way that race may be masking a socio-economic effect that we do not have good enough measures to detect, variables we did not measure may be the fundamental mediators between this intervention and our outcomes. Future research that provides data on other potential mediators (e.g., those not assessed in this study) will also prove tremendously helpful.

Other key future directions emerge out of the results themselves. First, the Teacher Treatment seemed to yield a greater effect on our outcomes than the Student Treatment. When teachers learned what they had in common with their students, they felt they had more in common with those students, perceived better relationships with them, and those students seem to have better grades. Although more speculative, it appears that the Teacher Treatment may primarily affect the underserved students. Thus, one set of future studies might investigate whether the effects of the intervention are really concentrated on teachers and underserved students, or whether this finding varies by context or population. Other studies could investigate whether the intervention might be adapted to improve students’ perceptions of the relationship or to make it effective for all students rather than just a subset of students. Additional research might investigate the role of teachers’ race and/or the congruence between students’ and teachers’ race on the effectiveness of the intervention.

Second, although consequential for students’ futures, grades have limitations as a key outcome variable. Specifically, they leave substantial ambiguity as to why the effects of the intervention occur – a question that will be especially important for future studies to address. One potential explanation is rooted in interactions. Many teachers may see it as a part of their role to connect with students and form a positive working relationship. Knowing what they have in common with their students provides them with a lever through which they can begin developing this relationship. For a group of predominantly white teachers, learning what they have in common with their underserved students may be critically important. Indeed, we find that teachers report interacting with these students more frequently. From this knowledge and the increased interactions, teachers may connect better with students at an interpersonal level and may be better equipped to connect their subject matter to students’ interests. If this scenario transpires, greater learning seems a likely consequence. By contrast, ninth graders (regardless of race) may
have little interest in connecting with their teachers or having any more interactions than necessary. They might be much more focused on connecting with their peers during this developmental stage. As a result, the students in this treatment group may find few effects of the intervention beyond greater perceived similarity with their teacher.

An alternative explanation is rooted in perceptual biases. Perhaps teachers typically perceive their students – particularly their underserved students – in stereotypical fashion. However, when they realize several domains in which they share some common ground with these students, the teachers perceive their relationships with these students in a new way – more like members of their own ingroup (Hewstone, Rubin, & Willis, 2002). A potential consequence is that teachers might assign these students higher grades as a consequence of perceiving them differently.

Our exploratory analyses suggest that the possibility of perceptual biases will also be an important, challenging area of future investigation. On the one hand, we might expect students, who are welcomed into a classroom where the teacher more frequently interacts with them in positive ways, to attend class more regularly and arrive on time more often. While we did not find much evidence congruent with this conjecture, there are many factors that affect a student’s presence in class.

On the other hand, the perceptual bias story may not be completely congruent with the finding that teachers report interacting more frequently with students in the Teacher Treatment Condition than with their control group peers. In other words, if teachers interact with these students more frequently, then the higher grades may partly be a function of learning. Thus, research that can begin to shed light on the mechanisms – be they teacher-student interactions, teacher perceptions, a combination of both, or other factors – through which this intervention affects these important outcomes of TSRs and grades will be especially fruitful.

In closing, this study shows that (perceptions of) real similarities can be influenced by a brief intervention that affects real relationships in a consequential setting like a high school. Our findings suggest that the improvements in TSRs may, in turn, cause downstream benefits for students’ grades. Finally, these results generate strong hypotheses that similar interventions in the future may be effective in helping to close achievement gaps between subgroups of students.

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We dedicate this article to the memory of Maureen Brinkworth (1983 – 2014). She passed away far too young and with far too much of her abundant promise unrealized. She was equal parts intellectual inspiration for this work and logistical wizard who made this study happen. Although she died while the article was under review, we hope she is smiling somewhere to see her work acknowledged.
References:


Brinkworth, M. E., McIntyre, J., Harris, A. D., & Gehlbach, H. (manuscript under review). Understanding teacher-student relationships and student outcomes: The positives and negatives of assessing both perspectives.


References:


Increasingly, scholars have voiced skepticism about the validity of nuanced findings in psychology that may have resulted from practices such as post-hoc data mining (Simmons, Nelson, & Simonsohn, 2011). Several approaches seem promising, although it is clear that more experience and research will need to occur before consensus can be reached on the optimal set of approaches. One particularly promising approach entails the development of registries where scientists would provide a brief description of their intended study before beginning the research, list the variables that they will collect, and, perhaps most importantly, list their hypotheses a priori – what Cumming (2014) would describe as "pre-specified" hypotheses.

Unfortunately, this approach is not possible for the present study. We collected the data for this research before becoming aware of the practice of registering studies (and still remain unaware of websites that facilitate the registration of psychological studies). We also have concerns about how this practice should play out ideally – particularly with regard to field experiments like the present investigation. Registering a study ahead of time should work well if random assignment works, if implementation of the intervention is high in its fidelity, and unforeseen circumstances do not arise. However, in the messiness of the real world, studies are rarely implemented perfectly and predicting all possible contingencies and compensatory steps that may be required seems unrealistic. Finally, it seems reasonable that scholars might want to weight their confidence in different hypotheses along a sliding scale. For example, a pre-specified hypothesis that a manipulation check will work seems much safer (and less interesting) than predicting that a particular treatment will be simultaneously moderated by race and mediated by a personality trait.

There is clear value to these new steps for the integrity of psychological science, yet there are challenges in figuring out how to adjust to these new recommendations. In the hopes of finding a middle ground, we are writing this statement of transparency on March 13, 2014 – the day before we begin any data analysis. We hope this step maximizes the integrity of our study. This statement will not be edited in any way once data analysis commences. We hope that this approach might have some strengths that other scholars may benefit from; undoubtedly, this approach will have weaknesses that we hope others help us to learn from. In this statement, we describe the following features of the study:

A background section which overviews the preliminary pilots that informed the current study.

A list of the variables collected which denotes the variables we intend to use in the analysis for the current study (in contrast to those we are interested in for other studies).

A list of hypotheses which denotes a set of clear "pre-specified" hypotheses. All other analyses conducted in our final manuscript should be viewed as exploratory, hypothesis-generating findings.

Key details of the analytic choices that we are making ahead of time.

Background to the present study

This data collection represents the third time we have implemented an intervention similar to the present one at the school in question. The basic procedure was always the same as what is described in the methods section: We give teachers and students a "get-to-know-you" survey, randomly assign them to get feedback (or not) about what they have in common with the other party, ask them to reflect on that feedback, follow up with a longer survey shortly before the end of the marking period, and collect grades and student record data after the quarter ends. We first ran this field experiment during the 2011-12 school year with a convenience sample of 10th grade classes. Overly confident from laboratory studies suggesting that even trivial commonalities could change individuals’ affect and behavior for each other (Ames, 2004; Burger, Messian, Patel, del Prado, & Anderson, 2004), we were relatively cavalier about what types of similarities we asked teachers and students about (e.g., favorite pizza toppings, preference for crunchy versus smooth peanut butter, etc.). We found no clear effects from this study. That spring we conducted an open-ended survey with 9th graders to learn what types of commonalities they might value having with their teachers. These data allowed us to substantially revise the "get-to-know-you" survey.

For the 2012-13 school year we conducted the study again with several important changes. First, we used the revised get to know you survey. Second, our Year 1 analyses yielded a suggestive finding that perhaps the student control group (who received feedback on what they had in common with other students in their grade) might have actually benefitted from a heightened sense of belonging at their school. So for Year 2, we changed the control group’s feedback to learning what they had in common with
students from a school in a different state. Third, we added an additional treatment group that would learn what they had in common with students from their own grade (i.e., to see whether the control condition from the previous year really was having a positive effect). Fourth, we ran the intervention with 9th graders, thinking that they might benefit the most given the often challenging transition to high school.

These data showed promising, though mildly vexing results. Specifically, the similarity manipulation seemed to work: both teachers and students in the treatment conditions perceived greater similarity to the other party. The intervention produced a clear boost in the positivity of the teacher-student relationship from the teacher’s perspective. However, the effect of the intervention on the students’ perceptions of their relationship with their teacher was much less clear. The intervention manifested a significant, positive effect on students’ mid-quarter grades. This trend dropped to non-significance by the end of the quarter, but the effects were still in the same direction. We found no suggestions that the sense of belonging intervention had any effect.

Based on these encouraging, but mixed findings from our similarity intervention and our modest sample size (N = 101, spread across four conditions), we decided to try to replicate the intervention in the current 2013-14 school year. We dropped the sense of belonging intervention conditions to maximize our power for the similarity intervention. However, no substantive changes were made to the similarity intervention itself from 2012-13 to 2013-14.

List of variables in the present study

<table>
<thead>
<tr>
<th>Student self-report measures:</th>
<th>Number of items</th>
</tr>
</thead>
<tbody>
<tr>
<td>Teacher-student relationship positivity</td>
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<tr>
<td>Teacher-student relationship negativity</td>
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<td>Similarity</td>
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<tr>
<td>Social perspective taking confidence</td>
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<tr>
<td>Social perspective taking effort</td>
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<td>Self-other description</td>
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<tr>
<td>Sense of belonging</td>
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<tr>
<td>Social perspective taking accuracy</td>
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<tr>
<td>Interest in subject matter</td>
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</tr>
<tr>
<td>Effort in class</td>
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<tr>
<td>Self-efficacy in class</td>
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<td>Memory check</td>
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<td>Donation</td>
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<td>Suspicion check</td>
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<td>Demographics</td>
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<table>
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<td>Effort</td>
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<td>Interest</td>
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<td>Demographics</td>
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<table>
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<tr>
<th>School record data</th>
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<td>Mid-term grade (in focal class and overall)</td>
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<tr>
<td>End-of-semester period grade (in focal class and overall)</td>
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<tr>
<td>Tardies (focal class and overall)</td>
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<td>Attendance (focal class and overall)</td>
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<td>Discipline</td>
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Note: Bolded variables indicate which variables that will be included in the analyses for this study.
Hypotheses

We will test the following prescriptive hypotheses:

Similarity:

- Students who receive feedback that they have commonalities with their teacher will report a greater sense of similarity to their teacher on the student-reported 6-item similarity scale.

- Teachers who receive feedback that they have commonalities with a particular student will report a greater sense of similarity to that student on the teacher-reported similarity item.

Teacher-student relationship:

- Students who receive feedback that they have commonalities with their teacher will report perceiving a more positive teacher-student relationship (i.e., the 9-item student-report measure).

- Teachers who receive feedback that they have commonalities with their student will report perceiving a more positive teacher-student relationship (i.e., the 9-item teacher-report measure).

Grades:

- Students of teachers who receive feedback that they have commonalities with their student will earn higher mid-term grades in their focal class.

- Students of teachers who receive feedback that they have commonalities with their student will earn higher final marking period grades in their focal class.

We have arrayed these prescriptive hypotheses such that we are most confident about the hypotheses listed towards the top (largely based on our prior pilot data and our previous correlational studies suggesting that the association between teacher-student relationships and students’ grades is due to the teacher’s perception of the relationship). We hope that this sliding scale helps readers calibrate their confidence in our findings accordingly. All other analyses that we present in the final manuscript should be viewed as exploratory, hypothesis-generating findings.

Analytic details

Data cleaning will be used to cull any students who changed classes during the first quarter of the school year (such that their focal teacher changed). In addition, we will remove teacher and student responses that show evidence of straight-line responding (Barge & Gehlbach, 2012). Specifically, sets of ten or more sequential responses on the same response anchor within the same section of the survey will be removed. Depending upon where in the survey this occurs (e.g., during the similarity and teacher-student relationship items), it may, for all practical purposes, have the effect of removing students from subsequent analyses.

With those students removed, we will examine the differences between four key conditions of interest in a 2 x 2 design: A control group (who learned that they had commonalities with students from another state), students who found out that they had commonalities with their teacher, students whose teacher learned that s/he had commonalities with the student, and student-teacher dyads who both knew that they had commonalities with each other. Then we will use multi-level structural equation modeling to test those hypotheses where latent variables are used (i.e., the students’ report of similarity and both teacher-student relationship outcomes).

Structural equation modeling will be used for the remaining tests (where the outcomes are not latent). The CLUSTER IS command will be used in Mplus account for students being nested within teacher (rather than within class). No covariates will be used in these primary analyses unless random assignment fails. The treatment predictor for each equation will be dichotomous – in other words, we are only hypothesizing main effects from the teacher receiving feedback or the student receiving feedback. However, as a complement to these results, figures will present mean-levels of perceived similarity, teacher-student relationship, students’ grades, and attendance/tardy outcomes unadjusted for nesting and broken out into the four different conditions described above. In line with Cumming’s (2014) recommendation, we will evaluate our hypotheses by presenting and discussing 95% confidence intervals and effect sizes (not by reporting p-values). Our basic model for the 6 hypotheses articulated above is:
The model above will be used to test the first Similarity hypothesis and both Teacher-student relationship hypotheses.

The model above will be used to test the second Similarity hypothesis and both Grade hypotheses.
Table 1: Results from re-analyses using a wild cluster bootstrap-t.

<table>
<thead>
<tr>
<th></th>
<th>Full sample β [95% CI]</th>
<th>Well-served students β [95% CI]</th>
<th>Under-served students β [95% CI]</th>
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<tbody>
<tr>
<td>Students’ Perceived Similarity for Student Treatment</td>
<td>0.18 [0.05, 0.33]</td>
<td>0.11 [-0.07, 0.29]</td>
<td>0.32 [0.08, 0.56]</td>
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<tr>
<td>Teachers’ Perceived Similarity for Students in Teacher Treatment</td>
<td>0.28 [0.07, 0.49]</td>
<td>0.05 [-0.21, 0.31]</td>
<td>0.50 [-0.04, 1.05]</td>
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<tr>
<td>Students’ Perceived Teacher-student Relationship for Student Treatment</td>
<td>0.05 [-0.13, 0.23]</td>
<td>-0.08 [-0.36, 0.15]</td>
<td>0.20 [-0.12, 0.47]</td>
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<td>Teachers’ Perceived Teacher-student Relationship for Students in Teacher Treatment</td>
<td>0.15 [-0.01, 0.31]</td>
<td>0.00 [-0.23, 0.23]</td>
<td>0.29 [-0.07, 0.67]</td>
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<tr>
<td>End-of-quarter Grades for Students’ in Teacher Treatment</td>
<td>0.21 [-0.01, 0.41]</td>
<td>-0.01 [-0.30, 0.28]</td>
<td>0.46 [-0.17, 1.04]</td>
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Notes: The standardized bootstrap relies on the bootstrap-implied distribution of a t-statistic rather than a beta estimate (Cameron, Gelbach, & Miller, 2008), and so we do not report the standard errors of the t-statistic; the bootstrap makes no assumptions about the normality or even symmetry of the sampling distribution, and so standard errors cannot be used to calculate confidence intervals or conduct hypothesis tests.

Table 2: Raw (unadjusted for nesting) means of key variables by Student and Teacher Conditions: Mean, (Standard Errors), and [95% Confidence Intervals].
### Results Appendix

<table>
<thead>
<tr>
<th></th>
<th>Control (n = 60)</th>
<th>Student Treatment Only (n = 63)</th>
<th>Teacher Treatment Only (n = 69)</th>
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<td>2.56 (0.09)</td>
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<td>3.36 (0.06)</td>
<td>3.33 (0.11)</td>
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<td>[3.71, 3.93]</td>
<td>[3.77, 4.02]</td>
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<td><strong>Students' Mid-quarter Grades</strong></td>
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<td>3.23 (0.11)</td>
<td>3.22 (0.12)</td>
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<td>[2.99, 3.45]</td>
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<td>3.00 (0.13)</td>
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<td>[2.74, 3.24]</td>
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<td>0.36 (0.08)</td>
<td>0.21 (0.09)</td>
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<td>1.18 (0.19)</td>
<td>1.41 (0.19)</td>
<td>1.24 (0.20)</td>
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<td>[1.03, 1.79]</td>
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Notes: To facilitate the review process, we are presenting these more comprehensive tables in place of the series of figures described in the Statement of Transparency. We are happy to include either for the final publication.

Figures 1a and 1b: Students' and teachers' perceptions of similarity to one another by condition (Mean and 95% CI).
Figures 1a and 1b: Students’ and teachers’ perceptions of similarity to one another by condition (Mean and 95% CI).

Figures 2a and 2b: Students’ and teachers’ perceptions of their teacher-student relationship by condition (Mean and 95% CI).
Figures 3a and 3b: Students’ mid-quarter and end-of-quarter grades in their focal class by condition (Mean and 95% CI).

Figures 4a and 4b: Students’ tardiness and attendance by condition (Mean and 95% CI).