Abstract – Recommendations for the aggressive recruitment of minority teachers are based on hypothesized role-model effects for minority students as well as evidence of racial biases among non-minority teachers. However, prior empirical studies have found little or no association between exposure to an own-race teacher and student achievement. This paper presents new evidence on this question by examining the test score data from Tennessee’s Project STAR class-size experiment, which randomly matched students and teachers within participating schools. Specification checks confirm that the racial pairings of students and teachers in this experiment were unrelated to other student traits. Models of student achievement indicate that assignment to an own-race teacher significantly increased the math and reading achievement of both black and white students.

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I. INTRODUCTION

The economic literature on the policy determinants of student achievement has focused largely on the possible benefits of new educational resources that reduce class size and improve teacher salaries and training. However, the literature from other social sciences provides provocative evidence that several other contextual factors might also substantively influence the relationship between teachers, students and student achievement. In particular, the conventional wisdom among educators is that minority students are more likely to excel educationally when matched with teachers who share their race or ethnicity. The frequent calls for aggressive recruitment of under-represented minority teachers are typically motivated by the specific claim that such teachers are better-equipped to deal with the special needs of at-risk minority students and that they provide more effective role models (e.g., U.S. Department of Education, 1997; Graham, 1987; Ladson-Billings, 1994; NCTAF, 1996). There is also corresponding evidence that the racial pairings of teachers and students influence how teachers allocate their time in the classroom as well as their expectations and evaluations of students (e.g., Ferguson, 1998; Casteel, 1998; Zimmerman et al., 1995; Ehrenberg, Goldhaber and Brewer, 1995). Yet, relatively few studies have attempted to identify the relationship between exposure to own-race teachers and subsequent levels of student achievement. And what evidence is available suggests that there is actually little association between student achievement and the racial match between teachers and students (Ehrenberg, Goldhaber and Brewer, 1995; Ehrenberg and Brewer, 1995).

However, the appropriate specification for econometric models of student achievement is a controversial issue. For example, the contentious literature over whether "money matters" in models of student achievement has in part focused on specification issues like functional form and the role of omitted, endogenous or poorly measured variables (e.g., Burtless, 1996). In a recent contribution to this literature, Krueger (1999) examined data from the Project STAR experiment in order to address some of these specification issues as well as the relationship between class size and test scores. Tennessee's Project STAR (Student Teacher Achievement
Ratio) was a large-scale randomized experiment on the achievement benefits of small class sizes. It began in the 1985-86 school year with a group of over 6,000 students from 79 participating schools. The experiment continued through the third grade and ultimately included over 11,000 students. A key feature of the experimental design was the random assignment of both students and teachers to small classes, regular-sized classes and regular-sized classes with teacher aides within each school. Project STAR was not designed to evaluate the relationship between own-race teachers and student achievement. Nonetheless, this experiment provides a novel and potentially compelling opportunity to do so since the putatively random pairings of students and teachers should circumvent the non-random and possibly confounding assignments inherent in conventional data on student achievement.

This study presents such test-score evaluations by relying on the recently released Project STAR Public Access Data. I begin by briefly summarizing and discussing the prior evidence on the racial dynamics between students and teachers (Section II). I then discuss the Project STAR experiment and data in more detail (Section III). Next, I present some initial evaluation results and evidence on whether the racial pairing of students and teachers actually satisfies the supposedly randomized experimental design (Section IV). This is a relevant concern since attrition from the experiment and classroom reassignments ("treatment crossover") could have compromised the integrity of the experimental design (Krueger, 1999; Hanushek, 1999). The empirical results presented here suggest that these issues are not problematic in this context. In particular, auxiliary regressions indicate that the within-school variation in exposure to an own-race teacher was uncorrelated with other important, student traits (e.g., small class assignment,

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1 Not surprisingly with an experiment of this scale and scope, there are a number of important issues with regard to its conduct that could threaten the resulting inferences (most notably, potentially non-random attrition and classroom reassignment). Krueger (1999) addressed these concerns in the context of the class-size results and found that the conventional findings linking small classes with higher levels of student achievement were robust. This study presents similar evidence on whether these violations may be confounding in this context.

2 Furthermore, because these students are “treated” and observed in their early stages of their formal schooling, the possible effects of own-race teachers may be more easily detectable in these data. Prior
age, free-lunch status). I then present a broader set of empirical evidence on the achievement effects of exposure to an own-race teacher (Section V). In brief, these results indicate that assignment to an own-race teacher was associated with substantive gains in achievement for both black and white students. More specifically, these results suggest that a year with an own-race teacher increased math and reading scores by roughly two to four percentile points. Notably, the estimated achievement gains associated with an own-race teacher exist for nearly all groups of students defined by race and gender. However, these effects also appear to vary in intriguing ways with respect to other student, teacher and classroom characteristics (e.g., free-lunch status, teacher experience and class size) and to be additive over time.

I also discuss and examine a fundamental ambiguity regarding the strong associations between student achievement and assignment to an own-race teacher. Specifically, a critically important concern is whether inferences about the effects associated with a teacher’s race are biased by the unobserved dimensions of teacher quality. The randomized pairing of students and teachers only assured that the students’ unobserved propensity for achievement is uncorrelated with their teacher’s race. However, the variation in unobserved teacher quality could still impart biases to the extent that it varies systematically with a teacher’s race. While it is difficult to address this concern definitively, I examine its empirical relevance in several ways, including

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3 Furthermore, concerns about the possible biases due to reassignment and attrition are also addressed here in two other ways. One is by evaluating test score equations that include imputed data for students who left the experiment (Krueger 1999). The other is by generating 2SLS estimates of the effect of own-race teachers where the instrumental variable is a measure of the teacher race a student would have had in the absence of treatment crossover (i.e., the “intent to treat”). The uniformity of the results based on these models suggests that experimental violations are not confounding in this context.

4 It is difficult to quantify the benefits of these test score gains. However, a comparison with other estimated effects suggest these effects are sizable. More specifically, these estimated effects are often comparable to those associated with a small-class assignment (around 4 percentile points) and are relatively large in comparison to the observed black-white test gap (around 6 to 9 percentile points) as well as the test differences between students who do and do not receive free lunches (around 11 percentile points).
evaluating specifications that introduce classroom fixed effects. The results suggest that unobserved teacher quality does not impart confounding biases, particularly with respect to mathematics achievement and white students. In Section VI, I conclude by summarizing this study’s results and speculating briefly about their implications for policy and further research.

II. TEACHERS AND RACE

Three concerns have dominated discussions of educational policy regarding teachers in recent years. First, a shortage of teachers has been observed since the 1980’s. The dearth of teachers has been exacerbated in recent years by the retirement of older “baby-boomer” teachers while the “echo” of the baby boom has simultaneously put upward pressure on school enrollments (U.S. Department of Education, 1997). Second, there is evidence that the ability of those choosing to become teachers has been in decline (e.g., Murnane et al., 1991; Corcoran, Evans and Schwab, 2002). The third, frequently cited concern is a decline in the proportion of teachers who are minorities. Minorities have been historically underrepresented among teachers but, with the pattern of projected retirements and the expected relative growth of minority enrollments, this ostensibly problematic situation is expected to worsen. These stylized facts regarding teachers have motivated recommendations for renewed efforts at recruiting and retaining teachers, particularly those who are racial and ethnic minorities (NCTAF, 1996; U.S. Department of Education, 1997; Graham, 1987; Ladson-Billings, 1994). A maintained assumption underlying these recommendations is that minority teachers would be particularly adept at educating the growing population of minority students. Given that minority students are

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5 I also present qualified, indirect evidence on this issue by discussing the racial variation in teacher observables associated with student achievement and by examining the estimated effects of own-race teachers across schools that may have different abilities to recruit and retain high-quality teachers.
more likely to be at-risk for academic failure, it is perhaps not surprising that less attention has been paid to whether minority teachers might be less effective teachers of non-minority students.\(^6\)

In a recent survey article, Ferguson (1998) concludes that the racial dynamics between students and teachers do appear to influence educational achievement. However, he also notes that the magnitudes of these effects are uncertain and that the possible structural mechanisms are complicated and sometimes based on thin evidence. The prior literature offers at least two general explanations for why the racial pairing of students and teachers might exert an important influence on student achievement. These explanations are not mutually exclusive. One class of explanations involves what could be called "passive" teacher effects. These effects are simply triggered by a teacher's racial presence and not by explicit teacher behaviors. For example, one frequently cited reason for the relevance of a teacher's race is that, by its mere presence, a teacher's racial identity generates a sort of role-model effect that engages student effort, confidence and enthusiasm (e.g., King, 1993; Clewell and Villegas, 1998). For underprivileged black students, the presence of a black teacher may encourage them to update their prior beliefs about their educational possibilities. Similarly, students may feel more comfortable and focused in the presence of an own-race teacher regardless of the teacher’s actual behavior. While the existence of such role-model effects is frequently assumed in commentaries on educational policy, there is actually little in the nature of direct empirical support (Cizek, 1995).

Another possibly relevant sort of passive teacher effect is "stereotype threat" (Steele, 1997). This hypothesized mechanism is based on the assumption that academic identification (i.e., seeing self-worth in academic achievement) is critically important for sustaining educational development. Stereotype threat refers to the possibility that, in situations where students perceive stereotypes might attach (e.g., black students with white teachers), they experience an apprehension that retards their academic identification and subsequent achievement.

\(^6\) However, since this study presents models of achievement among white and black students separately, it provides evidence on this issue.
Experimental evidence based on students at a selective university appears to confirm the existence of race-based stereotype threat. For example, black students underperformed relative to white students on items from the verbal Graduate Records Examination (GRE) when told beforehand that the test was diagnostic of ability (Steele, 1997). Furthermore, test-score comparisons indicated that stereotype threat was triggered for the black students merely by introducing a pre-test demographic questionnaire that recorded race (Steele, 1997). However, it is not yet clear whether this intriguing phenomenon exists among students in elementary and secondary settings (Ferguson, 1998).

A second class of explanations for the educational benefits of own-race teachers points to "active" teacher effects: race-specific patterns of behavior among teachers. In particular, it may be that in allocating class time, in interacting with students and in designing class materials, teachers are more oriented towards students who share their racial or ethnic background. The limited amount of evidence from experimental studies suggests that this does occur (Ferguson, 1998). For example, one study placed white teachers in a teaching environment where they could not observe the student directly and found that they provided less coaching and briefer, less positive feedback when told beforehand that the student was black (Taylor, 1979). Similarly, studies based on observations from actual classrooms often find that black students with white teachers receive less attention, are praised less and scolded more than their white counterparts.7 Unfortunately, the reduced-form test score models presented here cannot meaningfully distinguish between the passive and active hypotheses for why own-race teachers might be educationally relevant. In the concluding section, I argue that our limited understanding of why race might matter in classrooms implies an important caveat to related policy recommendations.

7 See Casteel (1998) for recent evidence on biases in teacher behavior as well as a brief overview of this literature. Ferguson (1998) is careful to point out that the non-experimental evidence suggests the absence of unconditional race neutrality but that it is difficult to assess teacher biases conditional on student behaviors.
While there is a seeming consensus on the importance of race in student-teacher interactions, there is surprisingly little clear evidence on the presumed implications for student achievement. For example, in a recent study, Hess and Leal (1997) found that the share of minority faculty in urban school districts is positively correlated with college matriculation rates. But Hess and Leal (1997) correctly noted that partial correlations linking own-race teachers with improved student outcomes may be very misleading. More specifically, they suggested that a high proportion of minority faculty can proxy for important but unobserved district-specific determinants of student achievement. Similarly, Ehrenberg and Brewer (1995), in a study examining the classic "Coleman Data" from the 1960's, demonstrated that black teachers are associated with improved test scores gains among black students. However, they also recognized the ambiguity of these partial correlations and actually found that these effects are not robust in models that correct for the simultaneous determination of teacher characteristics.\(^8\) Ehrenberg, Goldhaber and Brewer (1995) reconsidered these questions using more recent data from the National Education Longitudinal Study of 1988 (NELS-88). As in prior educational studies, they found evidence that subjective teacher evaluations of students are often higher when student and teacher race coincide.\(^9\) However, they found almost no evidence that the racial pairings of students and teachers influenced the test score gains among NELS-88 respondents. Since the public-use NELS-88 data do not include geographic identifiers, they did not attempt to address the endogeneity of teacher characteristics. It is important to note that the lack of a partial correlation between racial pairing and student achievement, such as that reported by Ehrenberg, Goldhaber and Brewer (1995), could also reflect a negative bias imparted by omitted or endogenous regressors. For example, if minority faculty sought out or were more likely to be

\(^8\) But they also find that the evidence of lower gain scores among white students with black teachers is more robust. However, in general, the quality of their identification strategy for endogenous teacher characteristics may be suspect since it relies on variables that could presumably influence student achievement (e.g., family traits, county or SMSA variables and starting teacher salary).
assigned to at-risk minority students, naïve estimates of their impact on student outcomes would understate the true effects. In the absence of compelling instrumental variables, the uncertain biases inherent in inferences based on observational data represent a seemingly intractable problem for evaluating the educational impact of own-race teachers. For this reason, the putatively random pairings of Project STAR students and teachers provide a unique opportunity to identify the possible links between student and teacher race and student outcomes.

III. TENNESSEE’S PROJECT STAR

In the spring of 1985, the Tennessee Legislature authorized $3 million for the first year of a four-year study of class size that began with kindergarten students that fall (Word et al., 1990). In the first year of the study, 79 schools (and over 6,000 students) participated. Over the four-year study, roughly 11,600 students participated with about 2200, 1600 and 1200 entering in the first, second and third grades respectively (Krueger, 1999). The participating schools were drawn from around the state and, by legislative mandate, included inner-city and suburban schools from larger metropolitan areas (e.g., Knoxville, Nashville, Memphis and Chattanooga) as well as rural schools and urban schools from smaller towns. Recognizing that schools around the state differed in substantive ways that are inherently difficult to quantify, a within-school experimental design was chosen. This implied that smaller schools were necessarily excluded. Participating schools had to have enough students in a given grade so that three class types, a small class of 15 students and two regular-sized classes of 22 (one with a teacher’s aide), could be formed (Mosteller, 1995). Students and teachers within participating schools were randomly matched to three class types. It was originally intended that, once assigned, a student would keep their class type through

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9 Actually, they exhaustively study the influence of interactions between race, gender and ethnicity. However, the focus of this study is race alone. Almost all Project STAR students and teachers are black or non-Hispanic whites and almost all teachers are female.

10 This is similar to the potential bias in conventional class-size studies: if at-risk students are more likely to be assigned to small classes, naïve evaluations can understate the impact of the smaller classes. Ehrenberg
third grade so that cumulative effects could be identified. However, over the course of the four-
year study, this pure experimental design was potentially compromised by both class-type
reassignment and student attrition. The next section discusses these issues in more detail and
presents evidence on the extent to which they might confound these evaluations. For more
detailed discussions of the Project STAR experiment in general, see Word et al. (1990), Mosteller

The empirical results presented here are based on the Project STAR Public Access Data.
Given the very limited number of Hispanic, Asian and American Indian Project STAR
participants, these data were edited to include only those observations from black and white non-
Hispanic students with black and white non-Hispanic teachers. The implied reductions in sample
size were quite modest. For example, among the 6,325 kindergarten students this eliminated only
95 observations. A small number of observations with missing data on key observed
c Characteristics (e.g., age, gender, free-lunch status) were also deleted (e.g., five of the
kindergarten students). Observations were also omitted simply because test score data were
unavailable (largely due to student absenteeism). For example, test score data were available for
only about 5,900 kindergarten students. The test scores available in these data are the scaled
scores from the Stanford Achievement Tests (SAT) in math and reading, which were given in
later March and early April of each study year. As in Krueger (1999), the test outcomes modeled
here are the percentile ranks based on these scores. More specifically, since the two tests differed
across grades, percentile ranks specific to each grade were computed for each subject test.
Pooling the individual observations over the four years leads to 23,883 observations on the math
test and 23,544 on the reading test (Table 1). Aggregating the data in this way is useful since it

et al. (1995) also note that the use of gains in student test scores (as opposed to test score levels) may
exacerbate measurement error and lead to attenuated estimates.

11 The number of first-grade entrants was fairly high since kindergarten was not required.
12 Only 30 kindergarten students were neither black nor white, 3 more had missing race/ethnicity data. All
of kindergarten teachers for whom data were available were identified as white or black. However, teacher
race was missing for 62 kindergarten students.
increases statistical precision and allows us to consider the effects of cumulative exposure to an own-race teacher.

Interestingly, the means in Table 1 uniformly indicate that test scores among both black and white students were higher for those assigned to an own-race teacher. However, those simple differences should be considered suspect since they rely in part on unadjusted comparisons across schools. The key observed student characteristics available in these data include a school identifier, class type assignment, student race, gender and age (here represented by a binary indicator for a birth year prior to 1980) as well as an indicator for whether the student received free lunches in their entry year (Table 1). Roughly two-thirds of the students in the sample were white. It should be noted that the free-lunch variable is based on receipt, not eligibility, but is still particularly useful since it provides the only available information on the students' socioeconomic background.

The available teacher information includes race, years of experience, education and merit pay status. Notably, teacher gender is not included on the public-use data. In all likelihood, this was intended to preserve confidentiality since almost none of the teachers were male. This does limit the generalizability of this study somewhat since gender may generate some heterogeneity in the race-based interactions between students and teachers. Each student's exposure to an own-race teacher is represented in two ways in this study. The main approach is simply to identify whether each student had an own-race teacher in their current academic year. However, some models will address the cumulative effects of student exposure to an own-race teacher through the use of an unrestrictive set of binary indicators for one to four years of exposure (with no exposure as the reference). On average, 94 percent of white students had an own-race teacher in their current academic year while only 45 percent of black students did (Table 1). However, these unconditional means mask an interesting pattern of heterogeneity across schools. For example, in

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13 Krueger (1999, appendix table) reports that none of the kindergarten or first-grade teachers are male while one and three percent of the second-grade and third-grade teachers are.
the 16 participating city schools, roughly 97 percent of the students and half of the teachers were black. While, in the 7 urban and 38 rural schools, 93 percent of the students and 97 percent of the teachers were white. The 18 suburban schools were more integrated with 38 percent of the students and 26 percent of the teachers being black.\(^{14}\)

An important specification issue in this study involves the potential bias in the estimated effects associated with a teacher’s race due to the unobserved dimensions of teacher quality.\(^{15}\) Therefore, the available data on observed teacher traits typically associated with quality (education, experience, merit pay status) may facilitate important robustness checks. Each teacher's education is measured here by a binary indicator for having some type of graduate degree (e.g., M.S., M.A., Ed.S. or Ph.D.) with the reference category including those who only have a bachelor's degree. Roughly 38 percent of students were assigned to a teacher with a graduate degree. Teaching experience is measured in years (a quadratic term is also included in regression models). On average, students were assigned to teachers with nearly 12 years of experience. Another binary indicator identifies teachers recognized by Tennessee's contemporaneous merit pay plan, the Career Ladder Evaluation System (Dee and Keys, 2003). Roughly 90 percent of students were assigned to teachers who participated in Tennessee's career ladder. Whether such merit pay programs can systematically identify and reward good teachers is actually a controversial issue (Murnane and Cohen, 1986; Ballou, 2001). However, Tennessee's program was considered a relatively sophisticated one since it blended pecuniary and professional rewards and relied on several teaching evaluation instruments (including classroom observation).\(^{16}\) Nonetheless, its usefulness as a proxy for teacher quality appears to be somewhat limited. Dee and Keys (2003) find that Tennessee's merit pay program had only mixed success in rewarding teachers who raised the average level of classroom achievement.

\(^{14}\) I discuss and examine how the effects associated with own-race teachers varies across more and less segregated schools.

\(^{15}\) The empirical relevance of this concern is discussed in more detail in Section V.
IV. THREATS TO VALIDITY

Project STAR arguably provides a unique and compelling opportunity for making reliable inferences about the determinants of student achievement. However, a number of factors could also attenuate the generalizability of inferences based on these experimental data. For example, the estimated effects associated with class-size assignments could be biased by the ways students and teachers adjust their behavior in response to the knowledge that they have been assigned to a particular experimental group (i.e., “Hawthorne” and “John Henry” effects). The “external validity” of the experiment may also be limited because the study had almost no male teachers, excluded smaller schools and included only young students in Tennessee. A caveat about the external validity of focusing on Tennessee students may be unusually relevant in this study than in a class-size study since the links between own-race teachers and student achievement could reflect Tennessee-specific cultural factors that to some degree will not generalize to other schools. But, perhaps even more important are the possible threats to the "internal validity" of causal inferences based on the experiment. As noted earlier, like any social experiment, Project STAR had some notable and potentially problematic complications in its execution. For example, because of parental complaints, students in the regular-sized classes were randomly reassigned to regular-sized classes with and without teacher aides at the beginning of first grade. Assignments to small classes were generally unaffected by this re-randomization. However, roughly 10 percent of students were also moved between small and regular class assignments, largely because of complaints or behavioral problems. Furthermore, attrition from the study was fairly high. This attrition could reflect conventional family mobility, grade repeating and advancing as well as possibly non-random responses to class assignments.

16 Brandt (1995) praised Tennessee's approach as "perhaps the country's most comprehensive experiment in summative evaluation."
17 Initial evaluations based on the kindergarten students indicated that the addition of teacher aides had no impact on student achievement.
18 Krueger (1999) concludes that the class-size effects in Project STAR are not compromised by these problems. The effects of sample attrition were addressed through the use of imputed test scores. The
Fortunately, the threats to the validity of inferences about class-size effects should not be as relevant for inferences about the effects associated with a teacher’s race. For example, non-random attrition or class reassignment would seem less likely to be problematic in this context than in a high-profile study of class-size effects. When parents chose a school, they presumably had fairly sound prior expectations regarding the conditional probability that their child would be assigned an own-race teacher. And, since teachers and students would be reassigned in the next academic year, the racial pairings in a given year do not provide a particularly strong incentive for attrition. In contrast, a student’s assignment to small or large classes was intended to persist through the third grade. Furthermore, the “reactive” effects associated with the participants’ knowledge that they have a particular experimental assignment should not be relevant since the racial dynamics between students and teachers were not a focus of the study. Nonetheless, it is important to consider whether Project STAR’s deviations from an ideal experimental design might confound this study’s inferences, which link assignments to an own-race teacher with higher test scores. Notably, the likely direction of biases introduced by non-random attrition or class reassignment is uncertain a priori. They could plausibly result in estimates that overstate or understate the true effect of an own-race teacher. For example, if the decision to move a child to an own-race teacher partly reflects unobserved family or parental priors that harm student achievement (e.g., ignorance, intolerance, poor socialization), the results presented here will understate the academic benefits of an own-race teacher. In contrast, to the extent that parents who tend to provide strong academic support for their child are also more likely to seek out an own-race teacher, the results presented here will overstate the benefits of such teachers. Clearly, the latter of these possibilities is the most relevant given the pattern of results reported here.

These concerns are evaluated here in several ways. One is by presenting results based only on the kindergarten data which were relatively unaffected by these experimental problems.
(Word et al., 1990; Krueger, 1999). A second approach is based on considering ad-hoc regressions that evaluate the randomness of the within-school racial pairings by assessing the association between observed student traits and assignment to an own-race teacher. These approaches are combined in the kindergarten results presented in Table 2. The first column of Table 2 reports the results of an OLS regression where the dependent variable is a binary indicator for whether the kindergarten student had an own-race teacher. The regressors in this model are five basic student traits and school fixed effects. These results indicate that black students are substantially less likely to have an own-race teacher. This is to be expected since relatively few black teachers are available among the participating schools. However, if the matching of students and teachers were indeed random, we should find no within-school association between the other observed student traits and exposure to an own-race teacher. The results in Table 2 indicate that this is so. For example, poorer students (i.e., those receiving free lunches) are actually less likely to have an own-race teacher. However, the estimated effect is both small and statistically indistinguishable from zero. Similarly, gender, age and a small-class assignment all exhibit small and statistically weak relationships with assignment to an own-race teacher. These four variables are jointly insignificant determinants as well (p-value of 0.35). But the test score results in Table 2 indicate that assignment to an own-race teacher was associated with higher achievement in both math and reading. Kindergarten students assigned to an own-race teacher had math scores that were a statistically significant 3.6 percentile points higher. The estimated increase in reading scores among these kindergarten students is also large (2.9 percentile points) but not statistically distinguishable from zero.

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19 Throughout this study, standard errors that allow for classroom-specific heteroscedasticity are reported. Classroom identifiers were not reported in the public-use data. However, I effectively identified them by concatenating school, grade, class type assignment, merit pay status and teacher education. The validity of this classroom identifier was confirmed by my ability to replicate the class size distribution reported by Krueger (1999, Table 3). The correspondence of these distributions was exact except for the apparent typographical error in the number of students in 22-student regular-sized classes with aides. Krueger (1999) reports 329 students while I identified 330 students in 15 separate classes.
The kindergarten results in Table 2 provide important evidence that assignment to own-race teachers appears to have been independently given and that this assignment increased math achievement. However, most of the results in this study will instead exploit the pooled data since they can generate more statistical precision as well as identify cumulative effects. The increase in sample size will also allow us to estimate some models separately by race and gender. The key variable in most models based on the pooled data will again be a binary indicator for whether the student had an own-race teacher in a given year. Other models will exploit a variable that reflects the probability a student would have had an own-race teacher if they hadn't changed their classroom assignment. This type of "intent to treat" variable provides a plausible instrumental variable for the actual race of a student's teacher. However, it should be noted that, because of data limitations, this variable is not an exact measure of intent to treat. In particular, we do not know the exact class each student would have attended in the absence of treatment crossover or attrition. Instead, we know each student’s entry school and the classroom type (small, regular-sized, regular-sized with aide) that they first attended. A measure of intended exposure to an own-race teacher was constructed by matching each student to the race of the teacher they would have had in a particular grade if they remained within their entry school and classroom-type assignment. This variable takes on the appropriate fractional value in the few cases (15 percent) where students from a given entry school and classroom type could have had a black or white teacher in subsequent years.\textsuperscript{20} Another complication is that the classroom assignments that we observe for students in their first year are the actual ones, not necessarily the originally intended ones (which are not available in the data). However, Krueger (1999) compared intended and actual classroom assignments for kindergarten students from 18 schools and found that they differed for only 0.3 percent of students.

\textsuperscript{20} Also, in the few cases where the class type or grade was not observed for an entry school in later study years, students were assigned the mean value for the school and grade or just for the school.
Table 3 presents some critical evidence on whether the within-school variation in these putatively random measures is independent of other student characteristics. The results in the top panel relate the within-school variation in current exposure to an own-race teacher to other student traits. In all of these models, we again see small and statistically insignificant relationships between the current assignment to an own-race teacher and the observed student characteristics. Furthermore, as the p-values indicate, these variables are jointly insignificant as well. The bottom panel relates the intended assignment to an own-race teacher to these student traits. Again, these models indicate that, within schools and entry waves, there was no association between assignment to an own-race teacher and other student characteristics. These results provide an important validation of the exogeneity of the experimental assignment to an own-race teacher. However, the subsequent empirical models also examine the relevance of these issues by exploiting an intended assignment to an own-race teacher as an instrumental variable for their actual assignment and by evaluating reduced-form models that include imputed test scores for students who left the experiment or were absent when a test was given.

V. RESULTS

A. OLS and 2SLS estimates

The basic econometric model presented here relates $Y_{isgc}$, the grade and subject-specific percentile test rank for student i in school s, grade g and class c, to student, teacher and classroom traits and fixed effects for the grade, entry wave (kindergarten, grades 1 through 3) and the school of entry. More specifically, this model takes the following form:

$$Y_{isgc} = Z_{isgc} \Pi + X_{isgc} \beta + \alpha_g + (\alpha_f \times \alpha_s) + \epsilon_{isgc}$$

where $\alpha_g$, $\alpha_f$ and $\alpha_s$ are grade, entry-wave and school-of-entry fixed effects and $\epsilon_{isgc}$ is a mean-zero random error. Because randomization occurred in the school of entry upon the year of entry,
a full set of interactions between $\alpha_t$ and $\alpha_s$ is included (Krueger and Whitmore, 2001). However, fixed effects specifications that exclude these interactions return similar results. The matrix, $Z$, includes variables that vary at the individual level such as race, gender, age and free-lunch status. In the long form of this model, $Z$ also includes several student-specific measures of peer group traits: the percent of classmates receiving free lunches, the percent who are black, the percent who are female and the percent who attended kindergarten.$^{22}$ The matrix, $X$, includes class-specific variables such an assignment to an own-race teacher and assignment to a small class.$^{23}$ In the long form of this model, $X$ also includes other class-specific measures: years of teaching experience and its square and binary indicators for whether the teacher has a graduate degree and for whether the teacher is in the merit pay program. The impact of introducing these controls is of particular interest since it could suggest whether the observed effects of teacher race simply reflect the systematic racial differences in the background of teachers. As noted earlier, since there is classroom-specific variation in class size and other unobserved determinants, classroom-specific heteroscedasticity is accommodated in this model through the use of Huber-White standard errors.

Tables 4 and 5 present the OLS and 2SLS estimates of the effect of current exposure to an own-race teacher on math scores in models broken out by the race and gender of the students.$^{24}$ The results in Table 4, which are for white males and females, indicate that assignment to an own-race teacher is associated with a statistically significant 4 to 5 percentile-point increase in math scores. These results are quite robust to 2SLS estimation and to the

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$^{22}$ This last measure is subject to measurement error since we only know kindergarten attendance for those who were in the experiment at that time.

$^{23}$ Most of the results presented here are based on race-specific models so assignment to an own-race teacher varies at the classroom level. However, in models based on pooled data from black and white students, assignment to an own-race teacher varies at the individual level since it reflects the interaction a student and a teacher trait.

$^{24}$ The first-stage effect of an intended assignment to an own-race teacher is not reported. But, not surprisingly, the marginal effect of an intended assignment is quite large and statistically significant. The first-stage coefficient is typically around 0.9 and about 20 times larger than its standard error. Since the
introduction of the other teacher variables. The results in Table 5 document similarly robust and statistically precise effects in models of the math scores of black students, male and female. Tables 6 and 7 present the results of similar evaluations for reading scores. For white and black males and black females, assignment to an own-race teacher is associated with a statistically significant 3 to 6 percentile point increase in reading scores. For white females, these effects are positive but not statistically distinguishable from zero. The results in Tables 4 through 7 provide rather consistent and robust evidence of the link between exposure to an own-race teacher and increased student achievement. Interestingly, these estimated effects generally change little (within a fraction of the relevant standard errors) when controls for other teacher and peer traits are introduced and in 2SLS models. However, it should also be noted that the observed teacher traits are not uniformly associated with gains in achievement. But these models do indicate that exposure to more experienced teachers often led to statistically significant increases in achievement for white students (but at a decreasing rate). Additionally, there were statistically significant gains in the mathematics scores of black females when assigned to a teacher receiving merit pay or to a teacher with a graduate degree.

**B. Reduced-form results**

The robustness of the results in Tables 4 through 7 to 2SLS estimation suggests that treatment crossover does not confound this study's main inferences. However, the other substantive experimental violation of concern involved attrition from the experiment. The amount of attrition from Project STAR in each year was fairly large, ranging from roughly 20 to 30 percent annually (Hanushek, 1999). Auxiliary regressions indicate that the intended assignment to a regular-sized class significantly increased the probability of attrition. However, assignment to an own-race teacher appears to have been largely unrelated to the probability that a student left
the experiment. Specifically, regressions based on the full sample and three of the four race-gender samples indicate that the intended assignment to an own-race teacher had small and statistically insignificant effects on the probability of attrition. The sole exception involves the black, male students among whom an intended assignment to a white teacher significantly increased the probability of attrition. The existence of this sort of increased attrition in response to other-race teachers would impart a confounding bias to this study’s inferences if the students who left had an unobserved propensity for higher achievement. The lack of an association between assignment to an own-race teacher and observed student traits (in particular, free-lunch status) suggests that this is not so (Table 3). However, in Table 8, I present alternative evidence on this question by summarizing the key evaluation results from reduced-form test score models that include imputed data for students who left the experiment. More specifically, test score outcomes were crudely imputed for students who were absent or left the experiment by relying on the prior and subsequent subject-specific test score rankings available in the data set.\(^{25}\) The regression models for these data include binary indicators for age, free-lunch status and the intent to assign to a small class in addition to the grade and entry school by entry wave fixed effects.\(^{26}\) The key independent variable is the intended assignment to an own-race teacher since the actual assignment is unavailable for students who left the experiment. For comparison purposes, Table 8 also presents the results of this model when applied to only the actual test score data. These results suggest that the test-score gains associated with assignment to an own-race teacher are quite robust both in the full sample and across the demographic subgroups, including black males. More specifically, the right panel of Table 8 uniformly indicates that an own-race teacher increased math and reading achievement by 2 to 4 percentile points among the four subgroups.

\(^{25}\) A missing test score was first imputed by the average of an individual’s scores from the prior and subsequent years. If still missing, the imputation relied on the most recent prior scores and, then, subsequent scores. Krueger (1999) adopted a similar "last observation carry forward" method. I also experimented with imputing missing scores by relying on the subject-specific scores among students who remained in the intended school/grade/class-size cell. The results were similar to those reported here.
Though these estimated effects tend to be somewhat smaller than the corresponding estimates based only on actual test scores, these differences are fairly small relative to the sampling variation.

C. The role of teacher unobservables

Overall, these results indicate that assignment to an own-race teacher was associated with large and statistically significant achievement gains for both black and white students. Furthermore, the randomized pairings of students and teachers that occurred as part of the Project STAR experiment allow us to be unusually certain that these robust associations do not merely reflect the unobserved, student-level determinants of educational achievement. However, these estimates do not provide entirely unambiguous evidence on the importance of racial dynamics in the classroom. In particular, a fundamental difficulty with these results is that the apparent effects of teachers' race could simply reflect unobserved dimensions of teacher quality that happen to vary with race. For example, the results for black students are also consistent with the plausible, alternative hypothesis that the predominantly black schools tend to attract and retain high-quality black teachers but only low-quality white teachers. Similarly, the results for white students could merely reflect the possibility that the black teachers in predominantly white schools tend to be of lower quality. Furthermore, given the sharp racial segregation of students across most of these schools, it is also possible that both types of bias (relatively low-quality white teachers in black schools and relatively low-quality black teachers in white schools) occurred simultaneously.

One straightforward approach to assessing whether unobserved teacher quality imparts large biases is to evaluate specifications that introduce classroom fixed effects (i.e., $\alpha_c$). Over the four study years, Project STAR included 1,307 unique classrooms. The effect of an own-race teacher on test scores can be identified in models that include classroom fixed effects when data

---

26 Since an actual classroom assignment was not available for the imputed data, the heteroscedasticity in these models is accommodated at the level of school/grade/class-size cells.
for the black and white students are pooled and when we assume that the effects of an own-race teacher are the same for black and white students. In such a specification, the effects of an own-race teacher are effectively separated from the unobserved determinants associated with a particular teacher and classroom by relying on within-classroom comparisons, conditional on a student’s race, of students who did and did not share the teacher’s race. In Table 9, I report the key results from specifications that introduce classroom fixed effects. The results for math scores declined only slightly (by less than one standard error) after introducing these fixed effects. And the estimated effect of an own-race teacher on math scores remains quite large (i.e., 3.1 percentile points or 35 percent of the corresponding black-white test score gap) and statistically significant (p-value = .001). The estimated effect of an own-race teacher on reading scores exhibited more sensitivity to the introduction of classroom fixed effects, declining from 3.1 to 1.6 percentile points. However, this point estimate is still quite large (i.e., roughly 29 percent of the corresponding black-white test score gap) and is weakly significant (p-value = .075).

The results in Table 9 suggest that unobserved teacher quality does not entirely explain the achievement gains associated with assignment to an own-race teacher, particularly with respect to mathematics. However, the assumptions implicit in relying on the pooled data from black and white students specification may obscure the race-specific biases associated with unobserved teacher quality. An alternative and indirect way to assess these concerns is to suppose that the racial differences in observed teacher quality provide a guide to the racial differences in

27 The results from Tables 4 through 7 suggest that the achievement consequences of an own-race teacher are generally similar for black and white students. More formally, I also estimated the race-specific effects associated with assignment to an own-race teacher in models based on the pooled data. The hypothesis that these effects were equal for black and white students could not be rejected both in models for math scores (p-value = .5022) and for reading scores (p-value = .8749).

28 The weak significance of this effect reflects in part the conservative approach to constructing standard errors. Specifically, allowing classroom-specific, not individual-specific, heteroscedasticity increases the standard error by roughly 8 percent. This effect is also somewhat larger and more statistically significant (p-value = .047) in models that exclude the roughly 300 entry-school/entry-wave fixed effects. Furthermore, the evidence on response heterogeneity (Table 11) indicates that effects of own-race teachers on reading achievement are larger and more precise in sub-samples defined by certain student, teacher and class traits.
unobserved teacher quality. More specifically, a comparison of how observed teacher traits that appear to influence student achievement vary by teachers’ race may suggest how unobserved teacher quality varies by race. The robustness of the results in Tables 4 through 7 to the introduction of observed teacher traits would appear to indicate that unobserved teacher quality is not imparting confounding biases. However, since the teacher observables (years of experience, graduate degree, merit-pay status) were not uniformly associated with significant gains in student achievement, these relationships merit more detailed scrutiny. For example, as noted earlier, the results in Tables 4 and 6 indicate that the white students in Project STAR benefited significantly from assignment to more experienced teachers, particularly in terms of reading achievement, but less so from merit-pay teachers and those with graduate degrees. These results imply that, for white students, at least one observed teacher trait, years of experience, is an important component of teacher quality. Interestingly, among white students, assignment to an own-race teacher actually implies a teacher with fewer years of experience. To the extent that this particular racial difference in teacher quality provides a guide to the variation in unobserved teacher quality, it implies that the earlier results understate the achievement gains associated with assignment to own-race teachers for white students. Unfortunately, this exercise is not quite as dispositive for the black students in Project STAR since the apparent effects of observed teacher traits are generally smaller and more imprecisely estimated (Tables 5 and 7). However, there is evidence that, for black females, assignment to a merit-pay teacher or to a teacher with a graduate degree significantly increased mathematics scores (Table 5). And auxiliary regressions indicated that, for black students, assignment to an own-race teacher is actually associated with reduced probabilities of having a teacher with a graduate degree or merit pay. Therefore, like the results

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29 See Murphy and Topel (1990) and Altonji, Elder and Taber (2000) for applications of this approach to bounding omitted variable biases.
30 An auxiliary regression based on the full set of controls indicates that, for white students, own-race teachers have 1.2 fewer years of experience. However, neither this relationship nor that of the other teacher observables is significantly related to teachers’ race.
for white students, this racial pattern in teacher observables implies that the bias attributable to unobserved teacher quality could be negative.

Another indirect and qualified way to assess the race-specific biases that might be associated with unobserved teacher quality is to consider how the estimated effects associated with own-race teachers vary across particular types of schools. For example, a plausible conjecture is that schools with relatively few disadvantaged students (e.g., schools with lower shares of students receiving free lunches) were more able to recruit teachers of higher and more uniform quality. Similarly, schools whose students are racially homogeneous may only be able to recruit other-race teachers whose unobserved quality is relatively low. If these assumptions are correct, evidence that the apparent effects of own-race teachers are concentrated in schools with more disadvantaged students or in more segregated schools would be consistent with the hypothesis that unobserved teacher quality is empirically relevant. However, the existence of this sort of response heterogeneity would not unambiguously suggest the importance of unobserved teacher quality. The achievement gains associated with own-race teachers could be concentrated in more disadvantaged and segregated schools for a number of other plausible reasons. However, the extent to which the effects associated with own-race teachers are similar across such schools would point more clearly to the irrelevance of the racial patterns in unobserved teacher quality.

In Table 10, I present evidence on this issue by reporting, separately for white and black students, the estimated effects of own-race teachers across different types of schools. First, for each sample, I identified schools as having high or low socioeconomic status (SES) based on whether the school-level percent of students receiving free lunches exceeded the median value for this variable. Similarly, for white students, I identified schools as more segregated if their school-level percent white exceeded the median value. And, for black students, I identified schools as more segregated if their school-level percent black exceeded the median value. I then estimated test-score equations in which the effects associated with own-race teachers were allowed to vary across school type. The results in the top panel of Table 10 indicate that, for white students, the
test-score consequences of assignment to an own-race teacher are quite similar across these types of schools. In each of the four regressions the hypothesis that the effects across the two types of schools are the same cannot be rejected. So, assuming that these schools differ in their capacities to hire good teachers, this evidence suggests that unobserved teacher quality does not impart a substantive bias to the effects associated with own-race teachers. However, the results in the bottom panel of Table 10 indicate that, for black students, the test-score benefits of own-race teachers are largely concentrated in schools with more disadvantaged students and in more segregated schools. This evidence is consistent with the hypothesis that the apparent test-score gains associated with matching black students with black teachers partly reflect the systematically lower quality of white teachers who work in more disadvantaged and segregated black schools. However, these results could also simply indicate that the racial dynamics in classrooms (e.g., role-model effects, stereotype threat) are particularly relevant in more disadvantaged and segregated black communities.

D. Response heterogeneity and cumulative effects

The results in Tables 4 through 7 indicate that the effects associated with assignment to an own-race teacher are generally similar by race and gender. However, there are a variety of reasons that the effects associated with own-race teachers might vary by other student, teacher and classroom traits. In Table 11, I present evidence on the patterns of response heterogeneity by reporting the estimated effects of own-race teachers in several different samples of students defined by student, teacher and classroom traits. These results are based on the pooled data from black and white students and specifications that include classroom fixed effects. As points of reference, the first row in Table 11 reports the estimated effects of own-race teachers on math and reading scores from models that include all the available data (i.e., the results from Table 9). The

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31 The pattern of results is quite similar for models estimated separately for black and white students except where noted.
next results indicate that the effects of own-race teachers are largely concentrated among those who were assigned to regular-sized classes. For those assigned to small classes, own-race teachers appear to have positive effects but these estimates are smaller and statistically insignificant. These results also indicate that the achievement gains associated with an own-race teacher are somewhat larger among students with lower socioeconomic status (i.e., those receiving free lunches). These gains are also isolated among students assigned to relatively inexperienced teachers (11 or fewer years of experience) and are much smaller and statistically insignificant among those assigned to more experienced teachers. The results in Table 11 also suggest that the effects of own-race teachers do not vary across teachers with and without graduate degrees.  

The results in Tables 10 and 11 indicate that other student, teacher, classroom and school traits may have important consequences for the racial interactions between students and teachers. However, it should be noted that these differences are not always statistically meaningful. Nonetheless, a natural and important question to consider is whether these types of response heterogeneity suggest the extent to which own-race teachers matter because of passive teacher effects (e.g., role model effects, stereotype threat) or active ones (e.g., teacher biases). Unfortunately, these results are arguably consistent with a variety of hypotheses. For example, the relative absence of these effects in small classes could be due to more personal student-teacher interactions that obviate the racially driven role-model effects that could occur in larger classes (a passive effect). However, the racial biases in teacher behavior may also be less severe in smaller classes where a teacher's finite resources are less scarce (an active effect). This pattern of results could also reflect violations of the class-size experiment. For example, the racial dynamics could be more relevant in regular-sized classrooms simply because principals sent their best teachers to

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32 However, these particular results mask further heterogeneity by students’ race. For black students, the test-score gains associated with own-race teachers are concentrated among teachers without graduate degrees. However, for white students, these effects are larger among students assigned to teachers with graduate degrees.
the small classes or because of non-random attrition of high-ability students from regular-sized classes. Similarly, the concentration of these effects among more inexperienced teachers could reflect the importance of a teacher's age for race-based role-model effects (a passive effect) and the role of experience in attenuating unintended racial biases by teachers (an active effect). Furthermore, the particular relevance of black teachers for the black students in more segregated schools could reflect a relative need for role models (a passive effect) as well as the existence of lower expectations among white teachers in those schools (an active effect).

The prior empirical models presented here have implicitly assumed that there is a constant effect associated with a year's exposure to an own-race teacher regardless of a student's cumulative exposure. The results presented in Table 12 are based on alternative models that identify the effects of years of cumulative exposure in a relatively unrestricted manner by using a set of four binary indicators indicating whether the student was in their first to fourth year with an own-race teacher. The results uniformly suggest that student achievement was monotonically increasing in years of exposure to own-race teachers. More specifically, the first year of exposure implied an increase in student achievement of roughly 2 to 3 percentile points. The gains implied by the second year of exposure were relatively small. However, the third and fourth year of exposure each implied additional increases in student achievement of roughly 2 to 3 percentile points. The hypotheses that the coefficients on these indicators are equal can be easily rejected. These results imply that exposure to an own-race teacher does not simply confer a fixed, one-time gain but rather can have additive effects on a student’s achievement as they age.

VI. CONCLUSIONS

Frequent recommendations for the aggressive recruitment of minority teachers have been motivated by the putative educational benefits for minority students. However, the available

33 It should be noted that, even if the class-size experiment were compromised in these ways, it would not necessarily be problematic for this study’s inferences. And the available evidence suggests that it is not.
evidence that own-race teachers actually improve student achievement has at best been limited and qualified. As recent studies have recognized, it is difficult to make reliable inferences about this relationship given the pervasive specification problems associated with standard observational data on educational outcomes and the absence of compelling natural experiments. It was suggested here that the Project STAR class-size experiment presents a unique opportunity to examine the putative educational benefits of own-race teachers since it generated ostensibly random pairings of the students and teachers under study. This study presented such evaluations and found consistent evidence that there are rather large educational benefits for both black and white students from assignment to an own-race teacher in these early grades.

These results clearly provide novel support for the conventional assumption that recruiting minority teachers can generate important achievement gains among minority students. However, these results also suggest that one of the real and typically overlooked costs of such efforts may be a meaningful reduction in the educational achievement of non-minority students. There are also several important caveats appropriate to considering the broader policy implications of this study’s results. For example, these findings cannot, of course, speak directly to whether these effects exist in regions outside of the Tennessee schools under study. These results also do not address the effects of own-race teachers on important long-term student outcomes such as educational attainment. Furthermore, there may also be a variety of general-equilibrium costs and benefits that are associated with aggressively recruiting minority teachers that this study does not identify (e.g., reductions in quality or improved socialization and training for all teachers).

But, perhaps the most important caveat is that this study does not provide evidence on the exact mechanisms by which own-race teachers influence student achievement (i.e., the varying types of passive and active teacher effects). This gap in our knowledge is noteworthy for several reasons. In particular, a better understanding of the racial dynamics within classrooms could suggest policies and practices that improve upon current recommendations to recruit minority
teachers more aggressively. More specifically, understanding the relative educational consequences of role-model effects, stereotype threat and teacher biases could suggest new approaches that make teachers effective for all students regardless of race. Our limited understanding of the racial dynamics within classrooms is also relevant since this study’s results could be narrowly construed to suggest that an increased racial segregation of teachers and students should be promoted to improve the overall levels of educational achievement. Such a recommendation could be criticized not only on normative grounds but also because it ignores the possibility of more balanced policies informed by an improved understanding of why the racial interactions between students and teachers matter for student outcomes. A more appropriate interpretation of this study’s results is that it underscores the possibly sizable educational relevance of the racial dynamics between students and teachers as well as the need for a better understanding of what actually drives this phenomenon.

REFERENCES


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<th>Black Students</th>
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<tr>
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<td>56.7 (27.8)</td>
<td>50.0 (28.5)</td>
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<tr>
<td>Reading Score</td>
<td>56.9 (28.0)</td>
<td>51.5 (28.9)</td>
</tr>
<tr>
<td>Small Class</td>
<td>.31 (.46)</td>
<td>.29 (.45)</td>
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<tr>
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<td>.48 (.50)</td>
</tr>
<tr>
<td>Born Before 1980</td>
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<td>.38 (.49)</td>
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<td>Free Lunch</td>
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<td>.33 (.47)</td>
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<td>922</td>
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Notes - Standard deviations are reported in parentheses. Observed student characteristics are defined for the 23,883 student observations with math scores. The reading test is defined for 23,544 student observations.
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<th>Variable</th>
<th>Teacher of Own Race</th>
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<th>Reading Score</th>
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<td>2.9 (2.1)</td>
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<td>Small Class</td>
<td>-0.010 (0.028)</td>
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<td>5.7‡ (1.2)</td>
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<td>-8.4‡ (1.8)</td>
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<td>Born Before 1980</td>
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<td>5.8‡ (0.7)</td>
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<td>R²</td>
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<td>.3131</td>
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Notes – Robust standard errors are reported in parentheses. All models include school fixed effects. Roughly 77 percent of the kindergarten students have an own-race teacher.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 3 – Within-School Association between Student Traits and Assignment to Own-Race Teachers, By Student Race and Gender

<table>
<thead>
<tr>
<th>Variable</th>
<th>White Males</th>
<th>Black Males</th>
<th>White Females</th>
<th>Black Females</th>
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<td><strong>Dependent Variable: Own-Race Teacher</strong></td>
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<tr>
<td>Small Class</td>
<td>.013</td>
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<td>.043</td>
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<td></td>
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<td>(.045)</td>
<td>(.013)</td>
<td>(.048)</td>
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<td>.002</td>
<td>-.003</td>
<td>.001</td>
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<td></td>
<td>(.005)</td>
<td>(.015)</td>
<td>(.005)</td>
<td>(.017)</td>
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<td>-.001</td>
<td>.006</td>
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<td></td>
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<td>(.021)</td>
<td>(.005)</td>
<td>(.021)</td>
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<td>.1952</td>
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<td>.1910</td>
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<td>0.37</td>
<td>0.94</td>
<td>0.84</td>
</tr>
</tbody>
</table>

| **Dependent Variable: Intended Own-Race Teacher** |             |             |               |               |
| Small Class               | .014        | .049        | .007          | .034          |
|                           | (.010)      | (.033)      | (.010)        | (.033)        |
| Born Before 1980          | .002        | -.004       | -.002         | .007          |
|                           | (.004)      | (.012)      | (.004)        | (.013)        |
| Free Lunch                | .001        | -.011       | -.001         | .019          |
|                           | (.004)      | (.016)      | (.004)        | (.015)        |
| R²                        | .3173       | .2698       | .3103         | .2577         |
| p-value                   | 0.44        | 0.42        | 0.88          | 0.44          |
| Sample Size               | 8,328       | 4,024       | 7,665         | 3,939         |

Notes – Robust standard errors are reported in parentheses. All models include grade fixed effects and the interactions of fixed effects for the entry school and entry wave. The p-value refers to an F-test of the joint significance of the three variables.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 4 - Estimated Effects of an Own-Race Teacher on the Mathematics Scores of White Students by Gender

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<td></td>
<td>OLS 2SLS OLS 2SLS OLS 2SLS OLS 2SLS OLS 2SLS OLS 2SLS</td>
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<tr>
<td>Teacher of Own Race</td>
<td>4.6‡ 5.0† 4.4† 4.7* 4.5† 5.1† 4.0† 4.5*</td>
<td>(1.7) (2.4) (1.8) (2.5) (1.9) (2.6) (1.9) (2.7)</td>
<td></td>
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<tr>
<td>Small Class</td>
<td>4.5‡ 4.4‡ 4.3‡ 4.3‡ 3.1‡ 3.0‡ 2.9‡ 2.9‡</td>
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<td>-0.3 -0.4 -0.4 -0.4 -0.4 -0.4 -0.5 -0.5</td>
<td>(0.7) (0.7) (0.7) (0.7) (0.7) (0.7) (0.7) (0.7)</td>
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<td>(0.9) (0.9)</td>
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<td>(1.8) (1.8)</td>
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<tr>
<td>Percent of Classmates</td>
<td>-- -- 1.4 1.5 -- --</td>
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<td>(3.8) (3.8)</td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td>on Free Lunch</td>
<td>-- -- 0.9 0.9 -- --</td>
<td>(3.2) (3.3) -- --</td>
<td>(3.4) (3.4)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Percent of Classmates in</td>
<td>-- -- -6.0 -6.0 -- --</td>
<td>(6.6) (6.6) -- --</td>
<td>(7.1) (7.0)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Kindergarten</td>
<td>-- -- 0.2 0.2 -- --</td>
<td>(4.5) (4.6) -- --</td>
<td>(4.7) (4.7)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Black</td>
<td>-- -- -14.9† -14.8† -- --</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Percent of Classmates Female</td>
<td>-- -- 0.2 0.2 -- --</td>
<td>(4.5) (4.6) -- --</td>
<td>(4.7) (4.7)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
R²                           | .1729 .1729 .1746 .1746 .1867 .1867 .1896 .1896 |       |       |       |       |       |       |
Sample Size                   | 8,310 8,310 8,310 8,310 7,645 7,645 7,645 7,645 |       |       |       |       |       |       |

Notes – Robust standard errors are reported in parentheses. All models include grade fixed effects and the interactions of fixed effects for the entry school and entry wave.
* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 5 - Estimated Effects of an Own-Race Teacher
on the Mathematics Scores of Black Students by Gender

<table>
<thead>
<tr>
<th>Variable</th>
<th>Black Males</th>
<th></th>
<th>Black Females</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>OLS (1)</td>
<td>2SLS (2)</td>
<td>OLS (3)</td>
<td>2SLS (4)</td>
</tr>
<tr>
<td>Teacher of Own Race</td>
<td>3.2†</td>
<td>4.0†</td>
<td>3.2†</td>
<td>3.9*</td>
</tr>
<tr>
<td></td>
<td>(1.5)</td>
<td>(2.0)</td>
<td>(1.6)</td>
<td>(2.1)</td>
</tr>
<tr>
<td>Small Class</td>
<td>7.2‡</td>
<td>7.1‡</td>
<td>4.9‡</td>
<td>4.9‡</td>
</tr>
<tr>
<td></td>
<td>(1.5)</td>
<td>(1.5)</td>
<td>(1.7)</td>
<td>(1.7)</td>
</tr>
<tr>
<td>Born Before 1980</td>
<td>-0.11</td>
<td>-0.11</td>
<td>-0.09</td>
<td>-0.08</td>
</tr>
<tr>
<td></td>
<td>(0.9)</td>
<td>(0.9)</td>
<td>(0.9)</td>
<td>(1.0)</td>
</tr>
<tr>
<td>Free Lunch</td>
<td>-8.4‡</td>
<td>-8.3‡</td>
<td>-8.4‡</td>
<td>-7.6‡</td>
</tr>
<tr>
<td></td>
<td>(1.2)</td>
<td>(1.2)</td>
<td>(1.2)</td>
<td>(1.1)</td>
</tr>
<tr>
<td>Teacher Experience</td>
<td>--</td>
<td>--</td>
<td>-.17</td>
<td>-.16</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.26)</td>
<td>(.26)</td>
</tr>
<tr>
<td>Teacher Experience Squared</td>
<td>--</td>
<td>--</td>
<td>.006</td>
<td>.006</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(.008)</td>
<td>(.008)</td>
</tr>
<tr>
<td>Graduate Degree</td>
<td>--</td>
<td>--</td>
<td>1.3</td>
<td>1.4</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(1.6)</td>
<td>(1.6)</td>
</tr>
<tr>
<td>Merit Pay</td>
<td>--</td>
<td>--</td>
<td>2.1</td>
<td>2.1</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(2.1)</td>
<td>(2.1)</td>
</tr>
<tr>
<td>Percent of Classmates on Free Lunch</td>
<td>--</td>
<td>--</td>
<td>7.6</td>
<td>7.7</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(7.0)</td>
<td>(7.0)</td>
</tr>
<tr>
<td>Percent of Classmates in Kindergarten</td>
<td>--</td>
<td>--</td>
<td>16.2‡</td>
<td>16.3‡</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(4.7)</td>
<td>(4.7)</td>
</tr>
<tr>
<td>Percent of Classmates Black</td>
<td>--</td>
<td>--</td>
<td>-3.3</td>
<td>-3.3</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(8.8)</td>
<td>(8.8)</td>
</tr>
<tr>
<td>Percent of Classmates Female</td>
<td>--</td>
<td>--</td>
<td>3.9</td>
<td>4.0</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(8.2)</td>
<td>(8.2)</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include grade fixed effects and the interactions of fixed effects for the entry school and entry wave.
* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
### Table 6 - Estimated Effects of an Own-Race Teacher on the Reading Scores of White Students by Gender

<table>
<thead>
<tr>
<th>Variable</th>
<th>White Males</th>
<th></th>
<th>White Females</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>OLS 2SLS</td>
<td>OLS 2SLS</td>
<td>OLS 2SLS</td>
<td>OLS 2SLS</td>
</tr>
<tr>
<td>Teacher of Own Race</td>
<td>4.1†</td>
<td>4.6†</td>
<td>3.9†</td>
<td>4.2*</td>
</tr>
<tr>
<td></td>
<td>(1.6)</td>
<td>(2.2)</td>
<td>(1.6)</td>
<td>(1.7)</td>
</tr>
<tr>
<td>Small Class</td>
<td>3.6‡</td>
<td>3.6‡</td>
<td>3.4‡</td>
<td>3.3‡</td>
</tr>
<tr>
<td></td>
<td>(0.8)</td>
<td>(0.8)</td>
<td>(0.9)</td>
<td>(0.8)</td>
</tr>
<tr>
<td>Born Before 1980</td>
<td>-2.7‡</td>
<td>-2.7‡</td>
<td>-2.7‡</td>
<td>-2.4‡</td>
</tr>
<tr>
<td></td>
<td>(0.7)</td>
<td>(0.7)</td>
<td>(0.7)</td>
<td>(0.7)</td>
</tr>
<tr>
<td>Free Lunch</td>
<td>-11.6‡</td>
<td>-11.6‡</td>
<td>-11.6‡</td>
<td>-13.3‡</td>
</tr>
<tr>
<td></td>
<td>(0.7)</td>
<td>(0.7)</td>
<td>(0.7)</td>
<td>(0.7)</td>
</tr>
<tr>
<td>Teacher Experience</td>
<td>--</td>
<td>--</td>
<td>.49‡</td>
<td>.49‡</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(.15)</td>
<td>(.15)</td>
</tr>
<tr>
<td>Teacher Experience Squared</td>
<td>--</td>
<td>--</td>
<td>-.014‡</td>
<td>-.014‡</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(.004)</td>
<td>(.004)</td>
</tr>
<tr>
<td>Graduate Degree</td>
<td>--</td>
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<td>-1.0</td>
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<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(0.8)</td>
<td>(0.8)</td>
</tr>
<tr>
<td>Merit Pay</td>
<td>--</td>
<td>--</td>
<td>2.1</td>
<td>2.1</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(1.6)</td>
<td>(1.6)</td>
</tr>
<tr>
<td>Percent of Classmates on Free Lunch</td>
<td>--</td>
<td>--</td>
<td>-1.9</td>
<td>-1.8</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(3.7)</td>
<td>(3.7)</td>
</tr>
<tr>
<td>Percent of Classmates in Kindergarten</td>
<td>--</td>
<td>--</td>
<td>1.5</td>
<td>1.5</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(3.2)</td>
<td>(3.2)</td>
</tr>
<tr>
<td>Percent of Classmates Black</td>
<td>--</td>
<td>--</td>
<td>-12.9†</td>
<td>-12.8†</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(6.2)</td>
<td>(6.1)</td>
</tr>
<tr>
<td>Female</td>
<td>--</td>
<td>--</td>
<td>2.0</td>
<td>2.0</td>
</tr>
<tr>
<td></td>
<td>--</td>
<td>--</td>
<td>(4.5)</td>
<td>(4.5)</td>
</tr>
<tr>
<td>R²</td>
<td>.1887</td>
<td>.1887</td>
<td>.1918</td>
<td>.1918</td>
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<tr>
<td>Sample Size</td>
<td>8,154</td>
<td>8,154</td>
<td>8,154</td>
<td>8,154</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include grade fixed effects and the interactions of fixed effects for the entry school and entry wave.
* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 7 - Estimated Effects of an Own-Race Teacher on the Reading Scores of Black Students by Gender

<table>
<thead>
<tr>
<th>Variable</th>
<th>Black Males</th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>OLS (1)</td>
<td>2SLS (2)</td>
<td>OLS (3)</td>
<td>2SLS (4)</td>
<td>OLS (5)</td>
<td>2SLS (6)</td>
<td>OLS (7)</td>
</tr>
<tr>
<td>Teacher of Own Race</td>
<td>3.3† (1.4)</td>
<td>4.7† (2.0)</td>
<td>3.0† (1.5)</td>
<td>4.5† (2.1)</td>
<td>3.7‡ (1.4)</td>
<td>6.0‡ (2.2)</td>
<td>3.7† (1.4)</td>
</tr>
<tr>
<td>Small Class</td>
<td>6.6‡ (1.4)</td>
<td>6.5‡ (1.5)</td>
<td>5.6‡ (1.6)</td>
<td>5.6‡ (1.6)</td>
<td>6.8‡ (1.4)</td>
<td>6.7‡ (1.5)</td>
<td>5.0‡ (1.6)</td>
</tr>
<tr>
<td>Born Before 1980</td>
<td>-3.2‡ (0.9)</td>
<td>-3.3‡ (0.9)</td>
<td>-3.3‡ (0.9)</td>
<td>-1.3 (0.9)</td>
<td>-1.3 (0.9)</td>
<td>-1.3 (0.9)</td>
<td>-1.3 (0.9)</td>
</tr>
<tr>
<td>Free Lunch</td>
<td>-6.8‡ (1.2)</td>
<td>-6.7‡ (1.2)</td>
<td>-6.7‡ (1.2)</td>
<td>-10.7‡ (1.3)</td>
<td>-10.7‡ (1.3)</td>
<td>-10.3‡ (1.3)</td>
<td>-10.3‡ (1.3)</td>
</tr>
<tr>
<td>Teacher Experience</td>
<td>--</td>
<td>0.13 (0.26)</td>
<td>0.13 (0.26)</td>
<td>--</td>
<td>--</td>
<td>0.44* (0.24)</td>
<td>0.44* (0.24)</td>
</tr>
<tr>
<td>Teacher Experience</td>
<td>--</td>
<td>-0.0003 (0.007)</td>
<td>-0.001 (0.007)</td>
<td>--</td>
<td>--</td>
<td>-0.009 (0.007)</td>
<td>-0.010 (0.007)</td>
</tr>
<tr>
<td>Squared</td>
<td>--</td>
<td>1.0 (1.5)</td>
<td>1.2 (1.5)</td>
<td>--</td>
<td>--</td>
<td>1.8 (1.5)</td>
<td>2.1 (1.5)</td>
</tr>
<tr>
<td>Graduate Degree</td>
<td>--</td>
<td>0.5 (2.0)</td>
<td>0.6 (2.0)</td>
<td>--</td>
<td>--</td>
<td>1.1 (2.1)</td>
<td>1.3 (2.2)</td>
</tr>
<tr>
<td>Merit Pay</td>
<td>--</td>
<td>3.9 (6.6)</td>
<td>4.1 (6.6)</td>
<td>--</td>
<td>--</td>
<td>2.6 (6.7)</td>
<td>3.0 (6.7)</td>
</tr>
<tr>
<td>Percent of Classmates</td>
<td>--</td>
<td>6.2 (4.4)</td>
<td>6.3 (4.4)</td>
<td>--</td>
<td>--</td>
<td>11.9‡ (4.7)</td>
<td>11.8† (4.8)</td>
</tr>
<tr>
<td>on Free Lunch</td>
<td>--</td>
<td>-7.5 (8.3)</td>
<td>-7.7 (8.4)</td>
<td>--</td>
<td>--</td>
<td>-12.8* (7.4)</td>
<td>-13.5* (7.4)</td>
</tr>
<tr>
<td>in Kindergarten</td>
<td>--</td>
<td>8.1 (7.6)</td>
<td>8.4 (7.6)</td>
<td>--</td>
<td>--</td>
<td>20.5‡ (7.2)</td>
<td>21.3‡ (7.1)</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include grade fixed effects and the interactions of fixed effects for the entry school and entry wave.
* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 8 - Estimated Effects of an Intended Own-Race Teacher, Actual and Imputed Test Scores

<table>
<thead>
<tr>
<th>Sample</th>
<th>Math Score</th>
<th>Sample Size</th>
<th>Reading Score</th>
<th>Sample Size</th>
<th>Math Score</th>
<th>Sample Size</th>
<th>Reading Score</th>
<th>Sample Size</th>
</tr>
</thead>
<tbody>
<tr>
<td>Full Sample</td>
<td>4.2‡</td>
<td>23,883</td>
<td>3.7‡</td>
<td>23,544</td>
<td>3.0‡</td>
<td>34,317</td>
<td>2.4‡</td>
<td>33,978</td>
</tr>
<tr>
<td>White Male</td>
<td>4.4†</td>
<td>8,310</td>
<td>4.0*</td>
<td>8,154</td>
<td>2.8*</td>
<td>11,679</td>
<td>2.8†</td>
<td>11,535</td>
</tr>
<tr>
<td>White Female</td>
<td>4.4†</td>
<td>7,645</td>
<td>1.0</td>
<td>7,518</td>
<td>4.0‡</td>
<td>10,506</td>
<td>1.8</td>
<td>10,379</td>
</tr>
<tr>
<td>Black Male</td>
<td>3.6*</td>
<td>4,005</td>
<td>4.2†</td>
<td>3,972</td>
<td>2.7†</td>
<td>6,270</td>
<td>3.1†</td>
<td>6,219</td>
</tr>
<tr>
<td>Black Female</td>
<td>4.4†</td>
<td>3,923</td>
<td>4.9‡</td>
<td>3,900</td>
<td>2.4*</td>
<td>5,862</td>
<td>2.3</td>
<td>5,845</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include binary indicators for race, gender, age, free lunch status, intended small class assignment, grade fixed effects and the interactions of fixed effects for the entry school and entry wave.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 9 - Estimated Effects of an Own-Race Teacher on Mathematics and Reading Scores, With and Without Classroom Fixed Effects

<table>
<thead>
<tr>
<th>Variable</th>
<th>Math Score (n=23,883)</th>
<th>Reading Score (n=23,544)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
</tr>
<tr>
<td>Teacher of Own Race</td>
<td>3.8‡</td>
<td>3.1‡</td>
</tr>
<tr>
<td></td>
<td>(1.0)</td>
<td>(0.9)</td>
</tr>
<tr>
<td>Black Student</td>
<td>-8.8‡</td>
<td>-8.9‡</td>
</tr>
<tr>
<td></td>
<td>(1.0)</td>
<td>(1.0)</td>
</tr>
<tr>
<td>Female Student</td>
<td>1.2‡</td>
<td>1.1‡</td>
</tr>
<tr>
<td></td>
<td>(0.4)</td>
<td>(0.3)</td>
</tr>
<tr>
<td>Born Before 1980</td>
<td>-0.2</td>
<td>-0.2</td>
</tr>
<tr>
<td></td>
<td>(0.4)</td>
<td>(0.4)</td>
</tr>
<tr>
<td>Free Lunch</td>
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<tr>
<td></td>
<td>(0.4)</td>
<td>(0.4)</td>
</tr>
<tr>
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</tr>
<tr>
<td>R²</td>
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<td>.3835</td>
</tr>
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<td></td>
<td>.2449</td>
</tr>
<tr>
<td></td>
<td></td>
<td>.3805</td>
</tr>
</tbody>
</table>

| Classroom Fixed Effects       | no                    | yes                      |

Notes – Robust standard errors are reported in parentheses. All models include the interactions of fixed effects for the entry school and entry wave. Models (1) and (3) also include grade fixed effects and the controls for peer, class and teacher traits.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 10 – Estimated Effects of an Own-Race Teacher on Mathematics and Reading Scores, By School Traits

<table>
<thead>
<tr>
<th>Variable</th>
<th>Mean</th>
<th>Math Score</th>
<th>Reading Score</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>White students</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>High-SES School x Own-Race Teacher</td>
<td>.48</td>
<td>4.5†</td>
<td>1.9</td>
</tr>
<tr>
<td>Low-SES School x Own-Race Teacher</td>
<td>.46</td>
<td>4.0†</td>
<td>2.9*</td>
</tr>
<tr>
<td>p-value</td>
<td></td>
<td>.8406</td>
<td>.6544</td>
</tr>
<tr>
<td>More Segregated School x Own-Race Teacher</td>
<td>.50</td>
<td>5.7†</td>
<td>3.5</td>
</tr>
<tr>
<td>Less Segregated School x Own-Race Teacher</td>
<td>.44</td>
<td>3.9†</td>
<td>2.3</td>
</tr>
<tr>
<td>p-value</td>
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<td>.5767</td>
<td>.6564</td>
</tr>
<tr>
<td><strong>Black students</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>High-SES School x Own-Race Teacher</td>
<td>.18</td>
<td>0.2</td>
<td>1.0</td>
</tr>
<tr>
<td>Low-SES School x Own-Race Teacher</td>
<td>.26</td>
<td>5.8‡</td>
<td>5.0‡</td>
</tr>
<tr>
<td>p-value</td>
<td></td>
<td>.0345</td>
<td>.1054</td>
</tr>
<tr>
<td>More Segregated School x Own-Race Teacher</td>
<td>.26</td>
<td>5.4‡</td>
<td>5.1‡</td>
</tr>
<tr>
<td>Less Segregated School x Own-Race Teacher</td>
<td>.18</td>
<td>0.6</td>
<td>0.8</td>
</tr>
<tr>
<td>p-value</td>
<td></td>
<td>.0655</td>
<td>.0687</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. The math and reading results are based on 15,955 and 15,672 observations, respectively, for white students and on 7,928 and 7,872 observations, respectively, for black students. All models include the controls for student, peer, class and teacher traits, grade fixed effects and the interactions of fixed effects for the entry school and entry wave. The p-value refers to an F-test of the null hypothesis that the effects associated with each type of school are the same.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 11 – Estimated Effects of an Own-Race Teacher on Mathematics and Reading Scores, By Class, Student and Teacher Traits

<table>
<thead>
<tr>
<th>Sample</th>
<th>Math Score</th>
<th>Sample Size</th>
<th>Reading Score</th>
<th>Sample Size</th>
</tr>
</thead>
<tbody>
<tr>
<td>Full Sample</td>
<td>3.1‡</td>
<td>23,883</td>
<td>1.6*</td>
<td>23,544</td>
</tr>
<tr>
<td></td>
<td>(1.0)</td>
<td></td>
<td>(0.9)</td>
<td></td>
</tr>
<tr>
<td>Regular-Sized Class</td>
<td>3.6‡</td>
<td>16,699</td>
<td>2.0†</td>
<td>16,437</td>
</tr>
<tr>
<td></td>
<td>(1.1)</td>
<td></td>
<td>(1.0)</td>
<td></td>
</tr>
<tr>
<td>Small Class</td>
<td>1.1</td>
<td>7,184</td>
<td>0.7</td>
<td>7,107</td>
</tr>
<tr>
<td></td>
<td>(1.7)</td>
<td></td>
<td>(2.0)</td>
<td></td>
</tr>
<tr>
<td>No Free Lunch</td>
<td>2.8*</td>
<td>12,214</td>
<td>0.9</td>
<td>12,074</td>
</tr>
<tr>
<td></td>
<td>(1.4)</td>
<td></td>
<td>(1.5)</td>
<td></td>
</tr>
<tr>
<td>Free Lunch</td>
<td>3.6†</td>
<td>11,669</td>
<td>2.2</td>
<td>11,470</td>
</tr>
<tr>
<td></td>
<td>(1.5)</td>
<td></td>
<td>(1.4)</td>
<td></td>
</tr>
<tr>
<td>Inexperienced Teachers</td>
<td>4.5‡</td>
<td>12,363</td>
<td>2.2*</td>
<td>12,227</td>
</tr>
<tr>
<td></td>
<td>(1.3)</td>
<td></td>
<td>(1.2)</td>
<td></td>
</tr>
<tr>
<td>Experienced Teachers</td>
<td>1.3</td>
<td>11,520</td>
<td>1.0</td>
<td>11,317</td>
</tr>
<tr>
<td></td>
<td>(1.3)</td>
<td></td>
<td>(1.3)</td>
<td></td>
</tr>
<tr>
<td>Graduate Degree</td>
<td>2.9*</td>
<td>8,991</td>
<td>1.5</td>
<td>8,855</td>
</tr>
<tr>
<td></td>
<td>(1.6)</td>
<td></td>
<td>(1.8)</td>
<td></td>
</tr>
<tr>
<td>No Graduate Degree</td>
<td>3.4‡</td>
<td>14,892</td>
<td>1.9*</td>
<td>14,689</td>
</tr>
<tr>
<td></td>
<td>(1.1)</td>
<td></td>
<td>(1.1)</td>
<td></td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include the controls for student traits, classroom fixed effects and the interactions of fixed effects for the entry school and entry wave. "Inexperienced" teachers are defined as those with 11 or fewer years of experience; experienced teachers have more than 11 years of experience.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level
Table 12 – Estimated Effects of Cumulative Years with an Own-Race Teacher on Mathematics and Reading Scores

<table>
<thead>
<tr>
<th>Variable</th>
<th>Math Score (n=23,833)</th>
<th>Reading Score (n=23,544)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cumulative Years with an Own-Race Teacher</td>
<td></td>
<td></td>
</tr>
<tr>
<td>One</td>
<td>2.9‡</td>
<td>1.6†</td>
</tr>
<tr>
<td></td>
<td>(0.8)</td>
<td>(0.8)</td>
</tr>
<tr>
<td>Two</td>
<td>3.2‡</td>
<td>1.8*</td>
</tr>
<tr>
<td></td>
<td>(1.0)</td>
<td>(1.1)</td>
</tr>
<tr>
<td>Three</td>
<td>5.3‡</td>
<td>4.8‡</td>
</tr>
<tr>
<td></td>
<td>(1.5)</td>
<td>(1.4)</td>
</tr>
<tr>
<td>Four</td>
<td>8.9‡</td>
<td>8.2‡</td>
</tr>
<tr>
<td></td>
<td>(1.9)</td>
<td>(1.9)</td>
</tr>
<tr>
<td>( R^2 )</td>
<td>.3840</td>
<td>.3812</td>
</tr>
</tbody>
</table>

Notes – Robust standard errors are reported in parentheses. All models include the controls for student traits, classroom fixed effects and the interactions of fixed effects for the entry school and entry wave.

* Statistically significant at 10-percent level
† Statistically significant at 5-percent level
‡ Statistically significant at 1-percent level