

Social Capital and Student Achievement: An Intervention-Based Test of Theory

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Abstract

Social capital is widely cited as benefiting children's school performance, but close inspection of existing research yields inconsistent findings. Focusing on intergenerational closure among parents of children in the same school, this article draws from a field experiment to test the effects of social capital on children's achievement in reading and mathematics. When children were in first grade, their schools were randomly assigned to an after-school family-based intervention that boosts social capital. A total of 52 schools in Phoenix, Arizona, and San Antonio, Texas, containing over 3,000 first graders, participated in the study, with half the schools in each city assigned to the treatment group and half serving as no-treatment controls. Two years later, no differences in third-grade achievement were evident between children who had been in treatment schools versus control schools. By contrast, nonexperimental analyses of survey-based measures of social capital suggest positive effects on achievement, indicating that naïve estimates based on survey measures may be upwardly biased by unobserved conditions that lead to both stronger ties among parents and higher test scores. This article adds to a growing literature that raises doubts about the effects of this type of social capital for achievement outcomes among young children.

Keywords

social capital, intergenerational closure, elementary school, achievement, parental educational expectations

The relation between social capital and student achievement has long been of interest to sociologists of education, especially since Coleman (1988) theorized that social capital plays a key role in the cultivation of human capital in children. Much of the attention centers on the concept and potential benefits of intergenerational closure, that is, the degree to which children belong to closed social networks that include their parents, their friends, and the parents of their friends (Coleman 1988, 1990). According to this view, when parents can surround their children with social networks characterized by trust and shared expectations, they may be better positioned to share valuable information and to establish norms that help their children succeed in school.

A sizable literature points to the promise of social capital for improving children's achieve-

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Adam Gamoran, William T. Grant Foundation, 60 East 42nd Street, 43rd floor, New York, NY 10165. Email: agamoran@wtgrantfdn.org ment, and multiple studies demonstrate a positive relationship between measures of intergenerational closure in particular and schoolchildren's academic performance (Carbonaro 1998; Dika and Singh 2002; Israel, Beaulieu, and Hartless 2001; Kao and Rutherford 2007; Pong 1998; Rodriguez Menés and Donato 2015; Sun 1998, 1999). However, growing evidence casts doubt on this positive relationship, pointing to the correlational nature of existing evidence (Durlauf 2002; Freeman and Condron 2011; Geven and van de Werfhorst 2020; Morgan and Sørensen 1999; Morgan and Todd 2009; Mouw 2006; Pribesh and Downey 1999).

At the heart of these conflicting findings is a thorny empirical challenge that is intertwined with social capital's theoretical underpinnings. Social capital theory is notoriously difficult to operationalize because the theory itself posits that people often selectively form relationships in the pursuit of anticipated resources or opportunities (Bourdieu 1986). Hence, it is hard to disentangle whether social capital promotes favorable educational outcomes or whether resources and attributes that promote educational success attract more social capital (Dika and Singh 2002; Durlauf 2002; Mouw 2006). Prior studies of social capital effects on students' achievement rely on observational data and are thus vulnerable to selection bias, offering little purchase to discern the causal direction of the relation between strong social networks among families and children's performance in school (Dika and Singh 2002).

This study offers a novel perspective on the social capital debate by using a school-randomized field experiment to stimulate social capital development and test for effects untainted by selection bias. Schools with large enrollments of low-income, Hispanic students were randomly assigned to an intervention-Families and Schools Together (FAST)-designed to boost social capital among families and between families and schools. Earlier analyses of these experimental data provide strong evidence that FAST strengthened the structure of parent-parent social networks as well as the quality of relationships among parents of first-grade students within schools (Rangel, Shoji, and Gamoran 2020). We examine the effects of this socialcapital-building intervention two years later on students' reading and mathematics achievement test scores in third grade.

Whereas most studies of the relation between families' social networks and children's academic outcomes focus on middle or high school students, our study examines these relationships among families of elementary school children. Implementing the intervention in first grade leverages a key time in the educational career, when families are just beginning to form relations with other parents in the school community and when parents have a substantial influence on children. Examining achievement in third grade focuses on academic growth in the key early childhood years and also gives social capital time to operate-for example, by building relationships that facilitate information exchanges, norm development, interpersonal support, and social control. Because the multiyear, multifamily intervention was randomly assigned, the analysis can distinguish the effects of social capital from unobserved characteristics of families or schools that may both enhance connections among parents and children in the same school and boost children's achievement. Our article adds new evidence to the debate over the relation between social capital and children's academic achievement, and our approach offers a powerful way to test a theoretical construct central to the sociology of education.

SOCIAL CAPITAL AND SCHOOLING OUTCOMES

Scholars have explored the role of social capital in children's school success for more than three decades (Bankston and Zhou 2002; Coleman 1988; Dika and Singh 2002; Dufur et al. 2016; Dufur, Parcel, and Troutman 2013; Furstenburg and Hughes 1995; Putnam 2015; Rodriguez Menés and Donato 2015). The concept of social capital rose to prominence among sociologists in the 1980s with the contributions of Bourdieu (1986) and Coleman (1988), who offered two distinct yet related theoretical frameworks (Portes 2000). Both frameworks highlight the benefits of social ties for individuals or groups, but Bourdieu's conceptualization of social capital is based in theories of social reproduction that emphasize not only the structure of networks but also the quality of resources inhering in networks and individuals' ability to access those resources. For Bourdieu, social capital can accentuate social stratification individuals with better access to because resource-rich networks and those with more power to access those resources within networks can obtain greater benefits.

In contrast, Coleman's conceptualization of social capital is based in rational choice theory and emphasizes the qualities of shared information, trust, and enforcement of norms that characterize these networks. Yet Coleman's (1990:302) approach also has functionalist roots in that he defined social capital "by its function" as it "facilitates certain actions of individuals within the [social] structure." For Coleman, social capital can mitigate stratification because the variety of individuals who share in a network's ties can all access and benefit from the resources available to the network. According to this view, social capital not only complements other forms of capital, but it can also help build human capital in circumstances where economic capital is in short supply (Coleman 1988).

Early research on social capital in education largely followed Coleman's approach (Dika and Singh 2002), likely because of Coleman's emphasis on the importance of families and schools for social capital effects on children and because of his own empirical research on the topic (Coleman and Hoffer 1987). Coleman's concepts and measures remain prevalent in this field, and our approach is likewise largely consistent with Coleman's work. His conception is well suited to our inquiry given our interest in whether it is possible to intentionally build social capital among families within schools and by doing so, to enhance the school performance of children from marginalized communities. However, like other recent studies, our approach responds to a salient critique of Coleman's conception: His defining social capital by its function would preclude empirical assessment of its impact because if the definition of social capital is that it "facilitates certain actions," one cannot test whether social capital is causally implicated (Durlauf 2002; Portes 1998). Instead, we distinguish between indicators of social capital such as the size, structure, and quality of relationships in a social network and the potential payoff for child development (Sampson, Morenoff, and Earls 1999).

Thus, in our conception—drawn largely from Coleman but avoiding the tautological logic of defining social capital by its function—when parents and educators belong to strong social networks characterized by trust and shared expectations for children, they can communicate more effectively and enforce norms that can help children succeed in school. A widely studied indicator of social capital, and a primary focus of this article, is *intergenerational closure*, the extent to which children are embedded in closed social networks consisting of their friends, their parents, and the parents of their friends (Coleman 1988, 1990). Commonly measured by asking parents how many of the parents of their children's friends they know (Carbonaro 1998), intergenerational closure is a marker of network structure, signaling both size and closure (Geven and van de Werfhorst 2020).¹ Other aspects of social capital include the quality of relationships among parents of children in the same school, such as the degree of trust and shared expectations and the extent to which parents regularly interact (Bryk and Schneider 2002; Gamoran et al. 2012).

How might social capital among families in a school elevate children's chances of academic success? Key mechanisms proposed in past research include a flow of useful information and social control. Isolated families may lack access to information about how to help their children succeed in school. Information that flows through a social network enables parents to help their children, whether by advocating for teachers who provide the best support, understanding teachers' expectations, or aiding children with homework (Horvat, Weininger, and Lareau 2003; Putnam 2015). Additionally, when parents are connected with other parents, they can act collectively to set and enforce norms for their children's behavior (Carbonaro 1998). From nutrition to bedtime expectations to keeping up with schoolwork, establishing collective norms makes enforcing rules less stressful and more taken for granted.

For young children, who are the focus of this investigation, social capital derived from parents' networks may be especially consequential for their schooling outcomes. Most studies of the relations between parents' and families' social networks and children's academic outcomes focus on middle or high school students. Yet as others have argued (Freeman and Condron 2011), parents' social networks may play an especially important role in early childhood, before peer effects become more salient and influential as children enter adolescence.

Does Social Capital Boost Achievement?

Much of the research on social capital among families and its relation to children's achievement relies on data from a national survey of the high school class of 1992, which began with eighth graders in 1988 and was the first large U.S. panel survey of youth to collect information on intergenerational closure (measured via parent surveys) and student achievement (measured with standardized achievement tests in math, reading, science, and history). Analyses of this survey, known as the National Educational Longitudinal Study of 1988, have reached contradictory conclusions. Despite using the same data, some studies report a positive relationship between intergenerational closure and students' achievement scores, particularly in math, but others do not. Analyses that focus on achievement growth (Carbonaro 1998; Kao and Rutherford 2007; Morgan and Sørensen 1999; Pribesh and Downey 1999) less often find positive effects compared to studies that focus on differences in achievement at a point in time (Carbonaro 1998; Israel et al. 2001; Kao and Rutherford 2007; Sun 1998, 1999).

Research using more recent data seems to reflect the same pattern. For example, crosssectional analyses of the 2006 Program for International Student Assessment survey data revealed positive associations between social capital and cognitive performance in science (Rodriguez Menés and Donato 2015), whereas longitudinal analyses of a U.S. national survey from the high school class of 2004 found no association between intergenerational closure and math score gains for public school students, although there was a positive association for Catholic school students (Morgan and Todd 2009). And using data on adolescents in Germany and the Netherlands, Geven and van de Werfhorst (2020) observed positive associations between intergenerational networks and school grades when comparing among students, but within-person estimates that controlled for all time-invariant unobserved confounders of closure and grades yielded null effects.²

Of particular note for our study on elementary school children, two studies using the Early Childhood Longitudinal Study-Kindergarten cohort (ECLS-K) also found no relationship between intergenerational closure and achievement gains among children in kindergarten and first grade (Condron 2009; Freeman and Condron 2011).³ Freeman and Condron (2011:541) concluded that their study "adds to the mounting research suggesting that [intergenerational] closure is not an important predictor of students' learning." However, another study using the ECLS-K reported significant and positive effect of а

intergenerational closure on reading gains (but not math gains) for a subset of children living in stepfamilies (Shriner, Mullis, and Schlee 2009).

In summary, numerous studies suggest parentparent social capital is beneficial to student achievement (Carbonaro 1998; Dika and Singh 2002; Israel et al. 2001; Kao and Rutherford 2007; Pong 1998; Rodriguez Menés and Donato 2015; Sun 1998, 1999), but other work challenges this conclusion, pointing to the correlational nature of existing evidence on social capital in general (e.g., Durlauf 2002; Mouw 2006) and intergenerational closure in particular (Freeman and Condron 2011; Geven and van de Werfhorst 2020; Morgan and Sørensen 1999; Morgan and Todd 2009; Pribesh and Downey 1999). With observational data, it is difficult to discern whether a positive association reflects an effect of social capital on schooling outcomes or whether relationships among families thrive in schools where children are performing well (Dika and Singh 2002). Our study design responds to this limitation.

RESEARCH QUESTIONS, STUDY DESIGN, AND METHODS

The most compelling design to test the causal effects of social capital would involve random assignment of families to varying conditions of social capital. Of course, this is neither feasible nor desirable. Instead, we adopted an intervention approach by randomly assigning schools to an intervention that strengthens relations among families within schools (Fiel, Shoji, and Gamoran 2015). The school-based intervention provides an exogenous "shock" to social capital that is independent of preexisting relationships among families and prior levels of student academic performance. If the intervention is successful and conditions of random assignment are met, social capital theory as expressed by Coleman (1988) predicts the intervention will positively affect schooling outcomes. Positive intervention effects would strengthen the credibility of social capital theory; null or negative effects would challenge the theory.

Field-based experiments to test theory are relatively rare in sociology, but they are increasingly common in economics (Duflo 2007). For example, Heckman's model of the importance of human capital investment in young children has been corroborated by experimental studies of high-quality early childhood education (Cunha and Heckman 2007). Likewise, the recent National Academies of Sciences, Engineering, and Medicine (2019) report on a "roadmap to reducing child poverty" relies substantially on field experiments to show that children who grow up in poverty suffer in their cognitive and social development and that tax and workforce interventions will reduce child poverty and improve short-term developmental outcomes and long-term economic results, consistent with theories from psychology and economics. In a similar vein, Banerjee and Duflo (2009) have crafted a renowned line of scholarship in development economics in which interventions put economic theory to the test.

Like human capital, social capital is a complex phenomenon that cannot be captured in a single intervention. But as economists do for human capital theory, sociologists can design an intervention whose effects can be predicted by social capital theory and then "make direct comparisons between observations and the predictions of a theory" (V. L. Smith 1989:153) to interpret the results. This is our aim in the present study.

We recognize that findings consistent with the predictions of social capital theory will not prove social capital theory. Our approach responds to an endemic problem of social capital research in education (the difficulty of distinguishing cause from effect), but it does not solve all problems. Notably, the complexity of any attempt to build social capital in the field leaves open the possibility that results may reflect other aspects of the intervention in addition to the components intended to build social capital. As a test of social capital theory, therefore, our approach is not as strong as if we could actually randomize social capital itself. Nonetheless, our study demonstrates the advantage of the intervention approach over prior work in the sociology of education. Prior research often represents social capital with proxy measures such as family structure, social mobility, and extracurricular participation, which are at least as removed from social capital (if not more so) and which may reflect unobserved correlates that predict both the social capital proxy and school performance. By contrast, our approach offers more confidence that we are studying social capital effects because (1) the intervention was randomized, (2) qualitative research at an early stage of the study documented how the intervention builds social capital in ways consistent with social capital theory (Shoji et al. 2014), and (3) the focal

intervention has demonstrated effects on widely accepted social capital indicators such as intergenerational closure and relationship quality among families (Rangel et al. 2020).

Conversely, a finding of null or negative intervention effects would align with empirical studies that raise doubts about whether social capital elevates achievement among young children (Condron 2009; Freeman and Condron 2011). Null or negative effect estimates, however, could be vulnerable to suppressor effects if negative intervention effects on achievement obscure positive social capital effects. It is therefore important to choose an intervention that not only builds social capital among families within the same school but also has a documented record of positively enhancing child development, a consideration we fulfilled in this study.

Following the logic of our intervention approach, our primary research question about social capital and achievement is an experimental one:

Research Question 1 (Experimental): Are third-grade test scores higher in schools that were randomly assigned to the socialcapital-building intervention in first grade?

A limitation of the intervention approach is that not all families participated in the intervention offered to their schools. We thus supplement our primary analyses based on experimental data with a set of quasi-experimental analyses that test whether actual participation in the intervention improves achievement outcomes. The quasiexperiment uses random assignment to the intervention as an instrument to test the effects of participation. Its ability to yield causal estimates rests on the assumption that effects of random assignment could occur only via participation in the intervention, a reasonable assumption in this case. A quasi-experimental approach allows us to test whether students who actually participated in the intervention, and thus were most likely to have their social capital boosted, had improved academic outcomes. Consequently, we pose a second, quasi-experimental question:

Research Question 2: (Quasi-experimental): Are third-grade test scores higher among students who actually participated in the social-capital-building intervention?

The Intervention: Families and Schools Together

To carry out the research, we selected a widely used family-school engagement program known as Families and Schools Together. Designed to build relationships among parents, between parents and school staff, and between parents and children, FAST places children in an ecological context in which they are embedded in supportive social networks. FAST has a record of successfully engaging disadvantaged parents (Caspe and Lopez 2006; McDonald et al. 2006). Distinctive features include (1) a leadership team of parents, school staff, and professionals who are demographically representative of the local community; (2) adaptation of the program to the local cultural context; and (3) active outreach to draw parents and children into FAST. FAST consists of eight weekly sessions led by a FAST leadership team in each school, followed by two years of monthly follow-up activities led by parents. In each city, a social service agency with past FAST experience implemented the program for this study, supported by the national FAST nonprofit organization.⁴ FAST trainers used the FAST Program Integrity Checklist in each site to ensure the program was implemented with fidelity (Shoji et al. 2014).

Elements of FAST. The eight FAST sessions occur in "hubs" of 5 to 10 families and last approximately 2.5 hours each week, usually beginning in the late afternoon and encompassing the dinner hour. Sessions are led by the trained leadership team, and all family members are invited to attend.5 The first activity of the first session consists of constructing a "Family Flag," in which children and parents talk about what symbols reflect their family and then explain those symbols to other families. For the next activity, which is also the first activity in succeeding weeks, families share a meal together. Directed by their parents, children serve the food and clean up after the meal. After the meal, family members play games with one another at their family tables as parents lead their children in taking turns, listening, and sharing feelings. This period also includes the "FAST Hello," when family members introduce themselves to other families, and communal singing of the "FAST Song." As Turley and colleagues (2017:209) noted, the first

activities of the evening are designed to strengthen relationships between parents and their own children, but they

also contribute to trust, shared expectations, and shared values across families and between families and school staff. Singing together, playing games, and sharing a meal comprise a positive, shared experience for families and school staff, providing a basis for the development of relationships and reducing anxiety about the school context. Because families take turns providing meals for the entire FAST hub, they develop reciprocity in their relationships with each other, which engenders trust across the social network.

After the family meal and games, activities move toward building relationships among families, supporting intergenerational closure and reciprocity in the school community. Children move to a separate area to play and complete homework while parents have "Buddy Time," a 15-minute period during which two parents from different families take turns talking with one another in a structured dialogue. The two parents each have seven uninterrupted minutes to speak about their day. Parents are asked to listen without giving advice or passing judgment, which helps build trust and shared values within new friendships. Then, parents gather for "Parent Group Time," joined by a leadership team member who participates as little as possible while parents are encouraged to take the lead in talking with one another. The aim of Parent Group Time is to create a peer network among parents to provide support for their children's success. This sets the stage for monthly meetings of parent-led activities following the end of FAST.

The next activity is "Special Play," which consists of 15 minutes of one-to-one parent-child interaction during which the child takes the lead in playing and the parents devote their full attention to the children. Parents are asked to follow four guidelines during Special Play: "(1) Don't boss. (2) Don't teach. (3) Don't judge. (4) Follow the child's lead." (Parents are also asked to repeat Special Play at home in between FAST sessions.) Finally, the entire group reassembles for closing activities that further strengthen relationships among families, including announcements about school and community activities, birthdays, and other special events and a final communal activity such as creating a "rain storm" by snapping, clapping, and foot-stomping. As recounted by Turley and colleagues (2017:210), "these closing activities are designed to reinforce togetherness through a routine and positive group experience."

FAST and social capital. Qualitative and quantitative evidence reported in previous articles based on these same data confirms that FAST triggered improvements in social capital in these school communities. In an interview-based qualitative study carried out in conjunction with the randomized trial, Shoji and colleagues (2014) identified four theory-based mechanisms through which FAST helped trigger social capital development among participating families. Two mechanisms were particularly salient in the Parent Time activity: responsive communication, in which parents reacted with interest or enthusiasm to other parents, and reciprocal communication, characterized by give-and-take among parents. These communication patterns helped establish a sense of bounded solidarity and reciprocity, conditions identified as supportive of social capital development (Portes 1998). Other mechanisms included the shared experience of attending FAST and engaging in its activities and rituals and institutional linkages fostered with the school and community partners via members of the FAST leadership team.

In the qualitative portion of a mixed-methods study, Rangel and colleagues (2020) found FAST not only helped parents get to know other parents more quickly than they would have otherwise but also enabled parents to determine one another's trustworthiness, express care and respect for one another, and demonstrate reciprocity through mutual support. They concluded that FAST affected not only the structure of parent networks but also the quality of relationships within the network.

Quantitative analyses of the randomized design have further established that in these same data, FAST strengthened ties among families (Gamoran et al. 2012; Shoji 2014; Turley et al. 2017). For example, intent-to-treat (ITT) estimates yielded significant FAST effect sizes of .21, .14, and .11 for measures of intergenerational closure, shared expectations, and reciprocity in relationships among parents, respectively (Turley et al. 2017). Complier average causal effect estimates for students who completed FAST were even larger, with effect sizes of .54, .40, and .36, respectively.⁶ Although FAST's effects on network structure (intergenerational closure) diminished over time, effects on the quality of relationships among parents (shared expectations and reciprocity) persisted through third grade: Rangel and colleagues (2020) found that for intergenerational closure, a first-grade ITT effect size of .26 declined to .14 by third grade, but an effect of .16 in relationship quality reached .19 by third grade.⁷

In these data, children in schools assigned to FAST exhibited fewer internalizing behavior problems during first grade after the intervention, and these effects were pronounced for students who completed the FAST program (Turley et al. 2017). The question for the present study is whether FAST, as an engine of social capital, affected academic outcomes in third grade. Earlier randomized trials indicated positive effects on atrisk children's behavioral and academic outcomes (Kratochwill et al. 2009, 2004; Layzer et al. 2001; McDonald et al. 2006). These prior studies, however, recruited only at-risk students, randomized at the student (or in one case, the classroom) level, and measured achievement using teachers' subjective judgments of students' academic performance. Ours was the first study to recruit students universally, randomize at the school level, and assess performance on standardized tests, design elements that were key to assessing FAST as a driver of social capital effects.⁸ For this article, FAST was implemented in first grade, and the follow-up ended when students were in third grade, the first year of statewide achievement testing.

Study Design

A total of 52 schools participated in the study, evenly divided between Phoenix, Arizona, and San Antonio, Texas. We selected these cities because they had social service agencies with experience implementing FAST and because they had many schools with high proportions of low-income Latino students. These contexts were of special interest given past research showing low-income Latino immigrant families' isolation from school communities (e.g., Valenzuela 1999). In Phoenix, three small districts participated, and in San Antonio, schools from one district were blocked into two groups based on proportions of students eligible for free and reduced-priced lunch. Each of these five units constituted blocks for randomization, and within each block, half the schools were assigned to treatment and half to control. Because the intervention was labor intensive, we further randomized schools to a two-year implementation cycle, with half the treatment schools initiating FAST in the 2008– 2009 academic year and half in 2009–2010.

A total of 3,084 families enrolled in the study, reflecting participation rates of about 60 percent in both treatment and control schools. Incomplete participation in the intervention may limit the generalizability of our findings, but it does not introduce bias into the treatment-control comparison. None of the families in control schools were exposed to FAST during our study period. Because assignment to FAST was by school, we have no reason to think a family in a control school was affected by the assignment of other families to FAST, and vice versa. Likewise, the assignment of schools to the treatment group did not affect resource allocation patterns or other potential differences between treatment and control schools within districts. Of the families that consented to the study in treatment schools, 73 percent attended FAST at least once, and the average family attended four sessions (Shoji et al. 2014). Data collection continued through 2010-2011 and 2011–2012, respectively, following children from first to third grade. All 52 schools and 70 percent (2,165 of 3,084) of the children from the original sample participated in the third-grade follow-up.

Measures. The dependent variables for this article consist of student scores on high-stakes tests of reading and mathematics in third grade.⁹ For each test, we use two measures. First, we obtained from districts students' scale scores, which we standardized-by state, test subject, and language in which the test was administered¹⁰—using the score mean and standard deviation of students in control schools. These scores capture the full range of student performance, but combining cases across states, years, and languages assumes the distribution of student scores is the same within each group. We have no reason to expect otherwise within states, but we are less certain of the homogeneity of distributions across Texas and Arizona.

Second, districts provided a binary indicator of whether students-based on their scale scoremet reading and math proficiency standards as defined by the state. Although these proficiency scores convey less information about the range of achievement, they have two advantages. First, the value of proficient is common across states, and because state tests in Arizona and Texas had similar relations to scores on the National Assessment of Educational Progress, proficiency levels in the two states represent similar absolute levels as well (Bandeira de Mello, Blankenship, and McLaughlin 2009). The proficiency measure provides insight into changes in achievement for students who are on the margins of meeting statewide standards. Second, the data on proficiency levels are more complete than the scale scores. Approximately 14 percent of scale scores provided by districts were outside the range of legitimate scale scores for a particular test and are thus likely misreported.

Control variables included student race/ethnicity, gender, free or reduced-price lunch status, English-language learner status, and special education status, all drawn from district records. Language use was indicated by whether parents chose to respond to baseline surveys in English or Spanish.

Measures of social capital are consistent with prior studies and are derived from parent surveys administered prior to treatment and approximately six weeks posttreatment. The structure of social relations is measured by intergenerational closure, using the familiar measure of how many parents of their children's friends at this school parents know (Coleman 1988). The indicator is coded 0 to 6 for responses of none to six or more. For some models, we dichotomize the variable as 0 for none or one and 1 for more than one parent known, based on the idea that being connected to at least a few parents may be very important but that connections to additional parents may have diminishing returns. Freeman and Condron (2011) also ran supplemental models with intergenerational closure coded dichotomously; they reported little difference in results. Measures of the quality of social relations include indicators of the extent to which parents at the school interacted with one another in areas such as babysitting, carpooling, and sharing meals together. Parents responded to six items measured on a 4-point scale from not at all to a lot; the items were summed, and then the

	Mean	SD	N
FAST	.50	_	2,165
Math standardized test score	02	1.02	I,867
Reading standardized test score	.02	1.03	1,854
Met proficiency level in math	.72	_	2,165
Met proficiency level in reading	.77	_	2,164
Race/ethnicity			
Hispanic	.78	_	2,140
Non-Hispanic white	.13	_	2,140
Black	.07	_	2,140
Asian, American Indian, or other race	.02	_	2,140
Female	.50	_	2,163
Eligible for free or reduced-price lunch	.78	_	2,140
English-language learner	.26	_	2,145
Special education student	.10	_	2,145
Parent completed baseline survey in Spanish	.28	_	2,165
Intergenerational closure (linear)	3.45	2.12	1,493
Knew two or more parents of child's friends (dichoto-	.78	_	1,493
mous measure of intergenerational closure)			
Reciprocity scale (standardized)	.02	1.01	1,509
Shared expectations			1,479
Not at all	.29	_	1,479
A little	.21	_	1,479
Some	.28	_	1,479
A lot	.21		1,479

Table 1. Student-Level Descriptive Statistics.

Note: Standard deviations are reported for continuous variables. All other variables are dichotomous. Social capital measures are collected in Year I, prior to treatment. FAST = Families and Schools Together.

measure was standardized to have a mean of 0 and standard deviation of 1. In addition, the data include a single-item indicator of shared expectations among parents, measured by responses to a scale from 1 to 4 on how much "other parents share your expectations for your child"; response options were none, a little, some, and a lot. Table 1 provides means and standard deviations of all variables. For some of our supplementary analyses, we also use factor analysis to create a composite parent-parent social capital scale that uses all these items.

Models. We weighed several options to best match the analytic approach to the study design. The preferred approach is one that makes the fewest assumptions: simply examine the experimental effect of FAST assignment on student achievement outcomes to answer the first research question of whether third-grade test scores are higher in schools assigned to the social-capital-building intervention. Prior analyses have established that FAST primarily improved the quality and quantity of parent-parent relationships, which is not surprising given that FAST programming focused heavily on these relationships. Hence, we first estimate ITT models that compare students who attended schools assigned to FAST to students in control schools. This approach minimizes potential problems of systematic selection among individuals into FAST participation as well as interference between treated and untreated units (in this case, schools). Because schools were the unit of assignment to treatment, we fit two-level mixed-effects linear regression models with students clustered by the schools they attended in first grade; for dichotomous outcomes, we substitute logit models. Our main analyses use standard listwise deletion approaches (Acock 2005; Allison 2001). We also show results with multiple imputation in which we impute the independent variables using the *mi impute* chained procedure in Stata (StataCorp 2019).

All models include randomization controls; results are shown with and without additional pretreatment controls. Comparison of treatment and control schools revealed no differences in school composition, size, or average prior achievement (Gamoran et al. 2012), but we did observe differences in pretreatment social capital measures favoring control schools (i.e., study enrollment rates were higher among disconnected families in the treatment group; see Appendix Table A1 in the online Supplemental Material and Turley et al. 2017). Controls for pretreatment social capital adjust for this discrepancy and improve precision in the estimates.

Individuals in treatment schools were encouraged to participate in the FAST intervention, but they did so to varying degrees. Hence, we also estimate treatment-on-treated (TOT) models that account for individual students' participation. One model considers the "dosage" (number of FAST sessions) students received, and another considers whether students "graduated" from FAST, where graduation is defined as attending six or more of the eight FAST sessions. We estimate the TOT models using an instrumental variables approach (two-stage least squares) wherein the endogenous variable (either FAST dosage or FAST graduation) is instrumented using FAST \times Randomization Block interactions (which are exogenous). This could help us detect treatment effects that might be suppressed by imperfect compliance and answers the second research question of whether third-grade test scores are higher among students who actually participated in the social-capitalbuilding intervention. Again, we fit these models with and without pretreatment controls.

An important question motivating this study was whether results from prior observational analyses of the effect of social capital on achievement might be upwardly biased by unobserved conditions that lead to both stronger ties among parents and higher test scores. Thus, we reanalyzed our data drawing on observational measures of social capital commonly used in nonexperimental studies to examine how those results compare to our experimental findings. To further explore this contrast, we briefly note findings from two quasi-experimental strategies that incorporate measures of social capital drawn from parent surveys. One applies a causal mediation analysis (Imai, Keele, and Tingley 2010) to estimate the portion of FAST ITT effects explained by a parent-parent social capital scale that combines the aforementioned parentparent social capital measures. The other uses school-level treatment assignment (FAST × Randomization Block interactions) as instrumental variables to estimate the effects of individual students' posttreatment parent-parent social capital, using the same parent-parent social capital scale. Both of these supplementary specifications include pretreatment social capital controls. These approaches entail more restrictive assumptions than the experimental approach. Causal mediation assumes sequential ignorability for the treatment (assignment is independent of mediator and outcome, a plausible assumption given random assignment) and for the mediator (a more challenging assumption that rests on inclusion of pretreatment controls for the mediator). The instrumental variables approach requires the exclusion restriction, that FAST effects operate only through the social capital indicator, a challenging assumption given the rich, multifaceted nature of FAST and social capital. Still, these alternative specifications can help us make sense of differences that may emerge between experimental, quasi-experimental, and observational results.

Finally, we conduct additional analyses to assess the sensitivity of our findings to modeling assumptions. First, we check the sensitivity of our main results to survey nonresponse by introducing weights designed to adjust the sample for nonresponse to the third-grade follow-up survey. Weights were calculated by predicting nonresponse using multilevel logistic regressions; the resulting models weight individual students by the inverse of their response propensity. Students who are missing item-level data on predictors included in the nonresponse model do not have nonresponse weights.

Second, we use a Heckman selection correction model to examine whether our results are robust to concerns about missing outcome data (Guo and Fraser 2010). Data for third-grade achievement are available only for students who reach third grade two years after first grade and are attending school in their original district. Any students who were retained in first or second grade would have missing data on third-grade achievement, yet retention may be endogenous to the intervention (Fiel et al. 2015). The selection model adjusts for missing third-grade test scores. Demographic variables and pretreatment social capital are used to predict achievement and selection. The proportion of students at the school level who made adequate yearly progress in math and

RESULTS

Table 2 provides the main results for achievement in mathematics and reading. Whether the dependent variable is measured as the standardized test score or proficiency level and whether we examine the ITT or TOT models, FAST exhibits no effects on test score outcomes. In the ITT analysis of math standardized scores, the effect size is -.01 without pretreatment controls and -.03 with pretreatment controls. The TOT effect sizes for math are -.01 and -.06 for the dosage models and .03 and -.06 for the graduation models, respectively. In reading, the effect sizes are .05 without pretreatment controls and .04 with pretreatment controls in the ITT models, and the TOT effect sizes are .14 and .06 for the dosage models and .20 and .07 for the graduation models, respectively. None of these coefficients are statistically significant. For the indicators of whether students met proficiency standards, the effects range from -.01 to .10, and again, none are significant.

Reanalysis using imputed data for missing pretreatment control variables yields virtually identical results, as seen in Appendix Table A2 in the online Supplemental Material. Moreover, interactions by race, ethnicity, gender, and free or reduced-price lunch status are also insignificant (not shown). Based on these results, our analyses indicate that FAST—and by implication, social capital—has no effect on third-grade achievement.

Contrast with Analysis Based on Observational Data

Past research on social capital effects on achievement, particularly effects of intergenerational closure, has yielded inconsistent results. In reviewing this literature, we detected a tendency for more rigorously controlled analyses to yield smaller or no effects, although the pattern was not universal. To consider this issue more fully, we reanalyzed our data using measures of social capital drawn from parent surveys instead of the randomly assigned FAST treatment. Table 3 reports these results. In contrast to our experimental findings, and yet consistent with some of the observational studies, this analysis yields statistically significant, positive social capital effects. Effects of the combined social capital scale are significant only in mathematics, with a small standardized effect of .07 with pretreatment controls included in the model. Effects of the reciprocity scale alone are similar. Effects of shared expectations are considerably larger, however, as are effects of intergenerational closure when specified as a dichotomous contrast between parents who know zero or one parent of their children's friends and parents who know two or more parents. Moreover, effects of shared expectations and intergenerational closure appear in both reading and math. Once again, the results are very similar for models that impute missing pretreatment control data (see Appendix Table A3 in the online Supplemental Material).

How are we to understand the positive effects of the survey social capital measures in contrast to the absence of experimental effects? The most likely explanation is that the survey measures are vulnerable to selection bias derived from unobserved predictors of both parent-reported social capital and student achievement. To further consider this explanation, we examine causal mediation analyses that place the survey measures as mediators of FAST ITT effects on standardized achievement. Table 4 displays the results, which yield no evidence of FAST effects operating through the composite parent-parent social capital scale. For example, with first-grade posttreatment social capital as the mediator and reading achievement as the outcome, the causal mediation analysis shows a total effect of .058, consisting of a direct effect of .054 and a mediation effect of .004, none of which are statistically significant. The coefficients are even smaller and likewise nonsignificant for mathematics. Similarly, as displayed in the second and third panels of Table 4, the analyses fail to detect mediation effects through the social capital composite when drawn from second- and third-grade parent surveys. All the mediation effects are close to zero and precisely measured, suggesting no positive social capital effects were suppressed by other intervention effects and raising further doubt about the causal role of social capital on third-grade achievement.11

We also fit models with FAST \times Randomization Block interactions as instruments for posttreatment social capital effects on achievement (and controls for pretreatment social capital) to leverage random assignment as an instrument for direct measures of social capital. However, these models are less informative because the standard

				- - -						-	-	
			Math Stand	Math Standardized Scores	ores			Σ	et Proficien	Met Proficiency Level in Math	Math	
	ΤI	Т	TOT E	TOT Dosage	TOT Graduation	lduation	LΤΙ	Ť	TOT Dosage	osage	TOT Graduation	aduation
FAST	01	03	01 181	06	.03 .77	06	01 10,-	03	.05	.04 (707)	60. 1017	.07
Pretreatment controls	No No	Yes	No No	Yes	No No	Yes	(^{02.)}	Yes	No No	Yes	No Vo	Yes
Z	1,867	1,706	1,712	I,559	1,712	I,559	2,165	I,963	I,953	1,772	1,953	1,772
		8	eading Star	Reading Standardized Scores	cores			Met	: Proficienc	Met Proficiency Level in Reading	Reading	
	TTI	F	TOT E	TOT Dosage	TOT Graduation	lduation	ΤΤΙ	F	TOT Dosage	osage	TOT Gr	TOT Graduation
FAST	.05	.04	<u>. 14</u>	90.	.20	.07	60.	01.	.06	.05	60.	80.
Pretreatment controls	(.08) No	(.07) Yes	(- 17) No	(.15) Yes	(.25) No	(.20) Yes	(.21) No	(.19) Yes	(07) No	(.06) Yes	(60.) No	(.09) Yes
Z	I,854	1,695	1,700	I,549	1,700	I,549	2,164	1,963	1,952	1,772	1,952	1,772
Note: Figures are regression coefficients, with standard errors in parentheses. All models include randomization controls. Pretreatment controls are race/ethnicity, sex, free or reduced-price lunch eligibility. English-language learner status, special education status, language of consent, and pretreatment social capital. ITT = intent to treat: TOT = treatment on	ר coefficients א English-lan	, with stand	lard errors ir er status, spe	n parentheses scial education	s. All models in status, langua	nclude randor ge of consent	mization cont t. and pretrea	trols. Pretrea	atment conti capital. ITT =	rols are race, = intent to tr	:/ethnicity, sex reat: TOT = tre	, free or eatment on

Achievement.
Reading
Math and
FAST on
Effects of
Experimental
Table 2.

treatment on to treat; IOI сарітаі. β eaulielli rt, and pretri language or conser Р status, spe reduced-price lunch eligibility, English-language learner treated; FAST = Families and Schools Together.

U		•		,				
		ith Irdized	Met Pro	oficiency el in	Rea Standa	0	Met Pro	
	Scanda			ith	Sco			ding
Model I: Social capital combined	.09*	.07*	.05	.00	.06	.03	.12	.10
scale	(.03)	(.03)	(.07)	(.08)	(.03)	(.03)	(.08)	(.09)
Model 2: Reciprocity scale	.07*	.05	.03	.00	.05	.03	.10	.08
	(.03)	(.03)	(.07)	(.07)	(.03)	(.03)	(.07)	(.08)
Model 3: Other parents share exp	oectation	าร						
A little	.20*	.21*	.43*	.49*	.13	.14	.46*	.53*
	(.08)	(.08)	(.19)	(.20)	(.08)	(.08)	(.20)	(.21)
Some	.14	.12	.19	.15	.15*	.12	.61*	.62*
	(.07)	(.07)	(.17)	(.18)	(.07)	(.07)	(.19)	(.20)
A lot	.26*	.21*	.39*́	.26	.25*	.20*	.49*	.48*
	(.08)	(.08)	(.19)	(.21)	(.08)	(.08)	(.20)	(.22)
Model 4: Intergenerational clo-	.03*	.03*	.06	.06	.01	.01	.07*	.07*
sure (linear)	(.01)	(.01)	(.03)	(.03)	(.01)	(.01)	(.03)	(.04)
Model 5: Intergenerational clo-	.18*	.18*	.28	.30	.13	ÌI.	.22	.19
sure (dichotomous)	(.07)	(.07)	(.16)	(.17)	(.07)	(.07)	(.17)	(.18)
Pretreatment controls	No	Yes	No	Yes	No	Yes	No	Yes
Ν	1,233	1,229	1,418	1,413	1,223	1,219	1,417	1,412

Table 3. Regressions of Math and Reading Achievement on Survey Measures of Social Capital.

Note: Each survey-based measure of social capital was entered in a separate two-level, mixed-effects regression model. Figures are regression coefficients, with standard errors in parentheses. All models include randomization controls. Pretreatment controls are race/ethnicity, sex, free or reduced-price lunch eligibility, English-language learner status, special education status, and language of consent.

*p < .05.

errors are large and the coefficients are not robust across social capital indicators and outcomes.¹² In any case, none of the instrumented social capital effects on achievement are statistically significant. Overall, the additional analyses fail to yield evidence that changes in social capital triggered by FAST improve children's test scores. Given experimental evidence of FAST effects on posttreatment social capital (Rangel et al. 2020; Turley et al. 2017), the results suggest social capital, at least as we have measured it, does not affect achievement. Rather, our findings suggest the observational estimates of social capital effects may be upwardly biased, and they reveal the value of the experimental approach to provide a more rigorous assessment of social capital theory.

Sensitivity Analyses

We conducted two additional analyses to assess the sensitivity of our results to sample specifications. First, we introduced weighting to adjust for sample nonresponse. Comparing Table 2 (main results) with Table 5 (weighted results), we observe no substantive differences. The weighted coefficients closely replicate the unweighted coefficients, and none are statistically significant. These comparisons suggest our main results are not biased by sample attrition. Second, we adjusted for selection to account for missing data on the dependent variable, which reflects students who were retained in grade and thus had no third-grade test scores as well as students who left the school districts. (Students who left the sample schools but remained in their districts were retained in the sample.) Here, too, the supplementary results lead to identical conclusions as the main results: The coefficients in Table 6 are all close to zero and are statistically insignificant.

DISCUSSION AND CONCLUSIONS

Since Coleman's (1988) foundational article on social capital and the creation of human capital, much attention and debate has focused on the effects of social capital on children's academic outcomes. This is particularly true with respect of Social Capital.

Table 4.	Causal Mediation	Effects of Families a	nd Schools Tog	ether on Achievem	ent via Survey Measures
of Social	Capital				

	First-G	rade (Posttreatn	nent) Social (Capital Comb	ined Scale as M	ediator
	Math	Standardized Sc	ores	Readin	g Standardized S	Scores
	Estimate	95% CI	þ Value	Estimate	95% CI	p Value
Mediation effect	.004	002, .014	.260	.004	002, .015	.228
Direct effect	028	203, .150	.732	.054	116, .222	.534
Total effect	024	202, .150	.770	.058	107, .225	.498
Proportion mediated	012	646, .860	.834	.025	650, .834	.574
N		1,142			1,134	

		Second-Grade S	ocial Capital	Combined Se	cale as Mediator	
	Math	Standardized Sc	ores	Readin	g Standardized S	Scores
	Estimate	95% CI	þ Value	Estimate	95% CI	þ Value
Mediation effect	.004	006, .019	.452	.007	003, .025	.210
Direct effect	.120	054, .288	.194	.125	068, .323	.210
Total effect	.124	051, .295	.178	.132	068, .333	.192
Proportion mediated	.016	253, .433	.558	.040	429, .483	.342
N		752			747	

		Third-Grade So	cial Capital (Combined Sc	ale as Mediator	
		Math			Reading	
	Estimate	95% CI	þ Value	Estimate	95% CI	p Value
Mediation effect	.003	006, .015	.498	.003	005, .015	.490
Direct effect	045	231, .127	.626	007	183, .176	.926
Total effect	042	227, .129	.644	004	179, .179	.950
Proportion mediated	006	–.657, .631	.854	.000	640, .607	.996
N		770			765	

Note: Analysis was conducted using the mediation package in R (Tingley et al. 2014). Models include randomization controls, pretreatment social capital, race/ethnicity, free or reduced-price lunch eligibility, English-language learner status, special education status, and language of consent. Cl = confidence interval.

to Coleman's notion of intergenerational closure, reflecting the benefits children may derive from their parents' connections with parents of their peers. The mixed empirical evidence has been enough to sustain many scholars' conviction in the benefits of this form of social capital while fueling skepticism among others. Much of the skepticism revolves around the difficulty of designing studies that can take into account differences among families that influence both their social capital and their children's outcomes (Durlauf 2002; Mouw 2006).

Our study adds to a growing literature challenging the notion that social capital boosts children's academic achievement with an experimental analysis that circumvents the potential selection bias plaguing observational studies. Given our novel approach to examining social capital effects, it is worth revisiting the logic. All first-grade students in some schools were randomly assigned to the FAST intervention, and students in other schools were assigned to a businessas-usual control condition. As shown through prior studies using the same data, FAST significantly enhances intergenerational closure and improves the quality of parents' relationships. If this sort of social capital improves children's academic achievement, then we would expect children in

ITT TOT FAST –.01 –.03 –.03 (.08) (.08) (.19) Pretreatment controls No Yes No N 1,815 1,706 1,663	TOT Dosage .0308 .(19) (.17) Vo Yes 663 1,559	TOT Graduation .0307 (.27) (.26 No Yes	luation						
		.03 (.27) No		LTI	L	TOT Dosage	osage	TOT Graduation	aduation
No Yes 1,815 1,706 1		No	07	.00	00.	.06	.05	II.	60.
1,815 1,706	_		Yes	e on N	(co.) Yes	No No	Yes	No No	Yes
		1,663	I,559	2,100	I,963	I,896	1,772	I,896	1,772
Reading Si	Reading Standardized Scores	cores			Met	Met Proficiency Level in Reading	r Level in F	teading	
111 101	TOT Dosage	TOT Graduation	luation	E	L	TOT Dosage	osage	TOT Gr	TOT Graduation
.03		.17	.07	00.	00.	.07	.07	Ξ.	Ξ.
(.17) (.08) (.17)	.17) (.15)	(.24)	(.21)	(.04)	(.03)	(.07)	(90.)	(.10)	(60.)
Yes		No	Yes	٩	Yes	No	Yes	٥N	Yes
N 1,802 1,695 1,651	651 1,549	1,651	I,549	2,099	1,963	I,895	1,772	1,895	1,772

Weights.	
Nonresponse	
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			Math Stand	lardized Score	es	
	TI	т	TOT D	Dosage	TOT Gr	aduation
FAST	08 (.13)	05 (.09)	07 (.31)	05 (.20)	.03 (.48)	01 (.29)
Pretreatment controls N	No 1,723	Yes 1,706	No 1,723	Yes 1,706	No 1,723	Yes 1,706
			Reading Star	ndardized Sco	res	
	ТІ	Т	TOT D	Dosage	TOT Gr	aduation
FAST	02 (.10)	02 (.08)	.03 (.24)	02 (.18)	.12 (.37)	.00 (.27)
Pretreatment controls N	No 1,712	Yes (1,695	No 1,712	Yes ´ 1,695	No 1,712	Yes 1,695

Table 6. Effects of FAST on Math and Reading Achievement with Selection C	Correction.
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Note: Figures are regression coefficients, with standard errors in parentheses. Pretreatment controls are race/ethnicity, sex, free or reduced-price lunch eligibility, English-language learner status, special education status, language of consent, and pretreatment social capital. For the TOT models, FAST attendance and graduation are predicted in a prior model using FAST \times Randomization Block interactions; predicted values from the prior model are used in these models. Selection equation estimates (not shown) are available on request. FAST = Families and Schools Together; ITT = intent-to-treat; TOT = treatment-on-treated.

the FAST schools to fare better on achievement tests in third grade. They did not. This is the case for both math and reading tests, and the findings are robust to a variety of sensitivity analyses. By failing to confirm the expectations of theory, this study adds to growing skepticism about whether social capital is a potent strategy for boosting school success among young children.

Our results are consistent with other recent studies that question the effects of social capital on achievement. We noted at the outset that more rigorously controlled studies tend to yield smaller effects of social capital on achievement. In well-controlled studies, the effect of intergenerational closure or similar constructs on achievement generally ranges from 0 to about .2 in effect-size units (Carbonaro 1998; Condron 2009; Dufur et al. 2013; Freeman and Condron 2011; Geven and van de Werfhorst 2020; Kao and Rutherford 2007; Morgan and Todd 2009),¹³ and the observational analyses we report in Table 3 mostly fall within this range (larger for shared expectations). With FAST exhibiting ITT effects on social capital measures of about .11 to .26 and TOT effects as high as .54 (Rangel et al. 2020; Turley et al. 2017), we were positioned to find significant effects of FAST on achievement. This seemed especially likely because based on

qualitative studies, FAST builds social capital among parents in ways that are not necessarily captured by survey measures, such as responsive and reciprocal communication, bounded solidarity, and the shared experience of participating (Shoji et al. 2014). The opportunity to capture social capital formation more fully and to eliminate selection bias through random assignment are strengths of our intervention approach. A limitation is that questions about our interpretation of intervention effects as social capital effects would be unavoidable had we found positive effects of FAST on achievement because, as we noted earlier, FAST might operate in other ways in addition to building social capital. But in the absence of positive effects and with no basis to expect suppressor effects, interpreting our results as inconsistent with the predictions of social capital theory appears to be our most logical conclusion.

Reconciling Our Results with Previous Studies of FAST

It is important to reconcile apparent inconsistencies between this study and prior studies of FAST, including evidence from the same sample that FAST reduces children's behavioral problems

(Turley et al. 2017). Despite recent attempts to link early behavioral outcomes with later academic results, it is plausible that social capital affects children's behavior but not their test scores. For example, McNeal (1999) and Domina (2005) examined a variety of parent-child, parentschool, and parent-parent connections and argued that the social control generated when parents are networked with the parents of their children's friends is likely to affect behavior, whereas only connections that involve school staff are likely to affect both behavior and achievement. Perhaps this pattern is at play in our findings. Although earlier studies of FAST in other cities reported academic benefits, the outcome measures were teacher-reported judgments of students' academic performance, which may reflect behavior as well as academic skills (McDonald et al. 2006). A recent UK-based study found social-behavioral but not achievement effects of FAST, mirroring our results (Lord et al. 2018).

Our findings are a reminder that improving social-behavioral skills may not necessarily lead to higher cognitive performance, at least not in the medium term and not among young children. In a definitive study of six large-scale databases, Duncan and colleagues (2007) showed that test scores among elementary students are not well predicted by social-behavioral skills in early childhood. Notably, the Early Childhood Longitudinal Study revealed zero effects of internalizing and externalizing behavior problems in kindergarten on reading and mathematics scores in third grade. In another, more recent study, an intervention known to elevate social and emotional learning in second grade yielded no effect on standardized test scores in third through fifth grades (Hart et al. 2020). Hence, the inconsistent findings for FAST effects-with positive effects on behavioral outcomes and no effects on achievement scores two years later-may actually align with other research.14

Implications for Future Research on Social Capital and Achievement

Our findings point toward several new directions for research on social capital and school success. First, relationships among parents in children's peer networks are but one dimension of social capital, and it may still be the case that improving relationships between children and their parents or teachers or between parents and school personnel may affect children's academic outcomes. Although we examined measures related to other forms of social capital, the FAST intervention had weak effects on them (Shoji 2014). It is therefore possible that the type of social capital engendered by FAST is not the sort that boosts achievement and that the sort that elevates test scores was not triggered by FAST. Raising doubts about achievement effects of the types of social capital instigated by FAST-relationships of trust and shared expectations in closed networks of parents and children-is significant because of its centrality to sociological theories of social capital. However, it does not close the book on understanding the relation between social capital and school performance.

Second, a possible explanation for why the interfamily networks strengthened by FAST did not improve children's academic achievement is that-although these networks may serve an important norm-enforcing function-they may not provide access to the resources necessary to alter children's success in school. The FAST intervention was fielded in schools with large enrollments of students from low-income families. Intergenerational closure has been characterized in some research as a "middle-class phenomenon" that mostly benefits children from advantaged families whose parents can draw on one another's resource-rich networks (Horvat et al. 2003). Resources inhering in the parental networks of lower-income families may have less potential to boost children's academic performance, a notion supported by research suggesting intergenerational closure boosts children's achievement in low-poverty schools but not in high-poverty schools (Fasang, Mangino, and Brückner 2014). Within diverse school settings, recent research points to the importance of having parent-parent networks that bridge social divisions, in addition to strong bonds forged within social groups, to generate equitable allocation of resources within schools (Murray et al. 2020). Commonly used quantitative measures of intergenerational closure, including the one used in this study, do not shed light on the diversity of parent-parent networks or on the extent to which these networks may bridge social lines. Future studies can improve on these measures.

Third, our findings for young children may not generalize to adolescents. On the one hand, parents' connections to the parents of their children's friends may be especially powerful when children are young and parents are better positioned to make use of information and exert social control, so the absence of achievement effects for young children is telling. On the other hand, precisely because peer networks become increasingly salient as children reach adolescence, intergenerational closure may be especially important in facilitating parents' access to information and social control over teenagers, and our evidence cannot speak to that possibility. This notion points to the need for further research on social capital with adolescents using designs that offer stronger causal warrant than the middle and high school survey analyses of the past.

It is reasonable to ask whether experiments such as this one provide useful information about the processes by which social capital operates naturally. However, if social capital can be a lever for change to improve children's lives, then it will likely be through interventions similar to the FAST program that such social engineering occurs. Over the past decades, countless interventions have been developed to strengthen relationships between parents in a community, to improve parent-school relationships, and to otherwise enhance parents' involvement in their children's education (T. E. Smith et al. 2020). These interventions differ in name, specific activities, dosage, and other features but share common underlying theoretical mechanisms. To advance science and to meet real-world challenges, researchers must identify the proposed underlying mechanism, such as social capital, and examine in the field whether the mechanism functions as predicted across interventions.

Indeed, new studies are essential because the problem that social capital formation seeks to address is more urgent than ever. As the events of 2020 clearly demonstrated, inequality faced by racial and ethnic minority and low-income communities persists at a staggering level with deep structural roots. This study took place in public schools with substantial enrollments of lowincome Hispanic students within two southwestern cities with large immigrant communities. Scholars have documented the social disconnect between many Hispanic families and their school communities and speculated that such isolation contributes to their children's educational struggles (Stanton-Salazar and Dornbusch 1995; Valenzuela 1999). This social isolation can be compounded by restrictive immigration policies that instill fear and distrust among immigrant families (Suárez-Orozco et al. 2011); such sentiments were documented for some of the immigrant families involved in the present study (Valdez, Padilla, and Valentine 2013). The question of whether school-based programs can build social capital and whether such social capital can help children from marginalized communities succeed academically remains pressing.

RESEARCH ETHICS

All human subjects research conducted for this article was approved by an institutional review board. The research was performed in a way that is consistent with the ethical standards laid out in the 1964 Declaration of Helsinki and Section 12 of the American Sociological Association's Code of Ethics. All human subjects gave their informed consent prior to their participation in the research (children's parents, in the case of 5- to 8-year-olds), and adequate steps were taken to protect participants' confidentiality. An earlier version of this article was presented at the annual meeting of the American Educational Research Association, Washington, DC, April 2016.

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DATA ACCESSIBILITY STATEMENT

De-identified data for this project may be obtained from the Inter-university Consortium for Social and Political Research (ICPSR 35481).

SUPPLEMENTAL MATERIAL

Supplemental material is available in the online version of the journal.

NOTES

- 1. *Intergenerational closure* refers to closure in a network composed of children and parents from the same families. Although commonly measured in education research as closed networks among families within the same school, intergenerational closure does not assume closure among all families in the same class or school.
- 2. All but one of these studies examined intergenerational closure, and most used the National Educational Longitudinal Study indicator of number of parents of children's friends known as reported by parent respondents. In the survey from Germany and the Netherlands analyzed by Geven and van de Werfhorst (2020), intergenerational networks were measured with a full sociometric map that asked students which classmates their parents know and with whose parents their own parents get together, divided by the sum of possible nominations. An exception is the Program for International Student Assessment study by Rodriguez Menés and Donato (2015), which used a constellation of indicators related to social cohesion among families and students.
- 3. The indicator of intergenerational closure was slightly different in the Early Childhood Longitudinal Study–Kindergarten cohort, asking parents to report the number of parents of their children's classmates they talk to regularly by phone or in person.
- Information on the Families and Schools Together (FAST) national organization is available at https://www.familiesandschools.org/.
- Descriptions of FAST activities have been provided by McDonald and Frey (1999), and the research basis for each FAST activity was laid out by McDonald and colleagues (1997) and McDonald (2002).
- 6. The complier average causal effect model uses data from the treatment group to simulate which members of the control group would have complied had they had the opportunity to do so. Effect estimates are based on comparisons of compliers in the treatment group to would-be compliers in the control group (Bloom 1984; Turley et al. 2017).
- Rangel, Shoji, and Gamoran (2020) reported metric coefficients; for comparability, we converted these to effect sizes by dividing the metric coefficients by their respective pooled standard deviations.
- Two subsequent studies have followed our design of universal recruitment and school-level randomization with mixed results. A UK-based study of Key Stage 1 (ages 5–7) pupils reported positive effects on socialbehavioral outcomes but no effects on achievement (Lord et al. 2018). A study of kindergarten students

in Philadelphia yielded no experimental effects on first-grade test scores but a positive quasi-experimental effect on reading in a small sample of students who complied with the treatment compared to a matched sample of controls (Bos et al. 2018).

- Arizona's statewide standardized test in 2011 and 2012 was Arizona's Instrument to Measure Standards. The statewide test administered in Texas changed between 2011 (Texas Assessment of Knowledge and Skills) and 2012 (State of Texas Assessments of Academic Readiness).
- Arizona's standardized test was administered in English only. Texas administered an English and a Spanish version of its statewide tests.
- 11. These results contrast with causal mediation effects on first-grade child behavior, which reveal statistically significant mediation effects for FAST effects on behavioral outcomes via a social capital composite (Turley et al. 2017:Note 4).
- 12. Model diagnostics indicate that FAST is an adequate instrument for social capital, with F statistics of 11.66 and 15.446 for the social capital composite in math and reading, respectively, and 34.49 and 29.19 for intergenerational closure on its own. None of these analyses yield significant effects on achievement.
- 13. Fasang, Mangino, and Brückner's (2014) results translate to an effect size of .43 for the effect of intergenerational closure on high school grades, but only in low-poverty schools; the effect in high-poverty schools was actually negative. (In this case and others in which coefficients were reported in their original metrics, we converted them to effect-size coefficients for comparability.)
- 14. Of course, social-behavioral skills gained in early childhood may be vital for long-term life chances even if they do not affect cognitive skills in the short term (Deming 2017; Kautz et al. 2014).

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