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Do Boys and Girls Use Computers Differently, and Does it Contribute to why Boys Do Worse in School than Girls?

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Robert W. Fairlie

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Abstract

Boys are doing worse in school than are girls, which has been dubbed “the Boy Crisis.” An analysis of the latest data on educational outcomes among boys and girls reveals extensive disparities in grades, reading and writing test scores, and other measurable educational outcomes, and these disparities exist across family resources and race. Focusing on disadvantaged schoolchildren, I then examine whether time investments made by boys and girls related to computer use contribute to the gender gap in academic achievement. Data from several sources indicate that boys are less likely to use computers for schoolwork and are more likely to use computers for playing games, but are less likely to use computers for social networking and email than are girls. Using data from a large field experiment randomly providing free personal computers to schoolchildren for home use, I also test whether these differential patterns of computer use displace homework time and ultimately translate into worse educational outcomes among boys. No evidence is found indicating that personal computers crowd out homework time and effort for disadvantaged boys relative to girls. Home computers also do not have negative effects on educational outcomes such as grades, test scores, courses completed, and tardies for disadvantaged boys relative to girls.

JEL-Code: C930, I240, J160.

Keywords: technology, computers, ICT, education, gender, field experiment.

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Do Boys and Girls Use Computers Differently, and Does It Contribute to Why Boys do Worse in School than Girls?

1. Introduction

Boys do worse in school than girls. They obtain lower grades, and are less likely to graduate from high school and attend college (NCES 2012).¹ These gender disparities in academic performance exist for minority and low-income schoolchildren as well as for more advantaged schoolchildren. One factor that might contribute to why boys and girls differ in academic performance is that they make different time investments after school.² For example, boys might spend more time playing video games, "playing around" on computers, watching TV, and using other forms of media than girls. All of these activities might crowd out time spent doing homework and studying for exams. A recent national time-use diary survey found that children consume 7 ½ hours of media a day and this level of use is 20 percent higher than it was only five years ago (Kaiser Family Foundation 2010). The average length of a school day in the United States is roughly 6 ½ hours (NCES 2012). These time investments made at young ages between educational and non-educational activities might have long lasting effects on educational attainment. Surprisingly, very little research has focused on the time investments made by children and their consequences for educational outcomes.

¹ The media has dubbed these disparities as the "Boy Crisis." See "The Boy Crisis. At Every Level of Education, They're Falling Behind. What to Do?" Newsweek (Jan. 30, 2006), "Raising Cain: Boys in Focus," PBS (Jan. 12, 2006), "The Boys Have Fallen Behind," NY Times (March 27, 2010), and "The Boys at the Back," NY Times (February 2, 2013) for example.

² The reversal of the gender gap in college education with women now earning more college degrees than men has been well documented (see Sundstrom 2004; Goldin, Katz, and Kuziemko 2006 for example), but disparities between boys and girls in academic performance have drawn much less research attention. The underlying causes of these disparities are not well known. Some of the potential explanations examined in recent studies include a disproportionate representation of female teachers in younger grades, girls are more self-disciplined, girls respond more to pre-school interventions, girls are more ready to learn, differential treatment by teachers, and larger positive impacts for girls by Teach for America teachers (Dee 2007; Duckworth and Seligman 2006; Anderson 2008; Cornwell, Mustard and Van Parys 2013; Malamud and Schanzenbach 2007; Antecol, Eren and Ozbeklik 2013).

Of particular concern are the potential consequences of computer use among boys and girls. Computer use is one of the largest types of media use among children (Kaiser Family Foundation 2010), and thus extensive use of computers after school for video games, social networking and other entertainment activities might crowd out homework and study time among schoolchildren. There is evidence in the previous literature of computer use crowding out schoolwork and negative effects on academic performance. For example, Malumed and Pop-Eleches (2011) find evidence of heavy game use of computers and negative effects of computers on reading, homework and grades, and Fuchs and Woessmann (2004) find a negative relationship between home computers and math and reading test scores, possibly due to the distracting effects of children "playing computer games."³ If boys have higher levels of access to home computers, use computers more for playing games, or use computers less for schoolwork than girls, computers may partly contribute to why boys do worse in school. A better understanding of these potential effects is especially important for low-income and minority schoolchildren because of the policy focus on expanding access to technology to reduce the digital divide.⁴

In this paper, I first examine the latest national data on academic performance among girls and boys focusing on disadvantaged schoolchildren. I then explore three hypotheses

³ Concerns over the negative effects of home computers have gained a fair amount of attention recently in the press. See, for example, "Computers at Home: Educational Hope vs. Teenage Reality," NY Times, July 10, 2010 and "Wasting Time Is New Divide in Digital Era," NY Times, May 29, 2012. Extensive use of social networking sites, such as Facebook, is one particular concern (e.g., see Karpinski 2009; Pasek and Hargittai 2009). These concerns are similar to those over television (e.g., see Zavodny 2006).

⁴ The U.S. federal government spends more than \$2 billion per year on the E-rate program, which provides discounts to schools and libraries for the costs of telecommunications services and equipment (Puma, et al. 2000, Universal Services Administration Company 2013). England provided free computers to nearly 300,000 low-income families with children at a total cost of £194 million through the Home Access Programme. Additional policies include tax breaks, special loans and Individual Development Accounts (IDA) for educational purchases of computers, community technology centers, and laptop checkout programs for students (Servon 2002, Lazarus 2006, and Gordo 2008).

regarding computer use and these differences. First, I examine whether girls and boys have differential access rates to personal computers at home. Using microdata from the Computer and Internet Supplement to the Current Population Survey (CPS), I examine whether computer access rates vary across boy-only, girl-only and boy-girl families, and whether there are gender differences for low-income and minority schoolchildren. Second, I explore whether disadvantaged boys and girls use computers differently. Are boys more likely to use computers for video games and other non-educational activities, and are they less likely to use computers for schoolwork? Although a few previous studies examine gender differences in computer and Internet use among adults, very little is known about gender differences in the use of computers among children.⁵ Using data from three sources, I conduct the first detailed examination of computer and Internet use among boys and girls. Computer use for game playing, social networking, schoolwork and other activities is examined.

Third, I explore whether boy-girl differences in computer use crowd out homework time and effort differently, and contribute to gender disparities in educational outcomes among disadvantaged schoolchildren. To test this hypothesis I estimate the effects of home computers on homework time and effort, grades, standardized test scores, and several additional educational outcomes for girls and boys. Given similar access rates, if home computers have a larger negative impact on educational outcomes for boys than for girls then differential computer use at home widens the achievement gap. If instead, home computers have a similar effect for boys and girls then differential computer use does not contribute to the achievement gap. To remove

⁵ Men and women are found to have very similar levels of access to computers and the Internet, but differ in intensity of use and activities, and possibly benefits (see Ono and Zavodny 2003; Hargittai 2007; Mossberger 2008; NTIA 2011; Figlio, et al. 2012 for example). Differences in access and use, however, are substantially larger by race and income (see Hoffman and Novak 1998; Mossberger, Tolbert, and Stansbury 2003; Mossberger, Tolbert, and Gilbert 2006; Ono and Zavodny 2007; Fairlie 2004; Goldfarb and Prince 2008 for example).

concerns about selection bias resulting from which families decide to purchase computers I use data from the largest-ever randomized control experiment providing free personal computers to U.S. schoolchildren for home use.⁶ Half of over one thousand schoolchildren grades 6-10 attending 15 different schools were randomly selected to receive computers to use at home. Previous findings for all schoolchildren participating in the field experiment indicate that the randomly selected group of students receiving free computers experienced no improvement in educational outcomes relative to the control group that did not receive free computers (Fairlie and Robinson 2013). Fairlie and Robinson (2013), however, does not explore whether boys and girls use computers differently and whether these differences contribute to the gender gap in academic performance among disadvantaged schoolchildren.

Briefly previewing the results, I find that girls outperform boys not only in grades, test scores in reading and writing, and high school graduation rates, but also in numerous other educational outcomes. The results show a remarkably consistent underperformance of boys relative to girls in school, which holds across race and family resources. Using microdata from the CPS computer supplement I find that boys and girls have very similar rates of access to home computers overall and by race and income even though disparities across these groups are large. Boys and girls use computers differently, however, with boys using computers more for video games and girls using computers more for schoolwork, email and social networking, which does not differ substantially by race or income. Estimates from the random experiment, however, do not provide evidence that computers crowd out homework time and effort for disadvantaged

⁶ If computers were exogenously assigned to children the question could be explored by simply comparing the girl-boy gap in educational outcomes among existing computer owners to the girl-boy gap in educational outcomes among existing non-computer owners. But, parents make decisions about computer purchases partly based on concerns about the non-educational uses of computers by children and partly based on the perceived educational benefits of computers to children (which might differ between boys and girls) raising concerns about selection bias.

boys relative to girls, or that home computers have negative effects on grades, test scores, and other educational outcomes for boys relative to girls. Although parents, schools, and policymakers may have other concerns about how boys and girls use computers, these patterns do not appear to contribute to why disadvantaged boys do worse in school than girls.

The remainder of the paper is organized as follows. In Section 2, I describe the data sources used to examine gender differences in computer access and use. I also describe the experiment used to test for gender differences in the impacts of computers on educational outcomes. Section 3 presents estimates of gender differences in academic performance. Section 4 presents estimates of gender differences in computer access and use. Section 5 presents the experimental results for the impacts of home computers on homework time and effort and educational outcomes for boys and girls. Section 5 concludes.

2. Data and Methods

Data on Computer Use

Data from three national sources are used to examine whether boys and girls use computers differently. I use data from the Current Population Survey Computer and Internet Use Supplements conducted by the U.S. Bureau of Labor Statistics and Census Bureau, a time use diary study of the use of technology by children conducted by the Kaiser Family Foundation, and surveys of teenagers conducted as part of the Pew Internet and American Life Project. The combination of data from these national sources provides the first comprehensive examination of gender differences in computer use among children.

The Internet and Computer Use Supplement to the Current Population Survey (CPS), conducted by the U.S. Census Bureau and the Bureau of Labor Statistics, is representative of the

entire U.S. population and interviews approximately 50,000 households and 130,000 individuals. The Internet and Computer Use Supplement to the CPS is the primary source of information on technology use collected by the federal government and has been conducted over the past three decades at irregular intervals. The information gathered differs in each survey. Estimates from the 2003 supplement include the latest information on computer use activities among children. The estimates reported later in Table 4 are from these CPS data. The 2011 Supplement is the latest available data, but does not allow one to examine activities of computer use among children (only adult householders). Because of the lack of published results from the survey, I use microdata from the 2011 CPS Supplement to examine overall access and use rates among boys and girls ages 5-17 (N=23,594). The microdata also allow for a more detailed examination of home computer access rates among girls and boys by child and family characteristics such as the gender composition of the household, race, income, and age.

The Kaiser Family Foundation surveyed 2,002 children ages 8-18 across the country in 2008 and 2009 on their use of media including detailed information on computer use (Kaiser Family Foundation 2010). Similar surveys were conducted in 1999 and 2004 by the Kaiser Family Foundation. Information on computer use activities is also reported from national surveys conducted as part of the Pew Internet and American Life Project (see Pew Internet Project 2008a, 2008b). Surveys on numerous topics related to computer, Internet and media use are conducted regularly by the Pew Research Center. The Pew Internet Project (2008a, 2008b) studies include nationally-representative samples of 1,102 children ages 12-17 in 2007 and 2008 and 700 children ages 12-17 in 2007, respectively. To our knowledge, these three sources of data represent all of the nationally representative sources of data providing detailed information on computer use among children.

Randomized Control Experiment

To explore the effects of personal computers on crowding out homework time and educational outcomes among disadvantaged boys and girls and whether these effects differ by gender I use data from a field experiment that provides free personal computers to schoolchildren for home use. The randomized control experiment involved 1,123 students in grades 6-10 attending 15 schools across California (see Fairlie and Robinson 2013 for more details). It represents the first field experiment involving the provision of free computers to schoolchildren for home use ever conducted, and the largest experiment involving the provision of free home computers to U.S. students at any level. The randomized control experiment removes concerns about selection bias resulting from which families decide to purchase computers. All of the students participating in the study did not have computers at baseline. Half were randomly selected to receive free computers, while the other half served as the control group. Outcomes were tracked for all participating students over an academic year.

The sample for this study includes 1,123 students enrolled in grades 6-10 in 15 different middle and high schools in 5 school districts in California. The project took place over two years: two schools participated in 2008-9, twelve schools participated in 2009-10, and one school participated in both years. The 15 schools in the study span the Central Valley of California geographically. Overall, these schools are similar in size (749 students compared to 781 students), student to teacher ratio (20.4 to 22.6), and female to male student ratio (1.02 to 1.05) as California schools as a whole (U.S. Department of Education 2011). Schools in the experiment, however, are poorer (81% free or reduced price lunch compared with 57%) and have a higher percentage of minority students (82% to 73%) than the California average reflecting the

requirement of not having a home computer for eligibility in the experiment. Participating students also have lower average test scores than the California average (3.2 compared with 3.6 in English-Language Arts and 3.1 compared with 3.3 in Math), but the differences are not large (California Department of Education 2010).

To identify children who did not already have home computers, we conducted an in-class survey at the beginning of the school year with all of the students in the 15 participating schools. The survey, which took only a few minutes to complete, asked basic questions about home computer ownership and usage. To encourage honest responses, it was not announced to students that the survey would be used to determine eligibility for a free home computer (even most teachers did not know the purpose of the survey). In total, 7,337 students completed in-class surveys, with 24 percent reporting not having a computer at home. This rate of home computer ownership is roughly comparable to the national average: – estimates from the 2010 Current Population Survey indicate that 27% of children aged 10-17 do not have a computer with Internet access at home (U.S. Department of Education 2011).

Any student who reported not having a home computer on an in-class survey was eligible for the study.⁷ All eligible students were given an informational packet, baseline survey, and consent form to complete at home. To participate, children had to have their parents sign the consent form (which, in addition to participating in the study, released future grade, test score and administrative data) and return the completed survey to the school. Of the 1,636 students eligible for the study, we received 1,123 responses with valid consent forms and completed

⁷ Because eligibility for the study is based on not having a computer at home, our estimates capture the impact of computers on the educational outcomes of schoolchildren whose parents do not buy them on their own and do not necessarily capture the impact of computers for existing computer owners. Schoolchildren without home computers, however, are the population of interest in considering policies to expand access.

questionnaires (68.6%). We randomized treatment at the individual level, stratified by school. In total, of the 1,123 participants, 559 were randomly assigned to the treatment group. For boys, there were 555 participants with 280 assigned to the treatment group, and for girls there were 568 participants with 279 assigned to the treatment group.

The computers provided through the experiment were purchased from or donated by Computers for Classrooms, Inc., a Microsoft-certified computer refurbisher located in Chico, California. The computers were refurbished Pentium machines with 17" monitors, modems, ethernet cards, CD drives, flash drives, Microsoft Windows, and Microsoft Office (Word, Excel, PowerPoint, Outlook). The computer came with a 1-year warranty on hardware and software during which Computers for Classrooms offered to replace any computer not functioning properly. In total, the retail value of the machines was approximately \$400-500 a unit. Since the focus of the project was to estimate the impacts of home computers on educational outcomes and not to evaluate a more intensive technology policy intervention, no training or assistance was provided with the computers. We did not provide Internet service as part of the experiment and found that about half of the students receiving computers subscribed to service.

The computers were handed out by the schools to eligible students in the late fall of the school year. Almost all of the students sampled for computers received them: we received reports of only 11 children who did not pick up their computers, and 7 of these had dropped out of their school by that time. As expected, we found that some of the control group students purchased home computers by the end of the school year. From a follow-up survey conducted at the end of the school year, we found that 27 percent of girls and 25 percent of boys in the control group purchased computers, and in most cases these computers were purchased later in the school year (thus having less potential impacts on measured outcomes). After the distribution,

neither the research team nor Computers for Classrooms had any contact with students during the school year. In addition, many of the outcomes were collected at least 6 months after the computers were given out (for example, end-of-year standardized test scores and fourth quarter grades). Thus, it is very unlikely that student behavior would have changed for any reason other than the computers themselves (for instance, via Hawthorne effects).

Data from the experiment were collected from four main sources. First, we administered a detailed baseline survey which was required to participate in the project (as that was where consent was obtained). The survey includes detailed information on student and household characteristics. Second, we administered a follow-up survey at the end of the school year, which included detailed questions about computer ownership, homework time, and homework effort allowing for a comparison of first-stage and homework crowd-out effects. The response rates for the follow-up survey were high: 78.2 percent for boys and 76.6 percent for girls.⁸ Third, each school provided us with detailed administrative data on educational outcomes for all students covering the entire academic year. These administrative data include grades in all courses taken and disciplinary information. Finally, schools provided us with standardized test scores from the California Standardized Testing and Reporting (STAR) program. A major advantage of the administrative data on test scores as well as grades and other outcomes is that they are measured without any measurement error, and attrition is virtually non-existent. The collection of these datasets provides an extensive amount of information on computer ownership and educational outcomes.

Randomization and Implementation Checks

⁸ The response rates are 76.7 percent for the control group and 79.6 percent for the treatment group for boys, and 75.4 percent for the control group and 77.8 percent for the treatment group for girls.

Appendix Table 1 reports summary statistics for the treatment and control groups and provides a balance check. Balance checks are reported for the total sample, the girl sample, and the boy sample. For each sample, means for the treatment and control groups and the p-value for a t-test of equality are reported. Overall, there is very little difference between the treatment and control groups in all three samples. The only variable with a difference that is statistically significant is that treatment children are less likely to live with their mother in the boy sample (although the difference of 0.06 is small relative to the base of 0.93). It is likely that this one difference is caused by random chance given the large number of comparisons being made – nevertheless, all of these covariates are controlled for in the regressions that follow.

As a check of the experimental implementation, I also examine whether there is a large relative increase in reported computer ownership and whether the effect is similar for boys and girls. Information on computer ownership is obtained from the follow-up survey conducted at the end of the school year. For girls, I find that 82% of the treatment group and 27% of the control group report having a computer at follow-up. For boys, the overall levels are similar and the treatment-control difference is identical, with 80% of the treatment group and 25% of the control group reporting having a computer at follow-up. While these treatment-control differences of 55 percentage points are very large, if anything they are understated because only a very small fraction of the 559 students in the treatment group did not receive one (as noted above, we had reports of only 11 students who did not pick up their computer). In fact, I find that one quarter of the boy and girl treatment groups report positive hours of computer use at home even though they report in a previous question that they do not have a computer at home and are supposed to skip the question. In addition, any measurement error in computer ownership would understate differences in reported ownership.

The follow-up survey also asked children a battery of questions about how much they use computers for five different types of activities at home, school, and other locations. With the resulting 15 different questions of types of use, the self-reported hours responses are noisy with many missing values and some inconsistencies in reporting. Appendix Table 2 reports the average value of responses to these questions. Overall, computer use increased for both boys and girls. This includes separate use for schoolwork, email, games, and social networking. Unfortunately, the estimates are not precise enough to identify girl-boy differences in use. With so many different categories to report hours of use there is likely to be a fair amount of measurement error in these estimates, and thus I do not place too much weight on them.

These results are suggestive of three important findings regarding first-stage effects. First, the experiment had a large effect for both boys and girls on increasing computer ownership. Second, the increase in computer ownership was similar for girls and boys. Third, the experiment also increased computer use for both boys and girls for numerous activities.

3. Girl-Boy Differences in Academic Performance

I first examine girl-boy differences in academic performance. Figure 1 reports average grades for boys and girls from the High School Transcript Study which is part of the National Assessment of Educational Progress (NAEP) conducted by the National Center for Educational Statistics (NCES). The latest available data is for 2009 and grades are available overall and in several different subjects. The estimates clearly indicate that girls obtain better grades than boys. They have higher overall grades than boys and obtain better grades in every core subject matter. The disparities are large. Even in math and science, girls obtain grades that are nearly 0.2 points

higher, and in English grades are over 0.3 points higher (which is equivalent to a + or – modifier on a letter grade).

Gender differences are large among disadvantaged schoolchildren. Table 1 reports average grades by parental education and race (which are the categories available in the NAEP grade data). Boys from families with low parental education or from underrepresented minority groups have lower grades in all subjects than girls from similar families. In fact, the grade underperformance of boys relative to girls is large across all parental education groups and all racial groups.

Figure 2 reports test score data for boys and girls collected as part of the 2011 NAEP. Overall girls score better, on average, than boys on reading and writing assessment tests, similarly on math assessment tests, but slightly lower on science assessment tests. Table 2 reports test scores by school lunch eligibility and race (which are the categories available in the NAEP test score data). The girl-boy patterns in test scores across subjects (i.e. girls score higher in reading and writing, similarly in math, and slightly lower in science than boys) hold for low-income and minority schoolchildren.

Using the NAEP data on test scores, I also examine girl and boy distributions of test scores. Figure 3 reports inverse cumulative distribution functions for each of the test scores reported in Figure 2 for both boys and girls. The distributional estimates are limited by NAEP reporting to showing the 10th, 25th, 50th, 75th, and 90th percentiles, but these percentiles characterize the full distribution reasonably well. Because I report inverse CDFs, the vertical difference at each of the reported points in the distribution is the equivalent to a quantile treatment effect (QTE) estimate. In addition to girls having higher average test scores in reading and writing, girls have higher test scores throughout the distribution. At each reported percentile

girls have higher scores than boys. For math test scores, girls and boys have roughly similar scores throughout the distribution with some slight differences at the reported tails. For science test scores, the distribution for boys is slightly higher at all points.

In addition to performing better on grades, and reading and writing test scores, girls outperform boys along numerous other measures of educational outcomes. Girl-boy disparities in many of these measures are not as well-known as those for grades, high school graduation rates, and college attendance. Appendix Table 3 provides evidence of consistent and sizeable gender disparities in several educational outcomes starting with outcomes relevant to school entry and ending with high school graduation.

The evidence clearly indicates that boys are doing worse in school than girls. Schoolchildren from disadvantaged families are no exception, with minority and low-income boys having lower grades in all subjects and test scores in some subjects than girls. Although some of these differences in academic performance have received attention recently, we know relatively little about differences in computer access, use, and a broader set of after-school time investments made by boys and girls.

4. Girl-Boy Differences in Computer Access and Use

Microdata from the computer supplement to the 2011 CPS is first used to explore whether girls and boys have similar levels of access to computers at home. Table 3 reports home computer access rates for girls and boys. Overall, boys and girls have identical rates of access to computers at home. The lack of differences by gender is not simply due to boys and girls living in the same households. Even in households with only boys or only girls access rates are identical. Among low-income families and disadvantaged minorities, boys and girls also have

similar access rates to home computers. This finding of lack of girl/boy differences in computer access differs substantially from the large differences found by race or income (see Hoffman and Novak 1998; Mossberger, Tolbert, and Stansbury 2003; Ono and Zavodny 2003b; Fairlie 2004; Goldfarb and Prince 2008 for example). Ruling out gender differences in access to home computers is important because it focuses the analysis on differential use of computers between boys and girls, which is examined next.

Data from three national sources are used to examine whether disadvantaged boys and girls use computers differently. I use microdata from the 2003 CPS Computer and Internet Use Supplement, microdata from surveys of teenagers conducted as part of the Pew Internet and American Life Project, and a time use diary study of the use of technology by children conducted by the Kaiser Family Foundation. Table 4 reports gender differences in use. Focusing on how boys and girls use computers reveals some interesting differences. I find consistent evidence of four main patterns across the datasets. First, I find that girls are more likely than boys to use computers for schoolwork (although these differences are not large). Second, boys spend more time playing video games on computers than do girls. Third, I find some evidence that girls use computers more for social networking, email, and other communication activities. Finally, these patterns are similar for low-income and high-income children.

These results provide some evidence that boys and girls use computers differently, but it is not clear whether these differences lead to differential effects of home computers on crowding out homework and educational outcomes, and thus contribute to the gender achievement gap. Unfortunately, this question is not an easy one to answer empirically. One possibility is to conduct an experiment in which computers are randomly taken away from schoolchildren who already own them and examine their subsequent academic performance. Another approach,

which is much more feasible, is to conduct a random experiment providing computers to schoolchildren who do not already own them and examine their subsequent academic performance.⁹ I take this approach next.

5. Girl-Boy Differences in Impacts of Computers on Crowding Out Homework and Educational Outcomes

In this section, I examine whether computers crowd out homework time and effort differently for boys and girls, and whether computers have differential impacts on the educational outcomes of boys and girls. Recent research focusing on all children finds mixed results on the impacts of home computers on educational outcomes, but none of these studies focus on differential impacts by gender and their implications for the gender gap in academic achievement.¹⁰ To examine whether home computers have differential educational impacts for boys and girls using the experimental data, I can simply calculate treatment-control differences in mean values of each measure for the boy and girl samples separately. To improve precision and confirm the robustness of the results to randomization, however, I estimate several regressions for homework time and educational outcomes. The regression equation is straightforward in the context of the random experiment:

$$(1) Y_i = \alpha + \beta^G T_i G_i + \beta^B T_i B_i + \theta G_i + \delta X_i + \varepsilon_i,$$

⁹ The alternative approaches also have different implications for external validity. An experiment taking the former approach focuses on a more advantaged group whereas as an experiment taking the latter approach focuses on a less advantaged group. The two groups may use computers differently.

¹⁰ See Schmitt and Wadsworth (2004); Fairlie (2005); Fuchs and Woessmann (2004); Fiorini (2010); Beltran, et al. (2010); Malamud and Pop-Eleches (2010); Beuermann et al. (2012); Fairlie and Robinson (2013); Vigdor, et al. (2014); Falck, et al. 2015 for example, and see Bulman and Fairlie (2015) for a recent review of the literature.

where Y_i is the outcome for student i (e.g. grade), T_i is an indicator variable for being in the treatment group, G_i is an indicator for girls, B_i is an indicator for boys, \mathbf{X}_i includes the baseline characteristics such as demographic and family characteristics reported in Appendix Table 1, and ε_i is an error term. The separate effects of becoming eligible for a free computer or the "intent-to-treat" estimate of the giveaway program are captured by β^G for girls and β^B for boys, respectively. The differential impact of home computers on educational outcomes between girls and boys is equal to $\beta^G - \beta^B$. All specifications are estimated using OLS and robust standard errors are reported with adjustments for multiple observations per student (i.e. clustered by student) when needed for grades.¹¹ The standard error for $\beta^G - \beta^B$ is estimated from the re-specified regression:

$$(2) Y_i = \alpha + \beta^B T_i + (\beta^G - \beta^B) T_i G_i + \theta G_i + \delta \mathbf{X}_i + \varepsilon_i.$$

Marginal effects estimates are similar from probit and logit models, and are thus not reported.

I first examine whether obtaining a computer crowds out homework time and effort for boys relative to girls using information collected from our follow-up survey. Table 5 reports estimates of home computer effects on self-reported measures of time spent doing homework, turning assignments in on time, and how much time is spent on the last essay or report. I report estimates of treatment effects separately for boys and girls and the difference between the two (which is estimated in a separate regression to obtain the standard error). Home computers do not crowd out homework time for boys relative to girls. The point estimate on the girl-boy treatment

¹¹ For all regressions for each educational outcome, I include controls for the sampling strata (school*year) in addition to the controls listed in Appendix Table 1. To avoid dropping observations, for each control variable, I create a dummy equal to 1 if the variable is missing for a student and code the original variable as a 0 (so that the coefficients are identified from those with non-missing values). Estimates are similar when I instead exclude these observations (there are only a few missing values).

is actually negative, although small and statistically insignificant. Obtaining a personal computer also has no negative effect on whether boys report turning in homework on time relative to girls. Finally, the computers did not appear to result in boys spending less time working on essays than girls.

Overall, there is no evidence indicating that computers crowd out homework time and effort for boys relative to girls. Another interesting finding is that in absolute terms I also do not find evidence that computers crowd out homework time and effort among boys (or girls).

Turning to educational outcomes, I examine whether home computers have a negative effect on grades, test scores, total courses completed and tardies obtained from administrative data from each of the schools. Table 6 reports estimates of the separate effects of home computers on grades for boys and girls and the difference between the two.¹² Panel A reports estimates of treatment effects on overall grades and grades in academic subjects (i.e. math, English, social studies, science) for boys and girls and the difference between the two. For the grade regressions, I pool the quarter 3 and 4 grades together. I find similar results when I estimate separate regressions for quarter 3 and quarter 4. I also include controls for quarter 1 grades, the subject and quarter in the regressions. Grades are coded as A-4, B-3, C-2, D-1, F-0. +/- modifiers are set equal to 0.33 points. In all cases, I find no evidence of a positive or negative effect of computers on grades for boys or girls.¹³ Similarly, home computers do not have a differential effect on grades for boys and girls.

¹² I focus on grades first because of their importance in determining high school graduation and college admissions (Betts and Morrell 2009).

¹³ LATE (or IV) estimates would be about twice as large (since the difference in computer usage is 55 percentage points). I do not report these estimates, however, because I cannot technically scale up the coefficients with the IV estimator because of differential timing of purchasing computers over the school year by the control group (two thirds of the control group with a home computer at follow-up obtained this computer after the fall). The finding that 80-82 percent of the treatment group reports having a computer at the end of the school year also creates difficulty in scaling up the ITT estimates because I

In Columns 3-4, I supplement the overall grade estimate by focusing on the effects of home computers on the pass/fail part of the grade distribution. In all of the schools, a grade of D- or higher is considered passing and provides credit towards moving to the next grade level and graduation. Again, I find no evidence of a differential effect of home computers between boys and girls. Expanding the distribution even further, I focus on the effects of home computers on the probability of obtaining a grade of B or higher (columns 5-6) and a grade of A or higher (columns 7-8). In both cases, I find no evidence of a differential effect of home computers for boys and girls. Estimates from quantile regressions for the full post-treatment achievement distribution confirm these findings (not reported). I do not find evidence of a clear pattern of differential treatment effects across the distribution.

I also examine whether treatment effects differ by subject. The finding for overall grades holds when examining courses separately by subject. Girls perform better in all subjects than boys, but home computers have no differential effect, either negative or positive, on grades for any subject. The lack of a negative *relative* effect for boys suggests that home computers are not contributing to why boys have lower grades in all subjects than girls. The finding holds for both average grades and along the pass/fail margin.

Related to this issue, I examine whether there are differential treatment effects across the pre-treatment grade distribution. There might be negative relative treatment effects for boys for some parts of the distribution that cannot be identified focusing on the average treatment effect. I estimate the following regression:

$$(2) Y_i = \beta_{pc} * D_{ip} * C_i + \beta_{pt} * D_{ip} * T_i + \delta \mathbf{X}_i + \varepsilon_i$$

know that essentially all treatment students picked up their computers and that many of the treatment group reporting not having a computer at follow-up indeed had a computer at home (based on subsequent conversations with the students by principals). For these reasons I focus on the ITT estimates.

In the regression, D_{ip} is an indicator for whether individual i is in the p th percentile of the pre-treatment GPA distribution. Percentiles are calculated within each school and are restricted to 20 different percentile categories. C_i is an indicator for the control group, and T_i is an indicator for the treatment group. Thus, β_{pc} and β_{pt} are estimates of the relationship between pre- and post-treatment performance in the control and treatment groups, respectively, and the difference, $\beta_{pt} - \beta_{pc}$ provides an estimate of the treatment effect at the p th percentile. \mathbf{X}_i is a minimal set of controls, including only subject and quarter indicators (so that the coefficients represent the unconditional relationship between pre- and post-performance for the treatment and control groups). β_{pc} and β_{pt} are reported in Figure 4a for girls and Figure 4b for boys. Standard errors are clustered at the individual level, and the 95% confidence interval of the difference between the treatment and control groups is plotted.

The estimates displayed in the figures indicate that treatment effects are indistinguishable from zero at almost all points of the pre-treatment grade distribution for both girls and boys. Thus, I do not find evidence of differential effects of home computers for boys and girls across the distribution.

I also estimate the impacts of home computers performance on the California Standardized Testing and Reporting (STAR) Program tests. As part of the STAR Program, all California students are required to take standardized tests for English-Language Arts and math each spring. For regressions in which test scores are the dependent variable, I focus on two key measures. First, I report estimates for a standardized test score based on raw values. These test scores are normalized to have mean 0 and standard deviation 1. I also report estimates for an indicator of proficiency. This variable is coded as 1 if the student receives a 4 or 5 (out of 5) on the test, and 0 otherwise. Proficiency and advanced scores meet state standards and are important

for schools to satisfy Adequate Yearly Progress (AYP) as part of the No Child Left Behind Act.

Table 7 reports estimates of the effects of home computers on test scores in English/Language Arts and mathematics. For both test scores, and whether I use a standardized test score or an indicator for meeting proficiency levels, I do not find evidence that home computers have a differential effect for boys and girls.

In addition to not finding a differential effect between girls and boys at the proficiency level I also do not find effects throughout the distribution. Plots of inverse cumulative distribution functions (CDFs) for both boys and girls reveal substantial overlap between the treatment and control groups for both test scores.¹⁴ The lack of treatment effects across the distribution implies that there are no differential effects between boys and girls. Similarly, Figures 5 and 6 examine the effects of home computers on STAR scores by prior achievement levels. Again, there is no discernible effect at almost any point in the pre-treatment STAR distribution. The finding holds for both English/Language Arts and math test scores. These figures suggest minimal effects of computers across the pre-treatment ability distribution and rule out the possibility that the null estimates of average treatment effects are due to offsetting positive and negative treatment effects at different parts of the pre-treatment achievement distribution. Most importantly, the lack of treatment effects for both boys and girls implies no differential effects throughout the distribution.

I also examine the effects of home computers on total courses completed and number of tardies. Estimates are reported in Columns 5-7 of Table 7. I find no evidence of a differential effect of home computers on total courses completed in the 3rd and 4th quarters of the academic

¹⁴ I examined inverse CDFs because the STAR scores are lumped into only 5 bins and thus I cannot estimate quantile treatment effects.

year. Estimates from the experiment also do not indicate that differential effects of home computers explain boy-girl differences in being tardy for school.

For all of the educational outcomes examined there is no evidence of a negative relative effect for boys suggesting that home computers and their use cannot explain why boys generally do worse in school than girls.

6. Conclusions

The results from this study provide the first evidence in the literature on whether disadvantaged boys and girls use computers differently, whether home computers crowd out homework time differently for boys and girls, and whether home computers have differential effects on educational outcomes among boys and girls. Although estimates from the CPS indicate that girls and boys have similar rates of access to home computers, evidence from several sources of data indicate that boys use computers differently than girls. Boys use computers less for schoolwork and more for playing games, but less for communication such as through social networking, email and instant messaging, than girls. Using data from a large field experiment that randomly provides free personal computers to schoolchildren for home use, I test the hypothesis that these gender differences in computer use partly explain why boys generally do worse than girls in school. I do not find evidence that computers crowd out homework time and effort more for boys than for girls. Examining impacts on grades, test scores and additional educational outcomes, the evidence does not indicate negative effects of home computers for boys relative to girls. I do not find differential effects at notable points in the distribution such as pass rates and meeting proficiency standards, or throughout the distribution of post-treatment outcomes.

Disadvantaged boys and girls differ in how they use computers, but these differences do not appear to lead to different levels of crowding out of homework and study time, and do not ultimately lead to different grades, test scores and other educational outcomes. Thus, gender differences in time investments in how personal computers are used at home do not appear to contribute to the achievement gap between disadvantaged boys and girls. This finding has implications for the general view that girls are more "self-disciplined" than are boys. Both girls and boys are found here to use computers for non-educational activities, but for both boys and girls these activities do not appear to crowd out homework time and negatively affect performance in school.

For the broader picture of the girl-boy achievement gap, identifying, or ruling out, potential explanations for why boys are doing worse in school than girls is extremely important. Some policy recommendations include increasing the number of male teachers at young grades, all-boy classrooms, more hands-on activities, and more frequent or longer recesses. Recent trends in educational outcomes do not show relative improvement for boys, and differences between boys and girls are quite large. The girl-boy difference in grades of 0.2 grade points is only slightly smaller than the white-Latino difference of 0.25 grade points and half the white-black difference of 0.4 grade points. The racial achievement gap, however, has attracted considerably more attention in the literature and policy arena (e.g. Jencks and Phillips 1998). Further research on the causes of gender differences in educational outcomes especially among disadvantaged and low-income children is clearly needed.

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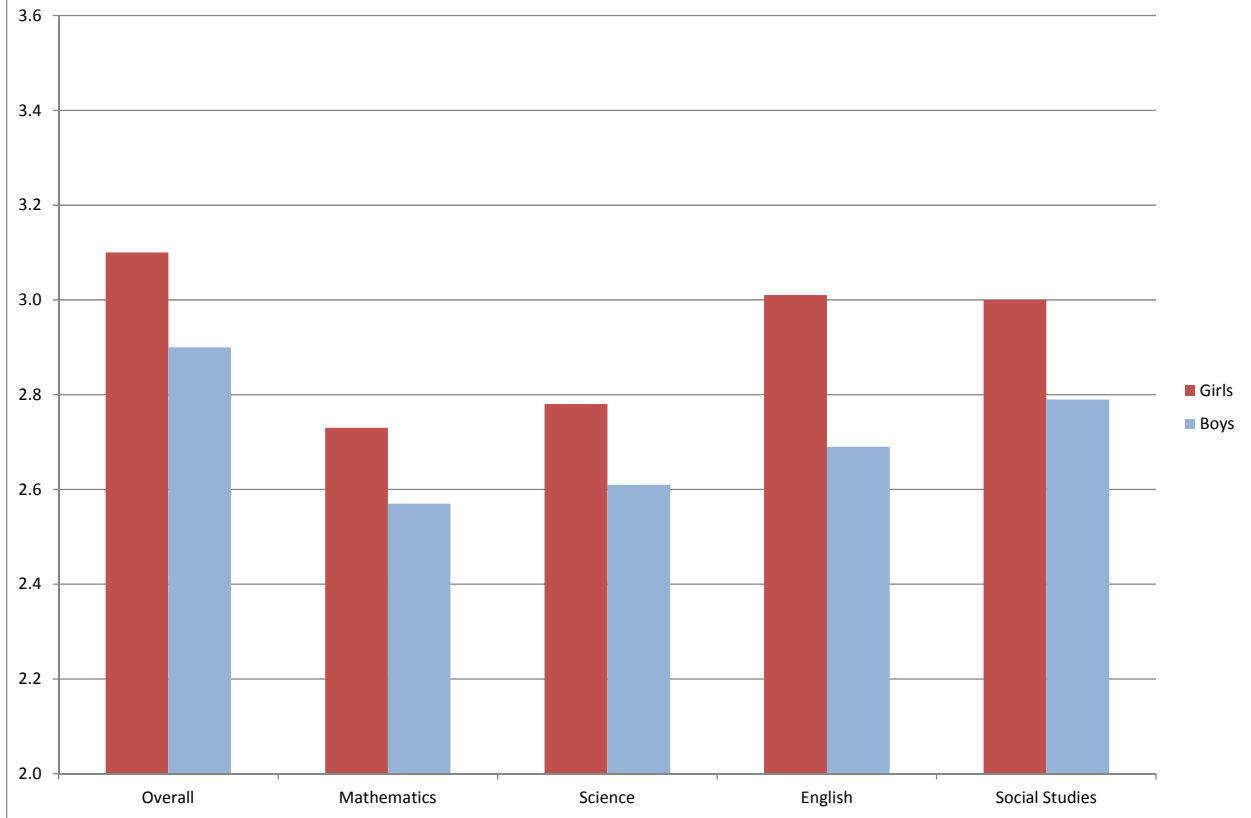
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Figure 1: Grade Point Average by Gender and Subject
High School Transcript Study, 2009



**Figure 2: Average Test Scores by Gender and Subject
National Assessment of Educational Progress, 2011**

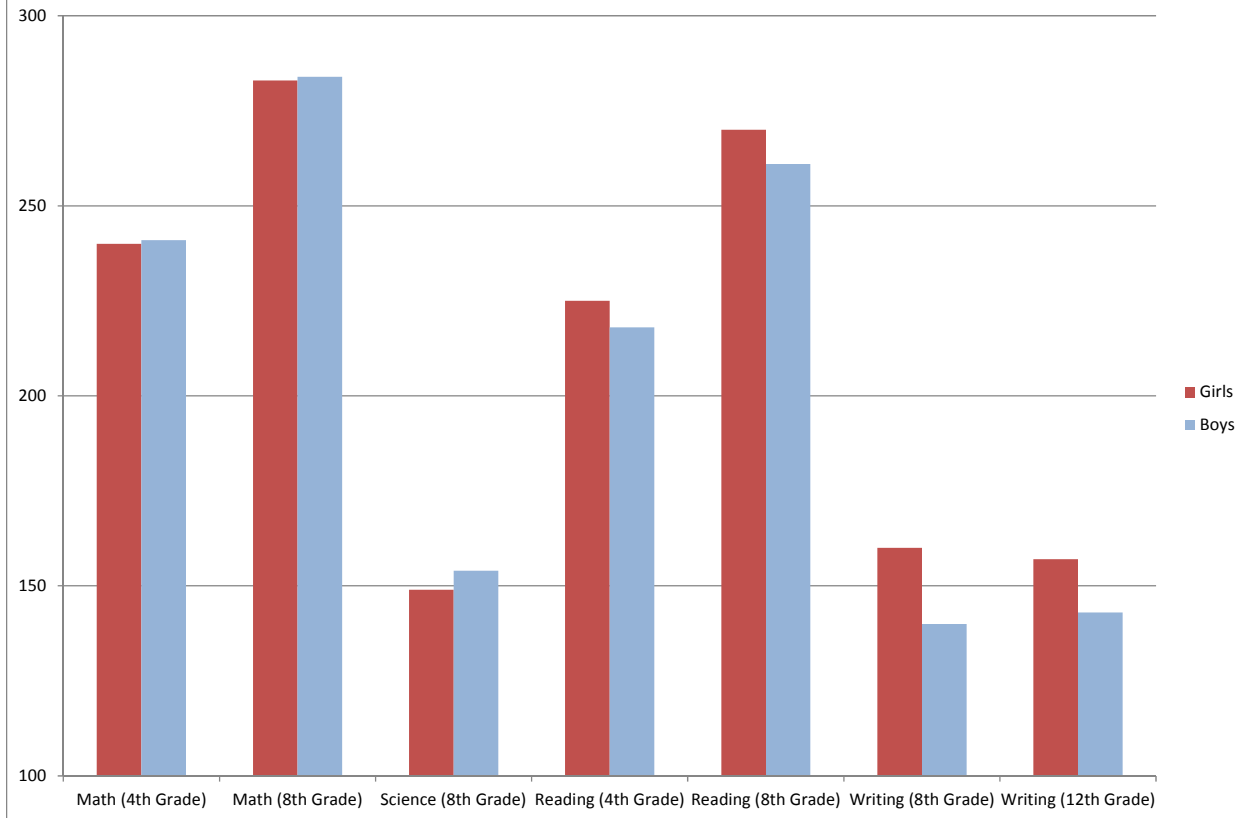


Figure 3.A: Inverse CDF for Test Scores (Reading and Writing)
National Assessment of Educational Progress, 2011

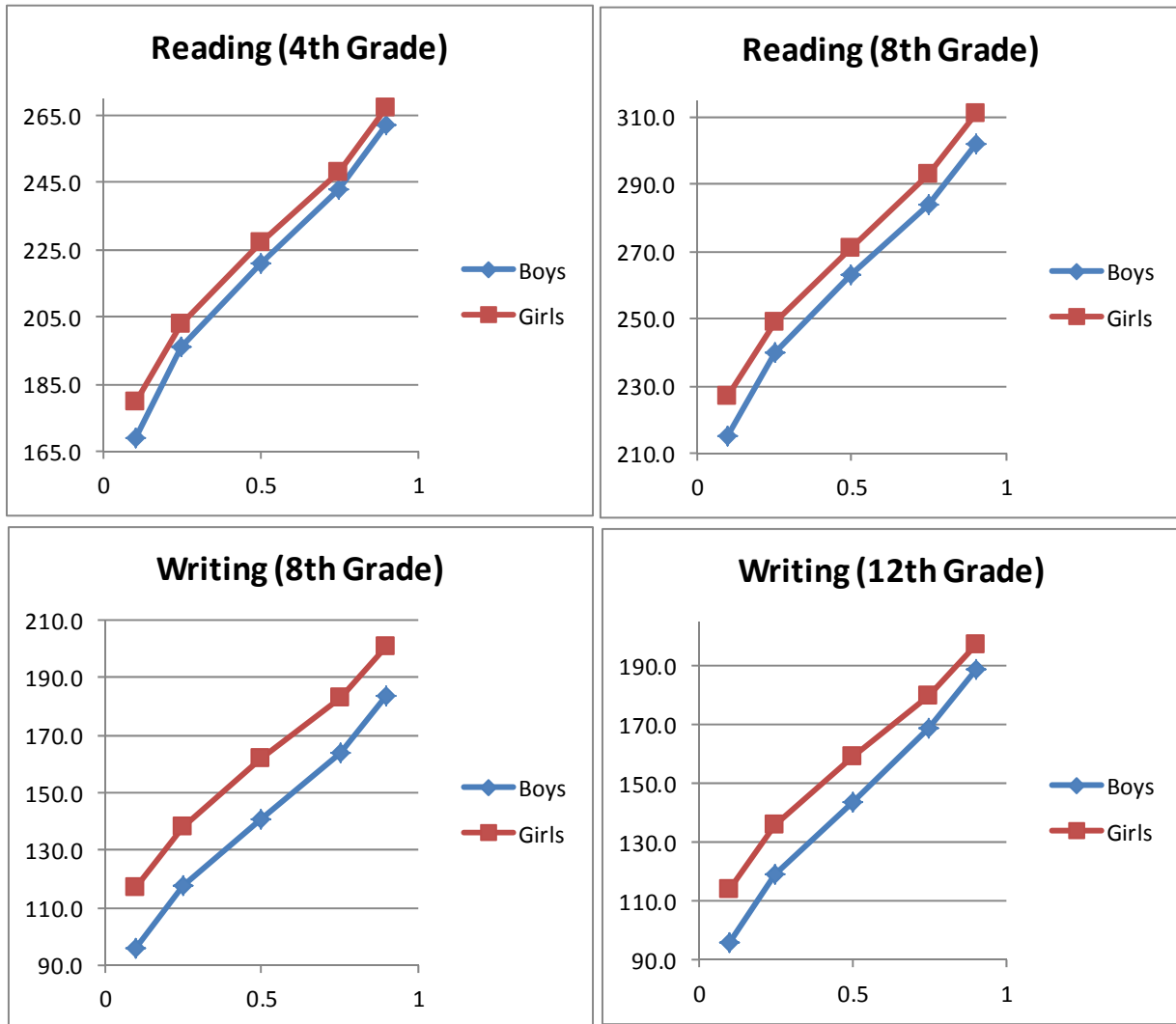


Figure 3.B: Inverse CDF for Test Scores (Math and Science)
National Assessment of Educational Progress, 2011

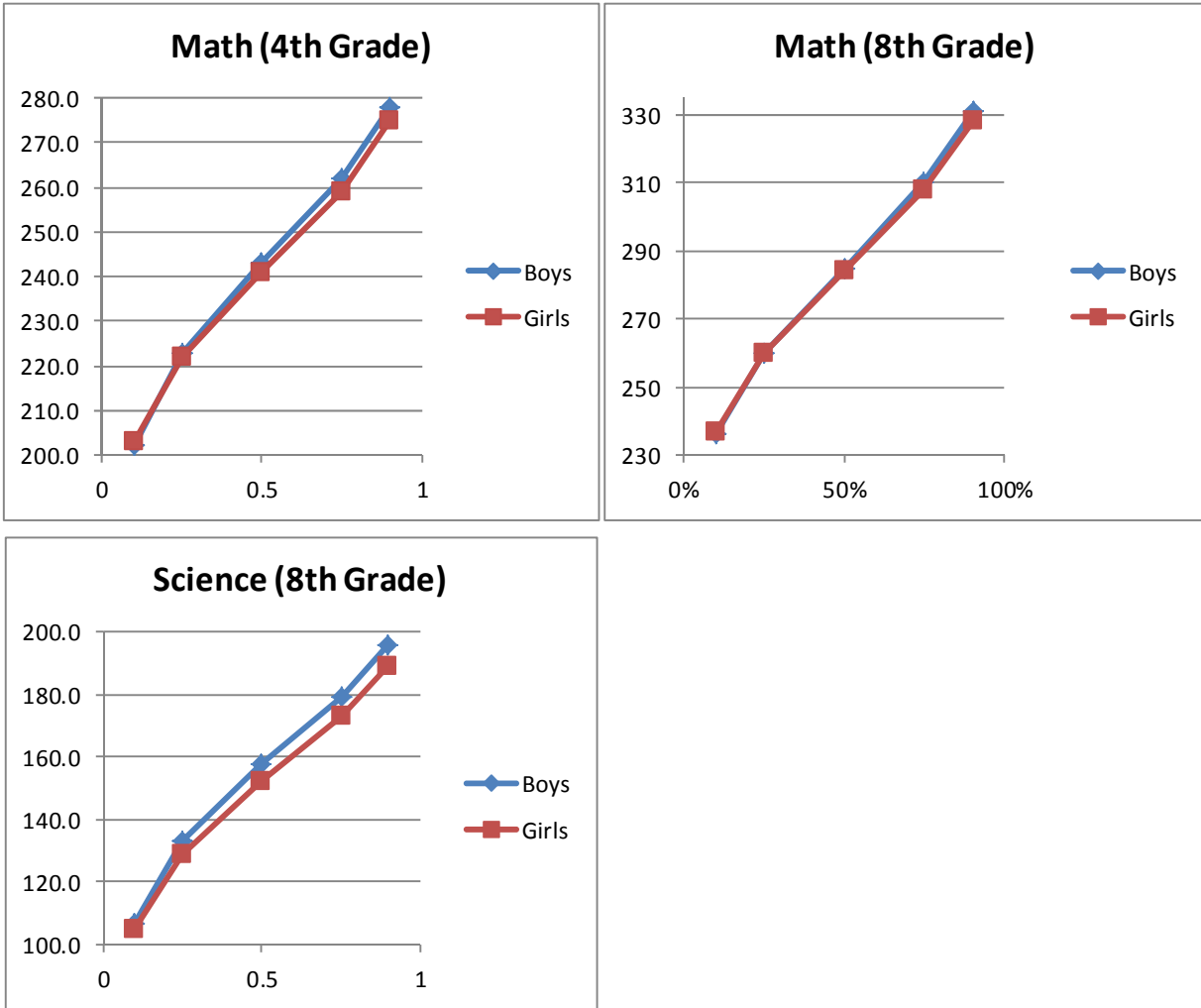
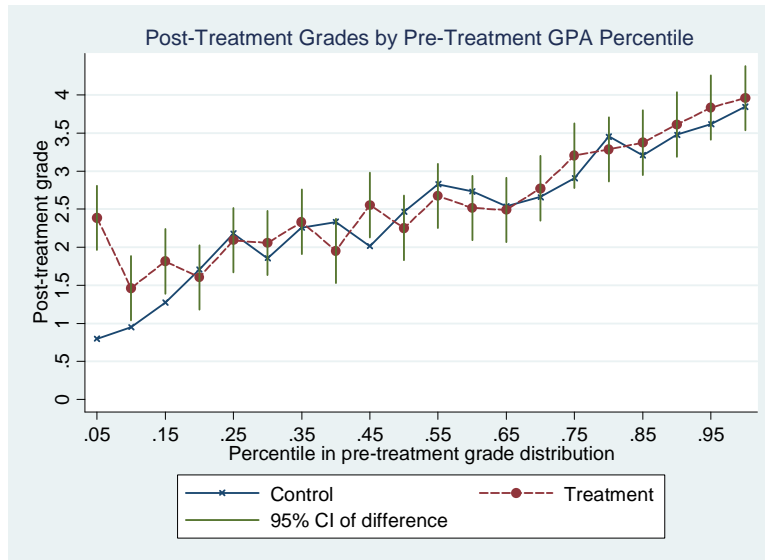
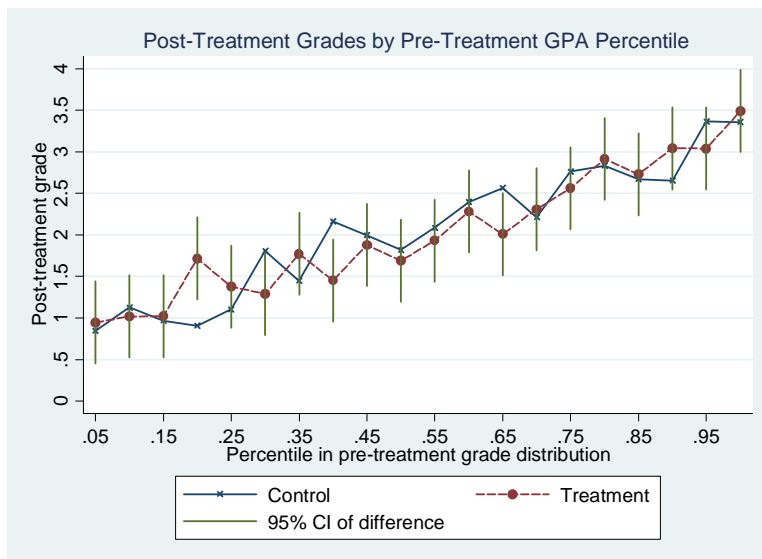


Figure 4. Post-Treatment Grades by Pre-Treatment GPA Percentile

Panel A. Girls



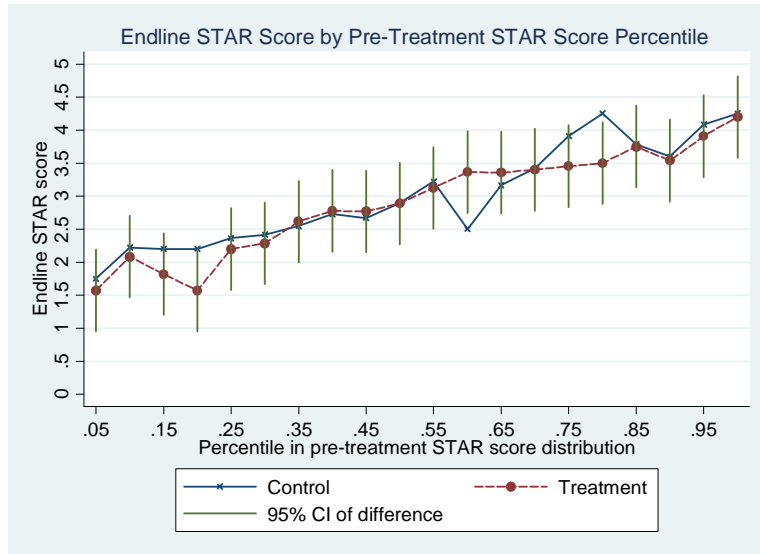
Panel B. Boys



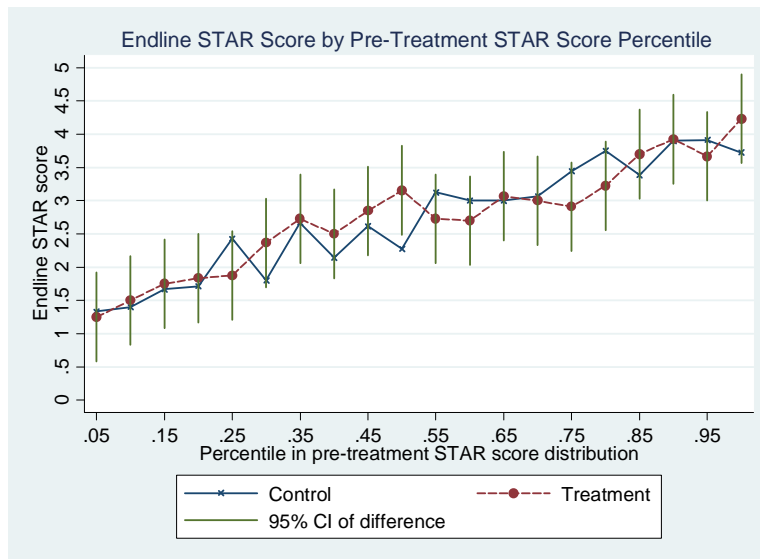
Notes: The graph shows estimated coefficients from a regression of post-treatment (quarters 3 and 4) grades on interactions between treatment and pre-treatment GPA percentile (in quarter 1, before the computers were given out). The vertical line is a 95% confidence interval for the difference between the treatment and control groups, at each percentile. See text for more details.

Figure 5. Post-Treatment English/Language Arts STAR scores by Pre-Treatment Star Percentiles

Panel A. Girls



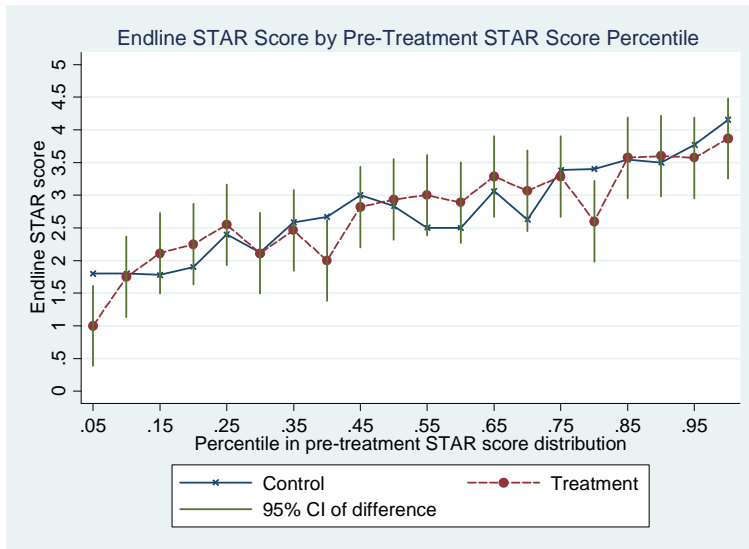
Panel B. Boys



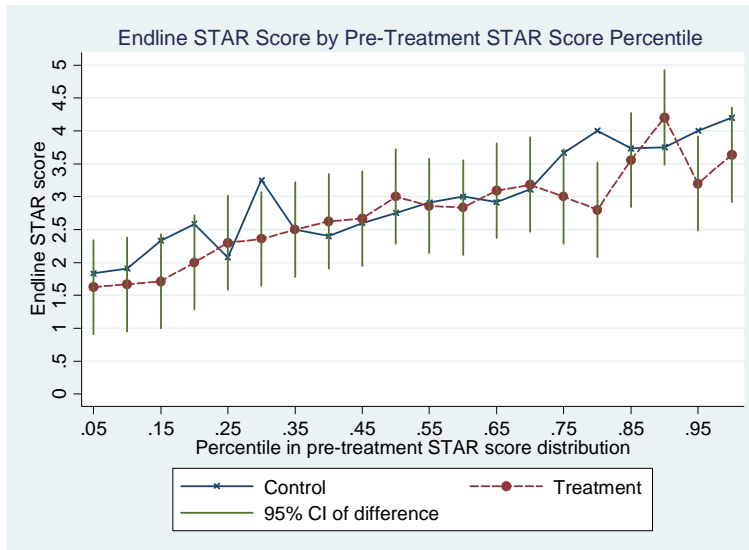
Notes: The graph shows estimated coefficients from a regression of endline STAR scores on interactions between treatment and pre-treatment STAR scores. The vertical line is a 95% confidence interval for the difference between the treatment and control groups, at each percentile. See text for more details.

Figure 6. Post-Treatment Math STAR scores by Pre-Treatment Star Percentiles

Panel A. Girls



Panel B. Boys



Notes: The graph shows estimated coefficients from a regression of endline STAR scores on interactions between treatment and pre-treatment STAR scores. The vertical line is a 95% confidence interval for the difference between the treatment and control groups, at each percentile. See text for more details.

Table 1. Grade Point Average by Gender, Parental Education and Race

	Girls	Boys	Girl-Boy Difference
<u>Grade Point Average (Overall)</u>			
All Students	3.10	2.90	0.20
Parental Education: High School Dropout	2.88	2.75	0.13
Parental Education: Graduated High School	2.98	2.77	0.21
Parental Education: Graduated College	3.27	3.04	0.23
Race: White	3.20	2.98	0.22
Race: Black	2.79	2.57	0.22
Race: Hispanic	2.91	2.75	0.16
Race: Asian	3.37	3.15	0.22
<u>Grade Point Average (Mathematics)</u>			
All Students	2.73	2.56	0.17
Parental Education: High School Dropout	2.51	2.42	0.09
Parental Education: Graduated High School	2.61	2.43	0.18
Parental Education: Graduated College	2.92	2.72	0.20
Race: White	2.84	2.63	0.21
Race: Black	2.41	2.23	0.18
Race: Hispanic	2.51	2.43	0.08
Race: Asian	3.09	2.94	0.15
<u>Grade Point Average (Science)</u>			
Average	2.78	2.61	0.17
Parental Education: High School Dropout	2.52	2.43	0.09
Parental Education: Graduated High School	2.64	2.47	0.17
Parental Education: Graduated College	2.99	2.78	0.21
Race: White	2.89	2.70	0.19
Race: Black	2.47	2.24	0.23
Race: Hispanic	2.53	2.44	0.09
Race: Asian	3.10	2.93	0.17
<u>Grade Point Average (English)</u>			
All Students	3.01	2.69	0.32
Parental Education: High School Dropout	2.74	2.49	0.25
Parental Education: Graduated High School	2.87	2.53	0.34
Parental Education: Graduated College	3.20	2.86	0.34
Race: White	3.11	2.77	0.34
Race: Black	2.71	2.37	0.34
Race: Hispanic	2.80	2.53	0.27
Race: Asian	3.30	2.97	0.33
<u>Grade Point Average (Social Studies)</u>			
All Students	3.00	2.79	0.21
Parental Education: High School Dropout	2.73	2.59	0.14
Parental Education: Graduated High School	2.86	2.62	0.24
Parental Education: Graduated College	3.20	2.96	0.24
Race: White	3.10	2.88	0.22
Race: Black	2.68	2.43	0.25
Race: Hispanic	2.78	2.60	0.18
Race: Asian	3.28	3.05	0.23

Source: High School Transcript Study, 2009.

Table 2. Average Test Scores by Gender, School Lunch Eligibility and Race

	Girls	Boys	Girl-Boy Difference
<u>Average Test Score (Math 4th Grade)</u>			
All Students	240	241	-1
Eligible for National School Lunch Program	229	229	0
Not Eligible for National School Lunch Program	251	253	-2
Race: White	248	251	-3
Race: Black	225	225	0
Race: Hispanic	231	232	-1
Race: Asian	260	260	0
<u>Average Test Score (Math 8th Grade)</u>			
All Students	283	284	-1
Eligible for National School Lunch Program	270	270	0
Not Eligible for National School Lunch Program	296	298	-2
Race: White	294	296	-2
Race: Black	265	262	3
Race: Hispanic	272	272	0
Race: Asian	310	311	-1
<u>Average Test Score (Science 8th Grade)</u>			
All Students	149	154	-5
Eligible for National School Lunch Program	135	139	-4
Not Eligible for National School Lunch Program	161	166	-5
Race: White	162	168	-6
Race: Black	127	130	-3
Race: Hispanic	136	140	-4
Race: Asian	161	164	-3
<u>Average Test Score (Reading 4th Grade)</u>			
All Students	225	219	6
Eligible for National School Lunch Program	211	204	7
Not Eligible for National School Lunch Program	239	233	6
Race: White	235	229	6
Race: Black	212	203	9
Race: Hispanic	214	207	7
Race: Asian	243	236	7
<u>Average Test Score (Reading 8th Grade)</u>			
All Students	270	261	9
Eligible for National School Lunch Program	259	249	10
Not Eligible for National School Lunch Program	284	273	11
Race: White	282	273	9
Race: Black	256	245	11
Race: Hispanic	261	252	9
Race: Asian	289	277	12
<u>Average Test Score (Writing 8th Grade)</u>			
All Students	160	140	20
Eligible for National School Lunch Program	144	125	19
Not Eligible for National School Lunch Program	171	151	20
Race: White	169	149	20
Race: Black	140	123	17
Race: Hispanic	147	129	18
Race: Asian	175	158	17
<u>Average Test Score (Writing 12th Grade)</u>			
All Students	157	143	14
Eligible for National School Lunch Program	140	126	14
Not Eligible for National School Lunch Program	165	150	15
Race: White	167	152	15
Race: Black	136	123	13
Race: Hispanic	142	130	12
Race: Asian	164	152	12

Source: National Assessment of Educational Progress, 2011.

Table 3. Access to Personal Computers at Home by Boys and Girls

	Percent with access to a home computer		
	Girls	Boys	Girl-Boy Difference
Total	84%	84%	0%
Girl only household	85%		
Boy only household		85%	
Girl and boy household	83%	83%	0%
Ages 5-9	81%	82%	0%
Ages 10-14	85%	84%	1%
Ages 15-17	86%	87%	0%
Family income < \$20,000	62%	60%	2%
Family income \$20,000-39,999	75%	75%	-1%
Family income \$40,000-74,999	91%	92%	-1%
Family income \$75,000-99,999	97%	96%	1%
Family income \$100,000 or more	98%	97%	1%
White, non-Hispanic	91%	91%	1%
Hispanic	70%	71%	-1%
Black	75%	75%	0%
Asian	93%	96%	-3%

Source: Current Population Survey, Computer and Internet Supplement 2011 Microdata.

Table 4. Computer Use by Gender and Family Income

	Girls	Boys	Girl-Boy Difference
Current Population Survey 2003			
Percent of Internet users using for playing games	61%	68%	-7%
Low-income children	59%	65%	-6%
High-income children	63%	69%	-6%
Percent of Internet users using for email and messaging	64%	57%	7%
Low-income children	52%	49%	3%
High-income children	69%	60%	9%
Percent of Internet users using for school assignments	79%	77%	2%
Low-income children	75%	73%	2%
High-income children	80%	78%	2%
Percent using Internet anywhere	61%	58%	3%
Low-income children	49%	46%	3%
High-income children	70%	69%	1%
Pew Internet Study (2007-08)			
Percent playing video games daily	22%	39%	-17%
Low-income children	23%	40%	-17%
High-income children	20%	37%	-17%
Percent sending email on a daily basis	20%	12%	8%
Low-income children	20%	11%	9%
High-income children	17%	10%	7%
Percent using Internet ever for school research	96%	92%	4%
Low-income children	89%	88%	1%
High-income children	97%	92%	5%
Kaiser Foundation Time Use Diary 2009			
Minutes of computer use for playing games	8	25	-17
Minutes of computer use for videos and other entertainme	19	23	-4
Minutes of computer use for social networking	25	19	6
Minutes of computer use for email and instant messaging	18	16	2
Minutes of computer use for other activities	12	14	-2
Minutes of computer use for schoolwork	19	13	6
Total minutes per day of computer use	101	110	-9

Sources: Kaiser Family Foundation (2010); Microdata from the Pew Internet Projects (2008a, 2008b); Current Population Survey, Computer and Internet Supplement 2003 microdata.

Table 5. Experimental Estimates of Computer Crowd-Out Impacts on Homework Time and Effort

	(1)	(2)	(3)	(4)	(5)
	How many hours per week do you spend on homework?	How often do you turn in homework on time?			How much time did you spend on last essay?
		Always	Usually	Sometimes	
Girl treatment	-0.38 (0.38)	-0.01 (0.05)	-0.02 (0.05)	0.03 (0.04)	-0.22 (1.15)
Boy treatment	0.20 (0.39)	-0.07 (0.05)	0.07 (0.05)	0.00 (0.04)	0.16 (1.17)
Girl-boy treatment diff.	-0.58 (0.56)	0.06 (0.07)	-0.10 (0.07)	0.04 (0.05)	-0.38 (1.66)
Observations	825	853	853	853	805
Girl control mean	2.79	0.48	0.37	0.15	4.02
Boy control mean	2.49	0.46	0.38	0.17	4.77

*Notes: Data is from follow-up survey completed by students at end of school year. Regressions include controls for sampling strata (school*year) and variables listed in Appendix Table 1. ***, **, * indicates significance at 1, 5 and 10%.*

Table 6. Experimental Estimates of Home Computer Impacts on Grades

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	Grades		Indicator for passing class		Indicator for B or Higher Grade		Indicator for A or Higher Grade	
	All subjects	Academic Subjects	All subjects	Academic Subjects	All subjects	Academic Subjects	All subjects	Academic Subjects
Panel A. Class Grades								
Girl treatment	0.00	0.04	0.01	0.02	-0.01	0.00	-0.02	0.00
	(0.05)	(0.06)	(0.01)	(0.01)	(0.02)	(0.02)	(0.02)	(0.02)
Boy treatment	-0.03	0.00	-0.01	-0.02	-0.01	0.01	0.00	0.01
	(0.05)	(0.06)	(0.01)	(0.02)	(0.02)	(0.02)	(0.02)	(0.02)
Girl-boy treatment diff.	0.03	0.04	0.02	0.03	0.00	-0.02	-0.01	-0.01
	(0.08)	(0.09)	(0.02)	(0.02)	(0.03)	(0.03)	(0.03)	(0.03)
Observations	11514	7820	11514	7820	11514	7820	11514	7820
Number of students	1036	1035	1036	1035	1036	1035	1036	1035
Girl control mean	2.58	2.40	0.90	0.89	0.59	0.53	0.33	0.27
Boy control mean	2.37	2.10	0.86	0.83	0.52	0.43	0.28	0.20
	Grade				Indicator for passing class			
	Math	English / Reading	Social Studies	Science	Math	English / Reading	Social Studies	Science
Panel B. Class Grades by Subject								
Girl treatment	0.03	-0.06	0.12	0.09	0.02	-0.01	0.05	0.01
	(0.09)	(0.08)	(0.09)	(0.08)	(0.03)	(0.02)	(0.02)**	(0.02)
Boy treatment	0.01	-0.14	0.07	0.08	0.00	-0.03	-0.01	-0.03
	(0.09)	(0.09)	(0.10)	(0.09)	(0.03)	(0.03)	(0.03)	(0.03)
Girl-boy treatment diff.	0.02	0.08	0.05	0.01	0.02	0.02	0.06	0.04
	(0.13)	(0.12)	(0.13)	(0.12)	(0.04)	(0.03)	(0.04)	(0.03)
Observations	1886	2121	1784	1895	1886	2121	1784	1895
Number of students	969	903	921	960	969	903	921	960
Girl control mean	2.05	2.65	2.45	2.42	0.83	0.92	0.88	0.89
Boy control mean	1.93	2.28	2.10	2.05	0.81	0.85	0.82	0.82

*Notes: Regressions include controls for sampling strata (school*year), variables listed in Appendix Table 1, and previous grades. ***, **, * indicates significance at 1, 5 and 10%.*

Table 7. Experimental Estimates of Home Computer Impacts on Test Scores and Other Educational Outcomes

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Standardized Score		Proficiency Indicator		Total Credits Earned		
	English / Lang. Arts	Math	English / Lang. Arts	Math	In 3rd quarter	In 4th quarter	Number of Tardies
Girl treatment	-0.10 (0.06)	0.02 (0.07)	-0.02 (0.03)	-0.03 (0.04)	-0.01 (0.13)	0.04 (0.14)	0.18 (1.31)
Boy treatment	0.00 (0.07)	-0.13 (0.07)*	0.02 (0.03)	0.00 (0.04)	0.09 (0.14)	-0.10 (0.14)	-0.72 (1.34)
Girl-boy treatment diff.	-0.10 (0.09)	0.15 (0.10)	-0.04 (0.05)	-0.03 (0.05)	-0.10 (0.19)	0.15 (0.20)	0.90 (1.89)
Observations	961	914	961	914	1123	1123	1104
Girl control mean	0.13	-0.04	0.33	0.29	5.42	5.54	9.65
Boy control mean	-0.13	0.04	0.25	0.31	5.30	5.43	13.49

Notes: Regressions include controls for sampling strata (school*year), variables listed in Appendix Table 1, and previous test scores. ***, **, * indicates significance at 1, 5 and 10%.

Appendix Table 1. Individual Level Summary Statistics and Balance Check

	Total			Girls			Boys		
	Control	Treatment	Equality of means <i>p-value</i>	Control	Treatment	Equality of means <i>p-value</i>	Control	Treatment	Equality of means <i>p-value</i>
Age	12.91 (0.87)	12.90 (0.84)	0.91	12.92 (0.91)	12.85 (0.78)	0.37	12.90 (0.83)	12.96 (0.89)	0.46
Female	0.51 (0.50)	0.50 (0.50)	0.66	1	1		0	0	
Ethnicity = African American	0.13 (0.34)	0.14 (0.34)	0.86	0.14 (0.35)	0.13 (0.34)	0.66	0.13 (0.33)	0.15 (0.35)	0.50
Ethnicity = Latino	0.56 (0.50)	0.55 (0.50)	0.76	0.55 (0.50)	0.57 (0.50)	0.66	0.57 (0.50)	0.54 (0.50)	0.38
Ethnicity = Asian	0.12 (0.33)	0.14 (0.34)	0.42	0.12 (0.33)	0.14 (0.35)	0.49	0.12 (0.33)	0.13 (0.34)	0.66
Ethnicity = White ¹	0.16 (0.36)	0.14 (0.35)	0.56	0.16 (0.37)	0.14 (0.35)	0.61	0.15 (0.36)	0.15 (0.35)	0.76
Immigrant	0.21 (0.41)	0.18 (0.38)	0.15	0.21 (0.41)	0.18 (0.39)	0.40	0.21 (0.41)	0.17 (0.38)	0.23
Primary language is English	0.43 (0.50)	0.43 (0.50)	0.97	0.43 (0.50)	0.43 (0.50)	0.98	0.43 (0.50)	0.43 (0.50)	0.97
Parent's education ²	12.81 (1.44)	12.76 (1.49)	0.64	12.71 (1.48)	12.77 (1.50)	0.67	12.94 (1.39)	12.75 (1.48)	0.22
Number of people living in household	4.98 (2.43)	5.02 (2.55)	0.79	5.03 (2.34)	5.13 (2.47)	0.62	4.93 (2.53)	4.92 (2.63)	0.95
Lives with mother	0.92 (0.28)	0.89 (0.32)	0.12	0.90 (0.30)	0.91 (0.29)	0.77	0.93 (0.25)	0.87 (0.34)	0.014**
Lives with father	0.58 (0.49)	0.58 (0.49)	0.90	0.53 (0.50)	0.56 (0.50)	0.48	0.64 (0.48)	0.60 (0.49)	0.33
Does your mother have job? ⁴	0.47 (0.50)	0.46 (0.50)	0.68	0.47 (0.50)	0.44 (0.50)	0.55	0.47 (0.50)	0.48 (0.50)	0.98
Does your father have a job?	0.73 (0.44)	0.70 (0.46)	0.36	0.70 (0.46)	0.65 (0.48)	0.41	0.77 (0.43)	0.74 (0.44)	0.66
Full sample size	564	559		289	279		275	280	

Notes: In Columns 1, 2, 4, 5, 7 and 8, means reported with standard errors in parentheses. Column 3, 6 and 9 report the *p-value* for the *t-test* for the equality of means. ***, **, * indicates significance at 1, 5 and 10%.

¹ Omitted ethnicity category is "not reported."

² This is the highest education level of either parent (which is the measure most schools in our sample collected).

³ The variables for mother's and father's job is reported only for households in which the given parent lives in the household.

Appendix Table 2. First-Stage Effects on Reported Hours of Computer Use by Location and Type of Activity

	(1)	(2)	(3)	(4)	(5)
	Hours of Computer Use Per Week				
	Schoolwork	Email	Games, Videos, Music, Ent.	Social Networking	Other
Panel A. Total					
Girl control group	2.10	0.27	0.58	0.44	0.54
Girl treatment group	2.84	0.59	1.27	1.07	0.77
Treatment-control	0.74	0.31	0.69	0.63	0.23
	(0.36)**	(0.17)*	(0.29)**	(0.24)**	-0.14
Boy control group	1.67	0.20	0.99	0.68	0.58
Boy treatment group	2.57	0.74	1.74	1.12	0.77
Treatment-control	0.90	0.54	0.75	0.44	0.19
	(0.35)**	(0.16)***	(0.28)***	(0.24)*	-0.14
Girl-Boy Treatment-Control	-0.16	-0.23	-0.06	0.19	0.04
	(0.51)	(0.23)	(0.40)	(0.34)	(0.20)
Observations	665	665	665	665	665
Panel B. At Home					
Girl control group	0.22	0.05	0.12	0.14	0.13
Girl treatment group	1.03	0.29	0.65	0.49	0.35
Treatment-control	0.81	0.24	0.54	0.35	0.22
	(0.16)***	(0.09)***	(0.15)***	(0.12)***	(0.07)***
Boy control group	0.21	0.08	0.23	0.12	0.10
Boy treatment group	0.98	0.33	0.81	0.62	0.34
Treatment-control	0.77	0.25	0.59	0.49	0.24
	(0.16)***	(0.09)***	(0.15)***	(0.12)***	(0.07)***
Girl-Boy Treatment-Control	0.04	-0.01	-0.05	-0.14	-0.02
	(0.23)	(0.13)	(0.21)	(0.17)	(0.10)
Observations	789	789	789	789	789
Panel C. At School					
Girl control group	1.11	0.06	0.16	0.05	0.23
Girl treatment group	0.97	0.05	0.12	0.03	0.17
Treatment-control	-0.15	-0.01	-0.04	-0.01	-0.06
	-0.16	-0.03	-0.11	-0.02	-0.07
Boy control group	0.91	0.07	0.30	0.08	0.27
Boy treatment group	1.11	0.05	0.39	0.04	0.13
Treatment-control	0.20	-0.02	0.09	-0.04	-0.15
	-0.16	-0.03	-0.11	(0.02)*	(0.07)**
Girl-Boy Treatment-Control	-0.35	0.01	-0.13	0.02	0.08
	(0.22)	(0.04)	(0.16)	(0.03)	(0.10)
Observations	771	771	771	771	771
Panel D. At Other Locations					
Girl control group	0.76	0.15	0.29	0.23	0.24
Girl treatment group	0.59	0.20	0.36	0.44	0.22
Treatment-control	-0.17	0.04	0.06	0.21	-0.02
	-0.19	-0.11	-0.10	-0.16	-0.08
Boy control group	0.53	0.09	0.48	0.47	0.23
Boy treatment group	0.43	0.36	0.48	0.46	0.30
Treatment-control	-0.10	0.27	0.01	-0.01	0.07
	-0.19	(0.11)**	-0.10	-0.16	-0.08
Girl-Boy Treatment-Control	-0.06	-0.23	0.06	0.22	-0.09
	(0.27)	(0.16)	(0.14)	(0.23)	(0.11)
Observations	771	771	771	771	771

Notes: Data are from follow-up survey completed by students at end of school year. ***, **, * indicates significance at 1, 5 and 10%.

Appendix Table 3. Educational Outcomes among Boys and Girls

	Girls	Boys	Girl-Boy Difference
Percent of 5-year olds enrolled in primary school (CPS 2011)	76.5%	73.4%	3.0%
Early-age test scores			
First-time kindergartners' mean reading scale score (ECLS-K 2011)	50.5	48.5	2.0
First-time kindergartners' mean mathematics scale score (ECLS-K 2011)	42.0	42.0	0.0
Average early reading scale score - children less than 48 months (ECLS-B 2007)	22.4	20.7	1.7
Average early reading scale score - children 48-57 months (ECLS-B 2007)	26.4	24.6	1.8
Average early reading scale score - children over 58 months (ECLS-B 2007)	30.4	29.0	1.4
Grade retention: 9th grade students ever retained (K-9) (HSLs 2009)	10.0%	15.0%	-5.0%
Learning difficulties			
Diagnosed with a learning disability (NHIS 2008)	7.3%	11.4%	-4.1%
Diagnosed with Attention Deficit Hyperactivity Disorder (NHIS 2008)	5.3%	16.6%	-11.3%
Percentage of gifted and talented students (CRDC 2006)	7.0%	6.3%	0.7%
Truancy and disciplinary			
Students aged 12-18 who skipped any classes in previous 4 weeks (SCS 2007)	4.7%	6.2%	-1.5%
Public school students in 9th-12th grade who had ever been suspended (SCS 2007)	16.9%	31.7%	-14.8%
Public school students in 9th-12th grade who had ever been expelled (SCS 2007)	1.9%	4.4%	-2.5%
Advanced placement and activities			
Advanced placement examination participation (CBNSR 2009-10)	56.0%	44.0%	12.0%
Average credits earned - advanced placement courses , grade 12 (CBNSR 2009-10)	1.20	0.96	0.24
High school seniors who participated in student council/government (MTF 2009)	13.1%	5.9%	7.2%
Average freshman graduation rate (CCD 2008-09)	79.0%	72.0%	7.0%
Percent of population aged 18-24 without a high school degree (CPS 2012)	17.2%	20.3%	3.1%

Sources: National Center for Education Statistics, Early Childhood Longitudinal Study, Kindergarten Class of 2010–11 (ECLS-K:2011); National Center for Health Statistics, National Health Interview Survey (NHIS 2008); Current Population Survey (CPS 2012); National Center for Education Statistics, Common Core of Data (CCD 2008-09); National Center for Education Statistics, Early Childhood Longitudinal Study, Birth Cohort (ECLS-B 2007); U.S. Department of Education, Office for Civil Rights, Civil Rights Data Collection (CRDC 2006); University of Michigan, Institute for Social Research, Monitoring the Future, (MTF 2009); The College Board, AP Program National Summary Reports (CBNSR 2009-10); U.S. Department of Justice, Bureau of Justice Statistics, School Crime Supplement (SCS) to the National Crime Victimization Survey (NCVS) (SCS 2007).