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School and drugs: Closing the gap - Evidence from a randomized trial in the US

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Evidence from a Randomized Trial in the US**

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ABSTRACT

School and Drugs: Closing the Gap – Evidence from a Randomized Trial in the US^{*}

We present evidence on how The Quantum Opportunity Program (QOP hereafter) worked in the US. While the program was regarded as successful in the short-term, in the long-run its educational results were modest and its effects on risky behaviors detrimental. Exploiting control group's self-reported drug use while in school, we evaluate whether the program worked best among those with high-predicted risk of problem behavior. We find QOP to be extremely successful among high-risk youths as it managed to curb their risky behaviors during high-school and, by doing so, it persistently improved high-school graduation by 20 percent and college enrollment by 28 percent. In contrast, QOP was unsuccessful among youths in the bottom-half of the risk distribution as it increased their engagement in risky behaviors while in high-school. Negative peer effects are possibly an explanation behind these results. Finally, negative peer effects also seem to explain the longer-run detrimental effects of QOP on risky behaviors.

JEL Classification: C93, I21, I22, I28, J24

Keywords: mentoring programs, peer effects, risky behaviors, educational programs, randomized trials

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Poor academic performance and engagement in risky behaviors are two of the most serious problems youths face today in the United States. Despite recent improvements, only 76 percent of young people graduated from high school in 2009 in the US, far from the OECD average of 82 percent. In addition, college completion among the 25- to 34-year-olds is also relatively low with the US ranking 15th among 34 OECD countries (*OECD in Figures*, 2011). At the same time, the US also underperforms in terms of 13- to 19-year-olds' engagement in risky behaviors as it ranks 15th among 30 OECD countries (OECD, 2009).¹ Estimates from the 2010 *National Survey on Drug Use and Health* reveal that as many as 10 and 14 percent of children 12- to 17-years-old reported illicit drug and alcohol use in the past 30 days, respectively. As a consequence, a large sum of the United States' budget is devoted to improve the social and educational outcomes of youths. The U.S. Federal government alone allocated \$14.5 billion to *Title I program*, the largest program under *No Child Left Behind*, and over \$500 million to *The Safe and Drug Free Schools Program* in fiscal year 2009.² In addition, many policy makers, practitioners and researchers have tried to develop programs whose main objective is to curb adolescents' problem behaviors such as drug use, and school dropout.

According to psychologists, there is a bidirectional relationship between risky behaviors and poor academic performance (Grant *et al.*, 2011; Guttman *et al.*, 2002; and Symons *et al.*, 1997). Moreover, economists have found that engaging in a variety of risky behaviors leads to poor academic achievement (Yamada *et al.*, 1996; Cook and Moore, 2000) and vice-versa (Sander, 1995a, b, de Walque, 2004; Kenkel *et al.*, 2006; and Grimard and Parent, 2007; and Jensen and Lleras-Muney, forthcoming). Thus, an interesting and timely policy question given today's limited resources is to ask whether it is more effective to target adolescents with high-risk of problem behavior or those at the margin of socially behaving. Empirical assessment of this question has proven difficult for the following two reasons. First, the way

¹ This ranking is based on the following 3 indicators of risk taking: 15-year-olds who smoke regularly, 13- and 15-year-olds who report having been drunk on more than two occasions, and rates of birth to females aged 15 to 19.

² These two programs aim at improving primary and secondary education of disadvantaged students and preventing violence and illegal use of tobacco, alcohol, and drugs by students, respectively.

psychologists study risk taking relies on asking individuals to respond to hypothetical dilemmas under conditions designed to minimize emotional influences on decision making. However, as Steinberg, 2004, points out: "...in the real world the risky, or potential risky, situations in which adolescents find themselves are anything but hypothetical....and they most likely occur under conditions of emotional arousal." Second, non-experimental evaluations of remedial programs are likely to bias the results since the reasons for which individuals self-select into programs are likely to be correlated with the underlying determinants of their outcomes.

In this paper we avoid the latter problem by using data from a randomized experiment in which some low-performing high-school students from low-performing high-schools in the US were offered mentoring, educational services and financial rewards during the 4 years of high-school (plus one additional year in case students fell behind one grade) while others were not (the control group).³ In addition, because the data collection included 3 surveys conducted 5, 7 and 10 years after random assignment on four key domains—high-school, post-secondary education, labor market, and risky behaviors—, we are able to use the control group's self-reported drug use at age 19 to estimate predicted probabilities of problem behavior. We then evaluate whether the program was effective for two distinct groups defined by whether their predicted probabilities of drug use at age 19 was below or above the control group's median. As young women may have more self-discipline (Duckworth and Seligman, 2006), be more likely to delay gratification (Silverman, 2003), or have lower discount rates than young men (Warner and Pleeter, 2001), implying a differential heterogeneity effect in drug use, the assignment to the top- or bottom-half of the risk distribution is done separately by gender.⁴

The research design in this paper is based on comparison of youths with high- (or low-) predicted chance of drug use who were randomly assigned to a treatment and control group when they first entered high-school in the *Quantum Opportunity*

³ Enrollees who graduated from high-school on time received some mentoring and assistance in enrolling in post-secondary education or training between graduation and the end of the fifth year of the experiment.

⁴ Because gender may lead to heterogeneity effects in programs effectiveness (Rodríguez-Planas, 2012), the analysis of the effectiveness of QOP was also done by gender. Results were similar to those presented in the main text, and available from the author upon request.

Program (QOP thereafter) social experiment, which was implemented at 11 high-schools across 7 sites in the United States. QOP's goals were to help youths graduate from high school and enroll in post-secondary education or training, and reduce risky behaviors. Students entering 9th grade in September 1995, except for those with an 8th grade point average (GPA) above the 66th percentile, were randomly assigned *within each school* to a treatment group or a control group. It is thus, important to note that the program targeted disadvantaged youths from low performing schools, and thus its targeted population had major difficult barriers to overcome, such as, substance abuse problems, or broken families. Over five years, the treatment group was offered substantial cash awards and an array of support services including mentoring by full-time social workers, academic tutoring and life skill activities. The control group was only eligible for standard youth programs offered in the community. The final sample for the QOP experiment consists of 1,069 students, 580 in the QOP group and 489 in the control group.⁵ Rodríguez-Planas (*forthcoming AEJ: Applied Economics*) has estimated the average effects of the program and found that while the program was regarded as successful in the short-term, in the long-run its educational results were modest and its effects on risky behaviors detrimental.

By analyzing whether the program worked best among those with high-predicted risk of problem behavior, this study shows that QOP was extremely successful in improving educational and behavioral outcomes for those most at risk, that is, youths in the top-half of the predicted drug use distribution. For this group, QOP increased high-school graduation by about 20 percent and college enrollment by 28 percent. Most importantly, these results persist over time, up to five years after the end of the program. Evidence shows that the program curbed risky behaviors (including substance abuse) among this group during high-school. QOP's mentors were social workers trained to identify and deal with youths' many structural barriers. They addressed any problems in any aspect of the treated youths' life, monitored their

⁵ The final sample includes a slightly larger number of youths in the treatment group (580) than in the control group (489) by design of the evaluation sample, where it was decided that each site would have either 50 or 100 youths in the treatment group (with the exception of the Washington D.C. site, which had 80 youths). The design of the evaluation sample is thoroughly explained in Rodríguez-Planas (*forthcoming AEJ: Applied Economics*).

progress, and advocated for them in matters pertaining to school, family, the juvenile justice system, and college. The mentor was expected to model appropriate behavior and attitudes, set disciplinary standards, and be continually available. Our estimates suggest that mentors succeeded in addressing treated youths' most severe problems at least while the program was operating, and that by doing so, they also succeeded in improving their high-school and college outcomes both in the short-, medium- and long-run. In contrast, we find suggestive evidence that QOP was unsuccessful among youths in the bottom-half of the distribution because it increased their engagement in risky behaviors while in high-school, and had no effect on educational outcomes in the short-, medium-, or long-run. The evidence presented is suggestive that negative peer effects may explain this lack of beneficial effects among this group. Finally, negative peer effects also seem to explain the longer-run detrimental effects of QOP on risky behaviors.

Recently, several researchers have found that targeting disadvantaged children before formal schooling begins promotes economic efficiency as the cognitive ability and character of children 0- to 5-years-old is the most malleable (Currie, 2001; Krueger and Whitmore, 2001; Heckman, 2000; Garces *et al.*, 2002; and Carneiro and Heckman, 2003). In this paper we take a different approach and contribute to the debate over whether investments in the later stages of a child's development have positive payoffs. While other rigorous studies have evaluated interventions targeted to youths at a relatively late stage of schooling (Machin, McNally and Meghir, 2004; and Lavy and Schlosser, 2005; Angrist and Lavy, 2009; and Rodríguez-Planas, *forthcoming AEJ: Applied Economics*), this paper is (to the best of our knowledge) the first one to explore the effectiveness of an intervention when targeting adolescents based on the predicted probability of problem behavior.⁶ Our findings provide evidence that QOP may be closing the gap between those most at-risk and those at the margin, and the mechanisms through which this may be.

⁶ To the best of my knowledge, Angrist and Lavy, 2009, are the only ones to analyze whether targeting based on the *predicted or fitted* chance of high-school matriculation certification is effective. Using a school-based randomization design offering cash incentives to all who passed their high-school matriculation certificate (Bagrut) in Israeli low-performing treated schools, the authors find that the experiment led to a substantial increase in certification rates for girls with high predicted Bagrut rates relative to other girls in the sample.

This paper is organized as follows. The next section describes the program implementation, the evaluation framework, and the data. Section III analyzes the results, and section IV concludes.

I. The Quantum Opportunity Program

A. Program Description

QOP engaged treated students in: (i) developmental activities intended to develop their social and employment-readiness skills; (ii) community service activities intended to develop their sense of community membership; and (iii) educational services designed to improve their academic performance. Examples of such activities are displayed in Appendix Table A.1. These activities were to be performed after school and during one half day on the weekend. The full treatment consisted of 750 hours of services per year (equally distributed among the three different activities), which (if achieved) would have represented around three-quarters of the hours required for in-school instruction per year.⁷

To encourage participation, QOP offered financial incentives to treated students. First, they received \$1.25 for each hour devoted explicitly to educational activities, developmental activities (excluding recreational activities) and community service. Second, they were promised if they obtained a high school diploma or GED *and* enrolled in post-secondary education or training they would receive an amount equal to their total earnings. The treated youths received on average over \$1,000 after high-school graduation and enrollment in post-secondary education. This aspect of the financial reward was partly motivated by evidence that a \$1,000 reduction in college tuition increases college enrollment by about 4 percentage points (Susan Dynarski 2003).

⁷In 2000, the average number of instructional hours spent in public school by 15-year-olds was 990 hours (U.S. Department of Education 2005; Table 26-2).

Case managers with prior expertise in social services were hired for the five-year period.⁸ Each had a caseload of only 15 to 25 treated students. The objective was to develop a highly personal, long-lasting connection with each youth that mirrored the relationship between a teenager and a nurturing, supportive older relative. Therefore, the case manager was instructed to focus on sustaining a strong relationship with the youth regardless of behavior, including whether the youth disengaged from the program, dropped out of school, became incarcerated, or moved out of the area. Case managers were also to manage the provision of supportive services to address personal, family and social barriers that might interfere with the treated student's ability to attend school and do well there.⁹

At almost \$25,000 per enrollee for the whole intervention, QOP was an expensive program. By comparison, the operating costs of the likewise-expensive Job Corps were approximately \$16,500 per participant in 1998 (Schochet, Burghardt, and McConnell 2008).

B. Average Results

This paper is closer to Rodríguez-Planas (*forthcoming AEJ: Applied Economics*) in that it analyzes the same randomized evaluation. Rodríguez-Planas estimates average effects of QOP and finds that the average beneficial effects of the program on high-school graduation were short lived, and that the program did *not* reduce risky behaviors during high-school. Although the program did managed to increase post-secondary education, many question whether the outcomes of the program warrant its intensity and high costs.

Since both psychologists and economists find evidence of a strong link between substance abuse and school performance, and given that QOP was mainly a mentoring program designed to identify and deal with students' many structural barriers--including substance abuse--, this paper analyzes whether the program had differential

⁸ Most mentors stayed with the program for several years and many stayed for the entire five years of the experiment. Unfortunately, no information on sex, race or ethnicity of mentors was collected.

⁹ These barriers could be addressed either directly by the case manager or by referral to a community resource, such as a substance abuse program or local agencies that provide housing, food, income support, or child care.

effects according to whether treated students were in the top- or bottom-half of the predicted drug use distribution at age 19. By doing so we provide novel valuable evidence that QOP worked extremely well for those most at risk, and offer additional information on the mechanisms through which the program may have worked (or failed to do so).

C. Evaluation Framework

The estimates reported are intent-to-treat (ITT) effects that make no adjustments for remaining involved or service participation in QOP. They are computed from estimating the following equation:

$$(1) \quad Y_{ist} = \alpha + \beta_1 D_t^T + \beta_2 X_{ist0} + \delta_s + \varepsilon_{ist}$$

where Y_{ist} denotes an outcome variable for youth i from school s at time t , D_t^T is a dummy variable that takes value one if the youth belonged to the treatment group, and X_{ist0} is a vector of controls including an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eight-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.¹⁰ δ_s are school dummies, and ε_{ist} is the error term. β_1 measures the treatment effect of the program's impact on outcome Y . Many of the outcome variables included in this study are binary. In such cases, we estimate a linear probability model and report the marginal effect of D_t^T for the impact of QOP on outcome Y .¹¹ As in Rodríguez-Planas (*forthcoming*) we decided against clustering the standard errors at the school level, which was the unit of randomization, as we were concerned that clustering would under-estimate standard errors as the

¹⁰ Since treatment was assigned randomly, the insertion of these covariates improves estimation precision without affecting the consistency of the parameter of interest. Precision is improved because the inclusion of the covariates accounts for chance differences between treated and control groups in the distribution of pre-random assignment characteristics, as well as for non-random attrition in the follow-up surveys.

¹¹We replicated our analysis using a Probit model and find very similar results.

number of schools is limited. Clustering standard errors at the school-level does not affect the main results as shown in the robustness section.

The analysis focuses on several outcomes from four key families of outcomes, high-school, post-secondary education, labor market, and risky behaviors, measured at three different points in time. The results are presented separately for youths at the bottom- and top-half of the predicted drug use at age 19 distribution. To address the problem of multiple inference, we follow Kling, Liebman, and Katz 2007 and construct a summary index, Y^* , as the unweighted average of all standardized outcomes within a family. That is, $Y^* = \frac{\sum_k Y_k^*}{k}$, where $Y_k^* = \frac{Y_k - \mu_k}{\sigma_k}$. Let Y_k be the k^{th} outcome of K variables within each family, with the sign for adverse outcomes (substance abuse, crime, welfare) reversed, so that a higher value of the normalized measure represents a more “beneficial” outcome. Standardization is performed using mean (μ_k) and standard deviation (σ_k) for the control group. The summary indices presented in this paper are constructed using the same specific outcome variables as in Rodríguez-Planas.

To classify youths by their risk of problem behavior we use a predictive regression that models the probability of problem behavior as a function of school dummies and individual covariates. The predictive model is:

$$Y_{is} = \alpha_0 + X_{is0}'\beta + \delta_s + \varepsilon_{is} \quad (2)$$

where X_{is0} and δ_s are as in equation (1), and Y_{is} a dummy variable that takes value one if the youth reported having used drugs in the past 30 days during the first survey, which was collected at age 19. Since model (2) is used to classify treated and control youths, we first estimate it separately for males and females using *only youths from the control group*. Appendix Table A.2 reports the coefficients estimated using equation (2). Males are more likely to have problem behavior than females as shown by the descriptive statistics at the top of Appendix Table A.2. We then estimated equation (1) using the following subgroup-classification scheme: we used the fitted values from model (2) to divide students into roughly equal-sized groups, those in the top and the bottom of the predicted-risk distribution. This classification was done by gender, as

the predicted drug use during high-school of females was considerably lower than that of males.¹²

D. The Data and Descriptive Statistics

The data for this study come from baseline information collected prior to random assignment, program implementation and participation, high-school transcripts, math and reading achievement tests conducted 4 years after random assignment, and three surveys conducted 5, 7 and 10 years after random assignment. Transcript data were collected 5 years after random assignment from *all* high-schools treated and control group individuals attended.

Table 1 reports means and differences in means by treatment status for our baseline variables.¹³ Columns 1 to 3 refer to the whole sample. They show that the sample is roughly equally divided between boys and girls. The median age of students was 14 when random assignment took place. The sample consists mainly of minority students: about two thirds are African-American and about one fourth are Hispanic (black or nonblack). There are no significant differences by treatment status.

Columns 4 to 9 present treatment and control means and differences for those in the top- and bottom-half of the predicted drug use distribution. Among youths in the bottom-half of the predicted risk distribution we observe that treated youths are less likely to be males and in the middle tier of the 8th grade GPA distribution than those in the control group. To account for these differences, estimation of equation (1) controls for all baseline characteristics as explained above.

Control response rates and treatment-control differences for the achievement test completion and the follow-up surveys are displayed in Table 2.¹⁴ The response rate to the achievement tests and the first telephone survey was 80 percent for the control group. The response rates to the second and third telephone surveys were 69 and 72

¹² For males, those with a predicted probability of drug use above 35 percent were classified in the top-half of the risk distribution, whereas the threshold for females was 17 percent.

¹³ As no baseline survey was collected, the pre-program information available is limited. However, the main baseline variable that is missing (compared to similar evaluations) is parent's education level. Fortunately, this variable is likely to be correlated with pre-program GPA, which we do have.

¹⁴ All individuals from the treatment and control group were surveyed, regardless of whether they failed to respond in one of the earlier surveys

percent for the control group, respectively. Attrition was higher among control group members than among treatment group members in all but the third telephone survey. The differential response rate between the treatment and control groups is 7 percentage points in the first telephone survey, 10 percentage points in the second telephone survey, and a statistically insignificant 3 percentage points in the third telephone survey. After dividing the sample by the predicted probability of drug use at age 19 we observe that most of the attrition differential between treated and control youths is driven by those in the top-half of the risk distribution. As these are the most difficult youths, higher attrition among them is expected. In addition, because those in the treatment group have a relationship with the mentor, they ought to also have been easier to contact in the follow-up surveys leading to a higher response rate relative to those in the control group. At the end of the results section, we discuss the consequences of differential attrition and non-response.¹⁵

The only available measure of program participation is the stipend or accrual accounts participants received. This is problematic as mentoring time did not count toward stipends or accrual account contributions. In addition, enrollees received “bonus hours” when they achieved a significant milestone, such as obtaining a B average or higher on his or her report card. These bonus hours cannot be distinguished from regular hours. Their inclusion overestimates time spent on program activities for some enrollees.

¹⁵We were also able to use transcripts data and exploit the fact that survey non-respondents did not necessarily overlap across surveys to build two variables measuring high-school completion and GED diploma receipt for most of the population at each point in time. We recoded as high-school graduates those survey non-respondents who had reported in an earlier survey being a high school graduate or for whom their transcript data clearly confirmed that they had indeed graduated from high-school. Similarly, we recoded as high-school dropouts those first- (or second-) survey non-respondents who had later responded to the second or third surveys and who had reported having dropped out of high-school. In addition, in the few cases where we found inconsistencies across surveys or across survey and transcripts, we used the answer that was corroborated at least twice. Several robustness checks have been performed using alternative definitions of high-school graduation and the results are robust to those reported in the main tables. For these two variables, differences in the response rates are considerably smaller (although they remain statistically significant at the 5 percent level as shown in Table 2).

E. Implementation of QOP and Service Use

Overall, QOP was successfully implemented. Most case managers reported developing close mentoring relationships with the majority of the individuals assigned to them, and they all provided access to services regardless of enrollees' behavior.¹⁶

Although the educational, community service and development activity component fell short of the target of 750 hours per year, enrollees still invested a substantial amount of time in QOP activities during the first four years. The average of 886 hours, including summers but not time spent with their mentors, corresponds to about 89 percent of an extra school year and more than half the average instruction time received by Job Corps participants, by far the most intensive education and training program for disadvantaged youths in the United States (Schochet *et al.*, 2008). The fact that QOP did not achieve its extremely ambitious target should not affect the external validity of this evaluation; if the program were implemented on a broader scale, it is likely that its implementation would be similar.

Enrollees spent an average of 61 hours per year on education, 76 hours on developmental activities, and 27 hours in community service.¹⁷ Not surprisingly, the average time spent on QOP activities fell steadily from 279 hours in the first year of the experiment to 125 hours in the fourth year (see Appendix Table A.3). Youths who had participated in QOP activities early during the experiment and then stopped or decreased participation over time gave leaving school, working, and family responsibilities as the main reasons for doing so. It is important to highlight, however, that almost the totality of QOP youths engaged in QOP activities, as all but 1 percent of enrollees spent some time on QOP activities in the first year.

Appendix Table A.3 also shows the distribution of hours over activities and year by whether youths were in the bottom- or top-half of the predicted drug use distribution. It is important to highlight that youths in the top-half of the risk distribution spent

¹⁶ See Maxfield *et al.* 2003a, and Maxfield *et al.*, 2003b, for further description of program design and implementation.

¹⁷In the case of community services, the lower intake was due to enrollees' lack of interest in this type of activities and case managers' belief that enrollees had a greater need for other QOP services. Most sites decided to reallocate their resources away from community service to developmental and educational activities.

slightly more hours in QOP activities than those in the bottom-half (albeit the difference is not statistically significant). In addition, although youths in the top-half of the risk distribution spent less hours during the first year than those in the bottom-half, the opposite is true during the fourth year, suggesting a more persistent involvement with QOP. Indeed we observe that while a higher percentage of them (44 percent) were no longer involved in QOP during the fourth year, of those who were still involved in QOP, the intensity of involvement was higher among those in the top-half of the risk distribution.

II. Results

A. Results

Tables 3 through 6 display results for the summary indices followed by each specific outcome that were components of the index. Each table covers each of the key domains under analysis, high-school, post-secondary education, labor market, and risky behaviors, measured in the short-, medium- and long-run. To draw general conclusions about the results of the experiment, we begin with the summary indices for a given category and then we discuss the individual outcomes as the magnitudes of these separate outcomes are often easier to interpret than those of the summary indices. Columns 1 to 3 present the control group outcome means for the whole sample and separately for those in the bottom- or top-half of the fitted distribution of drug use. Column 4 presents ITT estimates of the QOP effect for the whole sample. Separate ITT estimates for those in the bottom- or top-half of the fitted distribution of drug use are displayed in columns 5 and 6. Our discussion focuses on the effects of QOP on those in the top- and bottom-half of the risk distribution. It is important to keep in mind that a positive coefficient on the summary indices indicates a beneficial effect of the program, and a negative coefficient indicates a detrimental effect of QOP.

Control Group Means.— Column 1 of Tables 3 through 6 display mean outcomes for the control group. When youths were in their late-teens, we observe that close to 44 percent of those in the control group had graduated from high-school, and an

additional 7 percent had obtained a GED. At the same time, 27 percent of them were attending post-secondary education, and 19 percent were in college. In terms of risky behaviors, 19 percent reported binge drinking, 27 percent reported using illegal drugs in the past month, and a bit over one fourth reported committing a crime in the last year or ever being arrested. Three years later, when youths were in their early-twenties, both the percent of those with a high-school degree or a GED had gone up to 57 and 13.5 percent, respectively, and as many as 55 percent of youths had been enrolled in post-secondary education (34 percent in college). In terms of their labor force involvement, 73 percent were working (57 percent in a full-time job), and their average hourly wage was \$7 US dollars per hour. With the exception of binge drinking, risky behaviors had decreased: 16 percent report using illegal drugs and 8 percent report committing a crime. When we last surveyed them, 5 to 6 years after scheduled high-school graduation, we observe that 62 percent had graduated from high-school, 17 percent obtained a GED, and 56 percent had been enrolled in post-secondary education at some point in time. These averages are not far from the national average of 67 percent for high-school degree, 12 percent for GED, and 58 percent for similarly disadvantaged youths (U.S. Department of Education 2002b), but they are low compared to the national averages for the whole population discussed in the introduction. In addition, about 38 percent of youths in the control group had attended college, 30 percent had completed 2 years of college, and 7 percent had obtained a bachelor's or associate degree. Most of them (70 percent) were working at the time of the interview and their yearly earnings were, on average, \$13,470 US dollars. With age, youths' drug use and crime continues to go down with 12 percent reporting having used drugs in the past month and 2 percent having reported committing a crime.

Columns 2 and 3 of Tables 3 through 6 display control means by whether youths in the control group were in the top- or bottom-half of the predicted drug use distribution. We observe that average drug use for youths in the top-half of the predicted drug use distribution doubles that of youths in the bottom-half. In addition, reported crime is also higher for the former than the latter. These results persist over time as youths age. Focusing now on high-school outcomes, we also observe that,

when they were in their late teens, those in the top-half of the risk distribution are about 6 percentage points less likely to have graduated from high-school than those in the bottom-half. This differential persists and increases over time: when youths are in their mid-twenties, those in the top-half are 12 percentage points less likely to have graduated from high-school than those in the bottom-half. Similarly, we observe that youths in the top-half of the risk distribution are, on average, less likely to be employed and they earn on average lower earnings than those in the top-half. In contrast, we do not observe any differences in terms of math and reading test scores when youths were in their late-teens. Similarly, we do not observe any differences in terms of enrolling in college or post-secondary education.

High-School Outcomes.— Columns 5 and 6 of Table 3 bring to light a new result: QOP worked extremely well in terms of increasing high-school graduation for youths in the top-half of the predicted drug use distribution, and this result persist over time as youths age. Column 5 of Table 3 shows a zero ITT estimate of QOP on the high-school domain for youths with *low*-predicted drug use in the short-run that becomes negative although not statistically significant when youths are in their early- and late-twenties. In contrast, the ITT estimates of QOP for youths with *high*-predicted drug use during high-school show large, significant, and persistent ITT effects of QOP on the high-school domain. The estimates of 0.099, 0.101 and 0.085 in column 6 of Table 3 indicate that the mean effect of being in the treatment group for the high-school outcomes is one tenth to one thirteenth standard deviations for youths in the top-half of the predicted drug distribution. The individual outcomes reveal that these beneficial effects are driven by statistically significant increases in the incidence of graduating from high-school (by about 10 percentage points). As the high-school graduation rate in the control group among those in the top-half of the fitted distribution is 40 percent in the short-run, 51 percent in the medium-run and 55 percent in the long run, the treatment effect corresponds to a 24 percent increase when youths were in their late-teens, 20 percent increase when youths were in their early-twenties, and 18 percent increase when youths were in their mid-twenties. Because the control group means of those in the top-half of the distribution are 6, 10 and 12 percentage points lower than those in the bottom-half, it appears that QOP reduced

treated youths differences in terms of high-school graduation by increasing the performance of those with a high predicted risk of drug use while in school towards the level of those with a lower predicted risk.

College Outcomes.— Moving now to Table 4, we observe that the only statistically significant ITT estimate of the college domain is that of youths in the top-half of the risk distribution measured when youths were in their early-twenties (shown in column 6 of Table 4). For youths in the top-half of the risk distribution, the other estimates are not statistically significant although the size of the coefficients are quite large (and in the medium- and long-run double the estimates measured for youths in the bottom-half of the distribution). The individual outcomes reveals that the positive and statistically significant effect in the medium-run is driven by an 7.9 percentage points (or 24 percent) increase in the odds of attending college. This effect persist over time: by the time youths are in their mid-twenties, QOP increased the likelihood of ever attending college of those in the top-half of the risk distribution by 10.1 percentage points (or 28 percent). As there are few differences in the control group means between those in the bottom- and top-half of the risk distribution, QOP did not narrow the gap between the two groups in terms of post-secondary educational outcomes.

Employment Outcomes.—Information on employment was first collected at the time of the 2nd survey, when youths were in their early twenties. For results measured at that time, we observe a negative and statistically significant effect at the 5 percent level effect for males with low-predicted chance of using drugs during high-school. The ITT estimates of -0.205 in column 5 of Table 5 indicates that the mean effect of being in the treatment group for the six employment outcomes is one fifth of a standard deviation for those with low-predicted probability of using drugs. The individual estimates show that these treated youths were 11.5 percentage points *less* likely to have a job than youths in the control group, and that they worked on average about 5 less hours per week than youths in the control group. These effects are significant at the 5 percent level. These negative effects on employment may be explained by the fact that treated youths are more likely to be enrolled in post-secondary education (although effects on post-secondary attendance for this group are

not statistically significant as shown in column 5 of Table 4). It is also interesting to note that we do not observe negative effects on employment when youths were in their early twenties for those in the top-half of the risk distribution despite finding that QOP increased both their post-secondary education and college enrollment. By the time youths are in their mid-twenties, the negative ITT effect on the employment domain of youths in the bottom-half of the risk distribution is considerably smaller and no longer significant.

Risky-Behavioral Outcomes.— QOP managed to curb risky behaviors among those most at-risk at least while the program was operating. When youths were in their late-teens, the ITT estimate for the summary index in column 6 of Table 6 indicate a positive effect of 0.102 (significant at the 90 percent level) for youths in the top-half of the risk distribution. The individual variables indicate that QOP's positive effect on reducing risky behaviors for this group is driven by all of the individual estimates, albeit being measured with less precision. In contrast, the effect is -0.162 for those with low-predicted risk of drug use (significant at the 99 percent level) reveals that QOP increased risky behaviors among this group during their late-teens. This is driven by an increase in the likelihood of binge drinking, using drugs in the past month, and committing in crime in the past year of 12.2, 16.5, and 9.5 percentage points, respectively. As the control group means of those in the bottom-half of the predicted drug use distribution are 6 to 20 percentage points lower than those in the top-half, it appears that QOP again narrowed the gap between the two groups.

Over time, QOP's positive effects on reducing risky behaviors fade away. By the time youths were in their early-twenties, the ITT estimates for the risky behaviors domain are considerably smaller and no longer statistically significant. The only beneficial effect of QOP that remains is that the program continued to be effective at reducing drug use for youth with a high-risk of drug use. QOP decreased drug use among this group by 8.4 percentage points (or 40 percent)--statistically significant at the 95 percent level. It is important to note, though, that engagement in risky behaviors decreases as youths age and is at its peak when youths are in their late-teens. Thus, the fading positive effect of QOP on risky behaviors when youths are in their

twenties does not necessarily imply that QOP failed in its objective of reducing risky behaviors among youths in the top-half of the risk distribution.

QOP's negative effect on risky behaviors for youths in the bottom-half of the risk distribution also fades away when youths are in their early-twenties. At that point in time, the ITT estimate is close to zero and not statistically significant. In addition, the coefficients for the individual estimates are also close to zero (not significant) and sometimes negative (implying a reduction instead of an increase in substance abuse).

By the time youths are in their mid-twenties, we observe a negative and significant effect of QOP on the risky behaviors' domain among youths in the bottom-half of the risk distribution. This effect is driven by a 5.4 percentage points higher likelihood of arrests (significant at the 5 percent level). For youths in the top-half of the risk distribution, we also observe a negative (albeit non significant) effect of QOP. This is not the only intervention to find unanticipated adverse effect on crime in the long-run. Kling *et al.*, 2005 also find that the Moving to Opportunity (MTO) housing vouchers led to short-term reductions in violent crime arrests, but long-term *increases* in property crime arrests for males. Similarly, the randomized, experimental trial of the community-based treatment program *Cambridge-Somerville Youth Study*, whose objective was to prevent delinquency, found that youths in the treatment group were more likely in the long run to be rearrested for crimes and have further negative impacts on physical and psychological health compared to youths in the control group (McCord, 1978 and 1992). The impacts were measured up to 30 years later using official state records.

Potential Mechanisms.— Unfortunately, the evaluation of QOP was not designed to test alternative mechanisms. However, the analysis thus far reveals that QOP worked very well for youths with a high-predicted risk of drug use as it decreased their engagement in risky behaviors during high-school, and increased high-school graduation rate, and post-secondary educational involvement. Most importantly, the educational beneficial results persisted over time. Thus, it appears that QOP's mentors, who were social workers, did a good job at identifying youths most at-risk

and addressing their non-cognitive and structural barriers, and by doing so their educational outcomes improved. To corroborate this, we find evidence that treated youths in the top-half of the distribution became more optimistic about life at the end of the program than those in the control group. For instance, they were 6.5 percentage points (or 8 percent) more likely to believe that life will be happy, or 5.1 percentage points (or 6 percent) more likely to agree that they would *not* die before 30. Both estimates are statistically significant at the 90 percent level. In contrast, for those in the bottom-half of the distribution, QOP did not have any positive effect on their outlook of life.¹⁸

In contrast, QOP did not seem to work that well among treated youths in the bottom-half of the distribution as it had no effect on educational outcomes, and it increased engagement in risky behaviors while in high-school. So, what went wrong during implementation for youths with low-predicted chance of problem behavior? Negative peer effects are a plausible explanation. As discussed earlier, each mentor in QOP had assigned several mentees with whom he or she frequently worked at the same time as a group. In addition QOP offered cultural and recreational activities to treated youths to help them build strong relationships with mentors and peers. However, building strong cohesion within a group may have backfired. Evidence from psychologists and economists reveals that peer effects play an important role on getting youths involved in risky activities. This occurs because peers reinforce deviant conduct by responding with approval and attention (Dishion, *et al.*, 1999; Dishion, *et al.*, 1996; and Patterson, *et al.*, 2000).¹⁹

To test whether negative peer effects is a plausible story, we estimated the following peer effects variable. For each individual we estimated the proportion of students within the same school, treatment status, sex, and race, that reported doing drugs at

¹⁸ Control means for the control group in the top-half of the risk distribution are 80 percent for agreeing that life will be happy and 88 percent for agreeing that they will not die before 30. For those in the bottom half of the risk distribution, control group means are 79 percent and 91 percent.

¹⁹ See Brook *et al.*, 1998; Kandel, 1985; Jessor *et al.*, 1980 for peers' influence on marijuana use; Norton *et al.*, 1998, Jensen and Lleras-Muney, 2010 for drinking and peer effects; and Case and Katz, 1991 for peer-group interactions and criminal activity.

age 19. We then re-estimated equation (1) including this peer effect variable and its interaction with QOP. If the story is negative peer effects we would expect a negative and significant coefficient on this interaction term. Table 7 displays these estimates. Among youths in the bottom-half of the risk distribution, we observe that there is a negative peer effect for youths in the treatment group. For instance, the coefficient on the interaction between being treated and the peer variable is negative and statistically significant for the college outcome both in the short- and medium-run, and the high-school outcome in the short-run, indicating that having peers who engaged in drug use during high-school decreased the beneficial effect of QOP in this domain both in the short- and medium-run. Once we account for this peer effect, QOP improves the college outcomes in the short-run and medium-run for youths in the bottom-half of the distribution. Similarly, the estimate of QOP on risky behaviors when youths were in their late-teens, becomes positive, and the interaction between QOP and the peers effect variable is negative. Although these coefficients are estimated with less precision, this sign reversal on the treatment dummy suggests that negative peer effects among treated youths in the bottom-half of the distribution are a part of the story behind these treated youths' increased engagement in risky behaviors during high-school.

Finally, could negative peer effects explain the long-run detrimental effects of QOP on risky behaviors found earlier by Rodríguez-Planas (*forthcoming AEJ: Applied Economics*)? When looking at the whole sample, we observe that introducing our measure of peer effects based on drug use at age 19 has a negative and statistically significant impact on engagement in risky behaviors among treated youths when these were in their early- and mid-twenties. In addition, the coefficient on the treatment dummy becomes positive and significant in the medium-run and positive (albeit not significant) in the long-run. Table 7 also shows that these negative peer effects on risky behaviors are particularly large among youths in the top-half of the risk distribution. Thus, it seems that while mentors were able to curb risky behaviors among those at high-risk during high school, they did not succeed in changing behavior in the long run. Notice that while there are potentially alternative and

complementary explanations for this result, one would still need to explained why they were driven by those who had a higher percentage of peers during high-school doing drugs. The bottom line is that the evidence in Table 7 is suggestive that negative peer effects may well be behind QOP's detrimental long-run effect on risky behaviors.

B. Robustness Section

Clustering Standard Errors at the School Level.— Appendix Table A.4 presents estimates for each of the four domains clustering the standard errors at the school level, which was the unit of randomization. When doing so standard errors are usually a bit smaller and thus it increases the precision of our estimates. Clustering standard errors at the school-level does not affect the main results presented thus far.

Attrition from the Follow-up Surveys.—As we saw in the data section, differential attrition between treated and control groups exists especially among youths in the top-half of the predicted drug use distribution. Here, we analyze the sensitivity of the estimates to potential attrition bias. First, we set the response rate for the QOP group equal to the response rate for the control group *within* each of the 11 QOP schools and by whether youths were in the top- or bottom-half of the risk distribution. That is, we eliminated the last (and thus the most difficult to find) treatment group respondents until the “response rates” of the control and treatment groups were equalized at the school and top- (or bottom)-half of the risk distribution. The estimates obtained in this way (see Appendix Table A.5) are similar to those presented earlier, suggesting that higher response rates among the treatment group youths are not driving the results.

In addition, since we were able to construct high-school completion and GED diploma receipt for a considerably more complete sample using transcript data (as explained in footnote 15 in Section I.D and shown in Table 2), we compare these estimates of QOP (discussed in the main text and shown in Table 3) to those estimated using the attrited sample. Appendix Table A.6 displays these results, and shows that differences across the two estimates are small.

III. Conclusion

Rodríguez-Planas (*forthcoming AEJ: Applied Economics*) estimates average effects of QOP and finds that although the program was successful in the short-term, in the long-run its educational results were modest and its effects on risky behaviors detrimental. As a consequence, many question whether the outcomes of the program warrant its intensity and high costs. This paper takes a different approach and finds that QOP worked extremely well among youths with a high-predicted risk of drug use. By doing so, we unveil how QOP may have worked: it managed to curb risky behaviors among those most at risk during high-school and thus, it improved their educational outcomes both in the short-, medium-, and long-run. In contrast, we find suggestive evidence that QOP was unsuccessful among youths in the bottom-half of the distribution because it increased their engagement in risky behaviors while in high-school. The evidence presented seems to suggest that negative peer effects may explain this. Perhaps more importantly, we also find evidence suggesting that negative peer effects may also be behind the puzzling negative effects of QOP on long-run risky behavioral outcomes.

While we explored alternative explanations for these results, the evidence we found was weaker. For instance, an alternative explanation is that because QOP gave stipends, youths may have used this additional income to purchase alcohol and drugs.²⁰ However, on average treated youths got \$350 the first year and \$156 the fourth year. Per week, this represents \$6 and less than \$3 in the first and fourth year, respectively. While \$1.25 per hour may have encouraged participation, overall youths got much less than minimum wage, which went from \$4.25 in 1995 to \$5.15 in 1999. Thus, it is unlikely that QOP's stipends alone would explain the observed substance abuse. Another explanation is that QOP may have led to a substitution away from parents' attention, either because they trusted that the mentor was also watching over their children, or because QOP unintentionally weakened the ties between parents and enrollees, breaking important social bonds and thus leading to the observed perverse

²⁰ There is growing evidence showing that youths are very responsive to economic incentives, such as prices, when deciding to undertake risky behaviors (Nisbet and Vakil, 1972; Grossman *et al.*, 1994; Chaplounka and Wechler, 1996; Gruber, 2001; Pacula *et al.*, 2001).

effects. And that this happened mainly among youths in the bottom-half of the risk distribution (as those in the top-half did not have those strong family ties to start with). To explore whether this could be a possible explanation we re-estimated equation (1) adding a dummy equal to 1 if the youth reported having an influential adult during high-school and the interaction of this variable with the treatment effect.²¹ If the negative effect on risky behaviors for youths in the bottom-half of the distribution were caused by a substitution effect, we would find that this detrimental effect of QOP on risky behaviors would be partly explained by this interaction. We found no evidence of that. In the contrary, estimates suggests that, if anything, mentors decreased youths' engagement in risky behaviors.²²

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²¹ Clearly including having a dummy indicating presence of an influential adult while in high-school is endogenous, however, we do this only to see how much of the estimated causal detrimental effect of QOP could be driven by an influential adult substituting parents' attention. We find that if anything mentors decreased youths' engagement in risky behaviors.

²² The coefficient on the QOP dummy in the risky behaviors domain becomes -0.272 (s.e. 0.110) and the interaction between QOP and having an influential adult is 0.163 (s.e. 0.129) for youths in the bottom-half of the distribution. Similarly, for youths in the top-half of the distribution, the coefficient on the QOP dummy becomes 0.015 (s.e. 0.122) and the interaction term becomes 0.142 (0.143).

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TABLE 1. DESCRIPTIVE STATISTICS BY PREDICTED DRUG USE AT AGE 19

	<i>Low-risk of drug use at 19</i>					<i>High-risk of drug use at 19</i>			
	<i>Treatment means</i>	<i>Control means</i>	<i>Treatment - Control</i>	<i>Treatment means</i>	<i>Control means</i>	<i>Treatment - Control</i>	<i>Treatment means</i>	<i>Control means</i>	<i>Treatment - Control</i>
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
<i>Baseline Covariates</i>									
Male	0.522	0.558	-0.040 (0.030)	0.502	0.587	-0.078** (0.039)	0.544	0.526	-0.016 (0.038)
Age when entering 9 th grade									
< 14	0.107	0.110	-0.000 (0.019)	0.084	0.101	-0.022 (0.024)	0.132	0.121	0.018 (0.028)
14	0.533	0.575	-0.040 (0.030)	0.552	0.568	0.007 (0.039)	0.512	0.582	-0.069 (0.043)
> 14	0.360	0.315	0.041 (0.028)	0.364	0.331	0.015 (0.039)	0.356	0.297	0.050 (0.040)
Hispanic	0.262	0.257	0.005 (0.016)	0.411	0.381	0.023 (0.022)	0.103	0.121	0.001 (0.018)
Black	0.683	0.679	0.004 (0.013)	0.555	0.564	-0.003 (0.014)	0.818	0.806	0.001 (0.019)
Rank based on 8 th grade GPA									
Bottom third	0.365	0.329	0.037 (0.029)	0.415	0.361	0.048 (0.041)	0.313	0.293	0.015 (0.041)
Middle third	0.307	0.352	-0.044 (0.029)	0.224	0.288	-0.067* (0.037)	0.395	0.422	-0.018 (0.043)
Top third	0.328	0.319	0.008 (0.029)	0.361	0.350	0.019 (0.041)	0.292	0.284	0.003 (0.040)
<i>Additional Variables</i>									
Peers' drug use at 19	28.505 (13.158)	21.063 (13.581)	0.071*** (0.008)	26.916 (12.912)	15.554 (9.557)	0.112*** (0.009)	30.194 (13.230)	27.166 (14.739)	0.025** (0.012)
Predicted drug use at 19	0.272	0.274	-0.011 (0.007)	0.175	0.190	-0.017** (0.007)	0.375	0.368	0.004 (0.010)
<i>Sample size</i>	<i>580</i>	<i>489</i>	<i>1,069</i>	<i>299</i>	<i>257</i>	<i>556</i>	<i>281</i>	<i>232</i>	<i>513</i>

Note: Robust standard errors are reported in parentheses in columns 3, 6 and 9. Columns 3, 6 and 9 presents the coefficient on the Treatment dummy from a regression model with the Treatment dummy plus school dummies.

*, ** Estimate significantly different from zero at the 90%, or 95% confidence level.

TABLE 2. TREATMENT EFFECT ON RESPONSE RATES BY PREDICTED DRUG USE AT AGE 19

	<i>Low-risk of drug use at 19</i>				<i>High-risk of drug use at 19</i>				
	<i>Control means</i>	<i>Treatment - Control</i>		<i>CONTROL MEANS</i>	<i>TREATMENT-CONTROL</i>		<i>CONTROL MEANS</i>	<i>TREATMENT-CONTROL</i>	
		<i>School dummies</i>	<i>Full set of controls</i>		<i>School dummies</i>	<i>Full set of controls</i>		<i>School dummies</i>	<i>Full set of controls</i>
OUTCOMES									
Achievement tests	0.800	0.080*** (0.023)	0.081*** (0.022)	0.809	0.055* (0.032)	0.0512 (0.0317)	0.789	0.112*** (0.0322)	0.113*** (0.0317)
First telephone survey	0.795	0.071*** (0.023)	0.072*** (0.023)	0.817	0.028 (0.032)	0.0263 (0.0328)	0.772	0.117*** (0.0328)	0.119*** (0.0329)
Second telephone survey	0.685	0.096*** (0.027)	0.097*** (0.0273)	0.689	0.054 (0.039)	0.051 (0.039)	0.681	0.146*** (0.038)	0.148*** (0.038)
Third telephone survey	0.724	0.032 (0.027)	0.0326 (0.0270)	0.755	-0.010 (0.067)	-0.013 (0.037)	0.690	0.081** (0.040)	0.083** (0.040)
<i>High-school and GED recodes</i>									
First survey	0.926	0.031** (0.015)	0.032** (0.015)	0.934	0.010 (0.021)	0.010 (0.021)	0.918	0.053** (0.021)	0.054*** (0.021)
Second survey	0.856	0.046** (0.020)	0.049** (0.020)	0.872	0.004 (0.029)	0.008 (0.029)	0.841	0.094*** (0.028)	0.099*** (0.028)
Third survey	0.836	0.046** (0.021)	0.050** (0.021)	0.868	-0.007 (0.029)	-0.005 (0.029)	0.802	0.107*** (0.031)	0.113*** (0.031)

Notes: Robust standard errors are reported in parentheses. "Full set of controls" includes school dummies, an indicator for being male, an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eighth-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.

** , *** Estimate significantly different from zero at the 95%, or 99% confidence level.

TABLE 3. TREATMENT EFFECT ON HIGH-SCHOOL OUTCOMES BY PREDICTED DRUG USE AT AGE 19

OUTCOMES	CONTROL MEANS			TREATMENT-CONTROL		
	Full sample	Low-risk	High-risk	Full sample	Low-risk	High-risk
	(1)	(2)	(3)	Full set of controls	Full set of controls	Full set of controls
	(1)	(2)	(3)	(4)	(5)	(6)
<i>Youths Were in their Late Teens</i>						
<i>Summary measure</i>				0.056* (0.034)	0.0031 (0.049)	0.099** (0.047)
Obtained high-school Diploma	0.435	0.463	0.404	0.053* (0.029)	0.007 (0.041)	0.095** (0.041)
Obtained a GED	0.076	0.061	0.093	-0.009 (0.017)	-0.016 (0.021)	-0.001 (0.028)
Obtained HS diploma or GED or still in HS	0.636	0.662	0.606	0.043 (0.028)	-0.022 (0.039)	0.098** (0.041)
Math test scores (percentile)	40.541 [7.718]	40.741 [7.968]	40.316 [6.199]	0.368 (0.426)	0.632 (0.623)	-0.038 (0.557)
Reading test scores (percentile)	42.884 [7.540]	42.909 [7.849]	42.854 [7.196]	0.406 (0.484)	0.252 (0.692)	0.420 (0.647)
<i>Youths Were in their Early Twenties</i>						
<i>Summary measure</i>				0.044 (0.032)	-0.021 (0.043)	0.101***† (0.048)
Obtained high-school Diploma	0.568	0.616	0.513	0.032 (0.030)	-0.045 (0.041)	0.102***†† (0.043)
Obtained a GED	0.136	0.099	0.180	0.008 (0.023)	0.019 (0.029)	-0.002 (0.038)
<i>Youths Were in their Mid-Twenties</i>						
<i>Summary measure</i>				0.024 (0.030)	-0.038 (0.042)	0.085*†† (0.044)
Obtained high-school Diploma	0.616	0.673	0.548	0.021 (0.030)	-0.054 (0.040)	0.097***†† (0.044)
Obtained a GED	0.166	0.135	0.204	0.002 (0.024)	0.012 (0.031)	-0.011 (0.039)

Notes: Standard deviation in brackets. The table reports estimates of treatment effects on the dependent variables indicated in row headings. Robust standard errors are reported in parentheses. "Full set of controls" includes school dummies, an indicator for being male, an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eighth-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.

*, **, *** Estimate significantly different from zero at the 90%, 95% level, or 99% level.

†, ††, ††† indicates that the difference of the estimated effects between youths in the bottom- and top-half of the predicted drug use distribution is significant at the 90%, 95%, and 99% level.

TABLE 4. TREATMENT EFFECT ON POST-SECONDARY EDUCATION OUTCOMES BY PREDICTED DRUG USE AT AGE 19

OUTCOMES	CONTROL MEANS			TREATMENT-CONTROL		
	Full sample	Low-risk	High-risk	Full sample	Low-risk	High-risk
	(1)	(2)	(3)	Full set of controls	Full set of controls	Full set of controls
Youths Were in their Late Teens						
<i>Summary measure</i>				0.105* (0.059)	0.094 (0.082)	0.129 (0.084)
Attending or accepted in College	0.268	0.268	0.268	0.055* (0.029)	0.024 (0.040)	0.087** (0.042)
Attending college	0.189	0.192	0.184	0.030 (0.026)	0.031 (0.036)	0.036 (0.037)
Attending postsecondary Education	0.267	0.262	0.274	0.050 (0.030)	0.066 (0.041)	0.044 (0.044)
Youths Were in their Early Twenties						
<i>Summary measure</i>				0.108* (0.060)	0.061 (0.082)	0.177** (0.083)
Ever in college	0.338	0.345	0.331	0.060* (0.033)	0.048 (0.047)	0.079* (0.046)
Number of semesters in College	0.973 [1.790]	1.042 [1.879]	0.893 [1.687]	0.050 (0.126)	-0.119 (0.185)	0.250 (0.173)
Ever in postsecondary Education	0.549 0.729	0.531 0.768	0.571 0.684	0.085** (0.035)	0.0733 (0.049)	0.111** (0.050)
Youths Were in their Mid-Twenties						
<i>Summary measure</i>				0.083* (0.049)	0.062 (0.077)	0.112 (0.077)
Obtained a bachelor's Degree	0.020	0.005	0.038	0.011 (0.011)	0.014 (0.011)	0.005 (0.02)
Obtained a bachelor's or associate degree	0.071	0.062	0.082	-0.003 (0.018)	-0.006 (0.023)	0.001 (0.028)
Number of semesters in College	1.62 [3.40]	1.66 [3.69]	1.56 [3.02]	0.148 (0.232)	-0.113 (0.319)	0.460 (0.318)
Completed 2 years of college or training	0.301	0.285	0.320	0.070** (0.033)	0.080* (0.045)	0.069 (0.048)
Ever in college	0.377	0.392	0.358	0.043 (0.033)	-0.006 (0.046)	0.101**† (0.047)
Ever in post-secondary Education	0.558	0.541	0.579	0.074** (0.034)	0.089* (0.047)	0.070 (0.050)

Notes: See notes on Table 3. Post-secondary education includes two- and four-year college, vocational or technical school, and the armed forces.

TABLE 5. TREATMENT EFFECT ON EMPLOYMENT OUTCOMES BY PREDICTED DRUG USE AT AGE 19

OUTCOMES	CONTROL MEANS			TREATMENT-CONTROL		
	Full sample	Low-risk	High-risk	Full sample	Low-risk	High-risk
	(1)	(2)	(3)	Full set of controls	Full set of controls	Full set of controls
<i>Youths Were in their Early Twenties</i>						
<i>Summary measure</i>				-0.102* (0.058)	-0.205*** (0.078)	0.017† (0.084)
Has a job	0.730	0.768	0.686	-0.042 (0.033)	-0.115** (0.046)	0.047†† (0.048)
Has a full-time job	0.575	0.614	0.532	-0.089** (0.036)	-0.134*** (0.049)	-0.044 (0.052)
Has a full-time job with health insurance	0.404	0.448	0.353	-0.059* (0.035)	-0.121** (0.049)	0.0024† (0.051)
Attending postsecondary education or working	0.809	0.836	0.776	-0.006 (0.029)	-0.052 (0.040)	0.055†† (0.042)
Usual hours worked per week at main job	28.17 [19.98]	30.324 [19.854]	25.75 [19.907]	-1.528 (2.093)	-5.091** (2.054)	1.095†† (2.075)
Hourly wage at main Job	7.57 [12.15]	7.625 [5.952]	7.561 [5.627]	-0.274 (0.493)	-0.804 (0.777)	0.296 (0.561)
<i>Youths Were in their Mid-Twenties</i>						
<i>Summary measure</i>				0.011 (0.054)	-0.026 (0.075)	0.0566 (0.080)
Has a job	0.707	0.737	0.669	0.007 (0.033)	-0.0180 (0.044)	0.0397 (0.049)
Has a full-time job	0.553	0.593	0.503	0.013 (0.035)	-0.015 (0.048)	0.048 (0.052)
Has a job with health insurance	0.491	0.531	0.442	-0.004 (0.036)	-0.037 (0.049)	0.045 (0.053)
Hourly wage at main job (dollars)	9.36 [15.97]	9.89 [14.42]	8.72 [17.72]	-0.671 (1.008)	-0.622 (1.568)	-1.185 (1.610)
Usual hours worked per week at main job	27.53 [20.656]	29.25 [20.82]	25.43 [20.32]	0.511 (1.452)	-0.569 (2.003)	1.788 (2.102)
Total earnings in past 12 Months (dollars)	13,427 [13,291]	14,310 [12,635]	12,432 [14,027]	731.0 (1,121)	863.7 (1,647)	678.3 (1,461)

Notes: See notes on Table 3. No information on employment was collected when youths were in their late-teens. Earnings are coded as zero if the person is reported not working. This measure of earnings is one of *realized* earnings and is frequently used in the literature, despite being a crude measure of productivity—since earnings are only observed for employed individuals.

TABLE 6. TREATMENT EFFECT ON RISKY BEHAVIORS BY PREDICTED DRUG USE AT AGE 19

OUTCOMES	CONTROL MEANS			TREATMENT-CONTROL		
	Full sample	Low-risk	High-risk	Full sample	Low-risk	High-risk
	(1)	(2)	(3)	Full set of controls	Full set of controls	Full set of controls
Youths Were in their Late Teens						
<i>Summary measure</i>				-0.031 (0.039)	-0.162*** (0.052)	0.102*††† (0.058)
Binge drinking in the past 30 days	0.188	0.162	0.218	0.035 (0.027)	0.122*** (0.039)	-0.061††† (0.039)
Used any illegal drug in the past 30 days	0.270	0.177	0.376	0.063** (0.031)	0.165*** (0.040)	-0.051††† (0.047)
Committed a crime in past 12 months	0.284	0.246	0.326	0.024 (0.031)	0.095** (0.041)	-0.041†† (0.045)
Ever arrested or charged	0.276	0.263	0.291	-0.035 (0.029)	-0.008 (0.039)	-0.053 (0.043)
Have first child before age 18	0.240	0.230	0.251	-0.027 (0.028)	-0.024 (0.038)	-0.030 (0.042)
Youths Were in their Early Twenties						
<i>Summary measure</i>				0.012 (0.035)	-0.009 (0.046)	0.033 (0.054)
Binge drinking in the past 30 days	0.293	0.326	0.256	-0.041 (0.032)	-0.0742* (0.0449)	-0.010 (0.045)
Used any illegal drug in the past 30 days	0.159	0.113	0.212	-0.044* (0.025)	0.00251 (0.0307)	-0.084** (0.041)
Committed a crime in past 3 months	0.081	0.056	0.109	-0.011 (0.019)	0.0219 (0.0243)	-0.048† (0.030)
Arrested or charged in past 3 months	0.042	0.034	0.051	0.003 (0.015)	0.010 (0.018)	-0.004 (0.024)
Have first child before age 18	0.148	0.141	0.157	0.030 (0.026)	0.009 (0.035)	0.061 (0.039)
Currently on welfare	0.184	0.183	0.186	0.032 (0.028)	0.018 (0.038)	0.0297 (0.042)
Youths Were in their Mid-Twenties						
<i>Summary measure</i>				-0.078** (0.035)	-0.085* (0.047)	-0.065 (0.055)
Binge drinking in the past 30 days	0.283	0.325	0.243	0.012 (0.031)	-0.005 (0.042)	0.046 (0.046)
Used any illegal drug in the past 30 days	0.120	0.088	0.160	0.001 (0.022)	0.038 (0.030)	-0.050† (0.036)
Committed a crime in past 3 months	0.020	0.015	0.026	0.017 (0.012)	0.008 (0.014)	0.028 (0.020)
Arrested or charged in past 2 years	0.051	0.041	0.064	0.044** (0.018)	0.054** (0.025)	0.032 (0.028)
Have first child before age 18	0.162	0.160	0.166	0.014 (0.026)	-0.009 (0.035)	0.036 (0.040)
Currently on welfare	0.220	0.201	0.244	0.033 (0.029)	0.054 (0.039)	0.0026 (0.045)

Notes: See notes on Table 3.

TABLE 7. PEER EFFECTS BY PREDICTED DRUG USE AT AGE 19

OUTCOMES	<i>Full sample</i>			<i>Low-risk</i>			<i>High-risk</i>		
	<i>QOP</i>	<i>PEER</i>	<i>QOP*PEER</i>	<i>QOP</i>	<i>PEER</i>	<i>QOP*PEER</i>	<i>QOP</i>	<i>PEER</i>	<i>QOP*PEER</i>
<i>Youths Were in their Late Teens</i>									
High-school	0.092 (0.074)	0.139 (0.205)	-0.168 (0.261)	0.136 (0.103)	0.884** (0.382)	-0.864* (0.447)	0.137 (0.113)	0.0238 (0.278)	-0.126 (0.358)
College and Post- Secondary Education	0.199 (0.132)	-0.329 (0.336)	-0.243 (0.422)	0.411** (0.180)	0.547 (0.707)	-1.357* (0.774)	-0.0186 (0.231)	-0.625 (0.498)	0.496 (0.648)
Risky Behaviors	0.021 (0.081)	-1.091*** (0.253)	0.073 (0.309)	0.0831 (0.108)	-0.444 (0.386)	-0.687 (0.466)	0.0863 (0.135)	-1.151*** (0.352)	0.109 (0.451)
<i>Youths Were in their Early-Twenties</i>									
High-school	0.045 (0.069)	-0.173 (0.209)	0.038 (0.259)	-0.021 (0.091)	-0.120 (0.385)	0.054 (0.430)	0.158 (0.109)	-0.0538 (0.297)	-0.192 (0.375)
College and Post- Secondary Education	0.171 (0.122)	-0.179 (0.343)	-0.177 (0.419)	0.439*** (0.161)	1.086 (0.670)	-1.847** (0.733)	-0.0344 (0.207)	-0.376 (0.481)	0.745 (0.638)
Employment	-0.019 (0.128)	0.280 (0.400)	-0.379 (0.474)	-0.037 (0.167)	0.873 (0.669)	-0.996 (0.729)	0.140 (0.196)	0.525 (0.534)	-0.448 (0.659)
Risky Behaviors	0.145** (0.072)	0.074 (0.238)	-0.493* (0.275)	0.095 (0.089)	-0.234 (0.396)	-0.266 (0.411)	0.291** (0.138)	0.422 (0.340)	-0.907** (0.455)
<i>Youths Were in their Mid-Twenties</i>									
High-school	-0.019 (0.069)	-0.184 (0.204)	0.193 (0.262)	-0.147 (0.091)	-0.515 (0.356)	0.616 (0.418)	0.141 (0.111)	0.223 (0.289)	-0.197 (0.388)
College and Post- Secondary Education	0.161 (0.116)	0.009 (0.324)	-0.287 (0.426)	0.236 (0.152)	0.517 (0.556)	-0.860 (0.638)	0.157 (0.197)	0.0500 (0.462)	-0.153 (0.653)
Employment	0.116 (0.126)	-0.236 (0.375)	-0.326 (0.457)	-0.0277 (0.186)	-0.158 (0.903)	0.0748 (0.899)	0.234 (0.190)	-0.0364 (0.450)	-0.599 (0.628)
Risky Behaviors	0.040 (0.080)	0.254 (0.207)	-0.474* (0.294)	-0.0921 (0.111)	0.345 (0.391)	-0.128 (0.461)	0.260** (0.117)	0.277 (0.286)	-1.113*** (0.413)

Notes: See notes on Table 3. In addition to the covariates described in Table 3, the specifications from this Table include a peers' variable and a variable interacting peers' effects and the treatment dummy. The peers' variable is measured as follows: for each individual we estimated the proportion of students within the same school, treatment status, sex, and race, that reported doing drugs at age 19.

Online Appendix:

School, Drugs: Closing the Gap

Evidence from a Randomized Trial in the United States

By Núria Rodríguez-Planas

TABLE A.1

QOP'S DEVELOPMENTAL ACTIVITIES, COMMUNITY SERVICE ACTIVITIES AND EDUCATIONAL SERVICES

Activity	Examples of such types of activities
Developmental	Life skills activities/ discussion topics (such as, family planning, nutrition, personal hygiene, managing anger, avoiding drug behaviors, among others); pre-employment training; cultural activities; and recreational activities.
Community service	Visits to the residents of a local nursing home, or volunteering at a local food bank.
Educational services	Academic assessment, development of individualized education plans, one-on-one tutoring, and computer-assisted instruction in specific coursework as well as basic reading and mathematics. Making the youth aware of, and helping them plan for, college and other post-secondary education or training.

TABLE A.2

DETERMINANTS OF SOCIAL PROBLEMS IN THE CONTROL GROUP, BY GENDER

	Drug use	
	<i>Males</i>	<i>Females</i>
LHS variable means	0.335	0.192
Age when entering 9 th grade		
14	-0.692 (0.556)	0.401 (0.612)
> 14	-0.461 (0.596)	0.020 (0.806)
Rank based on 8 th grade		
Middle third	0.367 (0.376)	0.048 (0.569)
Top third	-0.045 (0.401)	-0.061 (0.545)
Black	-0.158 (0.862)	0.706 (1.078)
Hispanic	-0.330 (0.743)	-0.763 (1.057)
Observations	209	172
Pseudo R-squared	0.075	0.060

Notes: The table reports logit estimates. The estimates in this table are constructed using the sample of control group youths only. School dummies are included in the specification.

TABLE A.3. PARTICIPATION IN QOP ACTIVITIES BY PREDICTED DRUG USE AT AGE 19

				<i>Low-risk</i>			<i>High-risk</i>		
	<i>Cumulative Years 1 through 4</i>	<i>Year 1</i>	<i>Year 4</i>	<i>Cumulative Years 1 through 4</i>	<i>Year 1</i>	<i>Year 4</i>	<i>Cumulative Years 1 through 4</i>	<i>Year 1</i>	<i>Year 4</i>
Average Number of Hours	886	279	125	877	290	117	896	267	133
Average Hours on Educational Activities	245	97	26	247	101	23	244	92	30
Average Hours on Developmental Activities	302	114	27	110	39	13	108	36	16
Average Hours on Community Service Activities	109	37	14	298	114	22	306	113	32
No Hours of Participation (percent)	1	1	40	1	1	36	1	1	44
More Than 100 Hours (percent)	88	74	22	87	73	17	89	77	27
More Than 375 Hours (percent)	60	25	9	57	26	8	63	24	12
More Than 750 Hours (percent)	37	7	4	34	9	4	40	5	4
More Than 1,500 Hours (percent)	17	0	2	17	0	1	17	0	2
Total Amount Earned (dollars)	\$1,122	\$349	\$156	\$1,103	\$363	\$146	\$1,142	\$334	\$167

Note: Because QOP services in Period 5 differed substantially from those of the first four periods, we report hours over the first four periods. In Period 5, QOP offered enrollees who had graduated from high school only mentoring services, and hours spent being mentored were not recorded.

TABLE A.4. MEAN EFFECT SIZES FOR SUMMARY MEASURES OF OUTCOMES CLUSTERING STANDARD ERRORS AT THE SCHOOL LEVEL

OUTCOMES	TREATMENT-CONTROL			
	CLUSTERED STANDARD ERRORS			
	Low-risk Full set of controls	High-risk Full set of controls	Low-risk Full set of controls	High-risk Full set of controls
	(1)	(2)	(3)	(4)
<i>Youths Were in their Late Teens</i>				
High-school	0.0031 (0.049)	0.099** (0.047)	0.0031 (0.041)	0.099** (0.044)
College and Post-Secondary	0.094 (0.082)	0.129 (0.084)	0.094 (0.062)	0.129 (0.082)
Risky Behaviors	-0.162*** (0.052)	0.102*††† (0.058)	-0.162*** (0.037)	0.102*††† (0.034)
<i>Youths Were in their Early Twenties</i>				
High-school	-0.021 (0.043)	0.101**† (0.048)	-0.021 (0.046)	0.101* (0.046)
College and Post-Secondary	0.061 (0.086)	0.177** (0.082)	0.061 (0.084)	0.177* (0.082)
Employment	-0.205*** (0.078)	0.017† (0.084)	-0.205 (0.121)	0.017† (0.051)
Risky Behaviors	-0.009 (0.046)	0.033 (0.054)	-0.009 (0.048)	0.033 (0.041)
<i>Youths Were in their Mid-Twenties</i>				
High-school	-0.038 (0.042)	0.085*†† (0.044)	-0.038 (0.046)	0.085* (0.043)
College and Post-Secondary	0.063 (0.061)	0.112 (0.077)	0.063 (0.039)	0.112 (0.064)
Employment	-0.026 (0.075)	0.057 (0.080)	-0.026 (0.048)	0.057 (0.075)
Risky Behaviors	-0.085* (0.047)	-0.065 (0.055)	-0.085* (0.046)	-0.065 (0.054)

Notes: The table reports estimates of treatment effects on the dependent variables indicated in row headings. Robust standard errors are reported in parentheses. "Full set of controls" includes school dummies, an indicator for being male, an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eighth-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.

*, ** Estimate significantly different from zero at the 90%, or 95% confidence level.

TABLE A.5. MEAN EFFECT SIZES FOR SUMMARY MEASURES OF OUTCOMES USING SAME RESPONSE RATES ACROSS TREATMENT

OUTCOMES	TREATMENT-CONTROL			
	SAME RESPONSE RATE			
	Low-risk	High-risk	Low-risk	High-risk
	Full set of controls			
	(1)	(2)	(3)	(4)
Youths Were in their Late Teens				
High-school	0.0031 (0.049)	0.099** (0.047)	-0.003 (0.051)	0.149*** (0.049)
College and Post-Secondary	0.094 (0.082)	0.129 (0.084)	0.073 (0.083)	0.153* (0.089)
Risky Behaviors	-0.162*** (0.052)	0.102*††† (0.058)	-0.161*** (0.053)	0.127***††† (0.060)
Youths Were in their Early Twenties				
High-school	-0.021 (0.043)	0.101**† (0.048)	-0.034 (0.043)	0.091*†† (0.049)
College and Post-Secondary	0.061 (0.086)	0.177** (0.082)	0.067 (0.089)	0.180** (0.087)
Employment	-0.205*** (0.078)	0.017† (0.084)	-0.169** (0.081)	0.050† (0.086)
Risky Behaviors	-0.009 (0.046)	0.033 (0.054)	-0.009 (0.046)	0.014 (0.058)
Youths Were in their Mid-Twenties				
High-school	-0.038 (0.042)	0.085*†† (0.044)	-0.059 (0.049)	0.076†† (0.052)
College and Post-Secondary	0.063 (0.061)	0.112 (0.077)	0.059 (0.064)	0.143* (0.082)
Employment	-0.026 (0.075)	0.057 (0.080)	-0.019 (0.077)	0.081 (0.083)
Risky Behaviors	-0.085* (0.047)	-0.065 (0.055)	-0.087* (0.048)	-0.074 (0.058)

Notes: Same response estimates were obtained by making the response rate equal across treatment and control groups within school and within top- or bottom-half of the predicted drug use distribution. By restricting response rates by school and risk level to be equal the sample sizes got reduced by 59 observations in the first survey, 70 observations in the second survey and 53 observations in the last survey. The table reports estimates of treatment effects on the dependent variables indicated in row headings. Robust standard errors are reported in parentheses. "Full set of controls" includes school dummies, an indicator for being male, an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eighth-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.

*, ** Estimate significantly different from zero at the 90%, or 95% confidence level.

TABLE A.6. TREATMENT EFFECT ON HIGH-SCHOOL OUTCOMES USING COMPLETE AND ATTRITED SAMPLES

OUTCOMES	COMPLETE SAMPLE			ATTRITED SAMPLE		
	<i>Full sample</i>	<i>Low-risk</i>	<i>High-risk</i>	<i>Full sample</i>	<i>Low-risk</i>	<i>High-risk</i>
	<i>Full set of controls</i>					
	(1)	(2)	(3)	(4)	(5)	(6)
<i>Youths Were in their Late Teens</i>						
Obtained high-school Diploma	0.053*	0.007	0.095**	0.046	0.014	0.075*
	(0.029)	(0.041)	(0.041)	(0.031)	(0.044)	(0.043)
Obtained a GED	-0.009	-0.016	-0.001	-0.009	-0.016	-0.001
	(0.017)	(0.021)	(0.028)	(0.017)	(0.022)	(0.029)
<i>Sample size^a</i>	<i>1,008</i>	<i>523</i>	<i>484</i>	<i>892</i>	<i>464</i>	<i>428</i>
<i>Youths Were in their Early Twenties</i>						
Obtained high-school Diploma	0.032	-0.045	0.102**††	0.031	-0.045	0.101**††
	(0.030)	(0.041)	(0.043)	(0.033)	(0.045)	(0.047)
Obtained a GED	0.008	0.019	-0.002	0.010	0.018	-0.002
	(0.023)	(0.029)	(0.038)	(0.024)	(0.030)	(0.039)
<i>Sample size</i>	<i>942</i>	<i>486</i>	<i>456</i>	<i>786</i>	<i>398</i>	<i>388</i>
<i>Youths Were in their Mid-Twenties</i>						
Obtained high-school Diploma	0.021	-0.054	0.097**††	0.039	-0.049	0.138**††
	(0.030)	(0.040)	(0.044)	(0.032)	(0.045)	(0.047)
Obtained a GED	0.002	0.012	-0.011	-0.016	0.0033	-0.038
	(0.024)	(0.031)	(0.039)	(0.025)	(0.033)	(0.040)
<i>Sample size</i>	<i>919</i>	<i>480</i>	<i>439</i>	<i>792</i>	<i>417</i>	<i>375</i>

Notes: The table reports estimates of treatment effects on the dependent variables indicated in row headings. Robust standard errors are reported in parentheses. “Full set of controls” includes school dummies, an indicator for being male, an indicator for being 14 years old when entering ninth grade, an indicator for being over age 14 when entering ninth grade, an indicator for being in the middle third of the eighth-grade GPA distribution, an indicator for being in the top third of the eighth-grade GPA distribution, an indicator for being black and an indicator for being Hispanic.

* Estimate significantly different from zero at the 90% level.

^a Sample size differs from that of Table 3 in the main paper because when estimating summary indices if an individual has a valid response to at least one component measure of an index, then any missing values for other component measures are imputed at the random assignment group mean (as in Jeffrey Kling, Jeffrey Liebman, and Lawrence Katz 2007).