

College Consumption Amenities, Academic Performance, and Donation Behavior: Evidence from Big-Time College Sports Attendance

Joshua Hyman and Isaac McFarlin, Jr.*

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Abstract:

Colleges compete to attract students by investing in amenities such as athletics, dormitories, and student activities. We examine the effect of student consumption of postsecondary amenities on academic achievement and future donation behavior in the context of Big-Time college sports. We resolve the selection issue using data from a large, public university with a highly-ranked men's basketball team, where student season tickets are awarded by lottery. Game attendance has small negative effects on academic performance but no impact on donation behavior. Negative academic effects are concentrated at the bottom of the achievement distribution.

*Joshua Hyman, Department of Economics, Amherst College; jhyman@amherst.edu; Isaac McFarlin, Jr. NBER; isaacmcfarlin@gmail.com. We are grateful to Diego Briones for excellent research assistance. Thank you to Roger Blair, Eric Brunner, Charlie Clotfelter, Jeff Greenspan, Patrick Lapid, Jason Lindo, Paco Martorell, Brian McCall, Una Osili, Dan Rosenbaum, Stephen Ross, Hector Sandoval, Caroline Theoharides, Glen Waddell, seminar participants at Tufts University and the University of Michigan, and audience members at the 2020 Northeast Economics of Education Workshop, 2022 Association for Education Finance and Policy conference, and 2022 Society of Labor Economists for their helpful comments.

I. Introduction

Colleges compete to attract students by investing in consumption amenities such as athletics, dormitories, and student activities. Among the approximately 1,300 four-year non-profit postsecondary institutions in the U.S., for every dollar spent on academics, more than 50 cents is spent on amenities (Jacob, McCall, and Stange, 2018; henceforth, JMS). These investments have received criticism for being anti-academic and unproductive. For example, one article published in the *Wall Street Journal* highlights a recently constructed luxury dormitory at a public university featuring “gaming rooms” with wall-to-wall TVs and video game systems, questioning the academic benefits of these investments (Wotapka, 2012). On the other hand, while non-academic in nature, such amenities could provide well-needed breaks from the pressures of studying, improving student mental health, helping students develop social networks, and increasing students’ connection with the college, all of which could lead to improved academic performance.

Despite colleges’ large investments in amenities—and the uncertainty about whether these investments are beneficial to students—there is surprisingly little causal evidence on the impact of student consumption of these amenities on academic achievement. This is primarily due to the selection issue, whereby students who consume amenities differ from students who do not. For example, JMS (2018) show that the academically strongest students have a lower taste for said amenities, such that colleges competing for less-academically skilled students are those that invest in amenities most heavily. Without plausibly exogenous variation in student participation, it is challenging to separate the effects of consumption amenities on student outcomes from the underlying differences between students who do and do not prefer such amenities.

In this paper, we investigate the causal impact of student consumption of postsecondary amenities in the context of Big-Time college sports, one of the most prolific and well-funded amenities in postsecondary education (Clotfelter, 2011). We examine effects of game attendance on students’ academic outcomes and future donating behavior, solving the selection issue by leveraging student season ticket lotteries. Specifically, we use 9 years of student lottery data from a large, public flagship university with a consistently high-ranked men’s basketball team where, due to excess demand, student season tickets are assigned by lottery. We match all student lottery applicants to academic transcripts from the registrar and university foundation records to examine the effects of students’ game attendance on their grades, credits earned, and degree completion as well as on their likelihood and amount of future donations to the university.

Comparing lottery winners to lottery losers, we find that winning the lottery causes small reductions in GPA and credits earned during the current academic year. The GPA reduction is concentrated at the bottom of the achievement distribution and persists through students' final cumulative GPA. These declines in GPA and credits earned appear to be driven by in-state students and students who attend games during seasons in which the team reaches the March Madness postseason tournament, though these differences are merely suggestive as we do not have sufficient statistical power to reject equality across groups. When examining heterogeneity by student race, Asians, who comprise the largest racial minority group on campus, appear to experience the greatest negative impacts, though again the differences are statistically insignificant. These student subgroups whose academic performance appears to be most harmed by winning the lottery are ultimately less likely to complete college and earn a Bachelor's degree.

As to whether student consumption of postsecondary amenities increases their connection to the university, and thus their donations after graduating, we find no evidence that winning the lottery increases the likelihood or amount of donations by students, or their parents, after leaving the university. If anything, we find suggestive evidence that the parents of students experiencing the greatest academic declines due to game attendance *reduce* their giving to the university. One limitation of this analysis is that we only observe short- to medium-run donation behavior: we observe donations through ten years after graduating for our oldest student cohort, but through only a few months after graduating for our most recent cohort. While ideally we would observe donations further out in time, our null result may suggest that game attendance has no impact on longer-run donations as well, given habit formation in donating behavior and the strong correlation between giving over time (Meer, 2013).

The comparisons of lottery winners to lottery losers estimate the intent-to-treat (ITT) effect of game attendance. Lottery winners may choose not to attend games or may give away or resell their tickets to other students, though resale values are low, making financial motives an unlikely reason for application. Unfortunately, the university does not track which students actually use the ticket. To provide an informal scaling of our estimates into treatment-on-the-treated (TOT) effects, we conduct a survey distributed via email by the university athletics department in spring 2018 to all current and former students who applied to the fall 2014, 2015, 2016, or 2017 student season ticket lotteries. We ask about the number of games attended in each year, subsequently using the difference in attendance between lottery winners and lottery losers to scale our ITT effects into

TOT estimates. Our informal TOT effects suggest that attending an average number of games in a single season reduces students' current GPA during the semester after the lottery by 0.05 points (7% of a standard deviation) and their final cumulative GPA by 0.02 points (5% of a standard deviation).

What may explain the negative effects that we find on academic outcomes, and the null effect that we find on donations? Students who do not attend games may still watch the games on television, suggesting that our effects capture the impacts of physical presence at the games along with any change in behavior before or after physically attending games. As an attempt to examine mechanisms, we include two questions in the survey asking about how game attendance (relative to days when students watch the game on TV or not at all) affects student behavior, finding that students report having less time for studying and being more likely to skip class surrounding days they attended basketball games. We additionally ask students whether attending games made students feel more or less "connected" to the University. Students overwhelmingly responded positively, suggesting that game attendance may increase students' feelings of connectedness to the University in ways not reflected in their donation behavior.

This paper contributes to two literatures in the economics of higher education. First, JMS (2018) bring renewed economic attention to the literature examining markets in higher education by showing that consumption amenities are one of the primary attributes in which colleges invest and over which students sort across colleges. However, an open question in this literature is whether such college catering to demand-side market pressures for amenities helps or hinders student academic performance and whether these college investments have future payoffs in the form of increased developmental support from alumni. To the best of our knowledge, this paper is among the first to examine the causal impact of student participation in college consumption amenities on these outcomes. In related prior work, Webber and Ehrenberg (2010) use nationwide institution-level data from the Integrated Postsecondary Education Data System (IPEDS) to examine the effects of non-instructional spending on student completion rates, but lack plausibly exogenous variation in spending, relying on institution fixed effects for identification.

A few studies have shown that when college sports teams perform better, non-athlete academic performance declines (Clotfelter, 2011; Lindo, Swensen, and Waddell, 2012) and philanthropic contributions increase (Meer and Rosen, 2009; Anderson, 2017). For example, Clotfelter (2011) finds that the number of JSTOR articles viewed at university libraries decreases

around the time of the March Madness postseason basketball tournament for institutions with a team in the tournament. Lindo, Swensen, and Waddell (2012) use administrative data from a single university, examining effects of team success on students' GPA. Leveraging team success provides plausibly exogenous variation in non-athlete student engagement with Big-Time sports, but one disadvantage of this design is that the identifying variation is cross-seasonal. Therefore, all students in a given season are "treated," raising concern that estimates may be attenuated due to grade curving. To avoid this concern, the authors focus on the GPA gender gap, finding that GPA declines more for males than for females.

In contrast, our lottery design compares students within a given season, yielding estimates that are unlikely to be biased by grade curving. In addition, comparisons across years when a team performs better or worse capture one margin along which student engagement in Big-Time sports responds, but tell us little about the underlying effects when team success is relatively stable across seasons, or about effects of student engagement regardless of team success. Similar to Clotfelter (2011) and Lindo, Swensen, and Waddell (2012), we find that game attendance decreases academic performance more so during seasons when the team performs better. But in contrast to Meer and Rosen (2009) and Anderson (2017), we find no increase in philanthropic behavior due to game attendance, even during those better-performing seasons.

Our paper also contributes to the literature attempting to identify and reduce barriers to postsecondary academic success and completion. While rates of college entry have risen in recent decades, rates of college completion have not kept pace (Bound, Lovenheim, and Turner, 2010; Bailey and Dynarski, 2011). In response, economists and education researchers have turned their attention to identifying barriers to academic success, and to finding ways to dismantle these barriers (e.g., Stinebrickner and Stinebrickner, 2004; Bettinger and Baker, 2014; Holzer and Baum 2017; Andrew, Imberman, and Lovenheim, 2020; Oreopoulos, 2021; Oreopoulos et al., 2022). Academic coaching and financial support have had some success (Bettinger and Baker 2014; Goldrick-Rab et al., 2012) while efforts to increase study time have failed (Oreopoulos et al., Forthcoming). We show that consumption amenities may play a role in academic success and college completion. The ITT effects we estimate on current GPA are similar in magnitude to those found in Lindo, Swensen, and Waddell (2012) from a 25-percentage point increase in the football team's winning percentage. While our TOT estimates are informal, the declines in current GPA and credit completion that they suggest are of similar magnitudes as the increases brought about

by an offer of \$3,500 in financial aid in a randomized control trial examining the impacts of financial aid on student academic performance and persistence (Goldrick-Rab et al., 2012).

These results have important policy implications for the financing of public higher education. The overwhelming majority of postsecondary students in the U.S. attend public colleges or universities, where public funds support college investments. As public universities increase their investment in consumption amenities in response to demand-side pressure, it raises the question of whether such investments are an efficient use of taxpayer money in terms of their effects on both student academic success as well as future savings due to alumni giving. While we are unable to conduct a comprehensive welfare calculation, our results suggest that while such investments increase enrollment (JMS, 2018), they may in fact hinder student performance with little potential future financial benefit, and thus may be an inefficient use of taxpayer resources.

This study comes with some important caveats and raises some additional questions. First, Big-Time college sports attendance is just one specific type of consumption amenity in which colleges invest to attract students, and the effects of student participation in amenities may differ for various amenities. For example, many functional amenities such as parking, laundry services, and dining halls, may have little to no scope to affect academic achievement. Second, as noted by JMS (2018), it is possible that student participation in amenities could improve non-academic skills or social networks rewarded in the labor market that we cannot observe using our administrative data. Finally, like prior similar studies (e.g., Lindo, Swensen, and Waddell, 2012; Meer and Rosen, 2009), our paper focuses on a single institution, raising the question of whether the results would extrapolate to the broader set of colleges and universities across the U.S. Regarding this specific concern, it is worth noting that the representativeness of the large, public flagship university we study may compare favorably to the military academies and private elite colleges that have been the setting of many influential studies in the economics of higher education. While acknowledging these limitations, our study offers some of the first causal evidence that college catering to student preferences for consumption may hinder student academic success, with little future financial payoff to universities.

II. Background on the Season Ticket Lottery

Our study uses data from an unnamed large, public flagship university with a consistently well-ranked men's basketball team.¹ Due to greater demand for student tickets to attend games than there are seats, every fall the athletic ticket office holds a lottery to award student season tickets. All undergraduate and graduate students are emailed an invitation to apply electronically to the ticket lottery in mid-to-late September. Students have three days to decide whether to enter the lottery, which they can do by entering their student ID at the lottery website. There is no cost to enter the lottery and entering the lottery does not obligate students to purchase season tickets if they win the lottery. A few days after entering the lottery, winning students are emailed purchase instructions and losing students are also notified. Winners then have three days to purchase their season ticket package, which always almost sell out, but never do completely due to a small number of winners not claiming their season tickets. These unclaimed season tickets are then sold online, to students only, on a first-come basis and universally sell out. The basketball season runs from November through March, with 16 to 19 home games. Lottery winners must purchase the tickets to all of the home games. During our sample period, the price per ticket was \$5, for a total price for the season ticket package ranging between \$80 to \$95.

The student season ticket lottery is weighted by grade level such that first-years and graduate students have the lowest chance of winning and seniors have the greatest chance. The student IDs for seniors are entered five times into the lottery, juniors four times, sophomores three, first-years and most graduate students two, and medical, dental, and business graduate students once. The lottery process is conducted by a third-party consultant. After duplicating the student IDs the appropriate number of times, the consultant enters the IDs into a single column in Excel, then sorts the column by a randomly generated number, and denotes the top IDs as winners moving down the column until the number of unique winning IDs equals the number of available season tickets. While the number of winners is nearly identical across years, the number of students applying to the lottery varies, such that the probability of winning is higher in years when fewer students apply.

Appendix Table 1 shows the number of lottery applicants and fraction winning the lottery by grade level and year in our data. From fall 2009 through fall 2017 there were between about

¹ The university's football team is lower ranked, and sports culture at this university revolves around basketball. Men's basketball is the only team for which there is a season ticket lottery.

5,500 and 8,000 applicants per year for a total of 57,909 lottery applicants.² Some students apply for the lottery in more than one year; the total number of unique students in our sample is 30,299. Seniors (15,063 total applications) and juniors (15,919) were the most likely to apply, followed by sophomores (13,909), first-years (11,593), and graduate students (1,425). The winning percentage varies inversely with the number of applicants, ranging across years from 28% to 46%, with an average win percentage across all grades of 36%, and of 47% for seniors, 40% for juniors, 32% for sophomores, 22.4% for first-years, and 21.6% for graduate students. As discussed more in Section IV, we draw comparisons within grade and year so that we do not conflate the effect of winning the lottery with the effect of being in a grade or year with a higher likelihood of winning.

III. Data

Our study uses student data from several departments across an unnamed large, public flagship university. We merge 1) lottery data from the athletic ticket office, 2) housing assignments from the on-campus housing authority, 3) student transcripts from the Registrar, and 4) donation records from the University Foundation. We (rather than the University departments or a central office of Institutional Research) conduct all of the merges using Student ID as the matching variable. Below, we provide more detail about the information contained in each data source.

Season Ticket Lottery: We use nine years of lottery records provided by the athletic ticket office, from fall 2009 through fall 2017. The records contain the list of applicant names, student IDs, grade level, the rank of the IDs randomly assigned during lottery process, and which IDs were assigned as lottery winners.

Housing Assignments: We merge the lottery records with the universe of on-campus housing assignment records, containing dorm name, room number, and bed number. We include all lottery applicants' roommates during their lottery year in our dataset.

Student Transcripts: We then merge all lottery applicants and their roommates with transcripts from the Registrar as of July 2018. The transcript data contain basic demographics (i.e., race and sex), in-state versus out-of-state status, legacy status, course-level grades and credits

² Approximately a third of undergraduate students enrolled at the University applied to the lottery each year. We consider only undergraduate students for this calculation, because they make up nearly all of the lottery applicants, but only about 70% of all students at the University. Given the small share of the sample comprised by graduate students, in Table 4 we show robustness of our main results to excluding them.

earned, semester-level current and cumulative GPA and credits earned, final cumulative GPA, declared major, and degrees received with dates they were earned.

Donations: Finally, we merge all lottery applicants and their roommates to the universe of university giving records through August 2018. These data contain the complete history of all donations by each student and his or her parent(s), including date, amount, and any designation (e.g., athletics).

We combine these data sources into a student-by-lottery-application-year panel from 2009-2017 containing 57,909 observations and 30,299 unique students. Columns 1 and 2 of Table 1 show sample means and standard deviations for available pre-lottery sample characteristics. 70% of the sample is White, 9% Asian, 7% Hispanic, and 4% Black (with 9% unreported race). Almost half (49.5%) of the sample is female. Seventy-three percent of the sample is considered in-state while 12% comprise legacy students. Cumulative GPA and credits earned as of the semester prior to lottery application (dropping first-years) are 3.25 and 71 respectively. Thirty percent of parents donated to the University prior to the lottery semester. Summing across all past donations, the average unconditional donation is \$678, though this is highly skewed, with a standard deviation greater than \$22,000.³ Sixty-two percent of the sample lives on campus and has at least one roommate during their lottery semester, while 12% live on campus and have multiple roommates. Thirty-eight percent of the sample has at least one roommate who also applied to the lottery that year, and 14% has at least one roommate who applied to and won the lottery that year.

While we only have administrative data on the lottery applicants and their roommates, we compare the lottery sample to the overall university student population using data from IPEDS during our sample period to gain some understanding of how the lottery sample compares to the University population. In addition to the lottery sample being skewed toward undergraduates, and among undergraduates toward upper-classmen (as shown in Section II), it is also skewed somewhat toward in-state students (73% lottery sample versus 66% overall) and white students (70% lottery versus 62% overall). The lottery sample is nearly identical to the University population with respect to gender (49.5% lottery versus 49.6% overall).⁴

³ We show that our null result on donation amount is robust to several ways of addressing the skewed nature of the donation distribution.

⁴ We present these comparisons merely because it is interesting to understand how the lottery sample differs from the University population, but not because we believe any differences are a concern for external validity. The population for which we are interested in understanding the effects of sports attendance is those students who have some interest in attending games – the lottery sample is arguably a good proxy for that population. We would not

IV. Methodology and the Student Survey

To identify the causal impact of winning the lottery on student outcomes, we estimate the following OLS regression:

$$Y_{igy} = \beta_0 + \beta_1 WIN_{igy} + X_{igy} + \delta_{gy} + \varepsilon_{igy} \quad (1)$$

where Y_{igy} is the outcome for student, i , who was in grade, g , when they applied for the lottery during year, y , WIN is a dummy for whether the student won the lottery, X is the set of student and student-year level covariates presented in Table 1, and δ is a set of grade-by-year fixed effects. The fixed effects are necessary to ensure that comparisons are drawn within grade and year, as that was the level of random assignment. Omission of the fixed effects would cause β_1 , the parameter of interest, to conflate the effect of winning with the effect of being in a grade or year with a higher likelihood of winning. We cluster the standard errors at the grade-year level.

Table 1 shows that the lottery was well-implemented and that we achieve balance across our baseline characteristics. To test for balance for each covariate in X , we remove that covariate from X , estimate equation (1) where that covariate is the dependent variable, and report β_1 in column 3 and its standard error in column 4. All coefficients are near zero and statistically insignificant. An F-test of the joint significance of all of the covariates from a regression of WIN on X and the grade-by-year fixed effects fails to reject the null that they jointly equal zero.⁵

Equation (1) estimates the Intent-to-Treat (ITT) effect of game attendance, given the small amount of incomplete take-up of purchasing season tickets by winners, and more importantly, given the fact that the University does not require that student tickets be used by the purchaser. Student tickets must be used by a University student, and any person entering the sports arena with a student season ticket must present their student ID. However, there is no student name on the ticket or any other way to verify that the ticket purchaser is the student attending the game. Thus, there is an active secondary market for selling and purchasing student tickets to individual games

attempt to extrapolate our results to the population of college students who have no interest in attending a sports game, i.e., to the sample of students who did not enter the lottery.

⁵ One concern about using the full set of lotteries is that winning the lottery may influence students' future likelihood of applying, suggesting that lotteries after students' initial application year may be endogenous. To examine this concern, we restrict the sample to the first lottery in which each student participated and estimate equation (1) where the outcome is a dummy for whether the student applies in a later year. We find a precise zero effect of winning on future lottery applications (point estimate: -0.001; SE: 0.006), suggesting that this potential endogeneity is not a concern. We also show in Table 4 that our main results are robust to restricting the sample to the first lottery year for each student.

(and occasionally entire season ticket packages). For example, during our sample period there was a dedicated Facebook page titled “Buy and Sell [University] Basketball Tickets.”

In spite of this active secondary market, lottery winners made very little, if any, money selling their tickets, and therefore, financial motives are an unlikely reason for students applying to the lottery. To understand whether lottery winners made substantial amounts of money selling tickets, and whether this additional income could be one mechanism for the ITT effect estimated, we tracked all of the entries on the above-referenced Facebook page for an entire year (the 2017-18 season). Student postings to sell tickets usually asked for face value (\$5), with a small number of posts for particularly desirable games asking somewhat above face-value (\$8 or \$10), and on very rare occasion, requesting larger amounts, such as \$30, for the most desirable games. We also conducted several informal interviews with students who had applied to the lottery, and the winners described that they would typically receive face value for their sold tickets. No students we interviewed reported earning a profit of more than \$50 in a season for their tickets, with the overwhelming majority earning far less or even zero profit.⁶ Thus, given the relatively small cost of the season tickets (less than \$100), and relatively little profit made from selling tickets, we believe there is little room for an income effect to play much role in the effects we find.

Unfortunately, the University does not track who uses the tickets, and so we have no way to observe which of the lottery applicants in our data actually attend games. As such, in order to informally scale our estimates into a Treatment-on-the-Treated (TOT) effect of game attendance, we fielded a student survey. The email-based survey was conducted via Qualtrics in April 2018 (after the 2017-18 basketball season was complete) by the athletic ticket office. Every email included a unique link connected to each student ID so that we could merge our lottery data with the survey data. The survey was emailed to all 16,318 current and former students who applied to the ticket lottery during fall 2014, 2015, 2016, or 2017, comprising over half of the 30,299 students in our main sample. Of those emailed, 1,641 (10.1%) responded with a valid set of survey

⁶ Our understanding from our discussions with students and with staff at the athletic ticket office is that the low ticket price in the secondary market is because that while demand outstrips supply for the season tickets, it is does not for most of the individual games. During this period, students could only attend games by securing a student ticket that was part of the season ticket package. Students could not purchase single-game or “general admission” tickets from the athletic ticket office. While students earn relatively little from selling tickets, they may be earning good will from their peers by gifting their tickets. However, it is difficult to imagine this good will from their peers explaining the negative effect we find of winning the lottery on academic outcomes.

responses. Importantly, winning the lottery had no effect on whether a student responded to the survey, i.e., the survey data does not suffer from differential response by treatment status.⁷

The primary purpose of the survey was to query students on how many games they attended each year (e.g., 2014-15, 2015-16, etc.), so that we could compare the number of games attended by lottery winners to the number attended by lottery losers.⁸ To do this, we estimate equation (1) on the sample of survey respondents, reporting results in Appendix Table 2. We find that winning the lottery causes a 0.603 game (SE=0.151), or 20.7%, increase in the number of games attended, compared to a control mean of 2.9 games attended.⁹ Given this 20.7% effect of winning the lottery on game attendance, in our results presented below in Section V, we scale our ITT effects by 4.76 ($=1 / 0.207$) to provide suggestive estimates of the TOT effects. A similar approach is used by Hoxby and Turner (2013), who estimate TOT effects of their light-touch, college-going intervention by scaling their ITT estimates by the fraction of treated students who recall receiving the intervention. We acknowledge the informal nature of our TOT estimate, and believe that a conservative interpretation of our results would be to consider the ITT as a lower bound on the effects of game attendance, and the TOT as an upper bound.

V. Results

V.I Academic Performance

Table 2 presents effects of game attendance on student academic outcomes. Column 1 presents the ITT effect without covariates. Column 2, our preferred specification, presents the ITT effect including the covariates listed in Table 1. We focus on the preferred specification with covariates presented in column 2, noting that the results are similar in column 1.

We begin by examining effects on students' current GPA during the semesters that overlap with the basketball season, which include the fall semester of the lottery and the following spring semester. We find no statistically significant effect of winning the lottery on GPA during the fall

⁷ We estimate equation (1) on the sample of 2014, 2015, 2016, and 2017 lottery applicants, where the dependent variable was a dummy for responding to the survey. The point estimate is 0.004 (i.e., 0.4 percentage points) and the standard error is also 0.004.

⁸ We also asked a couple questions in the survey to explore possible mechanisms: we asked students whether they studied more or less, and were more or less likely to skip class, during days on or surrounding games they attended.

⁹ This nearly 3 game average attendance among lottery losers, and almost 21% greater attendance among lottery winners, is consistent with our understanding from the student interviews that most students attended a handful of games per season regardless of winning or losing the lottery.

semester of the lottery. However, we find a small decrease in the spring semester after the lottery of about 0.010 GPA points.¹⁰ Combining GPAs across both semesters, the impact on average GPA in the academic year of the lottery is negative 0.007 points. This effect on GPA persists through final cumulative GPA, with lottery winners experiencing a 0.005 point decline.

We find that the effects on current and final cumulative GPA are concentrated in the bottom of the GPA distribution. Figure 1 presents quantile treatment effects for both spring semester and final cumulative GPA, presenting effects at the 10th, 20th, ..., 90th deciles of the unconditional GPA distributions.¹¹ Figure 1a shows very small, statistically insignificant declines in spring semester GPA near the top of the GPA distribution (i.e., 70th-90th percentiles), about 0.01 point declines near the middle of the distribution (i.e., 30th-60th), and the greatest declines at the bottom: nearly 0.02 points at the 20th percentile and 0.03 points at the 10th percentile. Figure 1b shows a similar, though slightly less precise pattern for final cumulative GPA, with a -0.014 effect at the 10th decile. In both cases, the effect at the 10th decile of the GPA distribution is about three times as large as the mean effect, providing strong evidence that the academically weakest students are those whose academic performance is most harmed by game attendance.¹²

We next examine credits earned during the semesters that overlap with the basketball season. Game attendance could reduce credits earned if it causes students to fail or withdraw from classes, or to attempt fewer classes. Similar to GPA, we find no effect on credits earned during the fall semester of the lottery, and a small decline in the spring after the lottery of 0.05 credits (Table 2). The coefficient on total fall plus spring credits during the academic year of the lottery is -0.08. The point estimate for final total credits is similar in magnitude to that during the spring after the lottery, but is statistically insignificant.

¹⁰ One reason that negative GPA effects could emerge during the spring, but not fall semester, is that while there are a similar number of home games during the fall and spring semester, the games that are more desirable to attend occur primarily during the spring semester. The fall semester includes mostly nonconference games where teams play lesser known teams outside of their conference, whereas later in the season teams play other teams in their conference, including established rivals. Additionally, the postseason tournament, if the team performs well enough to play in it, occurs during the spring semester.

¹¹ Specifically, following Firpo, Fortin, and Lemieux (2009) we estimate unconditional quantile regressions using re-centered influence functions (RIFs) ordinary least squares (OLS), presenting partial effects of winning the lottery on each decile of the outcome. We follow equation (1), including the covariates and fixed effects, and clustering the standard errors at the grade-year level.

¹² Quantile treatment effects report the effect on that quantile of the GPA distribution, which is not identical to examining heterogeneity. An alternative approach to explicitly estimating heterogeneity by academic performance is to examine heterogeneity by GPA in the prior semester. However, we prefer the quantile treatment effect approach, because first-years, who comprise a fifth of our sample, have no such prior GPA.

Finally, given the negative academic effects found thus far, we examine whether winning the lottery reduces students' likelihood of completing college and earning a Bachelor's degree. We find a statistically insignificant 0.35 percentage point decline (standard error of 0.24 percentage points) in Bachelor's degree receipt.^{13, 14} Appendix Table 4, discussed further in the below section, examines effects on time to degree completion, finding suggestive evidence that winning the lottery increases time to degree.

Column 3 of Table 2 scales the point estimates from our preferred specification (column 2) to informally estimate TOT effects using results from the student survey. As reported in Section IV, winning the lottery increases the number of games attended in a season by 20.7%. Thus, we report in column 3 the estimates from column 2 multiplied by 4.76 ($=1/0.207$). These informal TOT estimates can be interpreted as the effects of attending the average number of games attended for a single basketball season relative to attending no games in that season.

Given the informal, "back-of-the-envelope" nature of the TOT calculation, these estimates should be considered as merely suggestive. With that caveat in mind, the effect of attending the average number of games in a season is approximately a 0.048 GPA point reduction during the spring semester following the lottery, and a 0.023 point reduction in final cumulative GPA. These represent small, but arguably non-trivial effects on current and final cumulative GPA equivalent to 7.2% and 4.8% of a standard deviation, respectively.¹⁵ For comparison, a randomized control trial evaluating the effects of a state scholarship program found that an offer of \$3,500 in financial aid for currently enrolled college students caused increases in final cumulative GPA of 0.05 points, about the same size as the negative effect we find on current spring GPA, and approximately twice as large as the negative effect we find on final cumulative GPA (Goldrick-Rab et al., 2012). We find that attending an average number of games in a season reduces credits earned during that year

¹³ Note that the sample size for our analysis of degree receipt is somewhat smaller (50,426 rather than 57,909), because we drop the recent cohorts for whom we cannot observe on-time degree receipt in our data, which ends during summer 2018. These cohorts include juniors in fall 2017, sophomores in fall 2016 and 2017, and freshmen in fall 2015, 2016, and 2017. We show the effects on GPA and credits earned for this sample in column 5 of Table 4.

¹⁴ We use linear probability models to examine effects on degree receipt and other binary outcomes, such as whether a student or their parent ever donated to the University. We show in Appendix Table 3 that our results for these outcomes are essentially identical when we instead report marginal effects from Logit or Probit regressions.

¹⁵ Among lottery losers the standard deviations of current GPA during the spring semester after the lottery, and of final cumulative GPA, are 0.664 and 0.475, respectively.

by 0.4, compared to a 0.5 credit per year increase in Goldrick-Rab et al. (2012).¹⁶ Given the suggestive nature of the TOT estimates, we also compare the ITT estimates to results from prior studies. Lindo, Swensen, and Waddell (2012) find that a 25 percentage point increase in a football team's winning percentage reduces current GPA among male students by approximately 0.010 to 0.015 points (row 1, Table 2), which is similar in magnitude to the 0.010 reduction in spring GPA that we estimate comparing lottery winners to lottery losers.

Given the negative effects of game attendance on student academic outcomes, and the large literature documenting the importance of peer effects in college (e.g., Sacerdote 2001; Carrell, Fullerton, and West, 2009; Carrell, Hoekstra, and West, 2011; Garlick 2018), we examine whether lottery applicants' roommates experience negative academic effects. These could occur either through traditional peer effects, or by lottery winners giving some of their tickets to their roommates. Appendix Table 7 presents effects of winning the lottery on the academic outcomes of lottery applicants' roommates, finding no evidence of any effects.¹⁷ We do not interpret these null results as evidence that peer effects are unimportant in college academics, but rather that the effects on academic performance are small enough in this context that: 1) they do not appear to generate any meaningful spillovers to lottery applicants' roommates, and 2) the tickets given to roommates are not enough to generate a detectable effect.

V.II Heterogeneity in the Academic Performance Effects

Some students may experience disproportionate effects of winning the lottery on their academic performance. We explore whether the effects of winning the lottery vary by four types of heterogeneity: 1) student race, 2) student gender, 3) student in-state residence, and 4) seasons when the basketball team made the March Madness postsecondary tournament compared with seasons that they did not. Any heterogeneity could be due to differing sensitivity to game attendance, or alternatively to differing effects of winning the lottery on game attendance. As an attempt to disentangle these two channels, we use the student survey data to examine heterogeneity across these groups in the effects of winning the lottery on game attendance.

¹⁶ An alternative way of characterizing the magnitude of effects on credits earned is that one in ten students attending an average number of games earns credit for one less class during that academic year (relative to the mean of approximately 7.2 classes taken per year).

¹⁷ This analysis is restricted to the 62% of the sample who live on campus and have a roommate. For the almost one-fifth of this sample (12% of the main sample) that have multiple roommates, the outcomes are averaged across the roommates.

Given the large number of outcomes and four different types of heterogeneity, to minimize the number of statistical tests conducted we begin by testing for differences jointly across all of the outcomes, for example, a joint test of gender differences across all seven academic outcomes. For all four measures of heterogeneity, we fail to reject the hypothesis that the differences are equal to zero, suggesting that there is no statistically significant evidence of heterogeneity in the effects of winning the lottery on academic outcomes.

In spite of this lack of statistically significant differences across groups, it may still be interesting to examine whether there are any observed patterns of heterogeneity, keeping in mind that these analyses are merely suggestive, given the large number of statistical tests performed. Columns 1 and 2 in Table 3 show effects by in-state status, revealing large, significant effects among in-state students. These students see a 0.013 GPA reduction in the spring semester, a 0.007 reduction in their final cumulative GPA, a 0.059 credit reduction in the spring semester, and a 0.6 percentage point reduction in degree completion, all of which are statistically significant at the 95% or 99% level. All of the coefficients are smaller (or positive), for out-of-state students, and none is statistically significant. Note that the only outcome for which the difference between the effects across columns 1 and 2 is statistically significant is degree receipt (p-value of 0.066), serving as a reminder that the observed pattern of heterogeneity by student in-state status is merely suggestive.

We next present effects separately for students applying to the lottery during seasons when the team made the March Madness postseason tournament (column 3) versus students applying when the team did not make the tournament (column 4). There are large, statistically significant effects among students applying during March Madness seasons. The pattern is most pronounced for credits earned and degree receipt: students during March Madness seasons experience a statistically significant 0.7 percentage point decline in degree receipt, while the point estimate for students during seasons when the team did not make the tournament is 0.000 (SE=0.003). However, none of the differences by March Madness are statistically significant, again reinforcing the fact these patterns are only suggestive.

We next examine effects by the two basic student demographic variables in our data: race and sex. To examine heterogeneity by student race, we combine Hispanic and Black students (column 7) into one subgroup given their relatively small sample sizes, also examining effects for White (column 5) and Asian (column 6) students, the largest minority group in our sample. We find large negative effects among Asian students. For example, Asian students experience a 0.040

GPA point reduction in the spring semester due to winning the lottery, four times greater than the 0.010 effect among all students. They experience a 0.032 point reduction in the average fall and spring GPA, more than four times greater than the 0.007 point effect among all students. They experience a 0.013 point reduction in final cumulative GPA, almost three times larger than the 0.005 point effect among all students. There are no statistically significant effects among Whites or Blacks and Hispanic students. However, as with the earlier results show in Table 3, most of the differences by race are not statistically significant, again reinforcing the suggestive nature of any patterns of heterogeneity. Finally, we examine heterogeneity by student gender, revealing no clear pattern.^{18,19}

Given the statistically significant declines in degree receipt that we observe among the most affected student groups (e.g., in-state students, males), we attempt to explore whether these groups also experience increases in their time-to-degree among those who eventually earn a degree. Nineteen percent of degree earners in our sample earn their degree in five or more years, as opposed to within four years. We restrict our sample for this analysis to only those students who we can observe for six or more years after their freshman year. While the results are relatively imprecise, we observe a pattern of declines in the likelihood of earning a degree within four years, and increases in the likelihood of earning a degree in five or more years, suggesting possible effects of winning the lottery not only on degree receipt, but also on time-to-degree (see Appendix Table 4).

As described at the beginning of this section, the larger ITT effect of game attendance for certain students could reflect either a greater sensitivity to game attendance or a larger effect of winning the lottery on game attendance. As one way to attempt to separate out these possible mechanisms, we examine effects of winning the lottery on game attendance for each of the student subgroups examined in Table 3. Appendix Table 2, columns 2-10 present these results. As a

¹⁸ We also examine heterogeneity by legacy status and by grade level (i.e., first-year, sophomore, junior, senior), and find no pattern of differential effects by these student characteristics. We similarly find no differential effect of winning the lottery by whether the applicant won the lottery in a previous year. While we would like to examine effects by other student characteristics, for example, whether students belong to a fraternity or sorority, we do not observe these characteristics in our data.

¹⁹ We also examine heterogeneity by student major. Unfortunately, we only observe major for degree earners, so this analysis requires conditioning on an endogenous variable. With that caveat in mind, we follow Dynarski et al. (2013) to categorize majors into three groups: 1) STEM, 2) economics/business, and 3) all other majors, following major designations in National Science Foundation (2011). We find suggestive evidence that the negative academic effects of winning the lottery are concentrated among students earning a degree in economics and business-related fields (see Appendix Table 5).

reminder, column 1 shows a 0.603 game, or 20.7%, increase in games attended by lottery winners relative to lottery losers.

Columns 2 and 3 show that the effect on games attended among in-state versus out-of-state students was very similar (0.558 and 0.635 games, or, 19% vs 23% respectively). This leads to similar TOT scale factors across these two groups, and thus, we conclude that the effects concentrated among in-state students reflect heterogeneity in the effects of game attendance rather than differing propensities to attend games after winning the lottery. On the other hand, students winning the lottery during March Madness seasons attended substantially more games than winners during other seasons, suggesting that much or all of the greater effect of winning the lottery during March Madness seasons could be due to a greater treatment intensity, rather than academic performance being more sensitive to game attendance during those seasons. Finally, effects by race lie somewhere in the middle, with Asian students experiencing a larger effect on game attendance than White students, but less than Black and Hispanic students, suggesting that some of the greater effects for Asians compared to Whites may be due to the greater number of games attended by Asian lottery winners. Black and Hispanic students see the largest effects on game attendance, but no effect on academic outcomes, suggesting their academic outcomes are not sensitive to game attendance.²⁰

V.III Robustness of Academic Performance Effects

We present several robustness checks for the effects of winning the lottery on students' academic outcomes in Table 4. Column 1 presents our baseline estimates using our preferred ITT specification with covariates (column 2 from Table 2). Given that we cluster our standard errors at the grade-year level, and that the number of clusters, 54 (6 lottery grade levels by 9 years), is not very large, we present in Column 2 estimates using heteroskedasticity-robust, but not clustered, standard errors. The precision of our estimates is similar, and all of our previously statistically significant estimates remain significant with the exception of the average fall and spring GPA, which loses statistical significance.

²⁰ Note that Asian students attend the fewest games (control mean of 1.8, compared to the overall control mean of 2.9). Note also that winning the lottery increases female game attendance more than male game attendance (columns 9 and 10).

Given that the overwhelming majority of students in our sample is undergraduates, and grading, credit accumulation, and degree receipt work differently for undergraduate and graduate students, in column (3) we drop the small number of lottery applications by graduate students (1,425 observations). The effects are nearly identical focusing only on undergraduates. Next, in column 4, we drop first-years from our sample, because they are missing prior cumulative GPA and credits earned, two covariates that we include in our estimating equation.²¹ Again, the effects are nearly identical on this reduced sample. One minor difference is that we now see a marginally significant negative effect on credits earned in the fall semester of the lottery. Moving to our next check, as described in footnote 6, our effects on degree receipt are estimated using a somewhat smaller sample, and so we show the GPA and credits earned results for this sample in column 5. Once again, the effects are very similar, and again, one minor difference is that there is a statistically significant decline in credits earned in the fall semester of the lottery. Finally, given that lottery applicants' subsequent decisions to apply to the lottery in later years is endogenous, we keep only students' first lottery application year. Column 6 shows the results estimated using this student-level sample, revealing similarly sized effects on cumulative GPA, final credits earned, and degree receipt, with somewhat larger effects (though not statistically significantly so) on contemporaneous GPA and credits earned. Overall, while there are small changes in the magnitude and precision of our point estimates, the effects of winning the lottery on students' academic outcomes are quite robust.²²

V.IV Donation Behavior

We present effects of game attendance on parental and student donations to the University in Table 5. Due to the timing of our data, we are limited to examining donations up to ten years after graduating for the oldest cohort in our sample, and for as little as a few months after graduating for our most recent observed cohort. While ideally we would measure donations further out in time, these first several years after graduating are a key period during which to study giving

²¹ In our main specification, we set missing values to zero and include a dummy variable for missing prior GPA and credits earned.

²² One additional concern is that our results may be biased due to grade curving if the lottery winners' worse academic performance causes lottery loser performance to mechanically increase (Waddell and Putz, 2023). We believe this would be a cause for concern if the lottery winners represented a substantial portion of the student population. However, only about 12% of undergraduate students ($=0.33 \times 0.36$, because 33% of students apply, and 36% of those win) win the lottery each year, so we think it is unlikely that their small GPA reduction would cause any meaningful increase among the other 88% of students.

behavior due to strong habit formation in alumni donations (Meer, 2013). For example, using giving histories back to 1982, Meer (2013) finds that for the University he studies, over 80% of alumni donors gave during the first five years after graduation.²³ Noting that the recent graduates in our data donate smaller amounts than their parents, we begin by combining student and parent donations. We examine whether winning the lottery affected whether either the student or parent donates, the total combined donation amount of the student and parents, and whether the total combined student and parent donations were at or greater than the 90th percentile of the conditional student plus parent donation distribution.

As noted in Section III, donations are extremely skewed: the (unconditional) mean combined student and parent donations is \$350, and the standard deviation is \$10,827. Most students have zero combined student and parent donations, the 90th percentile of the (unconditional) distribution is \$125, the 95th percentile is \$340 (smaller than the mean), the 99th percentile is \$3,190, and top values are in the millions of dollars. Clearly, this is not a distribution that lends itself to OLS regression. We use several approaches to estimate effects on donations in the face of this unique distribution. For the sake of simplicity and ease of interpretation, the approach that we use in our main analysis is to set donation values greater than the 99th percentile of the unconditional distribution to equal the value of the 99th percentile threshold, and then use OLS to estimate effects on this censored total donation variable. We show in Appendix Table 6 that our results are robust to estimating OLS censored at different percentiles, Poisson, zero-inflated Poisson, negative binomial, quantile treatment effects, and distribution regressions estimating effects on dummies for donating above quantiles other than the 90th. We also show in Appendix Table 6 zero effects on the number of donations by students and their parents, and on total donations, conditional on donating.

Table 5 reveals no effects of game attendance on student or parent donations. We find precisely estimated zero effects of game attendance on the likelihood of donating, either by students or by parents. For example, the effect of winning the lottery on either the student or parent donating (row 1, column 2) is -0.000 (SE=0.004). We can rule out with 95% confidence a 0.8

²³ While most alumni donors first donate soon after graduating, the likelihood of donating does increase with years since graduation. A large literature studies other factors affecting the timing of university donations, finding that alumni are more likely to donate when the stock market is up, during college reunion years, and when their university is conducting a fundraising campaign or recently experienced athletic success (Clotfelter, 2003; Monks, 2003; Holmes, 2009).

percentage point (2.2%) increase in the likelihood of donating. We estimate a statistically insignificant \$0.79 decline in total combined donations by the student and parents ($SE=\$3.76$), and a -0.000 effect on the likelihood of donating at or above the 90th percentile of donations. We can rule out with 95% confidence a \$6.58 (7%) effect on total donations and a 0.2 percentage point (5%) increase in the likelihood of donating at or above the 90th percentile.²⁴ Separately examining donations by students versus parents reveals a similar pattern of zero effects for both the student and parents on the likelihood of ever donating, donating at or above the 90th percentile, and on total donations.²⁵

We explore heterogeneity in the effects of winning the lottery on donation behavior by the same student subgroups explored previously. As with the academic outcomes, conducting joint tests of differences by group across all of the donation outcomes, we find no statistically significant evidence of significant differences. We generally find consistent evidence across all student subgroups of zero effect of winning the lottery on donations. The sole exception is Asian students, whose parents donate *less* when their child wins the lottery, however, as with the academic outcomes, these differences are not statistically significant, and so this pattern is merely suggestive. Specifically, Asian students' parents are 1.7 percentage points less likely to donate, their total donation amounts are \$12 smaller, and their likelihood of donating at or greater than the 90th percentile of parent donations is 0.4 percentage points lower. Asian students themselves see no effects of winning the lottery on donation behavior. One possible explanation for this negative effect among parents of Asian students, given the large negative academic effects on these students, is that their parents are unhappy with their students' experience of winning the lottery and attending games, and blame the University. Whatever behavioral effects attending games had on these students that caused their academic performance to decline, these effects may have soured their parents to the University, and subsequently caused these parents to reduce their giving.

²⁴ We calculate these statistics by adding the point estimate to the standard error multiplied by 1.96. For example, the point estimate on total donations is -0.79 and the standard error is 3.76, so $6.58 = -0.79 + (1.96 \times 3.76)$.

²⁵ While we observe gift designations, only 1% of our sample donates with a designation of "Athletics." We find no evidence that game attendance affects the likelihood of students or parents donating to Athletics or the total amount of donations toward Athletics.

VI. Conclusion

Consumption amenities are an important determinant of student preferences in sorting across colleges, and colleges cater to these preferences by investing heavily in such amenities in order to attract students (JMS, 2018). Yet little causal evidence exists on the effects of student consumption of postsecondary amenities due to a lack of exogenous variation. We examine the causal impact of student consumption of postsecondary amenities on academic performance and future donation behavior in the context of Big-Time college sports, one of the most prolific and well-funded college consumption amenities (Clotfelter 2011). We avoid selection bias by using 9 years of student lottery ticket data from a large, public flagship university with a consistently high-ranked men's basketball team, where student season tickets are awarded by lottery. Matching the lottery data to registrar and university donation records, we show that game attendance has small negative effects on GPA, concentrated in the bottom of the GPA distribution, and on credits earned, with no impact on donation behavior.²⁶ We find suggestive evidence that the negative academic effects are driven by in-state students, students attending during years when the basketball team makes the March Madness postseason, and Asian students. The stronger negative effects for these specific groups extend as far as reducing rates of degree completion. While the academic effects are small in magnitude, our results suggest that college investments in consumption amenities to attract students may hinder student academic success, with little upside in the form of increased future donations.

One shortfall of our study is that the university administrative data that we use, while rich in many dimensions, does not allow for an examination of mechanisms for the negative academic effects we find. In order to shed light on possible mechanisms, we include a couple questions on the student survey related to how game attendance might affect student behavior. When students were asked whether they studied more or less, or were more or less likely to skip class, on days on or surrounding games they attended, students responded that they were less likely to study, and more likely to miss class (Figures 2a and 2b reports these results). These responses support potential mechanisms for the negative academic effects we find, though we still lack a more comprehensive examination of mechanisms.

²⁶ While we do not find evidence of increases in donations, it is worth noting that aside from future donations, another benefit of consumption amenities like college athletics is the revenue that it generates for the school. For the University studied in this paper, the men's basketball program generates millions of dollars annually.

Another weakness of our study is that we only observe one measure of student connection to the University after they graduate – specifically, donation behavior – and game attendance may have increased students’ feelings of connection to the University in other ways not reflected in their donations. This seems to be the case: we also asked on the student survey whether game attendance made students feel more or less “connected” to the University, and students overwhelmingly responded positively (see Figure 2c). Thus, it seems plausible that attendance may have improved students’ feeling of connection to the University, in spite of our observed zero effect on donations.²⁷

Big-Time college sports is just one specific type of consumption amenity in which colleges invest to attract students, and the effects of student participation in amenities may differ for various amenities. Further, as highlighted by Jacob et al (2018), it is possible that student participation in such amenities could improve non-academic skills or social networks rewarded in the labor market that we cannot observe using our university administrative data. While we leave these questions for future research, our study offers some of the first causal evidence that college catering to student preferences for consumption may hinder student academic performance, with little future financial payoff to universities.

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²⁷ We estimate the effects of winning the lottery on all three of these outcomes from the student survey by including them as the dependent variable in Equation 1. Unfortunately, given that the sample size is less than one tenth of that of our main sample, the results are too imprecise to be informative. We also examine whether survey responses differ by student race, gender, in-state status, or whether the team makes the March Madness tournament, and find no evidence of heterogeneity in students’ responses.

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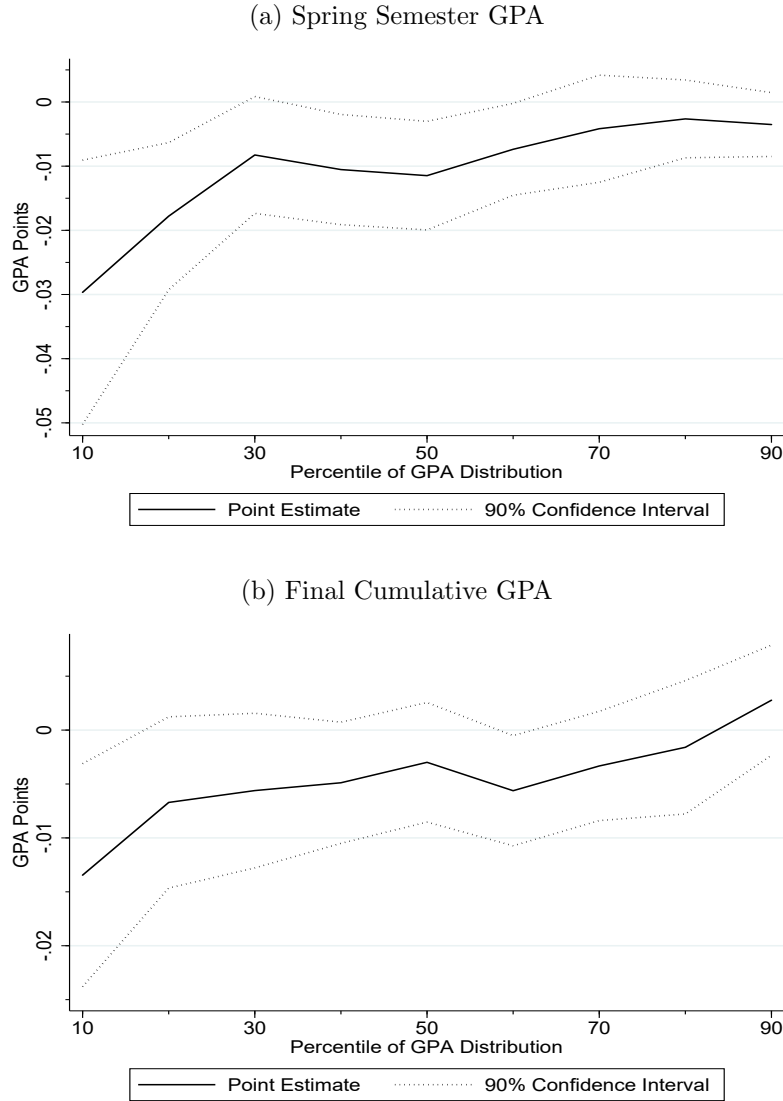
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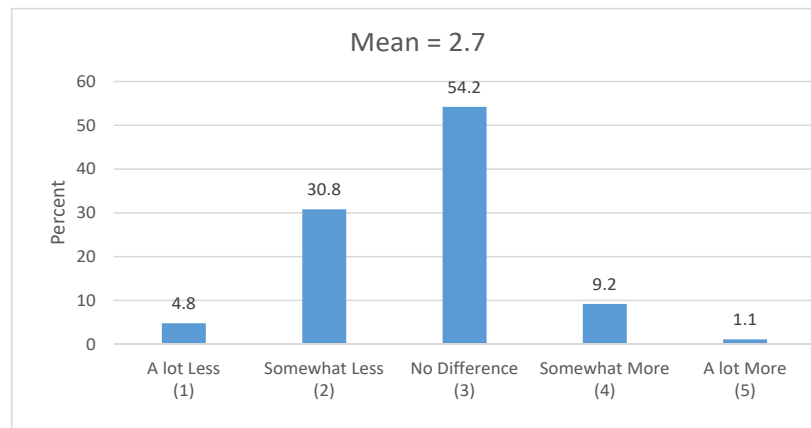
Figure I: Quantile Treatment Effects of Winning the Lottery on GPA



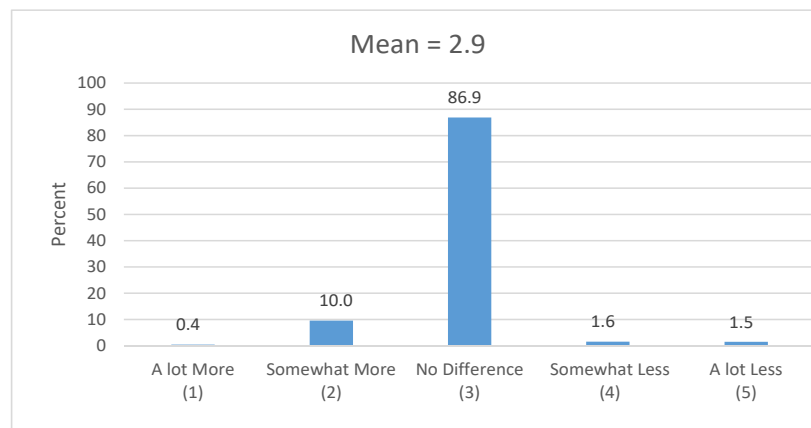
Notes: The figures present quantile treatment effects of winning the lottery on spring semester GPA (a) and final cumulative GPA (b) at the 10th, 20th,..., 90th deciles of the GPA distributions. We follow Firpo, Fortin, and Lemieux (2009) to estimate unconditional quantile regressions using re-centered influence functions (RIFs) ordinary least squares. See main text for more details.

Figure II: Effects of Game Attendance on Studying, Class Attendance, and Connectedness

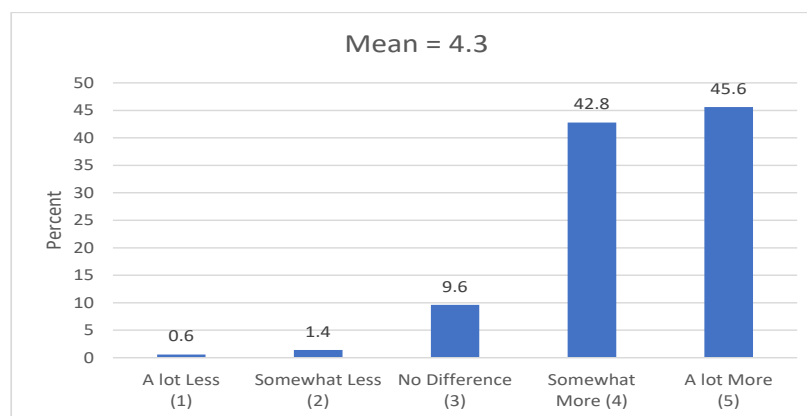
(a) More or Less Time Studying



(b) More or Less Likely to Skip Class



(c) More or Less Connected to University



Notes: The figures show responses to questions from the student and alumni survey about game attendance, student behavior, and feelings of connectedness to the University. Subfigure (a) shows responses to the question: “Do you study more or less on the days surrounding basketball games that you *do* attend compared to the days surrounding games that you do not attend (so when you either watch the game on TV or not at all)?” Subfigure (b) shows responses to the question: “Do you miss class more or less on the days surrounding basketball games that you *do* attend compared to the days surrounding games that you do not attend (so when you either watch the game on TV or not at all)?” Subfigure (c) shows responses to the question: “To what extent do you feel that attending basketball games has made you feel more or less connected to the University (for example, increased or decreased your “school spirit”)?”

Table 1. Sample Means and Balance Test

	Mean	Std. Dev.	Win Coef.	Std. Err.
	(1)	(2)	(3)	(4)
<u>Race</u>				
White	0.703		0.001	(0.005)
Black	0.042		0.000	(0.002)
Hispanic	0.067		-0.003	(0.003)
Asian	0.091		-0.001	(0.003)
Missing	0.090		0.003	(0.002)
Female	0.495		-0.001	(0.004)
In-State	0.733		-0.001	(0.004)
Legacy	0.116		0.001	(0.003)
Cum. GPA, Prior Sem.	3.25	0.45	-0.006	(0.004)
Cum. Credits, Prior Sem.	70.9	29.3	0.129	(0.121)
<u>Parent</u>				
Donated Prior	0.299		-0.002	(0.004)
Total Donations	678	22,383	-10.7	(171.1)
<u>Roommate</u>				
Has At Least One	0.617		-0.003	(0.003)
Has Multiple	0.124		-0.002	(0.003)
Applied to Lottery	0.378		0.003	(0.004)
Won Lottery	0.135		0.006	(0.004)
Cum. GPA, Prior Sem.	3.20	0.51	0.001	(0.007)
Cum. Credits, Prior Sem.	50.7	27.6	-0.038	(0.205)
Sample Size	57,909			
Number of Students	30,299			

Notes: The sample is at the student-year level and contains all men's basketball student season ticket lottery applicants from 2009-2017. Columns 1 and 2 provide sample means and standard deviations. Columns 3 and 4 show the coefficient and standard error on a dummy for winning the lottery from our preferred specification where the dependent variable is the characteristic in each row.

*** = 99% significance, ** = 95%, * = 90%

Table 2. Effects of Game Attendance on Academic Outcomes

Dependent Variable:	ITT (1)	ITT (2)	TOT (3)	Control Mean (4)
<u>Grade Point Average (GPA)</u>				
Fall of Lottery	-0.005 (0.003)	-0.003 (0.004)	-0.014	3.248
Spring after Lottery	-0.012** (0.005)	-0.010** (0.005)	-0.048	3.260
Average Fall and Spring	-0.008** (0.003)	-0.007** (0.003)	-0.031	3.254
Final Cumulative	-0.007** (0.003)	-0.005*** (0.001)	-0.023	3.272
<u>Credits Earned</u>				
Fall of Lottery	-0.030* (0.017)	-0.028 (0.017)	-0.134	14.44
Spring after Lottery	-0.053** (0.023)	-0.052** (0.021)	-0.249	14.34
Total Fall Plus Spring	-0.083*** (0.030)	-0.080*** (0.029)	-0.383	28.78
Final Credits Earned	-0.040 (0.154)	-0.054 (0.141)	-0.255	120.96
Earned Bachelors Degree	-0.0038 (0.0024)	-0.0035 (0.0024)	-0.017	0.924
Controls	N	Y	Y	

Notes: The sample is as in Table 1 (N=57,909 student-year observations), except for the bottom row (N=50,426), which drops cohorts who were too recent for us to observe their on-time graduation status in our data, which ends in summer 2018 (e.g., freshmen applying to the lottery in fall 2017). Column 3 is an informal TOT estimate calculated by dividing the ITT in column 2 by 0.207, given the 20.7% increase in games attended by lottery winners.

*** = 99% significance, ** = 95%, * = 90%

Table 3. Heterogeneity in the (ITT) Effects of Game Attendance on Academic Outcomes

Dependent Variable:	In-State Status		March Madness		Race			Sex	
	Yes	No	Yes	No	White	Asian	Black/Hisp	Female	Male
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
<u>Grade Point Average (GPA)</u>									
Spring after Lottery	-0.013**	-0.000	-0.010*	-0.011	-0.008	-0.040**	-0.001	-0.013*	-0.008
	(0.005)	(0.009)	(0.006)	(0.007)	(0.006)	(0.016)	(0.023)	(0.007)	(0.007)
Average Fall and Spring	-0.010**	0.004	-0.008*	-0.006	-0.004	-0.032**	-0.006	-0.007	-0.007
	(0.004)	(0.007)	(0.004)	(0.005)	(0.004)	(0.012)	(0.020)	(0.005)	(0.005)
Final Cumulative	-0.007***	0.002	-0.007***	-0.003*	-0.004	-0.013*	0.001	-0.007**	-0.003
	(0.002)	(0.006)	(0.002)	(0.002)	(0.003)	(0.007)	(0.014)	(0.003)	(0.003)
<u>Credits Earned</u>									
Spring after Lottery	-0.059**	-0.031	-0.079*	-0.029	-0.027	-0.172*	-0.051	-0.030	-0.072*
	(0.025)	(0.043)	(0.039)	(0.018)	(0.025)	(0.101)	(0.094)	(0.028)	(0.036)
Total Fall Plus Spring	-0.105***	-0.004	-0.118**	-0.048	-0.023	-0.242	-0.215	-0.045	-0.113**
	(0.035)	(0.070)	(0.048)	(0.032)	(0.040)	(0.168)	(0.159)	(0.048)	(0.046)
Final Credits Earned	-0.158	0.281	-0.099	-0.010	-0.107	-0.047	0.428	-0.127	0.040
	(0.169)	(0.218)	(0.242)	(0.162)	(0.182)	(0.631)	(0.674)	(0.207)	(0.230)
Earned Bachelors Degree	-0.006**	0.004	-0.007*	-0.000	-0.004	-0.006	0.002	0.001	-0.007**
	(0.003)	(0.004)	(0.004)	(0.003)	(0.002)	(0.008)	(0.011)	(0.003)	(0.003)
Sample Size	42,432	15,477	27,301	30,608	40,717	5,295	6,349	28,669	29,240

Notes: The sample is as in Table 1 (N=57,909 student-year observations), except for the bottom row (N=50,426) , which drops cohorts who were too recent to observe their on-time graduation status in our data, which ends in summer 2018 (e.g., freshmen applying to the lottery in fall 2017). The specification is as in column (2) from Table 2 (i.e., Intent-to-Treat with controls). In-state status refers to where a student was living when they applied to the University. March Madness refers to whether the basketball team earned a spot in the postseason. The sample sizes do not sum to 57,909 across the race subgroups, because 9% of the sample did not report their race.

*** = 99% significance, ** = 95%, * = 90%

Table 4. Robustness Checks for (ITT) Effects of Game Attendance on Academic Outcomes

Dependent Variable:	Baseline	Baseline, Robust SEs	Drop Grad. Students	Drop First- Years	Drop Recent Cohorts	Only First Application
	(1)	(2)	(3)	(4)	(5)	(6)
<u>Grade Point Average (GPA)</u>						
Fall of Lottery	-0.003 (0.004)	-0.003 (0.004)	-0.002 (0.003)	-0.002 (0.004)	-0.004 (0.004)	-0.006 (0.006)
Spring after Lottery	-0.010** (0.005)	-0.010** (0.005)	-0.010** (0.005)	-0.010* (0.005)	-0.008* (0.004)	-0.014** (0.006)
Average Fall and Spring	-0.007** (0.003)	-0.007 (0.004)	-0.006* (0.003)	-0.006* (0.003)	-0.006* (0.003)	-0.010* (0.006)
Final Cumulative	-0.005*** (0.001)	-0.005* (0.003)	-0.005*** (0.001)	-0.004*** (0.001)	-0.006*** (0.001)	-0.006 (0.003)
<u>Credits Earned</u>						
Fall of Lottery	-0.028 (0.017)	-0.028 (0.022)	-0.026 (0.017)	-0.031* (0.016)	-0.037** (0.018)	-0.051 (0.031)
Spring after Lottery	-0.052** (0.021)	-0.052** (0.025)	-0.055** (0.021)	-0.048* (0.025)	-0.057** (0.024)	-0.073*** (0.027)
Total Fall Plus Spring	-0.080*** (0.029)	-0.080** (0.040)	-0.081*** (0.027)	-0.079** (0.032)	-0.094*** (0.032)	-0.125*** (0.039)
Final Credits Earned	-0.054 (0.141)	-0.054 (0.151)	-0.076 (0.142)	-0.024 (0.151)	-0.032 (0.153)	0.056 (0.235)
Earned Bachelors Degree	-0.004 (0.002)	-0.004 (0.002)	-0.004 (0.002)	-0.004 (0.003)	-0.004 (0.002)	-0.003 (0.004)
Sample Size	57,909	57,909	56,484	46,316	50,426	30,297

Notes: Column (1) reports replicates column (2) from Table 2 (i.e., ITT with controls). Column (2) uses heteroskedasticity-robust standard errors instead of clustered standard errors. Column (3) drops graduate students from the sample. Column (4) drops first-years from the sample. Column (5) drops cohorts who were too recent for us to observe their on-time graduation status in our data, which ends in summer 2018 (e.g., first-years applying to the lottery in fall 2017). Column 6 keeps students' first lottery application year only, so includes one observation per student.

*** = 99% significance, ** = 95%, * = 90%

Table 5. Effects of Game Attendance on Donation Behavior

Dependent Variable:	ITT (1)	ITT (2)	TOT (3)	Control Mean (4)
<u>Student and Parent</u>				
Ever Donated	-0.001 (0.004)	-0.000 (0.004)	-0.001	0.366
Total Donations	-0.91 (4.08)	-0.79 (3.76)	-3.76	91.94
Donated at 90th Percentile	-0.000 (0.002)	-0.000 (0.001)	-0.001	0.037
<u>Student</u>				
Ever Donated	-0.001 (0.004)	-0.000 (0.003)	-0.001	0.220
Total Donations	0.12 (0.30)	0.15 (0.30)	0.70	8.05
Donated at the 90th Percentile	0.001 (0.001)	0.001 (0.001)	0.003	0.021
<u>Parent</u>				
Ever Donated	-0.001 (0.003)	0.000 (0.002)	0.001	0.215
Total Donations	-1.78 (3.75)	-1.71 (3.45)	-8.12	76.67
Donated at the 90th Percentile	-0.000 (0.002)	-0.000 (0.001)	-0.002	0.021
Controls	N	Y	Y	

Notes. The sample is as in Table 1 (N=57,909 student-year observations). Total donations is the sum across all donations after the lottery through summer 2018, censored at the 99th percentile. Donated at 90th percentile is a dummy equal to one if the total donation amount is greater than the 90th percentile of total donations among donors. Column 3 is an informal TOT estimate calculated by dividing the ITT in column 2 by 0.207, given the 20.7% increase in games attended by lottery winners.

*** = 99% significance, ** = 95%, * = 90%

Table 6. Heterogeneity in the (ITT) Effects of Game Attendance on Donation Behavior

Dependent Variable:	In-State Status		March Madness		Race			Sex	
	Yes	No	Yes	No	White	Asian	Black/Hisp	Female	Male
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
<u>Parent and Student</u>									
Ever Donated	0.003 (0.004)	-0.010 (0.007)	0.005 (0.006)	-0.005 (0.005)	0.002 (0.004)	-0.010 (0.013)	-0.012 (0.014)	0.004 (0.005)	-0.004 (0.005)
Total Donations	-0.424 (4.550)	-0.905 (4.248)	-6.043 (5.268)	4.122 (5.024)	-2.125 (4.786)	-12.939** (6.168)	0.143 (4.935)	-2.187 (4.231)	1.325 (5.504)
Donated at 90th Percentile	0.001 (0.002)	-0.002 (0.002)	-0.001 (0.002)	0.001 (0.002)	-0.002 (0.002)	-0.005 (0.003)	0.000 (0.002)	-0.001 (0.002)	0.001 (0.002)
<u>Student</u>									
Ever Donated	0.004 (0.004)	-0.010 (0.006)	0.004 (0.005)	-0.004 (0.005)	0.004 (0.005)	-0.006 (0.014)	-0.012 (0.011)	0.008 (0.005)	-0.007 (0.005)
Total Donations	0.308 (0.418)	-0.309 (0.476)	0.062 (0.547)	0.205 (0.299)	0.217 (0.371)	-0.407 (0.845)	-0.642 (0.703)	0.152 (0.391)	0.183 (0.396)
Donated at the 90th Percentile	0.001 (0.002)	-0.002 (0.002)	-0.001 (0.002)	0.002 (0.001)	0.001 (0.001)	-0.004 (0.004)	-0.002 (0.004)	0.001 (0.002)	0.001 (0.002)
<u>Parent</u>									
Ever Donated	0.001 (0.003)	-0.000 (0.006)	-0.000 (0.004)	0.001 (0.003)	0.001 (0.003)	-0.017** (0.008)	-0.005 (0.009)	-0.001 (0.003)	0.001 (0.004)
Total Donations	-1.034 (4.167)	-2.673 (3.939)	-7.682* (4.336)	3.843 (4.775)	-2.996 (4.536)	-11.908** (5.609)	1.929 (4.221)	-3.146 (3.950)	0.357 (4.817)
Donated at the 90th Percentile	0.000 (0.002)	-0.002 (0.002)	-0.003 (0.002)	0.002 (0.002)	-0.001 (0.002)	-0.004* (0.002)	0.001 (0.002)	-0.001 (0.002)	0.001 (0.002)
Sample Size	42,432	15,477	27,301	30,608	40,717	5,295	6,349	28,669	29,240

Notes. The sample is as in Table 1 (N=57,909 student-year observations). Total donations is the sum across all donations after the lottery through summer 2018, censored at the 99th percentile among donors. Donated at 90th percentile is a dummy equal to one if the total donation amount is greater than the 90th percentile of total donations among donors. The specification is as in column (2) from Table 5 (i.e., Intent-to-Treat with controls). In-state status refers to where a student was living when they applied to the University. March Madness refers to whether the basketball team earned a spot in the postseason. The sample sizes do sum to 57,909 across the race subgroups, because 9% of the sample did not report their race.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 1. Number of Lottery Applicants and Fraction Winners by Grade and Year

Panel A: Applicants

Grade Level	2009	2010	2011	2012	2013	2014	2015	2016	2017	Total
First-Years	1,781	1,200	1,426	1,132	1,191	1,302	1,507	948	1,106	11,593
Sophomores	2,111	1,815	1,793	1,420	1,143	1,562	1,614	1,233	1,218	13,909
Juniors	2,137	2,293	2,144	1,471	1,358	1,494	2,001	1,550	1,471	15,919
Seniors	1,630	2,032	2,206	1,439	1,213	1,554	1,703	1,647	1,639	15,063
Grad. Students	95	178	241	181	37	141	206	153	193	1,425
Total	7,754	7,518	7,810	5,643	4,942	6,053	7,031	5,531	5,627	57,909

Panel B: Fraction Won

Grade Level	2009	2010	2011	2012	2013	2014	2015	2016	2017	Total
First-Years	0.177	0.227	0.183	0.239	0.293	0.241	0.200	0.251	0.250	0.224
Sophomores	0.259	0.299	0.247	0.357	0.434	0.340	0.292	0.377	0.357	0.319
Juniors	0.315	0.392	0.318	0.432	0.515	0.435	0.352	0.452	0.473	0.398
Seniors	0.385	0.449	0.380	0.525	0.604	0.499	0.442	0.531	0.516	0.472
Grad. Students	0.137	0.225	0.141	0.215	0.297	0.199	0.243	0.268	0.269	0.216
Total	0.281	0.354	0.289	0.391	0.463	0.380	0.324	0.419	0.410	0.359

Notes. The sample is at the student-year level and contains all men's basketball student season ticket lottery applicants from 2009-2017. The table shows the number of lottery applicants (Panel A) and fraction winners (Panel B) by grade and year from 2009-2017.

Appendix Table 2. Effect of Winning the Lottery on Games Attended

Dep. Var.	All	In-State Status		March Madness		Race			Sex	
		Yes	No	Yes	No	White	Asian	Black/Hisp	Female	Male
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Games Attended	0.603*** (0.151)	0.558*** (0.158)	0.635** (0.261)	1.245** (0.418)	0.435*** (0.123)	0.494** (0.210)	0.609** (0.247)	1.093*** (0.310)	0.663*** (0.162)	0.479 (0.284)
Control Mean	2.912	2.978	2.749	3.264	2.795	3.238	1.778	2.128	2.112	4.256
Percent Effect	20.7	18.7	23.1	38.1	15.6	15.3	34.3	51.4	31.4	11.3
TOT Scale Factor	4.8	5.3	4.3	2.6	6.4	6.6	2.9	1.9	3.2	8.9
Sample Size	2,823	2,029	791	628	2,195	1,951	367	330	1,784	1,037

Notes: The sample is at the student-year level (N=2,823) and contains the 1,641 lottery applicants who responded to the survey asking about game attendance. Each column presents results from a separate regression of number of games attended in that season on whether the student won the lottery. The percent effect is the coefficient divided by the control mean. The TOT scale factor is 100 divided by the percent effect. In-state status refers to where a student was living when they applied to the University. March Madness refers to whether the basketball team earned a spot in the postseason. The sample sizes do not sum to 2,823 across the race subgroups, because 9% of the sample did not report their race.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 3. Effects of Game Attendance Using Logit / Probit

	OLS	Logit	Probit
Dependent Variable:	(1)	(2)	(3)
Earned Bachelors Degree	-0.004 (0.002)	-0.004 (0.003)	-0.004 (0.002)
<u>Student and Parent</u>			
Ever Donated	-0.000 (0.004)	-0.000 (0.004)	-0.000 (0.004)
Donated at 90th Percentile	-0.000 (0.001)	-0.001 (0.001)	-0.000 (0.001)
<u>Student</u>			
Ever Donated	-0.000 (0.003)	-0.000 (0.003)	-0.000 (0.003)
Donated at the 90th Percentile	0.001 (0.001)	0.001 (0.001)	0.001 (0.001)
<u>Parent</u>			
Ever Donated	0.000 (0.002)	0.000 (0.003)	0.001 (0.003)
Donated at the 90th Percentile	-0.000 (0.001)	0.001 (0.001)	0.000 (0.001)
Controls	Y	Y	Y

Notes. The sample is as in Table 1 (N=57,909 student-year observations).

Column 1 reports the main results from Tables 2 and 5. Columns 2 and 3 report marginal effects from Logit and Probit estimation, respectively, instead of OLS. Standard errors are reported in parentheses. Donated at 90th percentile is a dummy equal to one if the total donation amount is greater than the 90th percentile of total donations among donors.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 4. Heterogeneity in the (ITT) Effects of Game Attendance on Time-to-Degree

Dependent Variable:	All	In-State Status		March Madness		Race			Sex	
	(1)	Yes	No	Yes	No	White	Asian	Black/Hisp	Female	Male
Ever Earned a Degree	-0.000 (0.002)	-0.003 (0.003)	0.007* (0.004)	-0.003 (0.003)	0.003 (0.003)	-0.001 (0.003)	0.003 (0.008)	0.003 (0.013)	0.002 (0.003)	-0.003 (0.003)
Earned Degree Within 4 Years	-0.005 (0.004)	-0.007 (0.005)	-0.000 (0.006)	-0.012* (0.006)	0.003 (0.004)	-0.007 (0.004)	-0.006 (0.025)	-0.002 (0.019)	0.003 (0.006)	-0.012** (0.006)
Earned Degree In Exactly 5 Years	0.002 (0.003)	0.001 (0.004)	0.004 (0.006)	0.006 (0.004)	-0.003 (0.004)	0.003 (0.003)	0.010 (0.018)	0.001 (0.012)	-0.000 (0.005)	0.004 (0.004)
Earned Degree in 6 or More Years	0.003* (0.002)	0.003 (0.002)	0.004** (0.002)	0.003** (0.001)	0.003 (0.003)	0.003* (0.002)	0.001 (0.009)	0.004 (0.007)	-0.000 (0.002)	0.006*** (0.002)
Earned Degree in 5 or More Years	0.005 (0.003)	0.004 (0.004)	0.008 (0.006)	0.009* (0.005)	-0.000 (0.004)	0.006 (0.004)	0.011 (0.024)	0.005 (0.013)	-0.001 (0.005)	0.010* (0.005)
Sample Size	37,996	28,345	9,649	21,024	16,972	27,088	2,979	3,570	18,521	19,473

Notes: This sample includes the 37,996 student-lottery year observations for students who we can observe for at least 6 years after their freshman year. The control mean among this sample for ever earn a degree is 94.7%, for earn a degree within 4 years is 77%, for earn a degree in exactly 5 years is 15.4%, for earn a degree in 6 or more years is 2.3%, and for earn a degree in 5 or more years is 17.7%.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 5. Heterogeneity in the (ITT) Effects of Game Attendance by Student Major

Dependent Variable:	All Majors (1)	STEM (2)	Econ/Business (3)	Other Majors (4)
<i>Panel A: Academic Outcomes</i>				
<u>Grade Point Average (GPA)</u>				
Spring after Lottery	-0.007* (0.004)	-0.011 (0.007)	-0.023** (0.010)	0.004 (0.007)
Average Fall and Spring	-0.005 (0.003)	-0.005 (0.005)	-0.025*** (0.007)	0.005 (0.006)
Final Cumulative	-0.003* (0.002)	-0.005 (0.003)	-0.007 (0.005)	0.001 (0.004)
<u>Credits Earned</u>				
Spring after Lottery	-0.051** (0.022)	-0.026 (0.034)	-0.120** (0.048)	-0.039 (0.052)
Total Fall Plus Spring	-0.085*** (0.030)	-0.049 (0.057)	-0.231*** (0.056)	-0.041 (0.073)
Final Credits Earned	0.009 (0.144)	0.165 (0.209)	0.119 (0.188)	-0.030 (0.200)
<i>Panel B: Donation Outcomes</i>				
<u>Parent and Student</u>				
Ever Donated	-0.001 (0.004)	-0.005 (0.007)	-0.005 (0.009)	0.008 (0.006)
Total Donations	-3.909 (4.407)	-4.185 (5.048)	8.555 (10.404)	-4.700 (5.059)
Donated at 90th Percentile	-0.001 (0.002)	-0.002 (0.002)	0.003 (0.005)	-0.001 (0.002)
<u>Student</u>				
Ever Donated	0.001 (0.004)	-0.008 (0.006)	0.002 (0.010)	0.013** (0.007)
Total Donations	0.177 (0.368)	-0.505 (0.436)	1.045 (0.949)	0.531 (0.453)
Donated at the 90th Percentile	0.001 (0.002)	-0.003 (0.002)	0.005 (0.004)	0.002 (0.002)
<u>Parent</u>				
Ever Donated	-0.002 (0.002)	0.001 (0.004)	-0.006 (0.007)	-0.001 (0.006)
Total Donations	-4.738 (3.988)	-3.935 (4.519)	6.606 (9.181)	-7.070 (5.291)
Donated at the 90th Percentile	-0.002 (0.002)	-0.003 (0.002)	0.003 (0.004)	-0.002 (0.002)
Sample Size	46,817	20,739	10,444	15,631

Notes: The sample is at the student-lottery year level and includes all students who earned a Bachelors degree (N=46,817 student-year observations). The specification is as in column (2) from Table 2 (with controls). Column 1 reports the effect for the whole sample of degree earners, while columns 2, 3, and 4 split the sample by student major.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 6. Robustness Checks for (ITT) Effect of Game Attendance on Donations

	Student and Parent	Student	Parent
	(1)	(2)	(3)
<u>Total Donations</u>			
<u>OLS, Censoring at:</u>			
90th Percentile	0.108 (0.281)	0.025 (0.033)	0.075 (0.209)
95th Percentile	0.017 (0.679)	0.047 (0.080)	-0.055 (0.503)
99th Percentile	-0.790 (3.759)	0.147 (0.297)	-1.707 (3.447)
99.5th Percentile	-1.502 (7.322)	0.294 (0.473)	-2.877 (6.853)
99.9th Percentile	-21.394 (17.970)	1.208 (0.985)	-21.458 (17.095)
OLS, Not Censoring	-101.772* (57.759)	0.441 (2.271)	-102.214* (57.295)
OLS, Conditional on Donating	-7.18 (9.94)	0.57 (0.79)	-9.67 (9.25)
Poisson	-0.010 (0.041)	0.020 (0.037)	-0.024 (0.046)
Zero-Inflated Poisson	-0.024 (0.040)	0.010 (0.032)	-0.032 (0.043)
Negative Binomial	0.002 (0.043)	0.014 (0.035)	-0.039 (0.062)
<u>Quantile Treatment Effects at:</u>			
80th Percentile	-0.025 (0.114)	0.000 (0.006)	0.000 (0.000)
90th Percentile	-0.377 (0.488)	-0.129 (0.193)	-0.000 (0.029)
95th Percentile	1.469 (1.982)	-0.409 (0.528)	-0.000 (0.136)
<u>Total Donations Greater Than:</u>			
80th Percentile	0.000 (0.002)	0.000 (0.002)	-0.000 (0.002)
90th Percentile	-0.000 (0.001)	0.001 (0.001)	-0.000 (0.001)
95th Percentile	-0.000 (0.001)	0.000 (0.001)	-0.001 (0.001)
<u>Number of Donations</u>	-0.008 (0.028)	0.014 (0.027)	-0.019 (0.037)

Notes: The sample is as in Table 1 (N=57,909 student-year observations). Total donations is the sum across all donations after the lottery through summer 2018, with censoring conducted at percentiles of the unconditional distribution. Quantile treatment effects are conducted following Firpo, Fortin, and Lemieux (2009) using re-centered influence functions (RIFs)-OLS at percentiles of the unconditional distribution. "Total donations greater than" are dummies for donating greater than percentiles of the conditional distribution. Regression with number of donations as dependent variable conducted using Poisson.

*** = 99% significance, ** = 95%, * = 90%

Appendix Table 7. Effects of Game Attendance on Roommate's Academic Outcomes

Dependent Variable:	ITT (1)	ITT (2)	TOT (3)	Control Mean (4)
<u>Grade Point Average (GPA)</u>				
Fall of Lottery	0.002 (0.007)	0.002 (0.007)	0.011	3.168
Spring after Lottery	-0.001 (0.005)	-0.001 (0.004)	-0.003	3.204
Average Fall and Spring	0.001 (0.006)	0.001 (0.005)	0.007	3.177
Final Cumulative	-0.002 (0.006)	-0.001 (0.005)	-0.006	3.229
<u>Credits Earned</u>				
Fall of Lottery	0.022 (0.035)	0.021 (0.033)	0.100	14.273
Spring after Lottery	0.031 (0.031)	0.031 (0.031)	0.149	14.483
Total Fall Plus Spring	0.054 (0.042)	0.054 (0.040)	0.256	28.785
Final Credits Earned	0.116 (0.234)	0.093 (0.226)	0.442	115.771
Earned Bachelors Degree	-0.001 (0.004)	-0.001 (0.004)	-0.005	0.872
Controls	N	Y	Y	

Notes: The sample is all lottery applicants with an on-campus housing roommate (N=35,624 student-year observations), except for the bottom row (N=29,545), which drops cohorts who were too recent for us to observe their on-time graduation status in our data, which ends in summer 2018 (e.g., first-years applying to the lottery in fall 2017). Outcomes are averaged across roommates for applicants with multiple roommates. Column 3 is an informal TOT estimate calculated by dividing the ITT in column 2 by 0.207, given the 20.7% increase in games attended by lottery winners.

*** = 99% significance, ** = 95%, * = 90%