# Students' Preferences, Capacity Constraints and Post-Secondary Achievements in a Non-Selective System 

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## Préférences des étudiants, contraintes de capacité et réussite universitaire dans un système non-sélectif

Cet article étudie l'impact de la satisfaction des préférences des élèves concernant les formations universitaires sur leur parcours dans le supérieur. Pour ce faire, nous croisons les données SISE relatives à l'inscription dans les formations du supérieur en France avec les données administratives du système centralisé français de préinscription dans le supérieur Admission Post Bac (APB). Nous nous concentrons sur les licences universitaires en sous-capacité (dites «en tension »), qui ont eu recours au tirage au sort pour sélectionner leurs effectifs, pour les cohortes de 2013 à 2016. Cette composante aléatoire du système APB est utilisée comme un instrument de l'admission à la formation souhaitée par l'étudiant. Nous trouvons que l'admission à la licence universitaire classée en tête de la liste de vœux des candidats a un impact certain sur les chances de poursuite d'études supérieures : en moyenne, les candidats admis à la suite d'un tirage au sort favorable ont $10 \%$ de chances de plus de s'inscrire dans le supérieur par rapport aux candidats perdants du tirage au sort qui n'ont pu intégrer leur formation préférée. L'admission à la formation préférée a également un impact significatif sur d'autres aspects du parcours des étudiants, comme le maintien dans les études et la validation des années. Les effets estimés sont hétérogènes selon la filière d'étude souhaitée et le profil scolaire des candidats. En particulier, les élèves dont les chances initiales de réussite dans le supérieur sont les plus faibles sont davantage sensibles à leur admission dans leur premier choix.

Mots-clés: Enseignement supérieur, Préférences, Contraintes de capacité, Expérience randomisée, Procédures d'affectation.

Classification JEL : I23, I28, C21, D81.

## Students' Preferences, Capacity Constraints and Post-Secondary Achievements in a Non-Selective System

Using rich administrative data on the French centralized assignment system of admission in higher education Admission Post Bac (APB) paired with data on university enrollment, this article provides new evidence on the impact of satisfying students' stated preferences on their achievements in higher education. To do so, we exploit lotteries embedded in APB to prioritize applicants in oversubscribed university programs as an instrument for admission. Focusing on cohorts 2013 to 2016, we show that admission to one's top-ranked program has a large impact on the pursuit of post-secondary education: on average, it increases students' chances of enrollment into higher education by $10 \%$ from the baseline. It also affects other aspects of students' educational pathways such as persistence in higher education, choice of major and degree completion. Effects are heterogeneous both by programs' field of study and applicants' profile. In particular, students at the margin of pursuing higher education are more sensitive to capacity constraints in their favorite program.

Keywords: Centralized Matching Market, Higher Education, Preferences, Capacity Constraints, Randomized Control Trial.

## Introduction

Access to higher education is a crucial issue in labor economics, as differences in education levels translate into persistent earnings and employment inequalities. The swift rise of mass education at the tertiary level went hand in hand with high drop out and repetition rates, as well as increased strain on the capacity of programs. Among members of the Organization for Economic Co-operation and Development (OECD), the average completion rate of full-time tertiary students only reached $38 \%$ in 2017 (OECD, 2019).

Over the last decade, researchers and policy makers have worked jointly to improve admission systems in higher education, trying to achieve more equity, quality, and mobility for students (European Parliament, 2014). In this process, the quality of the studentprogram match has been identified as a key determinant of success in higher education Arcidiacono et al., 2014, Arcidiacono and Lovenheim, 2016; Hoxby and Avery, 2013 Black et al., 2015; Dillon and Smith, 2017), and assignment systems have been put at the center stage of the debate. Contributions from market design theory have shown that centralized assignment system and the use of specific matching mechanisms can be welfare improving (Niederle and Roth, 2003; Roth and Xing, 1997; Machado and Szerman, 2016, Abdulkadiroğlu et al., 2017a), leading to significant worldwide changes in the way students are allocated to schools and university programs. However, there are still some frictions in these processes, among which mismatch - that is to say the inadequacy between students' and programs' characteristics - has been singled out as a major issue. Still, little is known about the importance of satisfying students' preferences in this phenomenon, and its consequences on students' educational pathways has not been fully investigated. This should be particularly salient in a context where capacity constraints are binding, and highly demanded programs which lack available seats have to reject some applicants. Eventually, not satisfying students' preferences in default of capacity could be costly both individually and collectively. In France, the collective cost of setbacks in students' trajectory is estimated at 500 million euros per year, which corresponds to the entire budget of two middle-size universities (France Stratégie, 2017). Among others, this includes the cost of the inadequacy resulting from students' not having access to their desired curriculum. Opening new seats in specific fields of study still represents a substantial public cost ${ }^{1}$ and must be in line with actual employment prospects to facilitate students' transition into the world of work.

This paper evaluates the impact of satisfying students preferences under capacity constraints on students' achievement. We focus on higher education in France, where capacity constraints and low success rates in first year of university are a major policy concern.

[^1]Among students entering higher education in 2014, only $29 \%$ obtained their bachelor degree within three years, i.e without any grade repetition or interruption MESRI-SIES, 2019), and only one out of 10 students continue their second year of bachelor in the same program as the one chosen in their first year (France Stratégie, 2017).

We exploit two specific features of the French higher education system: i) its high degree of centralization, and ii) its predominant share of non-selective programs, where lotteries were implemented to break ties in programs facing capacity constraints.

From 2009 to 2017, students seeking admission to higher education had to go through a national centralized platform called Admission Post Bac (APB), which covered almost all public and private higher education programs. Students willing to pursue higher education were required to submit a rank-ordered list of programs, and were later allocated through a College-Proposing Deferred Acceptance Mechanism.

The French higher education system is dual. It is composed of two types of higher education programs: selective and university programs. On the one hand, selective programs are provided in various types of public and private institutions, and are allowed to freely rank their applicants. On the other hand, holding a secondary school certificate (Baccalauréat) is the only legal condition to get access to a university program: universities do not rank their applicants one the basis on any additional academic criteria. University programs gather half of high school graduates, and include some of the most popular programs such as medical studies, sports, law or psychology degrees. Since applicants could not be ranked based on their prior academic achievements, non-academic priorities rules were used to rank students applying to a given university program. In cases where capacities were insufficient to offer a seat to all priority applicants (i.e those with the highest level of priority), lotteries were used to break ties.

Using tie-breaking lotteries as an instrument for admission into oversubscribed university programs, we show that, on average, admission to one's first choice increases the probability to be enrolled in higher education within three years after high school by $10 \%$. As a benchmark, Fack and Grenet (2015) show that providing 1,500 euros cash allowances to a comparable population ${ }^{2}$ increases students' enrollment rates in higher education by 5 to $7 \%$. On top of this effect on the extensive margin, the admission to the top-ranked program also affects many aspects of students' educational trajectory, such as persistence in higher education, changes of field of study or degree completion. For instance, students admitted to their first choice are on average more likely to be enrolled in third year of bachelor within three years by $27 \%$. Finally, our results are neither driven by a single type of university program nor by a specific profile of applicant. Yet, our results are heterogeneous according to these two dimensions. In particular, students applying to Sports Studies or Social Sciences are more sensitive to the outcome of the admission procedure.

[^2]The analysis of the heterogeneity of effect by socio-demographic and schooling characteristics shows that low-achieving students and students from vocational high schools are more affected by the admission to their top-ranked program. Students with no relevant alternative are also more affected by the outcome of the assignment procedure. Overall, these results suggest that capacity constraints might be particularly detrimental to students at the margin of pursuing higher education.

Our contribution to the existing literature is twofold. First, our paper focuses on a larger set of programs and applicants than most studies using admission discontinuities do. Capacity constraints generate discontinuities in assignment systems. Applicants ranked close to the admission threshold (either just above or just below) have similar characteristics but different chances of being admitted. These discontinuities have been commonly used as a source of identification to estimate the effect of attending specific schools on academic outcomes. This strand of the literature, with landmark studies on Magnet and Charter schools in the United States, has notably shown that students can benefit from going to another school than their default option (Abdulkadiroğlu et al., 2014, 2017b), which can have a positive impact on college enrollment (Deming et al., 2014) and earnings (Hoekstra, 2009). However, the majority of these papers focus on specific kinds of programs, most of them being elite schools with high academic standards. By design, they also have to restrict their analysis to students at the margin of eligibility, and thus study a narrow set of students. Our paper focuses on a nationwide lottery for admission to the university, which is the main type of higher education programs in France. Since students could not be differentiated based on their grades, our sample consists of individuals of all abilities, with diverse socio-demographic characteristics and schooling preferences. It also encompasses all students with the same level of priority, included students far from the admission cutoff. This characteristic of the French system enables us to assess the effect of being admitted to one's first choice on a comprehensive set of students considering a first entry into higher education. To the best of our knowledge, this is a unique feature of our study, since most papers exploiting school lotteries deal with selective admission process on a narrower set of students. Second, our paper focuses on higher education, where rejections due to capacity constraints are more likely to be salient in students' decisions to stay or leave the educational system. At the end of compulsory schooling, going to the labor market rather than pursuing education becomes a more relevant outside option. To study this trade-off, we build a unique data-set based on linked individual-level administrative data, that allows us to track students every year they apply to and enroll in higher education, at least three years after high school graduation for cohorts 2013 to 2016. Such precision of the data enables us to study students' educational pathways on multiple dimensions such as persistence in higher education, the choice of major and degree completion.

The remainder of the paper is organized as follows. Section 1 describes the institutional background, Section 2 presents the data used for this study and some summary statistics. Section 3 presents the empirical strategy. Section 4 displays the estimates and a preliminary discussion of the results. Section 5 concludes.

## 1 Institutional Background

### 1.1 The French Higher Education System

General Context The French tertiary education system is predominantly public, and pursuing higher education is generally affordable. The majority of educational programs is publicly funded, and private institutions only amount to $18.2 \%$ of higher education students (DEPP and SD-SIES, 2017). Tuition fees for public university are very low (on average 170 euros per year for a bachelor degree). Each student can either pursue an academic or a vocational training, both provided by public institutions.

Until 2018 and the Loi relative à l'orientation et à la réussite des étudiants (ORE) university reform 3 , the French higher education system was composed of two kinds of programs. Undergraduate university programs (Licence) were generally not selective: every student holding the Baccalauréat was entitled to access to a first year of university ${ }^{4}$. Since university programs hosted about half of new entrants in higher education, a significant share of applicants was therefore not recruited on the basis of their academic records. By contrast, all the other available programs - later referred to as selective programs - were able to select applicants at their discretion, and thus to prioritize students with the highest academic standing. Selective programs typically included programs located in other types of educational institutions than universities, such as programs provided in high schools including preparatory classes (Classes Préparatoires aux Grandes Ecoles (CPGE)) and vocational training (Sections de Technicien Supérieur (STS)), or undergraduate business and engineering schools.

In university programs, non-academic priority criteria were used to distinguish between the applicants. Priorities were based on students' educational district and the rank of the program within students' rank-ordered list. Lotteries were then used to break ties among applicants with the same level of priority.

Over our period of study, the increase in available seats at the university only partly offset the growing trend in the number of applicants. In the meantime, concerns were raised regarding the functioning of the French higher education system. The debate mainly focused on oversubscribed university programs, where capacities were not large enough to offer a seat to all the applicants who had the highest level of priority. The use of lotteries to select students became more salient, which lead some students and parents

[^3]to question the fairness of the system. This was one of the main reasons advocated in favor of the introduction of a new platform called Parcoursup in 2018, which changed the matching mechanism and gave universities clearance to prioritize applicants based on their academic records.

Admission Post-Bac (APB) From 2009 to 2017, students seeking admission to higher education ${ }^{5}$ had to go through a national platform called Admission Post Bac (APB), where they could apply to both public and private institutions, including university programs and selective programs. The APB system gathered around 12,000 programs and 800,000 applicants every year. To fulfill their applications, students had to make a rank-ordered list (ROL) of programs, and were later allocated through a centralized matching mechanism, based on the College-Proposing Deferred Acceptance (DA) algorithm (Gale and Shapley, 1962). In the case of APB, applicants were requested to submit and to commit to their ROL from January to March. After having received all applications, programs also ranked their applicants. This ranking was made on a discretionary basis for selective programs, whereas it was automatically generated by the APB system according to both the legal priority rules and the lottery outcome for university programs.

The DA is an algorithm that enables to sequentially allocate seats to students while respecting as much as possible students' preferences and programs' priorities. It proceeds as follows:

Step 1 First choice applications are sent to the programs. Programs tentatively assign their seat to applicants one by one, while respecting priority order (detailed below), until their capacity is reached. Remaining applicants are rejected.

Step $k$ Rejected applicants apply to $k$-th-ranked programs. Once again, programs tentatively assign their seats to applicants one by one according to their priority. The procedure ends when there is no more applicants to assign or seat to fill.

Figure 1: Timeline of the Application and Matching Process in APB


The matching procedure involved three rounds of admission from June to July (Fig-

[^4]ure 11. At the beginning of the first round of admission, students were offered seats by programs based on the matching outcome determined by the DA algorithm, and each student received at most one offer. Students could accept the offer, turn it down or conditionally accept it, waiting for applicants selected by higher-ranked programs to withdraw from the procedure in the subsequent rounds. Applicants could exit the procedure at any point in time. The three-round articulated procedure was made to reallocate vacant seats. In particular, the final results of the Baccalauréat were published between the second and the third rounds of the procedure. Students who failed the exam were not able to compete for a seat anymore, and their seats were re-offered in the third round. Finally, applicants could participate in supplementary rounds, which took place between June and September, and helped students to apply to programs with remaining seats.

### 1.2 University Programs: Priorities and Lotteries

Definitions Our setting is composed of four main elements: applicants, programs, priorities and programs' capacities. Let $i \in\{1, \ldots, I\}$ denote applicants and $s \in\{0,1, \ldots, S\}$ programs, where $s=0$ is an outside option. Let $q_{s}$ stand for program $s$ 's capacity, where $n_{s}$ represents the number of applicants to the program $s$. Applicants have preferences over programs $s$, whereas programs have priorities over applicants. Applicants to a given university program are assigned to priority groups. Applicant $i$ 's priority group at program $s$ is written $p_{i s}$.

Let $C_{i}$ be $i$ 's ROL of programs, where $c_{i s} \in\{\emptyset, 1, \ldots, N\}^{2}$ denotes program $s^{\prime}$ position in $C_{i}$, and $c_{i s}=\emptyset$ means that the program is not ranked by student $i . c_{i s}$ is decomposed into two elements $c_{i s}^{r}$ and $c_{i s}^{a}$, which respectively stand for the relative and the absolute rank of program $s$ in applicant $i$ 's ROL. Program $s$ 's absolute rank corresponds to its position within applicant's $i$ ROL, whereas its relative rank represents its position with respect to the other programs in the same field of study in the ROL. The number of programs that applicants can rank is limited, though this figure varies over the period, from 46 in 2013 to 64 in 2017.

Priority Rules For each university program $s$, the priority group $p_{i s}$ for applicant $i$ is a function of $i$ 's educational district and ROL such that:

$$
p_{i s}=f(\underbrace{1\left\{\text { district }_{i}=\text { district }_{s}\right\}}_{\text {Educational district }}, \underbrace{c_{i s}^{r}, c_{i s}^{a}}_{s^{\prime} \text { position in } C_{i}})
$$

where $p_{i s}<p_{j s}$ means that school $s$ prioritizes $i$ over $j$. Priorities are lexicographic, which implies that $\forall\left(c, c^{\prime}\right) ; f(1, c)<f\left(0, c^{\prime}\right)$, meaning that applicants from the educational district always have priority over outside applicants. Students are then sorted sequentially by their relative and absolute rank: we have $\forall\left(c^{a}, c^{\prime a}\right) \in \mathbb{N}^{* 2}$, if $c^{r}<c^{\prime r}$, then $f\left(., c^{r}, c^{a}\right)<$ $f\left(., c^{\prime r}, c^{\prime a}\right)$. This last feature is unusual, as it has been shown that using applicants'

ROL as a criterion to prioritize applications cancels the strategy-proofness of the DA. We discuss the implication of this specificity for our study at the end of this section.

To illustrate priority rules, let's consider a simple example. Assume five applicants $a, b, c, d$ and $e$ who apply to two Law degrees $S_{1}$ and $S_{2}$, as displayed in table 1. $S_{1}$ is located in Paris, whereas $S_{2}$ is in Créteil, another French educational district. We focus on $S_{1}$ 's ranking of applicants. Applicants $a, b, c$ and $d$ are from Paris, so they have priority over $e$. Applicant $c$ ranked $S_{1}$ as her second absolute choice, so she will lose priority over $a, b$ and $d$. Finally, $b$ and $d$ have ranked $S_{1}$ as their first relative and absolute rank in Law. $a$ has put $S_{1}$ as her first relative rank in Law, but second absolute rank. Consequently, $b$ and $d$ will have priority over $a$.

Table 1: Priority and Random Draw in Non-Selective Programs: An Example

| Applicants | $\mathbf{a}$ | $\mathbf{b}$ | $\mathbf{c}$ | $\mathbf{d}$ | $\mathbf{e}$ |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Catchment area | Paris | Paris | Paris | Paris | Créteil |
| Students' rank-ordered-list |  |  |  |  |  |
| $1^{\text {st }}$ rank | Other | $S_{1}$ | $S_{2}$ | $S_{1}$ | $S_{1}$ |
| $2^{\text {nd }}$ rank | $S_{1}$ | $S_{2}$ | $S_{1}$ | Other | $S_{2}$ |
| $3^{\text {rd }}$ rank | $S_{2}$ | Other | Other | $S_{2}$ | Other |
| Program $S_{1}$ 's ranking |  |  |  |  |  |
| Priority group | 2 | 1 | 3 | 1 | 4 |
| Lottery number | $*$ | 2 | $*$ | 1 | $*$ |
| Final rank | $\mathbf{3}$ | $\mathbf{2}$ | $\mathbf{4}$ | $\mathbf{1}$ | $\mathbf{5}$ |

Notes: $S_{1}$ is located in the district of Paris, $S_{2}$ is located in the district of Créteil.

Lottery For applicants in the same priority group for program $s$ i.e., such as $p_{i s}=p_{j s}$, lotteries are used to break ties. The lottery ensures the "Equal Treatment of Equal" (ETE) property, meaning that students with the same stated preferences and priorities have the same chances of having access to the program. For each program $s$ and applicant $i$, the lottery number $\operatorname{Rand}_{i s}$ is drawn from a discrete uniform distribution $\mathcal{U}$ as follows:

$$
\operatorname{Rand}_{i s} \rightsquigarrow \mathcal{U}\left[1, \sum_{k=1}^{n_{s}} 1\left\{p_{k s}=p_{i s}\right\}\right]
$$

The lottery system works as a multiple tie-breaking rule, meaning that a student is assigned a lottery number for each of her application to a university program. Eventually, for a given program $s$, every applicant is given a rank according to her priority group
and the outcome of the lottery. Hence, $i$ 's position in program $s$ ' ranking of applicants is given by the number of students having priority over $i\left(A_{s}\right)$, plus the tie-breaking random component :

$$
\begin{equation*}
r_{i s}=\underbrace{\sum_{k=1}^{n_{s}} 1\left\{p_{k s}<p_{i s}\right\}}_{A_{s}}+\text { Rand }_{i s} \tag{1}
\end{equation*}
$$

Turning to our illustration (table 1), student $d$ belongs to the same priority group as $b$ : they both belong to $S_{1}$ educational district, and they put $S_{1}$ as their first absolute and relative rank. A random draw is used to break ties between $b$ and $d$ : the issue of the lottery is that $d$ is ranked first, and $b$ second. The final ranking of program $S_{1}$ is therefore: $d, b, a, c, e$.

Top-Priority Group and Oversubscribed Program We focus on the top-priority group of applicants, which we define as the set of applicants with the highest level of priority for a given university program $s$. In our previous example (table 11), the top-priority group would consist of $b$ and $d$.

Applicants from the top-priority group - later called priority applicants - are of particular interest for two reasons. First, they represent the best potential match between programs' institutional criteria and students' preferences according to the matching mechanism. Priority applicants stated their strong preference for entering this very program by ranking it at the top of their ROL.

Top-priority groups are also the only ones among which the probability to receive an offer in the first round of the procedure is only determined by the random draw, suggesting that the outcome of the lottery is very likely to affect students' final assignment. Indeed, priority applicants are also in first positions of programs' ranking, and thus automatically receive an admission offer if their lottery number is below the program's capacity ${ }^{[6]}$ Priority applicants are thus directly affected by capacity constraints, and would be the primary beneficiaries of the opening of new seats in oversubscribed programs.

Turning to equation (1), belonging to the top-priority group means that $A_{s}=0$. Within this group, the probability of a student $i$ to get an offer from program $s$ is straightforward and given by:

$$
\begin{aligned}
\operatorname{Prob}\left(\operatorname{Offer}_{i s} \mid A_{s}=0\right) & =\operatorname{Prob}\left(r_{i s} \leq q_{s} \mid A_{s}=0\right)=\operatorname{Prob}\left(\operatorname{Rand}_{i s} \leq q_{s}\right) \\
& =\frac{q_{s}}{\sum_{k=1}^{n_{s}} 1\left\{p_{k s}=p_{i s}\right\}} \quad \text { if } 0 \leq q_{s}<\sum_{k=1}^{n_{s}} 1\left\{p_{k s}=p_{i s}\right\} \\
& =1 \quad \text { otherwise }
\end{aligned}
$$

[^5]We define as oversubscribed a program $s$ whose capacity is lower than the number of its priority applicants i.e. $0 \leq q_{s}<\sum_{k=1}^{n_{s}} 1\left\{p_{k s}=p_{i s}\right\}$. In these cases, the lottery had a decisive impact on students' chances of admissions. Only priority applicants with a rank below the program's capacity received an offer from the university after the first admission round. This means that the probability of a priority applicant to receive an offer is strictly below one.

Figure 2 displays program's ranking of applicants for an illustrative oversubscribed program. We restrict the sample to students from the priority educational district, which is the first criterion to sort applicants. Shades depict the relative rank of the program in students' ROL, the second criterion to distinguish applicants: the lighter the shade, the larger the relative rank. Finally, the Y axis gives applicants' absolute rank, the last sorting criterion. As marked by the vertical dash line, the capacity of the program (200) was not large enough to satisfy every applicant from the top-priority group, that is to say students from the priority district who ranked the program as their first relative and absolute choice. Only priority applicants randomly ranked between 0 and 200 received an offer during the first round of the matching procedure.

Figure 2: Program's Ranking of Applicants and Capacity: The Case of an Oversubscribed Program


Source: Admission Post Bac data from 2013 to 2017 (MESRI-SIES).
Notes: The figure displays absolute rank in students' rank-ordered list (ROL) as a function of their ranking by the program. Shades depict relative ranks in student's ROL: the lighter the shade, the larger the relative rank. Priority applicants have ranked the program as their first absolute and relative rank.

Preference Revelation in APB The APB matching mechanism is based on the so-called DA algorithm. Among its main features, this mechanism is strategy-proof, meaning that ranking the programs truthfully is a dominant strategy for all applicants (Gale and Shapley, 1962). Nevertheless, the use of students' ROL in the priority rules of university programs raises the issue of truthful revelation of preferences in APB. For instance, one could argue that some applicants who think they have no chance to be admitted in their first-best program strategically ranked a second-best university program at the top of their ROL (which makes them part of the top priority group for this program) in order to maximize their chances to be recruited somewhere.
The question of the correspondence between submitted ROL and applicants' preferences is complex, but the principal concerns are addressed in the following lines. First, the APB system of priority rules in university programs has become public information only during summer 2016, which reduces the chances of applicants strategically reporting ROL that do not correspond to their true preferences. More specifically, if district priority was
a well known fact, the use of students' ROL to differentiate applicants was not official before the publication of the algorithm. Students were explicitly advised to rank programs according to their true preference, both on the official website and documentation available to students and teachers. Second, the APB system being a mix of a proper DA mechanism for selective programs and a rather different system (a variant of a Boston mechanism) for university programs, the window for strategic behaviors is not clear from a theoretical perspective. We cannot exclude that some students might have not reported their true preference as their first choice. However, since students are ranked according to their ROL in the Boston Mechanism, there is an important weight given to the first choice in the decision making process (Abdulkadiroğlu et al., 2011). This was a real incentive to put a program they preferred, at least among programs they considered as achievable.

## 2 Data and Summary Statistics

### 2.1 Data

Administrative sources Our analysis is based on linked individual-level administrative data which allows us to track students throughout their academic careers. We combine two main sources of information, the APB databases (2013-2017) and the Système d'Information sur le Suivi de l'Etudiant (SISE) databases (2013-2018), which were provided by the statistical service of the Directorate General of Higher Education (Sousdirection des systèmes d'Information et des Etudes Statistiques) (DGESIP/DGRI- SIES) and the Directorate of Evaluation, Prospective and Performance (Direction de l'Évaluation, de la Prospective et de la Performance) (MENJ - DEPP).

The APB files contain comprehensive information on students' ROL of programs, programs' rankings of applicants, and the matching outcome. They directly stem from memory files of the matching procedure. For each round of the procedure, we are able to know applicants' admission status for each of their applications, i.e. if the applicant has been rejected, if she has been offered a seat and accepted the offer, or if she chose to conditionally accept the offer to wait for other responses from higher-ranked programs. APB databases also provide information on students' socio-demographic characteristics (parents' socio-economic status, students' gender, age, place of birth, place of residence and scholarship status in high schoo ${ }^{7}$ ) as well as schooling characteristics, such as the type of Baccalauréat prepared (general or vocational) and distinctions obtained at the Baccalauréat. Additional details about the databases are displayed in the appendix A. The APB databases are matched with the SISE databases, using a unique student identifier, the Identifiant National Etudiant (INE), a 10 digit code assigned to every student who went through the French educational system. The SISE databases are records of all

[^6]students registered in public universities and of the majority of students enrolled in private institutions. In total, the SISE programs represent about $80 \%$ of students enrolled in higher education in France. We complete the SISE databases with the BPBAC data-sets, which record enrollment in selective programs located in high schools, namely Sections de Technicien Supérieur (STS) and Classes Préparatoires aux Grandes Ecoles (CPGE) ${ }^{8}$. These students represent more than $10 \%$ of the total number of students enrolled in higher education. For each year a student is enrolled in an educational institution, the SISE databases contain detailed information on the institution and degree attended. These data are available for school years 2013 to 2018. We have therefore information up to six years after APB admissions for the 2013 cohort, whereas we are limited to two years after the matching procedure for the 2017 cohort.

Sample Restriction Our study focuses on priority applicants to university oversubscribed programs between 2013 to 2017. We restrict our analysis to high school applicants who successfully complete their applications. We also restrict our sample to programs where at least 30 priority applicants received an offer in the first round of the procedure, and at least 30 others did not. Our final sample of analysis is a cross-sectional data-set of 102,072 high school priority applicants to oversubscribed programs.

### 2.2 Summary Statistics

Oversubscribed Programs' Characteristics University programs became increasingly attractive over the period (table 7 in appendix A). Their share in students' ROL went up from $28 \%$ in 2013 to $43 \%$ in 2017. More importantly, university programs constitute an increasing proportion of students' top-ranked program: they accounted for $39 \%$ of students' first choice in 2017, against $31 \%$ in 2013. Part of the explanation for this trend is the change in the rules of APB: during the period, students had to rank at least one university program for their ROL to be valid. Meanwhile, the number of seats filled in university programs throughout the three regular rounds of APB only increased slowly. This growing pressure on university programs mechanically increases the risk of seeing more oversubscribed programs over the years. Further details about applications in APB are displayed in appendix A.

Table 2 summarizes information on oversubscribed programs included in our sample, which are grouped into five main fields of study based on the French official classification of programs. Following the above mentioned general trend in applications, the number of oversubscribed programs in our sample has raised from 39 in 2013 to 110 in 2017, which corresponds to a $161 \%$ increase in the number of priority applicants over five years.

[^7]The greatest jump occurs between 2016 and 2017, the year before the implementation of Parcoursup. In relation to the total number of university programs, oversubscribed programs account for only a small fraction of the higher education supply. Nonetheless, their capacity represents 4 to $9 \%$ of university programs' total capacity, and students concerned by the lottery represent $10 \%$ of university's available seats on average over the period, which represents a sizable share of students entering in higher education.

Table 2: Summary Statistics of Oversubscribed Programs

|  | 2013 | 2014 | 2015 | 2016 | 2017 | All years |
| :--- | :--- | :--- | :--- | :--- | :--- | ---: |
| Oversubscribed Programs |  |  |  |  |  |  |
| Health \& Sciences | 5 | 2 | 3 | 3 | 10 | 23 |
| Humanities | 3 | 2 | 2 | 5 | 13 | 25 |
| Law \& Economics | 8 | 9 | 7 | 10 | 19 | 53 |
| Social Sciences | 7 | 8 | 16 | 11 | 26 | 68 |
| Sports Studies | 16 | 22 | 28 | 30 | 42 | 138 |
| All fields | 39 | 43 | 56 | 59 | 110 | 307 |
| Inertia | $82 \%$ | $66 \%$ | $86 \%$ | $100 \%$ | $100 \%$ | $89 \%$ |
| Priority Applicants |  |  |  |  |  |  |
| Lottery winners | 8,835 | 8,450 | 10,038 | 11,376 | 20,127 | 58,826 |
| Lottery losers | 5,235 | 5,042 | 6,987 | 9,370 | 16,612 | 43,246 |
| All | 14,070 | 13,492 | 17,025 | 20,746 | 36,739 | 102,072 |
| University programs' capacity | $8 \%$ | $7 \%$ | $8 \%$ | $9 \%$ | $15 \%$ | $10 \%$ |

Source: APB administrative data from 2013 to 2017, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Oversubscribed university programs with at least 30 lottery winners and 30 lottery losers.
Notes: Inertia refers to the share of programs which stay oversubscribed throughout the period. The last row of the table displays the percentage of total university programs' capacity that priority applicants represent.

Of the 102,072 applicants in our sample, $58 \%$ are randomly selected to receive an offer in the first round of the procedure. As the number of priority applicants increases and the number of seats available does so but not as rapidly - the share of lottery winners slightly decreases over the period, from $62 \%$ in 2013 to $54 \%$ in 2017. Finally, once oversubscribed, programs tend to remain as such in subsequent years. Around $43 \%$ of the programs in our sample appear more than one year and, on average, $89 \%$ of them stay oversubscribed until the end of the period from the moment they entered our sample.

Oversubscribed programs are spread all over the country, with a larger concentration for the region Ile-de-France and great urban centers as Lyon. Bachelor degrees specialized in Sports Studies are the most represented field in our sample. Sports programs prepare students not only for the Physical Education (PE) examination, but also for a variety of jobs related to sport (coach, sport management, bridges with physiotherapist degrees etc.).

Good labor market prospects for graduates has pushed demand up for such programs (France Stratégie and Cereq, 2019), with the number of oversubscribed sports degrees almost tripling over the period ${ }^{10}$ (Table 22). Sports is not only the most represented field of study, but also the category with the largest number of priority applicants. Sports degrees are followed by Social Sciences and Law and Economics. Social Sciences oversubscribed programs are mainly composed of Psychology degrees, while Law degrees represent the principal type of oversubscribed programs in Law and Economics (appendix A table 9).

Priority Applicants' Characteristics Table 3 compares priority applicants in our sample (left-hand side) with APB applicants who ranked another type of program at the top of their list as well as the whole set of APB applicants (right-hand side) along several sociodemographic and schooling characteristics. Academic ability is measured by distinctions obtained at the Baccalauréa ${ }^{117}$. There are five different type of distinctions, depending on the final GPA (over 20): Highest honors (above 16), High honors (from 14 to 16), Honors (from 12 to 14), Graduation with no distinction (from 10 to 12), and Failure (below 10). The other schooling characteristics are the age relative to the usual age of students preparing the Baccalauréat and the type of high school attended (general or vocationa ${ }^{[2]}$. Socio-economic status has been aggregated into four categories (High SES, Medium-High SES, Medium-Low SES, and Low SES) based on 2 digit socio-professional categories, following the usual definition of the French Ministry of Education. The other socio-demographic characteristics are gender and need-based scholarship status.

Even though our sample only covers a small fraction of the entire APB platform, it is still very representative of high school students applying through the platform, and exhibits a great variety of programs and applicants' profiles. On average, priority applicants have a very similar profile as the other applicants in APB, and especially with those who ranked an undersubscribed university program as their first choice. The composition of priority applicants also substantially varies across fields of study. Sports degrees only gather $25 \%$ of female students, instead of $84 \%$ for Social Sciences programs. Some oversubscribed programs tend to recruit higher-achieving students, or students from high socio-economic status. For instance, Science and Health programs consist of students with an excellent academic level, $14 \%$ of priority applicants having graduated high school with

[^8]the highest honors (against $8 \%$ for the entire APB platform). In contrast, $56 \%$ of Sports priority applicants ( $50 \%$ in Social Sciences) obtained their Baccalauréat with no distinction. Regarding students' social background, $42 \%$ of priority applicants to Humanities degrees come from very high socio-economic status (against $29 \%$ on average for all APB applicants). Sports and Social Sciences also recruit students from a relatively lower social background.

Table 3: Summary Statistics: Priority Applicants

|  | Priority Applicants in Oversubscribed University Programs |  |  |  |  |  | All Applicants, by Top-Ranked Program |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  <br> Sciences | Humanities | Field of st <br>  <br> Economics | dy <br> Social <br> Sciences | Sports <br> Studies | All <br> Fields | Undersubscribed <br> University <br> Programs | Selective <br> Programs | All <br> Programs |
| Count (k) | 10 | 4 | 16 | 18 | 55 | 102 | 827 | 1,843 | 2,934 |
| Socio-demographic characteristics |  |  |  |  |  |  |  |  |  |
| Female | 0.60 | 0.69 | 0.64 | 0.84 | 0.25 | 0.46 | 0.64 | 0.48 | 0.52 |
| Scholarship beneficiary | 0.14 | 0.15 | 0.21 | 0.2 | 0.16 | 0.17 | 0.2 | 0.2 | 0.2 |
| SES |  |  |  |  |  |  |  |  |  |
| Very High | 0.41 | 0.42 | 0.4 | 0.26 | 0.27 | 0.31 | 0.31 | 0.28 | 0.29 |
| Medium-High | 0.14 | 0.13 | 0.11 | 0.17 | 0.18 | 0.16 | 0.16 | 0.15 | 0.15 |
| Medium-Low | 0.27 | 0.29 | 0.25 | 0.32 | 0.32 | 0.31 | 0.29 | 0.3 | 0.30 |
| Low | 0.17 | 0.15 | 0.21 | 0.23 | 0.21 | 0.21 | 0.23 | 0.26 | 0.25 |
| No information | 0.01 | 0.01 | 0.02 | 0.01 | 0.01 | 0.01 | 0.01 | 0.01 | 0.02 |
| Schooling characteristics |  |  |  |  |  |  |  |  |  |
| General track | 0.94 | 0.86 | 0.78 | 0.77 | 0.71 | 0.76 | 0.84 | 0.48 | 0.58 |
| Age on time | 0.75 | 0.67 | 0.66 | 0.68 | 0.66 | 0.67 | 0.7 | 0.64 | 0.65 |
| Baccalauréat Distinctions |  |  |  |  |  |  |  |  |  |
| Highest honors | 0.14 | 0.04 | 0.06 | 0.02 | 0.01 | 0.03 | 0.07 | 0.09 | 0.08 |
| High honors | 0.2 | 0.13 | 0.15 | 0.1 | 0.06 | 0.1 | 0.15 | 0.16 | 0.15 |
| Honors | 0.26 | 0.28 | 0.27 | 0.27 | 0.21 | 0.24 | 0.26 | 0.29 | 0.28 |
| No distinction | 0.33 | 0.44 | 0.41 | 0.5 | 0.56 | 0.5 | 0.41 | 0.39 | 0.41 |
| Failed | 0.07 | 0.1 | 0.1 | 0.11 | 0.16 | 0.13 | 0.1 | 0.07 | 0.09 |
| Applications in the ROL |  |  |  |  |  |  |  |  |  |
| Average number | 6.4 | 7.8 | 9.6 | 5.3 | 6.8 | 6.9 | 4.6 | 7.8 | 6.9 |
| Same field as the $1^{\text {st }}$ choice | 0.77 | 0.69 | 0.65 | 0.71 | 0.63 | 0.66 |  |  |  |
| Only one program ranked | 0.19 | 0.09 | 0.04 | 0.09 | 0.06 | 0.07 | 0.22 | 0.06 | 0.11 |

Source: APB administrative data from 2013 to 2017, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Left-hand side: Priority applicants applying to an oversubscribed university program with at least 30 lottery winners and 30 lottery losers. Right-hand side: all the high school applicants preparing the Baccalauréat by the end of the year.
Notes: Socio-economic status has been divided into four categories (High SES, Medium-High SES, Medium-Low SES, and Low SES) based on two-digit socio-professional categories. Academic ability is measured by distinctions obtained at the Baccalauréat. The student gets the highest honors if she graduated with a GPA that is above 16 over 20 , high honors if her GPA is between 14 and 16 , and honors if her GPA is between 12 and 14. Students graduate with no distinction if their GPA is below 12 over 20, and fail the Baccalauréat if their GPA falls below 10.

Dynamics of Offers and Admissions Figure 3 depicts the dynamics of students' admission throughout the three rounds of the APB procedure. By construction, all lottery
winners received an offer from their top-ranked program in the first round, which they could accept or reject. At the end of the procedure, $76 \%$ of them have been admitted to the program. The rest of these applicants failed the Baccalauréat $(11 \%)$ or voluntary resigned (14 \%). Among lottery losers, $65 \%$ directly received an offer from a program ranked below in their ROL and could decide to accept the offer or postpone their answer to wait for a potential offer coming from their first choice in rounds 2 or 3. After the last round of admission, $29 \%$ of lottery losers were eventually admitted to their first choice, and $29 \%$ were admitted to a lower ranked program. Only $11 \%$ of lottery losers remained without any offer, which means that most of them had the possibility to pursue higher education through their application to APB.

Figure 3: Dynamics of Admission to Top-Ranked Program, by Treatment Assignment


Lottery losers ( $Z=0$ )


Lottery winners ( $Z=1$ )

Source: APB administrative data from 2013 to 2017, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Oversubscribed university programs with at least 30 lottery winners and 30 lottery losers.
Notes: The figure compares the admission status of applicants by treatment assignment at the beginning (programs' offers) and the end (applicants' responses) of the three regular rounds of the APB procedure.

Two-Sided Non Compliance As shown in Figure 3, there are different pathways from lottery assignment to admission ${ }^{[13}$. On the one hand, some applicants who received an offer might deliberately reject it, whereas other applicants who failed the Baccalauréat are not able to accept any offer. These students are not sensitive to the treatment assignment,

[^9]and should be seen as never-takers. On the other hand, some applicants who narrowly lost the lottery (i.e. those with a lottery number Rand $_{i s}$ that is just above the program's capacity $q_{s}$ ) still managed to enroll in their first choice during the subsequent rounds of the procedure by benefiting from seats freed by never-takers. Since they get the treatment without being assigned to it, these applicants are considered as always-takers. This group of students is likely to be specific, as vacant seats of the $2^{\text {nd }}$ and $3^{\text {rd }}$ rounds are offered to the pool of applicants who have not already accepted another offer (or resigned from the system) in the first round. For instance, these applicants waited for a potential new offer from their top-ranked program because they were more informed about the rules of the procedures, less risk-adverse, or more willing to enroll in the program. Conversely, a subset of unassigned students might also have cancelled their application due to endogenous factors. In particular, some applicants could have accepted the offer they received from a lower-ranked program (or an outside option) in the first round of admission only because they received it earlier. As shown by Grenet et al. (2019), it is costly to learn information about programs in a centralized system like APB, and students tend to accept early offers more often, even though they are not preferable. As applicants do not get any information about their rank and their probability to receive an offer in the next rounds of the procedure, some might directly accept the offer they received in the first round.

Eventually, the Equal Treatment of Equals (ETE) property ensures that conditional on the priority group, first-round offers are sent independently of confounding factors that could induce an omitted variables bias. However, this is not the case for final admission because of two-sided non-compliance. Applicants who accept the offer represent a selfselected subset of those who received it, and it is likely that they differ on unobserved characteristics that are both related to academic achievement and offer acceptance. Besides, the offers sent in the subsequent rounds of the procedure are only sent to applicants who are still competing for a seat, and cannot be considered as random anymore.

## 3 Empirical strategy

### 3.1 Design

Our purpose is to assess the causal effect of being admitted to one's top-ranked program on post-secondary educational outcomes. To do so, we take advantage of the exogenous variation embedded in the APB centralized assignment system. Lottery tie-breaking assigns priority applicants to oversubscribed programs as in a stratified randomized trial, where students exogenously received an offer (or not) from their top-ranked program.
Formally, let $Y_{i s c}$ be a post-secondary educational outcome for student $i$ from cohort $c$ (2013 to 2017) who put the program $s$ at the top of her ROL. We define the treatment $D_{i s c}$ as being admitted to one's top-ranked program. We therefore consider as assigned
to treatment lottery winners i.e, priority applicants who received an offer from their topranked program in the first admission round of the procedure ( Rand $_{i s c} \leq q_{s c}$ ). Lottery losers are considered as unassigned to the treatment. Treatment assignment is thus defined by the dummy $Z_{i s c}=1\left\{\operatorname{Rand}_{i s c} \leq q_{s c}\right\}$, which equals one if student $i$ is a lottery winner.

Local Average Treatment Effect To get an accurate picture of the effect of admission to one's top-ranked program, a simple comparison of students with respect to their admission status to the top-ranked program would be misleading, as well as a comparison of students with respect to the outcome of the lottery (the treatment assignment).

To overcome this issue, we estimate a Local Average Treatment Effect (LATE) focusing on the sub-population of compliers (Angrist et al. 1996), i.e., applicants who are sensitive to the treatment assignment in the sense that they are actually willing to enroll in the top-ranked program. We instrument admission $D_{i s c}$ to the top-ranked program by treatment assignment $Z_{i s c}$ in the first round of admission ${ }^{[14}$. The LATE, i.e., the effect of the treatment on the compliers, is given by the Wald estimator:

$$
\beta=\frac{E\left[Y_{i s c} \mid Z_{i s c}=1\right]-E\left[Y_{i s c} \mid Z_{i s c}=0\right]}{E\left[D_{i s c} \mid Z_{i s c}=1\right]-E\left[D_{i s c} \mid Z_{i s c}=0\right]}
$$

Estimation We estimate the following linear probability model by two-stage least squares (2SLS):

$$
\begin{equation*}
Y_{i s c}=\alpha+\beta D_{i s c}+\lambda_{s c}+X_{i}^{\prime} \mu+\nu_{i s c} \tag{2}
\end{equation*}
$$

where $X_{i}$ is a vector of individual constant socio-demographic and schooling characteristics (gender, referent parent's SES, need-based scholarship status, age when applying to APB, type of Baccalauréat - general or vocational - and Baccalauréat distinctions) ${ }^{15}$, $\nu_{i s c}$ are individual error terms, and $\lambda_{s c}$ are program $\times$ cohort fixed effects. These latter are used to distinguish between the applicants to the 307 different programs where the randomization has occurred. As there are different probabilities of assignment among our randomization units, we follow Abadie et al. (2017) and cluster our standard errors at the program $\times$ cohort level. Our first stage is given by the following equation:

$$
\begin{equation*}
D_{i s c}=\alpha+\beta Z_{i s c}+\lambda_{s c}+X_{i}^{\prime} \mu+\epsilon_{i s c} \tag{3}
\end{equation*}
$$

[^10]where $\epsilon_{i s c}$ are individual error terms, and our treatment $D_{i s c}$ is instrumented by our treatment assignment $Z_{i s c}$.

Outcomes of Interest We define four cumulative outcomes to investigate the different channels through which admission to one's first choice might affect students' trajectory in higher education. We set a three-year time frame from APB admissions to study three families of outcomes: (1) enrollment decision, (2) persistence in higher education and (3) completion of studies. Enrollment decision is captured by a dummy equal to one if the applicant has ever enrolled in higher education within three years. This provides evidence on whether capacity constraints can generate substantial discouragement effects on the pursuit of higher education. Persistence in higher education is measured by two outcomes: being enrolled every year in higher education, and being enrolled in higher education in the same field of study every year. Finally, students' completion of studies is measured by a dummy variable equal to one if the student is enrolled in her third/final year of bachelor within three years.
Due to data constraints, we exclude cohort 2017 from all the following estimations, since we are not able to track students three years after their high school graduation for the moment. Therefore, only 197 clusters (program $\times$ cohort) are considered to compute the LATE estimates ${ }^{16}$

### 3.2 Validity of the Strategy

Balancing Checks Since receiving an offer to a top-ranked oversubscribed program is the result of a lottery, priority applicants should have similar observable and unobservable characteristics with respect to treatment assignment $\left(Z_{i s c} \perp X_{i s c}, \epsilon_{i s c}\right)$. Figure 4 graphically summarizes the results of the balancing checks on observed socio-demographics and schooling characteristics defined in section 2.2, namely gender, students' social background, relative age, type of high school and distinctions at the Baccalauréat.

These tests are derived from linear probability models where our treatment assignment dummy $Z$ is regressed on a single observable characteristic and fixed effects for each cluster (program $\times$ cohort), our unit of randomization. These fixed effects are necessary to compare applicants with the same ex-ante probability of assignment to treatment $\sqrt{17}$.

[^11]Figure 4 displays the estimate and the $95 \%$ confidence interval for the characteristic of interest for each of the six regression models.

In almost all cases, lottery winners and losers do not significantly differ on observable characteristics, which suggests that students are homogeneously distributed between the two groups that were created by the random draw. It should be noted that the coefficient for female ( -0.008 ) is the only one with a p-value below 0.05 ( 0.048 ): this suggests that male were slightly more likely to be randomly drawn. Alternative specifications have been tested as robustness checks. We first have considered a joint model including the whole set of characteristics and failed to reject the joint hypothesis test of the nullity of all the characteristics. We also have considered nonlinear specifications, namely Logit and Probit. These estimates yield to the same conclusions as our linear specification (Table 10 in Appendix B).
This confirms the assumption that lottery winners and losers present also similar unobserved characteristics. Hence, it seems reasonable to think that lottery winners and losers started with equal chances of success in higher education before the beginning of the APB procedure, which strengthens the credibility of our estimates.

Figure 4: Balancing Checks of Selected Observed Characteristics in Our Final Sample


Source: APB administrative data from 2013 to 2016, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Oversubscribed university programs with at least 30 lottery winners and 30 lottery losers.
Notes: The figure gives the estimated effect of each observable characteristic on the probability to receive an offer from the program in the first round of APB (lottery winner), i.e. being assigned to the treatment group. The coefficients are estimated in separated linear probability models including program $\times$ cohort fixed effects. The red bars show $95 \%$ confidence intervals computed with robust standard errors clustered at the program $\times$ cohort level.

First Stage Our instrument meets the usual conditions to be valid. First, it satisfies the rank condition: it is positively correlated with the treatment, since only applicants being offered a seat at some point during the procedure can be admitted to their first choice. By design, our instrument is independent of the vector of potential outcomes and treatment assignment. We believe that it also satisfies the exclusion restriction and the monotonicity assumption. Indeed, we posit that there is no defiers in our setting, as it seems very unlikely that some applicant would accept an offer only because they have not been offered one in the first round of admission.

It has been shown that in some encouragement designs, random assignment to the treatment can have an effect in itself on the outcome of interest. In fact, lottery losers
could feel they have been treated unfairly, which could have an impact on their academic achievement. They could also misinterpret losing the lottery as a signal that they are not a good fit for the university, and lose confidence in their ability to succeed. Though we cannot prove formally that our instrument verifies the exclusion restriction assumption, we assume that the lottery is not likely to have an effect in itself on post-secondary outcomes, and only affects students' achievement through admission to one's first choice.

As previously discussed, there is imperfect compliance in our setting. However, this is not a threat to the internal validity of our design, since lottery winners are still more likely to be admitted to their first choice than lottery losers at the end of the procedure. Results for the first stage estimates are reported in Table 4 Our instrument turns out to be very strong, as the value of the F-stat is large (around 500). It is associated with a $120 \%$ increase from the baseline probability of being admitted to one's first choice for lottery losers (of $31 \%$ ). Because our instrument is a lottery, our estimates are robust to the inclusion of pre-treatment observed characteristics as control variables (see column 3).

Table 4: First Stage Estimates

|  | Outcome: Admission to $1^{\text {st }}$ choice (0/1) |  |  |
| :--- | :---: | :---: | :---: |
|  | Raw difference | Program $\times$ cohort <br> fixed effects | Full model |
| Marginal effect of the instrument |  |  |  |
| Baseline probability | 0.31 | 0.31 | 0.31 |
| Lottery winner | $0.47^{* * *}$ | $0.40^{* * *}$ | $0.40^{* * *}$ |
| F-stat | $(0.02)$ | $(0.02)$ | $(0.02)$ |
| Socio-demographic characteristics | 503 | 507 | 506 |
| Schooling characteristics |  |  | $\checkmark$ |
| Program $\times$ cohort fixed effects |  |  | $\checkmark$ |
| Nb. of applicants | 65,333 | 65,333 | $\mathbf{\checkmark}$ |
| Nb. of clusters (program $\times$ cohort $)$ | 197 | 197 | 65,333 |
| $R^{2}$ | 0.22 | 0.28 | 197 |

Notes: $\quad{ }^{*} \mathrm{p}<0.05 ;{ }^{* *} \mathrm{p}<0.01 ;{ }^{* * *} \mathrm{p}<0.001$; standard errors clustered at the program* cohort level in parenthesis. Socio-demographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions. The baseline probability is the mean of the outcome variable for lottery losers $(Z=0)$.
APB cohorts: 2013 to 2016.

## 4 Results

### 4.1 Main Results

Table 5 displays our main estimates of the treatment effect for our five outcomes of interest. The first column displays the results for the estimation of the effect of the lottery i.e., the effect of the Intent-to-Treat (ITT), and the second column shows the estimation of the LATE by 2SLS. The estimated coefficients are displayed in percentage points (p.p.).
For each specification and outcome, we provide a baseline probability of the outcome in order to give an idea of the magnitude of the treatment effect. For ITT estimates, the baseline probability is simply the observed mean of the outcome for lottery losers. Since the estimated LATE gives a treatment effect for the sub-population of compliers, we would like to display in the second column the baseline probability for compliers only i.e., the mean of the outcome for untreated compliers. This value is unknown, as it is not possible to individually identify compliers in our setting. As a second-best proxy, we compute a baseline probability that is the observed mean of the outcome for untreated lottery losers $\left(Z_{i s c}=0\right.$ and $\left.D_{i s c}=0\right)$ who obtained their Baccalauréat. This last filter helps reducing the share of potential never-takers, as failure at the Baccalauréat forces students to reject any admission offer regardless of the lottery outcome.

The treatment effect estimates are significantly different from zero at the $1 \%$ level for all higher education outcomes and all specifications. Overall, we find that having access to one's first choice improve students' chances to pursue and persist in higher education. Unsurprisingly, ITT estimates display lower coefficients due to imperfect compliance. Nevertheless, these coefficients are also significantly different from zero at the $1 \%$ level for all the outcomes we consider. Later in this section, we focus on the LATE estimates. The former estimates include all priority applicants. However, applicants who failed the Baccalauréat are not allowed to enroll in higher education. As a robustness check, we estimated the same models on the subset of students who actually obtained their diploma, and were thus entitled to enroll in higher education. Results are displayed in appendix in Table 11, and yield to the same conclusions, which is not surprising, since lotteries created balanced samples in terms of students' high school academic performances (Figure 4).

Students' Phasing out When outcomes become more difficult to achieve, the baseline probability of success declines. Only a few students manage to have an educational pathway without interruption, repetition or changes of field of study. If the baseline probability of ever being enrolled in higher education is $75 \%$, it falls to $27 \%$ for being enrolled three years successively without changing fields. This is evidence of students phasing out, as the baseline probability of having an uninterrupted trajectory (being enrolled in $3^{\text {rd }}$ year of bachelor after three years) is only $22 \%$. This has a direct consequence for the interpreta-
tion of our results: if our treatment effects are close in absolute terms for all the outcomes we consider, they represent a different relative impact if we take account of gaps in the baseline probability of success.

Enrollment Decision On average, students who were admitted to their first choice are more likely to have ever been enrolled in higher education within three years by 8 p.p.. This is a sizable effect, as it represents a $10 \%$ increase from the baseline probability of $75 \%$. Following students up to three years after their Baccalauréat enables us to capture not only a immediate disappointment effect the year following the end of the matching procedure, but what seems to be a lasting discouragement effect on students who did not obtained their first choice. This suggests that capacity constraints on students' preferred program might have a disincentive effect on the pursuit of higher education. This is true besides any consideration on the characteristics of the preferred program itself (quality of teaching, difficulty, student support etc.).

Persistence and Completion On average, students who were admitted to their first choice are more likely to have been enrolled every year successively by 8 p.p., and also to have done it without changing field of study by 12 p.p., which respectively corresponds to an increase by $20 \%$ and $44 \%$ from the baseline. In terms of academic success, students who were admitted to their first choice are more likely to be enrolled in $3^{\text {rd }}$ year of bachelor three years after their high school graduation by 6 p.p., which corresponds to a $27 \%$ increase from the baseline. Overall, students who obtained their first choice are thus less likely to change their field of study, but also less likely to have repeated a class or taken a gap year. In relative terms, these effects are larger than the one we observe on the decision to pursue higher education. This suggests that our estimates of the treatment effect on students' persistence and completion do not solely capture the lagging effect of enrollment decision, but something that could be the result of the quality of the match between students and programs, or due to the characteristics of the programs, namely the difficulty of their curriculum or their ability to support students.

Table 5: Effect of Admission to Top-Ranked Program on Various Higher Education Outcomes

|  | ITT | LATE |
| :--- | :---: | :---: |
| Treatment effects 3 years after admission |  |  |
| 1. At least one registration |  |  |
| Baseline probability | 0.78 | 0.75 |
| Treatment effect | $0.03(0.00)^{* * *}$ | $0.08(0.01)^{* * *}$ |
| 2. Enrolled every year | 0.41 | 0.40 |
| Baseline probability | $0.03(0.00)^{* * *}$ | $0.08(0.01)^{* * *}$ |
| Treatment effect |  |  |

3. Enrolled every year in the same field of study

| Baseline probability | 0.30 | 0.27 |
| :--- | :---: | :---: |
| Treatment effect | $0.05(0.01)^{* * *}$ | $0.12(0.01)^{* * *}$ |

4. Enrolled in $3^{\text {rd }}$ year of bachelor

| Baseline probability | 0.23 | 0.22 |
| :--- | :---: | :---: |
| Treatment effect | $0.02(0.00)^{* * *}$ | $0.06(0.01)^{* * *}$ |

## Controls

| Socio-demographic characteristics | $\checkmark$ | $\checkmark$ |
| :--- | :---: | :---: |
| Schooling characteristics | $\checkmark$ | $\checkmark$ |
| Program $\times$ cohort fixed effects | $\checkmark$ | $\checkmark$ |
| Nb. of applicants | 65,333 | 65,333 |
| Nb. of clusters (program $\times$ cohort) | $\mathbf{1 9 7}$ | 197 |

Notes: ${ }^{*} \mathrm{p}<0.05 ;{ }^{* *} \mathrm{p}<0.01 ;{ }^{* * *} \mathrm{p}<0.001$; standard errors clustered at the program $\times$ cohort level in parenthesis. ITT: Intent-to-treat, LATE: Local average treatment effect. Socio-demographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions. In column 1, the baseline probability is the mean of the outcome variable for lottery losers $(Z=0)$. In column 2 , the baseline probability is the mean of the outcome variable for untreated lottery losers $(Z=D=0)$ who obtained the Baccalauréat. APB cohorts: 2013 to 2016.

### 4.2 Treatment Effect Heterogeneity

To understand who are the individuals that are the most affected by the outcome of the admission procedure, we turn to an analysis of the heterogeneity of the treatment. We first focus on between-program differences in the intensity of the treatment, and then explore between-students heterogeneity based on a set of pre-treatment individual characteristics.

## Heterogeneity by field of study of top-ranked program

One of the main sources of heterogeneity in our sample lies in the different types of field of study represented among oversubscribed programs. Their recruitment largely differs in terms of socio-economic background and academic status. Indeed, as described in Table 3. oversubscribed programs in Social Sciences and Sports Studies tend to attract students coming from less favored social background and with lower academic level. Science and Health programs recruit students with very high academic level, whereas programs in Humanities are particularly demanded by applicants from upper social background. Fields of study also have different gender ratios, some being predominantly composed of male students (Sports Studies) or of female students (Social Sciences). Finally, each type of program is specific regarding the content of its curriculum, and does not necessarily have the same number of equivalent within selective programs. This represents another source of variation that might have a differentiated impact on students' trajectory in higher education.

We investigate this potential source of heterogeneity by separately re-estimating treatment effects for each of our five main fields of study. The results are illustrated in Figure 5. We find significant effects at the $5 \%$ level for all type of fields of study except Humanities (probably due to small sample size), which suggest that the effect of the treatment is not field-specific. Students applying to Sports seem to be the most affected by the outcome of the admission procedure. On average, priority applicants to Sports programs who obtained their first choice are more likely to ever be enrolled in higher education by 10 p.p. ( $13 \%$ from the baseline), and also to be enrolled three times successively without switching field of study by 15 p.p. ( $56 \%$ from the baseline). Regarding completion, we observe that on average, those who obtained their first choice are $37 \%$ more likely to be enrolled in $3^{\text {rd }}$ year of bachelor three years after high school (8p.p.). Interpretation is similar for students applying to Social Sciences.

The sign and the magnitude of these effects are consistent with the content and the difficulty of each curriculum. For instance, students who obtained their first choice in Science and Health are actually less likely to be enrolled in $3^{\text {rd }}$ of bachelor three years after high school (-9p.p., namely $-36 \%$ from the baseline). This stems from students going to medical school, whose first year of bachelor is a difficult exam based on a numerus clausus. This example shows that satisfying students preferences is not the sole determinant of students' success, especially in highly demanding curriculum. Sports Studies, in turn, have no close equivalent in the universe of higher education programs, and can only be provided by universities. This might explain the large effects on the extensive margin.

Figure 5: Heterogeneity of the LATE by Field of Study


Law \& Economics (34 clusters - 9,560 applicants) Social Sciences (42 clusters - 10,606 applicants)


Sports Studies (96 clusters - 36,804 applicants)
Notes: The figure reports LATE estimates given by equation (2) on different educational outcomes separately for each field of study. The filled bars indicate the baseline probability, which is the mean of the outcome variable for untreated lottery losers $(Z=D=0)$ who obtained the Baccalauréat. The solid dots show the estimated treatment effects (added to the baseline probability), with $95 \%$ confidence intervals denoted by vertical capped bars. Confidence intervals are computed with robust standard errors clustered at the program $\times$ cohort level. The LATE estimate differs significantly from 0 at the $5 \%$ confidence level if the capped bar around the point estimate does not overlap with the filled bar. Socio-demographic and schooling characteristics as well as program $\times$ cohort fixed effects are added in all models. Sociodemographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions.
APB cohorts: 2013 to 2016.

## Heterogeneity Among Applicants

Table 6 displays results for estimates on the whole sample of programs, where the treatment indicator has been interacted with some students' socio-demographic and schooling characteristics. We chose to focus on four features: students' gender, scholarship status, type of high school attended and academic level (Baccalauréat achieved with highest/high honors or not). P-values for the difference of the LATE estimates between the modalities of the interacted characteristic are displayed below the treatment effects.

Treatment effects are significantly different from zero for all sub-populations and all outcomes, except successive enrollment for high-achieving students, and enrollment in third year of bachelor for high-achieving students and vocational high school graduates. Once again, if treatment effects are of similar order of magnitude, they can substantially differ in relation to the baseline probability of success of each sub-population.

Significant differences between modalities are observed regarding enrollment in higher education. On average, male students are more affected than female students (9 p.p. against 6 p.p.), with respectively a $12 \%$ and a $8 \%$ increase from a baseline probability of $77 \%$ and $73 \%$. Students without a scholarship are more affected on the extensive margin: they are more likely to ever been enrolled in higher education by 8p.p., whereas the effect is not significant for students benefiting from a need-based scholarship. This could be explained by a higher entry cost for students with no scholarship, or by the fact that these students have more outside options. Students' benefiting from a need-based scholarship in high school are also eligible for a scholarship in higher education if they enroll in specific public or contract-based programs, which are mostly available through APB. This could deter these students with a potential scholarship in higher studies to enroll in a more expensive program that is outside the scope of APB programs. Vocational high school graduates are twice more affected than general graduates by the result of the matching procedure: on average, vocational graduates who obtained their first choice are more likely to be enrolled in higher education by 12 p.p. (against 6 p.p. for general graduates) which represent a $20 \%$ increase from a baseline probability $59 \%$ (respectively $7 \%$ from a baseline of $81 \%$ ). Interestingly, the treatment effects for high-achieving students do not differ significantly from the coefficients for low-achieving students, which suggests that capacity constraints do not have a differentiated effect along students' ability on the extensive margin.

Results are less straightforward regarding outcomes of persistence and completion in higher education. Most differences are not significant, and the heterogeneity of the treatment is difficult to interpret. Nonetheless, there are a few points to notice. First, treatments effects are larger for general high school graduates than vocational ones regarding enrollment every year in the same field of study (14 p.p. against 8 p.p.). However, in relative terms, vocational graduates are clearly more affected by their admission status
to their first choice, as the treatment effects represent a $88 \%$ increase from a baseline of $9 \%$ (against $41 \%$ from a baseline of $34 \%$ for general graduates). Eventually, admission to one's first choice seems to have no effect on completion on time for vocational graduates, but still does for general graduates (8p.p.). Treatment effects are also larger for low-achieving students regarding persistence in higher education: on average, they are more likely to enroll in the same field of study every year by 13 p.p., which represents a $52 \%$ increase from a baseline probability of $25 \%$. Rejection to one's first choice appears to exert a discouragement effect on more vulnerable students.

All in all, this analysis of the heterogeneity of the treatment suggests that students at the margin of pursuing higher education are the most affected by the outcome of the admission procedure. Being admitted to one's first choice has a larger impact on higher educational trajectories for low-achieving students and/or from vocational high schools, who are the individuals with the lowest probability to access and succeed in higher education in the first place.

Table 6: Treatment Heterogeneity by Selected Applicants' Characteristics

|  | Gender |  | Scholarship |  | Type of Baccalauréat |  | Baccalauréat Distinctions |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Male | Female | Yes | No | General | Vocational | Highest/High | Else |
| Treatment effects 3 years after admission <br> 1. At least one subscription |  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |
| Baseline probability | 0.77 | 0.73 | 0.79 | 0.74 | 0.81 | 0.59 | 0.78 | 0.74 |
| Treatment effect | $\begin{gathered} 0.09^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.04 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.12^{* * *} \\ (0.02) \end{gathered}$ | $\begin{aligned} & 0.06^{* *} \\ & (0.02) \end{aligned}$ | $\begin{gathered} 0.08^{* * *} \\ (0.01) \end{gathered}$ |
| P-value of the difference |  |  |  |  |  | 0.03 | 0.5 |  |
| 2. Enrolled every year |  |  |  |  |  |  |  |  |
| Baseline probability | 0.41 | 0.40 | 0.37 | 0.41 | 0.51 | 0.15 | 0.57 | 0.38 |
| Treatment effect | $\begin{gathered} 0.09^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.07^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.01) \end{gathered}$ | $\begin{aligned} & 0.06^{* *} \\ & (0.02) \end{aligned}$ | $\begin{gathered} 0.05 \\ (0.03) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.01) \end{gathered}$ |
| P-value of the difference |  |  |  |  |  | 0.31 | 0.4 |  |
| 3. Enrolled every year in the same field of study |  |  |  |  |  |  |  |  |
| Baseline probability | 0.26 | 0.27 | 0.23 | 0.28 | 0.34 | 0.09 | 0.36 | 0.25 |
| Treatment effect | $\begin{gathered} 0.13^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.12^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.13^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.12^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.14^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.02) \end{gathered}$ | $\begin{aligned} & 0.06^{*} \\ & (0.03) \end{aligned}$ | $\begin{gathered} 0.13^{* * *} \\ (0.01) \end{gathered}$ |
| P-value of the difference |  |  | 0. |  |  | 0.01 | 0.0 |  |
| 4. Enrolled in $3^{\text {rd }}$ year of bachelor |  |  |  |  |  |  |  |  |
| Baseline probability | 0.20 | 0.24 | 0.15 | 0.24 | 0.30 | 0.05 | 0.42 | 0.20 |
| Treatment effect | $\begin{gathered} 0.06^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.08^{* * *} \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.01 \\ (0.02) \end{gathered}$ | $\begin{gathered} 0.06 \\ (0.03) \end{gathered}$ | $\begin{gathered} 0.06^{* * *} \\ (0.01) \end{gathered}$ |
| P-value of the difference | 0.85 |  | 0.35 |  | $<0.01$ |  | $>0.99$ |  |
| Controls |  |  |  |  |  |  |  |  |
| Socio-demographic characteristics | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  |
| Schooling characteristics | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  |
| Program $\times$ cohort fixed effects | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  | $\checkmark$ |  |
| Nb. of applicants | 65,333 |  | 65,333 |  | 65,333 |  | 65,333 |  |
| Nb . of clusters (program $\times$ cohort) | 197 |  | 197 |  | 197 |  | 197 |  |

Notes: $\quad{ }^{*} \mathrm{p}<0.05 ;{ }^{* *} \mathrm{p}<0.01 ;{ }^{* * *} \mathrm{p}<0.001$; standard errors clustered at the program $\times$ cohort level in parenthesis. Each cell reports the LATE estimated coefficient of the dummy of admission to top-ranked program (instrumented by the treatment assignment) interacted with the characteristic mentioned above. One regression model is estimated for each characteristic, i.e the treatment dummy is only interacted with one variable. Socio-demographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions. The baseline probability is the mean of the outcome variable for untreated lottery losers $(Z=D=0)$ who obtained the Baccalauréat. APB cohorts: 2013 to 2016.

## Heterogeneity by Potential Admission Status for Other Applications

Though the definition of our treatment is clear, our strategy of identification does not allow us to disentangle between different counterfactual status. Not being accepted to one's first choice encompasses different situations, since applicants who finished the admission procedure with no offer are, for instance, mixed up with applicants eventually admitted to their second-ranked program. Yet, comparing treated applicants with a subset of applicants from the control group violates the causal chain between the treatment and our outcomes of interest. For instance, by considering only untreated applicants that are admitted to a program at the end of the procedure, we ignore the fact that these students represent a particular subset of our control group. These applicants: (i) ranked more than
one program in their ROL; (ii) were ranked high enough to receive an offer in at least one of these other programs and (iii) purposely accepted an offer coming from a second-best program.

If we cannot know the admission status of treated applicants if they had not been treated, it is still possible to take advantage of some predetermined characteristics to imagine what could have been the counterfactual outcome of the admission procedure. To understand to what extent the effect of the treatment might vary depending on this counterfactual, we split our sample along two variables, which are the type of program ranked in second place in the ROL, and the existence of plausible alternatives in the ROL. More specifically, the first variable is a dummy that equals to one if the second-ranked program belongs to the same field of study than the top-ranked program. The second variable is a dummy equal to one if the student was ranked below the first-round admission cutoff for at least one program (other than the top-ranked) in her ROL. These variables are defined before the treatment assignment, and give an insight on the extent to which students had a relevant alternative in their ROL. Details about their construction and their distribution are given in appendix C.

Figure 6 displays the LATE estimates on subsets of priority applicants based on the type of field of their second-ranked program and the existence of a plausible alternative in their ROL. For each subset of priority applicants and each higher education outcome, the effect of the treatment is significant at the $5 \%$ level. This suggests that, regardless of the potential counterfactual admission status, admission to the top-ranked program influences students' trajectories in higher education.
Treatment effects are similar between applicants whose second choice is in the same field of study as their first choice, and those who applied to another field of study. Nevertheless, baseline probabilities substantially differ between the two groups. Untreated applicants whose second choice was in the same field of study as their first choice are more likely to enroll in higher education within three years ( $79 \%$ against $71 \%$ ), to register every year in higher education ( $46 \%$ against $32 \%$ ), to keep on studying in the same field (32 \% against $19 \%$ ) and to enroll in third year of bachelor three year after graduation (27 \% against $16 \%$ ). This result suggests that admission to the top-ranked program might have a stable effect per se, this effect being relatively stronger for applicants who have a very different second-best option.

Comparisons between applicants with at least one or no plausible alternative lead to the same conclusion. Baseline probabilities are lower for applicants with no plausible alternative, and this is true for all higher education outcomes. Applicants with no alternative are less likely to be enrolled in higher education within three years (59 \% against $80 \%$ ), to register every year ( $27 \%$ against $45 \%$ ), to stay in the same field ( $17 \%$ against
$30 \%$ ) and to be enrolled in third year of bachelor within three years (14 \% against $25 \%$ ). Once again, LATE point estimates are similar for the two sub-samples, except for enrollment in higher education within three years, for which the effect of the treatment is larger for applicants with no alternative. Being accepted to one's first choice increases the probability of ever been enrolled in higher education by 15 p.p., this effect being much lower for applicants with plausible alternative programs ( +5 p.p.).

Figure 6: LATE Estimates by Potential Counterfactual Admission Status


Notes: The figure reports LATE estimates given by equation (2) on different educational outcomes for sub-set of priority applicants based on two variables: the field of the second-ranked program and the existence of plausible alternative program. The filled bars indicate the baseline probability, which is for each group the mean of the outcome variable for untreated lottery losers ( $Z=D=0$ ) who obtained the Baccalauréat. The solid dots show the estimated treatment effects (added to the baseline probability), with $95 \%$ confidence intervals denoted by vertical capped bars. Confidence intervals are computed with robust standard errors clustered at the program $\times$ cohort level. The LATE estimate differs significantly from 0 at the $5 \%$ confidence level if the capped bar around the point estimate does not overlap with the filled bar. Socio-demographic and schooling characteristics as well as program $\times$ cohort fixed effects are added in all models. Socio-demographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions.
APB cohorts: 2013 to 2016.

## 5 Conclusion

Our paper estimates the impact of satisfying students' preferences under capacity constraints on academic achievements. It concludes that limiting access to highly demanded programs can entail a substantial cost for students at the margin of pursuing higher education, as programs might be imperfect substitutes. Using tie-breaking lotteries as an instrument for final admission to French oversubscribed university programs, we show that being admitted to one's top-ranked program has a sizable impact on students' trajectory in higher education. Our results suggest that allocating students to their stated preferred program is an efficient way not only to increase students' probability to enroll in higher education, but also for them to study without interruption, repetition or change of field of study.

Our treatment effects have been estimated on a rather comprehensive set of students and university programs. However, the extrapolation of our results to a larger set of applicants might be questioned. The correspondence between stated preferences in applicants' ROL and true preferences is far from trivial in the framework of APB, and further investigation about the interaction between the DA for selective programs and the altered Boston for non-selective programs would be needed to clarify this issue. Nonetheless, this does not threaten the internal validity of our IV strategy. Applicants are randomly assigned to the treatment, we are therefore confident that lottery losers and winners do not differ on the way they conceived their ROL.

This empirical evidence of the impact of capacity constraints on students' achievement could have several policy implications in higher education. In the absence of openings of new university seats, providing information about programs' capacity constraints and helping students to find relevant alternatives could be a solution to circumvent students' drop out, in a context where success rates in bachelor in France are quite low: only $28 \%$ of students obtain their bachelor within three years spent at the university after high school, and $12 \%$ during their fourth year (DEPP and SD-SIES, 2019). Better accommodating students' preferences seems then a cost-effective way to limit substantial expenses, as it could mitigate changes of field of study, repetition and drop outs. For instance, Fack and Grenet (2015) show that providing 1,500 euros cash allowances to a comparable population increases students' enrollment rates in higher education by 5 to $7 \%$.

Our analysis of the heterogeneity of the treatment highlights two main points. First, all students are affected by their admission status to their first choice, regardless of their gender, their prior academic achievement, their type of Baccalauréat and their socioeconomic background. Second, more vulnerable high school students benefit the most from having access to their preferred program. Individuals at the margin of pursuing higher education are the most affected by the outcome of the admission procedure, as larger effects are observed for low-achieving students, students coming from less favored socio-economic background, and students from vocational high schools. These individuals
have the lowest baseline probability to access and succeed in higher education to start with. For instance, students coming from lower social background have a probability to achieve their bachelor within three or four years of $27 \%$, against $48 \%$ for students from upper social background ( DEPP and SD-SIES, 2019). These differences interact with program's curriculum peculiarities, with stronger effects for programs with a larger share of vulnerable students or fewer alternatives in higher education. Better accommodating stated preferences of vulnerable students would help to mitigate this discouraging effect on achievements in higher education.

Eventually, our attempts to qualify the counterfactual situation in case of non-admission to the top-ranked program suggest that, if all applicants significantly benefit from obtaining their first choice, students with no plausible alternative program in their ROL are those who gain the most. The estimated effects are also stronger for students whose second choice is in a different field of study as their first choice. Help in building consistent ROL, especially by making salient students' chances of admission and available alternative programs could level the playing field among students. These are relatively free of charges measures that could have a sizable effect on mismatch at the university, and so mitigate drops out and changes of field of study.

In 2018, the ORE reform introduced a new system of admission to higher education programs called Parcoursup. Under this system, universities are allowed to differentiate applicants based on their grades. Among oversubscribed programs, high-achieving applicants are now more likely to be admitted. Solving the capacity constraints dilemma by selecting students on their academic performances is expected to have positive consequences on the quality of the student-program match, assuming that grades are a good proxy for students' ability to succeed in higher education. However, in light of our results, this policy change might have a detrimental effect on the post-secondary trajectory of low-achieving students. This population is the most responsive to the outcome of the admission procedure and might be especially hurt by the rejection from their preferred program. This negative effect could be counterbalanced by other features of the reform such as the generalization of admission quotas for students benefiting from a scholarship. All these questions should be investigated to understand better the functioning of this new system.

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## Appendix

## A Data Appendix

Data Sources There are a few caveats to the APB databases, the main one being that the APB procedure is not fully comprehensive. Some private programs are not included in the platform, though they might represent a relevant outside option for some students. Consequently, this database does not allow us to distinguish between applicants who resigned from APB because they chose an external private program from those who actually stopped studying. This issue is partly resolved later by exploiting our second source of data, which enables us to track higher education students throughout the years on a broader set of programs. APB files are not databases that are meant to be directly used by researchers. Their primary use is to be inputs for the matching algorithm, therefore there is no comprehensive documentation to precisely identify each element of the databases. An important part of our work was thus to retrieve the structure of the files and to reconcile every element to understand how the matching procedure worked.

The SISE databases have a large coverage, but do not include every higher education program in France. In particular, students registered in some specific programs such as paramedical and social programs are absent from this source ${ }^{18}$. Moreover, students who go abroad to pursue higher education can only be retrieved if they are still registered in a French institution. It is the case for students traveling with the Erasmus program. Hence we do not know if students missing from the SISE databases are either studying in programs that are out of coverage or actually dropped out from higher education. We partly treat this issue by following students over multiple years. In our sample, about eight of ten students have enrolled in SISE higher education programs within three years following university admission.

Table 7 summarizes characteristics of APB applicants and programs for the years 2013 to 2017. For consistency reasons, we focus on students who successfully registered in the regular procedure of APB. The first element to notice is the increasing trend in the number of potential first entrants into higher education. The number of active applicants who obtained their Baccalauréat substantially increased over the period, from around 499,000 in 2013 to 569,000 in 2017. In the meantime, the estimated overall capacity of programs increased from around 416,000 to 477,000 . This discrepancy between the number of applicants and programs' size suggests that capacity constraints were increasingly binding within the APB system. However, this imbalance does not take into account voluntary resignations, neither additional seats that were filled later on during the supplementary round of the procedure. Eventually, a sizable share of applicants did not enroll in higher

[^12]Table 7: General Statistics - Applicants, Programs and Applications

|  | 2013 | 2014 | 2015 | 2016 | 2017 |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Applicants (k) |  |  |  |  |  |
| High school students | 549 | 572 | 582 | 605 | 626 |
| Baccalauréat obtained | 499 | 523 | 532 | 555 | 569 |
| All applicants | 710 | 763 | 788 | 814 | 843 |
| Capacity |  |  |  |  |  |
| Seats filled (k) |  |  |  |  |  |
| University programs | 183 | 195 | 216 | 230 | 238 |
| All programs | 416 | 434 | 461 | 480 | 477 |
| Nb. of programs |  |  |  |  |  |
| University programs | 2,068 | 2,201 | 2,242 | 2,060 | 2,036 |
| All programs | 9,705 | 10,092 | 10,435 | 10,428 | 10,005 |
| Applications in the ROL |  |  |  |  |  |
| Average number |  |  |  |  |  |
| University programs [s.d] | 1.7 [2.3] | 1.9 [2.5] | 2.1 [2.5] | 2.9 [3.7] | 3.2 [3.8] |
| All programs [s.d] | 5.9 [5.3] | 6.1 [5.6] | 6.2 [5.6] | 7.1 [6.1] | 7.4 [6.3] |
| Top-ranked program |  |  |  |  |  |
| University program | 0.31 | 0.33 | 0.34 | 0.35 | 0.39 |

Source: APB administrative data from 2013 to 2017, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Applicants successfully registered in the regular procedure of APB.
Notes: The number of seats filled is proxied by the number of applicants who were admitted through the regular procedure of APB. It does not take into account the seats filled during the complementary round of APB.
education, or chose a program outside the APB platform.
University programs became increasingly attractive over the period. Their share in students' ROL went up from $28 \%$ in 2013 to $43 \%$ in 2017. More importantly, university programs also represent an increasing proportion of students' top-ranked program: they accounted for $39 \%$ of students' first choice in 2017, against $31 \%$ in 2013. Part of the explanation for this trend is the change in the rules of APB: during the period, students had to rank at least one university program for their ROL to be valid. Meanwhile, the number of seats filled in university programs throughout the three regular rounds of APB has slowly moved from around 183,000 in 2013 to 238,000 in 2017. This growing pressure on university programs has mechanically increased the risk of seeing more oversubscribed programs over the years.

Dynamics of Admissions Table 8 displays students' final admission status and programs' characteristics by treatment assignment status. This table displays additional elements
on the discrepancy between top-ranked programs and the programs students are finally assigned to. Among lottery losers admitted to a program at the end of the procedure, half were admitted to their top-ranked program, $21 \%$ to their second choice, and the rest $(29 \%)$ to a program ranked below in their ROL. Eventually, $71 \%$ of lottery losers who are finally admitted to a program through APB end up in their top-ranked field of study. More than $65 \%$ of lottery losers were registered in SISE programs the year following the APB procedure. This share is higher than the one of students admitted through APB ( $58 \%$ ) since SISE data-sets recover some programs that are not included in the platform APB. Conversely, $72 \%$ of lottery winners are found in higher education a year after the APB procedure, though $75 \%$ of them were admitted to their first choice. This means that a sizable number of lottery winners did not register in their first choice after the summer holidays.

Table 8: Final Admission Status, by Treatment Assignment

|  | Treatment assignment |  |
| :--- | :---: | :---: |
|  | Lottery losers <br> $(Z=0)$ | Lottery winners <br> $(Z=1)$ |
| Count | 43,246 | 58,826 |
| Admission status (share of all applicants) |  |  |
| Admission | $58 \%$ | $75 \%$ |
| No offer | $11 \%$ | $0 \%$ |
| Failed the Baccalauréat | $7 \%$ | $11 \%$ |
| Resignation | $24 \%$ | $14 \%$ |
| Program of admission (share of admitted applicants) |  |  |
| Rank of admission | $50 \%$ | $100 \%$ |
| Top-ranked (T=1) | $21 \%$ | $0 \%$ |
| Ranked 2 | $29 \%$ | $0 \%$ |
| Ranked 3 or below | $94 \%$ | $100 \%$ |
| Other features | $71 \%$ | $100 \%$ |
| University programs |  |  |
| Preferred field of study | $114 \%$ | $96 \%$ |
| Registration in higher studies |  |  |
| Enrolment in $t+1$ |  |  |

Source: APB administrative data from 2013 to 2017 and SISE data from 2013 to 2018, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Oversubscribed university programs with at least 30 lottery winners and 30 lottery losers.

Table 9: Type of Oversubscribed Non-Selective Programs by Field of Study

| Field | Type of Programs | Oversubscribed Programs |  |
| :---: | :---: | :---: | :---: |
|  |  | Number | Priority Candidates |
| Law \& Economics | Economic © Social Administration | 5 | 813 |
|  | Law | 17 | 5,907 |
|  | Economics | 2 | 334 |
|  | Economics \& Management | 4 | 1,229 |
|  | Management | 11 | 3,740 |
|  | Political Sciences | 8 | 1,172 |
|  | Health $\mathcal{E}$ Social Sciences | 6 | 2,753 |
|  | Total | 53 | 15,948 |
| Humanities | Arts | 6 | 792 |
|  | Performing Arts | 4 | 578 |
|  | Plastic Arts | 1 | 292 |
|  | Cinema | 3 | 537 |
|  | Applied Foreign Languages | 4 | 558 |
|  | Foreign languages, Literature \& Civilizations | 7 | 1,106 |
|  | Total | 25 | 3,863 |
| Health \& Sciences | Mathematics | 3 | 1,275 |
|  | Medicine, Pharmacy, Dentistry, Midwifery | 12 | 5,485 |
|  | Physics | 1 | 212 |
|  | Earth $\mathcal{F}$ Life Sciences | 6 | 2,528 |
|  | Engineering | 1 | 100 |
|  | Total | 23 | 9,600 |
| Social Sciences | Geography | 1 | 67 |
|  | History | 1 | 128 |
|  | Information \& Communication | 8 | 1193 |
|  | Mathematics Applied to Social Sciences | 3 | 480 |
|  | Psychology | 27 | 9,751 |
|  | Sciences of Education | 23 | 4,235 |
|  | Language Sciences | 2 | 1,208 |
|  | Sociology | 3 | 661 |
|  | Total | 68 | 17,723 |
| Sports | Science $\mathcal{E}^{\text {T }}$ Technique of Physical 83 Sports Activities | 138 | 54,938 |
|  | Total | 138 | 54,938 |
| All Fields |  | 307 | 102,073 |

## B Robustness Checks

Table 10: Balancing Checks of Pre-treatment Characteristics for Alternative Specifications

|  | Model Specification |  |  |
| :---: | :---: | :---: | :---: |
|  | Linear | Logit | Probit |
| Average Marginal Effects |  |  |  |
| Gender |  |  |  |
| Male | ref. | ref. | ref. |
| Female | $-0.008^{*}(0.004)$ | $-0.008^{*}(0.004)$ | $-0.008^{*}(0.004)$ |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| $\overline{S E S}$ |  |  |  |
| Very High | ref. | ref. | ref. |
| Medium-High | -0.002 (0.005) | -0.002 (0.005) | -0.002 (0.005) |
| Medium-Low | -0.006 (0.004) | -0.006 (0.004) | -0.006 (0.004) |
| Low | 0.005 (0.005) | 0.005 (0.005) | 0.005 (0.005) |
| No information | 0.022 (0.015) | 0.022 (0.015) | 0.023 (0.015) |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| Scholarship beneficiary |  |  |  |
| No Scholarship | ref. | ref. | ref. |
| Scholarship | 0.002 (0.005) | 0.002 (0.005) | 0.002 (0.005) |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| Baccalauréat Distinctions |  |  |  |
| Highest honors | ref. | ref. | ref. |
| High honors | 0.011 (0.012) | 0.012 (0.012) | 0.010 (0.012) |
| Honors | 0.007 (0.011) | 0.007 (0.012) | 0.007 (0.012) |
| No distinctions | 0.002 (0.011) | 0.003 (0.011) | 0.002 (0.011) |
| Failure | 0.000 (0.012) | 0.000 (0.012) | 0.000 (0.012) |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| Type of Baccalauréat |  |  |  |
| General | ref. | ref. | ref. |
| Vocational | -0.001 (0.004) | -0.001 (0.004) | -0.001 (0.004) |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| Age |  |  |  |
| On time | ref. | ref. | ref. |
| In advance | -0.002 (0.011) | -0.002 (0.010) | -0.002 (0.011) |
| Late | 0.004 (0.004) | 0.004 (0.004) | 0.003 (0.004) |
| Quality of Fit | $R^{2}=0.18$ | $\log L / N=-0.58$ | $\log L / N=-0.58$ |
| Controls |  |  |  |
| Program $\times$ cohort fixed effects | $\checkmark$ | $\checkmark$ | $\checkmark$ |
| Nb of applicants | 65,333 | 65,333 | 65,333 |
| Nb. of clusters (program $\times$ cohort) | 197 | 197 | 197 |

Notes: $\quad{ }^{*} \mathrm{p}<0.05 ;{ }^{* *} \mathrm{p}<0.01 ;{ }^{* * *} \mathrm{p}<0.001$; standard errors clustered at the program $\times$ cohort level in parenthesis. For each characteristic, the table reports coefficients estimated in separated models where assignment to treatment is the dependent variable. All models include fixed effects for program $\times$ cohort, our randomization unit.
APB cohorts: 2013 to 2016.

Table 11: LATE estimates including or excluding applicants who failed the Baccalauréat

$$
\text { All applicants } \quad \text { Baccalauréat obtained }
$$

## Treatment effects 3 years after admission

1. At least one registration

| Treatment effect | $0.08^{* * *}$ | $0.08^{* * *}$ |
| :--- | :---: | :---: |
| 2. Registered every year | $(0.01)$ | $(0.01)$ |
| Treatment effect | $0.08^{* * *}$ |  |
|  | $(0.01)$ | $0.08^{* * *}$ |
|  | $(0.01)$ |  |

3. Registered every year in the same field of study

| Treatment effect | $\begin{aligned} & 0.12^{* * *} \\ & (0.01) \end{aligned}$ | $\begin{aligned} & 0.13^{* * *} \\ & (0.01) \end{aligned}$ |
| :---: | :---: | :---: |
| 4. Enrolled in $3^{\text {rd }}$ year of bachelor |  |  |
| Treatment effect | $\begin{gathered} 0.066^{* * *} \\ (0.01) \end{gathered}$ | $\begin{gathered} 0.066^{* * *} \\ (0.01) \end{gathered}$ |
| Controls |  |  |
| Sociodemographics | $\checkmark$ | $\checkmark$ |
| Schooling chr. | $\checkmark$ | $\checkmark$ |
| Program*cohort fixed effects | $\checkmark$ | $\checkmark$ |
| Nb. of applicants | 65,333 | 56,908 |
| Nb . of clusters (program $\times$ cohort) | 197 | 197 |

Notes: $\quad{ }^{*} \mathrm{p}<0.05 ;{ }^{* *} \mathrm{p}<0.01 ;{ }^{* * *} \mathrm{p}<0.001$; standard errors clustered at the program $\times$ cohort level in parenthesis. Socio-demographic characteristics include gender, age, SES and scholarship status. Schooling characteristics include the type of Baccalauréat (general or vocational) as well as Baccalauréat distinctions. APB cohorts: 2013 to 2016.

## C Heterogeneity of the Treatment: Details on Students' ROL Characteristics

Second-Ranked Program Second choices are an intuitive way to search for counterfactual; they suggest what kind of programs students are willing to go to if they are rejected from their first choice. This, by definition, their second best. Random assignment of the treatment allows us to have balanced samples with respect to the nature of the secondranked program. We therefore split our sample into three subsets according to the field of study of the second-ranked program. More precisely, we distinguish applicants who ranked in second position a program in the same field of study as their first choice from applicants who applied to another field of study. We also distinguish the few applicants who only applied to a single program only.

Table 12 compares the admission status of priority applicants depending on second choices' characteristics. Though having roughly the same probability of being admitted somewhere at the end of the procedure, unassigned students who ranked second a program in the same field of study as their top-ranked program (two thirds of the sample) are more likely to be admitted to this field ( $47 \%$ ) compared to unassigned students who ranked a different field of study ( $28 \%$ ). Conversely, those whose second choice is in another field of study are more likely to be admitted to their second-ranked program ( $20 \%$ against $10 \%{ }^{19}$. The few priority applicants whose ROL consists of a single program have no chance to get an offer from another program than the one they ranked first in their ROL. These applicants won't receive any offer from other programs through the three rounds of APB, meaning that the counterfactual situation for these applicants is undoubtedly non-admission. Only one quarter of untreated students with only one program ranked in their ROL are finally admitted.

Existence of Plausible Alternatives We do not know in advance students' actual chances to be admitted in another program in case of rejection from their first choice. These probabilities depend students' position in programs' ordered list of applicants with respect to their capacity. We create a dummy variable that allows to approximate applicants' ex-ante chances to be admitted in another program. For all the lower-ranked programs within priority applicants' ROL, we look in retrospect at the rank of the last applicant who actually received an offer in the first round of admission. If a priority applicant has in her list at least one lower-ranked program where she was ranked below the last applicant who received an offer in the first admission round, she would receive an offer from this program in case of non-assignment to the treatment. We say that these applicants have

[^13]at least one plausible alternative program. Conversely, priority applicants with a rank that is always too high to receive an offer from the other programs in their ROL have no plausible alternative. They may be admitted to a program by the end of the three rounds of APB, but won't receive any offer in the first round if they are randomly drawn by the top-ranked program.

Table 12 compares the admission status of priority applicants depending on the existence of a plausible alternative. Two thirds of applicants have a plausible alternative, regardless of their treatment assignment status. These applicants are more likely to be admitted somewhere at the end of the procedure ( $66 \%$, against $42 \%$ for those with no plausible alternatives). Naturally, applicants with a plausible alternative have also more chances to be admitted to their second-ranked program ( $17 \%$ against $4 \%$ ) or their preferred field of study ( $44 \%$ against $34 \%$ ).

Table 12: Admission Status by Treatment Assignment and ROL Characteristics

|  | Treatment assignment |  |
| :---: | :---: | :---: |
|  | Lottery losers $(Z=0)$ | Lottery winners $(Z=1)$ |
| All priority applicants | 43,246 | 58,826 |
| Second-ranked program in the ROL |  |  |
| Preferred field of study |  |  |
| Count | 29,337 | 38,336 |
| Admitted somewhere | 60\% | 78\% |
| Top-ranked program | 30\% | 78\% |
| Program ranked 2 in the ROL | 10\% | 0\% |
| Preferred field of study | 47\% | 78\% |
| Other field of study |  |  |
| Count | 11,447 | 15,483 |
| Admitted somewhere | $57 \%$ | $74 \%$ |
| Top-ranked program | 25\% | 74\% |
| Program ranked 2 in the ROL | 20\% | 0\% |
| Preferred field of study | 28\% | 74\% |
| Only one program ranked |  |  |
| Count | 2,462 | 5,007 |
| Admitted somewhere | 25\% | 58\% |
| Top-ranked program | 25\% | 58\% |
| Program ranked 2 in the ROL | 0\% | 0\% |
| Preferred field of study | 25\% | 58\% |
| Existence of plausible alternatives |  |  |
| At least one plausible alternative |  |  |
| Count | 28,042 | 37,682 |
| Admitted somewhere | 66\% | 80\% |
| Top-ranked program | 28\% | 80\% |
| Program ranked 2 in the ROL | 17\% | 0\% |
| Preferred field of study | 44\% | 80\% |
| No plausible alternative |  |  |
| Count | 15,204 | 21,144 |
| Admitted somewhere | 42\% | 68\% |
| Top-ranked program | 30\% | 68\% |
| Program ranked 2 in the ROL | 4\% | 0\% |
| Preferred field of study | $34 \%$ | 68\% |

Source: APB administrative data from 2013 to 2017, from the French statistical service of the ministry of higher education (MESRI-SIES).
Scope: Oversubscribed university programs with at least 30 lottery winners and 30 lottery losers.
Notes: A program is a plausible alternative for an applicant if her rank is below the rank of the last applicant who received an offer from that program in the first admission round.

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[^0]:    * Insee-Dese - Département des études économiques - Division «Redistribution et Politiques Sociales »
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[^1]:    ${ }^{1}$ In France, the state subsidizes each opening of a seat in an oversubscribed university program up to 1600 euros (Cours des Comptes, 2020).

[^2]:    ${ }^{2}$ These results are estimated on the universe of students applying for need-based grants in French higher education over the period 2008-2010 with data on all students enrolled in French public universities.

[^3]:    ${ }^{3}$ Law 2018-166, published the 8th of March 2018.
    ${ }^{4}$ Section L. 612-3 of the former Education code valid until the 10th of March 2018.

[^4]:    ${ }^{5}$ Students who were already enrolled in a first year of undergraduate studies and who wanted to switch field could also apply to another program through APB. In this paper, we focus on high school students who are potential new entrants into higher education.

[^5]:    ${ }^{6}$ For this set of applicants, the matching derived by the DA algorithm is equivalent to the one resulting from the Boston Mechanism (Abdulkadiroğlu et al. (2017b)).

[^6]:    ${ }^{7}$ Bourse sur critères sociaux, a national need-based scholarship program.

[^7]:    ${ }^{8}$ Hereafter, SISE programs refer to both SISE and BPBAC programs.
    ${ }^{9}$ The correspondence table between fields and type of programs at a disaggregated level is displayed in appendix A table 9

[^8]:    ${ }^{10}$ Over the period of study, the total number of Sport studies programs remained constant (about 85).
    ${ }^{11}$ This information is easily comparable between applicants from different high schools, though it should be noted that the final results of the Baccalauréat are published between the second and the third admission round. We assume that the outcome of the lottery does not affect students' achievements and believe that this hypothesis is very likely to be verified. This variable is thus considered to be a pre-treatment characteristic in this article. Using another proxy such as the continuous assessment in high school would not allow us to assess students' ability, as high schools have different strategies when it comes to grading their students.
    ${ }^{12}$ Students from vocational (lycées professionnels) and technological high-schools (lycées technologiques) belong to this group.

[^9]:    ${ }^{13}$ For the sake of the argument, we posit here the monotonicity assumption, also known as the "no defier" assumption. See section 3.2 about the validity of the strategy.

[^10]:    ${ }^{14}$ In this framework, another valid instrument is a dummy equal to 1 if the priority applicant had a lottery number Rand $_{i s c}$ that is lower than the one of the last applicant who received an offer at the end of the three rounds (Behaghel and De Chaisemartin, 2020). We do not use this instrument because in our setting almost all priority applicants would have been eventually considered as assigned to the treatment, as there is an important share of between-rounds rejections due to failures at the Baccalauréat.
    ${ }^{15}$ Even though we show in the following balancing checks section that the vector $X_{i s c}$ is almost perfectly orthogonal to our random assignment $Z_{i s c}$ in our sample of analysis, we include these characteristics in the equation to improve the accuracy of our estimates of treatment effects.

[^11]:    ${ }^{16}$ Results including cohort 2017 will be presented in future versions of this paper, as soon as the upcoming versions of SISE databases becomes available. As a robustness check, we have done the estimates of the LATE separately for each cohort from 2013 to 2016 and the coefficients (available on demand) do not qualitatively differ between cohorts. This suggests that our results are not likely to change by adding cohort 2017 in the estimates.
    ${ }^{17}$ Since the assignment rate differs from one program to another, and since all programs recruit students with different profiles, the raw balance between the pooled sample of assigned and unassigned students is not guaranteed. However, our sample appears to be fairly balanced when we do not add the program $\times$ cohort fixed effects.

[^12]:    ${ }^{18}$ Paramedical and social programs represent about $4 \%$ of students enrolled in higher education. Subsequent versions of the SISE database incorporates these programs.

[^13]:    ${ }^{19}$ This is partly due to the fact that the relative rank is part of the priority rules: the applicants who put second the same field of study have a relative rank of 2 (because they already applied to one program in that field), whereas students applying to another field wild get a relative rank of 1 that increases their chances to receive an offer.

