# Integer programming formulations for college admissions with ties 

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#### Abstract

When two students with the same score are competing for the last slot at a university programme in a central admission scheme then different policies may apply across countries. In Ireland only one of these students is admitted by a lottery. In Chile both students are admitted by slightly violating the quota of the programme. Finally, in Hungary none of them is admitted, leaving one slot empty. We describe the solution by the Hungarian policy with various integer programing formulations and test them on a real data from 2008 with around 100,000 students. The simulations show that the usage of binary cutoff-score variables is the most efficient way to solve this problem when using IP technique. We also compare the solutions obtained on this problem instance by different admission policies. Although these solutions are possible to compute efficiently with simpler methods based on the Gale-Shapley algorithm, our result becomes relevant when additional constraints are implied or more complex goals are aimed, as it happens in Hungary where at least three other special features are present: lower quotas for the programmes, common quotas and paired applications for teachers studies.


Keywords: integer programming • college admissions • stable matching.

## 1 Introduction

Gale and Shapley gave a standard model for college admissions [15], where stable matching is was the solution concept suggested. Intuitively speaking a matching is stable if the rejection of an application at a college is explained by the saturation of that college with higher ranked students. Gale and Shapley showed that a stable matching can always be found by their so-called deferred-acceptance algorithm, which runs in linear time in the number of applications, see e.g. [16]. Moreover, the student-oriented variant results in the student-optimal stable matching, which means that no student could get a better assignment in any other stable matching. The theory of stable matchings have been intensively studied since 1962 by mathematicians/computer scientists (see e.g. [16])
and economists/game theorists (see e.g. [20]). The Gale-Shapley algorithm has also been used in practice all around the world [8], first in 1952 in the US resident allocation programme, called NRMP [18], then also in school choice, e.g. in Boston [1] and New York [2]. In Hungary, the national admission scheme for secondary schools follows the original Gale-Shapley model and algorithm [9], and the higher education admission scheme also uses a heuristic based on the Gale-Shapley algorithm [10].

The Hungarian higher education admission scheme have at least four important special features: the presence of ties, the lower and common quotas, and the paired applications. Each of the latter three special features makes the problem NP-hard [11], only the case of ties is resolvable efficiently [12]. In a recent paper [4] we studied the usage of integer programming techniques for finding stable solutions with regard to each of these four special features separately, and we managed to solve the case of lower quotas for the real instance of 2008. In this follow-up work we develop and test new IP formulations for the case of ties. The ultimate goal of this line of work is to suggest a solution concept for the college admission problem where ties and common quotas are also present, together with providing integer programming formulations that are suitable to compute this solution for large scale applications, such as the Hungarian university admission scheme with over 100,000 students.

First we start by investigating the basic Gale-Shapley model and then we consider the case of ties. Due to the space limit we defer the description of IPs to the full version of the paper, here we present only the results of the simulations.

## 2 Model descriptions

In this section first we present the classical Gale-Shapley college admission problem and then the case of ties.

### 2.1 The Gale-Shapley model

In the classical college admissions problem by Gale and Shapley [15] the students are matched to colleges. ${ }^{5}$ In our paper we will refer the two sets as applicants $A=\left\{a_{1}, \ldots, a_{n}\right\}$ and colleges $C=\left\{c_{1}, \ldots c_{m}\right\}$. Let $u_{j}$ denote the upper quota of college $c_{j}$. Regarding the preferences, we assume that the applicants provide strict rankings over the colleges, where $r_{i j}$ denotes the ranking of the application $\left(a_{i}, c_{j}\right)$ in applicant $a_{i}$ 's preference list. We suppose that the students are ranked according to their scores at the colleges, so college $c_{j}$ ranks applicant $a_{i}$ according to her score $s_{i j}$, where higher score is better. Let $E \subseteq A \times C$ denote the set of applications. A matching is a set of applications, where each student is admitted to at most one college and each college has at most as many assignees

[^0]and its quota, $u_{j}$. For a matching $M$ let $M\left(a_{i}\right)$ denote the college where $a_{i}$ is admitted (or $\emptyset$ if $a_{i}$ is not allocated to any college) and let $M\left(c_{j}\right)$ denote the set of applicants admitted to $c_{j}$ in $M$. A matching $M \subset E$ is stable if for any application $\left(a_{i}, c_{j}\right)$ outside $M$ either $a_{i}$ prefers $M\left(a_{i}\right)$ to $c_{j}$ or $c_{j}$ filled its seats with $u_{j}$ applicants who all have higher scores than $a_{i}$ has. The deferred-acceptance algorithm of Gale and Shapley provides a student-optimal stable matching in linear time [15].

The notion of cutoff scores is important for both the classical Gale-Shapley model and its generalisations with ties and common quotas. Let $t_{j}$ denote the cutoff score of college $c_{j}$ and let $\mathbf{t}$ denote a set of cutoff scores. We say that matching $M$ is implied by cutoff scores $\mathbf{t}$ if every student is admitted to the most preferred college in her list, where she achieved the cutoff score. We say that a set of cutoff scores $\mathbf{t}$ corresponds to a matching $M$ if $\mathbf{t}$ implies $M$. For a matching $M$ an applicant $a_{i}$ has justified envy towards another applicant $a_{k}$ at college $c_{j}$ if $M\left(a_{k}\right)=c_{j}, a_{i}$ prefers $c_{j}$ to $M\left(a_{i}\right)$ and $a_{i}$ is ranked higher than $a_{k}$ at $c_{j}$ (i.e. $s_{i j}>s_{k j}$ ). A matching with no justified envy is called envy-free (see [22] and [21]).

It is not hard to see that a matching is envy-free if and only if it is implied by some cutoff scores [3]. Note that an envy-free matching might not be stable because of blocking with empty seats, i.e. when a student $a_{i}$ prefers $c_{j}$ to $M\left(a_{i}\right)$ and $c_{j}$ is not saturated (i.e. $\left|M\left(c_{j}\right)\right|<u_{j}$ ). In this case a matching is called wasteful. Again, by definition it follows that a matching is stable if and only if it is envy-free and non-wasteful (see also [6]). To achieve non-wastefulness we can require the cutoff of any unsaturated college to be minimum (zero in our case). Alternatively we may require that no cutoff score may be decreased without violating the quota of that college, while keeping the other cutoff scores. Furthermore, we may also satisfy the latter condition by ensuring that we select the student-optimal envy-free matching, which is the same as the student-optimal stable matching [22]. To return this solution we only need to use an appropriate objective function. We will use the above described connections when developing our IPs.

### 2.2 Case of ties

In many nationwide college admission programmes the students are ranked based on their scores, and ties may appear. In Hungary, for instance, the students can obtain integer points between 0 and 500 (the maximum was 144 until 2007), so ties do occur. When ties are present then one way to resolve this issue is to break ties by lotteries, as done in Ireland (so a lucky student with 480 point may be admitted to law studies, whilst an unlucky student with the same score may be rejected). However, the usage of lotteries can be seen unfair, so in some countries, such as Hungary [12] and Chile [17] equal treatment policies are used, meaning that students with the same score are either all accepted or all rejected. In case of such a policy, there are two reasonable variants when deciding about the last group of students without whom the quota is unfilled and with whom the quota is violated. In the restrictive policy, used in Hungary, the quotas are
never violated, so this last group of students is always rejected, whilst in Chile they use a permissive policy and they always admit this last group of students. For instance, if there are three students, $a_{1}, a_{2}$ and $a_{3}$, applying to a programme of quota 2 with scores 450,443 , and 443 , respectively then in Hungary only $a_{1}$ is admitted, whilst in Chile all three students are admitted. In Ireland, $a_{1}$ is admitted and they use a lottery to decide whether $a_{2}$ or $a_{3}$ will get the last seat.

Stable matchings for the case of ties were defined through the cutoff scores in [12]. The usage of cutoff scores in case of ties make the solution envy-free, meaning that no student $a_{i}$ may be rejected from college $c_{j}$ if this college admitted another student with score equal to or lower than the score of student $a_{i}$. This allocation concept is called also equal treatment policy, as the admission of a student to a programme implies the admission offer to all other students with the same score. A matching is envy-free for college admission problem with ties if and only if it is induced by cutoff scores [3]

For the restrictive policy used in Hungary, the stability of the matching can be defined by adding a non-wastefulness condition to envy-freeness. Namely, a matching induced by cutoff scores is stable if no college can decrease its cutoff score without violating its quota, assuming that the other cutoff scores remain the same. In the more permissive Chilean policy a matching is stable if by decreasing the cutoff score of any college there would be empty seats left there. (We note that the stability of a matching can be equivalently defined by the lack of a set of blocking applications involving one college and a set of applicants such that this set of applications would be accepted by all parties when compared to the applications of the matching considered. See more about this connection in [14].)

Biró and Kiselgof [12] showed two main theorems about stable matchings for college admissions with ties. In their first theorem they showed that a studentoptimal and a student-pessimal stable matchings exist for both the restrictive policy (Hungary) and the permissive policy (Chile), where the cutoff scores are minimal / maximal, respectively. Furthermore, they also proved the intuitive results that if we compare the student-optimal cutoff scores (or the studentpessimal ones) with respect to the three reasonable policies, namely the Hungarian (restrictive), the Irish (lottery), and the Chilean (permissive), then the Hungarian cutoff scores are always as high for each college than the Chilean cutoff scores and the Irish cutoff scores are in between these. When considering the student-optimal stable matching, it turns out to be also the student-optimal envy-free matching, as described in [3].

## 3 Simulations

In this section we present the main simulation results.

### 3.1 Gale-Shapley model

We took the 2008 data after breaking the ties randomly, by considering only the faculty quotas and keeping only the highest ranked application of each student
for every programme (i.e. the application for either the state funded or the privately funded seat). We used AMPL with Gurobi for solving the IPs.

| IP formulations | \#variables | \#constraints | \#non-0 elem. | size(Kb) | run time(s) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| SO-BB | 287,035 | 381,115 | $73,989,595$ | $1,319,663$ | 1,139 |
| SO-NW-CUT | 291,935 | 673,050 | $2,463,808$ | 69,464 | 81 |
| MIN-CUT | 289,485 | 668,150 | $2,169,423$ | 64,254 | 5,062 |
| MSMR-CUT | 289,485 | 668,150 | $2,169,423$ | 69,846 | 2,318 |
| SO-NW-BIN-CUT | 574,070 | 955,185 | $3,028,078$ | 75,810 | 107 |
| MIN-BIN-CUT | 574,070 | 952,735 | $2,738,593$ | 65,657 | 871 |
| MSMR-BIN-CUT | 574,070 | 952,735 | $2,738,593$ | 66,467 | 4,325 |
| MSMR-EF | n.a. | n.a. | n.a. | $8,667,403$ | n.a. |

### 3.2 Case of ties

We used the 2008 data with the original ties by considering again the faculty quotas and keeping only the highest ranked application of each student for every programme.

| IP formulations | \#variables | \#constraints | \#non-0 elem. | size(Mb) | run time(s) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| MIN-CUT | 289,485 | 668,150 | $2,169,423$ | 59,694 | 5,247 |
| MSMR-CUT | 289,485 | 668,150 | $2,169,423$ | 65,286 | 1,460 |
| MIN-BIN-CUT | 428,513 | 807,178 | $2,447,479$ | 53,548 | 982 |
| MSMR-BIN-CUT | 428,513 | 807,178 | $2,447,479$ | 57,106 | 1,362 |
| SO-H-NW-CUT | 578,970 | $1,694,333$ | $4,793,409$ | 114,882 | 1,310 |
| SO-H-NW-BIN-CUT | 861,105 | $1,813,840$ | $5,352,772$ | 118,828 | 165 |

Finally, we conducted the simulation on the same 2008 data, where we compared the results for the Hungarian, Irish and Chilean policies. The results indeed follow the theorems of [12] regarding the cutoff scores for the three different policies. The most interesting fact of the simulation is that for the Hungarian and Irish policies the difference between the student-optimal and student-pessimal solutions is minor, as demonstrated also in earlier paper for large markets, such as [19]. However, for the Chilean policy this difference was more significant.

| $\|$size of matching | average rank |  | average cutoffs |  | \# rejections |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| policies | A-opt. | C-opt. | A-opt. | C-opt. | A-opt. | C-opt. | A-opt. | C-opt. |
| Hungarian | 86,195 | 86,195 | 1.2979 | 1.2979 | 58.3931 | 58.3931 | 37,698 | 37,698 |
| Irish | 86,410 | 86,410 | 1.2916 | 1.2916 | 58.2090 | 58.2106 | 36,802 | 36,804 |
| Chilean | 86,650 | 86,614 | 1.2824 | 1.2844 | 57.2502 | 57.5200 | 35,668 | 35,901 |

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[^0]:    ${ }^{5}$ In the computer science literature this problem setting is typically called Hospital / Residents problem (HR), due to the National Resident Matching Program (NRMP) and other related applications.

