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# Improved approximation bounds for the Student-Project Allocation problem with preferences over projects $\stackrel{\star}{\sim}$

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# ABSTRACT

Manlove and O'Malley (2008) [8] proposed the Student-Project Allocation problem with Preferences over Projects (SPA-P). They proved that the problem of finding a maximum stable matching in SPA-P is APX-hard and gave a polynomial-time 2-approximation algorithm. In this paper, we give an improved upper bound of 1.5 and a lower bound of 21/19 (> 1.1052).

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#### 1. Introduction

Assignment problems based on the preferences of participants, which originated from the famous *Hospitals/Residents problem (HR)* [3], are important almost everywhere, such as in education systems where students must be allocated to elementary schools or university students to projects. In the university case, each student may have preferences over certain research projects supervised by professors and usually there is an upper bound on the number of students each project can accept. Our basic goal is to find a "stable" allocation where no students (or projects or professors if they also have preferences over students) can complain of unfairness. This notion of stability was first introduced by Gale and Shapley in the context of the famous *Stable Marriage problem* in 1962 [2].

The *Student-Project Allocation problem (SPA)* is a typical formulation of this kind of problem originally described by Abraham, Irving, and Manlove [1]. The participants here are *students, projects*, and *lecturers*. Each project is offered by a single lecturer, though one lecturer may offer multiple projects. Each project and each lecturer has a *capacity*. Students have preferences over projects, and lecturers have preferences over students. Our goal is to find a stable matching between students and projects satisfying all of the capacity constraints for projects and lecturers. They proved that all stable matchings for a single instance have the same size, and proposed linear-time algorithms to find one [1].

Manlove and O'Malley [8] proposed a variant of SPA, called *SPA with Preferences over Projects* (*SPA-P*), where lecturers have preferences over projects they offer rather than preferences over students. In contrast to SPA, they pointed out that the sizes of stable matchings may differ, and proved that the problem of finding a maximum stable matching in SPA-P, denoted

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*MAX-SPA-P*, is APX-hard. They also presented a polynomial-time 2-approximation algorithm. Specifically, they provided a polynomial-time algorithm that finds a stable matching, and proved that any two stable matchings differ in size by at most a factor of two.

#### 1.1. Our contributions

In this paper, we improve both the upper and lower bounds on the approximation ratio for MAX-SPA-P. We give an upper bound of 1.5 and a lower bound of 21/19 (> 1.1052) (under the condition that  $P \neq NP$ ). For the upper bound, we modify Manlove and O'Malley's algorithm SPA-P-APPROX [8] using Király's idea [7] for the approximation algorithm to find a maximum stable matching in a variant of the stable marriage problem (*MAX-SMTI*). We also show that our analysis is tight. For the lower bound, we give a gap-preserving reduction from (a variant of) MAX-SMTI. Our reduction also gives a lower bound of 1.25 under the Unique Games Conjecture.

### 2. Preliminaries

Here we give a formal definition of SPA-P and MAX-SPA-P, derived directly from the literature [8]. An instance *I* of SPA-P consists of a set *S* of *students*, a set *P* of *projects*, and a set *L* of *lecturers*. Each lecturer  $\ell_k \in L$  offers a subset  $P_k$  of projects. Each project is offered by exactly one lecturer, i.e.,  $P_{k_1} \cap P_{k_2} = \emptyset$  if  $k_1 \neq k_2$ . Each student  $s_i \in S$  has an *acceptable* set of projects, denoted  $A_i$ , and has a strict order on  $A_i$  according to preferences. Each lecturer  $\ell_k$  also has a strict order on  $P_k$  according to preferences. Also, each project  $p_j$  and each lecturer  $\ell_k$  has a positive integer, called a *capacity*, denoted  $c_j$  and  $d_k$ , respectively.

An assignment *M* is a subset of  $S \times P$  where  $(s_i, p_j) \in M$  implies  $p_j \in A_i$ . Let  $(s_i, p_j) \in M$  and  $\ell_k$  be the lecturer who offers  $p_j$ . Then we say that  $s_i$  is assigned to  $p_j$  in *M*, and  $p_j$  is assigned  $s_i$  in *M*. We also say that  $s_i$  is assigned to  $\ell_k$  in *M* and  $\ell_k$  is assigned  $s_i$  in *M*.

For  $s \in S$ , let M(s) be the set of projects to which s is assigned in M. For  $r \in P \cup L$ , let M(r) be the set of students assigned to r in M. If  $M(s_i) = \emptyset$ , we say that the student  $s_i$  is *unassigned* in M, otherwise  $s_i$  is *assigned* in M. We say that the project  $p_j$  is *under-subscribed*, full, or *over-subscribed* with respect to M according to whether  $|M(p_j)| < c_j$ ,  $|M(p_j)| = c_j$ , or  $|M(p_j)| > c_j$ , respectively, under M. If  $|M(p_j)| > 0$ , we say that  $p_j$  is *non-empty*, otherwise, it is *empty*. Corresponding definitions apply to each lecturer  $\ell$ .

A matching *M* is an assignment such that  $|M(s_i)| \leq 1$  for each  $s_i$ ,  $|M(p_j)| \leq c_j$  for each  $p_j$ , and  $|M(\ell_k)| \leq d_k$  for each  $\ell_k$ . For a matching *M*, if  $|M(s_i)| = 1$ , we may use  $M(s_i)$  to denote the unique project to which  $s_i$  is assigned. The size of a matching *M*, denoted |M|, is the number of students assigned in *M*.

Given a matching M, a (student, project) pair  $(s_i, p_j)$  blocks M, or is a blocking pair for M, if the following three conditions are met:

1.  $p_i \in A_i$ .

- 2. Either  $s_i$  is unassigned or  $s_i$  prefers  $p_i$  to  $M(s_i)$ .
- 3.  $p_i$  is under-subscribed and either
  - (a)  $s_i \in M(\ell_k)$  and  $\ell_k$  prefers  $p_i$  to  $M(s_i)$ , or
  - (b)  $s_i \notin M(\ell_k)$  and  $\ell_k$  is under-subscribed, or
  - (c)  $s_i \notin M(\ell_k)$ ,  $\ell_k$  is full, and  $\ell_k$  prefers  $p_i$  to  $\ell_k$ 's worst non-empty project,
  - where  $\ell_k$  is the lecturer who offers  $p_i$ .

Given a matching *M*, a *coalition* is a set of students  $\{s_{i_0}, s_{i_1}, \ldots, s_{i_{r-1}}\}$  for some  $r \ge 2$  such that each  $s_{i_j}$  is assigned in *M* and prefers  $M(s_{i_j+1})$  to  $M(s_{i_j})$ , where j + 1 is taken modulo *r*. A matching that has no blocking pair nor coalition is *stable*. Refer to [8] for the validity of this definition of stability. SPA-P is the problem of finding a stable matching, and MAX-SPA-P is the problem of finding a maximum stable matching.

We say that A is an r-approximation algorithm if it satisfies  $OPT(I)/A(I) \leq r$  for all instances I, where OPT(I) and A(I) are the sizes of the optimal and the algorithm's solutions for I, respectively.

#### 3. Approximability

#### 3.1. Algorithm SPA-P-APPROX-PROMOTION

Manlove and O'Malley's algorithm SPA-P-APPROX [8] proceeds as follows. First, all students are unassigned. Any student (*s*) who has non-empty preference list applies to the top project (*p*) on the current list of *s*. If the lecturer ( $\ell$ ) who offers *p* has no incentive to accept *s* for *p*, then *s* is rejected. When rejected, *s* deletes *p* from the list. Otherwise, (*s*, *p*) is added to the current matching. If, as a result,  $\ell$  becomes over-subscribed,  $\ell$  rejects a student from  $\ell$ 's worst non-empty project to satisfy the capacity constraint. This continues until there is no unassigned student whose preference list is non-empty. Manlove and O'Malley proved that the obtained matching is stable.

We extend SPA-P-APPROX using Király's idea [7]. During the execution of our algorithm SPA-P-APPROX-PROMOTION, each student has one of two states, "unpromoted" or "promoted". At the beginning, all of the students are unpromoted. The application sequence is unchanged. When a student (s) becomes unassigned with her preference list exhausted, s is promoted.

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