Stable Matching with Dynamic Priorities

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In many matching markets, agents may get prioritized depending on the assignment of other agents. Examples include (i) school choice, where students get prioritized if one of their siblings is assigned to a school; (ii) refugee resettlement, where the clearinghouse prioritizes the assignment of refugees with similar backgrounds (e.g., language) to the same cities; and (iii) the hospital-resident problem, where members of a couple may get prioritized depending on the assignment of their partner. In all these cases, priorities are dynamic and depend on the allocation. This paper introduces the problem of finding a student-optimal stable matching under dynamic priorities. We show that the problem is NP-hard, and we provide two integer programming formulations for the problem. Given the potentially high computational time to solve these formulations, we introduce several heuristics to preprocess the instances and solve them. Finally, we show that clearinghouses can significantly improve students' welfare when considering dynamic priorities using synthetic and real data from the Chilean school choice system.

Key words: stable matching, many-to-one matching, school choice, integer programming.

1. Introduction

The theory of two-sided many-to-one matching markets, introduced by Gale and Shapley (1962), provides a framework for solving many large-scale real-life assignment problems. Examples include entry-level labor markets for doctors and teachers, education markets from daycare, school choice and college admissions, and other applications such as refugee resettlement. A common feature in many of these markets is the use of mechanisms that find a stable assignment, as this guarantees that no coalition of agents has incentives to circumvent the match.

In many of these markets, the clearinghouse may be interested in finding a stable allocation, and individual agents care about their assignment and that of other agents. In the hospital-resident problem, couples jointly participate and must coordinate to find two positions that complement each other. In school choice, students may prefer to be assigned with their siblings or neighbors. In refugee resettlement, agencies may prioritize allocating families with similar backgrounds (e.g., from the same region or speaking the same language) to the same cities. One approach to accommodate these joint preferences is to provide priorities contingent on the assignment. For example, many school choice systems (including NYC, New Haven, Denver, Chile, etc.) consider sibling priorities,

by which students get prioritized in schools where they have a sibling currently assigned or enrolled. In refugee resettlement, families may get higher priority in localities where they have relatives based on *family reunification*. As a result, priorities may not be fixed and may depend on the current assignment. We refer to these as *dynamic priorities*.

In this paper, we study the problem of finding a stable matching under dynamic priorities, using school choice with siblings as a motivating example. To accomplish this, we first introduce a stylized model where students belong to (potentially different) grade levels and get higher priority in a given school if they have a sibling assigned to it. We show that approaches used in practice, such as sequentially solving the assignment for each grade level and updating priorities, can lead to suboptimal results. Moreover, we show that the problem of finding the maximum cardinality stable assignment under dynamic priorities is NP-hard. Given these results, we focus on deriving exact and near-optimal approaches to solve the problem. Specifically, we provide two exact formulations of the problem, and we introduce several heuristics to efficiently obtain near-optimal solutions. Finally, using data from the Chilean school choice system, we show that our framework can significantly increase the welfare of students by assigning more students with their siblings and increasing the overall number of students assigned.

1.1. Contributions

Our paper contributes to the literature in many dimensions:

Problem formulation. In this area:

- Identify issues with current practice: arbitrary preferences and priorities impact number of students assigned.
- Propose a model that captures main elements problems in practice, including dynamic priorities (e.g., siblings).
 - Show that problem is hard.

Stability concept. In this area:

- Introduce two notions of stability: dependent and independent.
- Show existence
- Strategy-proofness

Mathematical programming approach. In this area:

- Propose integer programming formulation to address the problem.
- Show correctness
- Show that different goals (maximal use of reservations, maximize size of match) can be achieved by modifying the objective.
 - Propose algorithm to reduce size of instance
 - Heuristics

Application and impact. In this area:

- Apply our framework to Chile.
- Show that the welfare of students can be significantly increase with our approach.

The remainder of this paper is organized as follows. In Section 2 we discuss the relevant literature. In Section 3 we introduce our model. In Sections 4 we study exact formulations of the problem. In Section 5 we describe heuristics to reduce the size of the problem and obtain near-optimal solutions in a relatively short time. In Section 6 we adapt our framework to the Chilean school choice problem and show its potential benefits compared to the current practice. Finally, in Section 7 we conclude.

2. Literature

Our paper is related to several strands of the literature.

School choice. Starting with Abdulkadiroğlu and Sönmez (2003), a large body of literature has studied different elements of the school choice problem, including the use of different mechanisms such as DA, Boston, and TTC (Abdulkadiroğlu et al. 2005b); the use of different tie-breaking rules (Arnosti 2015, Ashlagi et al. 2019); the handling of multiple and potentially overlapping quotas; the addition of affirmative action policies; and the implementation in many school districts and countries.

Abdulkadiroğlu and Sönmez (2003), (Abdulkadiroğlu et al. 2005a), (Abdulkadiroğlu et al. 2005b). Barcelona (Calsamiglia and Güell 2018), Amsterdam (?) and Chile Correa et al. (2022) have implemented centralized school choice systems using some variant of DA, BM, or Top-Trading Cycles (TTC). Abdulkadiroğlu and Sönmez (2003) also initiated a large literature that theoretically analyzes the school choice problem.

Optimization in stable matching.

- Starters: Gusfield and Irving (1989), Vate (1989), Rothblum (1992), Roth et al. (1993)
- Formulations: Baïou and Balinski (2000), Kwanashie and Manlove (2014), Ágoston et al. (2016), Ágoston et al. (2021), Delorme et al. (2019).
- Alternative objectives: Shi (2016), Ashlagi and Shi (2016), Bodoh-Creed (2020), Bobbio et al. (2022)

Siblings. Dur et al. (2022), Dur and Wiseman (2019), Correa et al. (2022) Complementarities.

3. Model

We formalize the *stable matching problem with dynamic priorities* using school choice with siblings as a motivating example.

Let S be the set of students, and let F be a partition of students into families, so that $f(s) \in F$ represents the family of student $s \in S$. Then, we say that students s and s' are siblings if f(s) = S.

f(s'), and we say that s has no siblings if $f(s) = \{s\}$. In addition, we assume that each family f has a strict preference order \succ_f over tuples in the set $(C \cup \{\emptyset\})^{|f|}$, where \emptyset represents that a student is unassigned. Specifically, if $f = (f_1, f_2)$, then $c, c' \succ_f c'', c'''$ implies that family f prefers that f_1 and f_2 go to schools c and c' over c'' and c''', respectively. Finally, let \mathcal{G} be the set of grade levels, g(s) the grade of student s, $\mathcal{S}^g \subseteq \mathcal{S}$ the set of students applying to grade level $g \in \mathcal{G}$, and S_c^g the set of students applying to school $c \in \mathcal{C}$ in grade $g \in \mathcal{G}$.

On the other side of the market, let C be the set of schools. Each school c offers q_{cg} seats on each grade $g \in \mathcal{G}$, with $q_{cg} = 0$ representing that school c does not offer grade g. For simplicity, we assume that there are no special reserves or quotas.³ In addition, we assume that all schools consider a sibling priority, i.e., they prioritize students who have a sibling assigned or enrolled in the school. As a result, each school has two priority groups: (i) students with siblings and (ii) students with no siblings in the school, where students in the former group are strictly preferred to students in the latter. Within each priority group, schools break ties with a random tie-breaker. It is important to notice that the random tie-breaker defines a strict order of students only within each priority group; however, it does not define an order among any two pair of students, as they may change from one priority group to the other depending on the assignment of their siblings. Nevertheless, we assume that each school c has preferences over sets of students in $2^{\mathcal{S}}$, which we denote by \succ_c .

Let $\mathcal{V} \subseteq \mathcal{S} \times \mathcal{C} \cup \{\emptyset\}$ be the seat of feasible pairs, i.e., $(s,c) \in \mathcal{V}$ implies that student s applied to c, satisfies all the admission requirements, and $q_{c,g(s)} > 0$. Similarly, let $\mathcal{V}^g \subseteq \mathcal{V}$ be the set of feasible pairs including students applying to grade $g \in \mathcal{G}$. A matching is an assignment $\mu \subseteq \mathcal{V}$ such that (i) each student is assigned to at most one school in \mathcal{C} , and (ii) each school is assigned at most its capacity in each grade level. Formally, given an assignment μ , let $\mu(s) \in \mathcal{C} \cup \{\emptyset\}$ be the school assigned to student s, and let $\mu(c)$ be the set of students assigned to school c. Then, a matching satisfies that (i) $\mu(s) \in \mathcal{C} \cup \{\emptyset\}$ for all students $s \in \mathcal{S}$ and (ii) $|\{s \in \mu(c) : g(s) = g\}| \leq q_{cg}$ for all schools $c \in \mathcal{C}$ and grade levels $g \in \mathcal{G}$.

As Roth (2002) discusses, a desired property of any assignment is stability, i.e., that there are no pair of agents that prefer to circumvent the match and be assigned to each other. Given an assignment μ , we say that student s has justified envy towards another student s' assigned to school c if (i) g(s) = g(s'), (ii) $(c, \mu(f \setminus \{s\})) \succ_f \mu(f)$, and(iii) $\mu(c) \cup \{s\} \setminus \{s'\} \succ_c \mu(c)$. In words, the first condition states that both students belong to the same grade level; the second condition implies

¹ In a slight abuse of notation, we use \sim_f to refer that family f is indifferent between two subsets of schools, i.e., for any $\vec{c}, \vec{c}' \in (\mathcal{C} \cup \{\emptyset\})^{|f|}, \vec{c} \sim_f \vec{c}'$ if neither $\vec{c} \succ_f \vec{c}'$ nor $\vec{c}' \succ_f \vec{c}$ hold.

² Notice that the model applies directly in cases where there is a "single grade level", as in refugee resettlement, college admissions and in the hospital-resident problem.

³ In the Appendix we extend our formulation to include overlapping types and reserves.

that the family prefer that $s \in f$ is assigned to c given the assignment of their siblings; and the third condition states that school c prefers the set of student that replaces s' with s. In addition, we say that an assignment μ is non-wasteful if there is no student $s \in \mathcal{S}$ and school c such that $\mu(c, \mu(f \setminus \{s\})) \succ_f \mu(f)$ and $|\{s' \in \mu(c) : g(s') = g(s)\}| < q_{cg}$. Then, we say that an assignment is stable if it is envy-free (i.e., no student has justified envy) and non-wasteful.

As Correa et al. (2022) discuss, a stable assignment may not exist (see Theorem 4 in Appendix A). Moreover, families may not be able to report preferences over tuples of schools, as the number of combinations increases exponentially in the number of siblings applying to the system. For these reasons, many school systems elicit a preference list for each family member, i.e., a set of preference orders $\{\succ_i\}_{i\in f}$ among schools in $\mathcal{C}\cup\{\emptyset\}$. Then, as proposed in Correa et al. (2022), one option to solve the problem is to define an order in which grades are processed and sequentially solve the assignment of each grade level using the student-optimal variant of DA. More specifically, the algorithm in Correa et al. (2022) starts processing the highest grade (i.e., 12th grade). Then, before moving to the next grade, the sibling priorities are updated, considering the assignment of the grade levels already processed. After processing the final grade level (i.e., Pre-K), this procedure finishes.

Notice that the heuristic mentioned above obtains a stable assignment is the preferences of families satisfy *higher-first*, i.e., each family prioritizes the assignment of their oldest member (see Proposition 2 in Correa et al. (2022)). However, this is not the case if some families' preferences do not satisfy this condition. In addition, as Example 1 illustrates, the order in which grades are processed matters.

EXAMPLE 1. Consider an instance with two grades $g_1 < g_2$, one family $f = \{f_1, f_2\}$, and two additional students $\{a_1, a_2\}$. Students f_1 and a_1 apply to grade g_1 , and f_2 and a_2 apply to grade g_2 . Finally, the preferences and priorities are:

$$\succ_{f_1} : c_2, c_1, \qquad \succ_{a_1} : c_2, c_1$$

$$\succ_{f_2} : c_1, c_2, \qquad \succ_{a_2} : c_1, c_2$$

$$\succ_{c_1}^{g_1} : a_1, f_1, \qquad \succ_{c_1}^{g_2} : a_2, f_2$$

$$\succ_{c_2}^{g_1} : a_1, f_1, \qquad \succ_{c_2}^{g_2} : a_2, f_2$$
(1)

We observe that, if grades are processed in decreasing order (as in Chile), we obtain the assignment $\mu = \{(f_1, c_1), (a_1, c_2), (f_2, c_1), (a_2, c_2)\}$. In contrast, if we process grades in increasing order, we obtain the assignment $\mu' = \{(f_1, c_2), (a_1, c_1), (f_2, c_2), (a_2, c_1)\}$.

Moreover, defining an order in which grades must be processed imposes an additional constraint that may rule out a stable assignment that is weakly preferred by every family (and strictly preferred by at least one family). This is illustrated in Example 2.

EXAMPLE 2. Consider an instance with two grades, $G = \{g_1, g_2\}$, $g_1 < g_2$, two families $f = \{f_1, f_2\}$, $f' = \{f'_1, f'_2\}$, and a set of students $S = \{f_1, f_2, f'_1, f'_2, a_1, a_2, a_3, a_4\}$. In addition, consider two schools $C = \{c_1, c_2\}$ offering two seats in each grade. Finally, suppose that the preferences of students are given by:

$$f: (c_{1}, c_{1}) \sim (c_{2}, c_{2}) \succ (c_{1}, c_{2}) \succ (c_{2}, c_{1}), \qquad a_{1}: c_{1}$$

$$f': (c_{1}, c_{1}) \sim (c_{2}, c_{2}) \succ (c_{1}, c_{2}) \succ (c_{2}, c_{1}), \qquad a_{2}: c_{2}$$

$$i: c_{1} \succ c_{2}, \forall i \in \{f_{1}, f'_{1}\}, \qquad a_{3}: c_{1}$$

$$i: c_{2} \succ c_{1}, \forall i \in \{f_{2}, f'_{2}\}, \qquad a_{4}: c_{2},$$

$$(2)$$

while schools priorities are:

$$c_1: f \succ f' \succ \{a_1, a_3\}$$

 $c_2: f \succ f' \succ \{a_2, a_4\}$ (3)

If we process grades starting with g_1 , we obtain the assignment

$$\mu = \{(f_1, c_1), (f_2, c_1), (f'_1, c_1), (f'_2, c_1), (a_1, \emptyset), (a_2, c_2), (a_3, \emptyset), (a_4, c_2)\},\$$

i.e., two students result unassigned. However, notice that the allocation

$$\mu' = \{(f_1, c_1), (f_2, c_1), (f'_1, c_2), (f'_2, c_2), (a_1, c_1), (a_2, c_2), (a_3, c_1), (a_4, c_2)\}$$

satisfies our notion (to be refined) of stability and leads to no students unassigned. \Box

Examples 1 and 2 show that the order in which grade levels are processed matters and that imposing an order may reduce the total number of students assigned. These two examples motivate Problem 1, which aims to find the student-optimal stable assignment of maximum cardinality by leveraging the siblings' priority.

3.1. Practical preferences and priorities

The definition of *justified envy* in the previous section assumes that schools have preferences over sets of students, and that families have joint preferences over sets of schools. However, in most clearinghouses, preferences and priorities are not as complex, and simply involve preferences by students and a combination of random tie-breakers and priority groups to define schools' priorities. For this reason, in the remainder of the paper we will assume a simplified structure of preferences and priorities, as formalized in Assumption 1.

Assumption 1. We assume that students' preferences and school priorities are such that:

- 1. On the students side, we assume that each student report a strict preference list.
- 2. On the schools side, we assume that each school ranks students according to a random tiebreaker, and that there is a single priority group that gives higher priority to students with siblings assigned or enrolled in the school.

Given this assumption, we will use \succ_c to represent ex-ante priorities at school c, i.e., the priorities derived from the random tie-breaker without considering siblings priorities. Although Assumption 1 simplifies the reporting of preferences, the siblings' priority needs some limitations to ensure the fairness of the assignment, as the following example illustrates.

EXAMPLE 3. Consider an instance with a single level, a set of students $S = \{a_1, a_2, f_1, f_2\}$ where f_1 and f_2 are siblings, and a single school with two seats. Moreover, suppose ex-ante priorities are $a_1 \succ_c a_2 \succ_c f_1 \succ_c f_2$. Then, one possible assignment is $\mu = \{(a_1, c), (a_2, c), (f_1, \emptyset), (f_2, \emptyset)\}$. However, the alternative assignment $\mu' = \{(a_1, \emptyset), (a_2, \emptyset), (f_1, c), (f_2, c)\}$ would still be feasible, as both f_1 and f_2 have siblings' priority and thus have the highest priority ex-post.

We claim that assignment μ' in Example 3 is not desirable, since neither f_1 nor f_2 would be admitted without siblings priority. To rule out this possibility, we restrict attention to assignments that satisfy *unilateral dynamic priorities*, which are formalized in Definition 1.

DEFINITION 1. Consider an assignment $\mu \subseteq \mathcal{V}$. Then, μ satisfies unilateral dynamic priorities if, for any student $s' \in f$ with a sibling $s'' \in f$ assigned to the same school c, either $s' \in \mu_{-s''}(c)$ or $s'' \in \mu_{-s'}(c)$, where μ_{-s} is the assignment that would be obtained if student s were excluded from the set of students.

In words, *unilateral priorities* ensure that at least one of the siblings is independently assigned to the school, potentially providing priority to their other siblings.

It remains to describe how ties are broken among students with siblings priority. One possible option, which we refer to as the *dependent rule*, is that students that ties among students that get prioritized are broken based on the priority of their family member that granted them their siblings priority. The other alternative, which we refer to as the *independent rule*, is that we break ties among prioritized students based on their own original priority. In Example 4 we illustrate these two rules.

EXAMPLE 4. Consider a single school c offering three seats, and two families, $f = \{f_1, f_2\}, g = \{g_1, g_2\}$, with all students applying to the same grade. Moreover, suppose that ex-ante priorities are $f_1 \succ_c g_1 \succ_c g_2 \succ_c f_2$. Then, $\mu(f_1) = c$ and $\mu(g_1) = c$. As a result, both f_2 and g_2 get siblings priority, but there is only one seat left. If the dependent rule is in place, then $\mu(f_2) = c$ and $\mu(g_2) = \emptyset$, since $f_1 \succ_c g_1$. On the other hand, if the independent rule is in place, $\mu(f_2) = \emptyset$ and $\mu(g_2) = c$, since $g_2 \succ_c f_2$.

Note that the *dependent* and the *independent* rules are used in practice. On the one hand, the former is used in Chile (Correa et al. 2022), as the clearinghouse breaks ties at the family level first, and then breaks ties within each family. On the other hand, the latter is used in NYC to break ties among students with siblings priority [Confirm that this is the case.]. Hence, which rule to use is a policy relevant question that we evaluate in Section 6.

Based on these two rules (i.e., dependent vs. independent rules), we define two notions of justifiedenvy: (i) dependent-justified envy and (ii) independent-justified envy. Before formally describing these, let $\zeta_{\mu}(s,c) = \arg\max_{k \in f(s) \setminus \{s\}} \{p_{k,c} : \mu(k) = c, k \succ_c s\}$ be the function that returns the sibling of student s assigned to c with the highest ex-ante priority. In a slight abuse of notation, we use $\zeta_{\mu}(s,c) = \emptyset$ if there is no such sibling, and thus $\zeta_{\mu}(s,c) \sim_c \zeta_{\mu}(s',c)$ if and only if both s and s' have no siblings assigned to school c.

DEFINITION 2. (Dependent-justified envy) A student s has dependent-justified envy toward another student s' assigned to a school c if (i) g(s) = g(s'), (ii) $c \succ_s \mu(s)$, and (iii) either $\zeta_{\mu}(s,c) \succ_c \zeta_{\mu}(s',c)$ or $\zeta_{\mu}(s,c) \sim_c \zeta_{\mu}(s',c)$ and $s \succ_c s'$.

DEFINITION 3. (Independent-justified envy) A student s has independent-justified envy toward another student s' assigned to a school c if (i) g(s) = g(s'), (ii) $c \succ_s \mu(s)$, and (iii) either $|\zeta_{\mu}(s,c)| > |\zeta_{\mu}(s',c)|$ or $|\zeta_{\mu}(s,c)| = |\zeta_{\mu}(s',c)|$ and $s \succ_c s'$.

Finally, we say that an allocation is (in)dependent-stable if it is (in)dependent-justify envy free and non-wasteful.

3.2. Maximum cardinality stable matching

We now formalize the problem that we focus on this paper.

PROBLEM 1. Given an instance Γ , the stable-matching with unilateral dynamic priorities aims to find a student-optimal stable-assignment with unilateral priorities of maximum cardinality. Problem 1 is relevant from a policy perspective because it combines two of the primary goals in many matching markets: (i) to find an assignment involving as many agents as possible, and (ii) incorporate dynamic priorities. However, in Theorem 1 we show that Problem 1 is NP-Hard.

Theorem 1. Problem 1 is NP-hard, independent of the rule used to break ties among students with siblings priority.

Proof. [TODO: I guess we should follow Manlove et al. 2002] [I think the easiest way to prove this is by showing that the problem with siblings can be modeled as an instance of the problem with ties and incomplete lists, which is known to be NP-complete. Manlove 2002 shows that the problem is NP-complete even if ties are in only one side and ties are of length max 2; if we think of siblings as being tied, then this would exactly apply.]

[Maybe add some result regarding approximability? Maybe we can find something similar to what is in Bobbio et al. 2022.]

4. Exact Formulations

[I'm working on this now]

Let $x_{s,c}$ be a binary variable equal to 1 if the clearinghouse assigns student $s \in \mathcal{S}$ to school $c \in \mathcal{C} \cup \{\emptyset\}$, and 0 otherwise. Then, the set of matchings can be represented as

$$\mathcal{P} = \left\{ \mathbf{x} \in \left\{0, 1\right\}^{\mathcal{V}} : \text{ for each } g \in \mathcal{G}, \sum_{c: (s, c) \in \mathcal{V}^g} x_{s, c} = 1 \text{ for all } s \in \mathcal{S}, \sum_{s: (s, c) \in \mathcal{V}^g} x_{s, c} \leq q_{c, g} \text{ for all } c \in \mathcal{C} \right\}.$$

The first condition captures that student s is assigned to one school in the set $\mathcal{C} \cup \{\emptyset\}$, while the second condition implies that each school c receives at most $q_{c,g}$ students. In addition, let $r_{s,c}$ be the position of school $c \in \mathcal{C}$ in the preferences of student $s \in \mathcal{S}$, and let $r_{s,\emptyset}$ be a parameter that captures the cost of having unassigned students. As Ágoston et al. (2016) and Bobbio et al. (2022) discuss, the problem of finding the student-optimal stable assignment in the standard setting (e.g., with no dynamic priorities) is equivalent to solving the following integer problem:

$$\max_{\mathbf{x} \in \mathcal{P}} \sum_{(s,c) \in \mathcal{V}} r_{s,c} \cdot x_{s,c} \tag{4a}$$

$$st. q_{c,g(s)} \left(1 - \sum_{c' \succeq_s c} x_{s,c'} \right) \le \sum_{\substack{s' \in S^g(s): \\ s' \succeq_c s}} x_{s',c}, \, \forall (s,c) \in \mathcal{V}$$
 (4b)

where the set of constraints 4b, first introduced in Baïou and Balinski (2000), ensure that the assignment is stable. Note that this problem is separable by grade level, and thus can be efficiently solved applying DA to each grade separately. However, as we will discuss next, considering dynamic priorities introduce dependencies between grade levels, and thus the problem is not separable anymore.

4.1. Base formulation

Given a family $f \in \mathcal{F}$ with |f| > 1, for any $s, s' \in f$ with $s \succ_c s'$, let $y_{s,s',c}$ be a binary variable equal to 1 if s gives siblings' priority to s', and 0 otherwise. In words, $y_{s,s',c}$ is equal to 1 if siblings s and s' are assigned to school c and s (i.e., $x_{s,c} = x_{s',c} = 1$) and s has the highest priority among all siblings of s' assigned to school c, and 0 otherwise.

To linearly incorporate these variables into our formulation, we consider the set

$$\mathcal{M}(\mathbf{x}) = \left\{ \mathbf{y} \in [0, 1]^{\mathcal{S} \times \mathcal{V}} : y_{s, s', c} \leq x_{s, c}, y_{s, s', c} \leq x_{s', c}, \sum_{s'' \in f(s'): s'' \succ s'} y_{s'', s', c} \leq 1, \right.$$

$$\left. \sum_{s'' \in f: s'' \succeq_{c} s} y_{s'', s', c} \geq x_{s, c} + x_{s', c} - 1, \ s, s' \in f \text{ with } s \succ_{c} s', \ f \in \mathcal{F}, \ c \in \mathcal{C} \right\}$$

[Explain why this set correctly defines y.]

4.1.1. Independent case. Based on the set of feasible assignments $\mathbf{x} \in \mathcal{P}$ and the joint assignment variables $\mathbf{y} \in \mathcal{M}(\mathbf{x})$, we can formulate Problem 1 by adapting (4) to incorporate the siblings priority.

$$\min_{\mathbf{x} \in \mathcal{P}, \mathbf{y} \in \mathcal{M}(\mathbf{x})} \qquad \sum_{(s,c) \in \mathcal{V}} r_{s,c} \cdot x_{s,c} \tag{5a}$$

$$st. q_{c,g(s)} \left(1 - \sum_{\substack{c' \succeq_s c}} x_{s,c'} \right) \leq \sum_{\substack{s' \in \mathcal{S}^{g(s)}: \\ s' \succeq_c s}} x_{s',c} + \sum_{\substack{(s',s'') \in \mathcal{S}^{g(s)} \times f(s'): \\ c' = s'}} y_{s'',s',c}, \forall (s,c) \in \mathcal{V},$$
 (5b)

$$\left(1 - \sum_{c' \succeq_s c} x_{s,c'}\right) + x_{s',c} \le 2 - x_{l,c} + \sum_{k \in f(l)} y_{k,l,c} \cdot \mathbb{1}_{\{l \succ_c s\}}, \forall c \in \mathcal{C}, \{s, s'\} \subseteq f \in \mathcal{F}, l \in \mathcal{S}^{g(s)} \quad (5c)$$

$$\left(1 - \sum_{c' \succeq_s c} x_{s,c'}\right) \le 1 - x_{l,c} \cdot \mathbb{1}_{\{s \succ_c l\}} + \sum_{k \in f(l)} y_{k,l,c} \cdot \mathbb{1}_{\{s \succ_c l\}}, \, \forall c \in \mathcal{C}, s, l \in \mathcal{S}^{g(s)} \tag{5d}$$

The sets of constraints (5b) adapt the stability constraints (4b) to incorporate dynamic priorities and eliminate independent justified-envy. Specifically, a student s may not get assigned to school c (or better) if there are at least $q_{c,g(s)}$ students with (i) higher priority assigned to c (first term on the right-hand side) or (ii) lower priority than s but that have a sibling assigned to the school, and thus obtain siblings' priority. The sets of constraints (5c) guarantee that ties among students with dynamic priority are broken according to their priorities, thus satisfying independent justified-envy. On the one hand, if student s has no sibling assigned to s, then the left-hand side of (??) is at most one, and thus the constraint is always satisfied. On the other hand, if $s \in f(s')$ is assigned to school s, s' gets dynamic priority. As a result:

- No student $l \in \mathcal{S}^(g(s'))$ without dynamic priority can be assigned to c unless s' is assigned to c or better. This is captured by the fact that, if l has no dynamic priority, then the summation on the right-hand size is equal to zero, and thus $x_{l,c} = 0$ if $x_{s,c} = 1$ and $\sum_{c' \succeq_{s} c} x_{s,c'} = 0$.
- Student s' may not get assigned to c or better only if student l has dynamic priority and is preferred by school c.

Finally, the set of constraints (5d) guarantees that priorities are respected among students with no dynamic priority. If student s is not assigned to c or better, the left-hand side is equal to 1, and thus a student less preferred l such that $s \succ_c l$ can only be assigned to c if they have dynamic priority. On the other hand, if $l \succ_c s$ then $\mathbb{1}_{\{s \succ_c l\}} = 0$ and thus the constraint is always satisfied.

In Theorem 2 we show that the integer programming formulation 5 solves Problem 1 under independent justified-envy.

Theorem 2. The formulation in (5) solves Problem 1 when we consider independent justifiedenvy. *Proof.* We need to show that

- 1. The formulation returns an allocation that is stable with unilateral dynamic priorities and that is free of independent justified-envy.
 - 2. The allocation obtained is student-optimal and maximizes the cardinality of the match.

In a slight abuse of notation, we will refer x^*, y^* to the optimal solution of (5), and we will denote by μ^* the corresponding matching, i.e., $\mu^*(s) = c$ if and only if $x^*_{s,c} = 1$. Similarly, $\mu^*(c) = \{s \in \mathcal{S} : x^*_{s,c} = 1\}$, and we say that $(s,c) \in \mu^*$ if $\mu^*(s) = c$ and $s \in \mu^*(c)$, respectively.

First, by definition of y we know that $y_{s,s',c} = 1$ only if $s \succ_c s'$ and both $x_{s,c} = x_{s',c} = 1$. Hence, by definition of y, it is direct that the allocation will satisfy that dynamic priorities are unilateral. Next, we show that the allocation that results from solving (5) is free of independent justified-envy. To accomplish this we proceed by parts.

Non-wastefulness. To find a contradiction, suppose there exists $(s,c) \notin \mu^*$ such that $c \succ_s \mu^*(s)$ and $|\mu^*(c) \cap \mathcal{S}^{g(s)}| < q_{c,g(s)}$. Then, we know that $\sum_{c' \succeq_{sc}} x_{s,c'}^* = 0$, and therefore the left-hand side of (5b) is equal to $q_{c,g(s)}$. As a result, we must have that

$$\sum_{\substack{s' \in \mathcal{S}^{g(s)}: \\ s' \succ_{c} s}} x_{s',c}^* + \sum_{\substack{(s',s'') \in \mathcal{S}^{g(s)} \times f(s'): \\ s' \prec_{c} s}} y_{s'',s',c}^* \ge q_{c,g(s)}.$$

Since there is no student s' satisfying that $s' \succ_c s$ and $s' \prec_c s$, we know that the two summations are over disjoint sets of students in $\mathcal{S}^{g(s)}$, and thus we know that

$$|\mu^*(c)| = \sum_{\substack{s' \in \mathcal{S}^{g(s)}: \\ s' \succ cs}} x_{s',c}^* + \sum_{\substack{(s',s'') \in \mathcal{S}^{g(s)} \times f(s'): \\ s' \succ cs}} y_{s'',s',c}^* \ge q_{c,g(s)} \ge q_{c,g(s)}.$$

Hence, we obtain a contradiction as we initially assumed that $|\mu^*(c) \cap \mathcal{S}^{g(s)}| < q_{c,g(s)}$.

No independent justified-envy. To find a contradiction, suppose there exists $(s, c) \notin \mu^*$ such that $c \succ_s \mu^*(s)$ and that s has independent justified-envy towards student $s' \in \mathcal{S}^{g(s)}$. Then, there are two options:

- 1. $|\zeta_{\mu^*}(s,c)| > |\zeta_{\mu^*}(s',c)|$, in which case s has a sibling assigned to school c and s' does not. Hence, we have that $x_{\tilde{s},c}^* = 1, \sum_{c' \succeq_s c} x_{s,c'} = 0$, and $x_{s',c} = 1$, where $\tilde{s} \in f(s)$. As a result, the left-hand side of (5c) is equal to 2 for (\tilde{s},s) , and therefore $-x_{s',c}^* + \sum_{k \in f(s')} y_{k,s',c} \cdot \mathbb{1}_{\{s' \succ s\}}$ must be equal to zero. However, by hypothesis we know that s' has no siblings assigned to school c, and thus the second term is equal to zero. Hence, for the constraint to be satisfied we need that $x_{s',c}^* = 0$, which contradicts the initial hypothesis that student s' is assigned to school c.
- 2. $|\zeta_{\mu^*}(s,c)| = |\zeta_{\mu^*}(s',c)|$ and $s \succ_c s'$, in which case both students have the same priority condition. Since s has independent justified-envy towards s', we know that $x_{s',c} = 1$, and $\sum_{c' \succeq c} x_{s,c'} = 0$. Then, we have two sub-cases:

- If both students do not have dynamic priority, then the left-hand side of (5d) is equal to 1. In addition, we know by hypothesis that $s \succ_c s'$, and thus $\mathbb{1}_{\{s \succ_c s'\}} = 1$. As a result, the right-hand side of (5d) becomes $1 x_{s',c} + \sum_{k \in f(s')} y_{k,s',c} = 1 x_{s',c}$, where the latter follows since both s and s' have no dynamic priority in school s'. Finally, this implies that s', which contradicts our hypothesis that s' has justified-envy.
- If both students have dynamic priority, then (5d) is always satisfied. On the other hand, the left-hand side of (5c) is equal to 2, and thus we must have that $-x_{s',c} + \sum_{k \in f(s')} y_{k,s',c} \cdot \mathbb{1}_{\{s' \succ_c s\}} = 0$. Since $\mathbb{1}_{\{s' \succ_c s\}} = 0$, we must have that $x_{s',c} = 0$, which leads to a contradiction.

Maximum caridnality. To find a contradiction, suppose there exists another student-optimal assignment μ' such that $\sum_{s \in \mathcal{S}} \mathbb{1}_{\{\mu'(s) \in \mathcal{C}\}} > \sum_{s \in \mathcal{S}} \mathbb{1}_{\{\mu^*(s) \in \mathcal{C}\}}$. Let $x'_{s,c} = 1$ if $\mu'(s) = c$, and $x'_{s,c} = 0$ otherwise, and let $\mu'(\emptyset) = \{s \in \mathcal{S} : \mu'(s) = \emptyset\}$ (similar definition applies for $\mu^*(\emptyset)$). Then,

$$\sum_{(s,c) \in \mu'} r_{s,c} = \sum_{(s,c) \in \mu'} x'_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu'(\emptyset) \right| < \sum_{(s,c) \in \mu'} x'_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mu^*} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{s,c} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{s,c} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right| = \sum_{(s,c) \in \mathcal{V}} x^*_{\emptyset} \cdot r_{\emptyset} + r_{\emptyset} \left| \mu^*(\emptyset) \right|$$

where the first equality follows by definition of x'; the inequality follows by hypothesis (i.e., that the cardinality of μ' is higher than that of μ^*); the second equality follows by student-optimality of both assignments, and the last equality follows by definition of x^* . As a result, we obtain that $\sum_{s,c\in\mathcal{V}} x'_{s,c} \cdot r_{s,c} < \sum_{(s,c)\in\mathcal{V}} x^*_{s,c} \cdot r_{s,c}$, which contradicts the optimality of x^* , leading to a contradiction. \square

4.1.2. Dependent case. In the dependent case,

$$\min_{\mathbf{x} \in \mathcal{P}, \mathbf{y} \in \mathcal{M}(\mathbf{x})} \qquad \sum_{(s,c) \in \mathcal{V}} r_{s,c} \cdot x_{s,c} \tag{6a}$$

$$st. q_{c,g(s)} \left(1 - \sum_{\substack{c' \succeq_s c \\ s' \succ_c s}} x_{s,c'} \right) \leq \sum_{\substack{s' \in \mathcal{S}^{g(s)} : \\ s' \succ_c s}} x_{s',c} + \sum_{\substack{(s',s'') \in \mathcal{S}^{g(s)} \times f(s') : \\ s'' \smile_c s}} y_{s'',s',c} , \forall (s,c) \in \mathcal{V},$$
 (6b)

$$x_{s,c} + \left(1 - \sum_{c' \succeq_{s'} c} x_{s',c'}\right) \le 2 - x_{l,c} + \sum_{k \in f(l)} y_{k,l,c} \cdot \mathbb{1}_{\{k \succ_{c} s\}}, \forall c \in \mathcal{C}, \{s, s'\} \subseteq f \in \mathcal{F}, l \in \mathcal{S}^{g(s')}$$

$$(6c)$$

 $\left(1 - \sum_{c' \succeq_s c} x_{s,c'}\right) \le 1 - x_{l,c} \cdot \mathbb{1}_{\{s \succ_c l\}} + \sum_{k \in f(l)} y_{k,l,c} \cdot \mathbb{1}_{\{s \succ_c l\}}, \, \forall c \in \mathcal{C}, s, l \in \mathcal{S}^{g(s)} \tag{6d}$

In Theorem 3 we show that the integer programming formulation 6 solves Problem 1.

Theorem 3. The formulation in (6) solves Problem 1 when we consider dependent justifiedenvy.

Proof. First, note that the objective of Problems 5 and 6 are the same. In addition, the sets of constraints (6b) and (6d) are equivalent to (5b) and (5d), respectively. Hence, the proofs for

non-wastefulness and maximum cardinality are the same. Hence, we only need to show that the optimal solution of Problem 6 is free of dependent justified-envy.

To show this, we proceed by contradiction. Suppose there exists $(s,c) \notin \mu^*$ such that $c \succ_s \mu^*(s)$ and that s has dependent justified-envy towards student $s' \in \mathcal{S}^{g(s)}$. As for the independent case, there are two options:

1. $\zeta_{\mu^*}(s,c) \succ_c \zeta_{\mu^*}(s',c)$, in which case s has a sibling assigned to school c and either s' has no siblings assigned to c (i.e., $\zeta_{\mu^*}(s',c) = \emptyset$) or s' has a sibling assigned to c that has lower priority compared to $\zeta_{\mu^*}(s,c)$. Then, we know that $x_{\zeta_{\mu^*}(s,c),c} = 1$, and thus the left-hand side of (6c) is equal to 2. On the other hand, since either s' has no sibling or has a sibling with lower priority than $\zeta_{\mu^*}(s,c)$, we know that $\sum_{k\in f(l)} y_{k,s',c} \cdot \mathbb{1}_{\{k\succ_c\zeta_{\mu^*}(s,c)\}} = 0$, and therefore the right-hand side is equal to $2-x_{s',c}$. As a result, we must have that $x_{s',c} = 0$, which contradicts the hypothesis that s has dependent-justified envy towards s' in school c.

2. $\zeta_{\mu^*}(s,c) \sim_c \zeta_{\mu^*}(s',c)$ and $s \succ_c s'$, in which case both students have no dynamic priority. Since s has independent justified-envy towards s', we know that $x_{s',c} = 1$, and $\sum_{c' \succeq c} x_{s,c'} = 0$. Then, the left-hand side of (6d) is equal to 1, while the right-hand side is equal to $1 - x_{s',c}$. Hence, we must have that $x_{s',c} = 0$, which leads to a contradiction since we assumed that s has dependent justified-envy towards s' in school c.

5. Heuristics

Given the complexity of the problem, solving the formulations described in Section 4 may take a significant amount of time. In this section, we introduce several heuristics that can reduce computational times. Specifically, in Section 5.1 we discuss preprocessing heuristics that can reduce the size of the problem while keeping the set of optimal solutions. For cases when this preprocessing is not enough to solve the problem reasonably fast, in Section 5.2 we propose heuristics to find near-optimal solutions quickly.

5.1. Preprocessing

[Not sure if these are definitions of propositions. I guess propositions because we need to show that they cannot belong to any stable matching.]

PROPOSITION 1. A pair (s,c) is school-dominated if there exists $c' \succ_s c$

$$|\{s' \in \mathcal{S} \setminus \{s\} : s' \succ_c s\}| \le q_{c,q(s)}$$

Proposition 2. A pair (s,c) is student-dominated if

$$|\{s' \in \mathcal{S} \setminus \{s\} : s' \succ_c s \text{ and } c \succ_{s'} c' \forall c' \in \mathcal{C}, \}| > q_{c,q(s)}$$

PROPOSITION 3. If a pair (s,c) is school-dominated in the case with no dynamic priorities, then it is also school-dominated in the case with dynamic priorities.

Proposition 3 implies that school-dominated nodes will remain dominated after dynamic priorities are taken into account. However, as the following example illustrates, this may not be the case when we consider student-dominated pairs.

EXAMPLE 5. Consider an instance with two levels, $G = \{g_1, g_2\}$, one family $f = \{f_1, f_2\}$, and two additional students, $\{a_1, a_2\}$. Students' subscripts indicate the level to which they belong, i.e., $S^{g_1} = \{f_1, a_1\}$ and $S^{g_2} = \{f_2, a_2\}$. In addition, suppose that there are two schools $C = \{c_1, c_2\}$, each offering one seat per level, and suppose that every student prefers c_1 over c_2 . Finally, suppose that $f_1 \succ_c a_1$ and $a_2 \succ_c f_2$ for every school $c \in C$.

In this case, if dynamic priorities did not exist, then the pair (f_2, c_1) would be student-dominated by the pair (a_2, c_1) , since student a_2 's top preference is c_1 and c_1 prefers a_2 over f_2 . However, with dynamic priorities, we know that student f_1 will be assigned to school c_1 , and thus f_2 will get dynamic priority and will no longer be student-dominated.

PROPOSITION 4. If a pair (s,c) is student-dominated under dynamic priorities if

$$|\{s' \in \mathcal{S} \setminus \{s\} : s' \succ_c s \text{ and } c \succ_{s'} c' \, \forall c' \in \mathcal{C}, \}| > q_{c,q(s)}$$

bla

5.2. Near-Optimal Approaches

We consider three families of heuristics:

- Sequential DA (Correa et al. 2022): This heuristic first defines an order among grade levels (e.g., decreasing from 12th grade to Pre-K), and then sequentially solves the assignment for each grade level using DA. Before moving to the next level, the algorithm updates priorities given by the assignments in the levels already processed, and continues until processing the last level.
- DA and re-allocation: This heuristic starts solving the assignment problem without considering any dynamic priorities. Then, the algorithm identifies students whose siblings were assigned to schools they prefer more, and assigns them to those schools unless they filled their quota with other families. After this, the algorithm computes the number of seats left in each school and level, and then computes the match on the residual graph that involves the remaining capacities and the students who were not previously assigned.

• NRMP algorithm: This algorithm works when participants submit joint lists, and it proceeds same as DA but couples are processed together and thus take two spots every time they propose. If one of the members of the couple is rejected, then both are rejected and they must move to their next joint preference. Note that this algorithm will not perform appropriately in our context because families do not report joint preferences. As a result, list are short, and thus families will most likely not get assigned.

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6. Experiments

- 6.1. Synthetic Data
 - 6.1.1. Setting and Data.
 - 6.1.2. Results.
- 6.2. Real Data: School Choice in Chile
 - 6.2.1. Setting and Data.
 - 6.2.2. Results.

7. Conclusions

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Appendix A: Additional Results

Theorem 4. A stable matching may not exist under dynamic preferences, even in the case where families independent.

Proof. Consider an instance with two families, $f = \{f_1, f_2\}$ and $f' = \{f'_1, f'_2\}$, two schools $\{c_1, c_2\}$ with two grades $G = \{g_1, g_2\}$, and suppose that each school offers one seat in each grade. We assume that g_1 is the lower grade, and g_2 the higher grade, and we assume that the subindex in the members of each family represents the grade to which they are applying. Finally, the preferences and initial priorities are:

$$\succ_{i} : c_{1} \succ_{i} c_{2}, \quad \forall i \in f \cup f'
\succ_{c_{1}}^{g_{1}} : f_{1} \succ_{c_{1}}^{g_{1}} f_{1}'
\succ_{c_{1}}^{g_{2}} : f_{2}' \succ_{c_{2}}^{g_{2}} f_{2}
\succ_{c_{2}}^{g} : f_{g}' \succ_{c_{2}}^{g} f_{g}, \quad \forall g \in \{g_{1}, g_{2}\}$$
(7)

There are two possible allocations:

- if $\mu(f_1) = c_1$, then f_2 gets dynamic priority, and thus $\mu(f_2) = c_1$. However, student f'_2 has justified envy in school (c_2) given the initial preferences.
- if $\mu(f_2') = c_1$, then f_1' gets dynamic priority, and thus $\mu(f_1') = c_1$. However, student f_1 has justified envy in school (c_2) given the initial preferences.

As a result, we observe that there is no envy-free stable assignment.