

Ekkyo Matching: How to Integrate Fragmented Matching Markets for Welfare Improvement

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EKKYO MATCHING: HOW TO INTEGRATE FRAGMENTED MATCHING MARKETS FOR WELFARE IMPROVEMENT

YUICHIRO KAMADA AND FUHITO KOJIMA

ABSTRACT. We consider a school-choice matching model that allows for inter-regional transfer of students (*ekkyo* admission), with the “balancedness” constraint: each student and school belongs to a region, and a matching is said to be balanced if, for each region, the outflow of students from that region to other regions is equal to the inflow of students from the latter to the former. Using a directed bipartite graph defined on students and schools, we characterize the set of Pareto efficient matchings among those that are individually rational, balanced and fair. We also provide a class of polynomial-time algorithms to compute such matchings. The outcome of an algorithm from this class weakly improves student welfare upon the one produced when each region independently uses a deferred acceptance algorithm.

JEL Classification Numbers: C70, D47, D61, D63.

Keywords: ekkyo, balancedness, fairness, algorithm, efficiency, daycare allocation, school choice, matching with constraints

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1. INTRODUCTION

Matching theory has been applied in numerous real-life markets with a purpose of centralizing market transactions, but the centralized clearinghouses are still often organized at a (often small) local level. As a consequence of such fragmentation, some efficiency gains are being missed. In Japan, for instance, allocation of slots at accredited daycares are conducted by individual municipal governments and, with few exceptions, a child can only attend a daycare in the municipality of their residence. The City of Tokyo, for example, is divided into 23 small municipalities with the average size being about a half of Manhattan’s, and each conducts a matching independently.¹ Due to the small sizes of the regions, many families find that daycare centers outside the regions of their residence are among the closest. Moreover, it is often convenient to put their children to a daycare center close to their workplace while many people cross a regional boundary to commute.² For these reasons, inter-regional admission—which is called the *ekkyo* admission—is an appealing option for many families and it is indeed legal. However, *ekkyo* matching is rare under the current practice.

We study how to integrate fragmented matching markets. Specifically, we analyze mechanisms that improve upon mechanisms organized at the local level and achieve desirable fairness and efficiency properties. To do so, we depart from the standard model of matching between students and schools (Balinski and Sönmez, 1999; Abdulkadiroğlu and Sönmez, 2003) by assuming that each school belongs to exactly one region while each student is a resident of exactly one region. We consider a balancedness constraint that requires that, for each region, the number of residents of other regions who are matched to a school in it, called the inflow, must be equal to the number of its residents who are matched to a school in other regions, called the outflow.

Why is balancedness a reasonable requirement? In the context of Japanese daycare allocation, for instance, although *ekkyo* matching is legal—in fact, the Japanese government amended its Children and Childcare Act to encourage *ekkyo* matching in 2018—the law makes it clear that it is each municipality’s responsibility to provide daycare services.³ Therefore, it is impractical to organize a daycare allocation in a fully integrated manner across multiple municipalities. Moreover, each municipality heavily subsidizes daycares,

¹The average area of each of the 23 municipalities is 10.4 square miles, while Manhattan’s area is 22.7 square miles.

²To get a sense of the magnitude, over 3.5 million people get on or get off a train in Tokyo’s Shinjuku Station *each day*.

³Revision to Article 14(4) of the Supplementary Provision to Children and Childcare Act of Japan, <https://elaws.e-gov.go.jp/document?lawid=424AC0000000065>.

so enrolling residents from other municipalities can be a severe financial burden. Our balancedness constraint is meant to alleviate municipalities’ concerns by guaranteeing that no municipality carries an excessive burden. Given those considerations, we design mechanisms that achieve desirable properties under the balancedness constraint. In other words, *we aim at partial integration of the regions instead of full integration.*

We find that there does not always exist a balanced matching that satisfies stability, a standard desideratum in school choice literature that is equivalent to individual rationality, fairness, and non-wastefulness. Given this observation, we weaken our desideratum to only require individual rationality, balancedness and fairness. Among the matchings satisfying all the three conditions (there always exists such a matching⁴), which we call the *iBFs*, we focus on the ones that cannot be further improved upon in terms of students’ welfare, which we call *efficient iBFs*.

We first characterize efficient iBFs. To this end, we define a novel bipartite graph called a fair improvement graph (FIG henceforth) on a matching, where the vertices on one side represent the students and those on the other side represent the schools, and existence of an arrow between two vertices depends on preferences, priorities, and the given matching. We show that an iBF is an efficient iBF if and only if there exists no “FIG cycle,” a cycle on the FIG, for the matching.

Based on our characterization of efficient iBF, we then provide a polynomial-time algorithm that finds an efficient iBF. The algorithm is called the FIG cycles algorithm and is illustrated in some detail with an example in Section 1.1. Roughly, each step of the FIG cycles algorithm checks if the current iBF allows for a FIG cycle and, if so, finds a relocation of students that improves outcomes for students while retaining individual rationality, balancedness and fairness.

To better understand the markets with the balancedness constraint, we provide further discussions. First, we characterize when balancedness and stability are compatible with each other. Second, we examine strategic properties. Third, comparative statics are provided to evaluate the effect of merging and splitting regions. Fourth, we consider the case with weak priority, which often arises in school choice applications.

We would like to emphasize that our analysis is applicable beyond daycare allocation. Also in Japan, choice systems for public elementary and middle schools are organized at the small municipal level as well. Naturally, there exist much potential demand for enrolling in schools in other municipalities, but opportunities for interdistrict school choice

⁴All the conditions are satisfied by the matching where no student is matched to any school.

are severely limited.⁵ The issue is not limited to Japan either. In the U.S., for example, school choice is basically organized at a highly local level, but some form of interdistrict school choice is practiced in 43 States.⁶ The analysis of our paper, and particularly the FIG cycles algorithm, could be applied to improve efficiency of school choice mechanisms.

At a high level, the fragmented nature of resource allocation is widespread beyond day-care allocation or school choice. In kidney exchange in the U.S., for instance, individual transplant centers often conduct exchanges on their own before sending remaining participants to national exchange, resulting in significant efficiency loss (Agarwal et al., 2019). In COVID-19 vaccine allocation in Japan in 2021, individual municipalities were charged with vaccinating their respective residents, which resulted in situations in which vaccine stocks run out quickly in one municipality while extra stocks remain unused in another.⁷ We envision that research is called for to understand how to overcome inefficiency from the fragmented nature of the allocation problems in a practical manner when existing legal, institutional and other constraints prohibit full integration.

1.1. Illustrative Example. In this paper we introduce an algorithm of inter-regional transfer that improves students' welfare while respecting the balancedness condition and fairness. To gain intuition for why inter-regional transfer may improve welfare, consider the following simple environment (see Figure 1 for an illustration). There are two regions, r and r' . One school s_1 and two students i_1 and i_2 reside in r while one school s_2 and one student i_3 reside in r' . School s_1 has the capacity of two while school s_2 has the capacity

⁵For instance, Nerima (one of the 23 municipalities of Tokyo City) declares that it admits a non-resident child “only in truly unavoidable circumstances.” See Nerima’s website, <https://www.city.nerima.tokyo.jp/kosodatekyoiku/kyoiku/shochu/8jyo-9jyo-n1.html>, last accessed on September 8, 2023.

⁶See the website of the Education Commission of the States, <https://ecs.my.salesforce-sites.com/mbdata/mbquest4e?rep=OE1705>, last accessed on September 12, 2023.

⁷For instance, in an interview at the time with NHK, the national public broadcasting organization, an epidemiology expert pointed out that “adjustment of vaccine supply across municipalities is limited, and it is necessary to introduce a centralized system at a larger scale.” See NHK’s website, <https://www3.nhk.or.jp/news/html/20210922/k10013272411000.html>, last accessed on September 21, 2023.

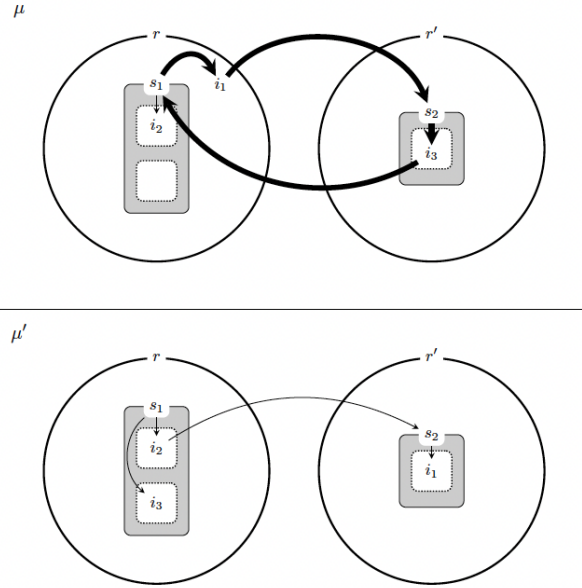


FIGURE 1. Illustrative Example. Each white dotted square in a school represents a seat in the school. The thick and thin arrows represent the FIG while the thick arrows represent a FIG cycle.

of one. Student preferences and school priorities are given as follows:

$$\begin{aligned}
 \gamma_{i_1} &: s_2, & \gamma_{s_1} &: i_1, i_2, i_3, \\
 \gamma_{i_2} &: s_2, s_1, & \gamma_{s_2} &: i_3, i_1, i_2. \\
 \gamma_{i_3} &: s_1, s_2,
 \end{aligned}$$

If an assignment of students to schools is determined region by region and there is no inter-regional transfer, the efficient matching for students is given by

$$\mu = \begin{pmatrix} s_1 & s_2 & \emptyset \\ i_2 & i_3, & i_1 \end{pmatrix},$$

which is realized by, for instance, running the student-proposing deferred acceptance algorithm in each region separately. However, with inter-regional transfer, a Pareto improvement for students is possible while respecting the balancedness condition. Specifically, if the two students i_1 and i_3 are sent to the schools in each other's regions, that is, if i_1 goes to s_2 and i_3 goes to s_1 , then in the realized matching,

$$\mu' = \begin{pmatrix} s_1 & s_2 & \emptyset \\ i_2, i_3 & i_1 & \emptyset \end{pmatrix},$$

these students are matched with their respective first-best school. In doing so, region r_1 takes in one student from outside (student i_3) while sending one student outside (student i_1), so the balancedness condition is satisfied.

Two things are noteworthy here. First, the number of students who are matched to some school is increased from μ to μ' . This is because, by swapping students between the two regions, the unmatched student i_1 was able to be matched with a school. We introduce an algorithm to make such an improvement possible.⁸

Second, there is another matching that respects the balancedness condition and Pareto dominates μ , which is

$$\mu'' = \begin{pmatrix} s_1 & s_2 & \emptyset \\ i_3 & i_2 & i_1 \end{pmatrix}.$$

However, this matching is not fair according to our definition, as i_1 is ranked higher than i_2 at s_2 . This suggests that care is needed about who can be moved to new schools across regions. The algorithm we introduce ensures that fairness is respected when an improvement is made.

We aim to achieve efficiency via inter-regional transfer like the one described in the above example. Specifically, we propose an algorithm that takes an arbitrary iBF as an input and achieves a Pareto improvement. The algorithm is based on a directed bipartite graph between students and schools that we call the *fair improvement graph* (or the *FIG*), and in each of its steps it “implements” a cycle in this graph—called a *FIG cycle*—, i.e., we move a student to a school that she points to. The outcome of repeatedly implementing FIG cycles turns out to be an *efficient iBF*. In fact, our main results characterize efficient iBF using cycles: we show that *an iBF is an efficient iBF if and only if there is no FIG cycle on it*.

Let us now discuss these results in the context of the aforementioned example. In our FIG, each school is pointed to by the top student (according to its priority) among those who strictly prefer the school to their current match and are acceptable to the school. This means that, under μ , s_1 is pointed to by i_3 while s_2 is pointed to by i_1 . Note that i_2 cannot point to s_2 as he is not the “top student” for s_2 (i_1 is). Also, each school points to the students that are matched to the school. So, s_1 points to i_2 and s_2 points to i_3 .

The above pointing rule is based on a standard idea in the previous literature that use cycles (e.g., Erdil and Ergin (2008) and Top Trading Cycles). However, in the present example, the graph constructed in this way does not have a cycle even though matching μ

⁸As we will explain in more detail later, this type of improvement is in a sharp contrast with most existing literature where the number of matched students is constant between before and after an improvement.

is not an efficient iBF. In order to achieve a Pareto improvement, we additionally require that each school with a vacancy points to any student matched to another school in the school's region as well as any unmatched students living in that region. In our example, this lets s_1 point to i_1 .⁹ With this, there is a cycle " $i_1 \rightarrow s_2 \rightarrow i_3 \rightarrow s_1 \rightarrow i_1$." Our characterization result shows that this implies μ is not an efficient iBF. Indeed, μ is not an efficient iBF because it is Pareto dominated by μ' , which is an iBF. In fact, the Pareto-improvement μ' is obtained by "implementing" this cycle. On μ' , the FIG now lets i_2 point to (only) s_2 and lets s_2 point to i_1 , but does not let i_1 point to any school (because i_1 is matched to her first choice school). Since i_3 does not point to any school either (because i_3 is matched to her first choice school), there is no FIG cycle on μ' . Our characterization result shows that this implies μ' is an efficient iBF. Indeed, one can verify that there is no iBF that Pareto dominates μ' .

The remainder of this paper proceeds as follows. Section 2 provides a model where we define efficient iBF. Section 3 introduces the fair improvement graph (FIG) and FIG cycles. Section 4 provides our main theorem, which characterizes efficient iBF by non-existence of FIG cycles. Section 5 defines the FIG cycles algorithm which outputs an efficient iBF. Section 6 provides various discussions, and Section 7 concludes. Proofs of all results are provided in the Appendix unless stated otherwise.

2. MODEL

2.1. Preliminary Definitions. Let there be a finite set of students I and a finite set of schools S . Each student i has a strict preference relation \succ_i over the set of schools and being unmatched (being unmatched is denoted by \emptyset). For any $s, s' \in S \cup \{\emptyset\}$, we write $s \succeq_i s'$ if and only if $s \succ_i s'$ or $s = s'$. Each school s has a strict priority order \succ_s over the set of students and leaving a position vacant (which is denoted by \emptyset).¹⁰ For any $i, i' \in I \cup \{\emptyset\}$, we write $i \succeq_s i'$ if and only if $i \succ_s i'$ or $i = i'$. Each school $s \in S$ is endowed with a (physical) **capacity** q_s , which is a nonnegative integer.

Student i is said to be **acceptable** to school s if $i \succ_s \emptyset$ (and unacceptable otherwise). Similarly, s is acceptable to i if $s \succ_i \emptyset$.¹¹ It will turn out that only rankings of acceptable

⁹In this example, the school with a vacancy points to an unmatched student. The benefit from requiring a school to point to a student matched to another school in the same region does not appear in the current example. We will explain this point in "Example 2, Continued" in Section 3.2.

¹⁰Strictness of priorities is assumed just for the sake of simplicity. In Section 6.4, we consider the case when indifferences are allowed and show that most results carry over to such a case.

¹¹In some applications, all schools may regard all students as acceptable. None of our results will hinge on the assumption that some students can be unacceptable to some schools.

partners matter for our analysis, so we often write only acceptable partners to denote preferences and priorities. For example,

$$\succ_i: s, s'$$

means that school s is the most preferred, s' is the second most preferred, and s and s' are the only acceptable schools under preferences \succ_i of student i .

A **matching** μ is a mapping that satisfies (i) $\mu_i \in S \cup \{\emptyset\}$ for all $i \in I$, (ii) $\mu_s \subseteq I$ and $|\mu_s| \leq q_s$ for all $s \in S$, and (iii) for any $i \in I$ and $s \in S$, $\mu_i = s$ if and only if $i \in \mu_s$. That is, a matching simply specifies which student is assigned to which school (if any).

A matching is **individually rational** if no student or school is matched with an unacceptable partner.

Given a matching μ , we say that a student i has **justified envy** to $j \in I$ if there is a school $s \in S$ such that (i) $\mu_j = s$, (ii) $s \succ_i \mu_i$, and (iii) $i \succ_s j$. We say that a matching μ is **fair** if there is no pair of students $(i, j) \in I^2$ such that i has justified envy to j .

Finally, a matching μ **weakly Pareto dominates** a matching μ' if $\mu_i \succeq_i \mu'_i$ for every $i \in I$. A matching μ **Pareto dominates** μ' if μ weakly Pareto dominates μ' and there exists $i \in I$ such that $\mu_i \succ_i \mu'_i$.

2.2. Regions, Balancedness, and Efficient iBF. There is a set of regions, denoted R , which is a partition of $I \cup S$. Formally, R satisfies the following conditions:

- (1) Each $r \in R$ is a nonempty subset of $I \cup S$.
- (2) $r \cap r' = \emptyset$ for any $r, r' \in R$ such that $r \neq r'$.
- (3) $\bigcup_{r \in R} r = I \cup S$.

The interpretation is that each s belongs to a single $r \in R$ and each i is a resident of a single $r \in R$. Let $r(s)$ be the region r such that $s \in r$, and similarly for $r(i)$.

We call $\mathcal{E} = (I, S, (\succ_a)_{a \in I \cup S}, (q_s)_{s \in S}, R)$ an **environment**.

We are now ready to introduce the key concept of this paper, “balancedness.”

Definition 1. A matching μ is **balanced** if for each $r \in R$,

$$(2.1) \quad \underbrace{\sum_{s \in r} |\{i | i \in \mu_s, i \notin r\}|}_{\text{inflow to } r} = \underbrace{\sum_{s \notin r} |\{i | i \in \mu_s, i \in r\}|}_{\text{outflow from } r}$$

As (2.1) shows, balancedness means that for any given region r , the inflow of students to r is the same as the outflow of students. Note that balancedness is not a “pairwise” notion, that is, it does not necessarily require that for every pair of regions r and r' , the number of students who live in r and are matched to a school in r' is the same as the

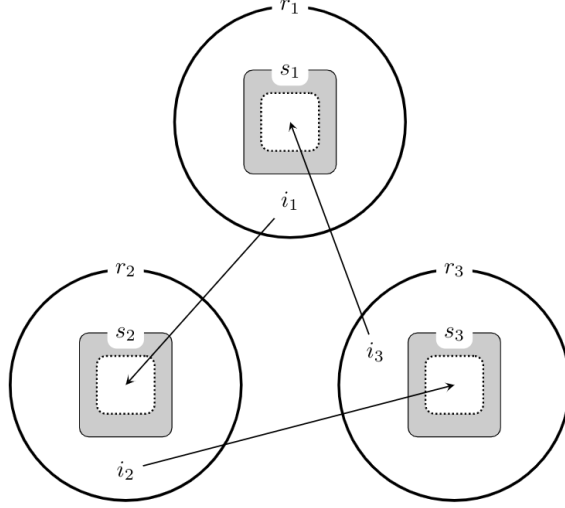


FIGURE 2. Three-way transfer (Example 1).

number of students who live in r' and are matched to a school in r . The next example illustrates.

Example 1 (Three-way transfer). Let $I = \{i_1, i_2, i_3\}$, $S = \{s_1, s_2, s_3\}$, $R = \{r_1, r_2, r_3\}$ and $r_k = \{i_k, s_k\}$ for each $k = 1, 2, 3$. Let

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_3 & i_1 & i_2 & \emptyset \end{pmatrix}.$$

See Figure 2 for a graphical representation. Note that under μ , the number of students who live in r_1 and are matched to a school in r_2 is 1 while the number of students who live in r_2 and are matched to a school in r_1 is 0. Matching μ is, however, balanced because it satisfies (2.1): For each region, the inflow and outflow are both 1. \square

We say that a matching is an **iBF** if it is individually rational, balanced, and fair. Below is the main solution concept of this paper.

Definition 2. A matching μ is said to be an **efficient iBF** if μ is an iBF and there is no iBF μ' that Pareto dominates μ .

In the standard environment without the balancedness constraint, there is a unique efficient iBF, which corresponds to a “student-optimal stable matching.” In our setting, there may be multiple efficient iBFs. The following example illustrates.

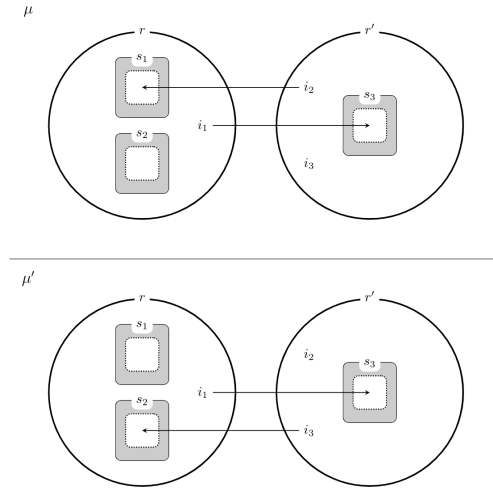


FIGURE 3. Multiple efficient iBFs (Example 2). Both μ and μ' are an efficient iBF.

Example 2 (Multiple efficient iBFs). Let $I = \{i_1, i_2, i_3\}$, $S = \{s_1, s_2, s_3\}$, $R = \{r, r'\}$ where $r = \{i_1, s_1, s_2\}$ and $r' = \{i_2, i_3, s_3\}$. Each school has the capacity of one. Student preferences and school priorities are given as follows:

$$\begin{aligned} \gamma_{i_1}: s_3, s_1, & & \gamma_{s_1}: i_2, i_1, \\ \gamma_{i_2}: s_1, & & \gamma_{s_2}: i_3, \\ \gamma_{i_3}: s_2, & & \gamma_{s_3}: i_1. \end{aligned}$$

Let

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_2 & \emptyset & i_1 & i_3 \end{pmatrix}, \quad \mu' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ \emptyset & i_3 & i_1 & i_2 \end{pmatrix}.$$

See Figure 3 for a graphical representation. We show that both μ and μ' are an efficient iBF. To see this, first notice that there are only four individually rational and balanced matchings. They are μ , μ' ,

$$\mu'' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_1 & \emptyset & \emptyset & i_2, i_3 \end{pmatrix},$$

and the empty matching (i.e., the matching where no student is matched to any school). The latter two matchings are Pareto dominated by μ and μ' while μ and μ' do not Pareto dominate each other, and one can verify by inspection that both μ and μ' are fair, and hence an efficient iBF. \square

Let us compare our notion of iBF with the standard notion of stability. Specifically, we say that a matching μ is **nonwasteful** if there is no pair (i, s) of a student and a school such that $s \succ_i \mu_i$, $i \succ_s \emptyset$, and $|\mu_s| < q_s$, and that μ is **stable** if it is individually rational, fair, and non-wasteful (but not necessarily balanced). A matching is the **student-optimal stable matching** if it is stable and is weakly preferred to every stable matching by all students. In the environment of Example 2, the unique stable matching (and hence the student-optimal stable matching) would be:

$$\mu''' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_2 & i_3 & i_1 & \emptyset \end{pmatrix}.$$

But μ''' is not balanced because the inflow to region r is 2 while the outflow from r is 1.

This example demonstrates that the balancedness condition is not necessarily compatible with stability. In Section 6.1, we characterize when those properties are compatible with each other.

3. FIG (FAIR IMPROVEMENT GRAPH) CYCLES

3.1. Definition of FIG Cycle. The key steps of our analysis involve defining bipartite directed graphs over the sets of students and schools, and identifying cycles on them. A bipartite directed graph on I and S , or simply a **graph**, $\mathcal{G} \subseteq (I \times S) \cup (S \times I)$, is a set of ordered pairs of agents in $I \cup S$. An interpretation is that if $(i, s) \in \mathcal{G}$, then there is an arrow pointing from i to s . In this case, we say “ i points to s .” The case of $(s, i) \in \mathcal{G}$ is analogous. Given a graph \mathcal{G} , a **cycle in \mathcal{G}** is any sequence of the form $(i_1, s_1, i_2, s_2, \dots, i_m, s_m)$ such that for each $k \in \{1, \dots, m\}$,

- (1) i_k points to s_k , i.e., $(i_k, s_k) \in \mathcal{G}$,
- (2) s_k points to i_{k+1} , i.e., $(s_k, i_{k+1}) \in \mathcal{G}$, where we set $i_{m+1} := i_1$,
- (3) $i_k \neq i_{k'}$ for all $k' \neq k$, and
- (4) $s_k \neq s_{k'}$ for all $k' \neq k$.

We will regard any two cycles as defined here, $(i_1, s_1, \dots, i_m, s_m)$ and $(i_{k+1}, s_{k+1}, \dots, i_m, s_m, i_1, s_1, \dots, i_k, s_k)$, as identical to each other.

Let $D_s^\mu := \{i \in I \mid s \succ_i \mu_i\}$, and $Top_s(I')$ be the student $i \in I'$ who has the highest priority among those in I' at \succ_s .

Now we define a particular type of a graph and a cycle on it. This cycle will be used to characterize efficient iBF as well as to define our algorithm.

Definition 3. Given a matching μ , the **fair improvement graph (FIG)** for μ is a graph such that, for any $i \in I$ and $s \in S$,

- (1) student $i \in I$ points to school $s \in S$ if $i = \text{Top}_s(D_s^\mu)$ and i is acceptable to s , and
- (2) school $s \in S$ points to student $i \in I$ if either
 - (a) $\mu_i = s$, or
 - (b) $|\mu_s| < q_s$ and, $[i \in r(s) \text{ and } \mu_i = \emptyset]$ or $\mu_i \in r(s)$.

A **fair improvement graph cycle (FIG cycle)** on μ is a cycle in the FIG for μ .

A student i points to a school s when she finds s to be better than her current outcome μ_i , and if she is acceptable to s and the best student for s among those who find s to be an improvement (so, under our assumption of strict priorities, each school can be pointed to by at most one student). In this sense, the students point to the schools in the most fair manner, and this is why we call the graph the “fair improvement graph.”

On the other hand, a school s can point to a student i in two different cases. The first case is as in other algorithms based on cycles in the literature such as the Top Trading Cycles (TTC) algorithm (Shapley and Scarf, 1974) and the stable improvement cycles algorithm (Erdil and Ergin, 2008). This is when i is currently matched to s , and it is described by part (2a) in Definition 3. The second case, described by part (2b) in Definition 3, depends on the notion of regions. School s can point to i if s has a vacancy and either i lives in the region of s and is unmatched, or i is matched to a school in the region. The need for this second case and the logic behind this particular pointing rule will become clear in a later example (Example 5 in Section 4).

Given a matching μ and a cycle of the form $\mathcal{F} = (i_1, s_1, i_2, s_2, \dots, i_m, s_m)$, call μ' the **matching generated by** (μ, \mathcal{F}) if

$$\mu'_{i_k} = s_k \text{ for each } k \in \{1, \dots, m\}, \text{ and } \mu'_j = \mu_j \text{ for all } j \in I \setminus \{i_1, \dots, i_m\}.$$

Given a matching μ and a cycle \mathcal{F} , we say that we **implement** \mathcal{F} on μ when we create the matching generated by (μ, \mathcal{F}) .

As we will show in Section 4, whether there exists a FIG cycle is crucial to the characterization of efficient iBF. Also, in the algorithm we define in Section 5, we repeatedly implement FIG cycles. But before stating the formal results that use the notion of FIG cycles, let us illustrate the concept of FIG cycle through a series of examples in the next subsection.

Remark 1. Erdil and Ergin (2017) consider a model with weak student preferences (and school priorities) and propose to improve students’ welfare by using *chains* in addition to cycles. A chain can start from a matched or unmatched student and ends at a school that has a vacant seat. We could define a chain in our model too, while restricting the pointing from schools to students to the one described in part (2a) of Definition 3.

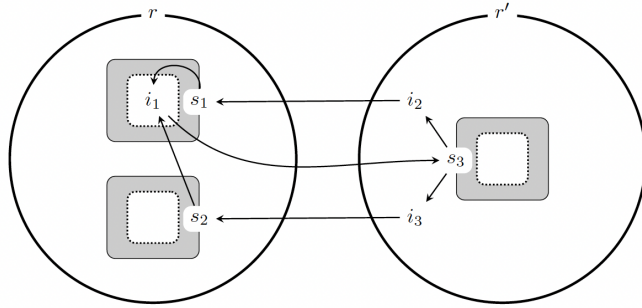


FIGURE 4. Example 2, Continued. The arrows represent the FIG. There are two FIG cycles.

Let us be forthcoming about the similarity and difference between their chains and our FIG cycles. First, in our model, implementing all chains and cycles under such a pointing rule would violate balancedness (there is no such issue in Erdil and Ergin (2017) as they have no balancedness constraint). For example, in Example 2, there would be one chain going out of r ($i_1 \rightarrow s_3$) while there would be two chains going into r ($i_2 \rightarrow s_1$ and $i_3 \rightarrow s_2$). As we will see in Theorem 1, balancedness is respected if we implement any FIG cycle. Second, one might argue that “connecting” chains might work. That is, we would start from a chain, and at the end of the chain (which is a school), we would find an unmatched student in the same region and see if there was a chain originating from that student. If there was such an arrow, then we would follow the arrows. Continuing this way, if an arrow eventually pointed to a school that had already appeared, then we would call the closed set of arrows a cycle. Cycles constructed in this way turns out to be the same as our FIG cycles. One can view that our pointing rule from schools to students, especially the part described in part (2b) of Definition 3, correctly captures how this “connecting” should be done. Third, the reasons behind why there are chains are different. In Erdil and Ergin (2017), a chain is implemented on a stable matching. For the existence of chains it is necessary that the student preferences are weak: if instead the students’ preferences are strict, then the last student on the chain would have strictly preferred to be matched to a vacant position in the last school in the chain under the original matching, so the original matching was not stable. In our model, there can exist a chain (defined in the absence of the pointing rule described in part (2b) of Definition 3) because of the balancedness constraint. \square

3.2. Examples of FIG Cycle.

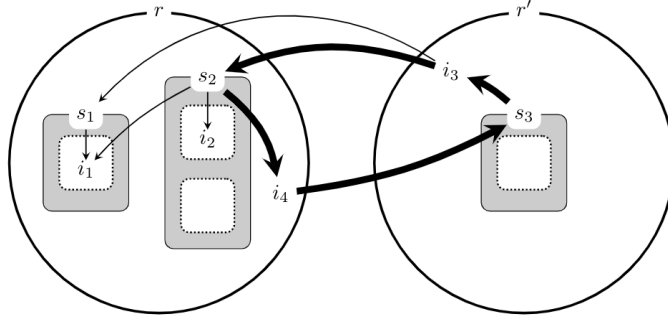


FIGURE 5. Example 3. The thick and thin arrows represent the FIG while the thick arrows represent a FIG cycle. School $s_2 \in r$ points to an unmatched student $i_4 \in r$ because s_2 's capacity is currently not full.

Example 2, Continued. Consider the same environment as in Example 2. The FIG for μ'' is drawn in Figure 4. Note that there are two FIG cycles:

$$\mathcal{F} := (i_1, s_3, i_2, s_1) \text{ and } \mathcal{F}' := (i_1, s_3, i_3, s_2).$$

Implementing the former cycle results in μ , and implementing the latter cycle results in μ' . Note that, in this example, it is important that the pointing rule for FIG lets a school point to a student matched to another school in the same region.¹² \square

Example 3 (Capacity matters for FIG cycle). Let $I = \{i_1, i_2, i_3, i_4\}$, $S = \{s_1, s_2, s_3\}$, $R = \{r, r'\}$ where $r = \{i_1, i_2, i_4, s_1, s_2\}$ and $r' = \{i_3, s_3\}$. Schools s_1 and s_3 have the capacity of one each while s_2 has the capacity of two. Student preferences and school priorities are given as follows:

$$\begin{aligned} \succ_{i_1} &: s_1, & \succ_{s_1} &: i_1, i_2, i_3, i_4, \\ \succ_{i_2} &: s_2, s_1, & \succ_{s_2} &: i_2, i_3, \\ \succ_{i_3} &: s_1, s_2, & \succ_{s_3} &: i_4. \\ \succ_{i_4} &: s_1, s_3, \end{aligned}$$

Consider the following matching:

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_1 & i_2 & \emptyset & i_3, i_4 \end{pmatrix},$$

which, by inspection, one can show to be fair. This matching as well as the FIG for μ is drawn in Figure 5. We note two features of this FIG. First, s_2 points to an unmatched

¹²This is the point alluded to in footnote 9.

student i_4 even though it is already matched with student i_2 . This is because s_2 's capacity is not filled under μ : the capacity is 2 while it is matched to only one student (student i_2). Second, there are multiple arrows from student i_3 . Although the rule for pointing in a FIG implies that each school is pointed to by at most one student, one student can point to multiple schools if she is the “number one” choice from multiple schools. In this example, i_3 ranks at the top among all students who prefer s_1 to their current match (i.e., among $\{i_3, i_4\}$), and she also ranks at the top among all students who prefer s_2 to their current match (i.e., among $\{i_3\}$). Note that there is a single FIG cycle:

$$\mathcal{F} := (i_3, s_2, i_4, s_3).$$

Implementing this cycle, we obtain the following matching:

$$\mu' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_1 & i_2, i_3 & i_4 & \emptyset \end{pmatrix},$$

which is an improvement over μ and is fair. One can show by inspection that μ' is also an efficient iBF. \square

4. CHARACTERIZATION OF EFFICIENT iBF

This is the main section of this paper, and we are going to formalize the following claims:

- (1) If we implement a FIG cycle on a given iBF μ , then it results in an iBF that Pareto dominates μ (Theorem 1).
- (2) If there is no FIG cycle on a given iBF μ , then μ is an efficient iBF (Theorem 2).

These results in particular imply the following characterization of efficient iBF: *Given an iBF, it is an efficient iBF if and only if there is no FIG cycle on it* (Corollary 1). In what follows, we will examine each claim and explain their intuition in detail.

Theorem 1. *Let μ be an iBF. If there exists a FIG cycle \mathcal{F} on μ , then a matching generated by (μ, \mathcal{F}) is an iBF and Pareto dominates μ .*

An implication of this theorem is that, if we can find a FIG cycle on a given iBF, then that matching cannot be an efficient iBF. Thus, one can think of this theorem as providing a necessary condition for an iBF to be an efficient iBF.

We note that the proof shows a stronger result: Implementing a FIG cycle on an arbitrary matching μ (which is not necessarily an iBF) results in a matching that is (i) Pareto superior to μ , (ii) individually rational if μ is individually rational, (iii) fair if μ is fair, and (iv) balanced if μ is balanced.

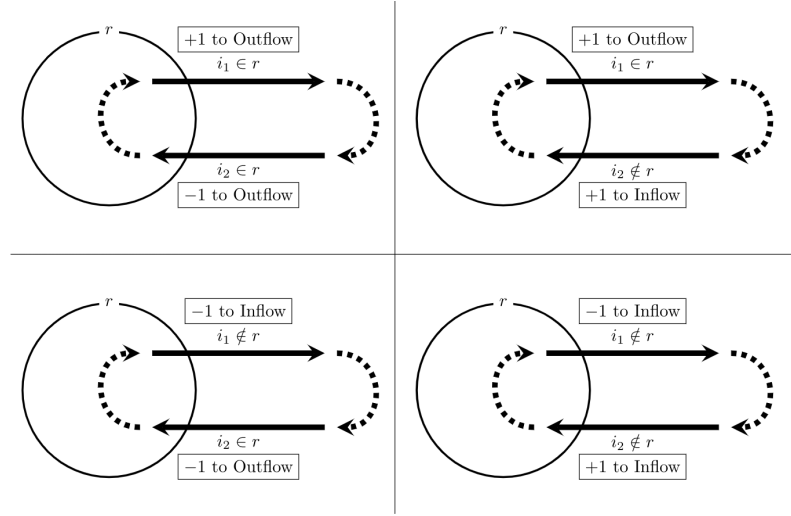


FIGURE 6. Why balancedness is maintained when a FIG cycle is implemented (proof intuition for Theorem 1).

Let us explain the intuition for the proof (of the stronger result above). For this, let μ be the original matching and μ' be the matching generated by (μ, \mathcal{F}) where \mathcal{F} is a FIG cycle on μ . The proof shows Pareto dominance, individual rationality, fairness and balancedness one by one. First, it is straightforward that μ' Pareto dominates μ by our pointing rule for students. Second, this observation also implies that μ' is individually rational if μ is individually rational.

Third, fairness of μ' is due to the pointing rule used in the FIG. A crucial step is to show that no student has justified envy to student i_k who is newly matched to s_k under μ' , where i_k and s_k appear in the FIG cycle (i.e., i_k points to s_k under the FIG).¹³ If some student i finds s_k to be better than her match under μ' , then she should have also found s_k to be better than her match under μ (because μ' Pareto dominates μ). But the pointing rule for FIG implies that i_k is the best student (according to s_k 's priority) among those who found s_k to be better than the match under μ . So, in particular, i is not higher than i_k under s_k 's priority. This implies that i cannot have justified envy to i_k under μ' .

Finally, balancedness of μ' holds because the FIG cycle is “closed,” that is, for any given region r , if an arrow from a student goes outside of r along the cycle, then another arrow from a student must come back to r and vice versa, which implies that the number of times the arrows go outside must be equal to the number of times the arrows come back to r . Whether balancedness is maintained by implementing a cycle may not be obvious

¹³Justified envy to other students matched to s_k or those that involve other schools can be shown not to exist by using fairness of μ and the fact that μ' Pareto dominates μ .

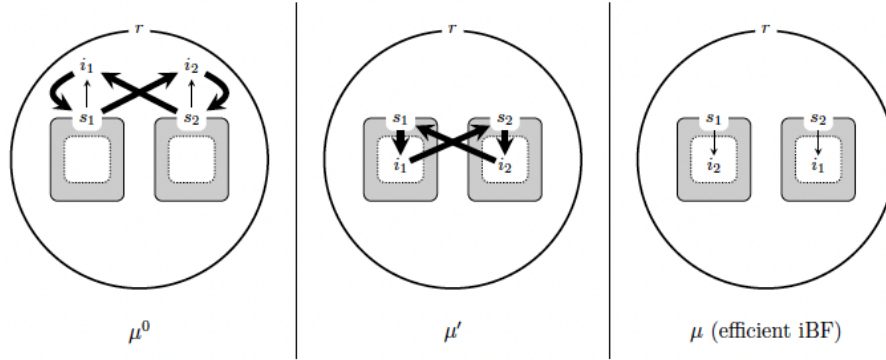


FIGURE 7. Example 4. The thick and thin arrows represent the FIG. μ is a unique efficient iBF, and there is no FIG cycle such that implementing it on μ^0 results in μ . Meanwhile, implementing a FIG cycle on μ^0 (indicated by thick arrows in the left panel of the figure) results in μ' , and implementing a FIG cycle on μ' (indicated by thick arrows in the middle panel of the figure) results in μ .

due to the fact that an outgoing arrow may carry a student who lives in r or one who does not live in r , and similarly an incoming arrow may carry either type of a student. Figure 6 shows that in every possible case, balancedness is maintained when a cycle is implemented.¹⁴ Feasibility would also be violated if an arrow that goes across regions can originate from a school, a possibility that the definition of FIG prohibits. Example 5 illustrates this point.

Two remarks are in order. First, we note that Theorem 1 does not assert that implementing a FIG cycle on a balanced and fair matching necessarily results in an efficient iBF. Indeed, the following example presents a case in which one needs to implement FIG cycles more than once to reach an efficient iBF.

Example 4 (Multiple rounds of FIG cycles). Let $I = \{i_1, i_2\}$, $S = \{s_1, s_2\}$, $R = \{r\}$. Each school has the capacity of one. Student preferences and school priorities are given

¹⁴Hafalir, Kojima and Yenmez (2022) study TTC under a variety of constraints, one of which is the balancedness constraints of the present paper. They verify that the balancedness constraints satisfy a condition called M-concavity, which Suzuki et al. (2023) showed is sufficient for the outcome of a certain version of TTC to satisfy balancedness. Although their TTC algorithm is substantially different from ours, one might also be able to use a similar indirect approach to establish balancedness of the outcome of our algorithm.

as follows:

$$\begin{array}{ll} \succ_{i_1}: s_2, s_1, & \succ_{s_1}: i_1, i_2, \\ \succ_{i_2}: s_1, s_2, & \succ_{s_2}: i_2, i_1. \end{array}$$

Let μ^0 be the empty matching. Starting from μ^0 , there are three FIG cycles:

$$\mathcal{F} := (i_1, s_1), \mathcal{F}' := (i_2, s_2), \text{ and } \mathcal{F}'' := (i_1, s_1, i_2, s_2).$$

It is straightforward to see that there is a unique efficient iBF, which is

$$\mu = \begin{pmatrix} s_1 & s_2 & \emptyset \\ i_2 & i_1 & \emptyset \end{pmatrix}.$$

However, μ is not a matching generated by $(\mu^0, \tilde{\mathcal{F}})$ where $\tilde{\mathcal{F}}$ is either \mathcal{F} , \mathcal{F}' or \mathcal{F}'' . The reason is that, for example, student i_1 does not point to s_2 in the FIG for μ^0 because she is not the top choice of s_2 among those who point to s_2 (s_2 's top choice is i_2).

The efficient iBF can be obtained by “two rounds of FIG cycles” starting from μ^0 : First, implementing the FIG cycle \mathcal{F}'' on μ^0 , we obtain:

$$\mu' = \begin{pmatrix} s_1 & s_2 & \emptyset \\ i_1 & i_2 & \emptyset \end{pmatrix}.$$

Second, implementing

$$\mathcal{F}''' := (i_1, s_2, i_2, s_1)$$

on μ' results in μ . □

The second remark is about the pointing rule for FIG. Note that Theorem 1 asserts that, among other things, implementing a FIG cycle on an iBF results in a balanced matching. This property depends on a somewhat subtle manner in which we define the FIG, as the following example illustrates.

Example 5 (Pointing rule for FIG). Let $I = \{i_1, i_2\}$, $S = \{s_1, s_2, s_3\}$, $R = \{r, r'\}$ where $r = \{i_1, s_1, s_2\}$ and $r' = \{i_2, s_3\}$. Each school has the capacity of one. Student preferences and school priorities are given as follows:

$$\begin{array}{ll} \succ_{i_1}: s_2, s_3, s_1, & \succ_{s_1}: i_1, i_2, \\ \succ_{i_2}: s_1, s_2, & \succ_{s_2}: i_2, i_1, \\ & \succ_{s_3}: i_1. \end{array}$$

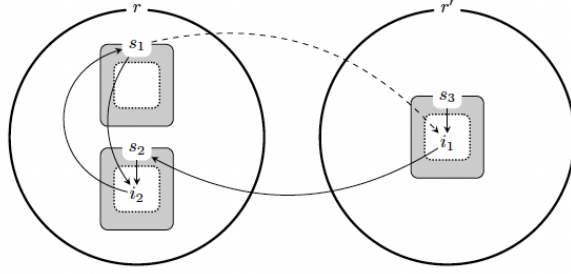


FIGURE 8. Example 5. The solid arrows represent the FIG. Student i_1 lives in r and student i_2 lives in r' . If we allowed s_1 to point to i_1 (the dashed arrow) in the FIG and implemented the resulting cycle (i_2, s_1, i_1, s_2) , the balancedness constraint would be violated.

Consider the following matching:

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ \emptyset & i_2 & i_1 & \emptyset \end{pmatrix}.$$

Note that this matching is a matching generated by (μ^0, \mathcal{F}) where μ^0 denotes the empty matching and we let $\mathcal{F} := (i_1, s_3, i_2, s_2)$. The cycle \mathcal{F} is a FIG cycle and μ is fair.

Now, consider the FIG on μ , which is depicted in Figure 8 by solid arrows. Suppose that we modify case (2b) of the definition of FIG (Definition 3) and hypothetically allow each school with a vacancy to point to *any* student that lives in the same region, irrespective of where that student is currently matched. This means that s_1 points to i_1 , although it is *not* allowed in the FIG according to its pointing rule. This additional pointing is depicted by a dotted arrow in Figure 8. With this arrow, there is a cycle $\mathcal{F}' := (i_1, s_2, i_2, s_1)$. However, the matching generated by (μ, \mathcal{F}') violates the balancedness constraint. The reason is that implementing this cycle on μ would result in one less outflow for region r while the inflow does not change. This happens because the arrow going out of region r starts from a school, not a student. In the definition of the FIG, all arrows going out of regions are from students. As we explained when providing the intuition for the proof of balancedness in Theorem 1 (see Figure 6), this feature ensures that implementing any FIG cycle on any balanced matching results in a balanced matching. \square

Theorem 2. *Let μ be an iBF. μ is an efficient iBF if there exists no FIG cycle on μ .*

Recall that Theorem 1 identifies a necessary condition for a given iBF to be an efficient iBF. Theorem 2 establishes that the same condition is in fact sufficient as well. Therefore, as we will formally state later in Corollary 1, combining those results provides a characterization of an efficient iBF.

The proof is by contraposition. That is, we take μ that is an iBF and assume that there is another iBF μ' that Pareto dominates μ . Then we show that there exists a FIG cycle on μ .

To find a FIG cycle, we construct a graph. In this graph, the students who are associated with arrows are those whose outcomes are different between μ and μ' . We denote by I' the set of those students. Meanwhile, the schools with arrows are the ones that are matched to students in I' under μ' (which means that these are the schools that are matched to some new students under μ' relative to μ). We denote by $S(\mu')$ the set of those schools. Also, let $S(\mu)$ be the set of schools that students in I' are matched with under μ .

To understand the complication in the proof, consider first a simple case: there is a single region r (i.e., a standard environment without a balancedness constraint), and μ is a stable matching (in the standard sense). Then, since stability implies non-wastefulness and μ' Pareto dominates μ , it follows that $S(\mu') = S(\mu)$. We can have each school s point to students in $\mu_s \cap I'$, and allow a student in I' point to s if she is the top student among I' who regard s to be an improvement relative to μ . This way, each school is pointed to by one student, and each student is pointed to by one school. Thus, there is a cycle, and with some work, one can prove that such a cycle must be a FIG cycle. This is essentially the same method as what is used in Erdil and Ergin (2008).¹⁵

In our problem, complication arises for two reasons. First, fairness alone does not imply non-wastefulness, so μ may have some waste (for example, consider the empty matching μ^0 in Example 4). This means that $S(\mu')$ may not be equal to $S(\mu)$, so a graph constructed in the above manner might not have a cycle (schools in $S(\mu') \setminus S(\mu)$ would have no outgoing arrow, and those in $S(\mu) \setminus S(\mu')$ would have no incoming arrow). This suggests we need an alternative way of forming a graph. Second, arrows might go in and out of any region, hence in defining the alternative graph, we must make sure that the balancedness constraint would be respected when implementing a cycle in the graph.

We overcome these difficulties by constructing a graph, denoted $\mathcal{G}(\mu, \mu')$, in the following manner. First, each school in $S(\mu')$ is pointed to by the top student among those in I' who

¹⁵We note that Erdil and Ergin (2008) allow school priorities to be weak, while we assume strict preferences here. However, as explained in Section 6.4, our analysis extends to the case with weak priorities without any significant change.

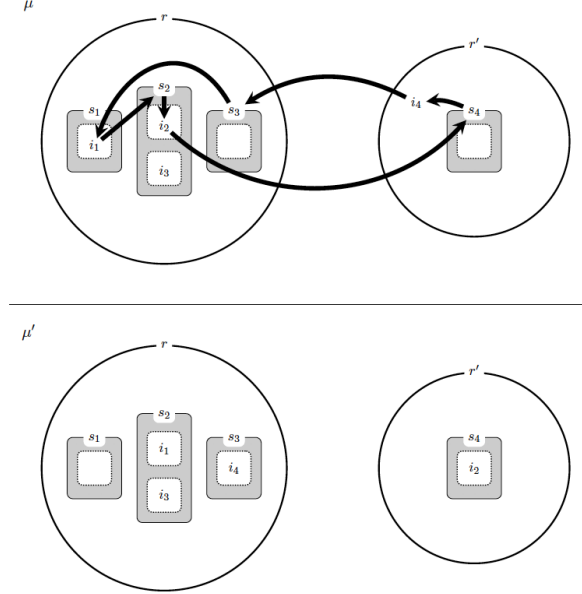


FIGURE 9. Example 6. Given an iBF μ and a Pareto dominating iBF μ' , the proof of Theorem 2 constructs a cycle (represented by the thick arrows) and shows that the constructed cycle is in fact a FIG cycle.

regard the school to be an improvement relative to μ , just as in the “simple case” above. Second, each student i in I' is pointed to by a school in the following three different ways depending on i 's outcome under μ :

- (1) If $\mu_i \in S(\mu')$, then we let μ_i point to i . This is the case that is analogous to the “simple case” explained above.
- (2) If $\mu_i \in S \setminus S(\mu')$, then we can show, using individual rationality and balancedness, that we can find a school that resides in μ_i 's region and belongs to $S(\mu')$ whose capacity is not filled under μ .¹⁶ We let such a school point to i .
- (3) If $\mu_i = \emptyset$, then we can show, again using individual rationality and balancedness, that we can find a school that resides in i 's region and belongs to $S(\mu')$ whose capacity is not filled under μ . We let such a school point to i .

Since each school is pointed to by one student, and each student is pointed to by one school, $\mathcal{G}(\mu, \mu')$ has a cycle. The proof shows that any cycle in this graph is a FIG cycle.

The next example illustrates how our construction works in a specific environment.

¹⁶Showing the existence of such a school is nontrivial. The proof constructs an additional graph for each region and uses that graph to establish the existence. See the Appendix for detail.

Example 6 (Construction of a FIG cycle). Let $I = \{i_1, i_2, i_3, i_4\}$, $S = \{s_1, s_2, s_3, s_4\}$, $R = \{r, r'\}$ where $r = \{i_1, i_2, i_3, s_1, s_2, s_3\}$ and $r' = \{i_4, s_4\}$. School s_2 has the capacity of two while all other schools have the capacity of one. Student preferences and school priorities are given as follows:

$$\begin{array}{ll} \succ_{i_1}: s_2, s_1, & \succ_{s_1}: i_1, \\ \succ_{i_2}: s_4, s_2, & \succ_{s_2}: i_2, i_3, i_1, i_4, \\ \succ_{i_3}: s_2, & \succ_{s_3}: i_4, \\ \succ_{i_4}: s_2, s_3, & \succ_{s_4}: i_2. \end{array}$$

Consider the following matchings:

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & s_4 & \emptyset \\ i_1 & i_2, i_3 & \emptyset & \emptyset & i_4 \end{pmatrix}, \quad \mu' = \begin{pmatrix} s_1 & s_2 & s_3 & s_4 & \emptyset \\ \emptyset & i_1, i_3 & i_4 & i_2 & \emptyset \end{pmatrix}.$$

They are both an iBF, and μ' Pareto dominates μ .

Let us explain how to obtain the graph $\mathcal{G}(\mu, \mu')$ and a FIG cycle in this environment (see Figure 9 for a graphical illustration). First, note that $I' = \{i_1, i_2, i_4\}$, $S(\mu) = \{s_1, s_2\}$, and $S(\mu') = \{s_2, s_3, s_4\}$. Recall that a student in I' points to a school in $S(\mu')$ if and only if she is the top student according to the school's priority among those who prefer the school to their current match. Hence, the edges originating at a student in $\mathcal{G}(\mu, \mu')$ are (i_1, s_2) , (i_2, s_4) , and (i_4, s_3) . Next, we illustrate how a student is pointed to by a school.

- (1) $\mu_{i_1} = s_1 \in S \setminus S(\mu')$. This is case 2 of the aforementioned pointing rule from a school to a student. School s_3 is the only school that resides in μ_{i_1} 's region r and belongs to $S(\mu')$ whose capacity is not filled under μ . Thus, (only) s_3 points to i_1 .
- (2) $\mu_{i_2} = s_2 \in S(\mu')$. This is case 1 of the pointing rule, and thus s_2 points to i_2 .
- (3) $\mu_{i_4} = \emptyset$. This is case 3 of the pointing rule. School s_4 is the only school that resides in i_4 's region r' and belongs to $S(\mu')$ whose capacity is not filled under μ . Thus, (only) s_4 points to i_4 .

Overall, the graph $\mathcal{G}(\mu, \mu')$ can be drawn as in Figure 9. There is a unique cycle on $G(\mu, \mu')$, which is $(i_1, s_2, i_2, s_4, i_4, s_3)$. One can check that this is a FIG cycle as well.

We note a special feature of this example: Given an iBF μ and a Pareto dominating iBF μ' , implementing a FIG cycle on μ constructed in the proof results in μ' . This feature turns out not to hold generally. This fact can be seen in Example 4, where two steps of FIG cycle implementations may be necessary to move between two iBFs. \square

Theorem 1 and Theorem 2 together imply the following characterization of efficient iBF

Corollary 1. *Let μ be an iBF. μ is an efficient iBF if and only if there exists no FIG cycle on μ .*

5. FIG CYCLES ALGORITHM

Let $\tilde{\mu}$ be an arbitrary iBF, e.g., the empty matching. Building on Corollary 1, we define a **FIG cycles algorithm** on $\tilde{\mu}$ as follows.

FIG Cycles Algorithm on $\tilde{\mu}$:

Step 0: Let $\mu^0 = \tilde{\mu}$ and move to Step 1.

Step l ($l \geq 1$): If there is no FIG cycle on μ^{l-1} , terminate the algorithm and output μ^{l-1} . Otherwise, choose a FIG cycle \mathcal{F} on μ^{l-1} arbitrarily and let μ^l be the matching generated by (μ^{l-1}, \mathcal{F}) , and go to Step $l + 1$.

Corollary 2. *Let $\tilde{\mu}$ be an arbitrary iBF. The FIG cycles algorithm on $\tilde{\mu}$ runs in polynomial time, and its output is an efficient iBF.*

Proof. Take an arbitrary iBF and denote it by $\tilde{\mu}$. Theorems 1 and 2 together show that the output of the FIG cycles algorithm on $\tilde{\mu}$ is an efficient iBF. To show that the algorithm runs in polynomial time, first note that each student can only become better off while running the algorithm, and at least one student must be made strictly better off at each step as long as the algorithm does not terminate in that step. Therefore, at most $|I| \times |S|$ steps are necessary for terminating the algorithm. Second, within each step, finding a FIG cycle can be done in polynomial time.¹⁷ These two observations show that the algorithm runs in polynomial time, as desired. \square

In some applications, a school may give a higher priority to residents of that school's region than non-residents. Let us now consider such a case. Formally, assume that $i \succ_s j$ for each $r \in R$, $s \in r$, $i \in r$, and $j \notin r$.

We consider a special fair matching, namely the matching that is produced by running the standard deferred acceptance algorithm of Gale and Shapley (1962), separately in each region. Formally, it is defined as follows.

Definition 4. A **region-wise student-optimal stable matching** μ^{RW} is a matching that satisfies the following:

- (1) For each $r \in R$ and each $i \in r$, we have $\mu_i^{RW} \in r \cup \{\emptyset\}$.
- (2) μ^{RW} is individually rational.

¹⁷There are well-known polynomial-time algorithms that identify a cycle if one exists and otherwise show that there is no cycle. The “depth-first search” algorithm, for example, has the running time of $O(|I| \times |S|)$ (Cormen et al., 2001).

- (3) For each $r \in R$, there is no pair of a student and a school $i, s \in r$ such that $s \succ_i \mu_i^{RW}$, $i \succ_s \emptyset$, and $|\mu_s^{RW}| < q_s$.
- (4) For each $r \in R$, there is no pair of students $i, i' \in r$ such that i has justified envy to i' under μ^{RW} .
- (5) μ^{RW} weakly Pareto dominates all matchings that satisfy conditions (1), (2), (3) and (4).

Intuitively, a region-wise student-optimal stable matching requires that a matching restricted to each region (i.e., consider the students and schools that reside in that region) is a student-optimal stable matching in the standard sense (Conditions (2), (3), and (4) correspond to individual rationality, non-wastefulness, and fairness within each region). By Gale and Shapley (1962), a region-wise student-optimal stable matching always exists, and it can be obtained by their deferred acceptance algorithm in polynomial time.

We note that μ^{RW} is an iBF. To see this, first note that μ^{RW} is individually rational (Condition (2)). Second, it is balanced because no student is matched to a school outside of her region (Condition (1)), and thus both the inflow and outflow for any given region is equal to zero. Third, μ^{RW} is fair by Condition (4) and the assumption in this section that schools rank students from the same region higher than those from other regions (so a student living in r cannot have justified envy to another student living in r' if the latter student is matched to a school in r').¹⁸

The next corollary considers the case when we run a FIG cycles algorithm starting from this matching.

Corollary 3. *The output of a FIG cycles algorithm starting from the region-wise student-optimal stable matching μ^{RW} is an efficient iBF and weakly Pareto dominates μ^{RW} .*

Proof. Let μ^{RW} be the region-wise student-optimal stable matching. Note that it is an iBF as we have explained after providing the statement of Definition 4. Hence, Corollary 2 implies that the output of a FIG cycles algorithm starting from μ^{RW} is an efficient iBF. Finally, Theorem 1 and the definition of FIG cycles algorithm imply that the output Pareto dominates μ^{RW} . \square

¹⁸We wrote Definition 4 following the most standard way, but some conditions and qualifiers are redundant because condition (5) holds and, as we explained, μ^{RW} is fair. A simplified (equivalent) definition requires the following three conditions:

- (1) For each $r \in R$ and each $i \in r$, we have $\mu_i^{RW} \in r \cup \{\emptyset\}$.
- (4') μ^{RW} is fair.
- (5') μ^{RW} weakly Pareto dominates all matchings that satisfy conditions (1) and (4').

6. DISCUSSIONS

This section provides a number of discussions. In Section 6.1, we characterize when balancedness and stability are compatible with each other. Section 6.2 examines the strategic properties. In Section 6.3, we provide comparative statics to evaluate the effect of merging and splitting regions. Section 6.4 considers the case with weak priority, which often arises in school choice applications. In Section 6.5, we review the related literature.

6.1. Balancedness and Stability. As discussed in Section 2.2, Example 2 demonstrates that the balancedness condition is not necessarily compatible with stability. This observation leads to the question of when those properties are compatible with each other. The following result provides a characterization and establishes, in particular, that those cases exactly coincides with cases in which the unique efficient iBF is the student-optimal stable matching.

Theorem 3. *The following statements are equivalent.*

- (1) *There exists a unique efficient iBF, and it coincides with the student-optimal stable matching.*
- (2) *There exists a stable and balanced matching.*
- (3) *Every stable matching is balanced.*

Proof. (1) \rightarrow (2): Obvious.

(2) \rightarrow (3): Let μ and μ' be two stable matchings. Define a graph as follows. The set of nodes are I and S . We let each $s \in S$ point to $i \in I$ if $\mu_i = s$ and let each $i \in I$ point $s \in S$ if $\mu'_i = s$. By the rural hospital theorem (Roth, 1986), for each school s , the number of arrows that point to s is equal to the number of arrows that point from s and, for each student i , the number of arrows that point to i is equal to the number of arrows that point from i . Therefore, the graph is partitioned into sets of edges that correspond to a finite number of disjoint cycles such that μ' is generated by implementing all those cycles on μ .¹⁹ Thus, by an argument analogous to the proof of Theorem 1, if μ is balanced, then μ' is balanced as well.

(3) \rightarrow (1): Suppose that every stable matching is balanced. We first observe that the student-optimal stable matching μ^* is stable and hence balanced. By Wu and Roth

¹⁹Formally, let $\mathcal{F}_1, \dots, \mathcal{F}_k$ be the set of cycles. Let $\mu^0 := \mu$ and μ^ℓ be generated by $(\mu^{\ell-1}, \mathcal{F}_\ell)$ for each $\ell = 1, \dots, k$. Then it follows that $\mu' = \mu^k$.

(2018), for any matching μ that is individually rational and fair, $\mu_i^* \succeq_i \mu_i$ for all $i \in I$. Therefore, μ^* is a unique efficient iBF. \square

6.2. Strategic Property. In this section, we investigate strategic properties. It turns out that our mechanism based on the FIG cycles algorithm is not strategy-proof. In fact, we show that there is no strategy-proof mechanism that always outputs an efficient iBF.

Formally, a mechanism φ is a function from the set of student preference profiles to the set of matchings. Mechanism φ is **strategy-proof** if

$$\varphi_i(\succ) \succeq_i \varphi_i(\succ'_i, \succ_{-i}),$$

for every student preference profile \succ , $i \in I$, and student preference \succ'_i .

Theorem 4. *There exists no strategy-proof mechanism that outputs an efficient iBF for all preference profiles.*

Proof. We prove the result by presenting an example. Let $I = \{i_1, i_2, i_3\}$ and $S = \{s_1, s_2, s_3\}$. Let there be two regions, $r = \{i_1, i_2, s_1\}$ and $r' = \{i_3, s_2, s_3\}$. Each school has the capacity of one. Student preferences and school priorities are given as follows:

$$\begin{array}{ll} \succ_{i_1}: s_2, & \succ_{s_1}: i_1, i_2, i_3, \\ \succ_{i_2}: s_3, & \succ_{s_2}: i_3, i_2, i_1, \\ \succ_{i_3}: s_1, & \succ_{s_3}: i_3, i_1, i_2. \end{array}$$

In this environment, there are two efficient iBFs:

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_3 & i_1 & \emptyset & i_2 \end{pmatrix}, \quad \mu' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_3 & \emptyset & i_2 & i_1 \end{pmatrix}.$$

Fix a mechanism φ that outputs an efficient iBF for all preference profiles. It must be either $\varphi(\succ) = \mu$ or $\varphi(\succ) = \mu'$.

Suppose $\varphi(\succ) = \mu$. Then, consider $\succ'_{i_2}: s_3, s_2$. The unique efficient iBF at $\succ' := (\succ'_{i_2}, \succ_{-i_2})$ is μ' , so $\varphi(\succ') = \mu'$. Noting that $\mu'_{i_2} = s_3 \succ_{i_2} \emptyset = \mu_{i_2}$, we have obtained that $\varphi_{i_2}(\succ) \not\succeq_{i_2} \varphi_{i_2}(\succ'_{i_2}, \succ_{-i_2})$, violating strategy-proofness.

Next, suppose $\varphi(\succ) = \mu'$. Then, by considering $\succ'_{i_1}: s_2, s_3$ and following a symmetric argument, we conclude that φ is not strategy-proof. This completes the proof. \square

Theorem 4 shows a sense in which the lack of strategy-proofness is not a drawback of the FIG cycles algorithm. Rather, it is inevitable in that any mechanism that outputs an efficient iBF necessarily fails to be strategy-proof.

6.3. Comparative Statics. Intuition suggests that merging regions will lead to a more efficient matching while splitting regions will have the opposite effect. To explore this intuition, first consider the case where all regions were merged, that is, the union of all regions in the original model is regarded as a single region after the mergers. In this case, there would be a unique efficient iBF, which coincides with the student-optimal stable matching. This matching weakly Pareto dominates any individually rational and fair matching, and hence any efficient iBF under any structure of regions. However, as we have argued in the Introduction, merging all regions may be impractical. A question of interest is whether mergers generally result in a more efficient matching even if there remain multiple regions after the mergers.

We say that an environment $\mathcal{E} = (I, S, (\succ_a)_{a \in I \cup S}, (q_s)_{s \in S}, R)$ is a **result of mergers from** another environment $\mathcal{E}' = (I', S', (\succ'_a)_{a \in I' \cup S'}, (q'_s)_{s \in S'}, R')$ if $I = I', S = S', \succ_a = \succ'_a$ for every $a \in I \cup S$, $q_s = q'_s$ for every $s \in S$ and, for each $r \in R$, r is a union of (possibly one) regions of R' . That is, some regions in \mathcal{E}' merge to form a region in \mathcal{E} , but otherwise all the primitives are unchanged between the two environments.

In the following simple example, merging regions Pareto-improves the outcomes for students.

Example 7 (An instance in which merging regions makes every student better off). Let $I = \{i\}$ and $S = \{s\}$. Let there be two regions, $r = \{i\}$, and $r' = \{s\}$.²⁰ School s has the capacity of one. Student i finds school s to be acceptable and school s finds student i to be acceptable. In this environment, there is a unique efficient iBF:

$$\mu = \begin{pmatrix} s & \emptyset \\ \emptyset & i \end{pmatrix}.$$

If regions r_1 and r_2 are merged, then there is a unique efficient iBF:

$$\mu' = \begin{pmatrix} s & \emptyset \\ i & \emptyset \end{pmatrix}.$$

Since $\mu'_i \succ_i \mu_i$ and i is the only student in the market, this means that merging the regions made every student strictly better off in this example. The intuition is simple: The merge reduced the constraint of balanced exchange between regions r and r' , so after the merger, the student i can go to school s .

It turns out that the finding of this example is not a coincidence. Indeed, the next proposition establishes a comparative statics regarding any mergers.

²⁰In this example, region r does not have a school and region r' does not have a student. These features are not necessary to make our point.

Proposition 1. *Suppose that \mathcal{E} is a result of mergers from \mathcal{E}' . Then, for any matching μ' that is an efficient iBF at \mathcal{E}' , there exists a matching μ that is an efficient iBF at \mathcal{E} such that μ weakly Pareto dominates μ' .*

Proof. We first show that μ' is an iBF at \mathcal{E} . Individual rationality and fairness are obvious. The following claim is proved in Appendix A.3:

Claim 1. *μ' is balanced at \mathcal{E} .*

The preceding arguments have established that μ' is an iBF at \mathcal{E} . Therefore, by definition of an efficient iBF, there exists a matching μ that is an efficient iBF at \mathcal{E} such that μ weakly Pareto dominates μ' . \square

We note that a kind of converse of Proposition 1 does not hold. Specifically, Example 8 in Appendix A.4 presents an instance in which, starting at a certain efficient iBF, some student is made strictly better off in any efficient iBF after splitting a region.

To summarize, the analysis in this section confirmed the intuition that merging regions improves student welfare. We also found that, however, a precise sense in which this intuition goes through is somewhat subtle.

6.4. Weak Priorities. In applications such as daycare allocations and school choice, schools are sometimes endowed with weak priorities. Erdil and Ergin (2008) consider weak priorities and propose an algorithm based on cycles to improve upon the deferred acceptance algorithm with tie-breaking, albeit in a setting without our balancedness constraints. Accordingly, a natural question would be whether our analysis extends to cases where priorities are allowed to be weak in the presence of balancedness constraints. As it turns out, all of our results go through. In particular, the conclusions of Theorems 1 and 2 hold without any change.²¹

Here we illustrate how the proofs of Theorems 1 and 2 change with weak priorities. The only difference from the case of strict priorities is the following. In the FIG or $\mathcal{G}(\mu, \mu')$ (a graph which appears in the proof of Theorem 2), each school can be pointed to by at most one student. This was because a student can point to a school only if she is the top student among those who regard the school as an improvement. Under strict priorities, there is a unique “top” student, which was why each school can be pointed to by at most one student. Under weak priorities, however, there can be multiple “top” students, so

²¹Under weak priority, we say that a student is acceptable to a school if she is ranked *weakly* higher than the outside option, and modify the definitions of individual rationality, efficient iBF and FIG by adopting this definition of acceptability in the relevant parts of those concepts. The proofs change accordingly in a straightforward manner.

a school can be pointed to by multiple students. But this change does not affect the remainder of the proof.

6.5. Related Literature. The aim of our study is to improve efficiency in situations in which matching is organized at a small local level. Sharing this interest with the present paper, Hafalir, Kojima and Yenmez (2022) study a variety of constraints in the context of interdistrict school choice, and one of the constraints studied there is the balancedness constraint. Differently from our analysis, they study conditions under which the outcome of the standard deferred acceptance algorithm satisfies the constraint, which turns out to be very restrictive. The present paper offers a way to improve upon matching mechanisms organized separately in each district even if the deferred acceptance algorithm does not produce a balanced matching. To our knowledge, the balancedness condition was introduced and studied first by Dur and Ünver (2019), although the setting of tuition exchange they study is different from ours or Hafalir, Kojima and Yenmez (2022) in several important respects, making their analysis and ours logically unrelated.

Our algorithm is based on a number of Pareto-improving cycles among students and schools. At a high level, this is a common idea and is shared by many other algorithms, including Gale’s celebrated TTC algorithm in Shapley and Scarf (1974). The difference of our algorithm from TTC is that we construct cycles in a more subtle and nuanced manner, taking school priorities into account in particular, so that implementing the cycles will keep fairness of the original matching. Closer to our algorithm are those due to Erdil and Ergin (2008, 2017) who, like us, provide iterative algorithms that improve efficiency while retaining fairness.²² Similarities and differences between those studies and ours are illustrated in detail in Section 1.1, Remark 1 in Section 3.1, and Section 4.

This paper can be regarded as part of the literature in matching with constraints. Research in this literature include Abdulkadiroğlu (2005), Ergin and Sönmez (2006), and Hafalir, Yenmez and Yildirim (2013) for school choice, Abraham, Irving and Manlove (2007) for project allocation, Westkamp (2013) and Aygün and Turhan (2020) for college admission, Pathak et al. (2021) on pandemic resource allocation, and Kamada and Kojima (2015, 2017, 2018, 2023), Goto et al. (2014), and Biro et al. (2010) on labor markets.²³ The main departure of the present paper is that we consider integration of multiple markets, while those earlier contributions treat the relevant market as given.

²²Algorithms based on analogous ideas have been adapted to other settings by, for example, Combe, Tercieux and Terrier (2022), Combe et al. (2022), and Erdil and Kumano (2019).

²³See also Kojima, Tamura and Yokoo (2018) who explore connection between matching with constraints and convexity and concavity notions in discrete mathematics.

The main applications of the present paper is allocation of daycare seats. Research in daycare allocation in the matching literature include Kennes, Monte and Tumennasan (2014), Veski et al. (2017), Herzog and Klein (2018), Okumura (2019), Delacrétaz (2019), and Kamada and Kojima (2023). While those papers and ours share interest in daycare, the overlap is rather limited, as none of those papers analyzes models with multiple regions as we do.

7. CONCLUSION

This paper studied a school-choice matching model that allows for *ekkyo* admission (inter-regional transfer of students). To accommodate practical concerns, we provided solutions for *partial* integration, namely a mechanism that satisfies the balancedness constraint. Given any matching, we defined a directed bipartite graph (the “FIG”) in which the nodes represent students and schools while the edges are constructed using student preferences, school priorities, and the information about the current matching. Using this graph, we characterized the set of efficient iBFs (individually rational, balanced, and fair matchings) by non-existence of a cycle that would improve the welfare of the students involved in the cycle. This led us to define the FIG cycles algorithm that computes an efficient iBF in polynomial time. In terms of application, our analysis provided a way to improve upon mechanisms organized in a fragmented manner.

Market fragmentation is prevalent in real markets: for example, in the City of Tokyo, each of the 23 municipalities independently organizes a daycare matching market. We started discussions with some of those municipalities about possible implementation of our FIG cycles algorithm. One of the appealing features of our method is that it does not require the 23 municipalities to fully integrate with one another: It would be impractical to require that transfer of students be allowed between different municipalities without restrictions. We, by contrast, only require partial integration where transfer can be made to the extent that the balancedness condition is respected. Given our experience with practitioners in daycare policies, we view such partial integration to be a practical solution.

REFERENCES

- Abdulkadiroğlu, A.** 2005. “College admissions with affirmative action.” *International Journal of Game Theory*, 33(4): 535–549.
- Abdulkadiroğlu, Atila, and Tayfun Sönmez.** 2003. “School Choice: A Mechanism Design Approach.” *American Economic Review*, 93: 729–747.

- Abraham, D. J., R.W. Irving, and D.F. Manlove.** 2007. “Two algorithms for the student-project allocation problem.” *Journal of Discrete Algorithms*, 5: 73–90.
- Agarwal, Nikhil, Itai Ashlagi, Eduardo Azevedo, Clayton R Featherstone, and Ömer Karaduman.** 2019. “Market failure in kidney exchange.” *American Economic Review*, 109(11): 4026–70.
- Aygün, Orhan, and Bertan Turhan.** 2020. “Dynamic reserves in matching markets.” *Journal of Economic Theory*, 188: 105069.
- Balinski, Michel, and Tayfun Sönmez.** 1999. “A tale of two mechanisms: student placement.” *Journal of Economic theory*, 84(1): 73–94.
- Biro, P., T. Fleiner, R.W. Irving, and D.F. Manlove.** 2010. “The College Admissions Problem with Lower and Common Quotas.” *Theoretical Computer Science*, 411(34-36): 3136–3153.
- Combe, Julien, Olivier Tercieux, and Camille Terrier.** 2022. “The design of teacher assignment: Theory and evidence.” *The Review of Economic Studies*, 89(6): 3154–3222.
- Combe, Julien, Umut Mert Dur, Olivier Tercieux, Camille Terrier, and M Utku Ünver.** 2022. *Market Design for Distributional Objectives in (Re) assignment: An Application to Improve the Distribution of Teachers in Schools*. MIT Department of Economics.
- Cormen, Thomas H, Charles E Leiserson, Ronald L Rivest, and Clifford Stein.** 2001. *Introduction to Algorithms*. MIT Press and McGraw-Hill.
- Delacrétaz, David.** 2019. “Stability in matching markets with sizes.” Working paper.
- Dur, Umut Mert, and M Utku Ünver.** 2019. “Two-sided matching via balanced exchange.” *Journal of Political Economy*, 127(3): 1156–1177.
- Erdil, Aytek, and Haluk Ergin.** 2008. “What’s the Matter with Tie-Breaking? Improving Efficiency in School Choice.” *American Economic Review*, 98: 669–689.
- Erdil, Aytek, and Haluk Ergin.** 2017. “Two-sided matching with indifferences.” *Journal of Economic Theory*, 171: 268–292.
- Erdil, Aytek, and Taro Kumano.** 2019. “Efficiency and stability under substitutable priorities with ties.” *Journal of Economic Theory*, 184: 104950.
- Ergin, Haluk, and Tayfun Sönmez.** 2006. “Games of School Choice under the Boston Mechanism.” *Journal of Public Economics*, 90: 215–237.
- Gale, David, and Lloyd S. Shapley.** 1962. “College Admissions and the Stability of Marriage.” *American Mathematical Monthly*, 69: 9–15.
- Goto, Masahiro, Atsushi Iwasaki, Yujiro Kawasaki, Yosuke Yasuda, and**

- Makoto Yokoo.** 2014. “Improving Fairness and Efficiency in Matching with Distributional Constraints: An Alternative Solution for the Japanese Medical Residency Match.” https://mpra.ub.uni-muenchen.de/53409/1/MPRA_paper_53409.pdf, last accessed on October 13, 2023.
- Hafalir, Isa E, Fuhito Kojima, and M Bumin Yenmez.** 2022. “Interdistrict school choice: A theory of student assignment.” *Journal of Economic Theory*, 201: 105441.
- Hafalir, Isa E, M Bumin Yenmez, and Muhammed A Yildirim.** 2013. “Effective affirmative action in school choice.” *Theoretical Economics*, 8(2): 325–363.
- Herzog, Sabrina, and Thilo Klein.** 2018. “Matching Practices for Childcare—Germany.” <http://www.matching-in-practice.eu/matching-practices-for-childcare-germany/>.
- Kamada, Yuichiro, and Fuhito Kojima.** 2015. “Efficient Matching under Distributional Constraints: Theory and Applications.” *American Economic Review*, 105(1): 67–99.
- Kamada, Yuichiro, and Fuhito Kojima.** 2017. “Stability concepts in matching with distributional constraints.” *Journal of Economic Theory*, 168: 107–142.
- Kamada, Yuichiro, and Fuhito Kojima.** 2018. “Stability and Strategy-Proofness for Matching with Constraints: A Necessary and Sufficient Condition.” *Theoretical Economics*, 13: 761–793.
- Kamada, Yuichiro, and Fuhito Kojima.** 2023. “Fair Matching under Constraints: Theory and Applications.” Forthcoming, *The Review of Economic Studies*.
- Kennes, John, Daniel Monte, and Norovsambuu Tumennasan.** 2014. “The day care assignment: A dynamic matching problem.” *American Economic Journal: Microeconomics*, 6(4): 362–406.
- Kojima, Fuhito, Akihisa Tamura, and Makoto Yokoo.** 2018. “Designing Matching Mechanisms under Constraints: An Approach from Discrete Convex Analysis.” *Journal of Economic Theory*, 176: 803–833.
- Okumura, Yasunori.** 2019. “School choice with general constraints: a market design approach for the nursery school waiting list problem in Japan.” *The Japanese Economic Review*, 70: 497–516.
- Pathak, Parag A, Tayfun Sönmez, M Utku Ünver, and M Bumin Yenmez.** 2021. “Fair allocation of vaccines, ventilators and antiviral treatments: leaving no ethical value behind in health care rationing.” Forthcoming, *Management Science*.
- Roth, Alvin E.** 1986. “On the allocation of residents to rural hospitals: a general property of two-sided matching markets.” *Econometrica*, 54(2): 425–427.

- Shapley, Lloyd, and Herbert Scarf.** 1974. “On cores and indivisibility.” *Journal of mathematical economics*, 1(1): 23–37.
- Suzuki, Takamasa, Akihisa Tamura, Kentaro Yahiro, Makoto Yokoo, and Yuzhe Zhang.** 2023. “Strategyproof allocation mechanisms with endowments and M-convex distributional constraints.” *Artificial Intelligence*, 315: 103825.
- Veski, André, Péter Biró, Kaire Pöder, and Triin Lauri.** 2017. “Efficiency and fair access in kindergarten allocation policy design.” *The Journal of Mechanism and Institution Design*, 2(1): 57–104.
- Westkamp, Alexander.** 2013. “An Analysis of the German University Admissions System.” *Economic Theory*, 53(3): 561–589.
- Wu, Qingyun, and Alvin E Roth.** 2018. “The lattice of envy-free matchings.” *Games and Economic Behavior*, 109: 201–211.

APPENDIX A. PROOFS

A.1. Proof of Theorem 1.

Proof. Fix μ and a FIG cycle $\mathcal{F} = (i_1, s_1, i_2, s_2, \dots, i_m, s_m)$ on μ . Let μ' be the matching that is generated by (μ, \mathcal{F}) . Clearly, μ' Pareto dominates μ .

To show that μ' is fair, notice that $\mu'_i \succeq_i \mu_i$ for all $i \in I$ by the definition of FIG cycle. This implies that, for every $i \notin \{i_1, \dots, i_m\}$, no one has justified envy to i under μ' because μ is fair. Thus, it remains to show that for each $k \in \{1, \dots, m\}$, no one has justified envy to i_k under μ' . To see this, note that, again by $\mu'_i \succeq_i \mu_i$ for all $i \in I$, $s \succ_i \mu'_i$ implies $s \succ_i \mu_i$. Hence, $D_s^{\mu'} \subseteq D_s^\mu$ for any $s \in S$ (and thus in particular for $s = s_k$).²⁴ Since $i_k = \text{Top}_{s_k}(D_{s_k}^\mu)$ for each k by definition, this implies $i \not\succeq_{s_k} i_k$ for every $i \in D_{s_k}^{\mu'}$. Hence, no one has justified envy to i_k under μ' .

To show that μ' is balanced, consider a sequence (r_1, \dots, r_m) such that $s_k \in r_k$ for each k . Fix $r \in R$ and let $K(r) = \{k \in \{1, \dots, m\} | r_k = r\}$. If $K(r) = \emptyset$, then (2.1) is satisfied for r under μ' because it is satisfied for r under μ . So suppose $K(r) \neq \emptyset$. Let

$$In_r = \{k \in K(r) | r_{k-1} \neq r\} \text{ and } Out_r = \{k \in K(r) | r_{k+1} \neq r\}.$$

Clearly we must have $|In_r| = |Out_r|$. Define $In'_r \subseteq In_r$ and $Out'_r \subseteq Out_r$ by

$$In'_r = \{k \in K(r) | r_{k-1} \neq r \text{ and } i_k \in r\} \text{ and } Out'_r = \{k \in K(r) | r_{k+1} \neq r \text{ and } i_{k+1} \in r\}.$$

On the one hand, the inflow to r has changed from μ to μ' by

$$\underbrace{(|In_r| - |In'_r|)}_{\text{the number of non-}r \text{ students coming to } r} - \underbrace{(|Out_r| - |Out'_r|)}_{\text{the number of non-}r \text{ students going out of } r}.$$

On the other hand, the outflow from r has changed from μ to μ' by

$$\underbrace{|Out'_r|}_{\text{the number of } r \text{ students going out of } r} - \underbrace{|In'_r|}_{\text{the number of } r \text{ students coming to } r}.$$

Note that these two values are equal because $|In_r| = |Out_r|$. Finally, since (2.1) holds for r under μ , this implies that (2.1) holds for r under μ' .

To show that individual rationality of μ implies individual rationality of μ' , note first that $\mu'_i \succeq \emptyset$ for each $i \in I$ because μ is individually rational and μ' Pareto dominates μ . Moreover, $i \succ_s \emptyset$ for every $s \in S$ and $i \in \mu'_s$ because (i) $i \succ_s \emptyset$ for every $i \in \mu'_s \cap \mu_s$ by the assumption that μ is individually rational, while (ii) $i \succ_s \emptyset$ for every $i \in \mu'_s \setminus \mu_s$ by the definition of FIG (more specifically, part (1) of Definition 3). \square

²⁴Recall that, for any matching $\tilde{\mu}$ and school s , $D_s^{\tilde{\mu}} = \{j \in I | s \succ_j \tilde{\mu}_j\}$.

A.2. Proof of Theorem 2.

Proof. We prove the contraposition, so fix an iBF μ and assume that there exists an iBF μ' that Pareto dominates μ .

Lemma 1. *For any $r \in R$, $|\{i \in I | \mu_i \in r, \mu'_i \notin r\} \cup \{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}|$ is equal to $|\{i \in I | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\} \cup \{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}|$.*

Proof. Since μ and μ' are both balanced, the change (from μ to μ') of the inflow of students to r and the change of the outflow are the same as each other. The change of the inflow is:

$$\underbrace{|\{i \in I \setminus r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{coming from an outside school to } r}} + \underbrace{|\{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{coming from being unmatched to } r}} - \underbrace{|\{i \in I \setminus r | \mu_i \in r, \mu'_i \notin r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{going out of } r}}.$$

The change of the outflow is:

$$\underbrace{|\{i \in r | \mu_i \in r, \mu'_i \in r' \text{ for some } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{going out of } r}} + \underbrace{|\{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{who was unmatched but is now matched to a non-}r \text{ school}}} \\ - \underbrace{|\{i \in r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \notin r' \text{ for all } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{who was matched to a non-}r \text{ school but now is not}}}.$$

Thus, we have:

(A.1)

$$\underbrace{|\{i \in r | \mu_i \in r, \mu'_i \in r' \text{ for some } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{going out of } r}} + \underbrace{|\{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{who was unmatched but is now matched to a non-}r \text{ school}}} \\ + \underbrace{|\{i \in I \setminus r | \mu_i \in r, \mu'_i \notin r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{going out of } r}} = \\ \underbrace{|\{i \in I \setminus r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{coming from an outside school to } r}} + \underbrace{|\{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}|}_{\substack{\text{the number of non-}r \text{ students} \\ \text{coming from being unmatched to } r}} \\ + \underbrace{|\{i \in r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \notin r' \text{ for all } r' \neq r\}|}_{\substack{\text{the number of } r \text{ students} \\ \text{who was matched to a non-}r \text{ school but now is not}}}.$$

Now, recall that μ is individually rational. Hence, $\mu'_i \succeq_i \mu_i \succeq_i \emptyset$ for every i , and thus we have that $\mu_i \in r$ implies $\mu'_i \neq \emptyset$. Therefore,

$$\{i \in r | \mu_i \in r, \mu'_i \in r' \text{ for some } r' \neq r\} = \{i \in r | \mu_i \in r, \mu'_i \notin r\}$$

Also, for the same reason, we have

$$\{i \in r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \notin r' \text{ for all } r' \neq r\} = \{i \in r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}.$$

Hence, (A.1) is equivalent to

$$\begin{aligned} & |\{i \in r | \mu_i \in r, \mu'_i \notin r\}| + |\{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}| + |\{i \in I \setminus r | \mu_i \in r, \mu'_i \notin r\}| = \\ & |\{i \in I \setminus r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}| + |\{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}| \\ & + |\{i \in r | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}|, \end{aligned}$$

or

$$\begin{aligned} & |\{i \in I | \mu_i \in r, \mu'_i \notin r\}| + |\{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}| = \\ & |\{i \in I | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\}| + |\{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}|. \end{aligned}$$

Since the two terms in each side of the above equation are disjoint from each other, we have

$$\begin{aligned} & |\{i \in I | \mu_i \in r, \mu'_i \notin r\} \cup \{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}| = \\ & |\{i \in I | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\} \cup \{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}|. \end{aligned}$$

This completes the proof. \square

Consider the following graph, in which the only agents associated with arrows are the schools in r and students who are matched to a school in r under μ and students living in r who are unmatched under μ . First, each $s \in r$ points to each student $i \in \mu_s \setminus \mu'_s$. Then, for each student who was pointed to by some school in r and each student living in r who are unmatched under μ , let her point to the school μ'_i if $\mu'_i \in r$. Moreover, by Lemma 1, there is a one-to-one and onto mapping from $\{i \in I | \mu_i \in r, \mu'_i \notin r\} \cup \{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}$ to $\{i \in I | \mu_i \in r' \text{ for some } r' \neq r, \mu'_i \in r\} \cup \{i \in I \setminus r | \mu_i = \emptyset, \mu'_i \in r\}$. Take one such mapping ϕ . Then, for each $i \in \{i \in I | \mu_i \in r, \mu'_i \notin r\} \cup \{i \in r | \mu_i = \emptyset, \mu'_i \in r' \text{ for some } r' \neq r\}$, let i point to $\mu'_{\phi(i)}$. This defines a directed graph, denoted $\mathcal{G}(\mu, \mu', r)$. By construction, only schools in r , students matched to a school in r under μ , and the students of r who are unmatched under μ may be associated with arrows in this graph.

Let $I' := \{i \in I | \mu'_i \succ_i \mu_i\}$. By the assumption that μ' Pareto dominates μ , we have $I' \neq \emptyset$. For any matching $\tilde{\mu}$, let $S(\tilde{\mu}) := \{s \in S | s = \tilde{\mu}_i \text{ for some } i \in I'\}$.

Lemma 2. *Suppose $s \in S(\mu) \setminus S(\mu')$. Then there exists a school $s' \in r(s) \cap S(\mu')$ such that $|\mu_{s'}| < q_{s'}$.*

Proof. Take an arbitrary s such that $s \in S(\mu) \setminus S(\mu')$ (If there is no such school, we are done). Starting from this school, follow the arrows in $\mathcal{G}(\mu, \mu', r(s))$ in an arbitrary manner without passing the same student twice (note that there is an outgoing arrow from s). Since there are a finite number of students, there is s' such that there is no more outgoing arrow from s' to a student who has not appeared in the path (note that, by definition,

this path cannot end at any student). Because $s \notin S(\mu')$ and hence $s' \neq s$, the number of incoming arrows to s' is greater than that of outgoing arrows from s' along this path. This implies that the number of students who are in $\mu'_{s'} \setminus \mu_{s'}$ is larger than the number of students who are in $\mu_{s'} \setminus \mu'_{s'}$ by at least one. Hence, we have $q_{s'} \geq |\mu'_{s'}| > |\mu_{s'}|$. Since $s' \in r(s)$ and $s' \in S(\mu')$ by the definition of the graph, this completes the proof. \square

Lemma 3. *Suppose $\mu_i = \emptyset$ and $i \in I'$. Then there exists a school $s' \in r(i) \cap S(\mu')$ such that $|\mu_{s'}| < q_{s'}$.*

Proof. Suppose there is i such that $\mu_i = \emptyset$ and $i \in I'$. Starting from this student i , follow the arrows in $\mathcal{G}(\mu, \mu', r)$ in an arbitrary manner without passing the same student twice (note that there is an outgoing arrow from i). Since there are a finite number of students, there is s' such that there is no more outgoing arrow from s' to a student who has not appeared in the path (note that, by definition, this path cannot end at any student). This implies that the number of students who are in $\mu'_{s'} \setminus \mu_{s'}$ is larger than the number of students who are in $\mu_{s'} \setminus \mu'_{s'}$ by at least one. Hence, we have $q_{s'} \geq |\mu'_{s'}| > |\mu_{s'}|$. Since $s' \in r(i)$ and $s' \in S(\mu')$ by the definition of the graph $\mathcal{G}(\mu, \mu', r(i))$, this completes the proof. \square

Next, we define a graph, denoted $\mathcal{G}(\mu, \mu')$, as follows. In this graph, only students in I' and schools in $S(\mu')$ may be associated with arrows. Formally, for any $s \in S(\mu')$, consider the set of students in I' who strictly prefer s to their match at μ , i.e., $D_s^\mu(I') := D_s^\mu \cap I'$ (or, equivalently, $D_s^\mu(I') := \{i \in I' \mid s \succ_i \mu_i\}$), and let $Top_s(D_s^\mu(I'))$ point to s . Note that $D_s^\mu(I')$ is nonempty by the definitions of I' , μ and μ' , and thus for any $s \in S(\mu')$, there exists some $i \in I'$ who points to s . Next, consider any $i \in I'$.

- (1) If $\mu_i \in S(\mu')$, then let μ_i point to i .
- (2) If $\mu_i \in S \setminus S(\mu')$, then $\mu_i \in S(\mu) \setminus S(\mu')$ by the definition of $S(\mu)$ and hence by Lemma 2, there exists a school $s' \in r(\mu_i) \cap S(\mu')$ such that $|\mu_{s'}| < q_{s'}$. Let any such school s' point to i .
- (3) If $\mu_i = \emptyset$, then by Lemma 3, there exists $s \in r(i) \cap S(\mu')$ such that $|\mu_s| < q_s$. Let any such school s point to i .

The graph must have a cycle because each school is pointed to by a single student, and each student is pointed to by at least one school.²⁵ Pick an arbitrary cycle and call it \mathcal{F}^* .

²⁵To find a cycle, take an arbitrary school and find the student pointing to that school. Then find the school pointing to that student. Then find the student pointing to that school, etc., until we find a school or a student that has already been visited. Since there are only finitely many students, this procedure ends in finite steps.

Lemma 4. \mathcal{F}^* is a FIG cycle.

Proof. Let $\mathcal{F}^* = (i_1, s_1, \dots, i_m, s_m)$.

It is straightforward to check that the last two conditions of a cycle are satisfied: By construction, each student appears only once in \mathcal{F}^* . Given this, since the pointing rule for $\mathcal{G}(\mu, \mu')$ implies that each school is pointed to only by a single student, each school appears at most once in \mathcal{F}^* .

To complete the proof, it suffices to show that $\mathcal{G}(\mu, \mu')$ is a subset of the FIG on μ . To show this, first we establish that, for any $k \in \{1, \dots, m\}$, i_k points to s_k according to the definition of pointing used for FIG. To do so, it suffices to show that $i_k = \text{Top}_{s_k}(D_{s_k}^\mu)$ and $i_k \succ_{s_k} \emptyset$ (part (1) of Definition 3). The latter holds for the following reason: we have $i \in D_{s_k}^\mu(I')$ for some $i \in \mu'_{s_k}$ because of the definitions of $S(\mu')$ and I' and the fact that $s_k \in S(\mu')$. Hence, by individual rationality of μ' , it follows that $i_k = \text{Top}_{s_k}(D_{s_k}^\mu(I')) \succeq_{s_k} i \succ_{s_k} \emptyset$. To show the former, note first that, by construction, $i_k = \text{Top}_{s_k}(D_{s_k}^\mu(I'))$ and hence $i_k \succ_{s_k} i$ for any $i \in (D_{s_k}^\mu \cap I') \setminus \{i_k\}$. Next, consider any $i \in D_{s_k}^\mu \setminus I'$. Because $\mu'_{i'} = \mu_{i'}$ for any $i' \in I \setminus I'$ by the definition of I' , it follows that $i \in D_{s_k}^{\mu'}$. This and the assumption that μ' is fair imply $j \succ_{s_k} i$ for every $j \in \mu'_{s_k}$. By the construction of the cycle, $i_k \succeq_{s_k} j$ for every $j \in \mu'_{s_k} \setminus \mu_{s_k} \neq \emptyset$ (the nonemptiness holds because $s_k \in S(\mu')$). Thus, we have $i_k \succ_{s_k} i$. Therefore, we have $i_k \succ_{s_k} i$ for any $i \in D_{s_k}^\mu \setminus \{i_k\}$, which implies $i_k = \text{Top}_{s_k}(D_{s_k}^\mu)$.

Second, we consider three cases of the definition of $\mathcal{G}(\mu, \mu')$ to show that, for any $k \in \{1, \dots, m\}$, s_k points to i_{k+1} according to the definition of pointing used for FIG, where $i_{m+1} = i_1$. Suppose first that i_{k+1} and s_k satisfy the condition described in Case 1 of the definition of $\mathcal{G}(\mu, \mu')$. This implies that i_{k+1} and s_k satisfy the assumption in part (2a) of Definition 3. Next, consider Case 2 of the definition of $\mathcal{G}(\mu, \mu')$. In this case, $|\mu_{s_k}| < q_{s_k}$ and $\mu_{i_{k+1}} \in r(s_k)$ hold, which satisfies the condition in part (2b) of Definition 3. Finally, consider Case 3 of the definition of $\mathcal{G}(\mu, \mu')$. In this case, $|\mu_{s_k}| < q_{s_k}$, $i_{k+1} \in r(s_k)$, and $\mu_{i_{k+1}} = \emptyset$ hold, which again satisfies the condition in part (2b) of Definition 3. This completes the proof. \square

Lemma 4 completes the proof. \square

A.3. Proof of Claim 1.

Proof. We first observe that the balancedness of μ' at \mathcal{E}' is equivalent to the property that, for each $r \in R'$,

$$\underbrace{\sum_{s \in r} |\{i | i \in \mu'_s, i \notin r\}|}_{\text{inflow to } r} - \underbrace{\sum_{s \notin r} |\{i | i \in \mu'_s, i \in r\}|}_{\text{outflow from } r} = 0.$$

Consider any $r^* \in R$. Then,

$$(A.2) \quad \sum_{r \in R', r \subseteq r^*} \left[\underbrace{\sum_{s \in r} |\{i | i \in \mu'_s, i \notin r\}|}_{\text{inflow to } r} - \underbrace{\sum_{s \notin r} |\{i | i \in \mu'_s, i \in r\}|}_{\text{outflow from } r} \right] = 0.$$

The left-hand side of (A.2) is equal to

$$\begin{aligned} & \sum_{r \in R', r \subseteq r^*} \left[\underbrace{\sum_{s \in r} (|\{i | i \in \mu'_s, i \notin r^*\}| + |\{i | i \in \mu'_s, i \in r^* \setminus r\}|)}_{\text{inflow to } r} - \underbrace{\left(\sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r\}| + \sum_{s \in r^* \setminus r} |\{i | i \in \mu'_s, i \in r\}| \right)}_{\text{outflow from } r} \right] \\ &= \sum_{r \in R', r \subseteq r^*} \left[\sum_{s \in r} |\{i | i \in \mu'_s, i \notin r^*\}| - \sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r\}| \right] + \sum_{r \in R', r \subseteq r^*} \left[\sum_{s \in r} |\{i | i \in \mu'_s, i \in r^* \setminus r\}| - \sum_{s \in r^* \setminus r} |\{i | i \in \mu'_s, i \in r\}| \right] \\ &= \left[\sum_{s \in r^*} |\{i | i \in \mu'_s, i \notin r^*\}| - \sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r^*\}| \right] + \sum_{r, r' \in R', r, r' \subseteq r^*, r \neq r'} \left[\sum_{s \in r} |\{i | i \in \mu'_s, i \in r'\}| - \sum_{s \in r'} |\{i | i \in \mu'_s, i \in r\}| \right] \\ &= \left[\sum_{s \in r^*} |\{i | i \in \mu'_s, i \notin r^*\}| - \sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r^*\}| \right] \\ &+ \sum_{k < \ell} \left(\left[\underbrace{\sum_{s \in r_k} |\{i | i \in \mu'_s, i \in r_\ell\}|}_{\text{flow from } r_\ell \text{ to } r_k} - \underbrace{\sum_{s \in r_\ell} |\{i | i \in \mu'_s, i \in r_k\}|}_{\text{flow from } r_k \text{ to } r_\ell} \right] + \left[\underbrace{\sum_{s \in r_\ell} |\{i | i \in \mu'_s, i \in r_k\}|}_{\text{flow from } r_k \text{ to } r_\ell} - \underbrace{\sum_{s \in r_k} |\{i | i \in \mu'_s, i \in r_\ell\}|}_{\text{flow from } r_\ell \text{ to } r_k} \right] \right) \\ &= \underbrace{\sum_{s \in r^*} |\{i | i \in \mu'_s, i \notin r^*\}|}_{\text{inflow to } r^*} - \underbrace{\sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r^*\}|}_{\text{outflow from } r^*}, \end{aligned}$$

where we let $\{r_1, r_2, \dots\}$ be a partition of r^* by regions in R' .

By (A.2), this implies that

$$\underbrace{\sum_{s \in r^*} |\{i | i \in \mu'_s, i \notin r^*\}|}_{\text{inflow to } r^*} - \underbrace{\sum_{s \notin r^*} |\{i | i \in \mu'_s, i \in r^*\}|}_{\text{outflow from } r^*} = 0.$$

Hence the inflow and outflow for region r^* are equal to each other. Since this relation holds for every $r^* \in R$, μ' is balanced at \mathcal{E} . \square

A.4. Comparative Statics: An Example. The following example shows that the converse of Proposition 1 does not hold. That is, starting at a certain efficient iBF, some student can be made strictly better off in any efficient iBF after splitting a region.

Example 8 (An instance in which splitting a region inevitably makes some student better off). Let $I = \{i_1, i_2\}$ and $S = \{s_1, s_2, s_3\}$. Let there be two regions, $r = \{i_1, s_1, s_2\}$ and $r' = \{i_2, s_3\}$. Each school has the capacity of one. Student preferences and school priorities are given as follows:

$$\begin{array}{ll} \succ_{i_1}: s_2, s_3, & \succ_{s_1}: i_2, \\ \succ_{i_2}: s_1, & \succ_{s_2}: i_1, \\ & \succ_{s_3}: i_1. \end{array}$$

In this environment, there is an efficient iBF:

$$\mu = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ \emptyset & i_1 & \emptyset & i_2 \end{pmatrix}.$$

If region r is split into two regions, $r_1 = \{i_1, s_1\}$ and $r_2 = \{s_2\}$, then there is a unique efficient iBF:

$$\mu' = \begin{pmatrix} s_1 & s_2 & s_3 & \emptyset \\ i_2 & \emptyset & i_1 & \emptyset \end{pmatrix}.$$

Note that i_2 is better off as a result of the split. The intuition is the following: Before the split, i_2 in r' was unable to match with s_1 in r as there was no student who wanted to come from r to his region r' . However, the split made it impossible for i_1 in r_1 (which was part of r) to go to s_2 , and she is now interested in coming to r' . This made it possible to implement a swap between r_1 and r' .

Notice that, in the above example, both μ and μ' are an efficient iBF before region r was split. In fact, multiplicity of efficient iBFs is necessary for the failure of the converse of Proposition 1: Indeed, Proposition 1 implies its converse if there exists a unique efficient iBF before the split of a region.