

Popular Matchings under Lower Quotas and their relationship to Stability under Classifications

Ankita Sarkar

Chennai Mathematical Institute

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Contents

- 1 Stable Marriage Problem
- 2 Hospital-Residents Problem
- 3 Popular Matchings
- 4 Hospital-Residents under Lower Quotas
- 5 Both-sided Lower Quotas
- 6 The Students-Courses Problem
- 7 Stability under Classifications
- 8 Generalising Further

Stable Marriage Problem

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- A bipartite graph, $G = (\mathcal{M} \uplus \mathcal{W}, E)$, representing men and women, with edges between men and women who are potential partners in marriage.

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Assumption : Strict preference lists, i.e. a vertex v has preference list of the form $(E(v), <_v)$.

Defining “Happiness”

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- no pair has incentive to deviate from the assigned matching.

Definition

(m, w) is a **blocking pair** with respect to a matching $M \subseteq E$ if the following conditions hold :

- $(m, w) \in E \setminus M$
- either m is unmatched or $w >_m M(m)$
- either w is unmatched or $m >_w M(w)$

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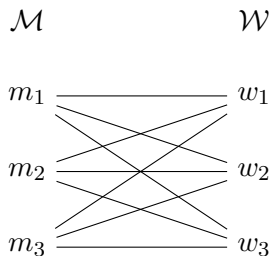
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Example



$m_1 : w_1 \ w_2 \ w_3$ $w_1 : m_2 \ m_3 \ m_1$
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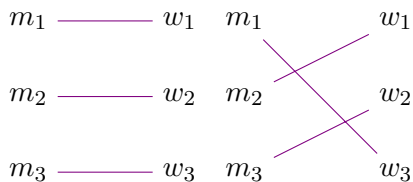
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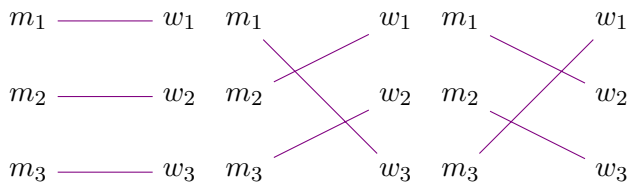


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Finding Stable Matchings

[Gale and Shapley, 1962] : showed existence of stable matchings via an algorithm to find them.

Gale-Shapley Algorithm

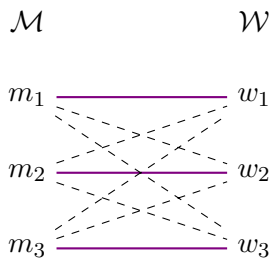
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- Each man proposes to women, in the order most to least preferred, until he finds a match or exhausts his preference list.
- Each woman provisionally accepts the first proposal she receives. Thereafter, she compares incoming proposals with her existing partner, and rejects the less preferred man.

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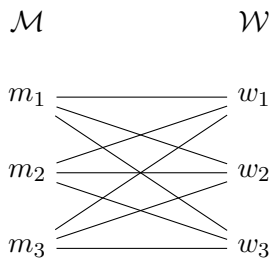


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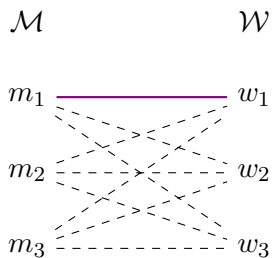


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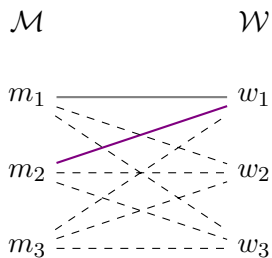


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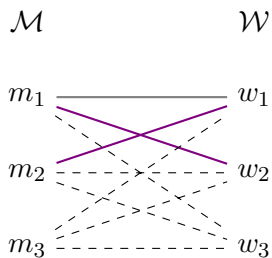


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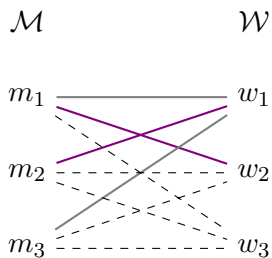


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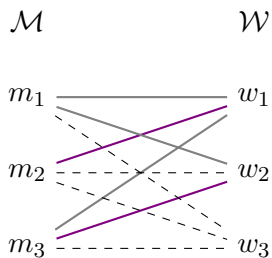
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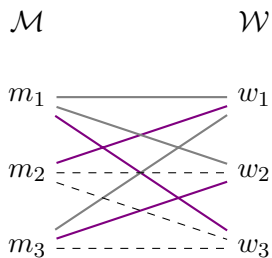


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Definition

A **matching** is $M \subseteq E$ such that $\forall h \in \mathcal{H}, |M(h)| \leq q^+(h)$ and $\forall r \in \mathcal{R}, |M(r)| = 1$.

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$(h, r) \in \mathcal{H} \times \mathcal{R}$ is a **blocking pair** with respect to a matching $M \subseteq E$ if the following conditions hold :

- $(h, r) \in E \setminus M$
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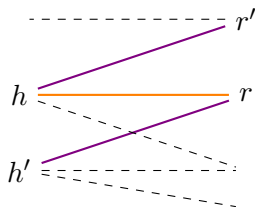
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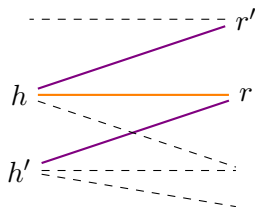
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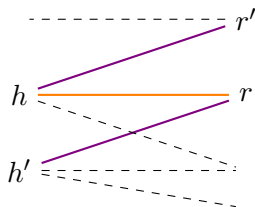
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- h must have proposed to r before r' . Thus h was rejected in favour of some $h'', h'' >_r h$.
- But **partners of residents only improve**. So $h' \geq_r h'' >_r h$, which is a contradiction.

Each hospital proposes at most once to each of its neighbour residents. Hence, the Gale-Shapley algorithm runs in $\mathcal{O}(|E|)$ time.

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- *All stable matchings have the same size.*
- *Undersubscribed hospitals are matched to **the same set of residents** in every stable matching.*

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- *Undersubscribed hospitals are matched to **the same set of residents** in every stable matching.*

This means that if the Gale-Shapley output is small, all stable matchings are small. They can in fact be as small as **1/2 the size of a maximum matching**.

Example : Small stable matching

$m_1 : w_1 w_2$

$w_1 : m_1 m_2$

$m_2 : w_1$

$w_2 : m_1$

The Gale-Shapley output is $\{(m_1, w_1)\}$ while the maximum matching is of size 2.

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- **Stability** : no pair has incentive to deviate from the assigned matching.
- **Popularity** : in a comparison with any other matching, a majority of vertices prefer this matching.

Definition

For a vertex v , its **vote** is defined as follows :

$vote_v : E(v) \times E(v) \rightarrow \{-1, 1\}$, where

$$vote_v(u, w) = \begin{cases} 1 & u >_v w \\ -1 & u <_v w \\ 0 & u = w \end{cases}$$

Voting between Matchings

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Definition

For a vertex v , wlog, $|M(v) \setminus M'(v)| \leq |M'(v) \setminus M(v)|$. Fix an arbitrary bijection between $M(v) \setminus M'(v)$ and an arbitrary subset of the same size in $M'(v) \setminus M(v)$. This bijection is $\mathbf{corr}_v|_{M(v) \setminus M'(v)}$ and for any remaining vertex u in $M'(v) \setminus M(v)$, $\mathbf{corr}_v(u) = \perp$.

Voting between Matchings

Once we fix an arbitrary correspondence as above for each hospital, we can calculate each vertex's vote between two matchings :

Definition

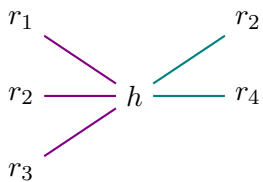
$$vote_v(M, M') = \sum_{u \in M(v) \setminus M'(v)} vote_v(u, corr_v(u))$$

Here we assume that $vote_v(u, \perp)$ is always 1 for any u .

Example

$M_1(h)$

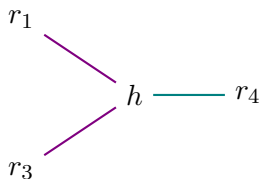
$M_2(h)$



$h : r_1 r_2 r_4 r_3$

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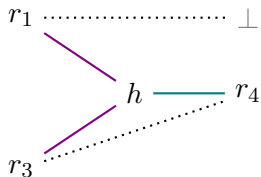
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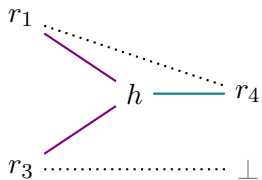


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$$h : r_1 \ r_2 \ r_4 \ r_3$$

$$\text{vote}_h(M_1, M_2) = 1 + 1 = 2$$

Now we can formalise popularity :

Definition

M' is **more popular than** M (denoted $M' \succ M$) if

$$\sum_{v \in \mathcal{H} \cup \mathcal{R}} \text{vote}_v(M', M) > 0$$

M is **popular** among a set of matchings $\mathcal{M} \subseteq 2^E$ if no matching in \mathcal{M} is more popular than M .

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- Recall : motivation for popularity is that stable matchings are small.
- **Task** : Find **maximum cardinality** popular matchings.

Algorithm in [Huang and Kavitha, 2011].

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Idea :

- Usual round of proposals as per Gale-Shapley.
- Undersubscribed hospitals propose again with **higher priority**, where residents always prefer higher priority hospitals.

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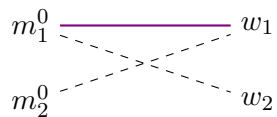
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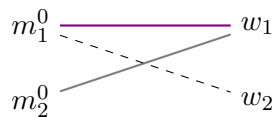
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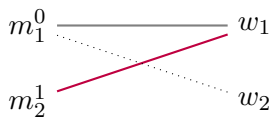
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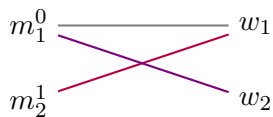
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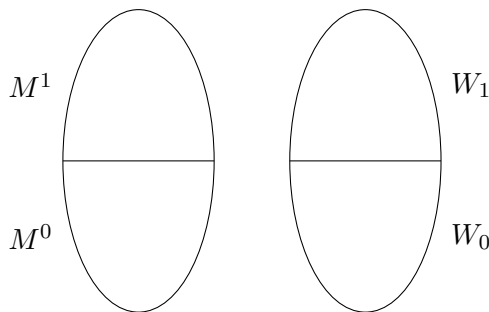
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- Label N -edges in the above paths and cycles as per the vote of each vertex v in $vote_v(N, M)$.
- Opposing votes along an edge cancel; we show that there are at most as many $(+1, +1)$ edges as $(-1, -1)$ edges.

Proof sketch :

- M^0 : Men matched at priority 0; M^1 : Men matched at priority 1
- W_0 : Unmatched women and women matched to priority-0 men; W_1 : women matched to priority-1 men.

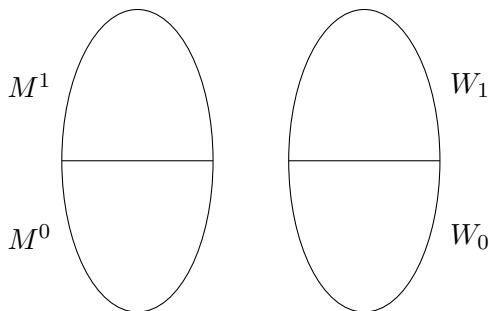
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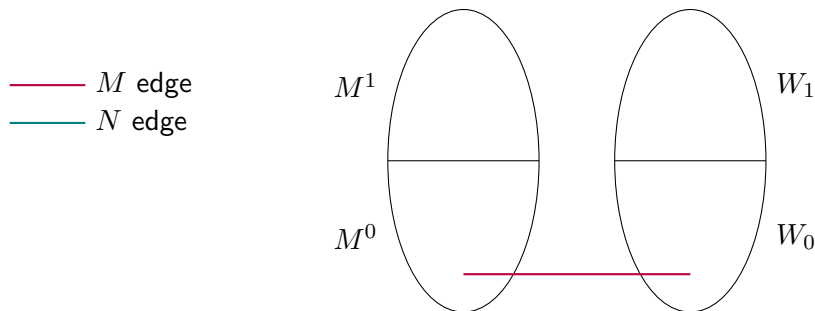
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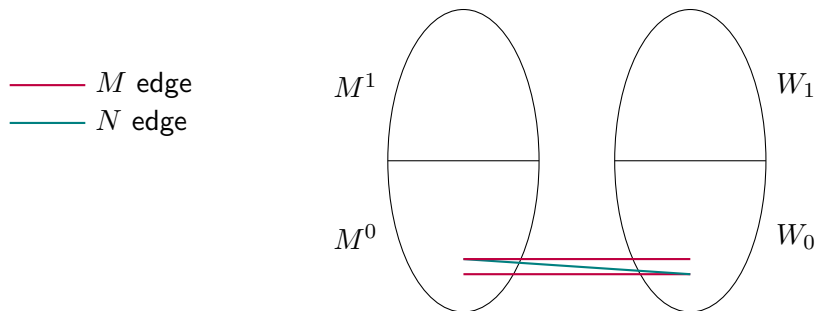
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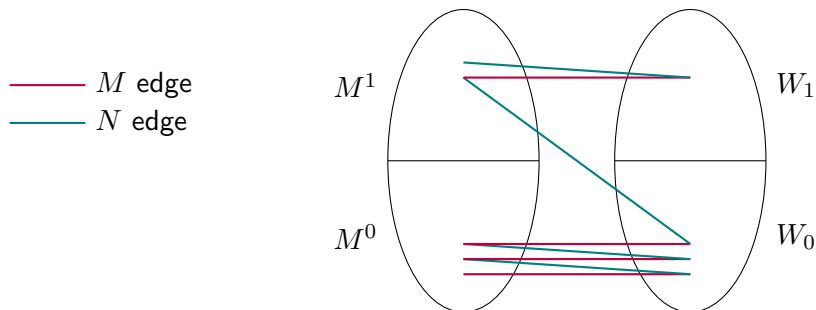
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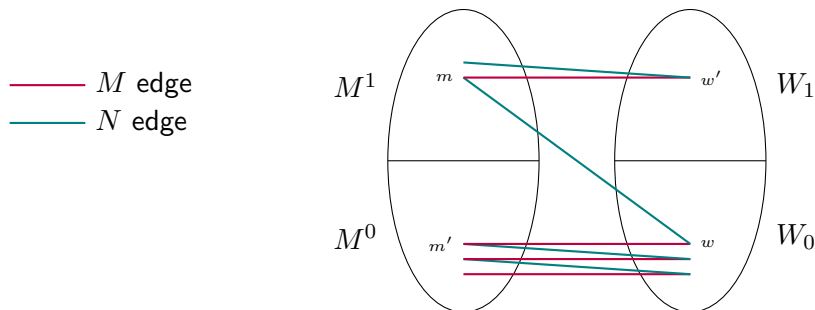
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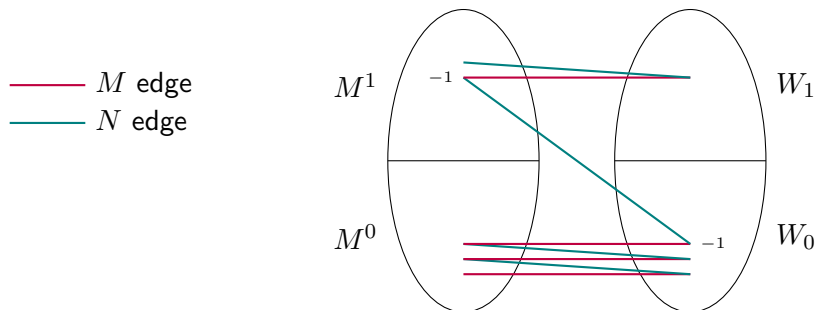
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- M^0 : Men matched at priority 0; M^1 : Men matched at priority 1
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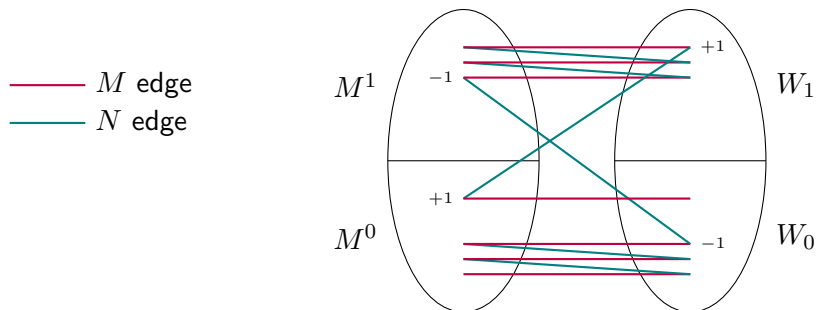
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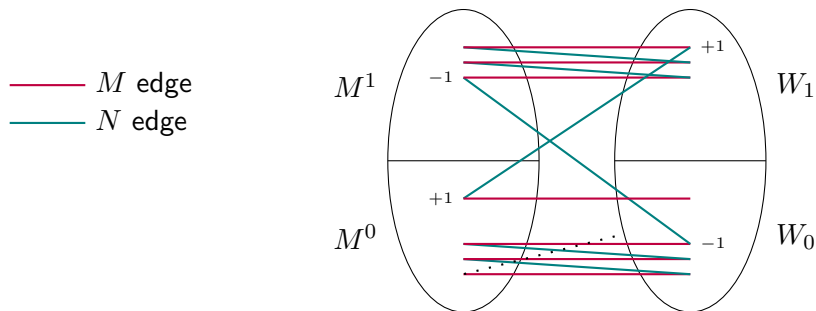
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Construction : we make two copies of each hospital h , and add dummy residents $d_h^1, \dots, d_h^{q^+(h)}$.

Original instance	New instance
$h : r_1 \dots r_k$	$h^0 : r_1 \dots, r_k d_h^1 \dots d_h^{q^+(h)}$ $h^1 : d_h^1 \dots d_h^{q^+(h)} r_1 \dots, r_k$
$r : h_1 \dots h_k$	$r : h_1^1 \dots h_k^1 h_1^0 \dots h_k^0$ $d_h^i : h^0 h^1$

Transformed instance :

$$m_1^0 : w_1 \ w_2 \ d_{m_1}^1$$

$$m_1^1 : d_{m_1}^1 \ w_1 \ w_2$$

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$$w_1 : m_1^1 \ m_2^1 \ m_1^0 \ m_2^0$$

$$w_2 : m_1^1 \ m_1^0$$

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$$d_{m_2}^1 : m_2^0 \ m_2^1$$

Example

Transformed instance :

$$\begin{array}{ll} m_1^0 : w_1 & w_2 & d_{m_1}^1 & & w_1 : m_1^1 & m_2^1 & m_1^0 & m_2^0 \\ m_1^1 : d_{m_1}^1 & w_1 & w_2 & & w_2 : m_1^1 & m_1^0 & & \\ m_2^0 : w_1 & d_{m_2}^1 & & & d_{m_1}^1 : m_1^0 & m_1^1 & & \\ m_2^1 : d_{m_2}^1 & w_1 & & & d_{m_2}^1 : m_2^0 & m_2^1 & & \end{array}$$

Gale-Shapley output is $\{(m_1^0, w_2), (m_1^1, d_{m_1}^1), (m_2^0, d_{m_2}^1), (m_2^1, w_1)\}$.
This gives the final matching $\{(m_1, w_2), (m_2, w_1)\}$.

Contents

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- Usual hospital-residents instance : $G = (\mathcal{H} \uplus \mathcal{R}, E)$;
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Question : Can we find a feasible stable matching?

Answer : Not always.

Example

$$h_1 : r_1 \ r_2 \ r_3$$

$$h_2 : r_2 \ r_3$$

$$h_3 : r_3$$

$$r_1 : h_1$$

$$r_2 : h_1 \ h_2$$

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Hospitals h_1, h_2, h_3 have upper quotas 3, 2, 1 respectively. All lower quotas are 1.

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Gale-Shapley matches all three residents with h_1 . This is infeasible. By Rural Hospitals Theorem, then **all stable matchings are infeasible**.

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Popularity among Feasible Matchings

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Observation : Feasible instance $\implies \mu - 2 \leq |\mathcal{R}|$. We can assume that trivially infeasible instances that violate this are ruled out before using the algorithm.

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The new edge set is of size $\mu |E| = \mathcal{O}(|\mathcal{R}| |E|)$.

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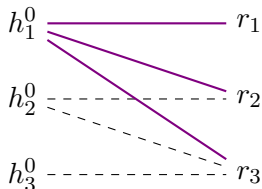
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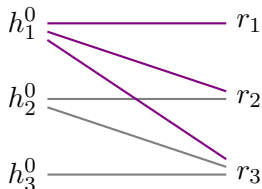
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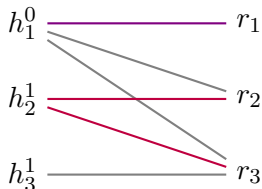
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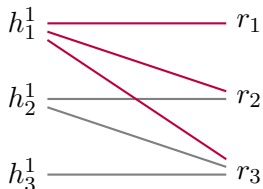
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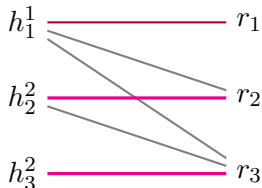
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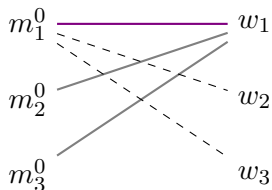
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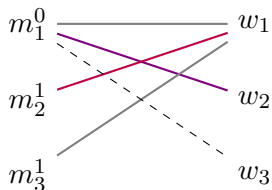
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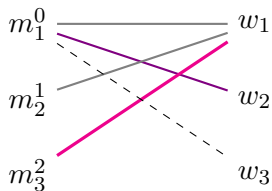
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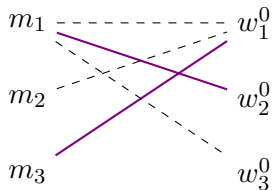
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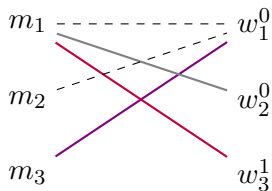
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If a vertex v had preference list $\mathbf{pref}(v)$, we give all its copies the concatenation of $\mathbf{pref}(v)^{\mu-1} \dots \mathbf{pref}(v)^0$ padded with appropriate number of dummies, where $\mu = \mu_H$ if $v \in \mathcal{R}$ and vice-versa.

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$$\begin{aligned}\mu_H \mu_R |E| &\leq (|\mathcal{R}| + 2) \left(\sum_{h \in \mathcal{H}} q^+(h) + 2 \right) |E| \\ &\leq (|\mathcal{R}| + 2)^2 |E| \\ &= \mathcal{O}(|\mathcal{R}|^2 |E|)\end{aligned}$$

Contents

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An edge $(s, c) \in E \setminus M$ **blocks** M when

- $|M(s)| < q^+(s)$ or $\exists c' \in M(s), c' <_s c$, and
- $|M(c)| < q^+(c)$ or $\exists s' \in M(c), s' <_c s$

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- [Brandl and Kavitha, 2017] give an algorithm for maximum-cardinality popularity, similar to the algorithm in [Huang and Kavitha, 2011].

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This runs in time

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- In [Huang, 2010], **laminar** classes are studied, which is more general than just partitions.

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- A hospital-residents instance with lower quotas.
- Each hospital h has **classes** over its neighbour set, $\mathcal{C}_h \subseteq 2^{E(h)}$, and class quotas $q_h^+, q_h^- : \mathcal{C}_h \rightarrow \mathbb{N}$

This problem is studied in [Huang, 2010]. What we require for our earlier algorithm is a many-to-many version, but for a very special case :

- There are no lower quotas, because we apply this to the transformed instance.
- In [Huang, 2010], **laminar** classes are studied, which is more general than just partitions.
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- All our class upper quotas are 1.
- All courses impose the same classes.

We utilise the first two properties to simplify Huang's algorithm to our purpose.

Stability under Classifications

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Definition

$C_h^r \in \mathcal{C}_h$ be the class in which h puts r . A **feasible blocking pair** is a blocking pair (h, r) such that either $|M(h) \cap C_h^r| < q_h^+(C_h^r)$, or $\exists r' \in M(h) \cap C_h^r, r' <_h r$.

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Definition

M is **class-stable** if it has no feasible blocking pairs.

- In the transformed instance with copies and dummies, each course considers all copies of a student to be a class with upper quota 1.

Application to Students-Courses with Lower Quotas

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- With a few minor observations, our earlier proofs go through for a class-stable (rather than stable) matching in the transformed instance.

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- With a few minor observations, our earlier proofs go through for a class-stable (rather than stable) matching in the transformed instance.
- The many-to-many version can be reduced to the case in Huang's algorithm. However it takes $\mathcal{O}(|E|^2)$ time, which due to successive reductions gives us $\mathcal{O}(|E|^8)$ in our application.
- We adapt Huang's algorithm to a linear-time version specific to our requirement.

Our scenario : Students-Courses, courses impose partition-classes on students, with class upper quotas.

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Huang's algorithm is resident-proposing, i.e. the side that imposes classes receives proposals from the side that is classified. So our algorithm is student-proposing. The algorithm is as follows :

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Correctness : similar to Gale-Shapley.

Contents

- 1 Stable Marriage Problem
- 2 Hospital-Residents Problem
- 3 Popular Matchings
- 4 Hospital-Residents under Lower Quotas
- 5 Both-sided Lower Quotas
- 6 The Students-Courses Problem
- 7 Stability under Classifications
- 8 Generalising Further

Students-Courses with both-sided Lower Quotas

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- But both-sided classifications are not enough – $(s^0, c^2), (s^1, c^3)$ might both be matched in the transformed instance. Fact : this only occurs with one side at priorities 0 and 1.
- We need more sophisticated constraints, like restricted edges or matroid constraints.

Questions

Thank you

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