

Popular Matchings and Optimality

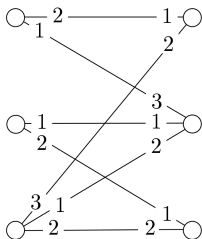
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Université Sorbonne Paris Nord

The input

A bipartite graph where every vertex has a strict ranking of its neighbors.



We seek a matching that pairs vertices **optimally**.

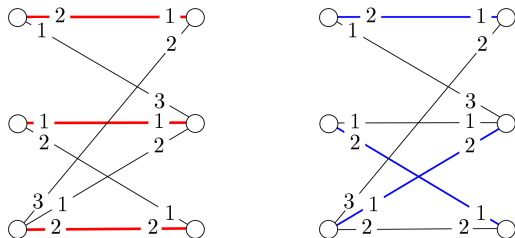
How do we formalize an optimal matching M ?

- ▶ Every agent a should be matched in M to the best possible job b such that b is willing to be matched to a .

Stability

The following property should hold true for any agent/job v :

- ▶ no neighbor ranked better than $M(v)$ is willing to be matched to v .



- ▶ The **red** matching satisfies this property; not the **blue** matching.
- ▶ The edge between the two middle vertices *blocks* the **blue** matching.
- ▶ A matching with no blocking edge is a **stable** matching.

Stable matchings

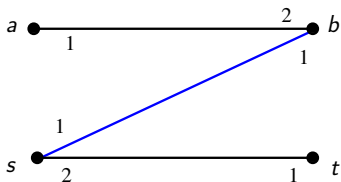
Stability is a very natural notion of optimality in matchings under preferences.

The Gale-Shapley algorithm: **agents propose and jobs dispose** — this is a very simple and clean algorithm.

Stable matchings

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As we saw, the above instance has a stable matching $\{sb\}$ of size 1.

- ▶ So the size of a stable matching might be only half the size of a maximum matching.

Beyond stability

Every vertex v has a ranking over all the possible matchings in G :

- ▶ $M \succ_v N$ if $M(v)$ is **better** than $N(v)$ in v 's ranking;
- ▶ $M \prec_v N$ if $M(v)$ is **worse** than $N(v)$ in v 's ranking;
- ▶ $M \sim_v N$ if $M(v) = N(v)$.

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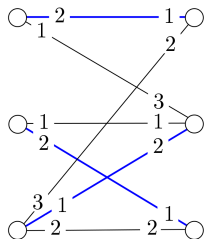
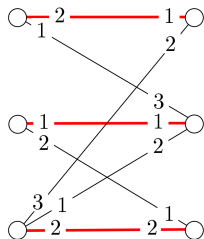
Comparing a pair of matchings:

- ▶ hold a head-to-head election between these two matchings.
- ▶ vertices are voters in this election.
- ▶ count the number of votes won by each matching.
- ▶ the matching with smaller number of votes loses this election.

The \succeq and \succ_v operators

Given two matchings M, N , we write $M \succeq N$ if

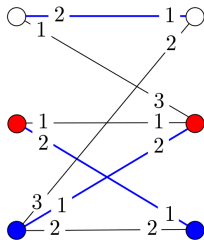
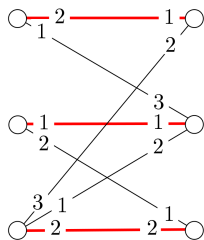
$$|\underbrace{\{v \in A \cup B : M(v) \succ_v N(v)\}}_{\# \text{ of votes for } M}| \geq |\underbrace{\{v \in A \cup B : N(v) \succ_v M(v)\}}_{\# \text{ of votes for } N}|$$



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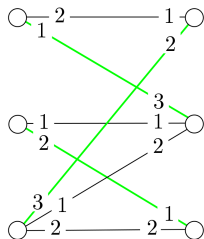
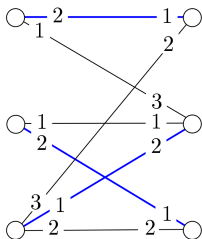
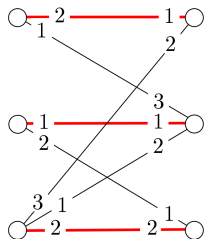
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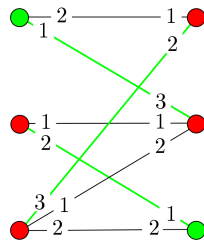
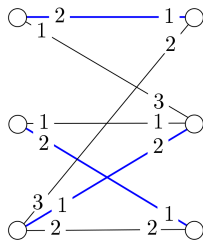
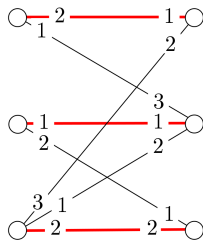
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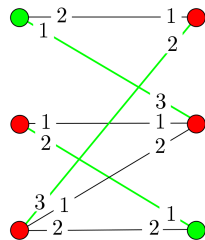
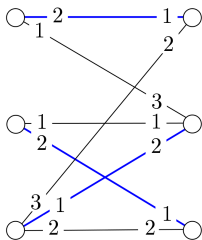
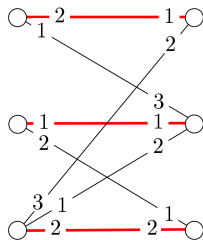


and $M \succ N$ if the inequality is strict (M **defeats** N).

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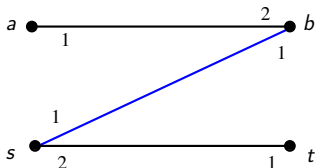
A matching M that does not lose any election is a popular matching.

Max-size popular matching

Every stable matching is popular [Gardenfors, 1975].

- ▶ Every stable matching is a **min-size** popular matching.

Our goal now is to find the matching $\{ab, st\}$ in the instance below.



We know that running the Gale-Shapley algorithm finds $\{sb\}$.

In order to find $\{ab, st\}$, a should get a second chance to propose to b .

- ▶ However b will again reject a .

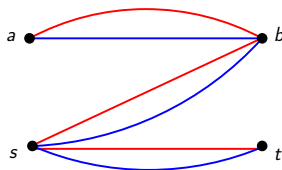
A new instance G'

IDEA. b should prefer a 's second proposal to s 's first proposal.

A new graph G' where every edge uv in G is replaced by uv and uv in G' :

- ▶ that is, by two parallel edges: one red and the other blue.

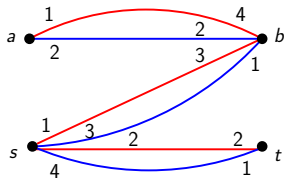
The corresponding graph G' is:



- ▶ Red edges correspond to *first-time* proposals.
- ▶ Blue edges correspond to *second-time* proposals.

A new instance G'

The graph G' with preferences is:



- ▶ Every agent prefers any red edge to any blue edge.
- ▶ Every job prefers any blue edge to any red edge.

Stable matchings in G'

Our algorithm in $G = (A \cup B, E)$

- ▶ Construct the red/blue graph $G' = (A \cup B, E')$.
 - ▶ Run Gale-Shapley algorithm in G' to compute M' .
 - ▶ Return the corresponding matching M in G .
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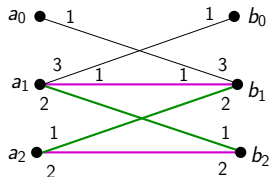
CLAIM. M is popular matching in G .

- ▶ Furthermore, M is more popular than any larger matching.

Thus M is a max-size popular matching in G .

Popular maximum matchings

Max-size popular matchings



There are two max-size popular matchings here: purple and green.

- ▶ Only the green matching occurs as a stable matching in the red/blue graph G' .
- ▶ Hence not every max-size popular matching occurs as a stable matching in the red/blue graph G' .

Dominant matchings

Any stable matching in the red/blue graph G' has an interesting property:

- ▶ M is a popular matching that is more popular than all larger matchings.

DEF. Call a popular matching that defeats all larger matchings dominant.

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Dominant matchings in $G \equiv$ Stable matchings in the red/blue graph G' .
[Cseh and K, 2018]

- ▶ Given $e \in E$: is there a popular matching in $G = (A \cup B, E)$ with e ?

Popularity with a forced edge

FORCED EDGE e

1. Check if there is a **stable matching** in G with edge e .
2. Check if there is a **dominant matching** in G with edge e .
3. If the answer in steps 1 and 2 is no then return “no”.

Popularity with a forced edge

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 2. Check if there is a **dominant matching** in G with edge e .
 3. If the answer in steps 1 and 2 is no then return “no”.
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- ▶ Any popular matching decomposes into a **stable** part and a **dominant** part.
 - ▶ This leads to the correctness of the above algorithm.

Popularity with a forced pair of edges

Given a pair of edges e_1, e_2 in G :

- ▶ Is there a popular matching in G with both e_1 and e_2 ?

There are polynomial time algorithms to determine if there is a stable matching with a given set $\{e_1, \dots, e_k\}$ of edges [Gusfield and Irving, 1989].

- ▶ there need not be a **stable** / **dominant** matching with both e_1 and e_2 .

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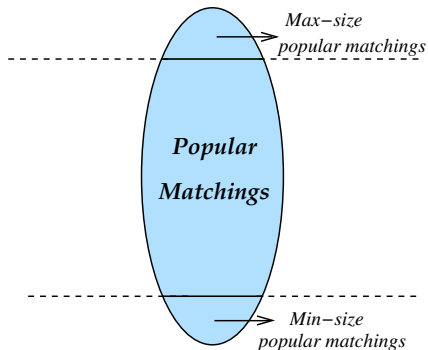
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THEOREM. The above problem is NP-hard [Faenza, K, Powers, Zhang, 2019].

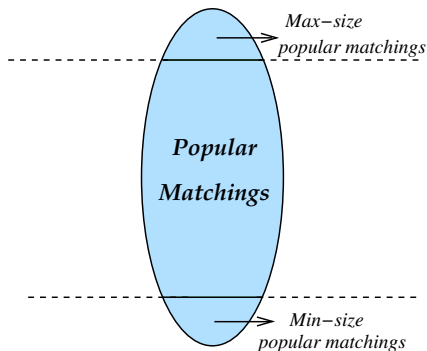
The easy subclasses

Finding a **max-size** (similarly, **min-size**) popular matching in G is easy.



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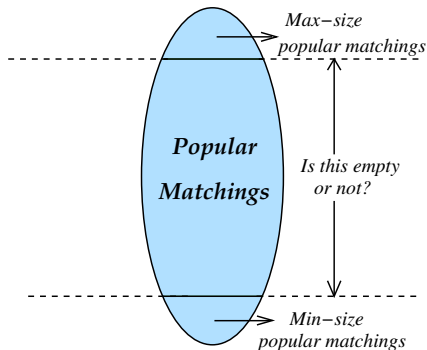
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Is there *any* popular matching in G that is not **max-size** or **min-size**?

The easy subclasses

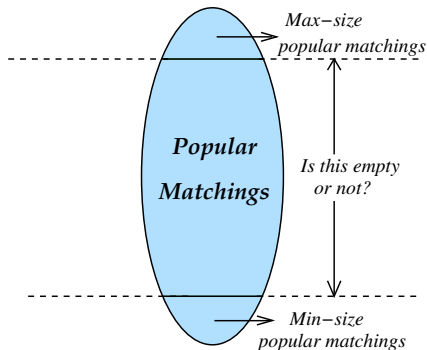
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Surprisingly, this is an **NP-hard** problem.

Min-cost popular matchings

Suppose there is a function $\text{cost} : E \rightarrow \{0, 1\}$.

- ▶ It is NP-hard to find a min-cost popular matching.

Finding a min-cost stable matching in G is easy.

- ▶ Efficient combinatorial [Irving, Leather, and Gusfield, 1987] and LP-based [Vande Vate, 1987; Rothblum, 1992] algorithms.
- ▶ The stable matching polytope has a linear-size description.

A relaxation of the min-cost popular matching problem:

- ▶ the min-cost popular **mixed matching** problem.

Mixed matchings

A mixed matching is a probability distribution over matchings, i.e.,

$$\Pi = \{(M_0, p_0), \dots, (M_k, p_k)\},$$

where M_0, \dots, M_k are matchings in G and $\sum_i p_i = 1$ and $p_i \geq 0 \forall i$.

- ▶ A mixed matching is a lottery over matchings.

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- ▶ A mixed matching is a lottery over matchings.

For any two matchings M and N :

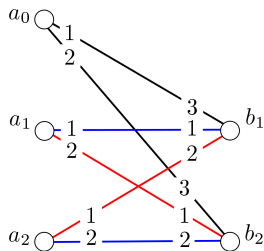
$$\text{let } \Delta(N, M) = \# \text{ of votes for } N - \# \text{ of votes for } M.$$

- ▶ Define $\Delta(N, \Pi) = \sum_i p_i \cdot \Delta(N, M_i)$.

DEFINITION. Call a mixed matching Π popular if $\Delta(N, \Pi) \leq 0 \forall$ matchings N .

An interesting example

This instance has only one popular matching $S = \{a_1 b_1, a_2 b_2\}$.



$M = \{a_1 b_2, a_2 b_1\}$ is unpopular; it is defeated by $N = \{a_0 b_2, a_1 b_1\}$.

▶ However the mixed matching $\Pi = \{(S, \frac{1}{2}), (M, \frac{1}{2})\}$ is popular:

▶ observe that $\Delta(N, \Pi) = \frac{1}{2} \cdot \Delta(N, S) + \frac{1}{2} \cdot \Delta(N, M) = \frac{1}{2} - \frac{1}{2} = 0$.

Thus a popular mixed matching need not be a probability distribution over popular matchings.

Finding a min-cost popular mixed matching

In bipartite graphs: a mixed matching \equiv a fractional matching
[Birkhoff-von Neumann theorem].

A fractional matching is a point \vec{x} in the matching polytope \mathcal{M}_G ,
i.e., $\vec{x} \in \mathbb{R}_{\geq 0}^m$ and $\sum_{e \in \delta(u)} x_e \leq 1$ for all $u \in A \cup B$.

Let $\mathcal{F}_G = \{\vec{x} \in \mathcal{M}_G : \Delta(N, \vec{x}) \leq 0 \text{ for all matchings } N\}$

$$\text{where } \Delta(N, \vec{x}) = \sum_{u \in A \cup B} \underbrace{\left(\sum_{v \prec_u N(u)} x_{uv} - \sum_{v \succ_u N(u)} x_{uv} \right)}_{u\text{'s vote for } N(u) \text{ vs } \vec{x}_u}$$

- ▶ \mathcal{F}_G is the popular fractional matching polytope of G .

Towards a compact extended formulation for \mathcal{F}_G

For any $\vec{x} \in \mathcal{M}_G$ and $e = (a, b) \in E$:

- ▶ Let $\text{wt}_x(e) = (a\text{'s vote for } b \text{ versus } \vec{x}_a) + (b\text{'s vote for } a \text{ versus } \vec{x}_b)$.

$$\text{So } \text{wt}_x(e) = \left(\sum_{b' \prec_a b} x_{ab'} - \sum_{b' \succ_a b} x_{ab'} \right) + \left(\sum_{a' \prec_b a} x_{a'b} - \sum_{a' \succ_b a} x_{a'b} \right).$$

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Similarly, let $\text{wt}_x(uu) = (u\text{'s vote for itself versus } \vec{x}_u)$ for any $u \in A \cup B$.

- ▶ So $\text{wt}_x(uu) = - \sum_{e \in \delta(u)} x_e$.

Thus $\Delta(N, \vec{x}) = \text{wt}_x(N)$ for any perfect matching N .

(note that N is augmented with self-loops to make it perfect)

- ▶ Hence \vec{x} is popular $\iff \text{wt}_x(N) \leq 0$ for all perfect matchings N .

LP for max-weight perfect matching

$$\max \sum_e \text{wt}_x(e) \cdot y_e$$

$$\sum_{e \in \delta(u) \cup \{uu\}} y_e = 1 \quad \forall u \in A \cup B$$

$$y_e \geq 0 \quad \forall e \in E \cup \{\text{self-loops}\}.$$

\vec{x} is popular \iff $\text{opt} = 0$.

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\vec{x} is popular \iff $\text{opt} = 0$.

The above LP allows us to test if the fractional matching \vec{x} is popular or not.

- ▶ But we want to describe the polytope of all popular fractional matchings.

So we also have to add the constraints $\sum_{e \in \delta(u) \cup \{uu\}} x_e = 1 \quad \forall u$ and $x_e \geq 0 \quad \forall e$.

- ▶ The objective function becomes *quadratic* in variables x_e and y_e .

LP for max-weight perfect matching

Dual LP

$$\min \sum_u \alpha_u$$

$$\alpha_a + \alpha_b \geq \text{wt}_x(ab) \quad \forall ab \in E$$

$$\alpha_u \geq \text{wt}_x(uu) \quad \forall u \in A \cup B.$$

\vec{x} is popular $\iff \exists$ dual feasible $\vec{\alpha}$ with $\sum_u \alpha_u = 0$.

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Let us add the constraints $\sum_{e \in \delta(u) \cup \{uu\}} x_e = 1 \quad \forall u$ and $x_e \geq 0 \quad \forall e$ to this LP.

- ▶ So this will be an LP in variables x_e 's and α_u 's.
- ▶ The set of optimal solutions $(\vec{x}, \vec{\alpha})$ will be an extension of the polytope \mathcal{F}_G .

The popular fractional matching polytope \mathcal{F}_G

A compact extended formulation

$$\begin{aligned}\alpha_a + \alpha_b &\geq \text{wt}_x(ab) \quad \forall ab \in E \\ \alpha_u &\geq \text{wt}_x(uu) \quad \forall u \in A \cup B \\ \sum_{e \in \delta(u) \cup \{uu\}} x_e &= 1 \quad \forall u \in A \cup B \\ x_e &\geq 0 \quad \forall e \in E \cup \{\text{self-loops}\} \\ \sum_{u \in A \cup B} \alpha_u &= 0.\end{aligned}$$

We can optimize over the above polytope (call it \mathcal{F}'_G) in polynomial time.

Thus we can find a **min-cost popular mixed matching** in polynomial time.

- ▶ A drawback of generalizing to mixed matchings is that the solution has become more complex to describe and difficult to implement.

The polytope \mathcal{F}'_G

The LP whose set of optimal solutions is \mathcal{F}'_G has an unusual property:

- ▶ It is **self-dual**.
- ▶ So $x_{ab} > 0 \implies \alpha_a + \alpha_b = \text{wt}_x(ab)$.
(by complementary slackness)

Suppose G admits a **perfect stable matching**.

- ▶ Then every popular matching in G has a witness in $\{\pm 1\}^n$.

We show that \mathcal{F}'_G (and thus \mathcal{F}_G) is integral in this special case.

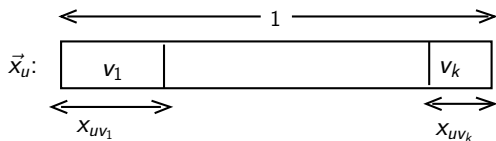
[Huang and K, 2021]

Our method is inspired by the proof of integrality of the formulation of the stable matching polytope [Teo and Sethuraman, 1998].

Integrality of \mathcal{F}_G in the special case

Let $\vec{x} \in \mathcal{F}_G$. Then $(\vec{x}, \vec{\alpha}) \in \mathcal{F}'_G$ for some $\vec{\alpha} \in [-1, 1]^n$.

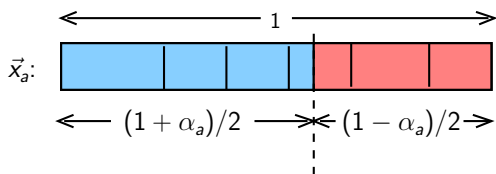
$\vec{x}_u = (x_{uv_1}, \dots, x_{uv_k})$ is u 's allocation in \vec{x} , for every $u \in A \cup B$.



So v_1, \dots, v_k are the neighbors of u such that $x_{uv_i} > 0$ for $i = 1, \dots, k$.

- ▶ For $a \in A$: arrange the entries of \vec{x}_a in **decreasing** order of a 's preference.
- ▶ For $b \in B$: arrange the entries of \vec{x}_b in **increasing** order of a 's preference.

Reordering the array \vec{x}_a



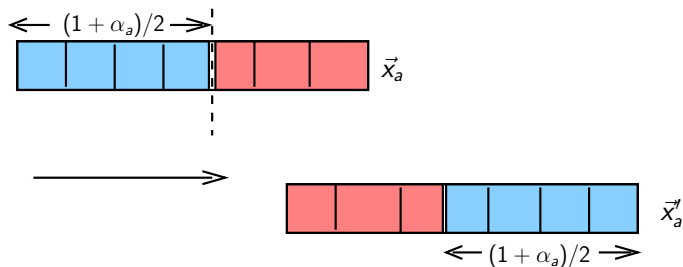
Call the initial $(1 + \alpha_a)/2$ fraction of \vec{x}_a the **blue sub-array** of \vec{x}_a .

Call the remaining $(1 - \alpha_a)/2$ fraction of \vec{x}_a the **red sub-array** of \vec{x}_a .

Swap the **blue** and **red** sub-arrays to form a reordered array \vec{x}'_a .

Reordering the array \vec{x}_a

Thus $\vec{x}_a \rightsquigarrow \vec{x}'_a$ by this swap.



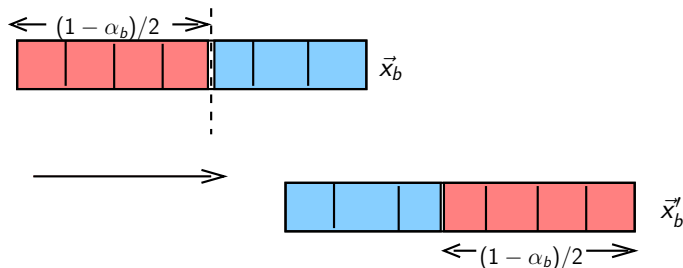
The order within the **blue** sub-array (similarly, the **red** sub-array) is preserved.

Reordering the array \vec{x}_b

We do an analogous transformation for $\vec{x}_b \rightsquigarrow \vec{x}'_b$ for any $b \in B$.

The initial $(1 - \alpha_b)/2$ fraction of \vec{x}_b will be the **red sub-array** of \vec{x}_b .

The latter $(1 + \alpha_b)/2$ fraction of \vec{x}_b will be the **blue sub-array** of \vec{x}_b .



Swap the **red** and **blue** sub-arrays to form a reordered array \vec{x}'_b .

The table T

Form a table T whose rows are the reordered arrays.

\vec{x}'_a					
\vec{x}'_b					
\vec{x}'_c					
\vec{x}'_u					
\vec{x}'_v					

Sweep a vertical line along the table T decomposing it into columns.

$M_i =$ pairing defined by the i th column.

The table T

Form a table T whose rows are the reordered arrays.

\vec{x}'_a					
\vec{x}'_b					
\vec{x}'_c					
\vec{x}'_u					
\vec{x}'_v					

Sweep a vertical line along the table T decomposing it into columns.

$M_1 =$ pairing defined by the first column.

The table T

Form a table T whose rows are the reordered arrays.

\vec{x}'_a							
\vec{x}'_b							
\vec{x}'_c							
\vec{x}'_u							
\vec{x}'_v							

Sweep a vertical line along the table T decomposing it into columns.

$M_2 =$ pairing defined by the second column.

The table T

Form a table T whose rows are the reordered arrays.

\vec{x}'_a	█	█	█						
\vec{x}'_b	█	█	█						
\vec{x}'_c	█	█	█						
\vec{x}'_u	█	█	█						
\vec{x}'_v	█	█	█						

Sweep a vertical line along the table T decomposing it into columns.

$M_3 =$ pairing defined by the third column.

The table T

Form a table T whose rows are the reordered arrays.

\vec{x}'_a							
\vec{x}'_b							
\vec{x}'_c							
\vec{x}'_u							
\vec{x}'_v							

Sweep a vertical line along the table T decomposing it into columns.

$M_4 =$ pairing defined by the fourth column.

Popularity of M_i

Thus $\vec{x} = \sum_i p_i \cdot M_i$, where p_i is the width of M_i 's column.

- ▶ Self-duality of our LP \Rightarrow each M_i is a matching in G .

We show a dual certificate $\vec{\alpha}_i \in \{\pm 1\}^n$ for each M_i .

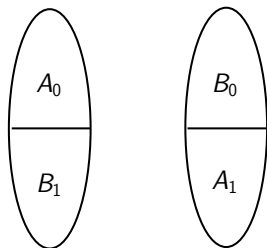
- ▶ Hence each M_i is a popular matching.

Thus we have $(\vec{x}, \vec{\alpha}) = \sum_i p_i \cdot (M_i, \vec{\alpha}_i)$ where $\sum_i p_i = 1$ and $p_i \geq 0 \forall i$.

- ▶ Hence \mathcal{F}'_G (and thus \mathcal{F}_G) is integral.
- ▶ This is for the special case when G admits a perfect stable matching.

Half-integrality of \mathcal{F}_G in the general case

Corresponding to $G = (A \cup B, E)$, let us build the following graph H :



H is made up of 2 copies of G along with the “self-loop” edges $u_0 u_1 \forall u$.

- ▶ H admits a perfect stable matching.
- ▶ So \mathcal{F}_H is integral.

Half-integrality of \mathcal{P}_G in the general case

We can define simple maps $f : \mathcal{F}_G \rightarrow \mathcal{F}_H$ and $h : \mathcal{F}_H \rightarrow \mathcal{F}_G$ such that
$$h \circ f(\vec{x}) = \vec{x} \text{ for any } \vec{x} \in \mathcal{P}_G.$$

Integrality of $\mathcal{F}_H \implies f(\vec{x}) = \{(M_i, p_i) : i = 1, \dots, k\}$ for popular matchings
 M_1, \dots, M_k in H .

► So $\vec{x} = \{(h(M_i), p_i) : i = 1, \dots, k\}$.

Each $h(M_i)$ will be a popular half-integral matching in G .

- Thus the popular fractional matching polytope \mathcal{F}_G is half-integral.
- Hence there is always a min-cost popular mixed matching
 $\Pi = \{(M_0, \frac{1}{2}), (M_1, \frac{1}{2})\}$.
- To implement the lottery given by Π , we need just one random bit.

The popular matching polytope

Let \mathcal{P}_G be the popular matching polytope of G .

- ▶ The extension complexity of \mathcal{P}_G is $2^{\Omega\left(\frac{m}{\log m}\right)}$, where $|E| = m$.
[Faenza and K, 2022]

Can we relax *popular* to “approximately popular” for the sake of tractability?

- ▶ M is popular \implies no matching wins more votes than M .

DEFINITION. Call a matching M **quasi-popular** if no matching wins more than twice as many votes as M .

- ▶ Can a min-cost quasi-popular matching be efficiently computed?
- ▶ No, we show it is **NP-hard** to compute a min-cost quasi-popular matching.

Quasi-popular matchings

The min-cost **popular** / **quasi-popular** matching problems are NP-hard to approximate up to any factor.

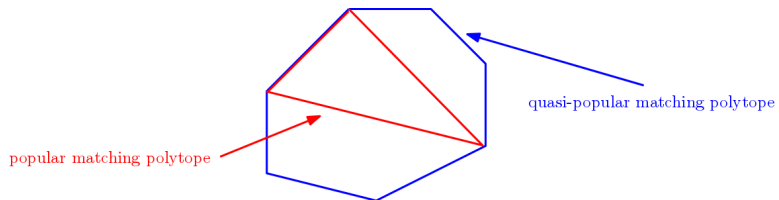
- ▶ A BICRITERIA APPROXIMATION. Can we efficiently find a **quasi-popular** matching of cost at most that of a min-cost **popular** matching?

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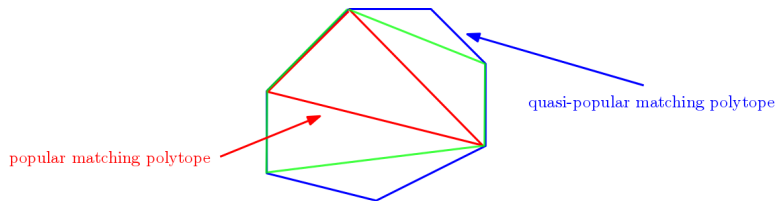
- ▶ A BICRITERIA APPROXIMATION. Can we efficiently find a **quasi-popular** matching of cost at most that of a min-cost **popular** matching?
- ▶ Interestingly, the answer is yes.

Our technique



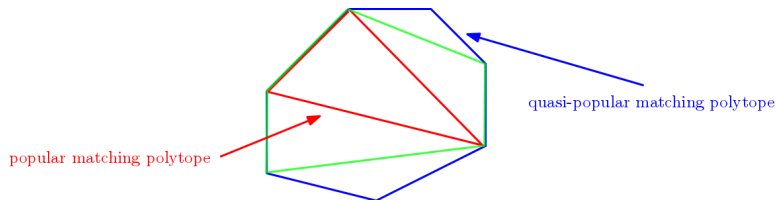
- ▶ The polytopes \mathcal{P}_G and \mathcal{Q}_G have near-exponential extension complexity.

Our technique



- ▶ The polytopes \mathcal{P}_G and \mathcal{Q}_G have near-exponential extension complexity.
- ▶ We show an **integral polytope** \mathcal{C} sandwiched between \mathcal{P}_G and \mathcal{Q}_G such that \mathcal{C} has a compact extended formulation.

Our technique



- ▶ The polytopes \mathcal{P}_G and \mathcal{Q}_G have near-exponential extension complexity.
- ▶ We show an **integral polytope** \mathcal{C} sandwiched between \mathcal{P}_G and \mathcal{Q}_G such that \mathcal{C} has a compact extended formulation.
- ▶ Optimizing over \mathcal{C} leads to the efficient bicriteria algorithm.

The popular matching polytope in a special case

Special case:

- ▶ G has a perfect stable matching $\implies \mathcal{F}_G$ is integral.

From general case to special case:

- ▶ Assume for simplicity that $|A| = |B|$.
- ▶ **Our idea:** augment G with some **new edges** so that the resulting graph G^* has a perfect stable matching.

The popular matching polytope in a special case

Special case:

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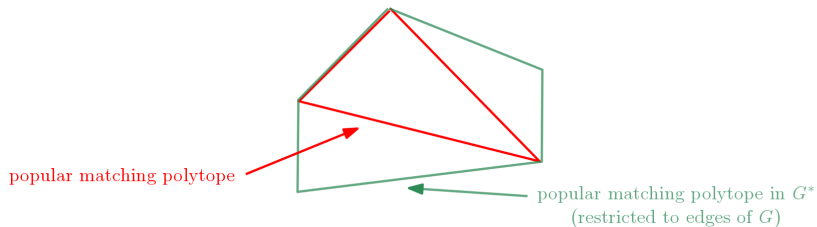
From general case to special case:

- ▶ Assume for simplicity that $|A| = |B|$.
- ▶ **Our idea:** augment G with some **new edges** so that the resulting graph G^* has a perfect stable matching.
- ▶ Pair up unstable vertices appropriately; add a new edge between each pair.

So the popular matching polytope of G^* has a compact extended formulation.

The sandwiched integral polytope

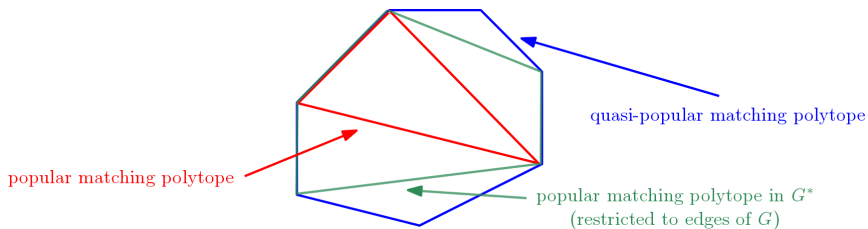
Every **popular** matching in G can be extended to a perfect popular matching in G^* using these new edges.



The sandwiched integral polytope

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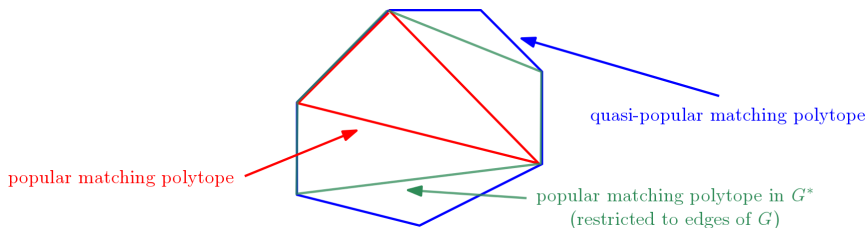
Every popular matching in G^* , when restricted to edges of G , is **quasi-popular** in G .



The sandwiched integral polytope

Every **popular** matching in G can be extended to a perfect popular matching in G^* using these new edges.

Every popular matching in G^* , when restricted to edges of G , is **quasi-popular** in G .



G^* has a perfect stable matching, so $\mathcal{F}_{G^*} =$ popular matching polytope of G^* .

The compact extended formulation of \mathcal{F}_{G^*} is an extension of an **integral polytope** \mathcal{C} where: $\mathcal{P}_G \subseteq \mathcal{C} \subseteq \mathcal{Q}_G$.

Another relaxation of popularity

Suppose N is a “very unpopular” matching.

- ▶ Let us not give N the power to block other matchings.
- ▶ That is, we waive the constraint $\Delta(N, \vec{x}) \leq 0$ in the popular matching polytope formulation.

How do we define “very unpopular” matchings?

Another relaxation of popularity

Suppose N is a “very unpopular” matching.

- ▶ Let us not give N the power to block other matchings.
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How do we define “very unpopular” matchings?

Call a matching S supporting if \exists a popular mixed matching Π whose support contains S .

So \exists popular $\Pi = \{(M_0, p_0), \dots, (M_k, p_k)\}$ such that $S = M_i$ for some i .

- ▶ A matching that is not supporting will be considered “very unpopular”.

Call a matching M **fairly popular** if $\Delta(S, M) \leq 0 \forall$ supporting matchings S .

Fairly popular matchings

Can a min-cost fairly popular matching be computed in polynomial time?

- ▶ Interestingly, the answer is “yes”.

The following statements are equivalent.

- ▶ S is a supporting matching.
- ▶ No popular mixed matching defeats S .
- ▶ S matches all stable vertices and $S \subseteq E_p$ where E_p is the set of popular fractional edges in G .

The set E_p is the set of popular edges in the graph H (this is two copies of G).

Fairly popular matchings

This characterization of supporting matchings allows us to show:

- ▶ M is fairly popular $\iff M$ can be realized as a stable matching in a certain colorful multigraph.
 - ▶ Thus the min-cost stable matching algorithm in this multigraph finds a min-cost fairly popular matching in G in polynomial time.
-

Fairly popular matchings

This characterization of supporting matchings allows us to show:

- ▶ M is fairly popular $\iff M$ can be realized as a stable matching in a certain colorful multigraph.
- ▶ Thus the min-cost stable matching algorithm in this multigraph finds a min-cost fairly popular matching in G in polynomial time.

Thus it is easy to decide if a given matching is undefeated by all popular mixed matchings.

- ▶ What is the complexity of deciding if a given matching is undefeated by all popular matchings?
- ▶ This is a coNP-hard problem.

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Thank you! Any questions?