

Matching Games

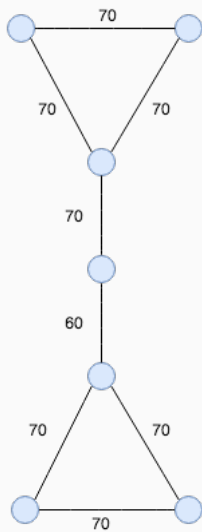
Computing the Nucleolus in Polynomial Time

By Justin Toth

Joint work with Jochen Könemann and Kanstantsin Pashkovich

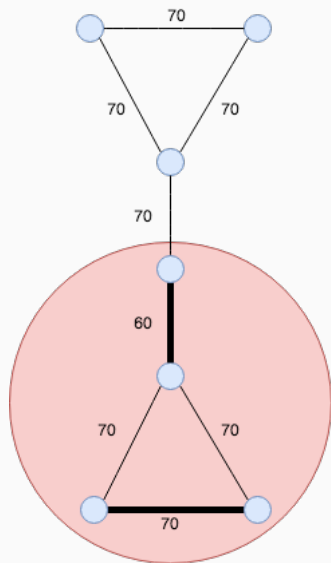
University of Waterloo

Weighted Matching Games



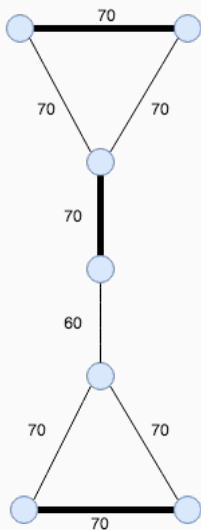
- **Vertices:** Players
- **Edges:** Potential Collaborations
- Have ≤ 1 collaborator
- Models Network Bargaining setting of [Kleinberg and Tardos '08]

Weighted Matching Games



- Value $\nu(S) = \max\{w(M) : M \text{ is a matching on } G[S]\}$
- $S \subseteq T \implies \nu(S) \leq \nu(T)$
- So assume the **grand coalition** forms.

Least Core



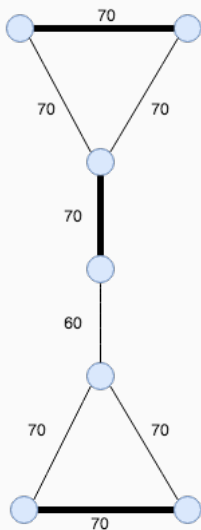
How to **fairly distribute** matching value to the players?

$$\sum_{i \in S} p_i \geq \nu(S) \quad \forall S \subseteq V$$

$$\sum_{i \in V} p_i = \nu(V)$$

$$p_i \geq 0 \quad \forall i \in V$$

Least Core



$$\nu(V) = 70 + 70 + 70 = 210$$

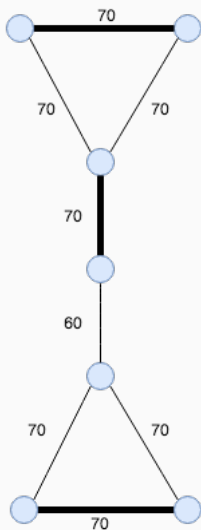
$$\nu_{\text{frac}}(V) = 7 \cdot 35 = 245$$

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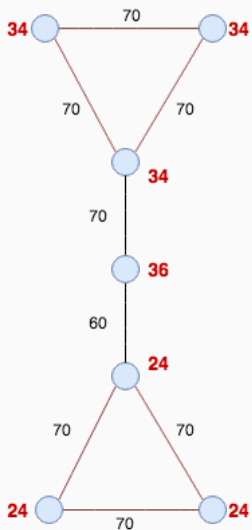
min ϵ

$$\text{s.t. } \sum_{i \in S} p_i \geq \nu(S) - \epsilon \quad \forall S \subseteq V$$

$$\sum_{i \in V} p_i = \nu(V)$$

$$p_i \geq 0 \quad \forall i \in V$$

An Outcome



$$\nu(V) = 210$$

$$\epsilon_1 = 70 - 2 \cdot 34 + 70 - 2 \cdot 24 = 24$$

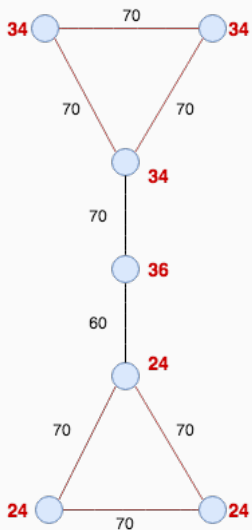
$$\min \epsilon$$

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An Outcome



Problems with the Least Core:

- outcomes are **not unique**
- **Excess:** $p(S) - \nu(S)$ only controlled for worst coalitions

The Nucleolus

The *nucleolus* is *fairest* allocation p^*

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Sort excesses: $ex_p(S_1) \geq ex_p(S_2) \geq \dots \geq ex_p(S) \geq \dots =: \theta(p)$

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p^* lexicographically minimizes $\theta(p)$

The *nucleolus* is fairest allocation p^*

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p^* lexicographically minimizes $\theta(p)$

And it is **unique!**

Hardness of Computing the Nucleolus

- Shortest Path Games
- Node-Weighted Matching
- Convex Games
- Spanning Tree Games
- Flow Games
- Weighted Voting Games
- Linear Production Games

Hardness of Computing the Nucleolus

In P

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NP-hard

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Open for 15 years: How about Weighted Matching Games?

[Kern and Paulusma '03]

Hardness of Computing the Nucleolus

In P

- Shortest Path Games
- Node-Weighted Matching
- Convex Games
- **Our Result:** Matching Games

NP-hard

- Spanning Tree Games
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- Weighted Voting Games
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Open for 15 years: How about Weighted Matching Games?

[Kern and Paulusma '03]

Maschler's Scheme

Solve a **hierarchy** of linear programs

$$P_1 \supseteq P_2 \supseteq \cdots \supseteq P_{|V|} \equiv p^*$$

Subsequent LPs **minimize excess** over coalitions **not fixed**

$$P_2 : \min \epsilon$$

$$\text{s.t. } p \in P_1$$

$$p(S) \geq \nu(S) - \epsilon \quad \forall S \notin \text{Fix}(P_1)$$

$$\text{Fix}(P_1) = \{S \subset V : \exists c \text{ s.t. } p(S) = c, \quad \forall p \in P_1\}$$

Maschler's Scheme

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Each LP **fixes** new coalitions \implies **dimension** of P_{i+1} reduces

$$P_2 : \min \epsilon$$

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$O(2^{|V|})$ constraints in **naive formulation** of each P_i

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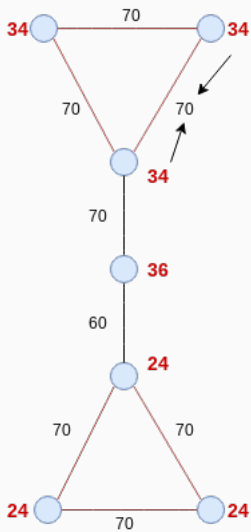
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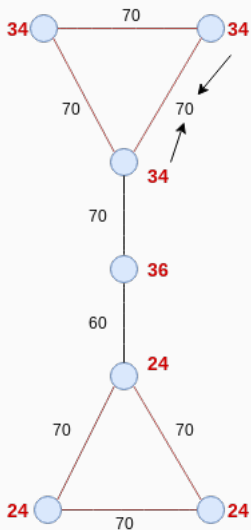
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Separation for P_1



$$\begin{aligned}
 & \min \epsilon \\
 \text{s.t. } & \sum_{i \in S} p_i \geq \nu(S) - \epsilon \quad \forall S \subseteq V \text{ (???)} \\
 & \sum_{i \in V} p_i = \nu(V) \quad \text{(Easy!)} \\
 & p_i \geq 0 \quad \forall i \in V \text{ (Easy!)}
 \end{aligned}$$

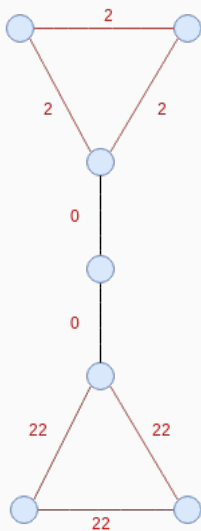
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$$\bar{w}(uv) = w(uv) - p_u - p_v$$

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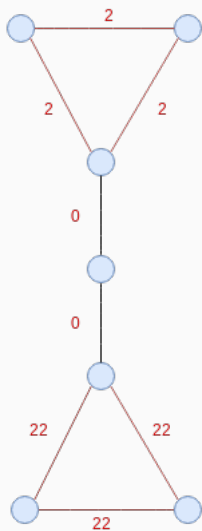


$$\bar{w}(uv) = w(uv) - p_u - p_v \text{ Min}$$

Excess Matchings: $\max \bar{w}(M)$

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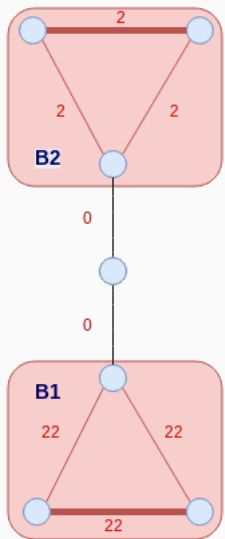
$$\bar{w}(uv) = w(uv) - p_u - p_v \text{ Min}$$

Excess Matchings: $\max \bar{w}(M)$

Solve using your favourite **max weight matching** algorithm!

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Describing Constraints of P_1



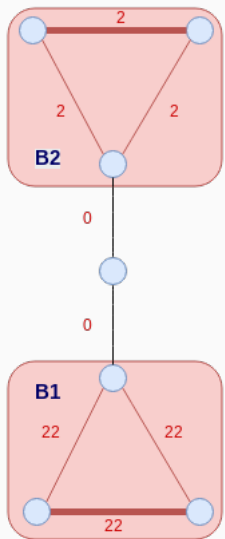
Tight P_1 constraints \equiv optimal \bar{w} matchings

Optimal Matchings defined by:

$$x(E(B)) = \frac{|B| - 1}{2} \quad \forall B \in \mathcal{L}$$

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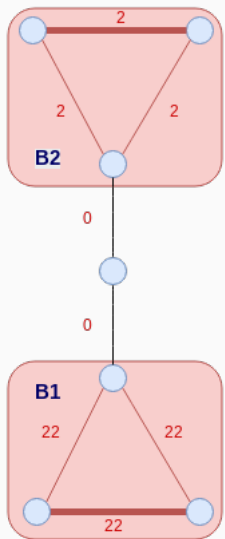
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Can show $T = \emptyset$ when $\epsilon_1 > 0$

Describing Constraints of P_1



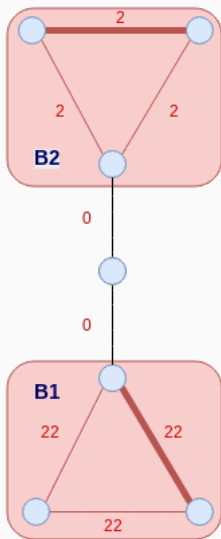
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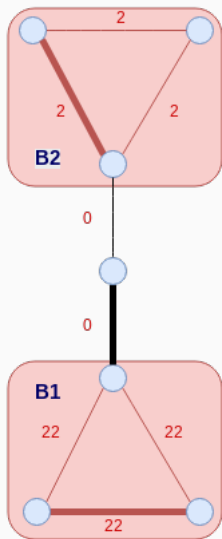
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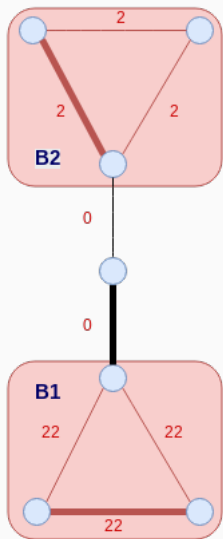
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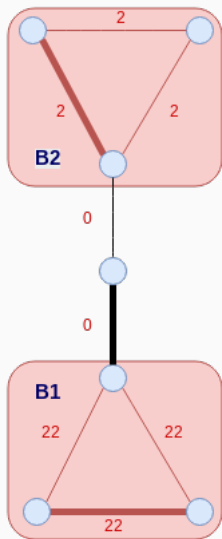
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Need something for all $p \in P_1$

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Universal Matchings

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Max $\bar{w}(uv) := w(uv) - \bar{p}_u - \bar{p}_v$ weight matchings describe *universal matchings*.

Studying the structure of *universal matchings* will be key!

Three Ideas for Description

Consider laminar blossom family \mathcal{L} for \bar{p}

(Recall: Matching M is optimal $\iff |M \cap E(B)| = \frac{|B|-1}{2} \quad \forall B \in \mathcal{L}$)

Three Ideas for Description

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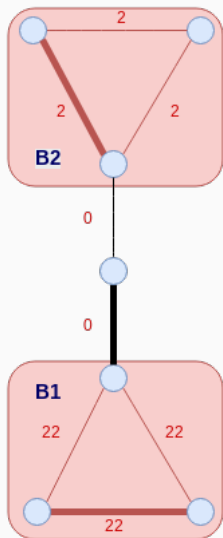
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Theorem: These linear constraints suffice to describe Least Core P_1

Observations 1 and 2

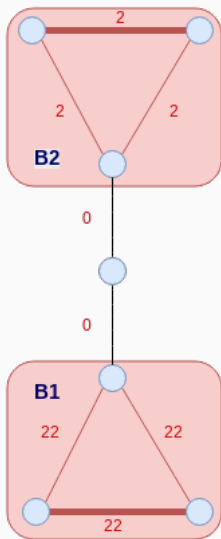


1. \bar{w} values > 0 on blossom edges.

Proof Idea: Drop non-positive edge e from M : Optimal but
 $|M - e \cap E(B)| < |M \cap E(B)| = \frac{|B|-1}{2}$



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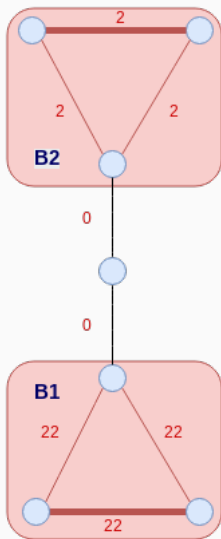
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2. \bar{w} values ≤ 0 on non-blossom edges.

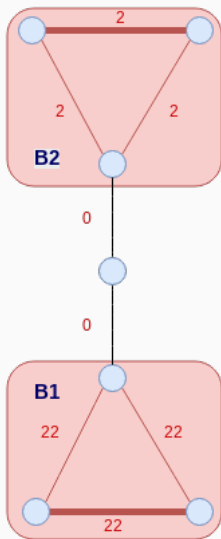
Proof Idea: Can satisfy \mathcal{L} constraints without such edges. ■

Observation 3



Need to understand \bar{w} distribution
over each $B \in \mathcal{L}$

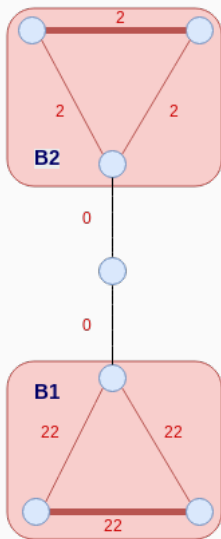
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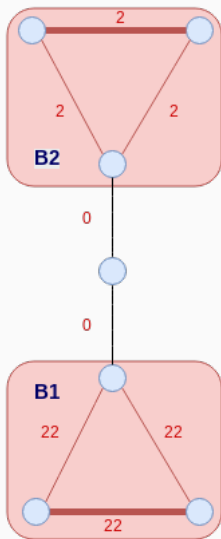


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Fails for larger examples!

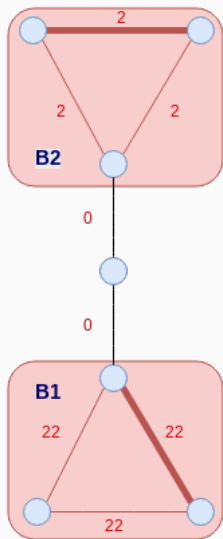
Observation 3



Actually: $\forall u, v \in B, \forall B \in \mathcal{L}$

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Technical Proof. **Key Idea:**
 M exposing u can be modified
within B to expose v .

Least Core Description

Obtain a nice P_1 description

$$\begin{array}{ll} \min & \epsilon \\ \text{s.t.} & p_v - \bar{p}_v = p_u - \bar{p}_u \quad \text{for all } u, v \in S_i^*, i \in [k] \\ & p(e) - w(e) \leq 0 \quad \text{for all } e \in E^* \\ & p(e) - w(e) \geq 0 \quad \text{for all } e \in E^+ \\ & p(M^*) - w(M^*) = -\epsilon \quad M^* \text{ a universal matching} \\ & p(V) = \nu(G) \\ & p \geq 0 \end{array}$$

Where S_1^*, \dots, S_k^* are maximal sets of \mathcal{L} ,

$$E^* := \left(\bigcup_{i \in [k]} E(S_i^*) \right) \cap \left(\bigcup_{M \in \mathcal{M}^*} M \right),$$

and E^+ is the edges with at most one endpoint each S_i^* .

Implementing Maschler's Scheme

Solve a **hierarchy** of linear programs

$$P_1 \supseteq P_2 \supseteq \cdots \supseteq P_{|V|} \equiv p^*$$

To find **nucleolus** = p^* give a polynomial sized formulation for each P_i

$$\begin{aligned} P_2 : \min \epsilon \\ \text{s.t. } p \in P_1 \\ p(S) \geq \nu(S) - \epsilon \quad \forall S \notin \text{Fix}(P_1) \end{aligned}$$

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Implementing Maschler's Scheme

Observation 3 (Blossom Symmetry) plays a major role.

$$\forall u, v \in B, \forall B \in \mathcal{L}, p_u - \bar{p}_u = p_v - \bar{p}_v$$

Lemma: $B \in \mathcal{L} \setminus \text{Fix}(P_1) \implies v \notin \text{Fix}(P_1) \quad \forall v \in B$

$$P_2 : \min \epsilon$$

$$\text{s.t. } p \in P_1$$

$$p(e) - w(e) \geq \epsilon - \epsilon_1 \quad \forall e \in E^+, e \notin \text{Fix}(P_1)$$

$$p_v \geq \epsilon - \epsilon_1 \quad \forall v \notin \text{Fix}(P_1)$$

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