Breaking the Envy Cycle: Best-of-Both-Worlds Guarantees for Subadditive Valuations

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Joint work with







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Problem

Allocating limited resources via fair randomized lotteries

Affordable Housing



School Admissions



Residence Permits



Problem

Allocating limited resources via fair randomized lotteries

Our model: agents with equal entitlement and diverse preferences over resources

Simple Model: Unit-Demand

Agents have unit-demand valuations:

 $\boldsymbol{\nu}_i(X_i) = \max_{j \in X_i} \boldsymbol{\nu}_{i,j}$

 X_i is agent *i*'s allocation

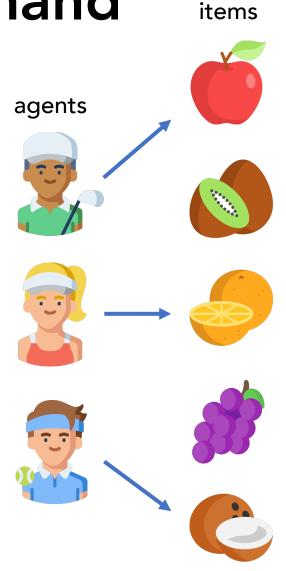
A randomized lottery X is a probability distribution over integral allocations.

X is ex-ante envy-free (EF) if: $\mathbb{E}[v_i(X_i)] \ge \mathbb{E}[v_i(X_j)]$

ex ante = before randomization

Ex-ante EF exists via simultaneous eating.

[Bogomolnaia, Moulin, 2001]



Intermediate Model: Additive

Agents have additive valuations:

$$v_i(X_i) = \sum_{j \in X_i} v_{i,j}$$

 X_i is agent *i*'s allocation

X is ex-ante envy-free (EF) if $\mathbb{E}[v_i(X_i)] \ge \mathbb{E}[v_i(X_j)]$.

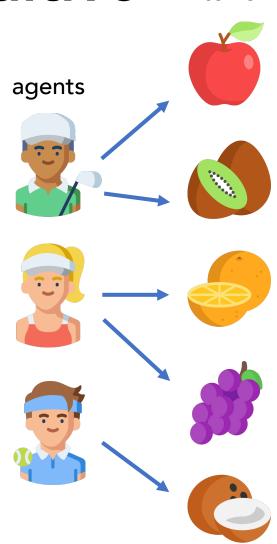
ex-ante = before randomization

X is ex-post EF1 if every realization satisfies: $v_i(X_i) \ge v_i(X_j - g)$ for some $g \in X_i$

ex-post = after randomization

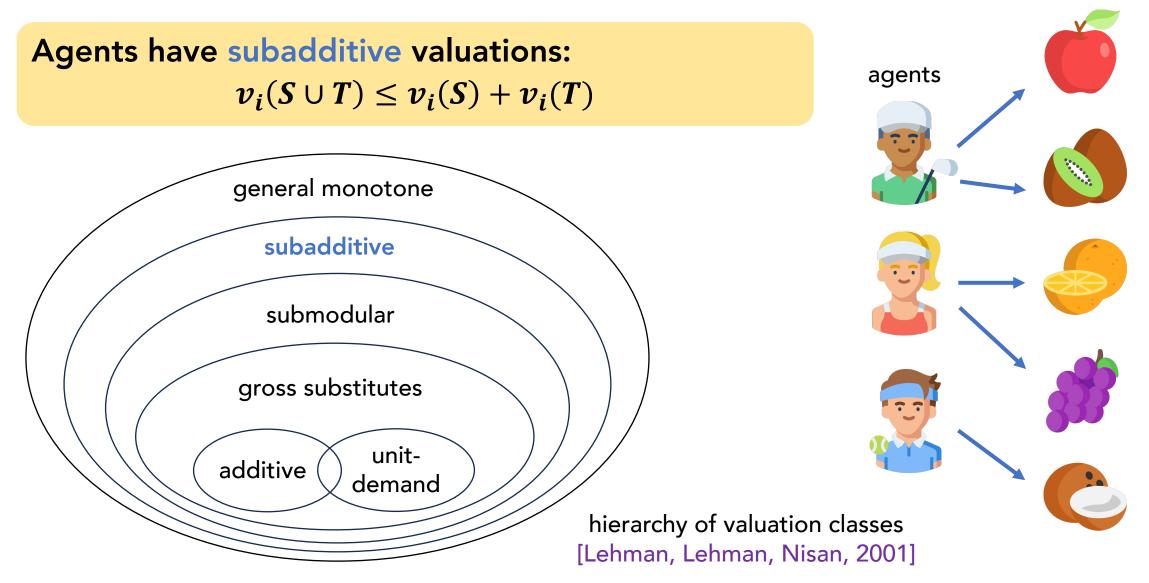
Best-of-both-worlds guarantee: Ex-ante EF and ex-post EF1 exists for additive vals.

[Freeman, Shah, Vaish, 2020] and [Aziz, 2020]



items

Combinatorial Model: Subadditive



Starting point: [Freeman, Shah, Vaish, 2020] and [Aziz, 2020] Ex-ante EF and ex-post EF1 exists for additive vals.

 $\left(\mathbb{E}[v_i(X_i)] \ge \mathbb{E}[v_i(X_j)] \right) \quad \left(v_i(X_i) \ge v_i(X_j - g) \text{ for some } g \in X_i \right) \quad \left(v_i(X_i) = \sum_{X_i} v_{i,j} \right)$

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Q1: Can we get subadditive vals?

 $(v_i(S \cup T) \le v_i(S) + v_i(T))$

Starting point: [Freeman, Shah, Vaish, 2020] and [Aziz, 2020] Ex-ante EF and ex-post EF1 exists for additive vals.

 $\left(\mathbb{E}[\boldsymbol{v}_i(X_i)] \ge \mathbb{E}[\boldsymbol{v}_i(X_j)] \right) \quad \left(\boldsymbol{v}_i(X_i) \ge \boldsymbol{v}_i(X_j - \boldsymbol{g}) \text{ for some } \boldsymbol{g} \in X_i \right) \quad \left(\boldsymbol{v}_i(X_i) = \sum_{X_i} \boldsymbol{v}_{i,j} \right)$

Q1: Can we get subadditive vals?

 $\left(\boldsymbol{v}_i(\boldsymbol{S}\cup\boldsymbol{T})\leq\boldsymbol{v}_i(\boldsymbol{S})+\boldsymbol{v}_i(\boldsymbol{T})\right)$

Q2: Can we get ex-post EFX?

 $(v_i(X_i) \ge v_i(X_j - g) \text{ for all } g \in X_i)$

Starting point: [Freeman, Shah, Vaish, 2020] and [Aziz, 2020] Ex-ante EF and ex-post EF1 exists for additive vals.

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Q1: Can we get subadditive vals?

 $\left(v_i(S \cup T) \le v_i(S) + v_i(T)\right)$

Q2: Can we get ex-post EFX?

$$(v_i(X_i) \geq v_i(X_j - g) \text{ for all } g \in X_i)$$

Theorem (impossibility result): Ex-ante EF and ex-post EFX does not exist for subadditive vals.

Main Result

Theorem:

Ex-ante $\frac{1}{2}$ -**EF** and **ex-post** $\frac{1}{2}$ -**EFX** + **EF1** exists for subadditive vals.

 $\left(\mathbb{E}[\nu_i(X_i)] \ge 1/2 \cdot \mathbb{E}[\nu_i(X_j)]\right) \quad \left(\nu_i(X_i) \ge 1/2 \cdot \nu_i(X_j - g) \text{ for all } g \in X_i\right) \quad \left(\nu_i(S \cup T) \le \nu_i(S) + \nu_i(T)\right)$

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- More general setting.
- Stronger ex-post guarantee.
- Weaker ex-ante guarantee.

Main Result

Theorem:

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- Removing both approximation factors (1/2 and 1/2) is impossible.
- Going beyond subadditive to general monotone is impossible.
- The first best-of-both-worlds guarantee for subadditive valuations.
- ¹/₂-EFX is the best known approximation of EFX for subadditive.
- Proof via a careful randomization of the Envy Cycles procedure.

Phase I: Allocate one item per agent. ^[Lipton, Markakis, Mossel, Saberi, 2004] Phase II: In each iteration, give unallocated item to unenvied agent. If all agents are envied, exchange bundles along envy cycle.



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[Lipton, Markakis, Mossel, Saberi, 2004] and [Plaut, Roughgarden, 2020]

For subadditive vals, under mild assumptions for Phase I, the outcome of Envy Cycles always satisfies ex-post EF1 and ½-EFX.

 $(v_i(X_i) \ge v_i(X_j - g) \text{ for some } g \in X_i) (v_i(X_i) \ge 1/2 \cdot v_i(X_j - g) \text{ for all } g \in X_i)$

The main difficulty is to guarantee ex-ante ¹/₂-EF.

 $\left(\mathbb{E}[v_i(X_i)] \geq 1/2 \cdot \mathbb{E}[v_i(X_j)]\right)$

Phase I: Allocate one item per agent. ^[Lipton, Markakis, Mossel, Saberi, 2004] Phase II: In each iteration, give unallocated item to unenvied agent. If all agents are envied, exchange bundles along envy cycle.

Guarantee ex-ante ¹/₂-EF by randomizing the choice of:

- 1. one item per agent
- 2. unallocated item
- 3. unenvied agent
- 4. envy cycle

 Phase I: Allocate one item per agent. ^[Lipton, Markakis, Mossel, Saberi, 2004]
Phase II: In each iteration, give unallocated item to unenvied agent. If all agents are envied, exchange bundles along envy cycle.

Guarantee ex-ante ¹/₂-EF by randomizing the choice of:

1. one item per agent - simultaneous eating and Birkhoff rounding

(see paper for details)

2. unallocated item

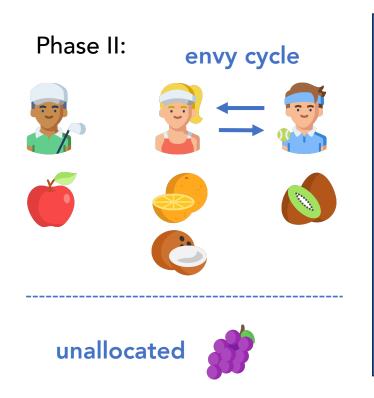
3. unenvied agent

4. envy cycle

arbitrary choices

---- crucial part

 Phase I: Allocate one item per agent. ^[Lipton, Markakis, Mossel, Saberi, 2004]
Phase II: In each iteration, give unallocated item to unenvied agent. If all agents are envied, exchange bundles along envy cycle.



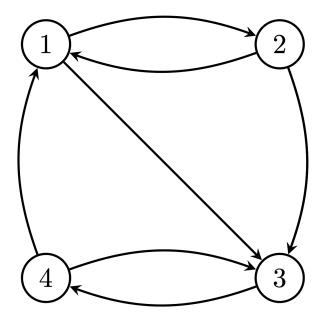
Which envy cycle to choose?

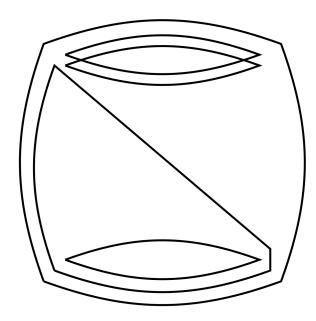
Next few slides:

- 1. Define key property for envy cycle distribution
- 2. Intuition behind the key property
- 3. Construction satisfying the key property

Envy graph: $(i \rightarrow j)$ if $v_i(X_i) < v_i(X_j)$

Distribution over envy cycles



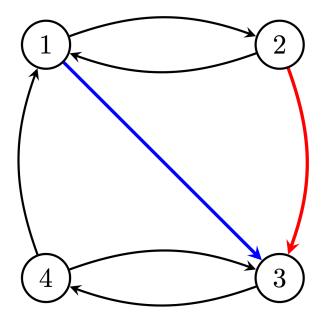


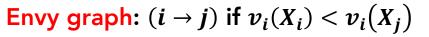
uniform distribution over 5 cycles

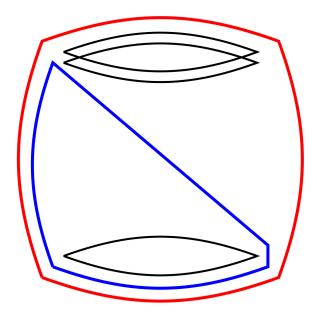
Key property:

If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

example: agents 1 and 2 both envy 3, and they get X_3 with probability 1/5 each.





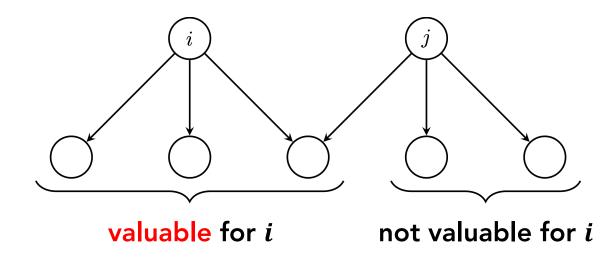


uniform distribution over 5 cycles

Key property:

If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

Intuition: An envy cycles distribution satisfying key property is envy-free.



i is at least as likely as *j* to get any of the valuable bundles

Key property:

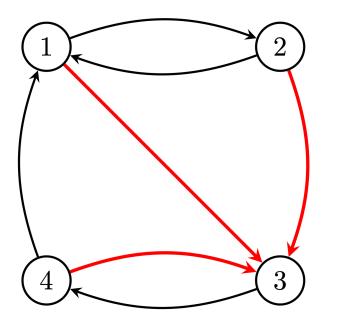
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Stationary distribution of the random walk on the transposed envy graph

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Stationary distribution of the random walk on the transposed envy graph

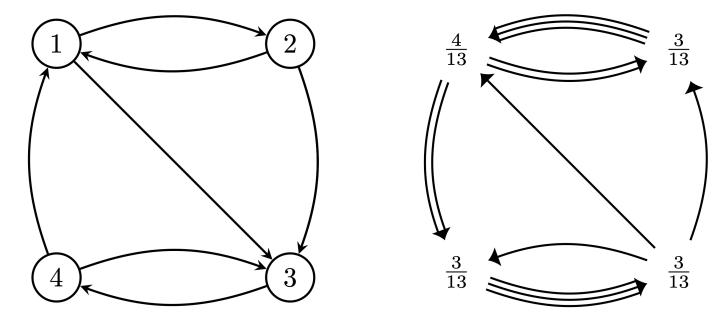


3 goes to 1, 2, 4 with probability 1/3 each

Key property:

If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

Stationary distribution of the random walk on the transposed envy graph

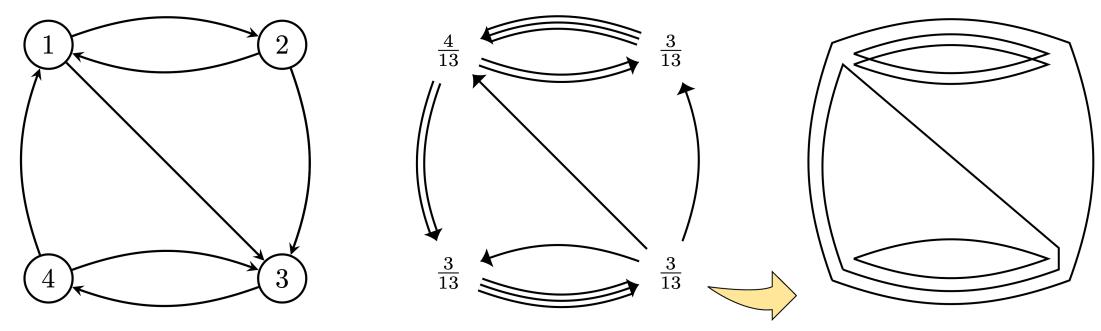


each edge = probability of 1/13

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If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

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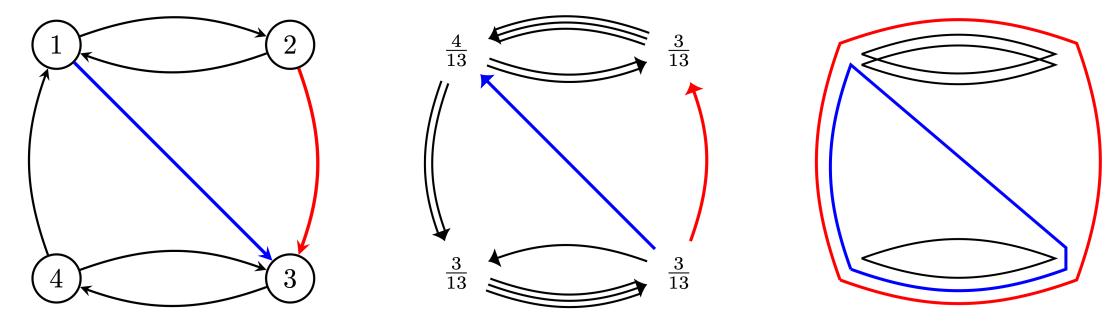


cycle decomposition by [MacQueen, 1981]

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If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

Stationary distribution of the random walk on the transposed envy graph

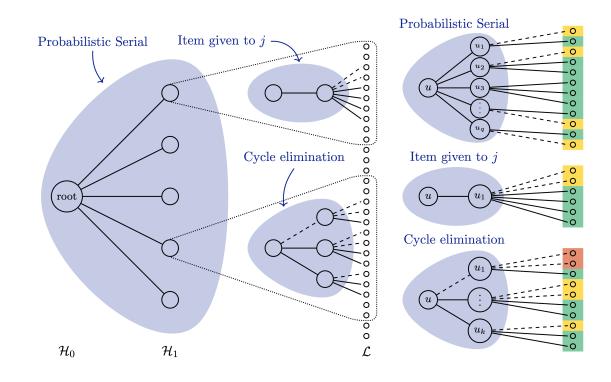


the key property holds because the random walk picks edges uniformly

Key property:

If agents *i* and *j* both envy *k*, they are equally likely to get X_k .

See the paper for the proof that key property implies ex-ante ½-EF



Open Problems

Main result:

Ex-ante $\frac{1}{2}$ -**EF** and **ex-post** $\frac{1}{2}$ -**EFX** + **EF1** exists for subadditive vals.

 $\left(\mathbb{E}[v_i(X_i)] \ge 1/2 \cdot \mathbb{E}[v_i(X_j)]\right) \quad \left(v_i(X_i) \ge 1/2 \cdot v_i(X_j - g) \text{ for all } g \in X_i\right) \quad \left(v_i(S \cup T) \le v_i(S) + v_i(T)\right)$

Open problem 1: Does ex-ante EF and ex-post EF1 exist for subadditive vals?

Can we randomize Envy Cycles to get ex-ante EF?

Open problem 2: Does ex-ante EF and ex-post EFX exist for additive vals?

Can we get a new impossibility result about EFX?

images: flaticon.com