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Topics in Computational Social Choice Theory

Lecture 01: Introduction on Discrete Fair Division

Hannaneh Akrami

Organization

Seminar: 2+0, 7 CPS

Organized by Kurt Mehlhorn, Nidhi Rathi, and Hannaneh Akrami

When? Every Tuesday 14:15 - 15:45

Requirements: Basic algorithms lecture
(Introduction to Algorithms and Data Structures)

- Your task:**
- Present a paper from the list in 50-85 minutes.
 - Write a summary of the paper by August 2nd.
 - The presentation needs to be discussed with us at least one week before your scheduled talk.
 - Send us your preferred order of the papers by April 30th.



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Computational Social Choice Theory

Social Choice Theory: Making a **collective** decision from **individual** preferences.



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Computational Social Choice Theory

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Voting



mp

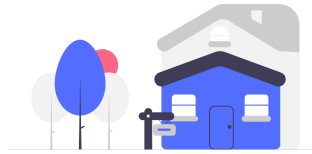
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Computational Social Choice Theory

Social Choice Theory: Making a **collective** decision from **individual** preferences.



Voting



Resource Allocation



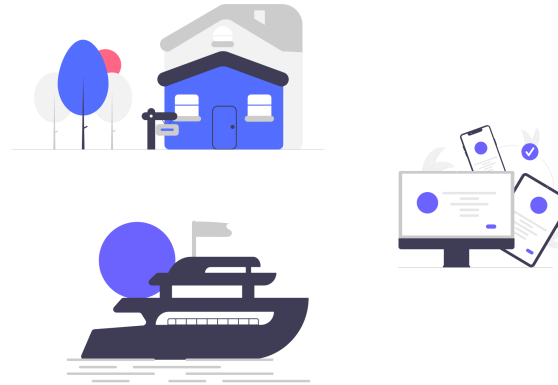
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Computational Social Choice Theory

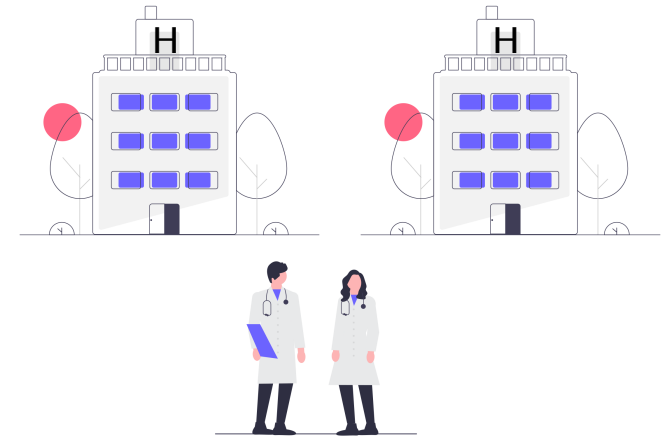
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Voting



Resource Allocation



Stable Matchings



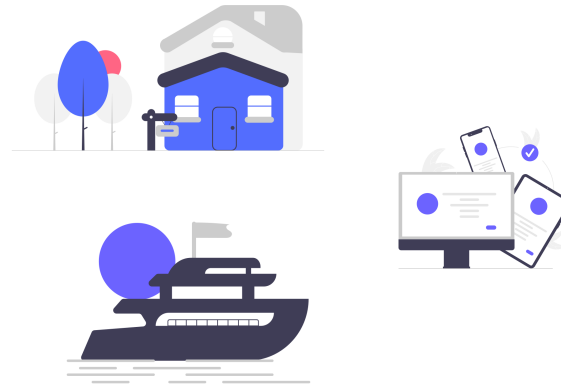
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Computational Social Choice Theory

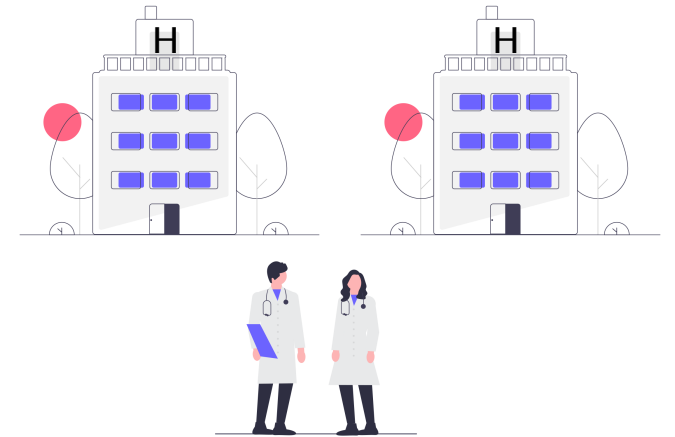
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Voting



Resource Allocation



Stable Matchings

Economists and Politicians: Does there exist a **social choice** mechanism with the desired economic properties?

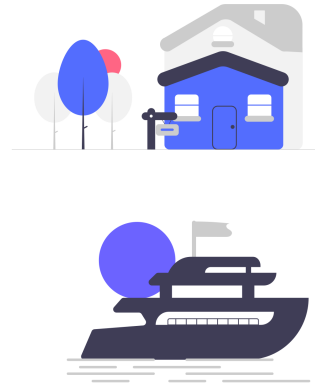


Computational Social Choice Theory

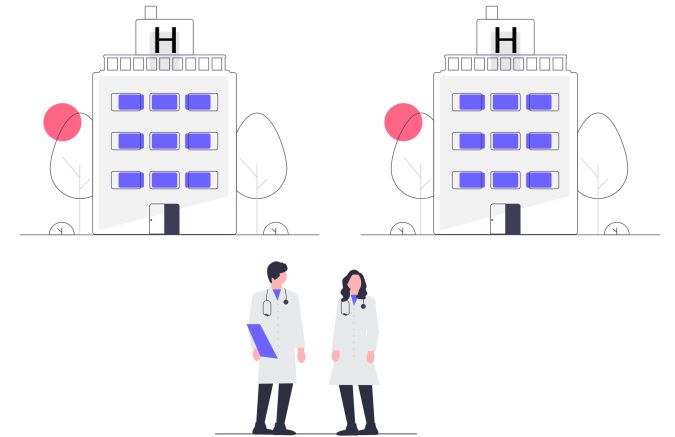
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Computer Scientists: How to efficiently **compute** such a mechanism?



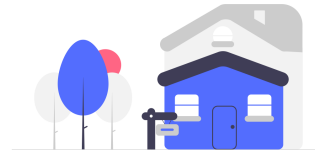
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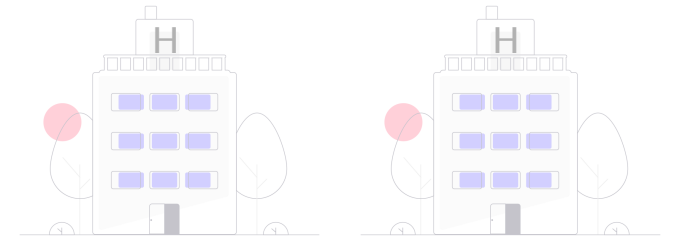
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Fair Division

Divide **items** among **agents** in a **fair** manner.

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Applications:



Partnership
dissolution



Divorce
settlements



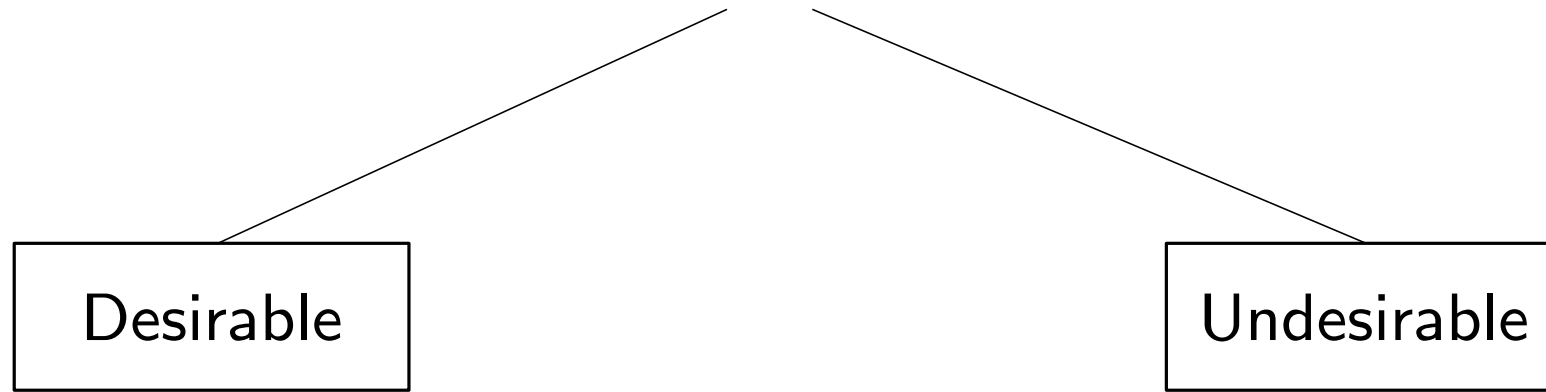
Household
chores



Air traffic
management



Items



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Items

Desirable



Divorce
settlements

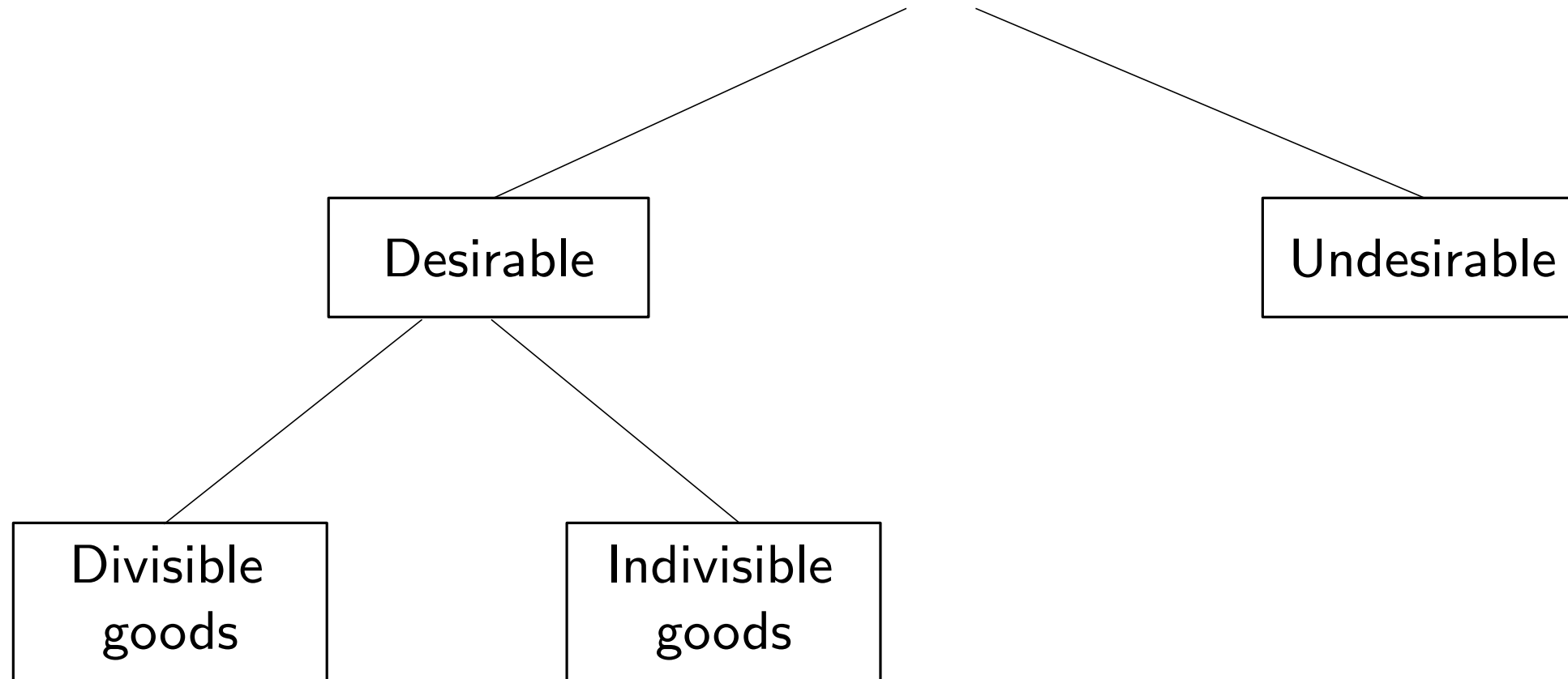
Undesirable



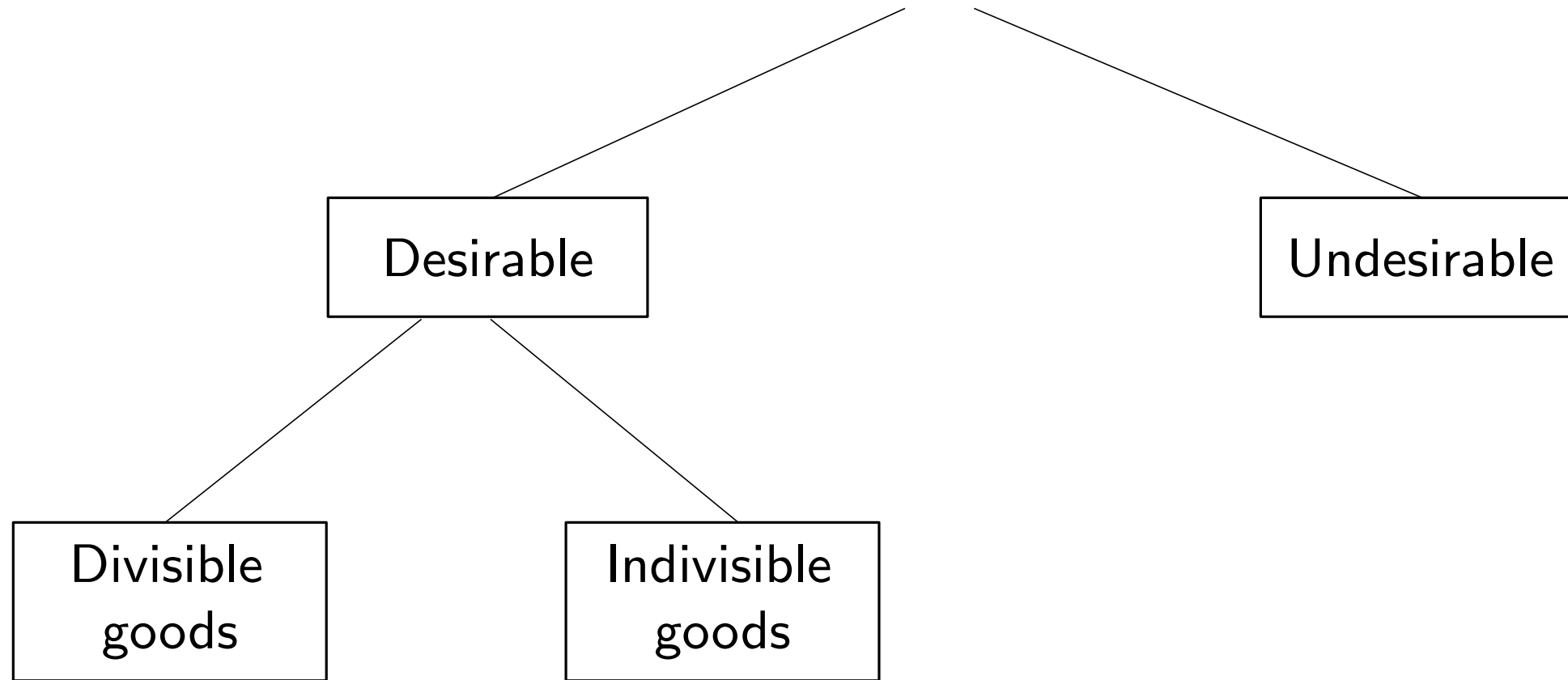
Household
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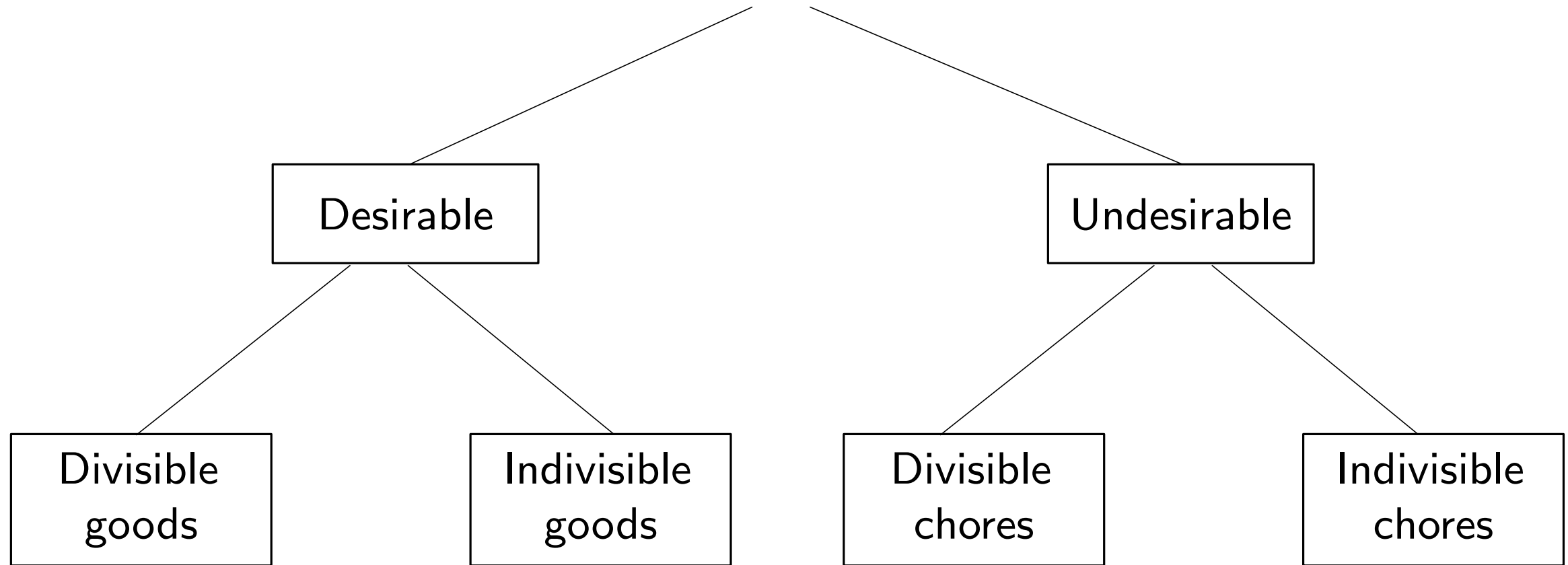
Items



Items



Items



Items

Desirable

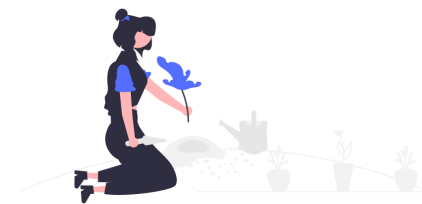
Undesirable

Divisible
goods

Indivisible
goods

Divisible
chores

Indivisible
chores



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Items

Desirable

Undesirable

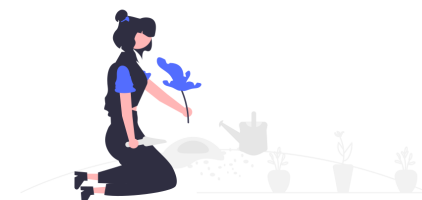
Today

Divisible
goods

Indivisible
goods

Divisible
chores

Indivisible
chores



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Items

Desirable

Undesirable

Next week

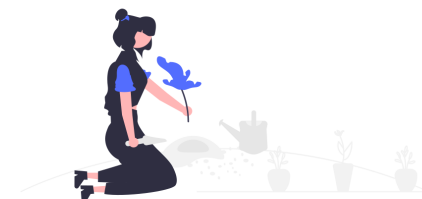
Today

Divisible
goods

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goods

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Discrete Fair Division

Divide **indivisible items** among **agents** in a **fair** manner.

Input: $\mathcal{I} = (N, M, V)$

- N : set of n agents
- M : set of m indivisible goods
- Valuation functions $v_i : 2^M \rightarrow \mathbb{R}_{\geq 0}$



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Discrete Fair Division

Divide **indivisible items** among **agents** in a **fair** manner.

Input: $\mathcal{I} = (N, M, V)$

- $N = \{a_1, a_2, a_3\}$
- $M = \{g_1, g_2, g_3, g_4, g_5\}$
- $X_1 = \{g_1\}, X_2 = \{g_2, g_5\},$
 $X_3 = \{g_3\}, P = \{g_4\}$
- $v_1(X_1) = 4, v_1(X_2) = 3$

	g_1	g_2	g_3	g_4	g_5
a_1	4	1	2	2	2
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assuming v_1 is additive: for all $S \subseteq M$, $v_1(S) = \sum_{g \in S} v_i(\{g\})$



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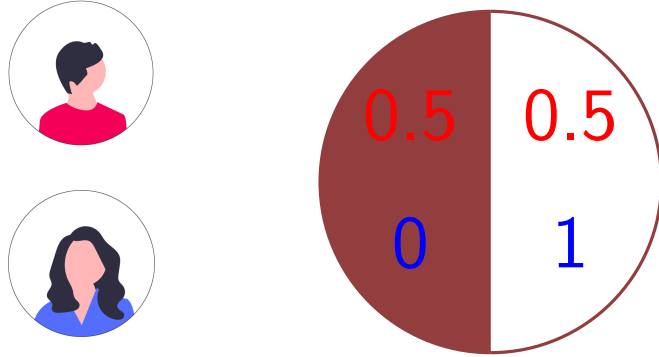
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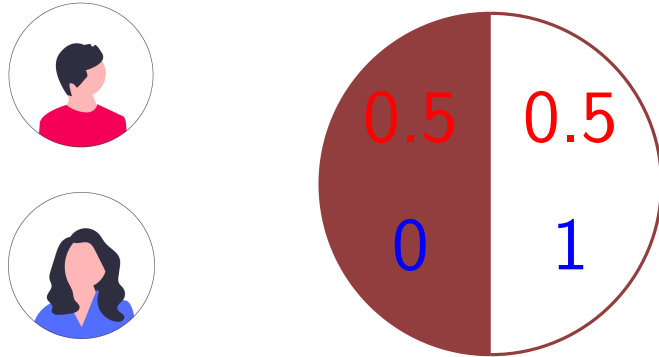
An allocation is **complete**, if $P = \emptyset$ and **partial** otherwise.



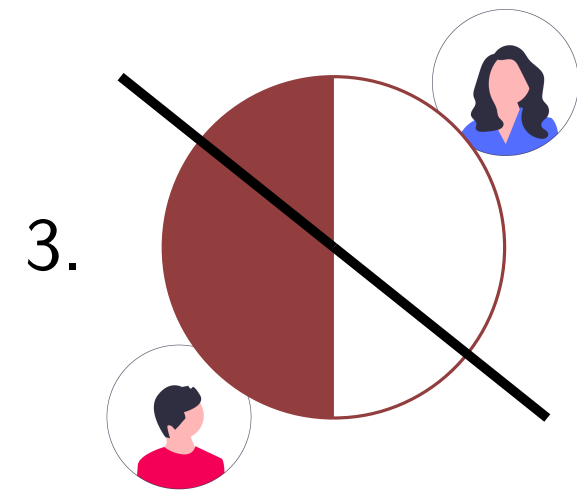
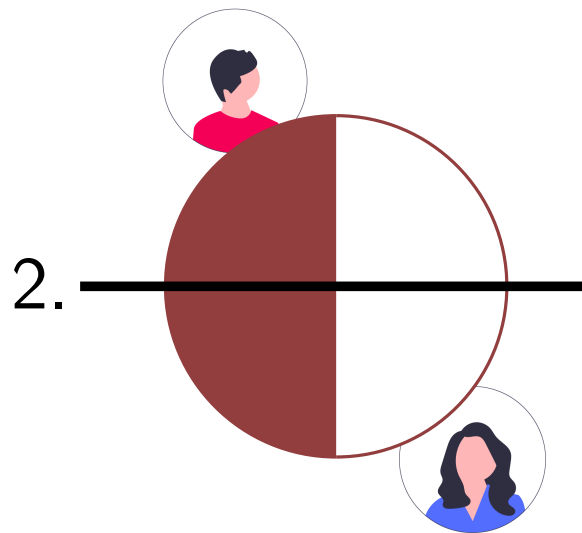
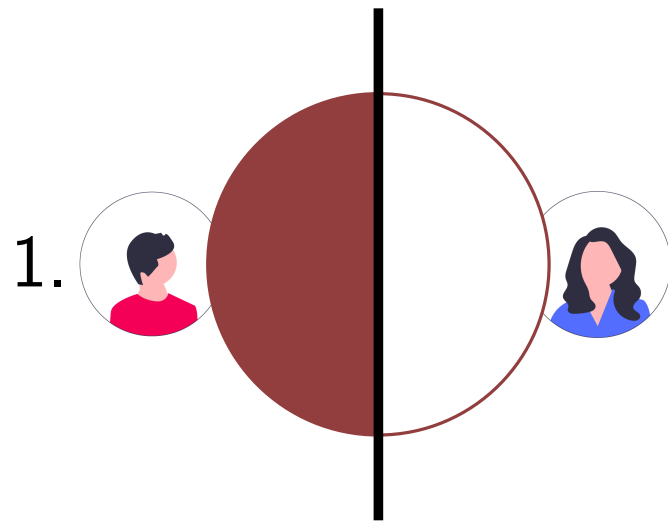
Fairness



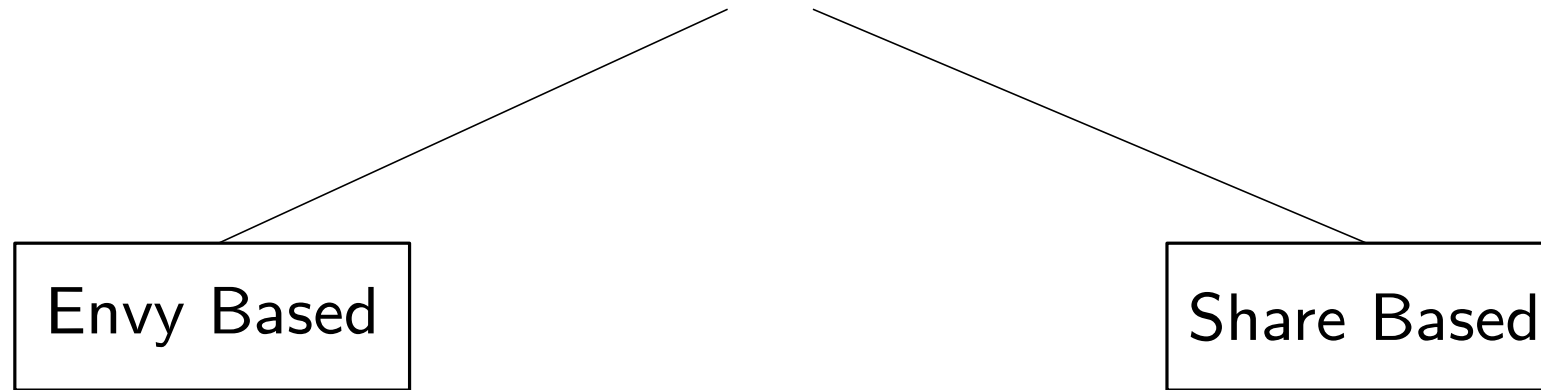
Fairness



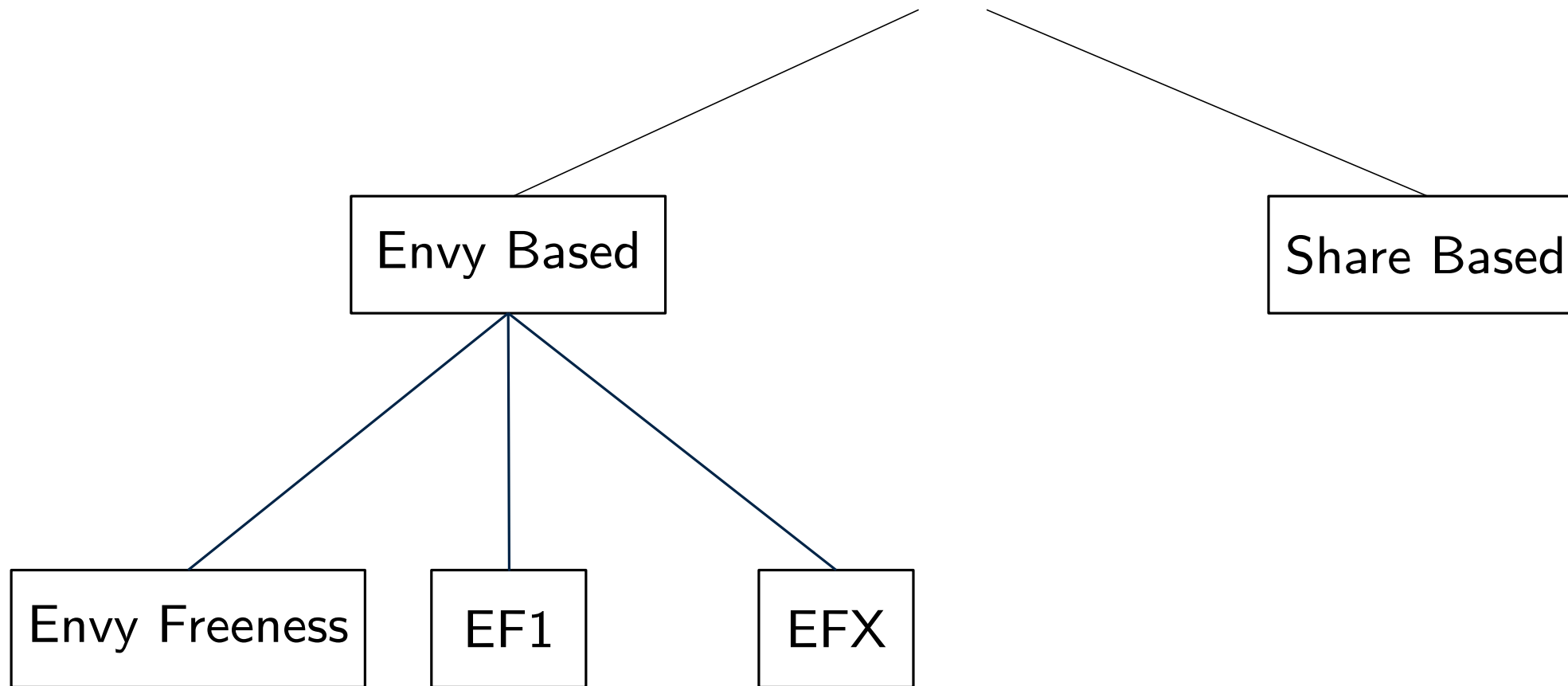
Which allocation is fair?



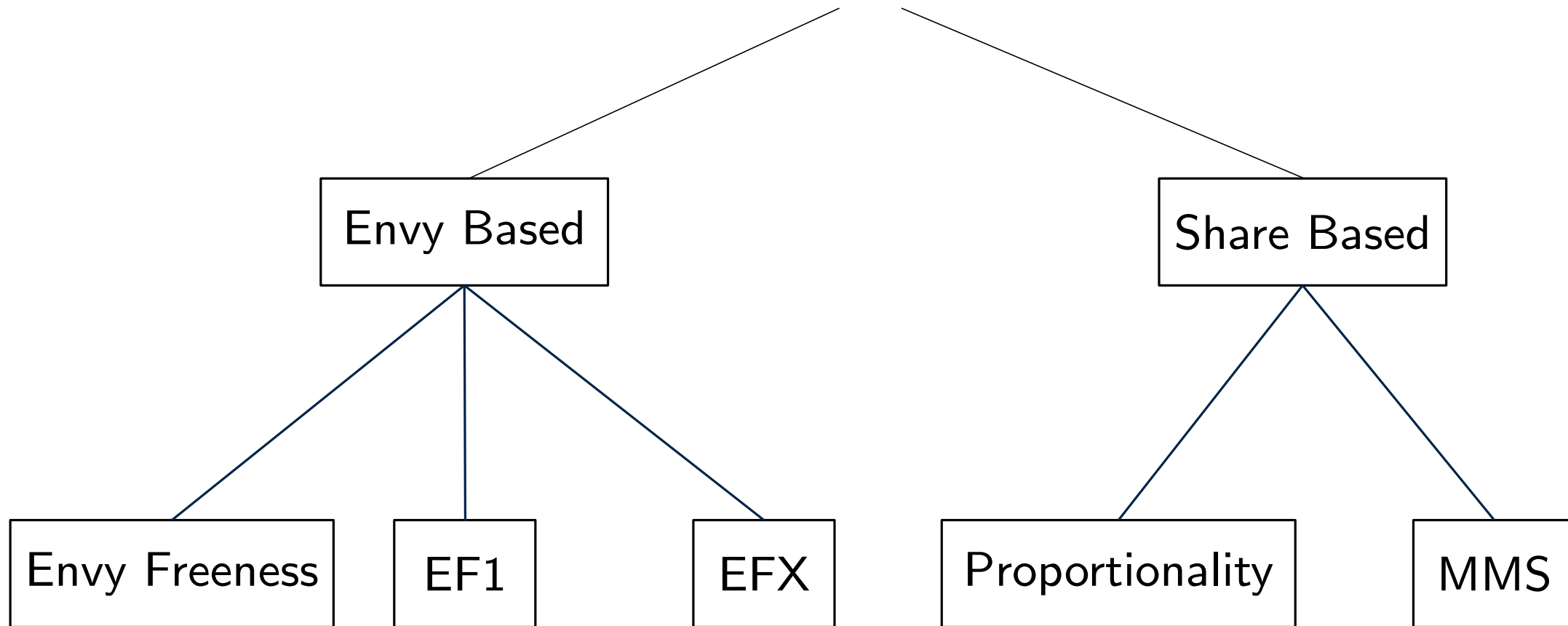
Fairness



Fairness



Fairness



Envy Freeness

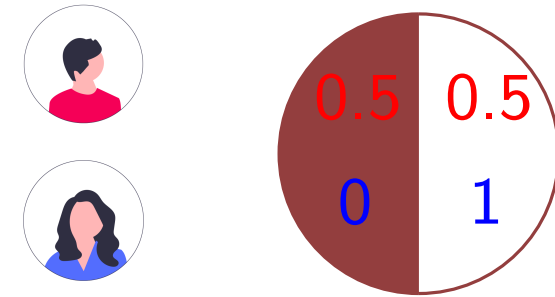
Definition: An allocation X is **envy free**, if and only if for all agents a_i, a_j :
 $v_i(X_i) \geq v_i(X_j)$. [\[Foley 1967\]](#)



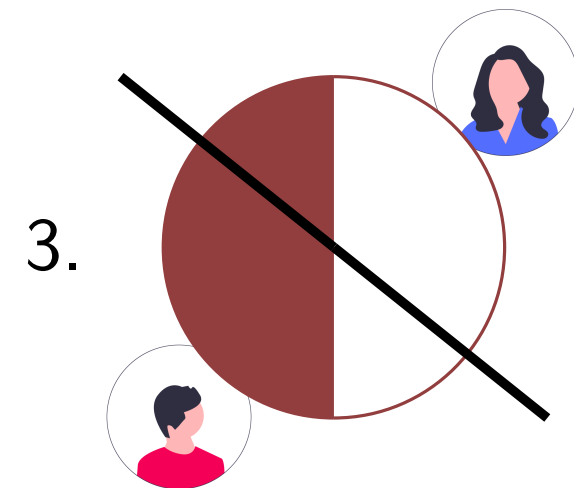
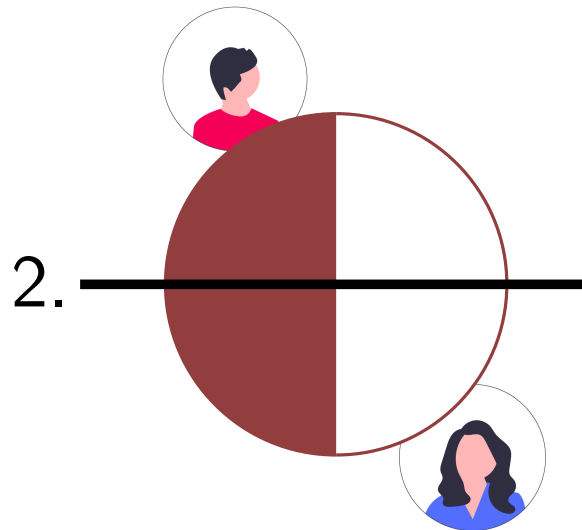
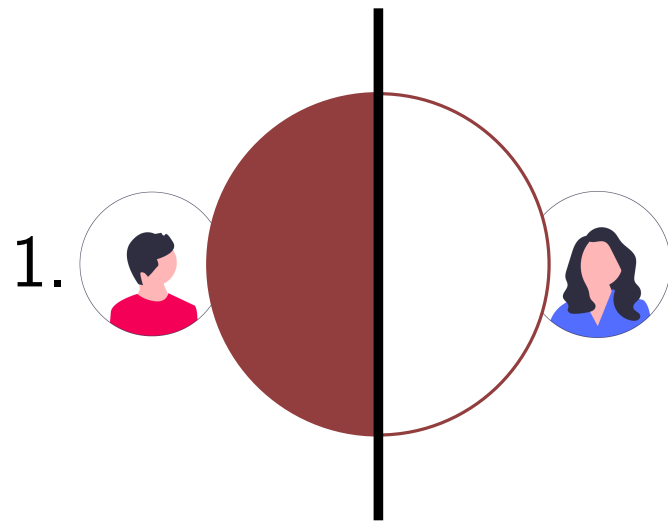
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Envy Freeness

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Which allocation is envy free?



Envy Freeness

Definition: An allocation X is **envy free**, if and only if for all agents a_i, a_j :
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Do complete envy free allocations always exist?



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- For divisible goods, YES! (Next weeks)



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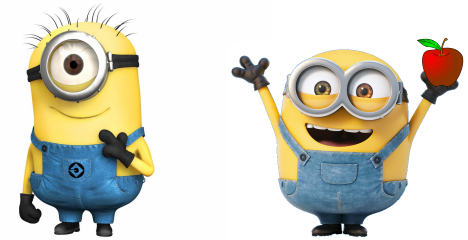
EF1

Definition: An allocation X is **envy free up to one item** or **EF1**, if and only if for all agents a_i, a_j , there exists a good $g \in X_j$ (if $X_j \neq \emptyset$): $v_i(X_i) \geq v_i(X_j \setminus \{g\})$.



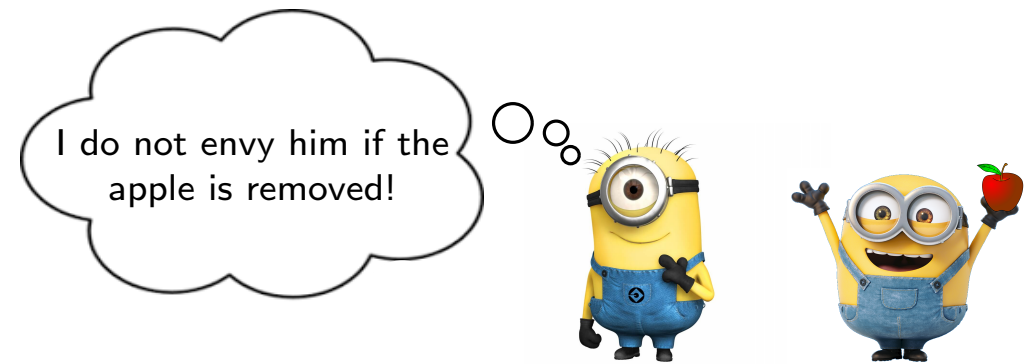
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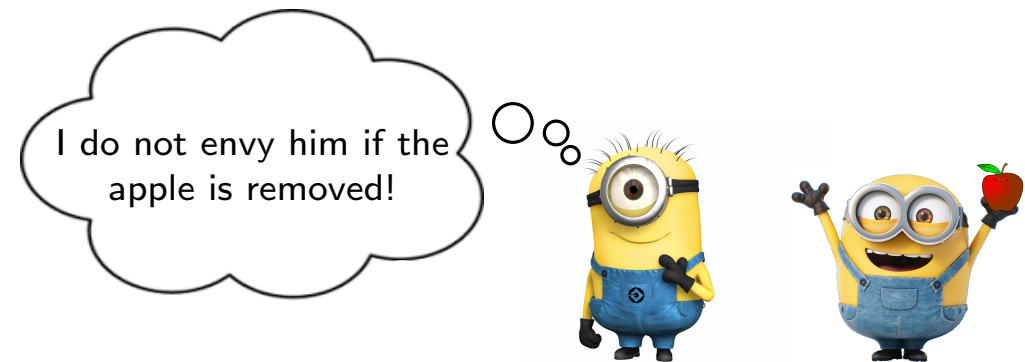
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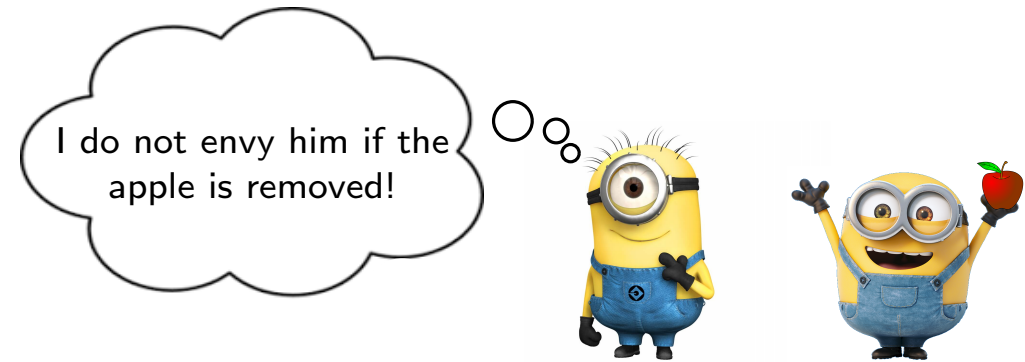


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Do complete EF1 allocations always exist?

- YES for monotone valuations!



EF1

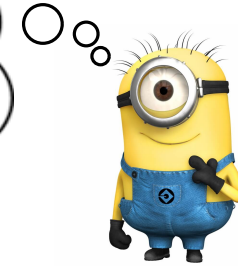
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for all $S \subseteq M$ and $g \in M$, $v(S \cup \{g\}) \geq v(S)$

I do not envy him if the apple is removed!



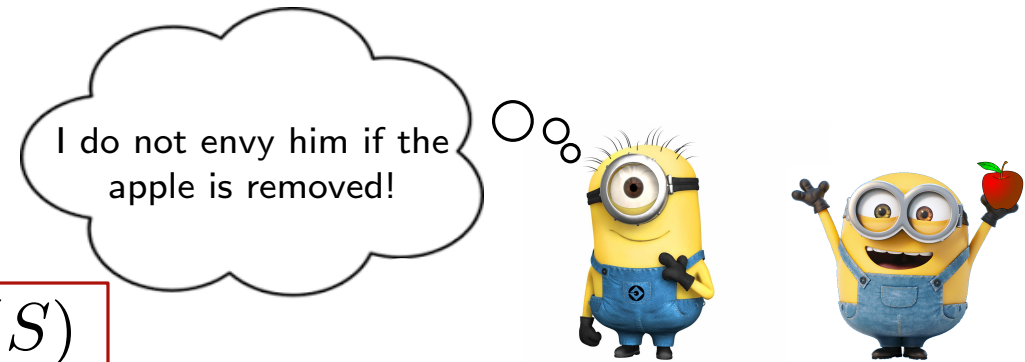
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- A complete EF1 allocation can be found in polynomial time.

[Lipton, Markakis, Mossel, Saberi 2004]



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[Lipton, Markakis, Mossel, Saberi 2004]

- Today: A polynomial time algorithm to find a complete EF1 allocation for additive valuations.











Round-Robin Algorithm

- Fix an ordering of the agents, say a_1, a_2, \dots, a_n .
- Agents take turns according to the ordering $(a_1, a_2, \dots, a_n, a_1, a_2, \dots, a_n, \dots)$ to pick their favorite items from the set of the remaining items.



Round-Robin Algorithm









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	4	1	2	2	2
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Round-Robin Algorithm

Theorem: For additive valuations, Round-Robin returns an EF1 allocation in polynomial time.

a_1 a_2 a_3 \dots a_n



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First round:



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a_1 a_2 a_3 \dots a_n

First round: ●





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Round-Robin Algorithm

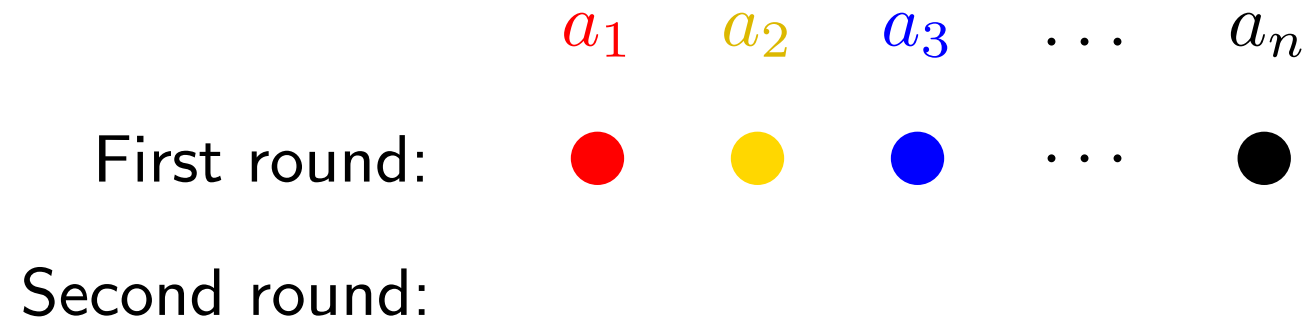
Theorem: For additive valuations, Round-Robin returns an EF1 allocation in polynomial time.

First round: a_1 a_2 a_3 \dots a_n
● ● ● ● ●



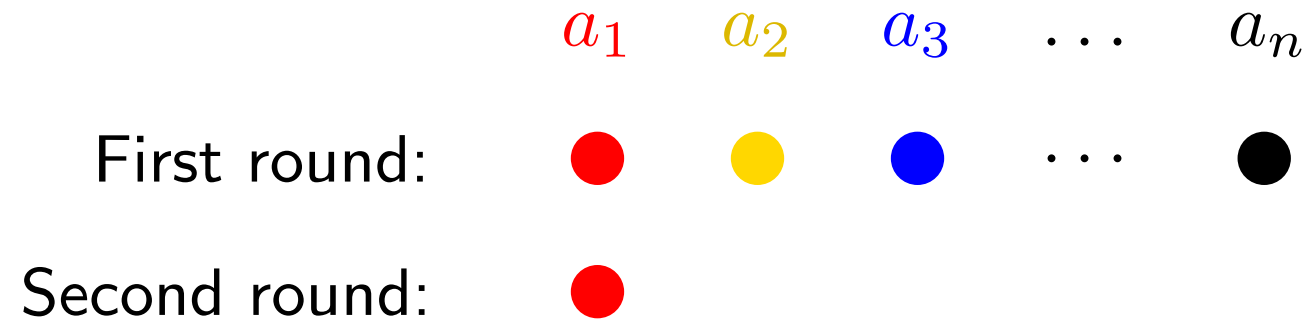
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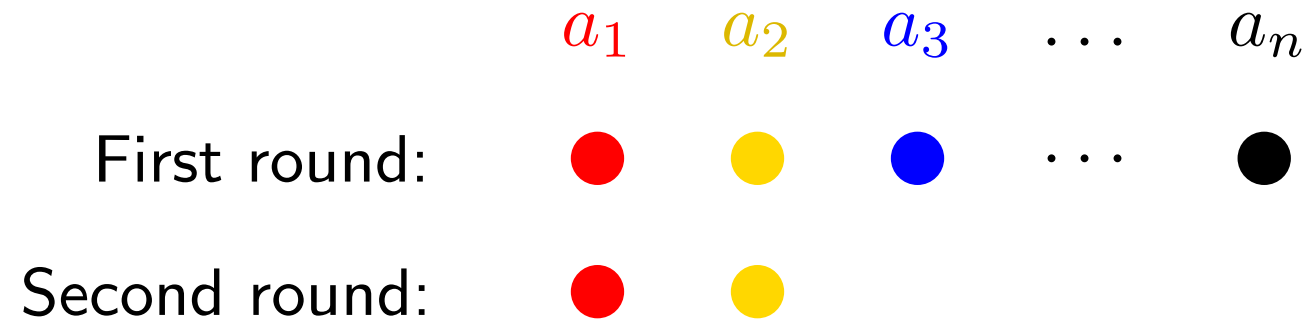
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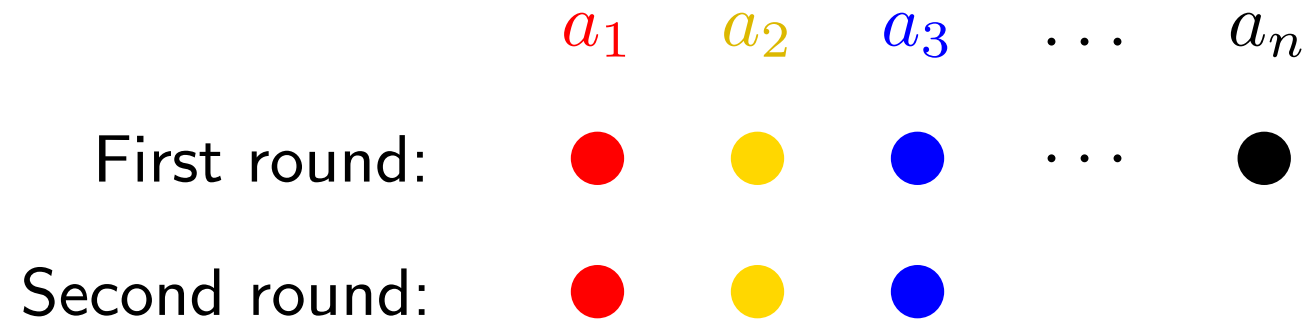
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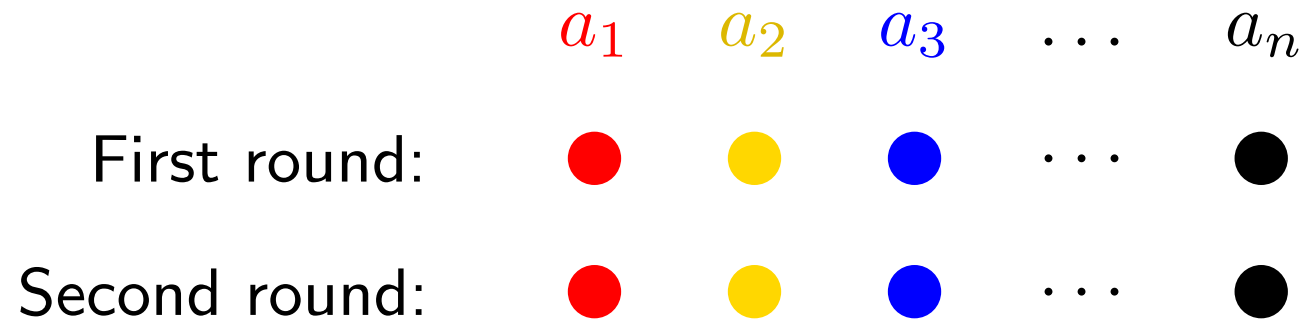
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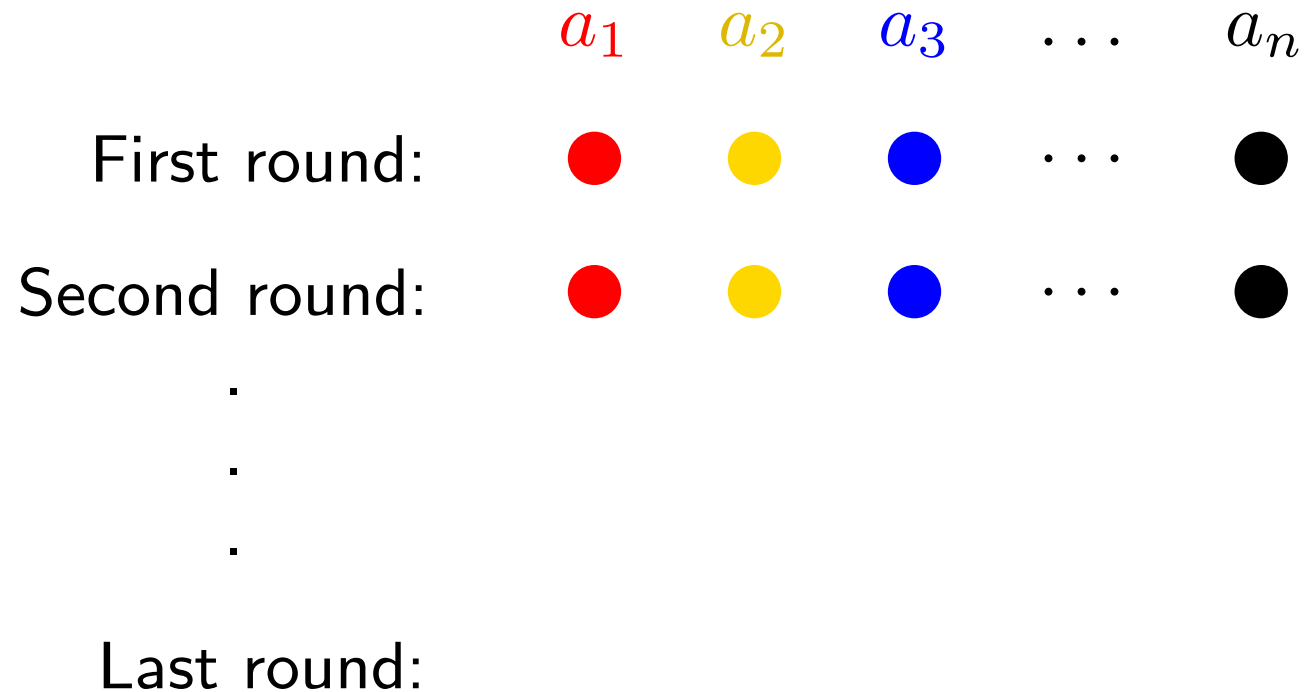
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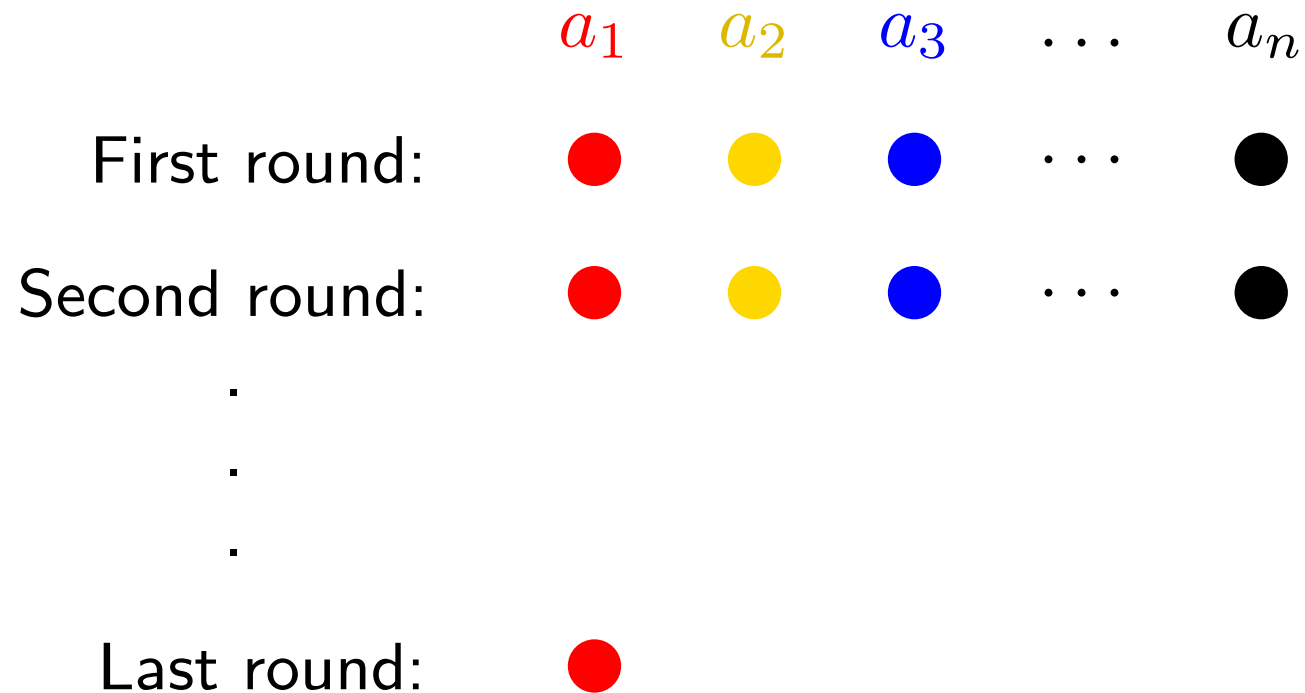
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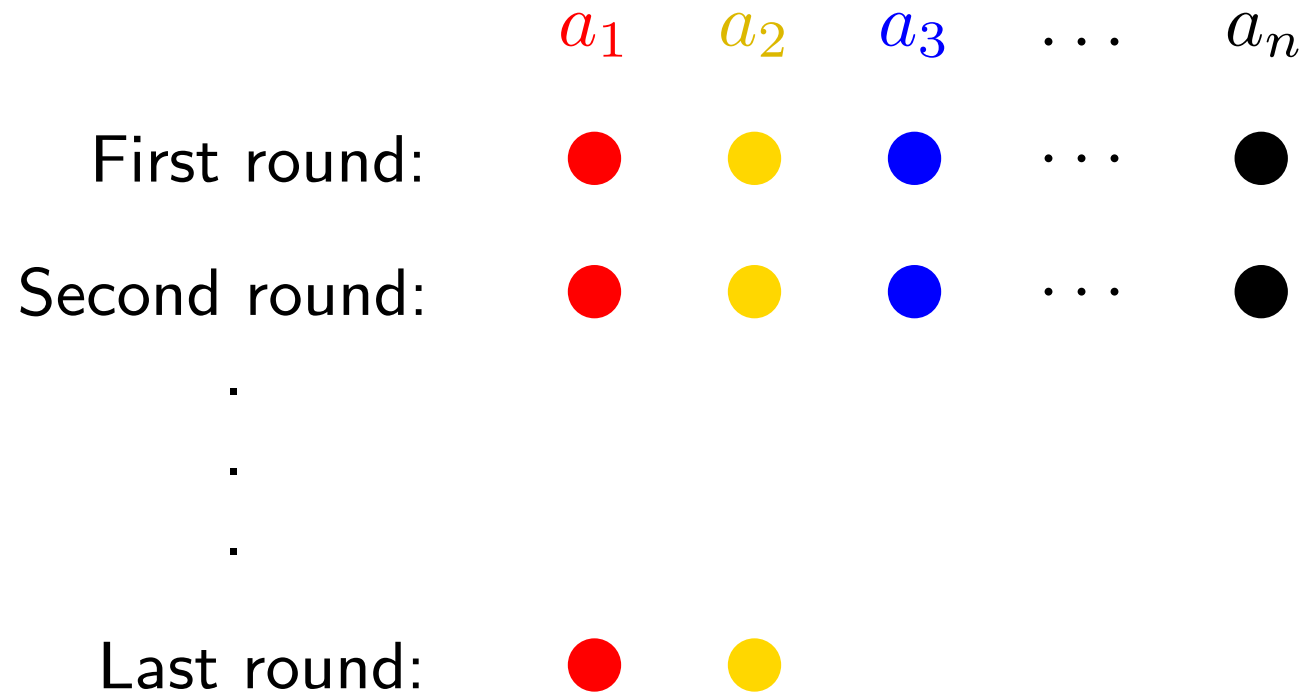
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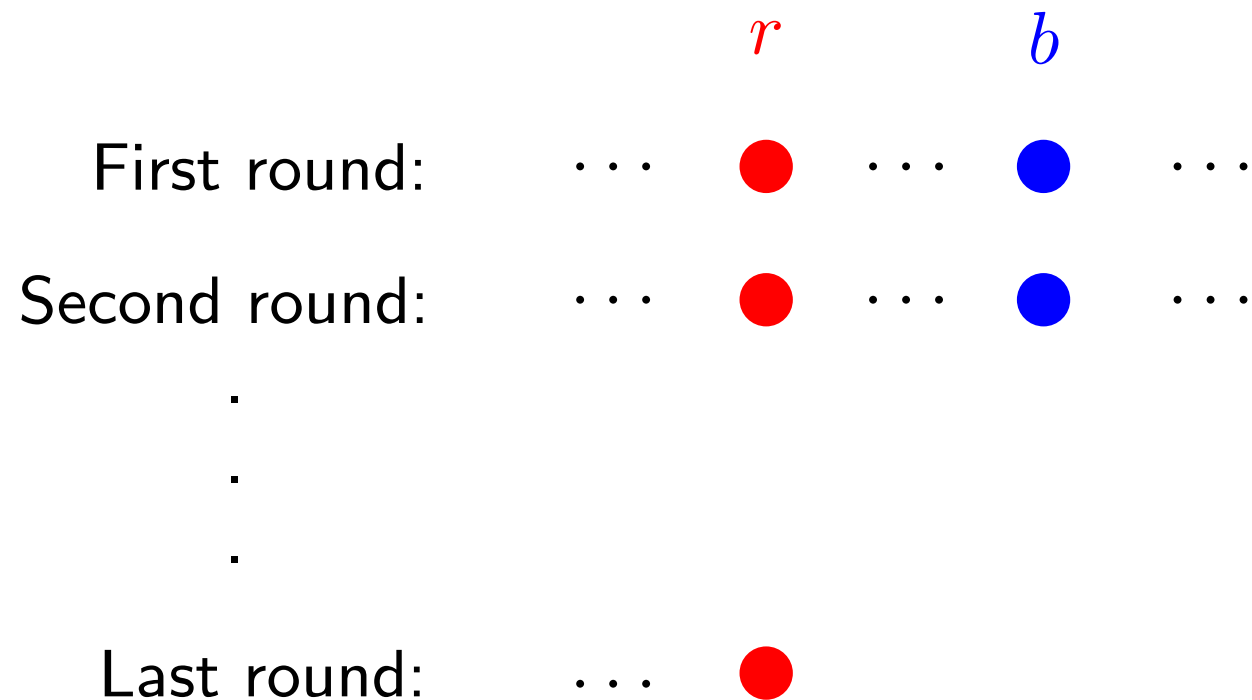


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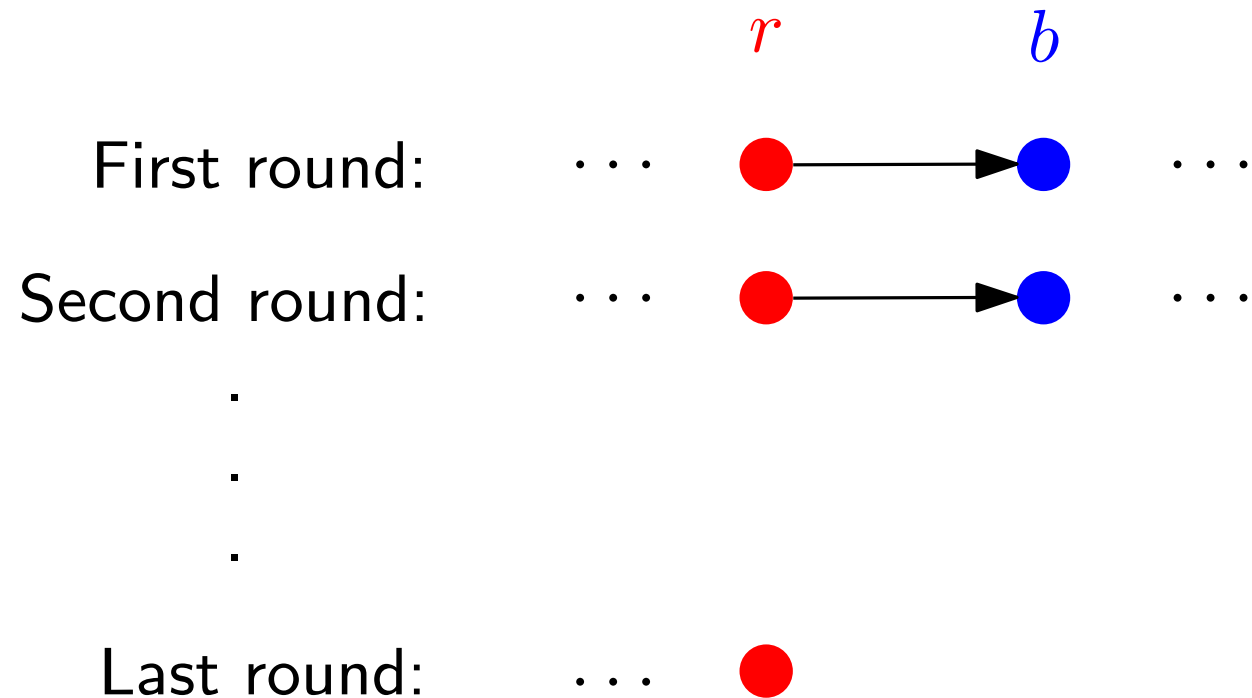
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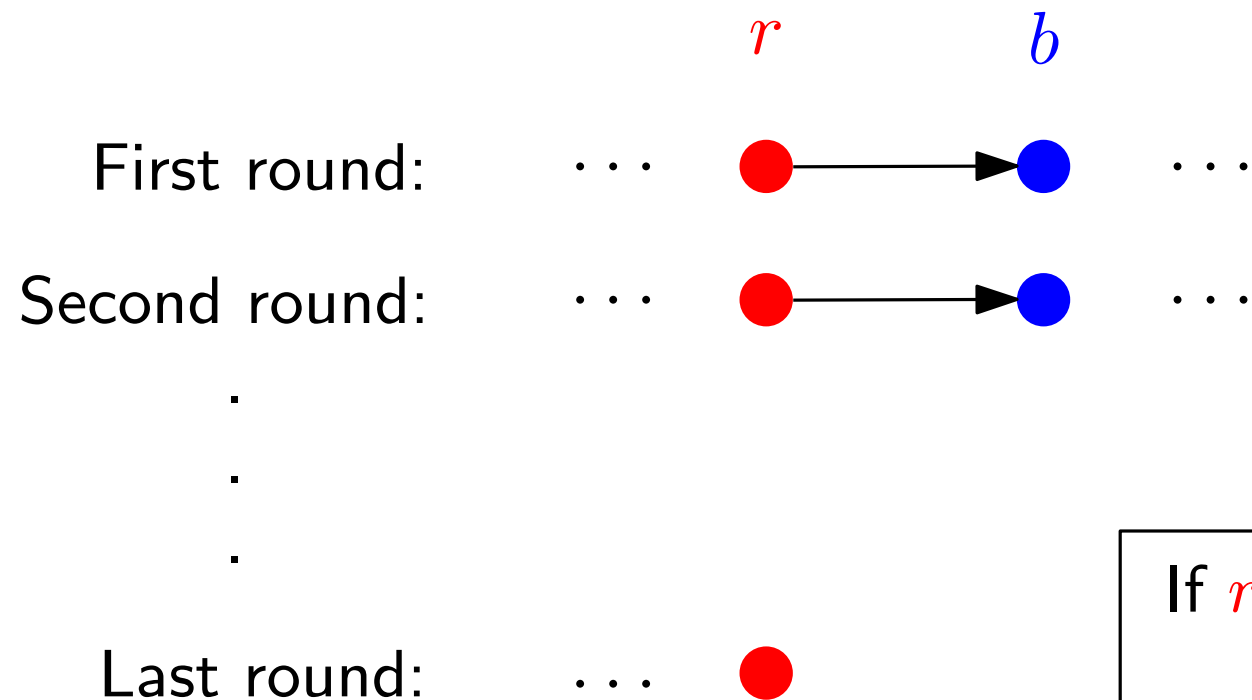
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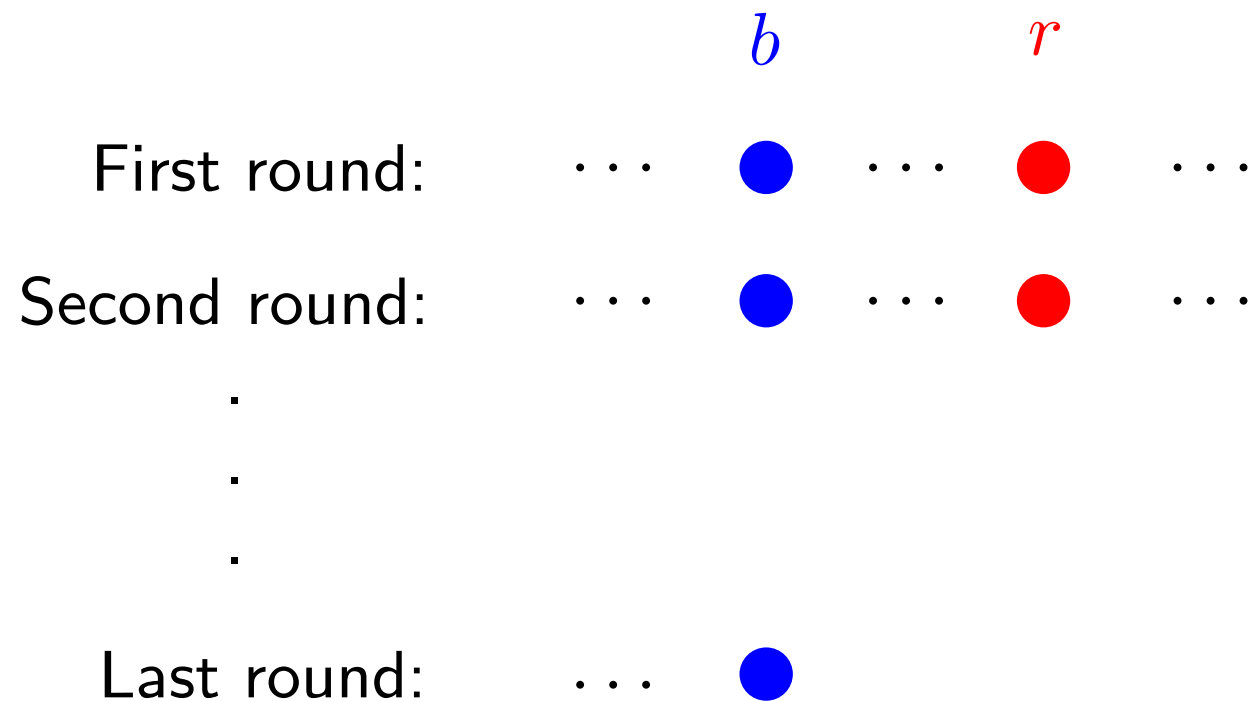


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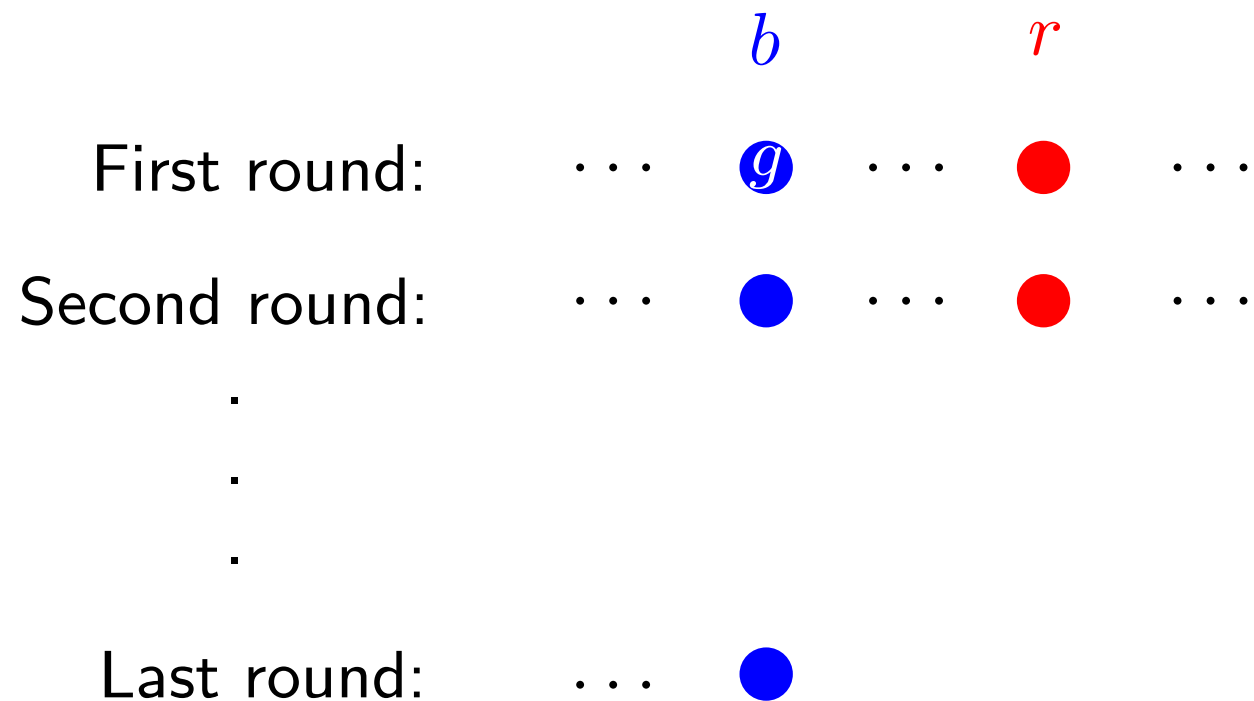
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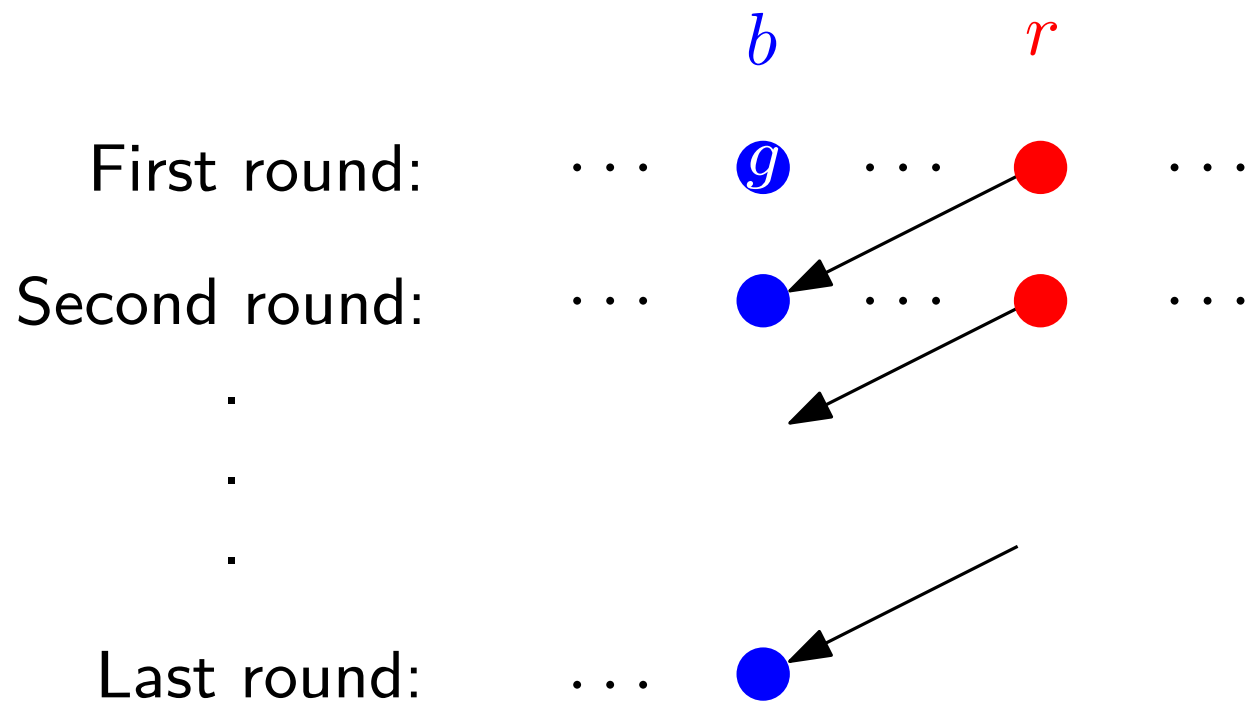
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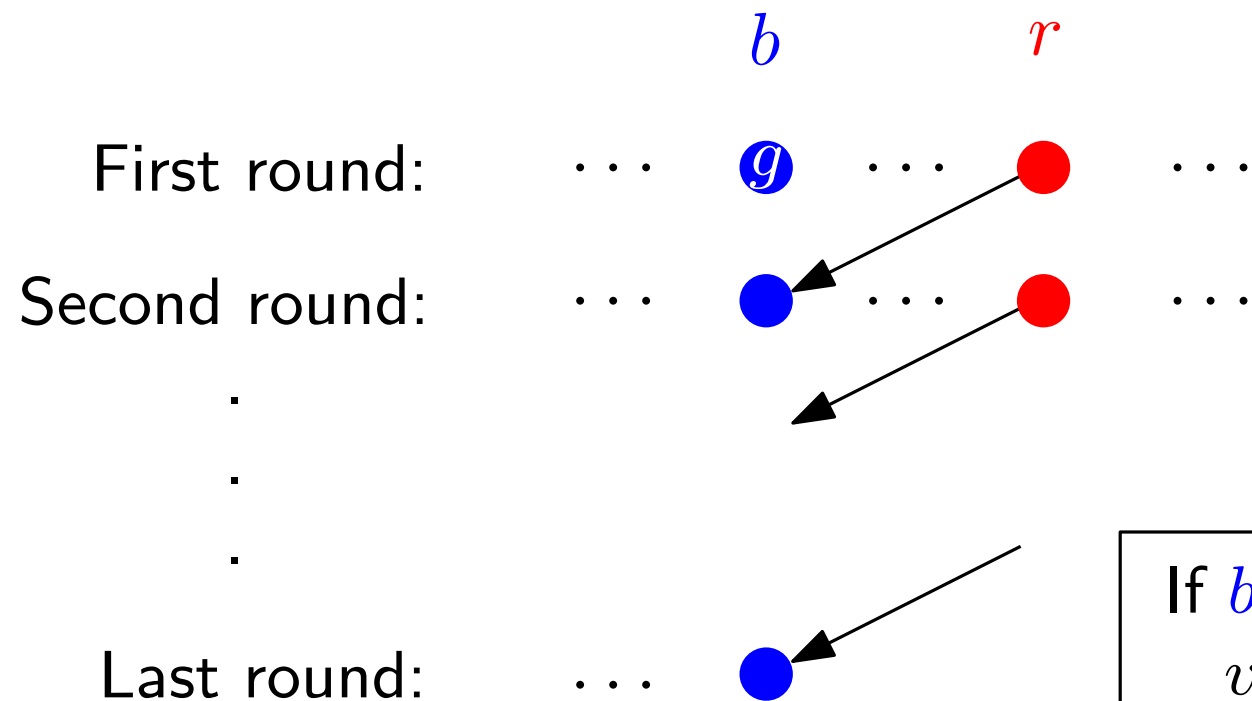
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EFX

Definition: An allocation X is **envy free up to any item** or **EFX**, if and only if for all agents a_i, a_j , and for all goods $g \in X_j$: $v_i(X_i) \geq v_i(X_j \setminus \{g\})$.

[Caragiannis, Kurokawa, Moulin, Procaccia, Shah, Wang 2016]



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Is the following allocation EFX?

					
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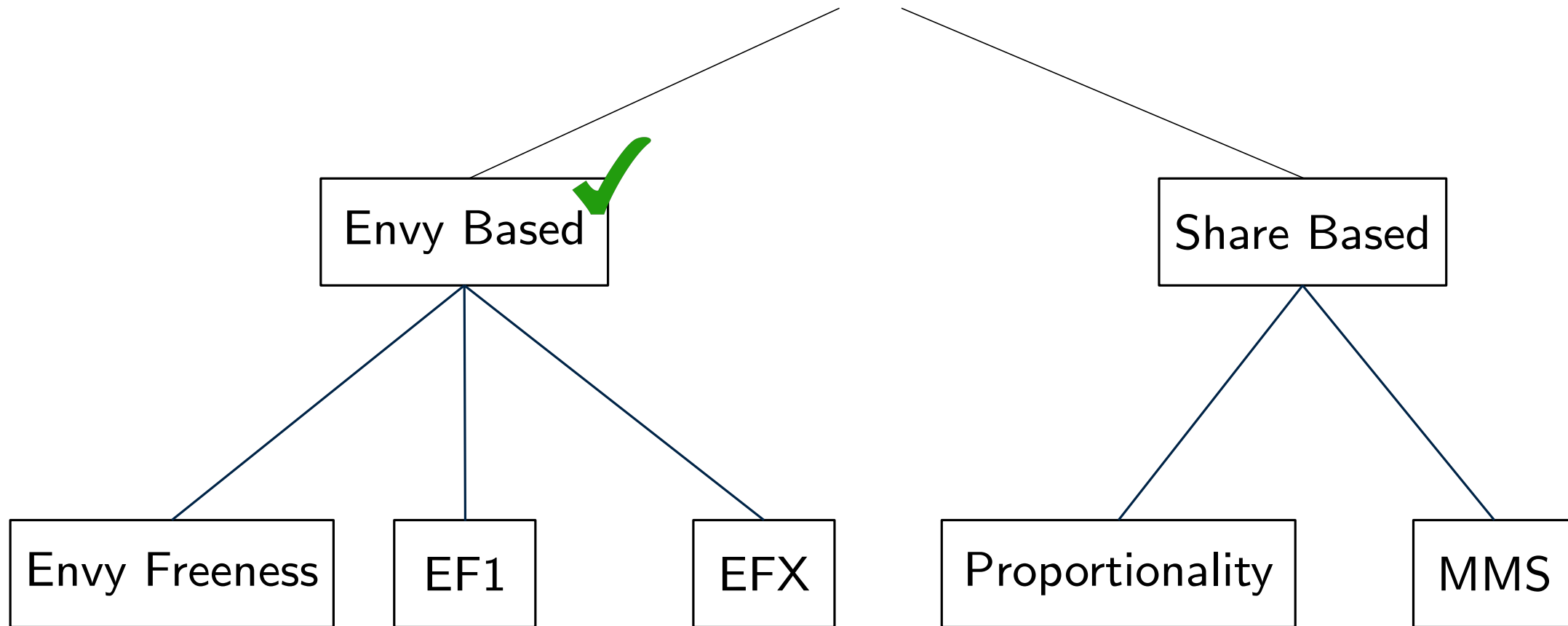
Fair division's biggest problem!

In this seminar we will see:

- Complete EFX allocations exist for 3 agents if at least one has an additive valuation. [Akrami, Alon, Chaudhury, Garg, Mehlhorn, Mehta 2023]
- “Good” partial EFX allocations exists. [Chaudhury, Kavitha, Mehlhorn, Sgouritsa 2020]



Fairness



Proportionality

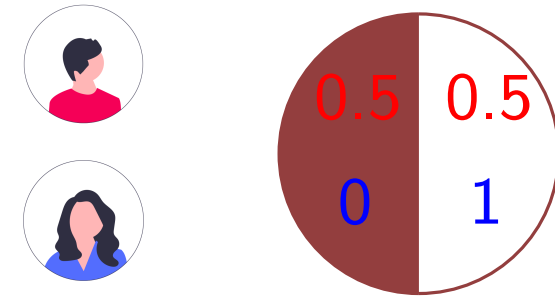
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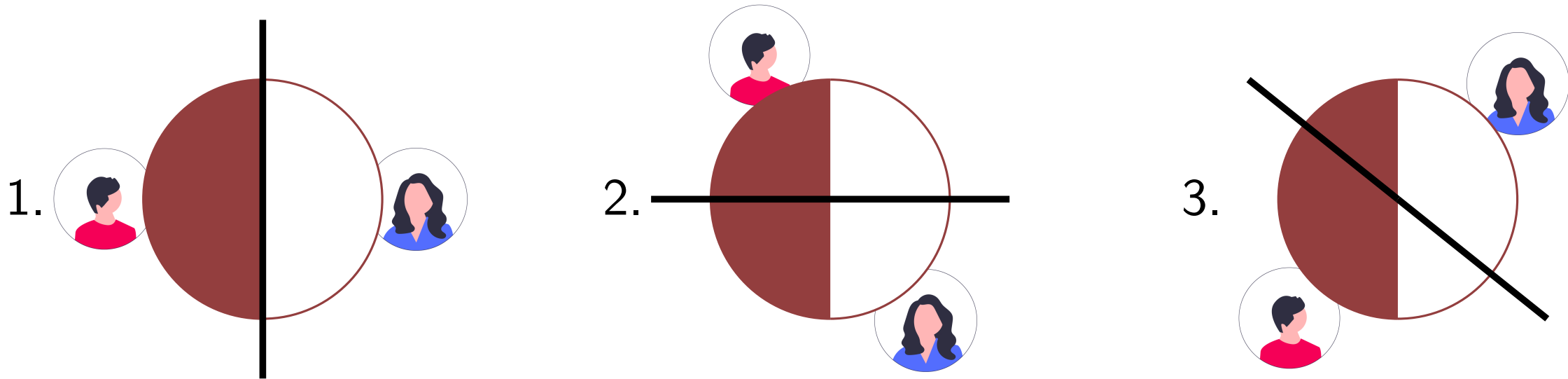
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Which allocation is proportional?



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Maximin Share

- What value can I guarantee for myself if I divide the items into n bundles and receive the least valuable bundle?



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Definition: For all agents a_i , maximin share of agent i is

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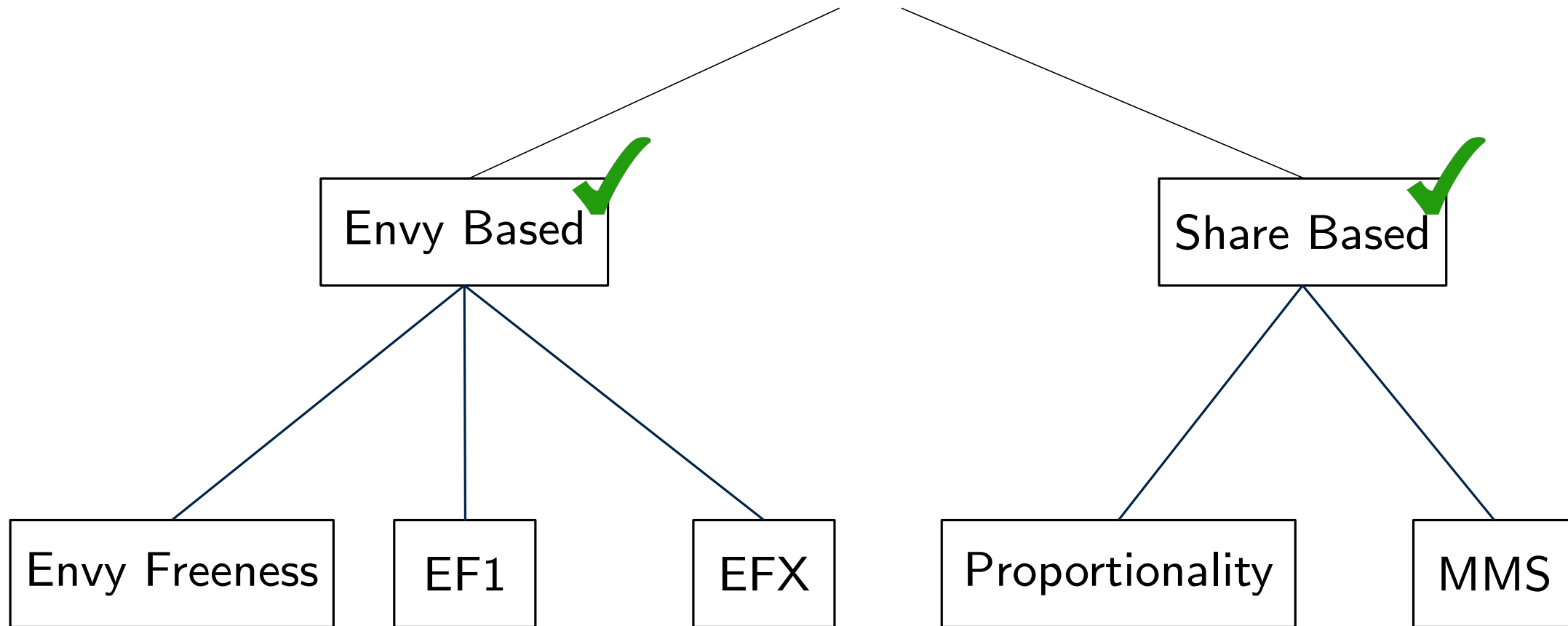
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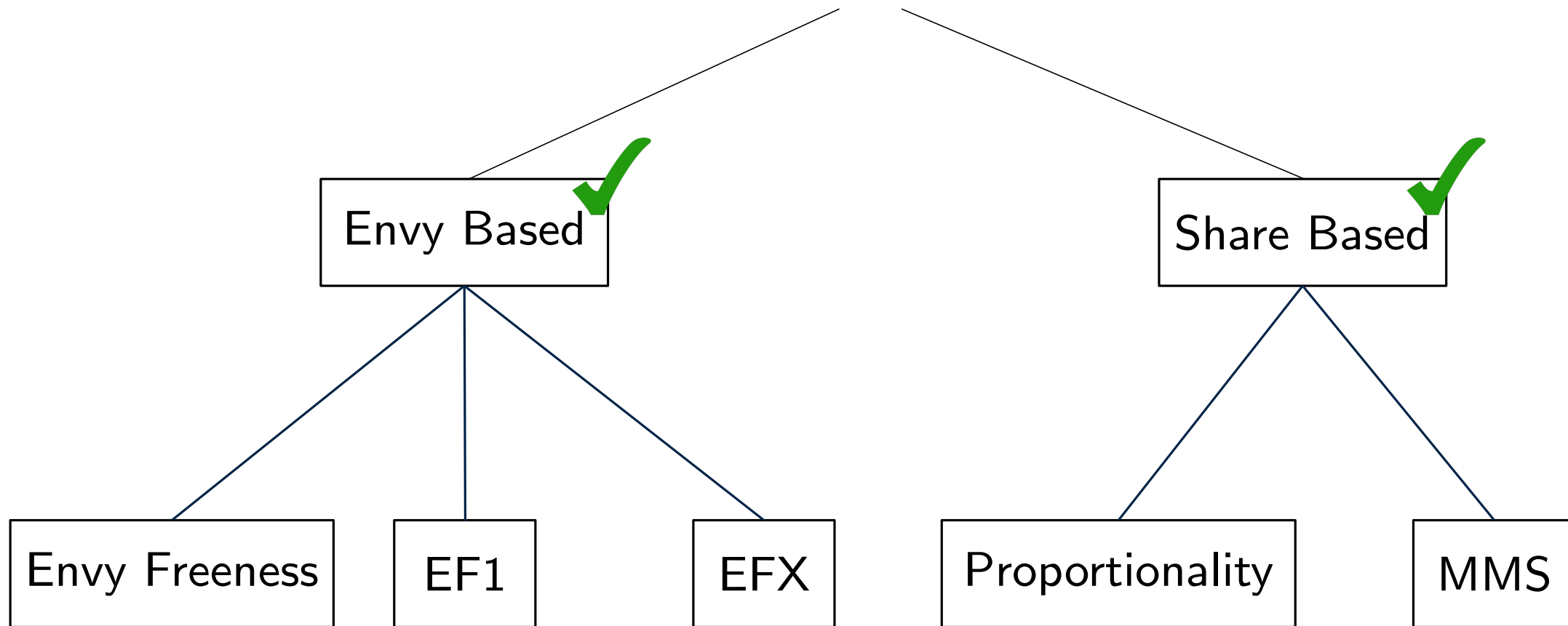
- $3/4$ -MMS allocations exist. [Ghodsi, Hajiaghayi, Seddighin, Seddighin, Yami 2018] [Garg, Taki 2020] [Akrami, Garg, Taki 2023]



Fairness







Fairness



Are we done?



Are we done?





		
	100	1
	1	100



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Are we done?

		
	100	1
	1	100





Is the allocation “fair”?



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Are we done?

		
	100	1
	1	100

Is the allocation “fair”?





- EF1?



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Is the allocation “fair”?





- EF1?
- EFX?



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- MMS?



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Efficiency

Divide **indivisible items** among **agents** in a **fair** and **efficient** manner.



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Definition: Allocation X **pareto dominates** allocation Y , if and only if

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Fairness and Efficiency

		
	100	1
	1	100





✓ Fair

✗ Efficient

		
	100	1
	1	100

✗ Fair

✓ Efficient





		
	100	1
	1	100

✓ Fair

✓ Efficient







Fairness and Efficiency

		
	100	1
	1	100





✓ Fair

✗ Efficient

		
	100	1
	1	100

✗ Fair

✓ Efficient

		
	100	1
	1	100

✓ Fair

✓ Efficient

In this seminar we will see:

- EF1+PO allocations exist and can be computed in pseudopolynomial time.

[Barman, Krishnamurthy, Vaish 2018]



Nash Social Welfare

Definition: Nash social welfare of an allocation X is

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In this seminar we will see:

- $\text{MNW} \implies \text{EF1} + \text{PO}$ [Caragiannis, Kurokawa, Moulin, Procaccia, Shah, Wang 2016]



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Definition: Allocation X is α -**MNW**, if $\text{NSW}(X) \geq \alpha \cdot \text{NSW}(Y)$ for all allocations Y and $\alpha \in [0, 1]$.

In this seminar we will see:

- $\text{MNW} \implies \text{EF1} + \text{PO}$ [Caragiannis, Kurokawa, Moulin, Procaccia, Shah, Wang 2016]
- 1.45^{-1} -MNW allocations can be computed in polynomial time.

[Barman, Krishnamurthy, Vaish 2018]



Recap

Divide **items** among **agents** in a **fair** and **efficient** manner.

Notions of fairness: envy freeness, EF1, EFX, proportionality, MMS, ...

Notions of efficiency: pareto optimality, MNW ...



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Seminar Overview

- 23.04: Introduction on Discrete Fair Division (HA)
- 30.04: Introduction on Cake Cutting (NR)
- 07.05: EFX: A Simpler Approach and an (Almost) Optimal Guarantee via Rainbow Cycle Number [Akrami, Alon, Chaudhury, Garg, Mehlhorn, Mehta 2023] (HA)
 - EFX for 3 agents
- 14.05: Rental Harmony: Sperner's Lemma in Fair Division [Su 1999] (NR)
 - Existence of EF for cake
- 21.05: no lecture
- 28.05: Fair and Efficient Cake Division with Connected Pieces [Arunachaleswaran, Barman, Kumar, Rathi 2019] (student talk)
 - $1/2$ -EF in polytime for cake



Seminar Overview

04.06: The Unreasonable Fairness of Maximum Nash Welfare [Caragiannis, Kurokawa, Moulin, Procaccia, Shah, Wang 2016] (student talk)

- MNW \implies EF1+PO

11.06: A Little Charity Guarantees Almost Envy-Freeness [Chaudhury, Kavitha, Mehlhorn, Sgouritsa 2020] (student talk)

- “good” partial EFX allocation

18.06: no lecture

25.06: Existence and Computation of Epistemic EFX Allocations [Caragiannis, Sharma, Garg, Rathi, Varricchio 2023] (student talk)

- a relaxation of EFX



Seminar Overview

- 02.07: Simplification and Improvement of MMS Approximation [Akrami, Garg, Sharma, Taki 2023] (student talk)
- $3/4$ -MMS
- 09.07: Finding Fair and Efficient Allocations [Barman, Krishnamurthy, Vaish 2018] (student talk)
- 1.45^{-1} -MNW + EF1 + PO
- 16.07: On Approximate Envy-Freeness for Indivisible Chores and Mixed Resources [Bhaskar, Sricharan, Vaish 2021] (student talk)
- EF1 for chores
- 23.07: Best of Both Worlds: Ex-Ante and Ex-Post Fairness in Resource Allocation [Freeman, Shah, Vaish 2020] (student talk)
- randomized allocations



Don't forget!

Send us your preferred list of the student papers by
April 30th.



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