

CSC2556

Lecture 7

Fair Division 2: Indivisible Goods Leximin Allocation

Cake-Cutting (contd)

Indivisible Goods

Pareto Optimality (PO)

- **Definition**

- We say that an allocation $A = (A_1, \dots, A_n)$ is PO if there is no alternative allocation $B = (B_1, \dots, B_n)$ such that

1. Every agent is at least as happy: $V_i(B_i) \geq V_i(A_i), \forall i \in N$

2. Some agent is strictly happier: $V_i(B_i) > V_i(A_i), \exists i \in N$

- I.e., an allocation is PO if there is no “better” allocation.

- **Q:** Is it PO to give the entire cake to player 1?

- **A:** Not necessarily. But yes if player 1 values “every part of the cake positively”.

PO + EF

- **Theorem [Weller '85]:**

- There always exists an allocation of the cake that is both envy-free and Pareto optimal.

- One way to achieve PO+EF:

- **Nash-optimal allocation:** $\operatorname{argmax}_A \prod_{i \in N} V_i(A_i)$
- Obviously, this is PO. The fact that it is EF is non-trivial.
- This is named after John Nash.
 - Nash social welfare = product of utilities
 - Different from utilitarian social welfare = sum of utilities

Nash-Optimal Allocation



- **Example:**

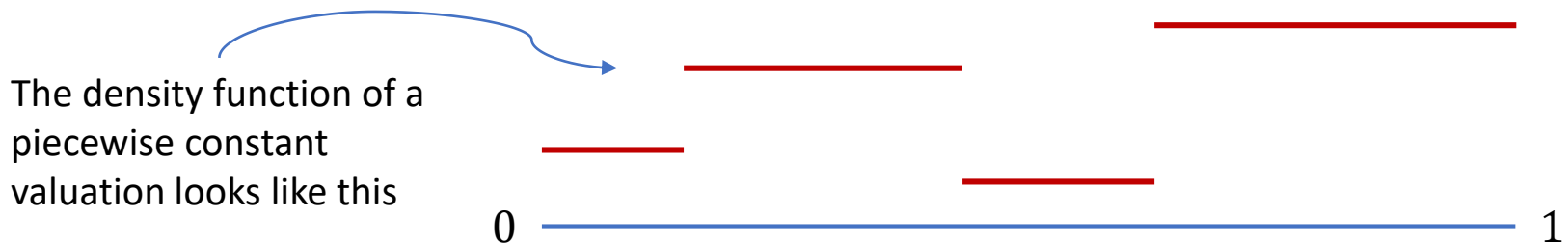
- Green player has value 1 distributed over $[0, 2/3]$
- Blue player has value 1 distributed over $[0, 1]$
- Without loss of generality (why?) suppose:
 - Green player gets x fraction of $[0, 2/3]$
 - Blue player gets the remaining $1 - x$ fraction of $[0, 2/3]$ AND all of $[2/3, 1]$.
- Green's utility = x , blue's utility = $(1 - x) \cdot \frac{2}{3} + \frac{1}{3} = \frac{3-2x}{3}$
- Maximize: $x \cdot \frac{3-2x}{3} \Rightarrow x = 3/4$ ($3/4$ fraction of $2/3$ is $1/2$).



Green has utility $\frac{3}{4}$
 Blue has utility $\frac{1}{2}$

Problem with Nash Solution

- Difficult to compute in general
 - I believe it should require an unbounded number of queries in the Robertson-Webb model. But I can't find such a result in the literature.
- **Theorem [Aziz & Ye '14]:**
 - For *piecewise constant* valuations, the Nash-optimal solution can be computed in polynomial time.



Interlude:

Homogeneous Divisible Goods

- Suppose there are m homogeneous divisible goods
 - Each good can be divided fractionally between the agents
- Let $x_{i,g}$ = fraction of good g that agent i gets
 - Homogeneous = agent doesn't care which "part"
 - E.g., CPU or RAM
- Special case of cake-cutting
 - Line up the goods on $[0,1]$ → piecewise uniform valuations

Interlude: Homogeneous Divisible Goods

- Nash-optimal solution:

Maximize $\sum_i \log U_i$

$$U_i = \sum_g x_{i,g} * v_{i,g} \quad \forall i$$

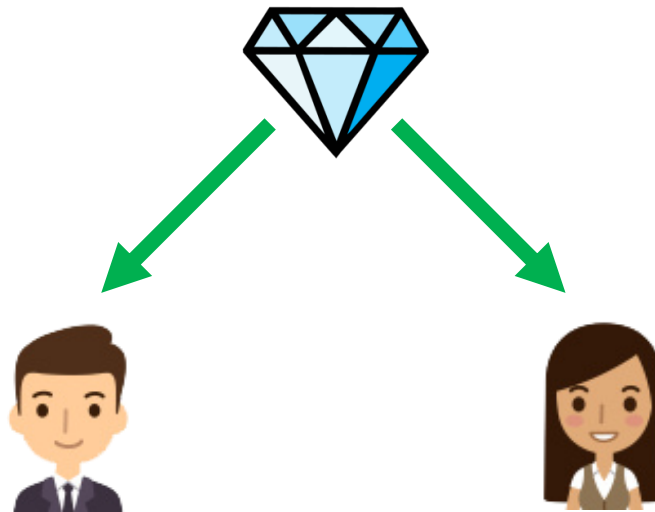
$$\sum_i x_{i,g} = 1 \quad \forall g$$

$$x_{i,g} \in [0,1] \quad \forall i, g$$

- Gale-Eisenberg Convex Program
 - Polynomial time solvable

Indivisible Goods

- Goods which cannot be shared among players
 - E.g., house, painting, car, jewelry, ...
- **Problem:** Envy-free allocations may not exist!





Indivisible Goods: Setting

				
	8	7	20	5
	9	11	12	8
	9	10	18	3

Given such a matrix of numbers, assign each good to a player.

We assume additive values. So, e.g., $V_{\text{man}}(\{\text{painting}, \text{car}\}) = 8 + 7 = 15$

Indivisible Goods

				
	8	7	20	5
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Indivisible Goods

- Envy-freeness up to one good (EF1):

$$\forall i, j \in N, \exists g \in A_j : V_i(A_i) \geq V_i(A_j \setminus \{g\})$$

- Technically, we need either this or $A_j = \emptyset$.
 - “If i envies j , there must be some good in j ’s bundle such that removing it would make i envy-free of j .”
- Does there always exist an EF1 allocation?


EF1

- Yes! We can use **Round Robin**.
 - Agents take turns in cyclic order: $1, 2, \dots, n, 1, 2, \dots, n, \dots$
 - In her turn, an agent picks the good she likes the most among the goods still not picked by anyone.
- Observation: This always yields an EF1 allocation.
 - Informal proof on the board.
- Sadly, on some instances, this returns an allocation that is **not Pareto optimal**.








EF1+PO?

- Nash welfare to rescue!
- **Theorem [Caragiannis et al. '16]:**
 - The allocation $\operatorname{argmax}_A \prod_{i \in N} V_i(A_i)$ is EF1 + PO.
 - Note: This maximization is over only “integral” allocations that assign each good to some player in whole.
 - Note: Subtle tie-breaking if all allocations have zero Nash welfare.
 - Step 1: Choose a subset of players $S \subseteq N$ with largest $|S|$ such that it is possible to give a positive utility to every player in S simultaneously.
 - Step 2: Choose $\operatorname{argmax}_A \prod_{i \in S} V_i(A_i)$








Integral Nash Allocation

				
	8	7	20	5
	9	11	12	8
	9	10	18	3








$$20 * 8 * (9+10) = 3040$$

				
	8	7	20	5
	9	11	12	8
	9	10	18	3




$$(8+7) * 8 * 18 = 2160$$

				
	8	7	20	5
	9	11	12	8
	9	10	18	3

$$8 * (12+8) * 10 = 1600$$

				
	8	7	20	5
	9	11	12	8
	9	10	18	3

$$20 * (11+8) * 9 = 3420$$

				
	8	7	20	5
	9	11	12	8
	9	10	18	3

Computation

- For indivisible goods, Nash-optimal solution is strongly NP-hard to compute
 - That is, remains NP-hard even if all values in the matrix are bounded
- **Open Question:** If our goal is EF1+PO, is there a different polynomial time algorithm?
 - Not sure. But a recent paper gives a pseudo-polynomial time algorithm for EF1+PO
 - Time is polynomial in n , m , and $\max_{i,g} V_i(\{g\})$.

Other Fairness Notions

- **Maximin Share Guarantee (MMS):**

- Generalization of “cut and choose” for n players
- MMS value of player i =
 - The highest value player i can get...
 - If *she* divides the goods into n bundles...
 - But receives the worst bundle for her (“worst case guarantee”)
- Let $\mathcal{P}_n(M)$ denote the family of partitions of the set of goods M into n bundles.

$$MMS_i = \max_{(B_1, \dots, B_n) \in \mathcal{P}_n(M)} \min_{k \in \{1, \dots, n\}} V_i(B_k).$$

- An allocation is **α -MMS** if every player i receives value at least $\alpha * MMS_i$.

Other Fairness Notions

- Maximin Share Guarantee (MMS)

- [Procaccia, Wang '14]:

There is an example in which no MMS allocation exists.

- [Procaccia, Wang '14]:

A $2/3$ - MMS allocation always exists.

- [Ghodsi et al. '17]:

A $3/4$ - MMS allocation always exists.

- [Caragiannis et al. '16]:

The Nash-optimal solution is $\frac{2}{1+\sqrt{4n-3}}$ -MMS, and this is the best possible guarantee.

Stronger Fairness

- **Open Question:** Does there always exist an EFX allocation?
- **EF1:** $\forall i, j \in N, \exists g \in A_j : V_i(A_i) \geq V_i(A_j \setminus \{g\})$
 - Intuitively, i doesn't envy j if she gets to **remove her most valued item** from j 's bundle.
- **EFx:** $\forall i, j \in N, \forall g \in A_j : V_i(A_i) \geq V_i(A_j \setminus \{g\})$
 - Note: Need to quantify over g such that $V_i(\{g\}) > 0$.
 - Intuitively, i doesn't envy j even if she **removes her least positively valued item** from j 's bundle.

Stronger Fairness

- The difference between EF1 and EFX:
 - Suppose there are two players and three goods with values as follows.

	A	B	C
P1	5	1	10
P2	0	1	10

- If you give $\{A\} \rightarrow P1$ and $\{B,C\} \rightarrow P2$, it's EF1 but not EFX.
 - EF1 because if P1 removes C from P2's bundle, all is fine.
 - Not EFX because removing B doesn't eliminate envy.
- Instead, $\{A,B\} \rightarrow P1$ and $\{C\} \rightarrow P2$ would be EFX.

Allocation of Bads

- **Negative utilities** (costs instead of values)
 - Let $c_{i,b}$ be the cost of player i for bad b .
 - $C_i(S) = \sum_{b \in S} c_{i,b}$
 - **EF**: $\forall i, j \ C_i(A_i) \leq C_i(A_j)$
 - **PO**: There should be no alternative allocation in which no player has more cost, and some player has less cost.
- **Divisible bads**
 - **EF + PO allocation always exists**, like for divisible goods.
 - One way to achieve is through “Competitive Equilibria” (CE).
 - For divisible goods, Nash-optimal allocation is the unique CE.
 - For bads, exponentially many CE.

Allocation of Bads

- **Indivisible bads**

- **EF1:** $\forall i, j \exists b \in A_i \ c_i(A_i \setminus \{b\}) \leq c_i(A_j)$

- **EFx:** $\forall i, j \ \forall b \in A_i \ c_i(A_i \setminus \{b\}) \leq c_i(A_j)$

- Note: Again, we need to restrict to b such that $c_{i,b} > 0$

- **Open Question 1:**

- Does an EF1 + PO allocation always exist?

- **Open Question 2:**

- Does an EFx allocation always exist?

- More open questions related to relaxations of proportionality

Leximin (DRF)

Computational Resources

- **Resources:** Homogeneous divisible resources like CPU, RAM, or network bandwidth
- **Valuations:** Each player wants the resources in a **fixed proportion** (Leontief preferences)
- **Example:**
 - Player 1 requires (2 CPU, 1 RAM) for each copy of task
 - Indifferent between (4,2) and (5,2), but prefers (5,2.5)
 - “fractional” copies are allowed

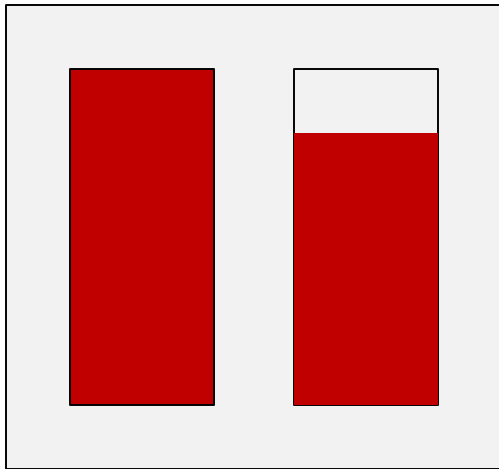
Model

- Set of **players** $N = \{1, \dots, n\}$
- Set of **resources** $R, |R| = m$
- **Demand** of player i is $d_i = (d_{i1}, \dots, d_{im})$
 - $0 < d_{ir} \leq 1$ for every $r, d_{ir} = 1$ for some r
 - “For every 1% of the total available CPU you give me, I need 0.5% of the total available RAM”
- **Allocation**: $A_i = (A_{i1}, \dots, A_{im})$ where A_{ir} is the fraction of available resource r allocated to i
 - Utility to player i : $u_i(A_i) = \min_{r \in R} A_{ir} / d_{ir}$.
 - We’ll assume a **non-wasteful** allocation
 - Allocates resources proportionally to the demand.

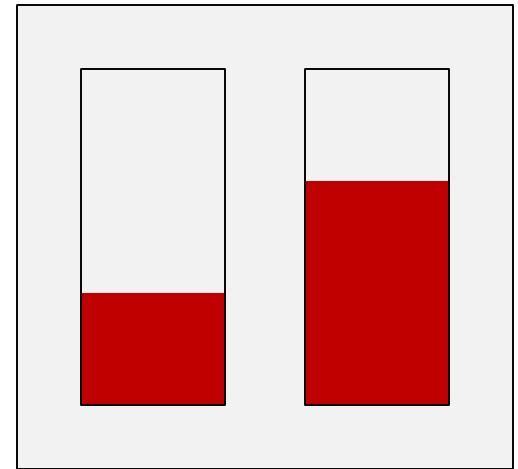
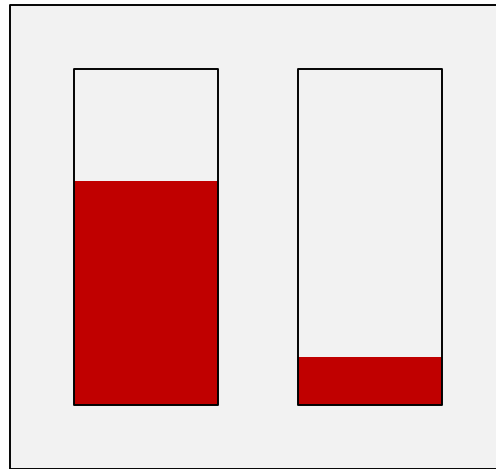
Dominant Resource Fairness

- **Dominant resource** of i is r such that $d_{ir} = 1$
- **Dominant share** of i is A_{ir} , where $r =$ dominant resource of i
- Dominant Resource Fairness (**DRF**) Mechanism
 - Allocate maximal resources while maintaining equal dominant shares.

DRF animated



Total



Properties of DRF

- **Envy-free:** $u_i(A_i) \geq u_i(A_j), \forall i, j$
 - Why? [Note: EF no longer implies proportionality.]
- **Proportionality:** $u_i(A_i) \geq 1/n, \forall i$
 - Why?
- **Pareto optimality** (Why?)
- **Group strategyproofness:**
 - If a group of players manipulate, it can't be that none of them lose, and at least one of them gains.
 - We'll skip this proof.

The Leximin Mechanism

- Generalizes the DRF Mechanism
- Mechanism:
 - Choose an allocation A that
 - Maximizes $\min_i u_i(A_i)$
 - Among all minimizers, breaks ties in favor of higher second minimum utility.
 - Among all minimizers, breaks ties in favor of higher third minimum utility.
 - And so on...
- Maximizes the egalitarian welfare

The Leximin Mechanism

- DRF is the leximin mechanism
 - In the previous illustration, we didn't need tie-breaking because we assumed $d_{ir} > 0$ for every $i \in N, r \in R$.
 - In practice, not all the players need all the resources.
 - When $d_{ir} = 0$ is allowed, we need to continue allocating even after some agents are saturated.
 - Not all agents have equal dominant shares in the end.
- **Theorem [Parkes, Procaccia, S '12]:**
 - When $d_{ir} = 0$ is allowed, the leximin mechanism still retains all four properties (proportionality, envy-freeness, Pareto optimality, group strategyproofness).

A Note on Dynamic Settings

- We assumed that all agents are present from the start, and we want a one-shot allocation.
- Real-life environments are dynamic. Agents arrive and depart, and their demands change over time.
- **Theorem [Kash, Procaccia, S '14]:**
 - A dynamic version of the leximin mechanism satisfies proportionality, Pareto optimality, and strategyproofness along with a relaxed version of envy-freeness when agents arrive one-by-one.

A Note on Dynamic Settings

- Dynamic mechanism design
 - Designing fair, efficient, and game-theoretic mechanisms in dynamic environments is a relatively new research area, and we do not know much.
 - E.g., what if agents can depart, demands can change over time, or agents can submit and withdraw multiple jobs over time?
 - Lots of open questions!

Leximin (Dichotomous Matching)

Matching + Dichotomous Prefs

- Recall the stable matching setting of **matching n men to n women**.
 - We assumed ranked preferences, and showed that the Gale-Shapley algorithm produces a stable matching.
 - What if agent preferences weren't ranked?
- Suppose the men and women have **dichotomous preferences** over each other.
 - Each man finds a subset of women “acceptable” (utility 1), and the rest “unacceptable” (utility 0).
 - Same for women's preferences over men.

Matching + Dichotomous Prefs

- Dichotomous preferences induce a bipartite graph between men and women.
 - If a perfect matching exists, it's awesome.
 - What if there is no perfect matching?
 - Any deterministic matching unfairly gives 0 utility to some agents.
 - Solution: randomize!
- Under a random matching, utility to an agent = probability of being matched to an acceptable partner.

Matching + Dichotomous Prefs

- (Integral) Matching:
 - “Select” or “not select” each edge such that the number of selected edges incident on each vertex is at most 1.
- Fractional Matchings:
 - “Put a weight” on each edge such that the total weight of edges incident on each vertex is at most 1.
- Birkoff von-Neumann Theorem:
 - Every fractional matching can be “implemented” as a probability distribution over integral matchings.

Matching + Dichotomous Prefs

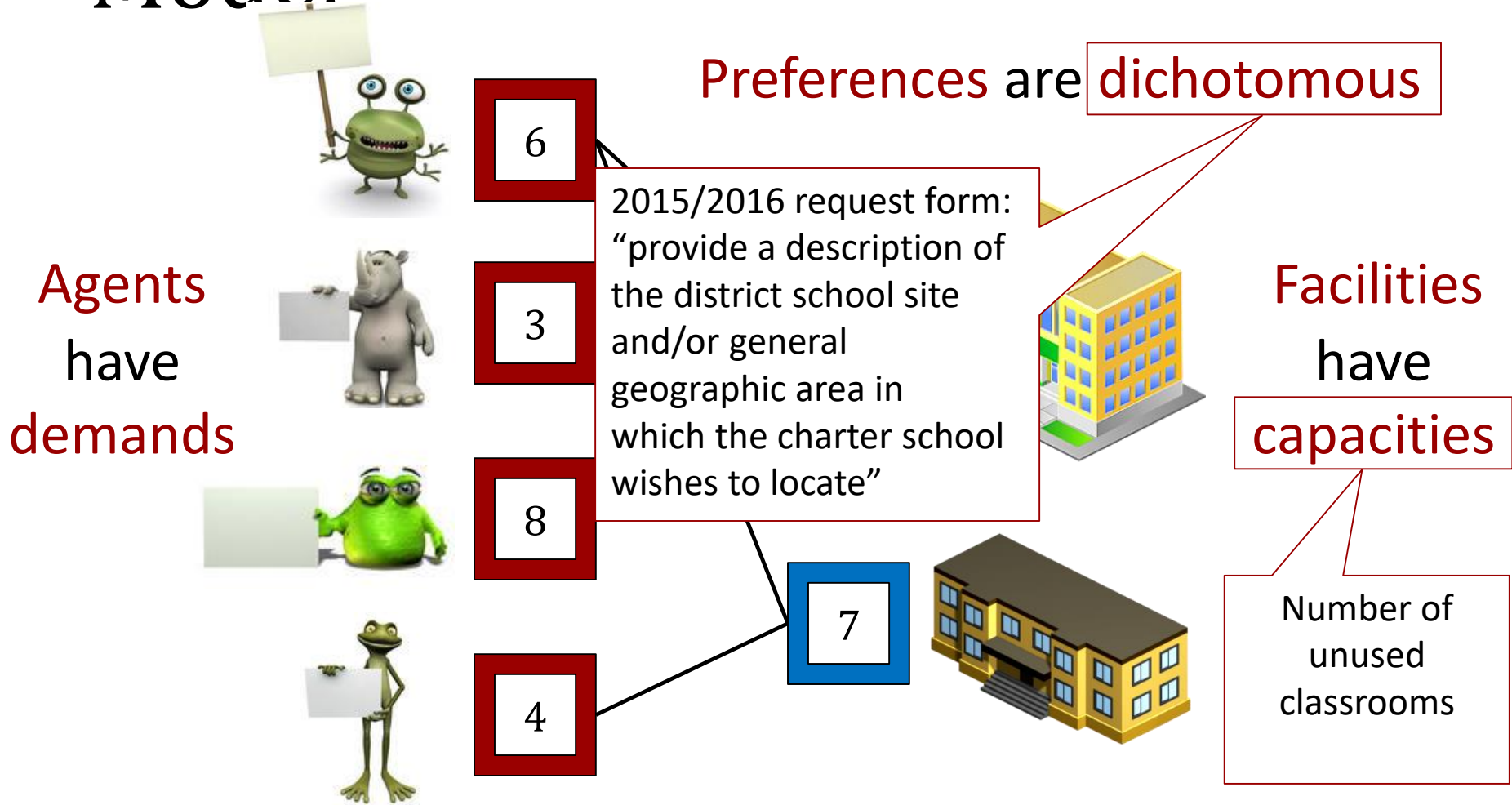
- **Randomized leximin mechanism:**
 - Compute the leximin fractional matching, and implement it as a distribution over integral matchings.
 - Both steps are doable in polynomial time!
- **Theorem [Bogomolnaia, Moulin '04]:**
 - The randomized leximin mechanism satisfies proportionality, envy-freeness, Pareto optimality, and group-strategyproofness (for both sides).
- In contrast: For ranked preferences, no algorithm can be strategyproof for both sides.

Matching with Capacities

- **Proposition 39** in California
 - “Unused resources in public schools should be *fairly* allocated to local charter schools that desire them.”
- Each charter school (**agent**) i wants d_i unused classrooms at one of the acceptable public schools (**facilities**) F_i .
 - If the demand is met, the charter school can relocate to the public school facility.
- Each facility j has c_j unused classrooms.
 - We assume facilities don't have preferences over agents.

Leximin (Classroom Allocation)

Model



Leximin Strikes Again

- Utility of agent i under a randomized allocation = probability of being allocated d_i classrooms at one of the facilities in F_i .
- **Theorem [Kurokawa, Procaccia, S '15]:**
 - The randomized leximin mechanism satisfies proportionality, envy-freeness, Pareto optimality, and group strategyproofness.
- Computing this allocation is NP-hard.
 - Unlike DRF and matching under dichotomous preferences.

Leximin Strikes Again

- The result holds in a generic domain which satisfies:
 - **Convexity:** If two utility vectors are feasible, then so should be their convex combinations.
 - Holds if fractional or randomized allocations are allowed.
 - **Equality:** The maximum utility of each agent should be the same.
 - Normalize utilities.
 - **Shifting Allocations:** Swapping allocations of two agents should be allowed.
 - **Maximal Utilization:** No agent should have a higher utility for agent i 's allocation than agent i has.
 - This should hold after the normalization. This is the most restrictive assumption.
- Captures DRF, matching with dichotomous preferences, classroom allocation, and many other settings from the literature.