

Recent Advances in Fair Resource Allocation

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Disclaimer

- In this tutorial, we will NOT
 - Assume any prior knowledge of fair division
 - Walk you through tedious, detailed proofs
 - Claim to present a *complete* overview of the entire fair division realm
- Instead, we will
 - Focus mostly on the case of “additive preferences” for coherence
 - With some results for and pointers to domains with non-additive preferences
- If you spot any errors, missing results, or incorrect attributions:
 - Please email nisarg@cs.toronto.edu or Rupert.Freeman@microsoft.com

Outline

- Fairness Axioms
 - Proportionality
 - Envy-freeness
 - Maximin share guarantee
 - Groupwise fairness
 - Core
 - Group envy-freeness
 - Groupwise MMS
 - Group fairness
- Implications of fairness
 - Price of fairness
 - Interplay with strategyproofness and Pareto optimality
 - Restricted cases
- Settings
 - Cake-cutting
 - Homogeneous divisible goods
 - Indivisible goods

A Generic Resource Allocation Framework

- A set of **agents** $N = \{1, 2, \dots, n\}$
- A set of **resources** M
 - May be finite or infinite
- **Valuations**
 - Valuation of agent i is $v_i : 2^M \rightarrow \mathbb{R}$
 - Range is \mathbb{R}_+ when resources are *goods*, and \mathbb{R}_- when they are *bads*
- **Allocations**
 - $A = (A_1, \dots, A_n) \in \Pi_n(M)$ is a partition of resources among agents
 - $A_i \cap A_j = \emptyset, \forall i, j \in N$ and $\cup_{i \in N} A_i = M$
 - A **partial allocation** A may have $\cup_{i \in N} A_i \neq M$

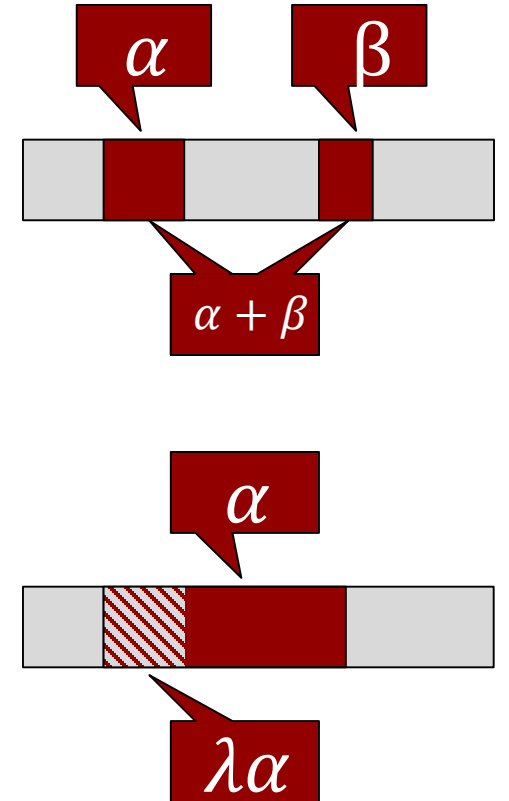
Cake Cutting

- Formally introduced by Steinhaus [1948]
- Agents: $N = \{1, 2, \dots, n\}$
- Resource (cake): $M = [0, 1]$
- Constraints on an allocation A
 - The entire cake is allocated (**full** allocation)
 - Each $A_i \in \mathcal{A}$, where \mathcal{A} is the set of finite unions of disjoint intervals
- **Simple** allocations
 - Each agent is allocated a single interval
 - Cuts cake at $n - 1$ points



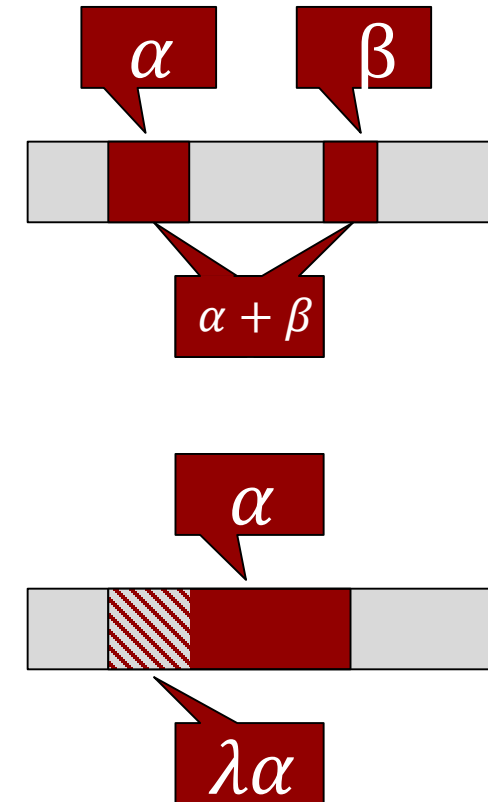
Agent Valuations

- Each agent i has an integrable density function $f_i: [0,1] \rightarrow \mathbb{R}_+$
- For each $X \in \mathcal{A}$, $v_i(X) = \int_{x \in X} f_i(x) dx$
- For normalization, we require $\int_0^1 f_i(x) dx = 1$
 - Without loss of generality



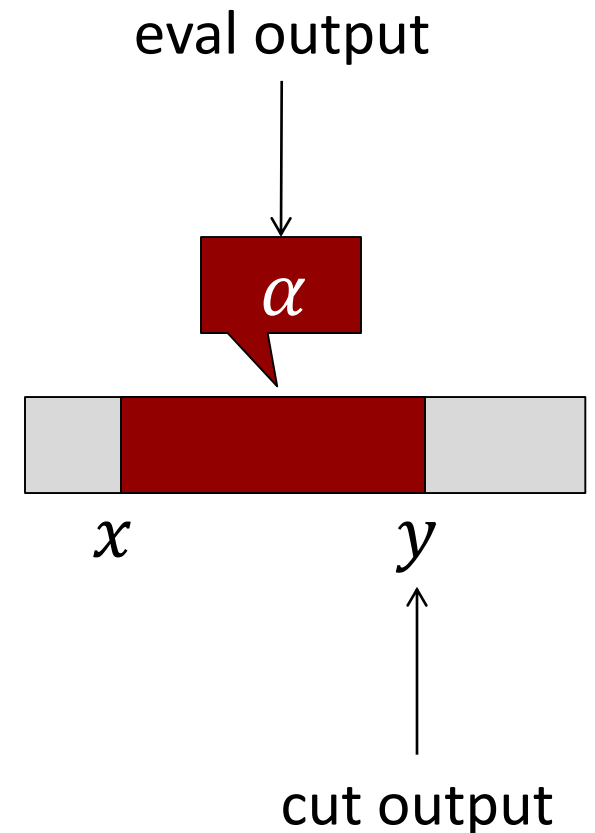
Agent Valuations

- In this model, the valuations satisfy the following properties
- **Normalized:** $v_i([0,1]) = 1$
- **Divisible:** $\forall \lambda \in [0,1]$ and $I = [x, y]$,
 $\exists z \in [x, y]$ s.t. $v_i([x, z]) = \lambda v_i([x, y])$
- **Additive:** For disjoint intervals I and I' ,
 $v_i(I) + v_i(I') = v_i(I \cup I')$



Complexity

- Inputs are functions
 - Infinitely many bits may be needed to fully represent the input
 - Query complexity is more useful
- **Robertson-Webb Model**
 - $\text{Eval}_i(x, y)$ returns $v_i([x, y])$
 - $\text{Cut}_i(x, \alpha)$ returns y such that $v_i([x, y]) = \alpha$

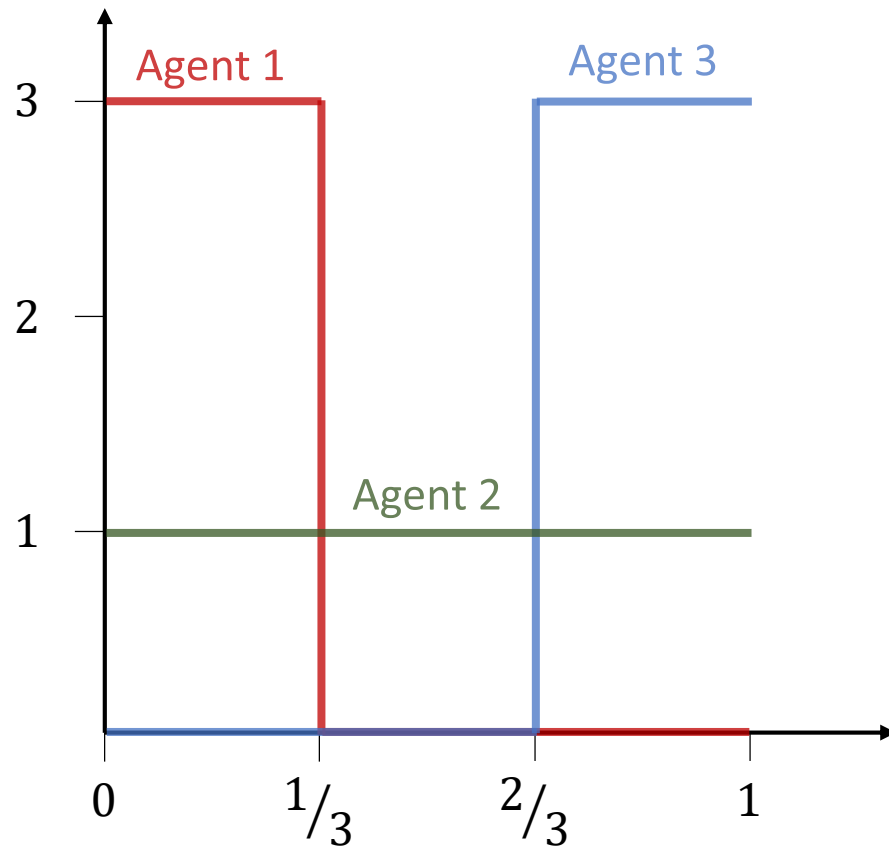


Three Classic Fairness Desiderata

- **Proportionality (Prop):** $\forall i \in N: v_i(A_i) \geq 1/n$
 - Each agent should receive her “fair share” of the utility.
- **Envy-Freeness (EF):** $\forall i, j \in N: v_i(A_i) \geq v_i(A_j)$
 - No agent should wish to swap her allocation with another agent.
- **Equitability (EQ):** $\forall i, j \in N : v_i(A_i) = v_j(A_j)$
 - All agents should have the exact same value for their allocations.
 - No agent should be jealous of what another agent received.

Example

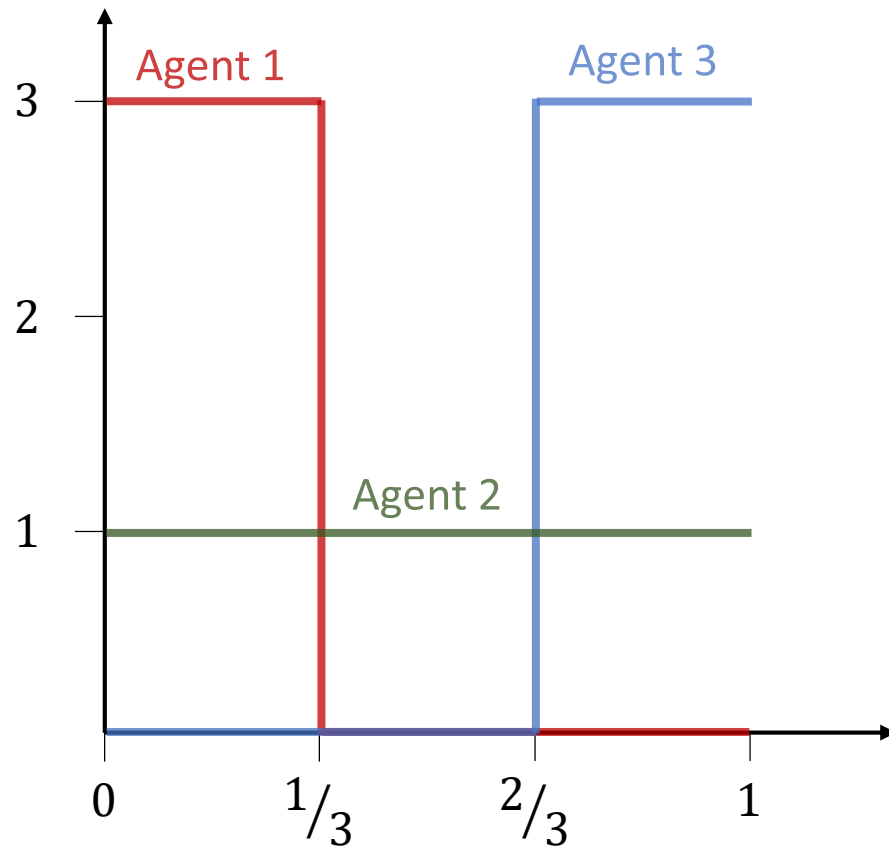
- Value density functions



- Agent 1 wants $[0, 1/3]$ uniformly and does not want anything else
- Agent 2 wants the entire cake uniformly
- Agent 3 wants $[2/3, 1]$ uniformly and does not want anything else

Example

- Value density functions

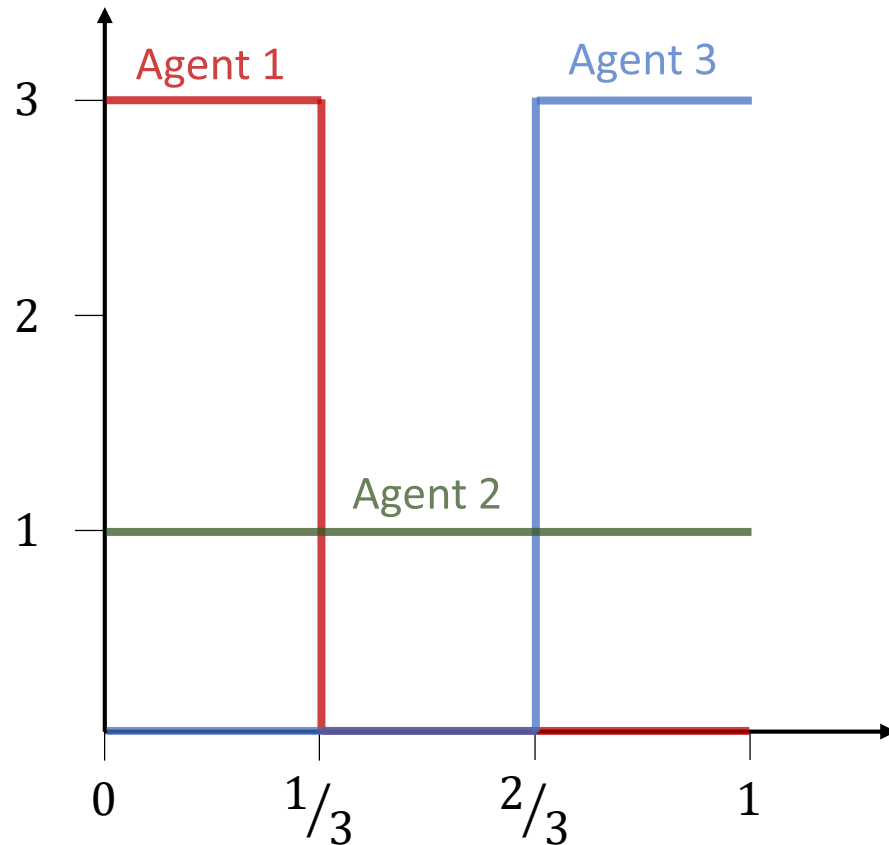


- Consider the following allocation

- $A_1 = [0, 1/9] \Rightarrow v_1(A_1) = 1/3$
- $A_2 = [1/9, 8/9] \Rightarrow v_2(A_2) = 7/9$
- $A_3 = [8/9, 1] \Rightarrow v_3(A_3) = 1/3$
- The allocation is proportional, but not envy-free or equitable

Example

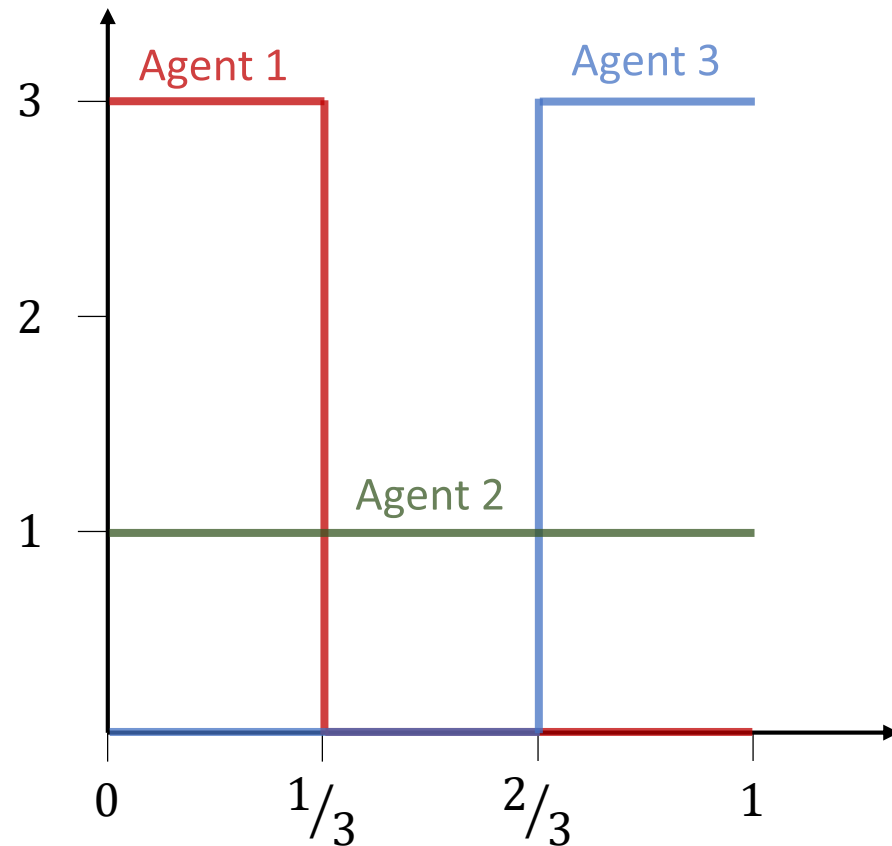
- Value density functions



- Consider the following allocation
- $A_1 = [0, 1/6] \Rightarrow v_1(A_1) = 1/2$
- $A_2 = [1/6, 5/6] \Rightarrow v_2(A_2) = 2/3$
- $A_3 = [5/6, 1] \Rightarrow v_3(A_3) = 1/2$
- The allocation is proportional and envy-free, but not equitable

Example

- Value density functions



- Consider the following allocation

- $A_1 = [0, 1/5] \Rightarrow v_1(A_1) = 3/5$
- $A_2 = [1/5, 4/5] \Rightarrow v_2(A_2) = 3/5$
- $A_3 = [4/5, 1] \Rightarrow v_3(A_3) = 3/5$
- The allocation is proportional, envy-free, and equitable

Relations Between Fairness Desiderata

- Envy-freeness implies proportionality
 - Summing $v_i(A_i) \geq v_i(A_j)$ over all j gives proportionality
- For 2 agents, proportionality also implies envy-freeness
 - Hence, they are equivalent.
- Equitability is incomparable to proportionality and envy-freeness
 - E.g. if each agent has value 0 for her own allocation and 1 for the other agent's allocation, it is equitable but not proportional or envy-free.

Existence

- **Theorem [Alon, 1987]**

Suppose the value density function f_i of each agent valuation v_i is continuous. Then, we can cut the cake at $n^2 - n$ places and rearrange the $n^2 - n + 1$ intervals into n pieces A_1, \dots, A_n such that

$$v_i(A_j) = 1/n, \forall i, j \in N$$

- This is called a “**perfect partition**”
 - It is trivially envy-free (thus proportional) and equitable
- As we will later see, this cannot be found with finitely many queries in Robertson-Webb model

Proportionality

PROPORTIONALITY : $n = 2$ AGENTS

- CUT-AND-CHOOSE

- Agent 1 cuts the cake at x such that $v_1([0, x]) = v_1([x, 1]) = 1/2$
- Agent 2 chooses the piece that she prefers.

- Elegant protocol

- Proportional (equivalent to envy-freeness for 2 agents)
- Needs only one cut and one eval query (optimal)

- More agents?

PROPORTIONALITY: DUBINS-SPANIER

- **DUBINS-SPANIER**

- Referee starts a knife at 0 and moves the knife to the right.
- Repeat: When the piece to the left of the knife is worth $1/n$ to an agent, the agent shouts “stop”, receives the piece, and exits.
- When only one agent remains, she gets the remaining piece.

- Can be implemented easily in Robertson-Webb model

- When $[x, 1]$ is left, ask each remaining agent i to cut at y_i so that $v_i([x, y_i]) = 1/n$, and give agent $i^* \in \arg \min_i y_i$ the piece $[x, y_{i^*}]$.

- **Query complexity: $\Theta(n^2)$**

PROPORTIONALITY: EVEN-PAZ

- EVEN-PAZ
- Input:
 - Interval $[x, y]$, number of agents n (assume a power of 2 for simplicity)
- Recursive procedure:
 - If $n = 1$, give $[x, y]$ to the single agent.
 - Otherwise:
 - Each agent i marks z_i such that $v_i([x, z_i]) = v_i([z_i, y])$
 - $z^* = (n/2)^{\text{th}}$ mark from the left.
 - Recurse on $[x, z^*]$ with the left $n/2$ agents, and on $[z^*, y]$ with the right $n/2$ agents.
- Query complexity: $\Theta(n \log n)$

Complexity of Proportionality

- Theorem [Edmonds and Pruhs, 2006]:
 - Any protocol returning a proportional allocation needs $\Omega(n \log n)$ queries in the Robertson-Webb model.
- Hence, EVEN-PAZ is provably (asymptotically) optimal!

Envy-Freeness

Envy-Freeness : Few Agents

- $n = 2$ agents : CUT-AND-CHOOSE (2 queries)
- $n = 3$ agents : SELFRIDGE-CONWAY (14 queries)

Gets complex pretty quickly!

Suppose we have three players **P1**, **P2** and **P3**. Where the procedure gives a criterion for a decision it means that criterion gives an optimum choice for the player.

1. **P1** divides the cake into three pieces he considers of equal size.
2. Let's call **A** the largest piece according to **P2**.
3. **P2** cuts off a bit of **A** to make it the same size as the second largest. Now **A** is divided into: the trimmed piece **A1** and the trimmings **A2**. Leave the trimmings **A2** to the side for now.
 - If **P2** thinks that the two largest parts are equal (such that no trimming is needed), then each player chooses a part in this order: **P3**, **P2** and finally **P1**.
4. **P3** chooses a piece among **A1** and the two other pieces.
5. **P2** chooses a piece with the limitation that if **P3** didn't choose **A1**, **P2** must choose it.
6. **P1** chooses the last piece leaving just the trimmings **A2** to be divided.

It remains to divide the trimmings **A2**. The trimmed piece **A1** has been chosen by either **P2** or **P3**; let's call the player who chose it **PA** and the other player **PB**.

1. **PB** cuts **A2** into three equal pieces.
2. **PA** chooses a piece of **A2** - we name it **A21**.
3. **P1** chooses a piece of **A2** - we name it **A22**.
4. **PB** chooses the last remaining piece of **A2** - we name it **A23**.

Envy-Freeness : Few Agents

- [Brams and Taylor, 1995]
 - The first finite (but unbounded) protocol for any number of agents
- [Aziz and Mackenzie, 2016a]
 - The first bounded protocol for 4 agents (at most 203 queries)
- [Amanatidis et al., 2018]
 - A simplified version of the above protocol for 4 agents (at most 171 queries)

Envy-Freeness

- Theorem [Aziz and Mackenzie, 2016b]
 - There exists a bounded protocol for computing an envy-free allocation with n agents, which requires $O(n^{n^{n^{n^{\mathbf{n}}}}})$ queries
 - After $O(n^{2n+3})$ queries, the protocol can output a **partial** allocation that is both proportional and envy-free
- What about lower bounds?

Complexity of Envy-Freeness

- Theorem [Procaccia, 2009]

Any protocol for finding an envy-free allocation requires $\Omega(n^2)$ queries.

Open Problem

Bridge the gap between $O(n^{n^{n^{n^{\dots}}}})$ upper bound and $\Omega(n^2)$ lower bound for envy-free cake-cutting

- Theorem [Stromquist, 2008]

There is no finite (even unbounded) protocol for finding a simple envy-free allocation for $n \geq 3$ agents.

Equitability

Upper Bound: $n = 2$ Agents

- Existence
 - Suppose we cut the cake at x to form pieces $[0, x]$ and $[x, 1]$
 - Let $f(x) = v_1([0, x]) - v_2([x, 1])$
 - Note that $f(0) = -1$, $f(1) = 1$, and f is continuous
 - By the intermediate value theorem: $\exists x^*$ such that $f(x^*) = 0$
 - Allocation $A_1 = [0, x^*]$ and $A_2 = [x^*, 1]$ is equitable
- Theorem [Cechlárová and Pillárová, 2012]
 - Using binary search for x^* , we can find an ϵ -equitable allocation for 2 agents with $O(\ln(1/\epsilon))$ queries.

Upper Bound: $n > 2$ Agents

- Theorem [Cechlárová and Pillárová, 2012]
 - This technique can be extended to n agents to find an ϵ -equitable allocation in $O(n \ln(1/\epsilon))$ queries.
- Theorem [Procaccia and Wang, 2017]
 - There exists a protocol for n agents which finds an ϵ -equitable allocation in $O(1/\epsilon \ln(1/\epsilon))$ queries.
 - Intuition:
 - If $n \leq 1/\epsilon$, use above protocol for finding an equitable ϵ -equitable allocation.
 - If $n > 1/\epsilon$, use a variant of the Evan-Paz algorithm to find an *anti-proportional* allocation where $n' = \lceil 1/\epsilon \rceil$ agents get value *at most* $1/n'$, and the rest receive nothing.
 - While this is a “bad” allocation, it is ϵ -equitable.

Lower Bound

- Theorem [Procaccia and Wang, 2017]

Any protocol for finding an ϵ -equitable allocation must require $\Omega\left(\frac{\ln(1/\epsilon)}{\ln \ln(1/\epsilon)}\right)$ queries.

- Theorem [Procaccia and Wang, 2017]

There is no finite (even if unbounded) protocol for finding an equitable allocation.

- Non-existence of bounded protocols follows from the previous result.
- But their proof works for non-existence of unbounded protocols as well.

Price of Fairness

Price of Fairness

- Measures the **worst-case loss in social welfare** due to requirement of a fairness property X
- **Social welfare** of allocation A is the sum of values of the agents
 - Denoted $sw(A) = \sum_{i \in N} v_i(A_i)$
- Let \mathcal{F} denote the set of feasible allocations and \mathcal{F}_X denote the set of feasible allocations satisfying property X

$$PoF_X = \max_{v_1, \dots, v_n} \frac{\max_{A \in \mathcal{F}} sw(A)}{\max_{A \in \mathcal{F}_X} sw(A)}$$

Price of Fairness

- Theorem [Caragiannis et al., 2009]

For cake-cutting, the price of proportionality is $\Theta(\sqrt{n})$, and the price of equitability is $\Theta(n)$.

- Because EF implies Prop, clearly price of EF is $\Omega(\sqrt{n})$
- But there is no $o(n)$ upper bound known

Open Problem

Close the gap between $\Omega(\sqrt{n})$ and $O(n)$ for the price of EF

Efficiency

Efficiency

- Weak Pareto optimality (WPO)

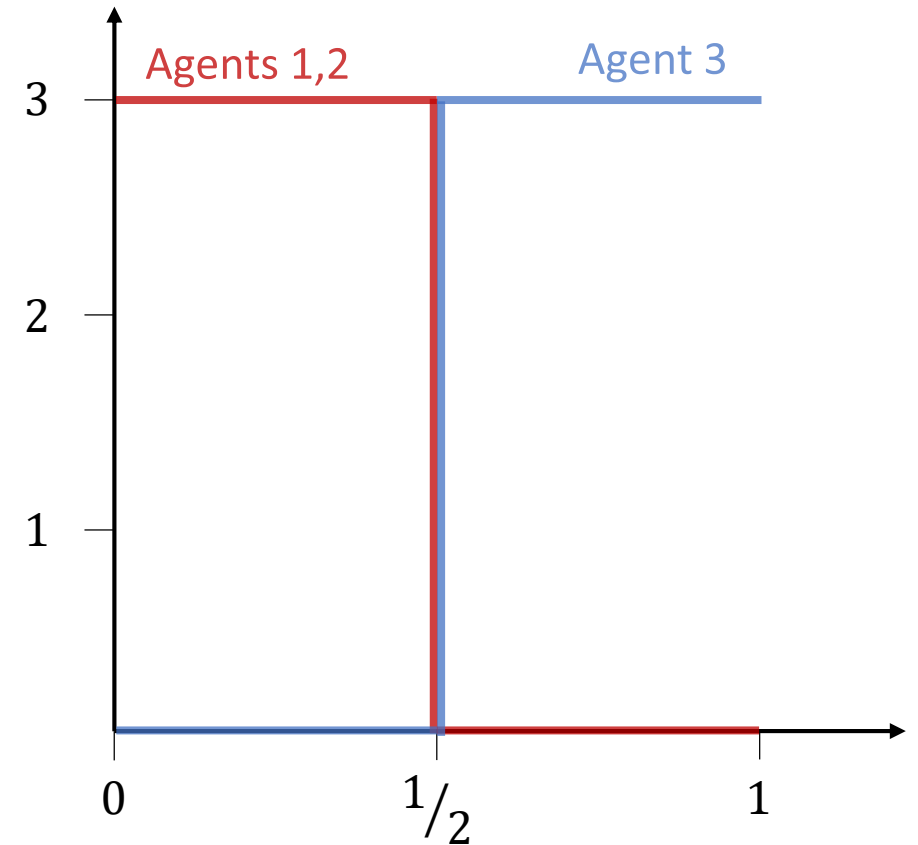
- Allocation A is weakly Pareto optimal if there is no allocation B such that $v_i(B_i) > v_i(A_i)$ for all $i \in N$.
- “Can’t make everyone happier”

- Pareto optimality (PO)

- Allocation A is Pareto optimal if there is no allocation B such that $v_i(B_i) \geq v_i(A_i)$ for all agents $i \in N$, and at least one inequality is strict.
- “Can’t make someone happier without making someone else less happy”
- Easy to achieve in isolation (e.g. “serial dictatorship”)

PO+EF+EQ: (Non-)Existence

- **Theorem [Barbanel and Brams, 2011]**
With two agents, there always exists an allocation that is envy-free (thus proportional), equitable, and Pareto optimal.
 - Their algorithm has similarities to the more popular “adjusted winner” algorithm, which we will see later in the tutorial.
- **With $n \geq 3$ agents, PO+EQ is impossible**



What about PO+EF?

- Competitive Equilibrium from Equal Incomes (CEEI)

- At equilibrium: there is an additive price function P on the cake, and each agent gets to buy their best piece from a budget of one unit of fake currency

- WCE: $\forall i \in N, Z \subseteq [0,1]: P(Z) \leq P(A_i) \Rightarrow v_i(Z) \leq v_i(A_i)$

- EI: $\forall i \in N: P(A_i) = 1$

- Theorem [Weller, 1985]

For cake-cutting, a CEEI always exists. Every CEEI is both envy-free and weakly Pareto optimal.

s-CEEI

- **Strong Competitive Equilibrium from Equal Incomes (s-CEEI)**

- A positive slice Z is a subset of the cake valued positively by at least one agent
- Allocation A is called s-CEEI allocation if there exists an additive price function P satisfying

- $P(Z) > 0$ iff Z is a positive slice

- SCE: $\forall i \in N$, and positive slices $Z \subseteq [0,1]$ and $Z_i \subseteq A_i$: $\frac{v_i(Z_i)}{P(Z_i)} \geq \frac{v_i(Z)}{P(Z)}$

- EI: $\forall i \in N: P(A_i) = 1$

Maximum bang-per-buck

- **Theorem [Segal-Halevi and Sziklai, 2018]**

For cake-cutting, an s-CEEI allocation always exists. Every s-CEEI allocation is envy-free and Pareto optimal.

s-CEEI and Nash-Optimality

- An allocation A^* is called **Nash-optimal** if

$$A^* \in \arg \max_A \prod_{i \in N} v_i(A_i)$$

- **Theorem [Segal-Halevi and Sziklai, 2018]**

For cake-cutting, the set of s-CEEI allocations coincide with the set of Nash-optimal allocations.

Nash-Optimality Example



- Due to PO, suppose:
 - Agent 1 gets x fraction of $[0, 2/3]$
 - Agent 2 gets $1 - x$ fraction of $[0, 2/3]$ and all of $[2/3, 1]$
 - $v_1(A_1) = x$
 - $v_2(A_2) = (1 - x) \cdot 2/3 + 1/3 = (3 - 2x)/3$
- Maximize $x \cdot (3 - 2x)/3 \Rightarrow x = 3/4$
 - Nash-optimal allocation:
 - $A_1 = [0, 1/2]$, $v_1(A_1) = 3/4$
 - $A_2 = [1/2, 1]$, $v_2(A_2) = 1/2$

Strategyproofness

Strategyproofness

- Direct-revelation mechanisms

- A direct-revelation mechanism h takes as input all the valuation functions v_1, \dots, v_n , and returns an allocation A
- Notation: $h(v_1, \dots, v_n) = A, h_i(v_1, \dots, v_n) = A_i$

- Strategyproofness (deterministic mechanisms)

- A direct-revelation mechanism h is called strategyproof if

$$\forall v_1, \dots, v_n, \forall i, \forall v'_i : v_i(h_i(v_1, \dots, v_n)) \geq v_i(h_i(v_1, \dots, v'_i, \dots, v_n))$$

- That is, no agent i can achieve a higher value by misreporting her valuation, regardless of what the other agents report

Strategyproofness (SP)

- Strategyproofness (randomized mechanisms)

- Technically, referred to as “truthfulness-in-expectation”
 - When referring to SP for randomized mechanisms, we will refer to this concept

- A randomized direct-revelation mechanism h is called strategyproof if

$$\forall v_1, \dots, v_n, \forall i, \forall v'_i : E[v_i(h_i(v_1, \dots, v_n))] \geq E[v_i(h_i(v_1, \dots, v'_i, \dots, v_n))]$$

- That is, no agent i can achieve a higher *expected* value by misreporting her valuation, regardless of what the other agents report
 - Expectation is over the randomness of the mechanism

Deterministic SP Mechanisms

- **Theorem [Menon and Larson '17]**
No deterministic SP mechanism is (even approximately) **proportional**.
 - Since EF is at least as strict as Prop, SP+EF is also impossible.
- SP+PO is easy to achieve through serial dictatorship
 - SP+PO+EQ is impossible, because, as we saw, EQ+PO allocations may not exist

Open Problem

Does there exist a direct revelation, deterministic SP+EQ mechanism?

Randomized SP Mechanisms

- We want the mechanism *always* return an allocation satisfying a subset of {EQ,EF,PO}, and be SP in expected utilities
- Recall: PO+EQ allocations may not exist
 - Hence, we can only hope for SP+PO+EF or SP+EF+EQ
 - The first is an open problem, but the second combination is achievable!

Open Problem

Does there exist a randomized SP mechanism which always returns a PO+EF allocation?

Randomized SP Mechanisms

- **Theorem [Mossel and Tamuz, 2010; Chen et al. 2013]**

There is a randomized SP mechanism that *always* returns an EF+EQ allocation.

- Recall: In a perfect partition B , $v_i(B_k) = 1/n$ for all $i, k \in N$
- **Algorithm:** Compute a perfect partition and return allocation A which randomly assigns the n pieces to the n agents
- **SP:** Regardless of what the agents report, agent i receives each piece of the cake with probability $1/n$, and thus has expected value exactly $1/n$
- **EF:** Assuming agents report truthfully (due to SP), agent i always receives a cake she values at $1/n$, and according to her, so do others.

Existential Summary

✗ = Impossibility
✓ = Possibility

SP+PO+EF+EQ

✗ Rand

SP+PO+EF

✗ Det

? Rand

SP+PO+EQ

✗ Rand

SP+EF+EQ

✗ Det

✓ Rand

PO+EF+EQ

✗ Rand

SP+PO

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SP+EQ

? Det

✓ Rand

PO+EF

✓ Det

PO+EQ

✗ Rand

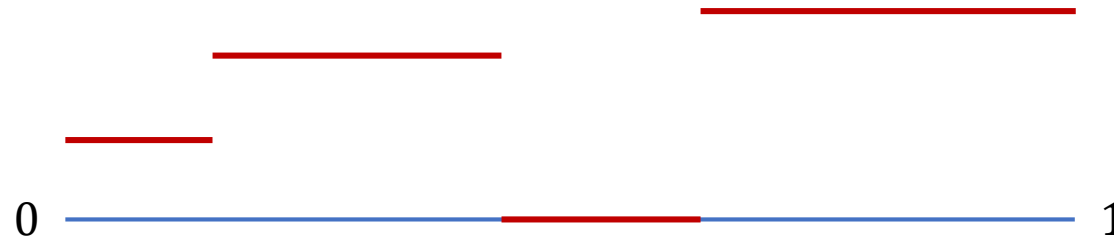
EF+EQ

✓ Det

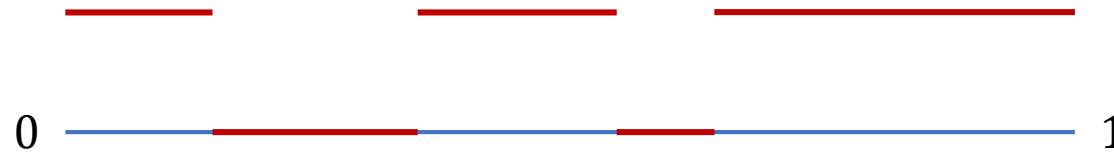
Special Cases

Piecewise Constant/Uniform Valuations

Piecewise constant
density function



Piecewise uniform
density function



Special case of piecewise constant

Possibilities

- Theorem [Chen et al., 2013]

For piecewise uniform valuations, there exists a deterministic SP mechanism which returns an EF+PO allocation.

➤ Recall that for general valuations, even deterministic SP+EF is impossible.

- Theorem [Aziz and Ye, 2014]

For piecewise constant valuations, an s-CEEI (i.e. Nash-optimal) allocation can be computed in polynomial time.

➤ Recall that this is EF (thus Prop) and PO.

➤ But this is not SP.

EF in Robertson-Webb

- Theorem [Kurokawa et al., 2013]

If an algorithm computes an envy-free allocation for n agents with piecewise uniform valuations with at most $g(n)$ queries, then it can also compute an envy-free allocation for n agents with general valuations with at most $g(n)$ queries.

- Let the same algorithm interact with general valuations v_1, \dots, v_n via CUT and EVAL queries and return an allocation A
- The proof constructs piecewise uniform valuations u_1, \dots, u_n which would have resulted in the same responses and $u_i(A_i) = v_i(A_i)$ for each agent $i \in N$

PO in Robertson-Webb

- **Non-wastefulness**

- An allocation A is called non-wasteful if no piece of the cake that is valued positively by at least one agent is assigned to an agent who has zero value for it
- PO implies non-wastefulness

- **Theorem [Ilanovski, 2012; Kurokawa et al., 2013]**

No finite protocol in the Robertson-Webb model can always produce a non-wasteful allocation, even for piecewise uniform valuations.

- This is the reason we did not provide query complexity results when discussing PO

Burnt Cake Division

Model

- Same as regular cake, except agents now have non-positive valuation for every piece of the cake
 - $f_i(x) \leq 0, \forall x \in [0,1]$
 - Hence, $v_i(X) \leq 0, \forall X \in \mathcal{A}$
- Equitability and perfect partitions carry over from the goods case
 - Simply use $-f_i$ and $-v_i$



Dividing a Burnt Cake

- Theorem [Peterson and Su, 2009]

For burnt cake division, there exists a finite (but unbounded) protocol for finding an envy-free allocation with n agents.

- Builds upon the Brams-Taylor protocol for dividing a good cake
- But certain operations require non-trivial transformations to the world of chores

Open Problem

Is there a bounded envy-free protocol for burnt cake division?

Allocating Divisible Goods + Bads

Model

- Agents: $N = \{1, 2, \dots, n\}$
- Resources: Set of divisible “items” $M = \{o_1, o_2, \dots, o_m\}$
- Allocation $A = (A_1, \dots, A_n)$
 - $A_i = (A_{i,j})_{j \in [m]}$
 - $\forall i, j: A_{i,j} \in [0, 1]$
 - $\forall j: \sum_i A_{i,j} \leq 1$
- Assume additive valuations: $v_i(A_i) = \sum_j A_{i,j} v_i(o_j)$
 - However, $v_i(o_j)$ can be positive, zero, or negative
- We’ll refer to s-CEEI simply as CEEI in this case

Achieving EF+PO

- **Theorem [Bogomolnaia et al. 2017]**
 - There always exists a CEEI allocation, which is envy-free and Pareto optimal.
 - The CEEI solution is “welfarist”, i.e., the set of feasible utility profiles is enough to identify the set of CEEI utility profiles.
 - The CEEI utility profile is given by the following:
 1. If it is possible to give a positive utility to each agent (who can receive a positive utility), then maximizing the Nash welfare gives the unique CEEI utility profile.
 2. Else, if the all-zero utility profile is feasible and Pareto optimal, then it is the unique CEEI utility profile.
 3. Else, there can be **exponentially many** CEEI utility profiles, which give non-positive utility to each agent.
 - Their actual result is stronger and in a more general model

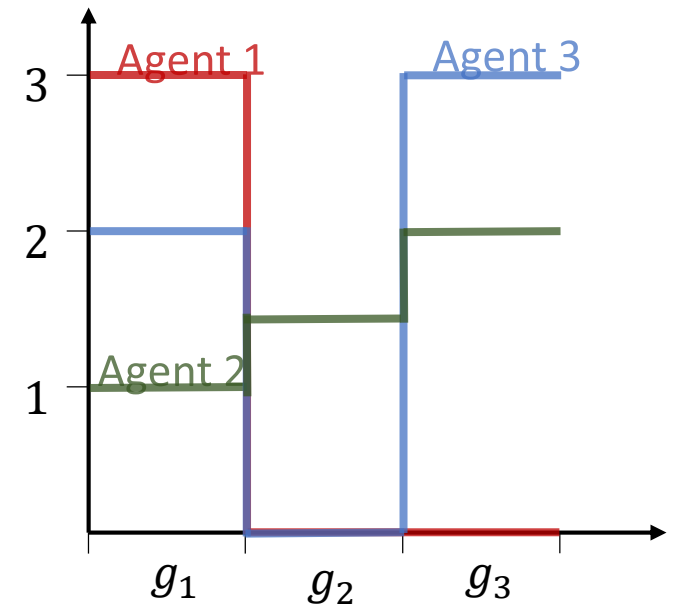
Not Covered

- Nash equilibria of cake-cutting
- Optimal cake-cutting
 - Algorithms for maximizing social welfare subject to fairness constraints
- Number of cuts and moving knives protocols
 - Possibility and impossibility results for $n - 1$ cuts
- Multidimensional cakes
- Randomized or strategyproof Robertson-Webb protocols
- Non-additive valuations
- ...

Divisible Goods

Model

- Agents: $N = \{1, 2, \dots, n\}$
- Resource: Set of divisible goods $M = \{g_1, g_2, \dots, g_m\}$
- Allocation $A = (A_1, \dots, A_n)$
 - $A_i = (A_{i,j})_{j \in [m]}$
 - $\forall i, j: A_{i,j} \in [0, 1]$
 - $\forall j: \sum_i A_{i,j} \leq 1$
- Assume additive valuations $v_i(A_i) = \sum_j A_{i,j} v_i(g_j)$
- Special case of cake cutting (up to normalization)



$n = 2$: Adjusted Winner Procedure

[Brams and Taylor 1996]

- Input: **Normalized** valuation functions
- Order the goods by ratio $v_1(g)/v_2(g)$.

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	20	30	15 30	10	5	5
a_2	10	15	20 10	15	10	30

$v_1(g)/v_2(g)$ high \longleftrightarrow $v_1(g)/v_2(g)$ low

- Divide the goods so that agent 1 receives goods g_1, \dots, g_{j-1} , agent 2 receives goods g_{j+1}, \dots, g_m for some j , and $v_1(A_1) = v_2(A_2)$
 - g_j is divided between the agents, if necessary

$n = 2$: Adjusted Winner Procedure

[Brams and Taylor 1996]

- **Theorem [Brams and Taylor 1996]:**

- The adjusted winner procedure is envy-free (and therefore proportional), equitable and Pareto optimal

- Breaks down for $n > 2$

- As in cake cutting, EF + EQ + PO is impossible, what about two of the three?
- EF+EQ: Divide each good equally among agents (“perfect partition”)
- EQ + PO: Impossible
- EF + PO: Can achieve with CEEI

	g_1	g_2
a_1	1	0
a_2	1	0
a_3	0	1

CEEI

- With a fixed set of items, the definition of s-CEEI (that we will now call just CEEI) becomes simpler.

- Equilibrium price $p_j > 0$ for each good g_j
 - Assume for simplicity that $\forall j \exists i$ with $v_i(g_j) > 0$
- CE: If $A_{i,j} > 0$ then $\frac{v_i(g_j)}{p_j} \geq \frac{v_i(g_k)}{p_k}$ for all k
- EI: $\sum_j p_j A_{i,j} = 1$ for all i

Example

- CEEI allocation:
 - $A_1 = (1, 0.75, 0)$
 - $A_2 = (0, 0.25, 1)$
- Prices $(p_1, p_2, p_3) = (0.4, 0.8, 0.8)$
- Check CE condition:
 - $\frac{v_1(g_1)}{p_1} = \frac{v_1(g_2)}{p_2} = 25 > 15 = \frac{v_1(g_3)}{p_3}$
 - $\frac{p_1}{v_2(g_2)} = \frac{p_2}{v_2(g_3)} = 20 > 12.5 = \frac{p_3}{v_2(g_1)}$
- Check EI condition:
 - $\sum_{j \in [m]} p_j A_{1,j} = 0.4 + 0.6 + 0 = 1$
 - $\sum_{j \in [m]} p_j A_{2,j} = 0 + 0.2 + 0.8 = 1$

	g_1	g_2	g_3
a_1	10	20	12
a_2	5	16	16

Note that $v_1(A_1) = 25$ and $v_2(A_2) = 20$. No other allocation yields a higher product.

Eisenberg-Gale convex program

- Can compute a CEEI allocation as the solution to the Eisenberg-Gale [1959] convex program:

$$\begin{aligned} & \max \sum_{i \in N} \log u_i \quad s.t. \\ & \forall i: u_i \leq \sum_{g_j \in M} A_{i,j} v_i(g_j) \\ & \forall j: \sum_{i \in N} A_{i,j} \leq 1 \\ & \forall i, j: A_{i,j} \geq 0 \end{aligned}$$

- **Theorem [Orlin 2010, Végh 2012]:**
 - The Eisenberg-Gale convex program can be solved in strongly polynomial time.

Strategyproofness

- CEEI solution is fair and efficient but not strategyproof.
 - It is **strategyproof in the large (SP-L)** [Azevedo and Budish 2018] though
- **Theorem [Han et al. 2011]:**
 - No strategyproof mechanism that always outputs a complete allocation can achieve better than a $1/m$ approximation to the optimal **social welfare** for large enough n .
 - Social welfare = $\sum_{i \in N} v_i(A_i)$
- **Theorem [Cole et al. 2013]:**
 - There is a strategyproof mechanism that provides every agent with a $1/e$ fraction of their CEEI utility.
 - Allocation is envy-free but not proportional
 - Does not allocate all resources, so envy-freeness does not imply proportionality

SP + Prop + EF

- SP + Prop + EF is trivial! Just allocate everyone an equal fraction of each good.
 - What if we also want PO?
- **Theorem [Schummer 1996]:**
 - It is impossible to achieve SP + Prop + PO.
 - SP + PO: Serial dictatorship.
- SP + Prop + EF can also be achieved non-trivially [Freeman et al. 2019]
 - Additionally achieves strict SP: agents always achieve strictly higher utility by reporting their beliefs truthfully than by lying.
 - Exploits a correspondence between fair division and wagering mechanisms [Lambert et al. 2008] to utilize proper scoring rules (e.g. Brier score)

Indivisible Goods

Model

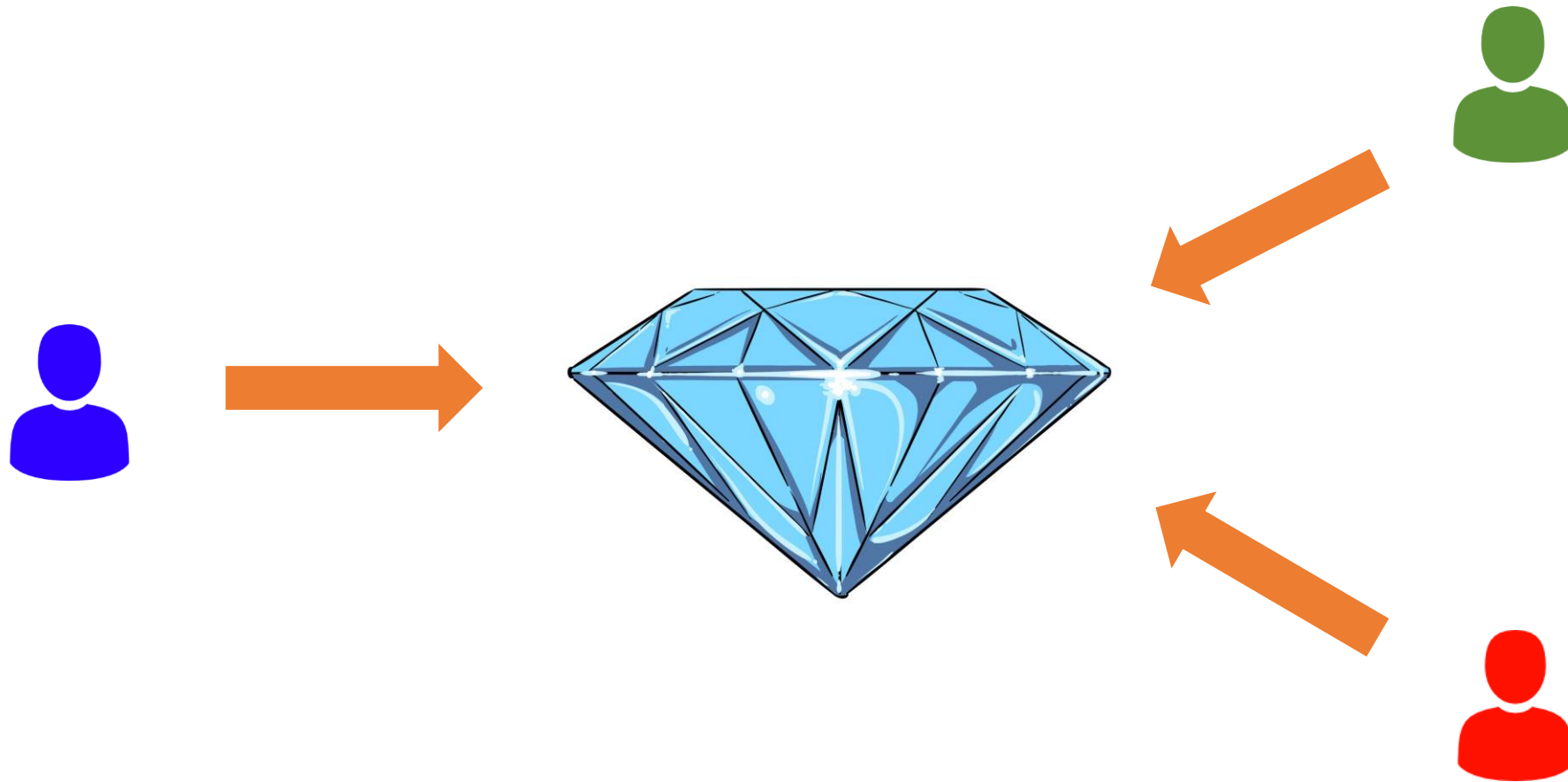
- Agents: $N = \{1, 2, \dots, n\}$
- Resource: Set of **indivisible goods** $M = \{g_1, g_2, \dots, g_m\}$
- Allocation $A = (A_1, \dots, A_n) \in \Pi_n(M')$ is a partition of M' for some $M' \subseteq M$.
- Each agent i has a valuation $v_i : 2^M \rightarrow \mathbb{R}_+$

Valuation Functions

- **Additive:** $\forall X, Y$ with $X \cap Y = \emptyset$: $v_i(X \cup Y) = v_i(X) + v_i(Y)$
 - Equivalently: $v_i(X) = \sum_{g \in X} v_i(g)$
 - Value for a good independent of other goods received
- **Submodular:** $\forall X, Y : v_i(X \cup Y) + v_i(X \cap Y) \leq v_i(X) + v_i(Y)$
 - Equivalently: $\forall X, Y$ with $X \subseteq Y$: $v_i(X \cup \{g\}) - v_i(X) \geq v_i(Y \cup \{g\}) - v_i(Y)$
- **Subadditive:** $\forall X, Y$ with $X \cap Y = \emptyset$: $v_i(X \cup Y) \leq v_i(X) + v_i(Y)$
- Submodular and subadditive definitions capture the idea of diminishing returns.

Most results for additive valuations unless stated otherwise

Need new guarantees!



Envy-Freeness up to One Good

Envy-Freeness up to One Good (EF1)

[Lipton et al 2004, Budish 2011]

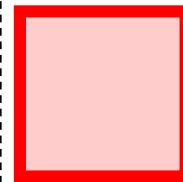
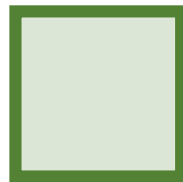
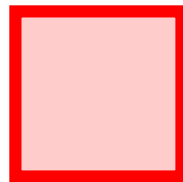
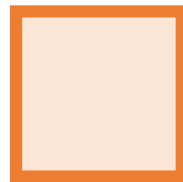
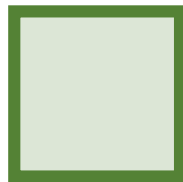
- An allocation is **envy-free up to one good (EF1)** if, for all agents i, j , there exists a good $g \in A_j$ for which

$$v_i(A_i) \geq v_i(A_j \setminus \{g\})$$

- “Agent i may envy agent j , but the envy can be eliminated by removing a single good from j ’s bundle.”
 - Note: We don’t consider $A_j = \emptyset$ a violation of EF1.

Round Robin Algorithm

- Fix an ordering of the agents σ .
- In round $k \bmod n$, agent σ_k selects their most preferred remaining good.
- **Theorem:** Round robin satisfies EF1.



Phase 1

Phase 2

Animation Credit: Ariel Procaccia

Algorithm for Achieving EF1

- Greedy algorithm [Lipton et al. 2004]
 - One at a time, allocate a good to an agent that no one envies
 - While there is an envy cycle, rotate the bundles along the cycle.
 - Can prove this loop terminates in a polynomial number of steps
- Removing the most recently added good from an agent's bundle removes envy towards them.
- Neither this algorithm nor round robin is Pareto optimal.

Maximum Nash Welfare

- **Maximum Nash Welfare (MNW):** Select the allocation that maximizes the geometric mean of agent utilities (more on this later).

$$A = \arg \max \left(\prod_i v_i(A_i) \right)^{1/n}$$

- This is just Nash-optimality from earlier
- What if $\prod_i v_i(A_i) = 0$ for all allocations?
 - Find an allocation that maximizes $|\{v_i(A_i) > 0\}|$, and subject to that maximizes

$$\left(\prod_{i:v_i(A_i)>0} v_i(A_i) \right)^{1/n}$$

EF1 + PO

- **Theorem [Caragiannis et al. 2016]:**
 - The MNW allocation satisfies EF1 and PO.
 - PO: A Pareto-improving allocation would have higher geometric mean of utilities for agents with non-zero utility or more agents with non-zero utility.
 - EF1: Let $g_i^* = \arg \max_{g \in A_i} v_i(g)$. Not-too-hard proof shows $v_j(A_j) \geq v_j(A_i \setminus g_i^*)$ for all j .

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	2	1	3	0	1	2
a_2	10	1	1	1	2	5
a_3	3	1	3	0	5	2

Computing EF1 + PO

- The MNW allocation is strongly NP-hard to compute (reduction from X3C).
 - Actually, it's APX-hard [Lee 2017].
- Special case: Binary valuations
 - MNW allocation can be computed in polynomial time [Darmann and Schauer 2015, Barman et al. 2018].
 - However, round robin already guarantees EF1 + PO in this setting.

Computing EF1 + PO

- Theorem [Barman et al. 2018]:

- There exists a pseudo-polynomial time algorithm for computing an allocation satisfying EF1 + PO
- Algorithm uses local search (sequence of item swaps and price rises) to compute an integral competitive equilibrium that is **price envy-free up to one good**.
- Price envy-free up to one good: $\forall i, k, \exists j: p(A_i) \geq p(A_k \setminus \{g_j\})$
- Need different entitlements because CEEI might not exist with indivisibilities
 - Two agents, one item...

Computing EF1 + PO

Open Problem:
Complexity of computing an EF1 + PO allocation

Open Problem:
Does there always exist an EF1 + PO allocation for
submodular valuation functions?

Proportionality up to One Good

Proportionality up to One Good (Prop1)

[Conitzer et al. 2017]

- An allocation is **proportional up to one good (Prop1)** if, for every agent i , there exists a good g for which

$$v_i(A_i \cup \{g\}) \geq \frac{v_i(M)}{n}$$

	g_1	g_2	g_3
a_1	1	3	3
a_2	1	3	3

$$v_1(A_1 \cup \{g_2\}) = 4 \geq \frac{7}{2} = \frac{v_i(M)}{n}$$

Prop1 + PO

- Any algorithm that satisfies EF1 + PO is also Prop1 + PO.
 - MNW
 - Barman et al. [2018] algorithm
- Theorem [Barman and Krishnamurthy 2019]:
 - An allocation satisfying Prop1 + PO can be computed in strongly polynomial time.
- Allocation is a careful rounding of the fractional CEEI allocation.
 - In contrast, there exist instances in which no rounding of the fractional CEEI allocation will give EF1 [Caragiannis et al., 2016].

Envy-Freeness up to the Least Valued Good

Envy-Freeness up to the Least Valued Good

[Caragiannis et al. 2016]

	g_1	g_2	g_3
a_1	10	5	5
a_2	10	ϵ	ϵ

- An allocation is **envy-free up to the least valued good (EFX)** if, for all agents i, j , and every $g \in A_j$ with $v_i(g) > 0$,

$$v_i(A_i) \geq v_i(A_j \setminus \{g\}).$$

Leximin Allocation

- Leximin allocation:
 - First, maximize the minimum utility any agent receives. Subject to this, maximize the second-minimum utility. Then the third-minimum utility, etc.

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	2	1	3	0	1	2
a_2	10	1	1	1	2	5
a_3	3	1	3	0	5	2

Satisfying EFX

	g_1	g_2	g_3	g_4
a_1	4	1	2	2
a_2	4	1	2	2
a_3	4	1	2	2

- Theorem [Plaut and Roughgarden, 2018]:
 - The Leximin allocation satisfies EFX + PO for agents with (general) identical valuations.
- Theorem [Plaut and Roughgarden, 2018]:
 - The Leximin allocation satisfies EFX + PO for two agents with (normalized) additive valuations.

Open Problem:
Does there always exist a complete allocation satisfying EFX?

Satisfying EFX

- What about partial allocations satisfying EFX?
 - Easy! We can just throw all goods away and take the empty allocation.
- Theorem [Caragiannis et al. 2019]:
 - There exists a partial allocation that satisfies EFX and achieves a 2-approximation to the optimal Nash welfare.
 - No (complete or partial) EFX allocation can achieve a better approximation.

	Existence		Computation	
	Without PO	With PO	Without PO	With PO
Envy-Freeness	No	No	NP-hard	NP-hard
EFX	Open	Open	Open	Open
EF1	Yes	Yes	Polytime	Open
Prop1	Yes	Yes	Polytime	Polytime

Maximin Share

Maximin Share [Budish 2011]

- “If I partition the goods into n bundles and receive an adversarially chosen bundle, how much utility can I guarantee myself?”
- Define $MMS_i^k(S) = \max_{(P_1, \dots, P_k) \in \Pi_k(S)} \min_{1 \leq j \leq k} v_i(P_j)$
- MMS allocation: One for which $v_i(A_i) \geq MMS_i^n(M)$
- Note that $MMS_i^n(M) \leq \frac{v_i(M)}{n}$, so Proportionality implies MMS

Maximin Share [Budish 2011]

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	2	1	3	0	1	2
a_2	10	1	1	1	2	5
a_3	3	1	3	0	5	2

$$MMS_1^n(M) = \min(3, 3, 3) = 3$$

$$MMS_2^n(M) = \min(10, 5, 5) = 5$$

$$MMS_3^n(M) = \min(4, 5, 5) = 4$$

Achieving Maximin Allocations

- **Theorem [Procaccia and Wang 2014]:**
 - There exist instances for which no allocation satisfies MMS.
- Instead, consider approximations.
 - c-MMS: allocation for which $v_i(A_i) \geq c \cdot MMS_i^n(M)$
 - Guarantee $v_i(A_i) \geq MMS_i^k(M)$ for some $k > n$
- **Theorem [Budish 2011]:**
 - There always exists an allocation that satisfies $v_i(A_i) \geq MMS_i^{(n+1)}(M)$ for every agent i .

c-MMS Allocations

- Theorem [Procaccia and Wang 2014]:
 - A $(2/3)$ -MMS allocation always exists.
- Theorem [Amanatidis et al. 2017]:
 - A $(2/3-\epsilon)$ -MMS allocation can be computed in polynomial time.
- Theorem [Ghodsi et al. 2018]:
 - A $(3/4)$ -MMS allocation always exists and a $(3/4-\epsilon)$ -MMS allocation can be computed in polynomial time.
- Theorem [Garg and Taki, Manuscript]:
 - A $(3/4)$ -MMS allocation can be computed in polynomial time.

c-MMS Allocations

[Ghodsi et al. 2018]

	Additive	Submodular	Subadditive
Lower bound (existence)	$\frac{3}{4}$	$\frac{1}{3}$	$\frac{1}{10} \lceil \log m \rceil$
Lower bound (polynomial algorithm)	$\frac{3}{4}$	$\frac{1}{3}$	-
Upper bound	$1 - \frac{1}{n^{n+1}}$	$\frac{3}{4}$	$\frac{1}{2}$

Open Problem:
Close the gaps!

Groupwise MMS [Barman et al. 2018]

- Idea: MMS_i^k should be guaranteed for all groups J of agents of size k and set of goods $\cup_{i \in J} A_i$

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	5	5	$5 + \epsilon$	$5 - \epsilon$	$5 + \epsilon$	$5 - \epsilon$
a_2	5	5	$5 + \epsilon$	$5 - \epsilon$	$5 + \epsilon$	$5 - \epsilon$
a_3	10	10	0	0	ϵ	ϵ

- $v_3(A_3) \geq MMS_3^3(M)$ but $v_3(A_3) < MMS_3^2(A_1 \cup A_3)$

Groupwise MMS [Barman et al. 2018]

- Allocation A satisfies Groupwise Maximin Share (GMMS) if,

$$\forall i: v_i(A_i) \geq \max_{J \subseteq N} MMS_i^{|J|}(\cup_{j \in J} A_j)$$

- **Theorem [Barman et al. 2018]:**
 - When valuations are additive, a 0.5-GMMS allocation exists and can be found in polynomial time.
 - Algorithm: Select an agent who is not envied by any other agent, and allocate her her most preferred unallocated good.
 - Small refinement of EF1 algorithm from earlier

(Relaxed) Equitability

Equitability

- Recall equitability:

$$\forall i, j \in N: v_i(A_i) \geq v_j(A_j)$$

- We can relax it in the same way we did for envy-freeness [Gourves et al. 2014, Freeman et al. 2019].

- **Equitability up to one good (EQ1):**

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

- **Equitability up to any good (EQX):**

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

Algorithm for Achieving EQX

- Greedy Algorithm [Gourves et al. 2014]:
 - Allocate to the lowest-utility agent the unallocated good that she values the most.
- Almost the same as EF1 algorithm, but achieves EQX!
 - Compare to EFX, existence still unknown

EQ1/EQX + PO

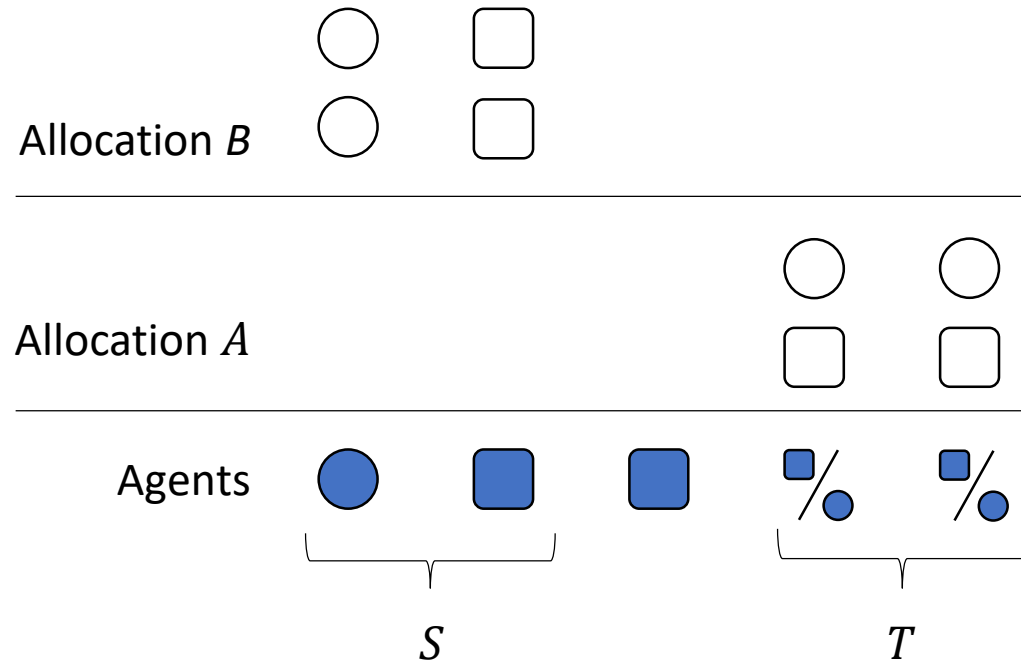
- Theorem [Freeman et al. 2019]:
 - An allocation satisfying EQ1 and PO may not exist.
 - Compare to EF1 + PO always exists

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	1	1	1	0	0	0
a_2	0	0	0	1	1	1
a_3	0	0	0	1	1	1

- Theorem [Freeman et al. 2019]:
 - When valuations are strictly positive, the Leximin allocation is EQX + PO

Group Fairness

Beyond Individual Fairness



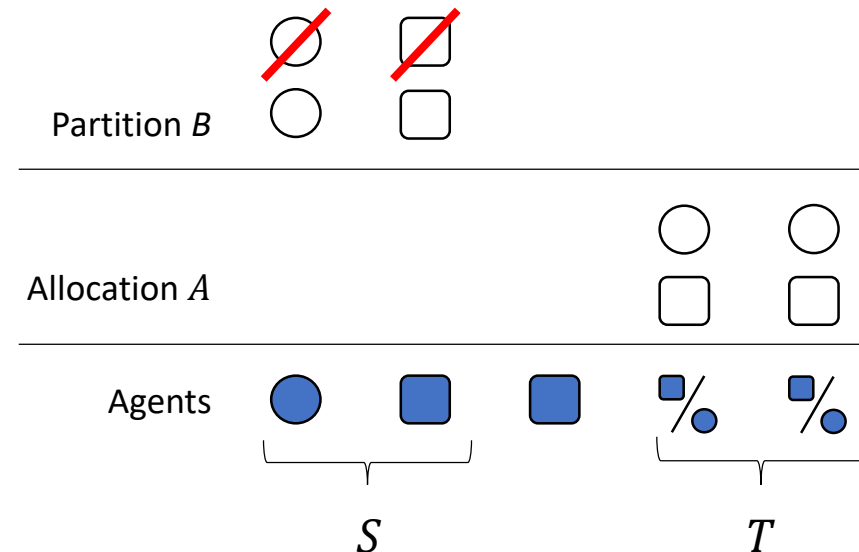
Envy-Free up to One Good (EF1)

Group Fairness

- An allocation A is **group fair** if for every non-empty $S, T \subseteq N$ and every partition $(B_i)_{i \in S}$ of $\cup_{j \in T} A_j$, $\left(\frac{|S|}{|T|}\right) \cdot (v_i(B_i))_{i \in S}$ does not Pareto dominate $(v_i(A_i))_{i \in S}$
- “It should not be possible to **redistribute the goods allocated to group T amongst group S** in such a way that every member of group S is (weakly, with at least one strictly) **better off**, for group sizes”
- Group Fairness \Rightarrow EF1 + PO

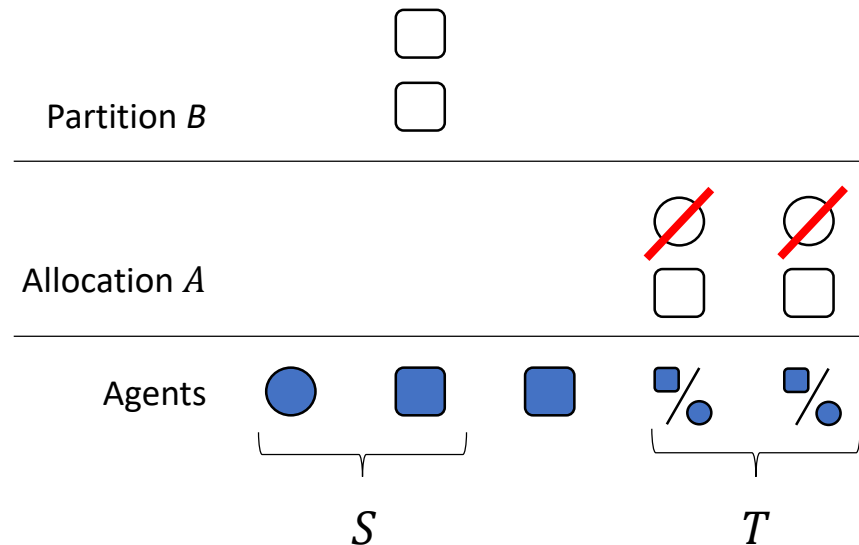
Group Fairness Relaxations

- **Group Fairness up to One Good, After (GF1A) [Conitzer et al. 2019]**
 - “It should not be possible to redistribute the goods allocated to group T amongst group S in such a way that every member of group S is (weakly, with at least one strictly) better off, **even when one good is removed from each agent in S**, with utilities adjusted for group sizes”



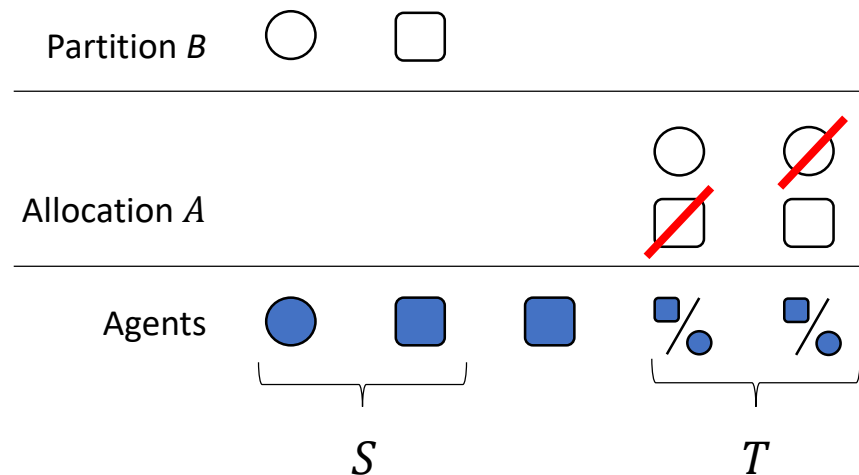
Group Fairness Relaxations

- **Group Fairness up to One Good, Before (GF1B) [Conitzer et al. 2019]**
 - “It should not be possible to redistribute the goods allocated to group T, **with one good per agent in T removed**, amongst group S in such a way that every member of group S is (weakly, with at least one strictly) better off, with utilities adjusted for group sizes”



Group Fairness Relaxations

- **Group Fairness up to One Good, Before (GF1B) [Conitzer et al. 2019]**
 - “It should not be possible to redistribute the goods allocated to group T, **with one good per agent in T removed**, amongst group S in such a way that every member of group S is (weakly, with at least one strictly) better off, with utilities adjusted for group sizes”



Achieving GF1A/GF1B

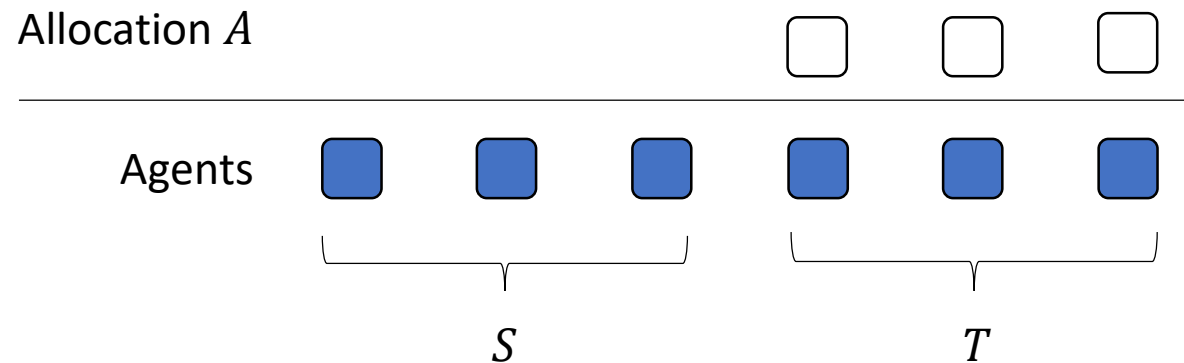
- **Locally Nash-optimal allocation:** Product of utilities cannot be improved by moving a single good.
 $\forall i, j, g \in A_j: v_j(g) > 0 \text{ and } v_i(A_i) \cdot v_j(A_j) \geq v_i(A_i + g) \cdot v_j(A_j - g)$
- **Theorem [Conitzer et al. 2019]:**
 - Any locally Nash-optimal allocation satisfies GF1A and GF1B.
 - Can be computed in pseudo-polynomial time by local search
 - When valuations are identical, an allocation is locally Nash-optimal iff it is EFX/EQX.

Open Problem:

Can we compute a locally Nash-optimal allocation in polynomial time?

Known Groups

- When we want to provide guarantees for **all** subsets of agents, “up to one good per agent” guarantees are the best we can give.



- What if we know S and T in advance?

Known Groups

- Let \mathcal{S} be a partition of N . Say that N is **Fixed-Group Fair up to One good, Before (FGF1B)**, if, for all $S, T \in \mathcal{S}$, there exists $g \in \cup_{i \in T} A_i$ such that for all partitions $(B_i)_{i \in S}$ of $\cup_{i \in T} A_i \setminus \{g\}$, $(B_i)_{i \in S}$ does not Pareto dominate $(A_i)_{i \in S}$.
- Can modify the definition to get FGF1A and up to any good (FGFXA/B) variants.

Open Problem:

Does there always exist an allocation satisfying FGF1A/B?

Nash Welfare Approximation

Nash Welfare Approximation

- We have seen that MNW satisfies several nice properties.
 - GF1A/B (\Rightarrow EF1) + PO
 - Scale-free
 - Natural fairness/efficiency tradeoff
- But NP-hard to optimize. Can we approximate?
- **Theorem [Lee 2017]**
 - Computing an allocation that maximizes the geometric mean of agent utilities under additive valuation functions is APX-hard.
 - Approximating to within a factor of 1.00008 is NP-hard.

Nash Welfare Approximation

- Theorem [Cole and Gkatzelis 2015, Cole et al 2017]:
 - There exists a polynomial time algorithm that approximates the MNW objective to within a constant multiplicative factor of 2.
- Theorem [Barman et al. 2018]:
 - There exists a polynomial time algorithm that approximates the MNW objective to within a constant multiplicative factor of 1.45.

Open Problem:

Close the gap between the 1.00008 lower bound and 1.45 upper bound.

Nash Welfare Approximation

- Approximate MNW solutions may not retain the nice properties of the exact solution.
- **Theorem [Garg and McGlaughlin 2019]:**
 - There exists a polynomial time algorithm that approximates the MNW objective to within a constant multiplicative factor of 2 and achieves Prop1, $(1/2n)$ -MMS and PO.
- And recall, there exists a partial allocation that satisfies EFX and is a 2-approximation to MNW objective [Caragiannis et al 2019].

Price of Fairness

Price of Fairness

- What effect does requiring a fairness property have on the social welfare?
- **Price of Fairness [Bertsimas et al. 2011, Caragiannis et al. 2012]:**
 - The price of fairness of fairness property P is defined as the ratio of the maximum possible social welfare and the maximum social welfare of an allocation that satisfies P .
- **Strong Price of Fairness [Bei et al. 2019]:**
 - The strong price of fairness of fairness property P is defined as the ratio of the maximum possible social welfare and the minimum social welfare of an allocation that satisfies P .
- Cf. Price of Stability and Price of Anarchy

Price of Fairness

- **Theorem [Caragiannis et al. 2012]:**
 - The price of fairness for proportionality, envy-freeness and equitability for are:

	Indivisible Goods	Cake Cutting
Proportionality	$\Theta(n)$	$\Theta(\sqrt{n})$
Envy-freeness	$\Theta(n)$	$\Omega(\sqrt{n}), O(n)$
Equitability	∞	$\Theta(n)$

- Caragiannis et al. also studied divisible items, and bads.

Price of Fairness

- Theorem [Bei et al. 2019]:

- The following bounds on the (strong) price of fairness apply for indivisible goods

	Price of P	Strong Price of P
EF1	LB: $\Omega(\sqrt{n})$ UB: $O(n)$	∞
Round Robin	n	n^2
Max Nash Welfare	$\Theta(n)$	$\Theta(n)$
Leximin	$\Theta(n)$	$\Theta(n)$
Pareto optimality	1	$\Theta(n^2)$

Strategyproofness

Adding Strategyproofness

- None of the rules we have considered so far are strategyproof
- For divisible goods, structure of strategyproof mechanisms is fairly rich
 - Impossibilities from the divisible realm carry over
- What about indivisible goods?

	g_1	g_2	g_3
a_1	1	x	0
a_2	0	y	1

Picking-Exchange Mechanisms

[Amanatidis et al. 2017]

- **Picking Mechanism:**

- Partition $M = N_1 \cup N_2$
- Agent 1 receives a subset of offers $O_1 \subseteq 2^{N_1}$. Let $S_1 = \arg \max_{S \in O_1} v_1(S)$.
- Agent 2 receives a subset of offers $O_2 \subseteq 2^{N_2}$. Let $S_2 = \arg \max_{S \in O_2} v_2(S)$.
- $A_1 = S_1 \cup (N_2 \setminus S_2)$ and $A_2 = S_2 \cup (N_1 \setminus S_1)$

- $N_1 = \{g_1, g_2, g_3, g_4\}, N_2 = \{g_5, g_6\}$

- $O_1 = \{\{g_1, g_2\}, \{g_2, g_3\}, \{g_4\}\}, O_2 = \{\{g_5\}, \{g_6\}\}$

	g_1	g_2	g_3	g_4	g_5	g_6
a_1	3	5	5	10	4	2
a_2	2	3	6	1	5	3

Picking-Exchange Mechanisms

[Amanatidis et al. 2017]

- **Exchange Mechanism:**

- Partition $M = E_1 \cup E_2$
 - Set of exchange deals $D = \{(S_1, T_1), \dots, (S_k, T_k)\}$, where each $(S, T) \subseteq (E_1, E_2)$
 - Agent i receives allocation E_i by default, with exchanges performed if they are mutually beneficial
- $E_1 = \{g_1, g_2, g_3\}, E_2 = \{g_4, g_5\}$
 - $D = \{(\{g_2, g_3\}, \{g_4\})\}$

	g_1	g_2	g_3	g_4	g_5
a_1	6	2	3	7	1
a_2	1	6	1	4	7

Picking-Exchange Mechanisms

[Amanatidis et al. 2017]

- **Picking-Exchange Mechanism:** Run a picking mechanism on $N_1 \cup N_2 \subseteq M$ and an exchange mechanism on $E_1 \cup E_2 \subseteq M$, where $N_1 \cup N_2 \cup E_1 \cup E_2 = M$ and N_1, N_2, E_1, E_2 are pairwise disjoint.
 - Up to tiebreaking technicalities...

Picking-Exchange Mechanisms

[Amanatidis et al. 2017]

- **Theorem [Amanatidis et al. 2017]:**
 - For $n = 2$ an allocation mechanism that allocates all goods is strategyproof if and only if it is a picking-exchange mechanism
- **Corollary [Amanatidis et al. 2017]:**
 - For $n = 2$, any strategyproof mechanism that allocates all goods does not achieve any positive approximation of the minimum envy or best proportionality guarantee.
 - For $n = 2$ and $m \geq 5$, no strategyproof mechanism can allocate all items and satisfy EF1.
 - For $n = 2$, no strategyproof mechanism guarantees better than $\frac{1}{\lfloor m/2 \rfloor}$ -MMS.
 - This is a tight bound [Amanatidis et al. 2016]

More General Strategyproof Mechanisms

Open Problem:

What is the structure of strategyproof mechanisms for $n = 2$ when not all goods have to be allocated?

Open Problem:

What is the structure of strategyproof mechanisms for $n > 2$?

Allocating Bads

Allocating Bads/Chores

- Bads: $v_i : 2^M \rightarrow \mathbb{R}_-$
- Techniques/analyses for goods often do not carry over.
- **Theorem [Aziz et al. 2017]:**
 - A 2-MMS allocation always exists and can be computed in polynomial time when dividing bads.
- **Theorem [Barman and Krishnamurthy 2017]:**
 - A $(4/3)$ -MMS allocation always exists and can be computed in polynomial time when dividing bads.

Allocating Goods and Bads

- Mixed setting: $v_i : 2^M \rightarrow \mathbb{R}$, assume additive valuations
 - Items can be goods for some agents and bads for others
 - More general than only goods or only bads
- An allocation is envy-free up to one item (EF1) if, for all agents i, j , there exists an item $g \in A_j \cup A_i$ for which
$$v_i(A_i \setminus \{g\}) \geq v_i(A_j \setminus \{g\})$$
- Other definitions (like Prop1) also need to be generalized appropriately

Allocating Goods and Bads

- Theorem [Aziz et al. 2019]:

- When items can be either goods or bads, an EF1 allocation always exists and can be computed in polynomial time.
- Double round robin algorithm: first allocates (unanimous) bads via round robin, then allocates remaining items via round robin in the reverse order.

- Theorem [Aziz et al. 2019]:

- When items can be either goods or bads and $n = 2$, an EF1 + PO allocation always exists and can be found in polynomial time

Open Problem:

Does an EF1 + PO allocation always exists for bads? Prop1 + PO? EFX?

What's Not Covered

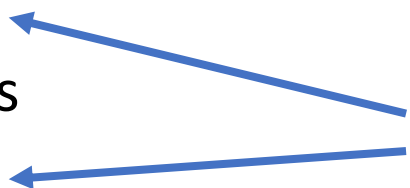
- Envy-freeness up to one less-preferred item (EFL) [Barman et al. 2018]
 - Stronger than EF1 and guaranteed to exist
 - Existence of EFL + PO allocations is an open question
- Various constraints and additional features
 - Agent social network structure
 - Connectivity constraints when items lie on a graph
- Asymptotic results
- ...

Leontief Preferences

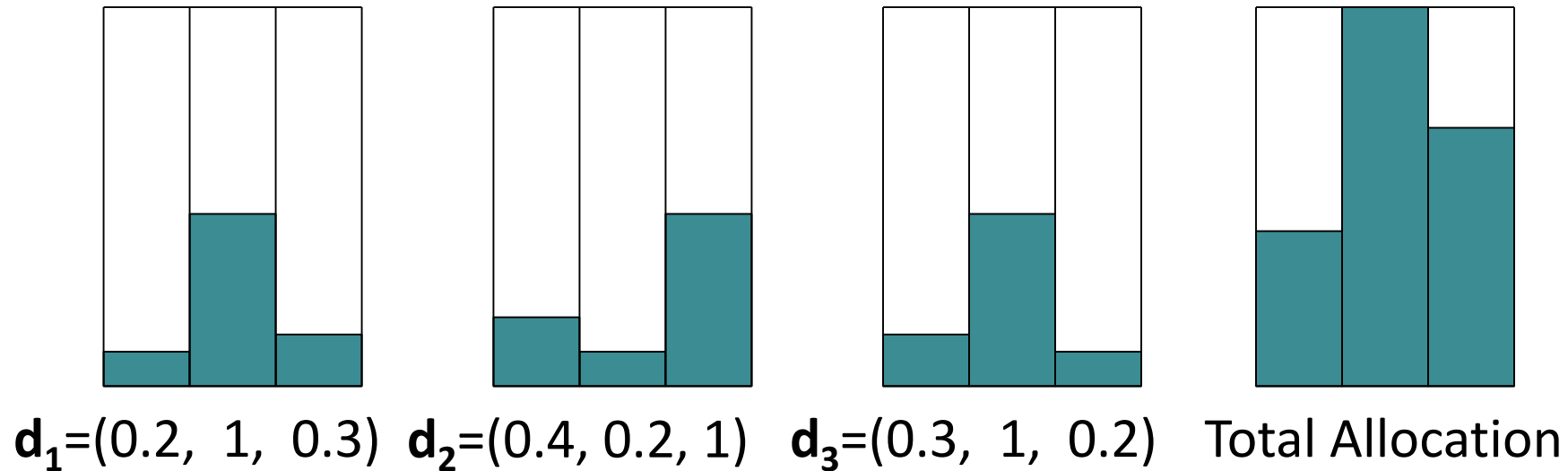
Leontief Preferences

- Set of resources M
- Agents require resources in a certain ratio $\mathbf{d}_i = (d_{i,1}, \dots, d_{i,m})$
- Fractional allocation \mathbf{A} where $A_{i,j}$ is the fraction of resource j that agent i receives.
- Leontief preferences: $v_i(A_i) = \max\{y \in \mathbb{R}_+ : \forall j \in M, A_{i,j} \geq y \cdot d_{i,j}\}$
 - Number of “complete sets” of resources received by i
- E.g. $M = \{\text{CPU}, \text{RAM}\}$, an agent with tasks that require 2 units of CPU for every unit of RAM has $\mathbf{d}_i = (2, 1)$.

Dominant Resource Fairness

- Dominant resource for agent $i = \arg \max_{j \in M} \{d_{i,j}\}$
 - Dominant Resource Fairness (DRF) [Ghodsi et al. 2011, Parkes et al. 2012]:
 - Equalize the allocation each agent receives of their dominant resource
 - Dominant resource fairness satisfies:
 - Proportionality
 - Strategyproofness
 - Envy-freeness
 - Pareto optimality
- Preferences no longer additive, so envy-freeness does not imply proportionality
- 

Dominant Resource Fairness



Ordinal Preferences

Ordinal Preferences

- Instead of valuation functions, take in preference orderings \succsim_i over items
 - E.g. $g_2 \succsim_i g_3 \succsim_i g_1 \succsim_i g_4$
- Agents are assigned fractions of each item
 - $A = (A_{i,j})_{i \in [n], j \in [m]}$
 - Can be interpreted as lotteries over integral allocations

Ordinal Preferences

- Partial preferences over bundles defined via **stochastic dominance** extension

$$A \succsim_i^{SD} B \quad \text{iff} \quad \forall k: \sum_{j \succsim_i k} A_{i,j} \geq \sum_{j \succsim_i k} B_{i,j}$$

- Many other extensions possible
 - Upper/downward lexicographic [Cho 2012]
 - Pairwise comparison [Aziz et al. 2014]
 - Bilinear dominance [Aziz et al. 2014]
- Can also elicit ordinal information over subsets directly [Bouveret et al. 2010]

Two Mechanisms

- **Random Priority**

- Select a random ordering of the agents. Agents select their favorite m/n goods in order.

- **Probabilistic Serial [Bogomolnaia and Moulin 2001]**

- Agents “eat” at a constant (equal) rate. At any time, agents eat their most preferred good that is not completely consumed.

- $\succsim_1: g_1 \succsim_1 g_2 \succsim_1 g_3 \succsim_1 g_4$

Random Priority

	g_1	g_2	g_3	g_4
a_1	1	1/2	0	1/2
a_2	0	1/2	1	1/2

$$\succsim_2: g_2 \succsim_2 g_3 \succsim_2 g_1 \succsim_2 g_4$$

Probabilistic Serial

	g_1	g_2	g_3	g_4
a_1	1	0	1/2	1/2
a_2	0	1	1/2	1/2

SD-efficiency

- SD-efficiency: There should not exist an alternative allocation that all agents weakly prefer and some agent strictly prefers.
- **Theorem [Bogomolnaia and Moulin 2001]:**
 - Probabilistic Serial satisfies SD-efficiency
- Random Priority is not SD-efficient
 - $\succsim_1: g_1 \succsim_1 g_2 \succsim_1 g_3 \succsim_1 g_4$ $\succsim_2: g_2 \succsim_2 g_1 \succsim_2 g_4 \succsim_2 g_3$

	g_1	g_2	g_3	g_4
a_1	1/2	1/2	1/2	1/2
a_2	1/2	1/2	1/2	1/2

SD-strategyproofness

- SD-strategyproofness: No agent should be able to improve their allocation by misreporting their preferences.

- **Theorem:**

- Random Priority is SD-strategyproof.

- Probabilistic Serial is not SD-strategyproof

- $\succsim_1: \cancel{g_1} \succsim_1 \cancel{g_2} \succsim_1 g_3 \succsim_1 g_4$

- $\succsim_2: g_2 \succsim_2 g_3 \succsim_2 g_1 \succsim_2 g_4$

$g_2 \succsim_1 g_1$

	g_1	g_2	g_3	g_4
a_1	1	0	1/2	1/2
a_2	0	1	1/2	1/2

	g_1	g_2	g_3	g_4
a_1	1	1/2	0	1/2
a_2	0	1/2	1	1/2

SD-Efficiency + SD-Strategyproofness

- Theorem [Bogomolnaia and Moulin 2001]:
 - No mechanism satisfies SD-efficiency, SD-strategyproofness, and equal treatment of equals
- We can get SD-efficiency + SD-envy-freeness
 - SD-envy-freeness: $\forall i, j: \sum_{j=1}^m A_{i,j} g_j \succsim_i^{SD} \sum_{j=1}^m B_{i,j} g_j$
 - Probabilistic Serial is SD-envyfree

Public Decisions










Public Decisions Model

- Set of agents N
- Set of issues T
- Each issue has associated set of alternatives $C^t = \{c_1^t, \dots, c_{k_t}^t\}$
- Agents have utility functions $u_i^t: A^t \rightarrow \mathbb{R}_+$

	Issue 1			Issue 2			...	Issue T		
	c_1^1	c_2^1	c_3^1	c_1^2	c_2^2	c_3^2		c_1^T	c_2^T	c_3^T
a_1	3	1	0	2	5	1		6	5	5
a_2	2	2	1	3	4	1		2	4	3
a_3	0	0	4	4	3	2		5	4	5

Public Decisions Model

- Set of agents N
- Set of issues T
- Each issue has associated set of alternatives $C^t = \{c_1^t, \dots, c_{k_t}^t\}$
- Agents have utility functions $u_i^t: A^t \rightarrow \mathbb{R}_+$

	Monday			Tuesday			...	Sunday		
										
a_1	3	1	0	2	5	1		6	5	5
a_2	2	2	1	3	4	1		2	4	3
a_3	0	0	4	4	3	2		5	4	5

Item Allocation as a Special Case

- Define the set of issues $T = M = \{g_1, \dots, g_m\}$
- Alternatives $C^t = N = \{a_1, \dots, a_n\}$
- $u_i^t(a_j) = \begin{cases} v_i(g_t) & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$

	g_1	g_2	g_3
a_1	5	2	3
a_2	0	3	1
a_3	2	3	4



	g_1			g_2			g_3		
	a_1	a_2	a_3	a_1	a_2	a_3	a_1	a_2	a_3
a_1	5	0	0	2	0	0	3	0	0
a_2	0	0	0	0	3	0	0	1	0
a_3	0	0	2	0	0	3	0	0	4

Fairness for Public Decisions

- Envy-freeness (and relaxations) not sensible in the general case
 - Decisions are public, all agents receive the same outcome
- Proportionality is still sensible
 - Each agent should receive their “dictator utility” multiplied by $1/n$
- **Proportionality up to one *issue* (Prop1)**
 - Each agent would receive their proportional share if they were allowed to change the outcome on a single issue
- **Theorem [Conitzer et al. 2017]:**
 - The MNW outcome satisfies Prop1 + PO in the public decisions setting
- Other fairness desiderata ((approximate) core, round robin share,...)

Allocation of Public Goods

	Issue 1			Issue 2		
	c_1^1	c_2^1	c_3^1	c_1^2	c_2^2	c_3^2
a_1	3	1	0	2	5	1
a_2	2	2	1	3	4	1
a_3	0	0	4	4	3	2



	g_1	g_2	g_3	g_4	g_5	g_6
a_1	3	1	0	2	5	1
a_2	2	2	1	3	4	1
a_3	0	0	4	4	3	2

- Generalizes public decisions
- A set of **public goods** $\{g_1, \dots, g_m\}$
 - Each good can give a positive utility to multiple agents simultaneously
- Constraints on which subsets of public goods are feasible

Allocation of Public Goods

	Issue 1			Issue 2		
	c_1^1	c_2^1	c_3^1	c_1^2	c_2^2	c_3^2
a_1	3	1	0	2	5	1
a_2	2	2	1	3	4	1
a_3	0	0	4	4	3	2



	g_1	g_2	g_3	g_4	g_5	g_6
a_1	3	1	0	2	5	1
a_2	2	2	1	3	4	1
a_3	0	0	4	4	3	2

- Public decision example:
 - Exactly one of $\{g_1, g_2, g_3\}$ and exactly one of $\{g_4, g_5, g_6\}$ must be chosen
 - Partition matroid constraint

Fairness Guarantees

- (δ, α) -Core

- An allocation of public goods C is in (δ, α) -core if for every subset of agents $S \subseteq N$, there is no feasible allocation of public goods C' such that

$$\frac{|S|}{n} \cdot u_i(C') \geq (1 + \delta) \cdot u_i(C) + \alpha$$

for all $i \in S$, and at least one inequality is strict.

- Valuations are normalized so that $\max_j u_i(g_j) = 1$
- Core (i.e. $(0,0)$ -core) generalizes proportionality
 - $(0,1)$ -core generalizes a guarantee very similar to Prop1

Fair Allocation of Public Goods

- **Matroid constraints**
 - Public goods are ground set elements
 - Feasible allocations are basis of a matroid
 - Generalizes public decisions (thus goods allocation) and multiwinner voting
- **Theorem [Fain et al. 2018]**
 - For matroid constraints, a $(0,2)$ -core allocation exists, and for constant $\epsilon > 0$, a $(0,2 + \epsilon)$ -core allocation can be computed in polynomial time.
 - **Algorithm:** Maximize **smooth Nash welfare** $\prod_{i \in N} (1 + u_i(C))$
 - For $\epsilon > 0$, $(0,1 - \epsilon)$ -core allocations may not exist.
- **Open problem:** Does there always exist a $(0,1)$ -core allocation?

Fair Allocation of Public Goods

- Theorem [Fain et al. 2018]

- For “matching constraints” and constant $\delta \in (0,1]$, a $(\delta, 8 + 6/\delta)$ -core allocation can be computed in polynomial time.
- Algorithm: Maximize a slightly different smooth NW $\prod_{i \in N} (1 + 4/\delta + u_i(C))$
- For $\delta > 0$ and $\alpha < 1$, a (δ, α) -core allocation may not exist.
- Open problem: Does there always exist a $(0,1)$ -core allocation?

- A slightly worse guarantee with logarithmically large α in case of “packing constraints”

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