



# Fair Division of Indivisible Items (Part II)

CS 580

Instructor: [Ruta Mehta](#)



Most slides are curtesy Prof. J. Garg

- $N$ : set of  $n$  agents,  $1, \dots, n$ ,
- $M$ : set of  $m$  **indivisible** items (like cell phone, painting, etc.)



- Agent  $i$  has a **valuation** function  $v_i : 2^m \rightarrow \mathbb{R}$  over **subsets of items**
  - **Monotone**: the more the happier



# Proportionality

- A set  $N$  of  $n$  agents, a set  $M$  of  $m$  indivisible goods
- **Proportionality:** Allocation  $A = (A_1, \dots, A_n)$  is proportional if each agent gets at least  $1/n$  share of all items:

$$v_i(A_i) \geq \frac{v_i(M)}{n}, \quad \forall i \in N$$

Cut-and-choose?



## Maximin Share (MMS) [B11]

### Cut-and-choose.






- Suppose we allow agent  $i$  to propose a partition of items into  $n$  bundles with the condition that  $i$  will choose at the end.
- Clearly,  $i$  partitions items in a way that **maximizes** the value of her **least preferred bundle**.
- $\mu_i :=$  Maximum value of  $i$ 's least preferred bundle


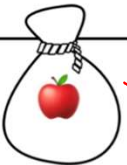

# Maximin Share (MMS) [B11]


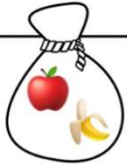

## Cut-and-choose.

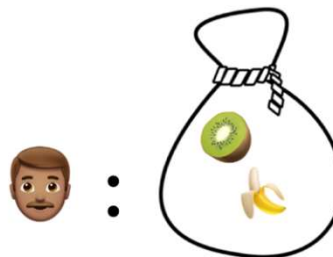
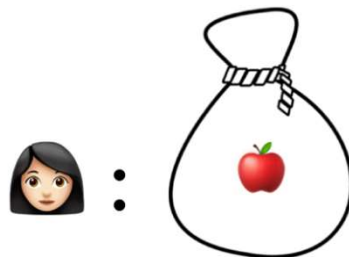
- Suppose we allow agent  $i$  to propose a partition of items into  $n$  bundles with the condition that  $i$  will choose at the end
- Clearly,  $i$  partitions items in a way that maximizes the value of her least preferred bundle
- $\mu_i :=$  Maximum value of  $i$ 's least preferred bundle
- $\Pi :=$  Set of all partitions of items into  $n$  bundles
- $\mu_i := \max_{A \in \Pi} \min_{A_k \in A} v_i(A_k)$
- **MMS Allocation:**  $A$  is called MMS if  $v_i(A_i) \geq \mu_i, \forall i$
- **Additive valuations:**  $v_i(A_i) = \sum_{j \in A_i} v_{ij}$

# MMS value/partition/allocation



Agent\Items			
	3	1	2
	4	4	5


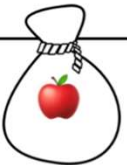

	 A	 B
Value	3	3
MMS Value	3	


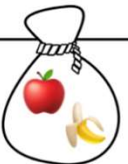

		
Value	8	5
MMS Value	5	

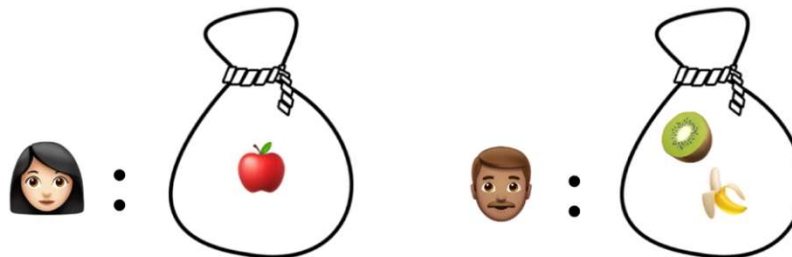


# MMS value/partition/allocation

Agent\Items			
	3	1	2
	4	4	5

		
Value	3	3
MMS Value	3	

		
Value	8	5
MMS Value	5	




**Finding MMS value is NP-hard!**

# What is Known?

- PTAS for finding MMS value [W97]

**Existence** (MMS allocation)?


- $n = 2$  : yes   
 $\Rightarrow$  A PTAS to find  $(1 - \epsilon)$ -MMS allocation for any  $\epsilon > 0$
- $n \geq 3$  : NO [PW14]



# What is Known?

- PTAS for finding MMS value [W97]

Existence (MMS allocation)?

- $n = 2$  : yes   
 $\Rightarrow$  A PTAS to find  $(1 - \epsilon)$ -MMS allocation for any  $\epsilon > 0$
- $n \geq 3$  : NO [PW14]
- $\alpha$ -MMS allocation for  $\alpha \in [0,1]$ :  $v_i(A_i) \geq \alpha \cdot \mu_i$ 
  - $2/3$ -MMS exists [PW14, AMNS17, BK17, KPW18, GMT18]
  - $3/4$ -MMS exists [GHSSY18]
  - $(3/4 + 1/(12n))$ -MMS exists [GT20]

$\frac{39}{40}$  - MMS does not exist

# Properties

## ■ Normalized valuations

□ Scale free:  $v_{ij} \leftarrow c \cdot v_{ij}, \forall j \in M$

□  $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$  ?  
 $\quad \quad \quad \text{"}$   
 $\quad \quad \quad v_i(M)$

MMS partition to agent i:

$v_i(A_1) \geq \dots \geq v_i(A_n) = \mu_i > 1$

$\sum_K v_i(A_K) = v_i(M) > n!$



# Properties

- **Normalized valuations**

- **Scale free:**  $v_{ij} \leftarrow c \cdot v_{ij}, \forall j \in M$

- $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$

- **Ordered Instance:** We can assume that agents' order of preferences for items is same:  $v_{i1} \geq v_{i2} \geq \dots v_{im}, \forall i \in N$








# Properties



## ■ Normalized valuations

□ Scale free:  $v_{ij} \leftarrow c \cdot v_{ij}, \forall j \in M$

□  $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$

■ **Ordered Instance:** We can assume that agents' order of preferences for items is same:  $v_{i1} \geq v_{i2} \geq \dots v_{im}, \forall i \in N$

					
	3	1	2	5	4
	4	4	5	3	2

	1	2	3	4	5
	5	4	3	2	1
	5	4	4	3	2

# Challenge

- Allocation of **high-value items!**
- If for all  $i \in N$ 
  - $v_i(M) = n \Rightarrow \mu_i \leq 1$
  - $v_{ij} \leq \epsilon, \forall i, j \Rightarrow \max_{i,j} v_{ij} \leq \epsilon$

$$v_{ij} \leq \epsilon, \forall i, j$$

$$g_1 > g_2 > \dots > g_m$$

Claim: After round  $k$ , if  $i$  remains then  $v_i(\text{remaining goods}) \geq n - k$ .

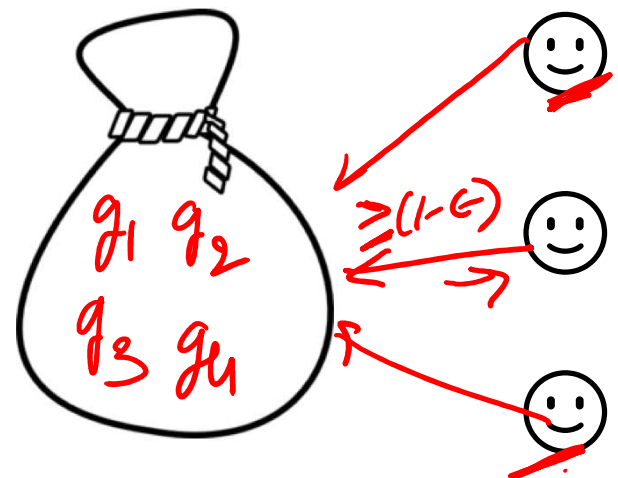
$$v_i(M \setminus \{g_1, g_2, g_3, g_4\}) > (n-1)$$

$$\uparrow \uparrow$$

$$v_i(\{g_1, g_2, g_3\}) < (1-\epsilon)$$

$$v_i(g_4) \leq \epsilon$$

$$v_i(\{g_1, g_2, g_3, g_4\}) < (1-\epsilon) + \epsilon < 1.$$



## Bag Filling Algorithm:

Repeat until every agent is assigned a bag

- Start with an empty bag  $B$
- Keep adding items to  $B$  until some agent  $i$  values it  $\geq (1 - \epsilon)$
- Assign  $B$  to  $i$  and remove both

$$v_{ij} \leq \epsilon, \forall i, j \quad v_i(M) = n$$

**Claim:** After round  $k$ , if  $i$  remains then  $v_i(\text{remaining goods}) \geq n - k$ .

PS: In every round, value of the assigned set of items for agent  $i < 1$ .



### Bag Filling Algorithm:

Repeat until every agent is assigned a bag

- Start with an empty bag  $B$
- Keep adding items to  $B$  until some agent  $i$  values it  $\geq (1 - \epsilon)$
- Assign  $B$  to  $i$  and remove them both

$$v_{ij} \leq \epsilon, \forall i, j$$

**Thm:** Every agent gets at least  $(1 - \epsilon)$ .



### Bag Filling Algorithm:

Repeat until every agent is assigned a bag

- Start with an empty bag  $B$
- Keep adding items to  $B$  until some agent  $i$  values it  $\geq (1 - \epsilon)$
- Assign  $B$  to  $i$  and remove them both

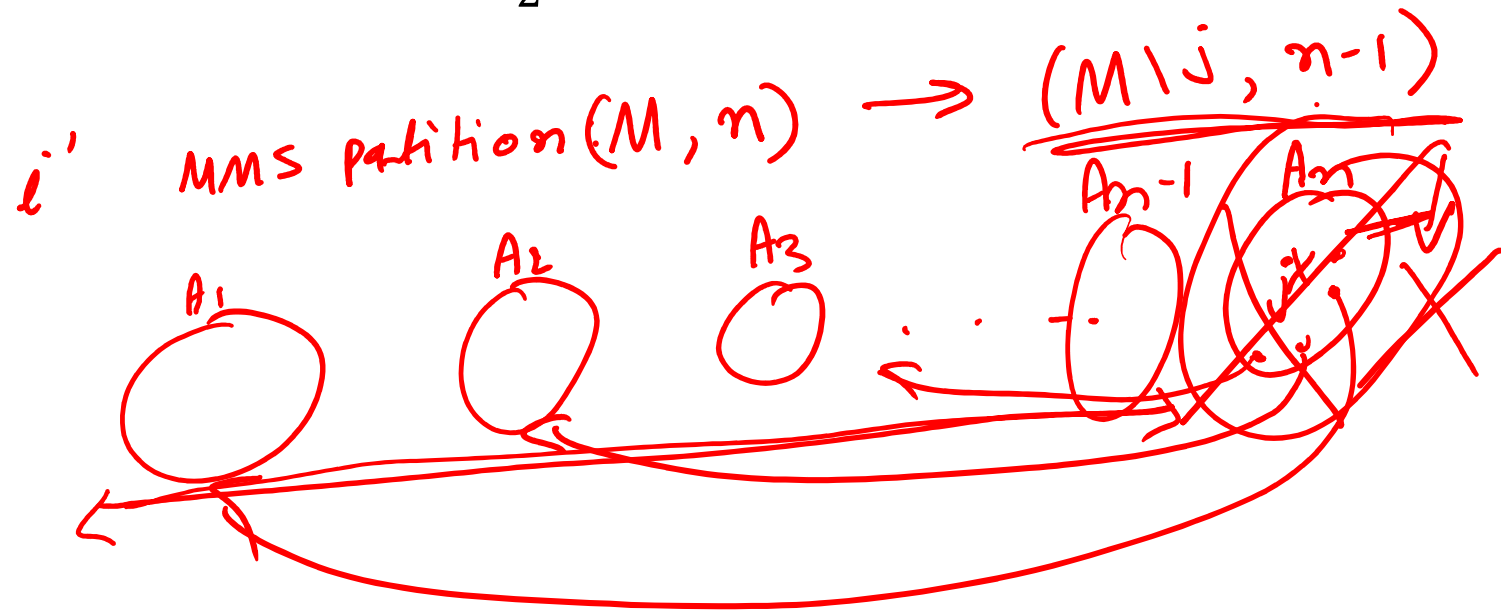


# Warm Up: 1/2-MMS Allocation

- If all  $v_{ij} \leq \frac{1}{2}$  then?

□ Done, using bag filling.

- What if some  $v_{ij} > \frac{1}{2}$ ?



# Valid Reductions

- Normalized valuations
  - Scale free:  $v_{ij} \leftarrow c \cdot v_{ij}, \forall j \in M$
  - $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$
- Ordered Instance: Agents' order of preferences for items is same:  $v_{i1} \geq v_{i2} \geq \dots v_{im}, \forall i \in N$
- **Valid Reduction ( $\alpha$ -MMS):** If there exists  $S \subseteq M$  and  $i^* \in N$ 
  - $v_{i^*}(S) \geq \alpha \cdot \mu_{i^*}^n(M)$
  - $\mu_i^{n-1}(M \setminus S) \geq \mu_i^n(M), \forall i \neq i^*$

$\Rightarrow$  reduce the instance size!

**Claim.** Suppose agent  $i \neq i^*$  gets  $A_i$  in the be an  $\alpha$ -MMS allocation of  $M \setminus S$  to agents  $N \setminus \{i^*\}$ , then  $(A_1, \dots, A_{i^*-1}, S, A_{i^*+1}, \dots, A_n)$  is an  $\alpha$ -MMS allocation in the original instance.

pf:

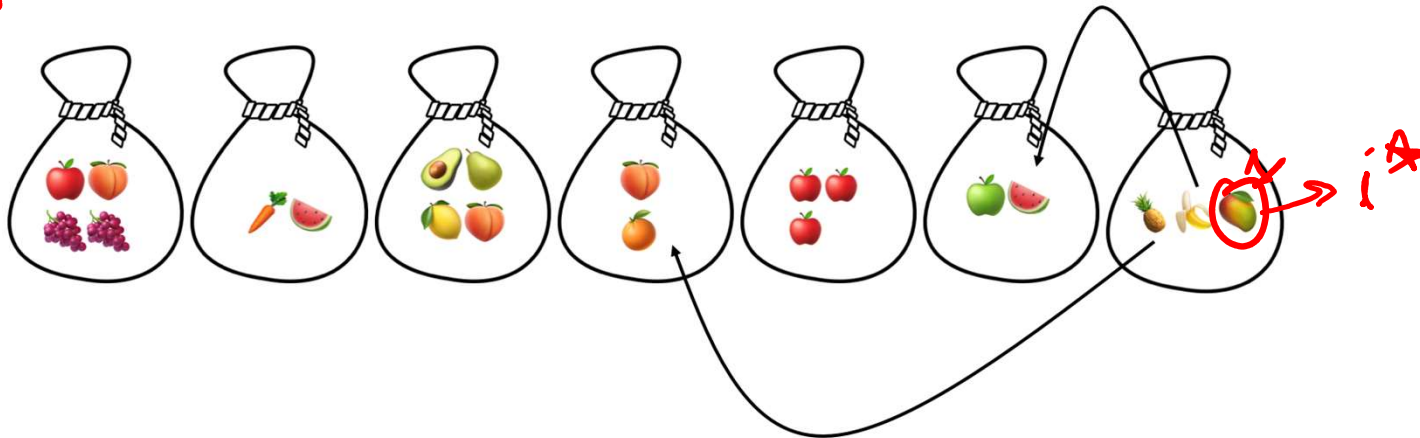
$$\frac{v_i(A_i)}{v_{i^*}(S)} \geq \frac{\alpha \mu_i^{n-1}(M \setminus S)}{\alpha \mu_{i^*}^n(M)} \geq \frac{\alpha \mu_i^{n-1}(M \setminus S)}{\alpha \mu_i^n(M)}$$

# 1/2-MMS Allocation

## Step 1: Valid Reductions

- If  $v_{i^*1} \geq 1/2$  then assign item 1 to  $i^*$

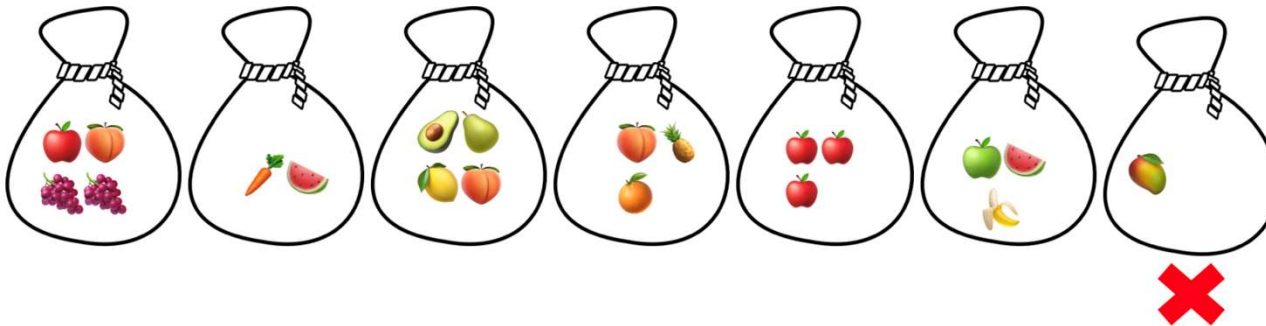
$i$ :



# 1/2-MMS Allocation

## Step 1: Valid Reductions

- If  $v_{i^*1} \geq 1/2$  then assign item 1 to  $i^*$



# 1/2-MMS Allocation

## ■ Re-normalization

$$\epsilon = 1/2$$

Step 0: Normalized Valuations:  $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$

Step 1: Valid Reductions  $\rightarrow (1-\epsilon)$

- If  $v_{i^*1} \geq 1/2$  then assign item 1 to  $i^*$ . Remove good 1 and agent  $i^*$
- After every valid reduction, normalize valuations

Step 2: Bag Filling  $\leftarrow v_{ij} < \frac{1}{2} = \epsilon$

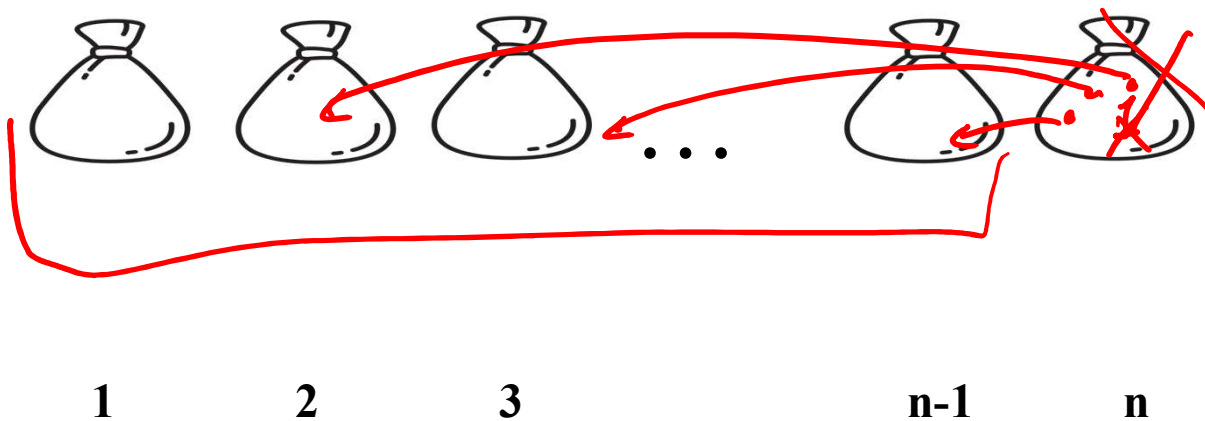
## 2/3-MMS Allocation [GMT19]

- If all  $v_{ij} \leq 1/3$  then?  $\epsilon = 1/3$

### Step 1: Valid Reductions

- If  $v_{i^*_1} \geq 2/3$  then assign item 1 to  $i^*$

agent  $i$ :  
MMS part.



# 2/3-MMS Allocation [GMT19]

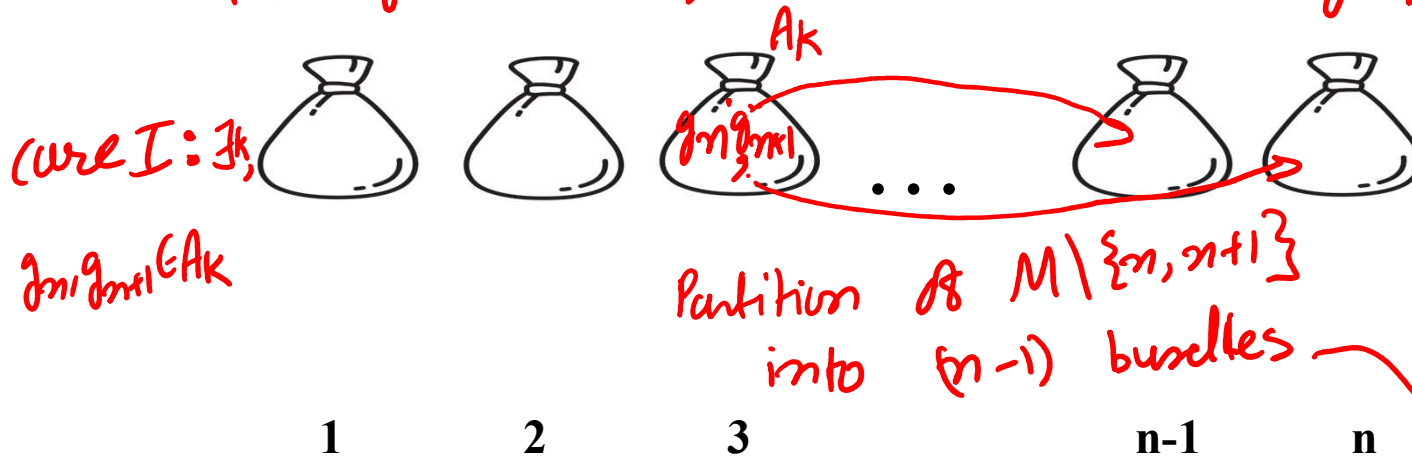
## Step 1: Valid Reductions

- If  $v_{i^*1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$

Q:

why valid reduction?

For agent  $i \neq i^*$ , let the MMS defining partition be



Then move  $A_k \setminus \{n, n+1\}$  to other bundles & remove  $A_k$

each with value  $\geq u_i$

# 2/3-MMS Allocation [GMT19]

## Step 1: Valid Reductions

- If  $v_{i^*1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$

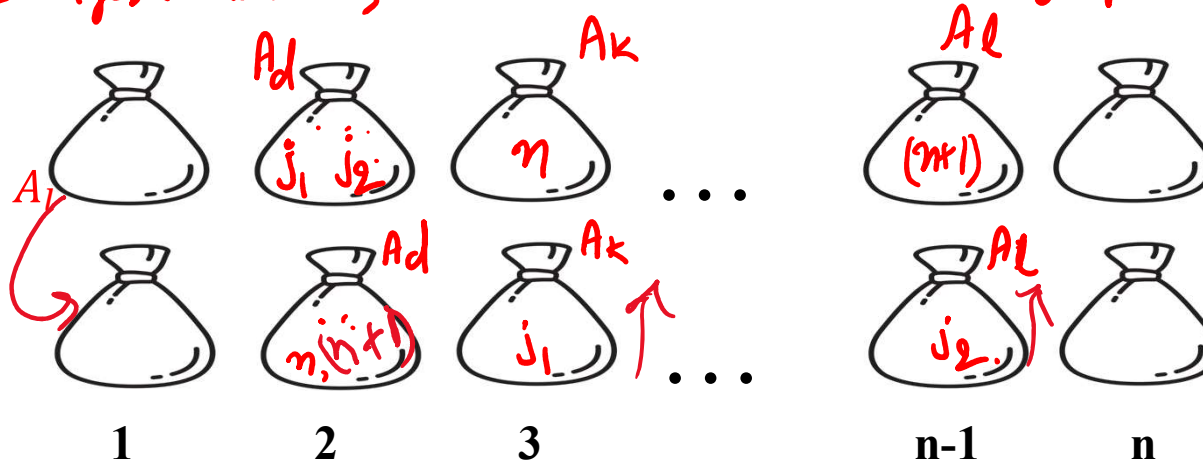
Q: Why valid reduction?

For agent  $i \neq i^*$ , let the MMS defining partition be

Case II:

$n \in A_k$

$(n+1) \in A_l$



$\exists A_d$ , with items  $j_1 < j_2 \leq (n+1)$ . Then, swap items  $j_1$  and  $n$ , and items  $j_2$  and  $(n+1)$ . This may only increase  $v_i(A_k)$  &  $v_i(A_l)$  because  $v_i(j_1) \geq v_i(n)$  &  $v_i(j_2) \geq v_i(n+1)$ .



# 2/3-MMS Allocation [GMT19]

## Step 1: Valid Reductions

- If  $v_{i^*1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$

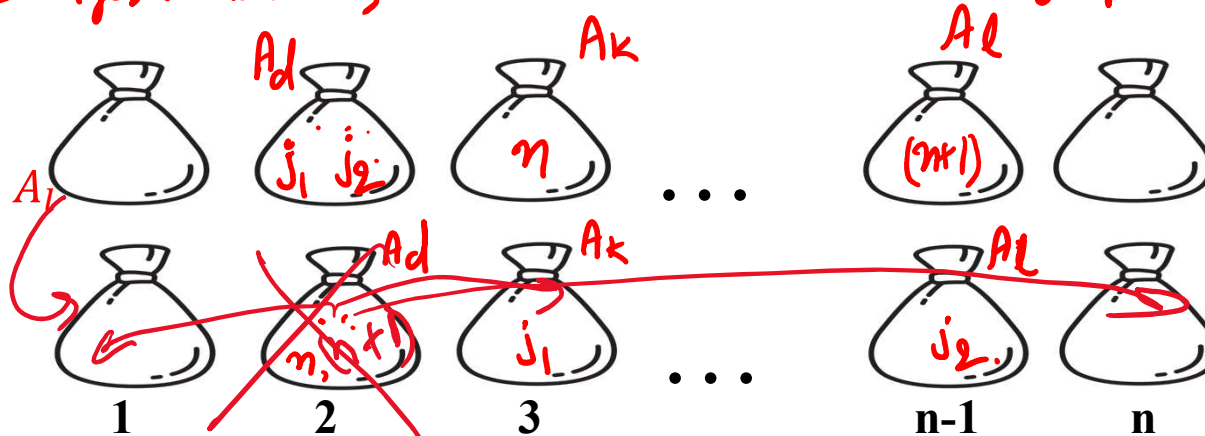
Q: Why valid reduction?

For agent  $i \neq i^*$ , let the MMS defining partition be

Case II:

$n \in A_k$

$(n+1) \in A_d$



$\exists A_d$ , with items  $j_1 < j_2 \leq (n+1)$ .

Then, swap items  $j_1$  and  $n$ , and items  $j_2$  and  $(n+1)$ .

Move remaining items of  $A_d$  to other bundles and remove  $A_d$ .

# 2/3-MMS Allocation [GMT19]

## Step 1: Valid Reductions

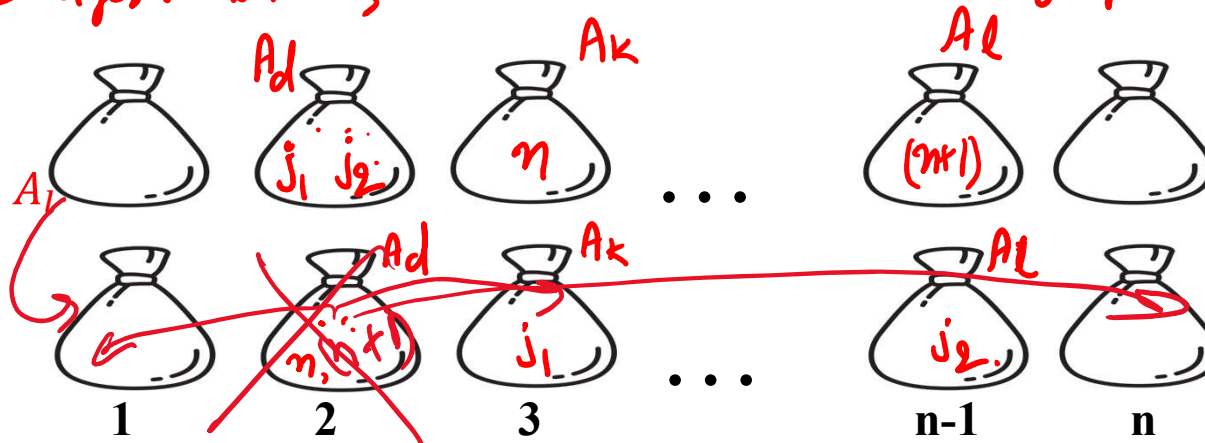
- If  $v_{i^*1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$

Q: Why valid reduction?

For agent  $i \neq i^*$ , let the MMS defining partition be

Case II:

$n \in A_k$   
 $(n+1) \in A_d$



$\exists A_d$ , with items  $j_1 < j_2 \leq (n+1)$ .

Then, swap items  $j_1$  and  $n$ , and items  $j_2$  and  $(n+1)$ .

Move remaining items of  $A_d$  to other bundles and remove  $A_d$ .

Again, value of none of the remaining bundles has decreased.

$\Rightarrow$  MMS value of agent  $i$  has only increased in the reduced instance.

## Step 1: Valid Reductions

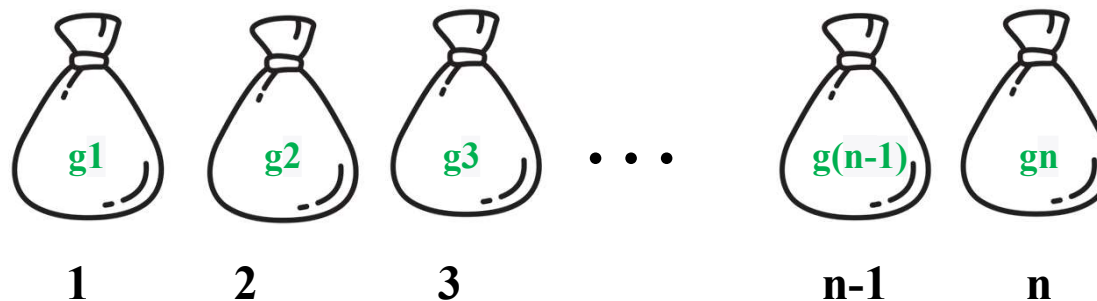
- If  $v_{i^*_1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*_n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$

## Step 2: Generalized Bag Filling with $\epsilon = \frac{1}{3}$

- Initialize  $n$  bags  $\{B_1, \dots, B_n\}$  with  $B_k = \{k\}, \forall k$ .
- Assign items starting from  $(n+1)$ th to the first available bag, and give it to the first agent who shouts (values it at least  $2/3 = (1 - \epsilon)$ ).

After Step 1,  
For each agent  $i$ ,  
 $v_{ij} < \frac{2}{3}, \forall j \leq n$   
 $v_{ij} < \frac{1}{3}, \forall j > n$

**Claim.** If agent  $i^*$  is the first to shout, then for any agent  $i \neq i^*$  the bag is of value at most 1.



## 2/3-MMS Allocation [GMT19]

### ■ (Re)normalization

**Step 0: Normalized Valuations:**  $\sum_j v_{ij} = n \Rightarrow \mu_i \leq 1$

**Step 1: Valid Reductions**

- If  $v_{i^*_1} \geq 2/3$  then assign item 1 to  $i^*$
- If  $v_{i^*_n} + v_{i^*(n+1)} \geq 2/3$  then assign  $\{n, n+1\}$  to  $i^*$
- After every valid reduction, normalize valuations

**Step 2: Generalized Bag Filling with  $\epsilon = \frac{1}{3}$**

- Initialize  $n$  bags  $\{B_1, \dots, B_n\}$  with  $B_k = \{k\}, \forall k$




# Chores

- $N$ : set of  $n$  agents,  $1, \dots, n$ ,
- $M$ : set of  $m$  **indivisible** chores



- Agent  $i$  has a **disutility** function  $d_i : 2^m \rightarrow \mathbb{R}_+$  over **subsets of items**
  - **Monotone**: the more the **un**-happier
- **Additive**:  $d_i(S) = \sum_{j \in S} d_{ij}$ , for any subset  $S \subseteq M$

- 
- $N$ : set of  $n$  agents,  $1, \dots, n$ ,
  - $M$ : set of  $m$  **indivisible** chores
  - Agent  $i$  has a **disutility** function  $d_i : 2^M \rightarrow \mathbb{R}_-$  over **subsets of items**
    - **Additive**:  $d_i(S) = \sum_{j \in S} d_{ij}$ , for any subset  $S \subseteq M$

Allocation  $A = (A_1, \dots, A_n)$

**EF1: No agent envies another after removing one of her chores.**

$$\forall i, k \in N, \quad d_i(A_i \setminus c) \leq d_i(A_k), \quad \exists c \in A_i$$

# EF1: Algorithms

## Round Robin

1. Order agents arbitrarily.
2. Let them pick their best chore (least painful chore), one-at-a-time, in that order.

### Observations:

- If agent  $k$  picks the last chore, then agent  $(k + 1)$  does not envy anyone. Why?



# EF1: Algorithms

## Envy-cycle-elimination

1.  $A = (\emptyset, \dots, \emptyset)$
2. While there are unassigned chores
  1. Construct envy-graph of  $A$  and remove any cycles.
  2. Give an unassigned chore to ..... ??

Observations:

- Cycle elimination does not increase any agent's disutility.
- Giving a chore to sink maintains EF1. Why?

# MMS

- $N$ : set of  $n$  agents,  $1, \dots, n$ ,
- $M$ : set of  $m$  **indivisible** chores
- Agent  $i$  has a **disutility** function  $d_i : 2^M \rightarrow \mathbb{R}_+$  over **subsets of items**
  - **Additive**:  $d_i(S) = \sum_{j \in S} d_{ij}$ , for any subset  $S \subseteq M$

- $\Pi :=$  Set of all partitions of items into  $n$  bundles

**MMS value:** 
$$\text{MMS}_i = \mu_i = \min_{A \in \Pi} \max_{A_k \in A} d_i(A_k)$$

$\alpha$ -MMS allocation for  $\alpha \geq 1$ :  $\forall i, d_i(A_i) \leq \alpha \mu_i$

1-MMS allocation may not exist!

# EF1 to $\alpha$ -MMS

**Claim.** If  $(A_1, \dots, A_n)$  is EF1 then it is 2-MMS

**Observations:**  $\mu_i \geq \frac{d_i(M)}{n}$  and  $\mu_i \geq \max_{j \in M} d_{ij}$

**Proof.**



# Summary

## Covered

- Additive Valuations:
  - $\frac{1}{2}$ -MMS allocation (poly-time algorithm)
  - $\frac{2}{3}$ -MMS allocation (polynomial-time algorithm)

## State-of-the-art

- $\left(\frac{3}{4} + \right)$ -MMS allocation [GT20]
- More general valuations
  - MMS [GHSSY18]
- Groupwise-MMS [BBKN18]
- Chores:  $\frac{11}{9}$ -MMS [HL19]

## Major Open Questions (additive)

- $c$ -MMS + PO: polynomial-time algorithm for a constant  $c > 0$
- Existence of  $\frac{4}{5}$ -MMS allocation? For 5 agents?

# References (Indivisible Case).

- [AMNS17] Georgios Amanatidis, Evangelos Markakis, Afshin Nikzad, and Amin Saberi. "Approximation algorithms for computing maximin share allocations". In: *ACM Trans. Algorithms* 13.4 (2017)
- [BBKN18] Siddharth Barman, Arpita Biswas, Sanath Kumar Krishnamurthy, and Y. Narahari. "Groupwise maximin fair allocation of indivisible goods". In: *AAAI* 2018
- [BK17] Siddharth Barman and Sanath Kumar Krishna Murthy. "Approximation algorithms for maximin fair division". In *EC* 2017
- [BK19] Siddharth Barman and Sanath Kumar Krishnamurthy. "On the Proximity of Markets with Integral Equilibria" In *AAAI* 2019
- [BKV18] Siddharth Barman, Sanath Kumar Krishnamurthy, and Rohit Vaish. Finding fair and efficient allocations. In: *EC* 2018
- [B11] Eric Budish. "The combinatorial assignment problem: Approximate competitive equilibrium from equal incomes". In: *J. Political Economy* 119.6 (2011)
- [CKMPSW14] Ioannis Caragiannis, David Kurokawa, Herve Moulin, Ariel Procaccia, Nisarg Shah, and Junxing Wang. "The Unreasonable Fairness of Maximum Nash Welfare". In: *EC* 2016
- [GMT19] Jugal Garg, Peter McGlaughlin, and Setareh Taki. "Approximating Maximin Share Allocations". In: *SOSA@SODA* 2019
- [GT20] Jugal Garg and Setareh Taki. "An Improved Approximation Algorithm for Maximin Shares". In: *EC* 2020
- [GHSSY18] Mohammad Ghodsi, MohammadTaghi HajiAghayi, Masoud Seddighin, Saeed Seddighin, and Hadi Yami. "Fair allocation of indivisible goods: Improvement and generalization". In *EC* 2018
- [HL19] Xin Huang and Pinyan Lu. "An algorithmic framework for approximating maximin share allocation of chores". In: *arxiv:1907.04505*
- [KBKZ09] Bart de Keijzer, Sylvain Bouveret, Tomas Klos, and Yingqian Zhang. "On the Complexity of Efficiency and Envy-Freeness in Fair Division of Indivisible Goods with Additive Preferences". In: *Algorithmic Decision Theory (ADT)*. 2009
- [KPW18] David Kurokawa, Ariel D. Procaccia, and Junxing Wang. "Fair Enough: Guaranteeing Approximate Maximin Shares". In: *J. ACM* 65.2 (2018), 8:1–8:27
- [PW14] Ariel D Procaccia and Junxing Wang. "Fair enough: Guaranteeing approximate maximin shares". In *EC* 2014
- [W97] Gerhard J Woeginger. "A polynomial-time approximation scheme for maximizing the minimum machine completion time". In: *Operations Research Letters* 20.4 (1997)