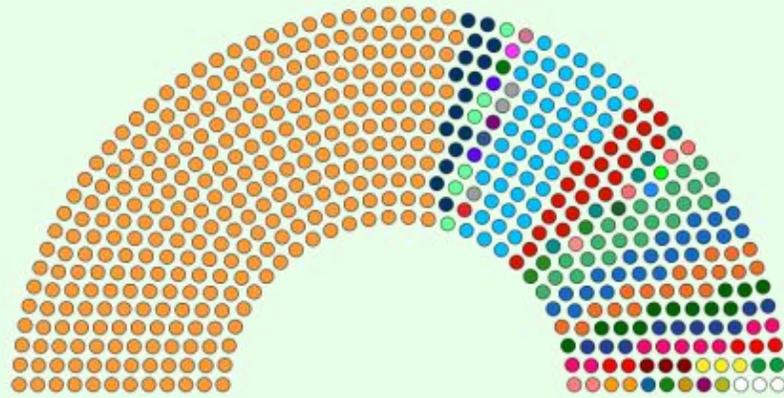


**f**airness through

**r**andomness

Rohit Vaish

# INDIVISIBLE



# The Model

	(A)	(B)	(C)	(D)	(E)
	4	1	2	2	2
	1	0	5	1	1
	1	1	5	1	1

# The Model

	(A)	(B)	(C)	(D)	(E)
	4	1	2	2	2
	1	0	5	1	1
	1	1	5	1	1

# The Model

	(A)	(B)	(C)	(D)	(E)
	4	1	2	2	2
	1	0	5	1	1
	1	1	5	1	1

Additive  
valuations

$$\begin{aligned} \triangle \{ \text{(B)} \text{(D)} \text{(E)} \} &= \triangle \{ \text{(B)} \} + \triangle \{ \text{(D)} \} + \triangle \{ \text{(E)} \} \\ &= 0 + 1 + 1 = 2 \end{aligned}$$

# Envy-Freeness [Gamow and Stern, 1958; Foley, 1967]

Each agent prefers its own bundle over that of any other agent.

# Envy-Freeness [Gamow and Stern, 1958; Foley, 1967]

Each agent prefers its own bundle over that of any other agent.

	(A)	(B)	(C)
My bundle is the best	4	1	2
My bundle is the best	1	1	5

Allocation  $A = (A_1, A_2, \dots, A_n)$  is EF if for every pair of agents  $i, k$ , we have  $v_i(A_i) \geq v_i(A_k)$ .

# Envy-Freeness [Gamow and Stern, 1958; Foley, 1967]

Each agent prefers its own bundle over that of any other agent.

	(A)	(B)	(C)
My bundle is the best	4	1	2
My bundle is the best	1	1	5

Allocation  $A = (A_1, A_2, \dots, A_n)$  is EF if for every pair of agents  $i, k$ , we have  $v_i(A_i) \geq v_i(A_k)$ .



Not guaranteed to exist (two agents, one good)



Checking whether an EF allocation exists is NP-complete

# Envy-Freeness Up To One Good [Budish, 2011]

Envy can be eliminated by removing some good in the envied bundle.

# Envy-Freeness Up To One Good [Budish, 2011]

Envy can be eliminated by removing some good in the envied bundle.

	(A)	(B)	(C)
My bundle is better if (A) is removed	4	1	2
My bundle is better if (C) is removed	1	1	5

Allocation  $A = (A_1, \dots, A_n)$  is EF1 if for every pair of agents  $i, k$ , there exists a good  $j \in A_k$  such that  $v_i(A_i) \geq v_i(A_k \setminus \{j\})$ .



Guaranteed to exist and efficiently computable

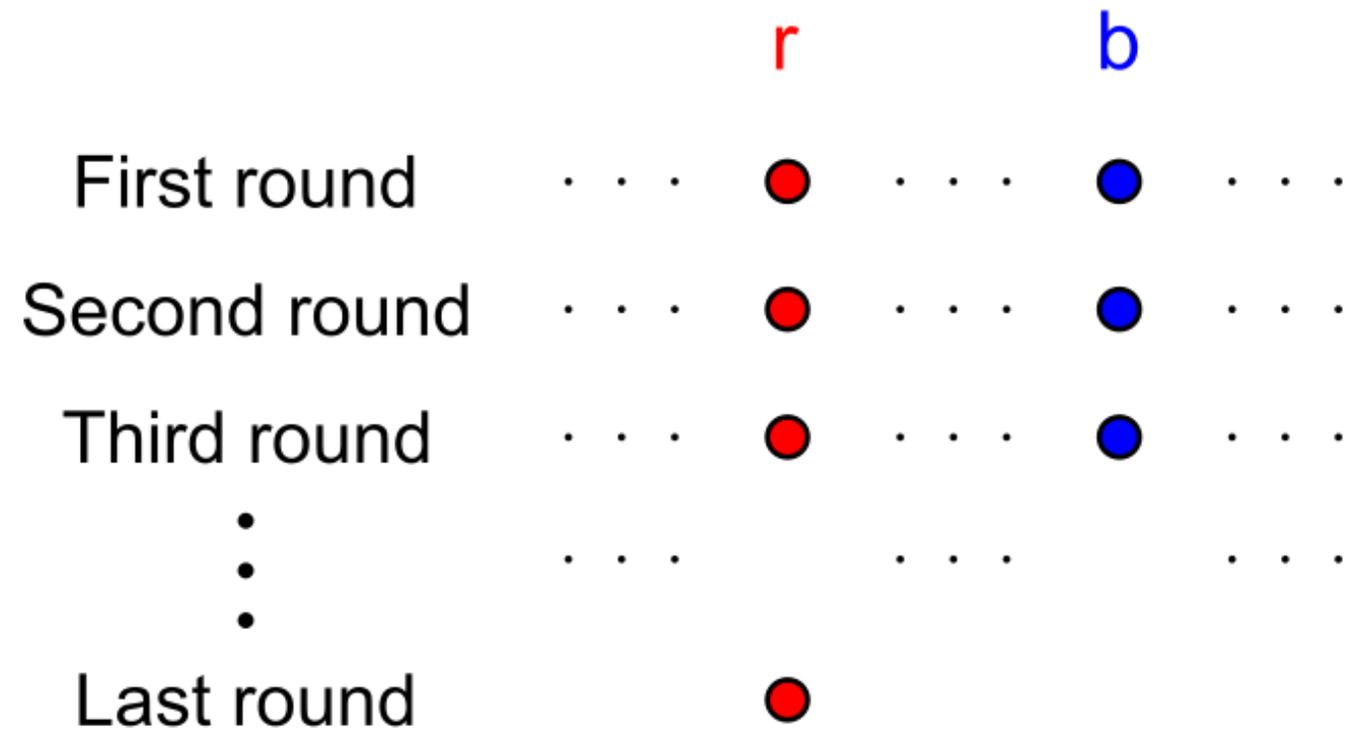
# Round-robin algorithm

- Fix an ordering of the agents, say  $a_1, a_2, a_3, \dots, a_n$ .
- Agents take turns according to the ordering  $(a_1, a_2, \dots, a_n, a_1, a_2, \dots, a_n, \dots)$  to pick their favorite item from the set of remaining items.

For additive valuations, the allocation computed by round-robin algorithm satisfies EF1.

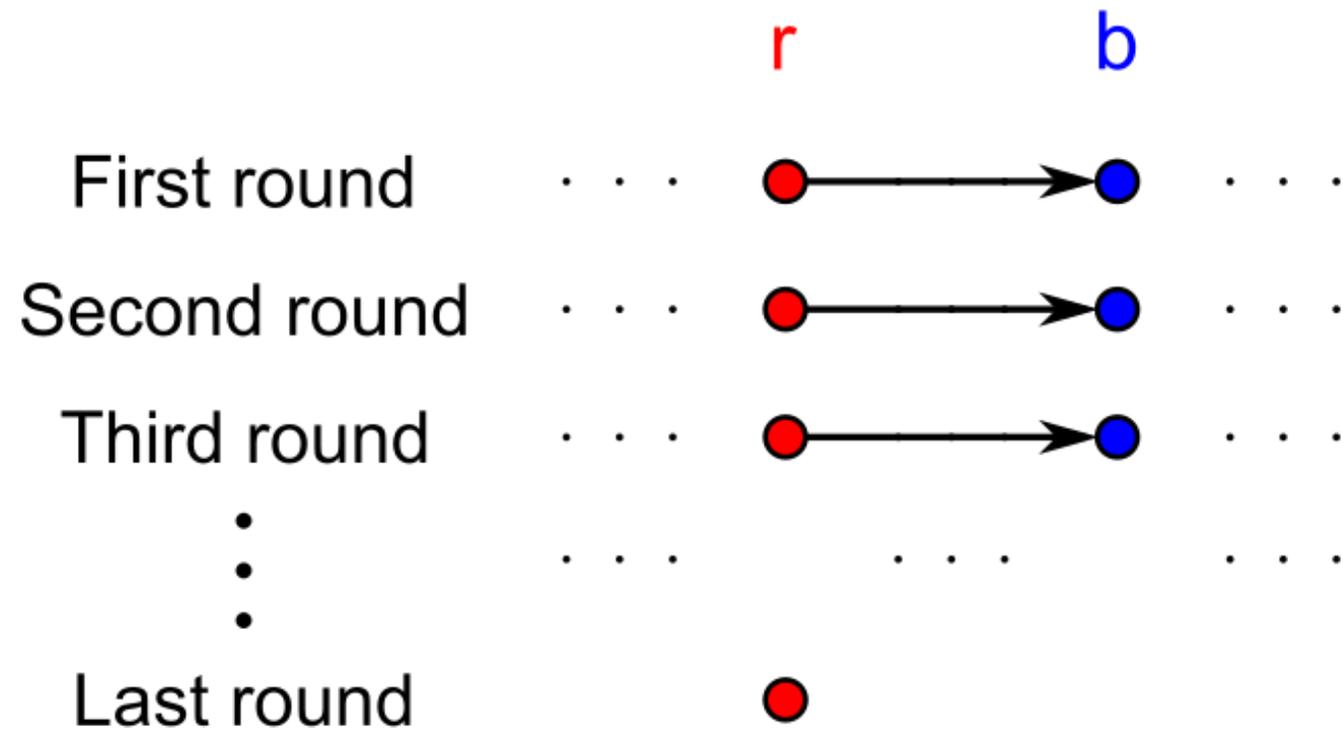
For additive valuations, the allocation computed by round-robin algorithm satisfies EF1.

Fix a pair of agents ( $r, b$ ). Analyze envy of  $r$  towards  $b$ .



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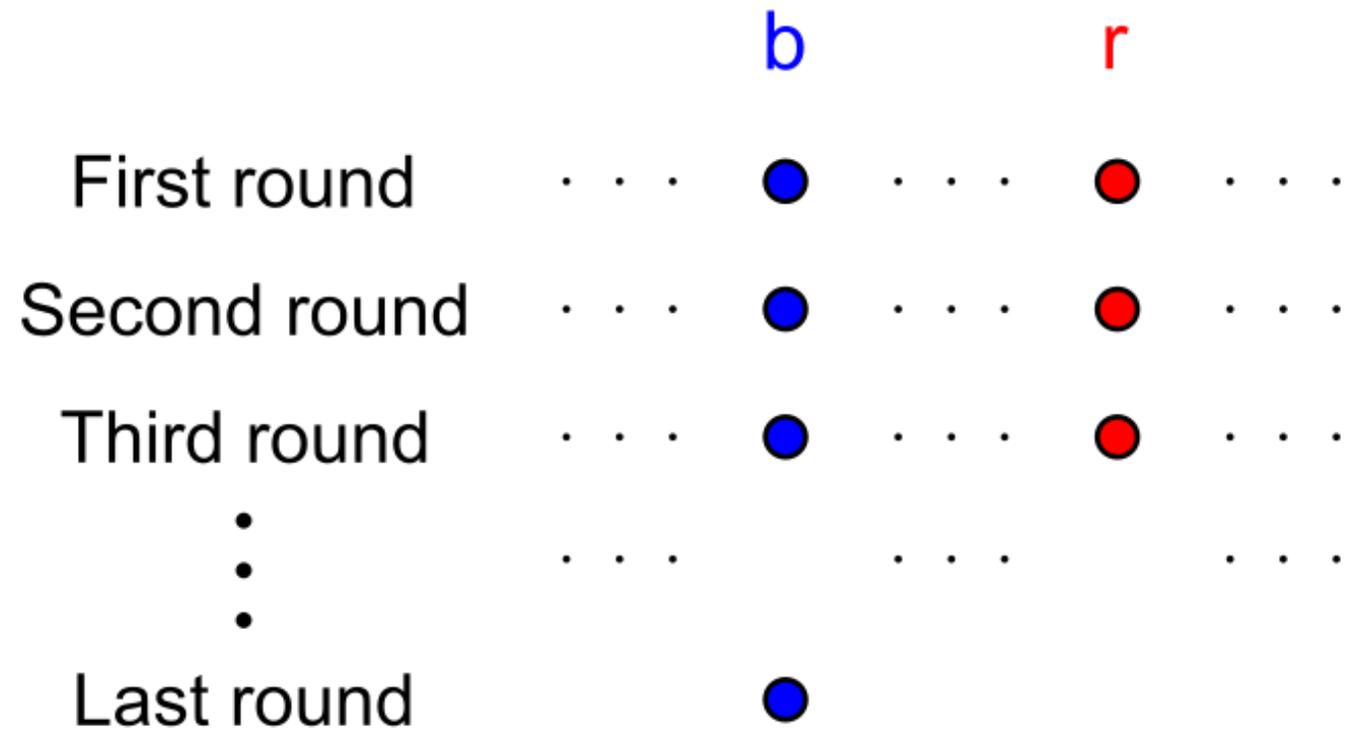
Fix a pair of agents ( $r, b$ ). Analyze envy of  $r$  towards  $b$ .



If  $r$  precedes  $b$ : Then, by additivity,  $v_r(A_r) \geq v_r(A_b)$ .

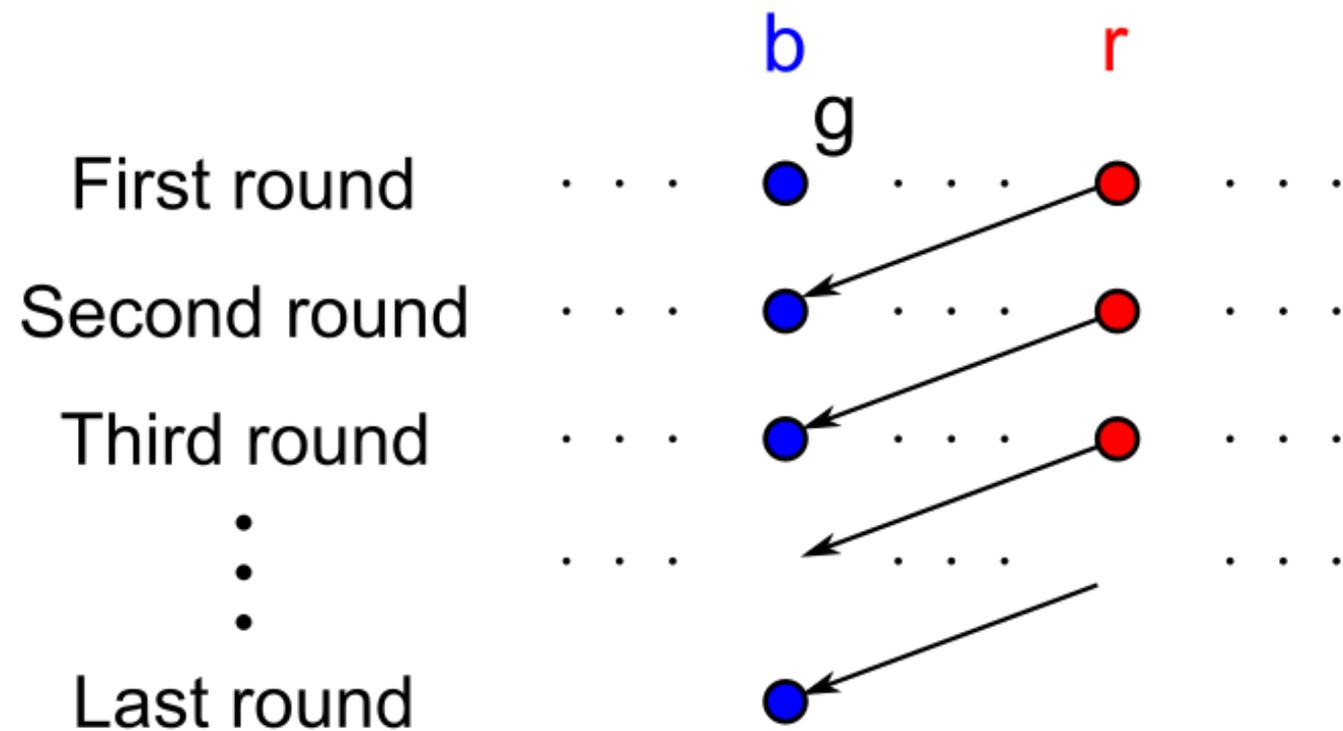
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If  $b$  precedes  $r$ : Again, by additivity,  $v_r(A_r) \geq v_r(A_b \setminus \{g\})$ .

**WHEN APPROXIMATE ENVY-FREENESS**



**SIMPLY ISN'T ENOUGH**



Day 1



Day 1



Day 1

Day 2



Day 1

Day 2



Day 1

Day 2

Day 3



Day 1

Day 2

Day 3

Day 4



**Deterministic** algorithms can systematically disadvantage certain agents.

A natural workaround: **Randomization**

# A natural workaround: Randomization

Pick a uniform distribution over all round-robin orderings

# A natural workaround: Randomization

Pick a uniform distribution over all round-robin orderings



Can still be **unfair**

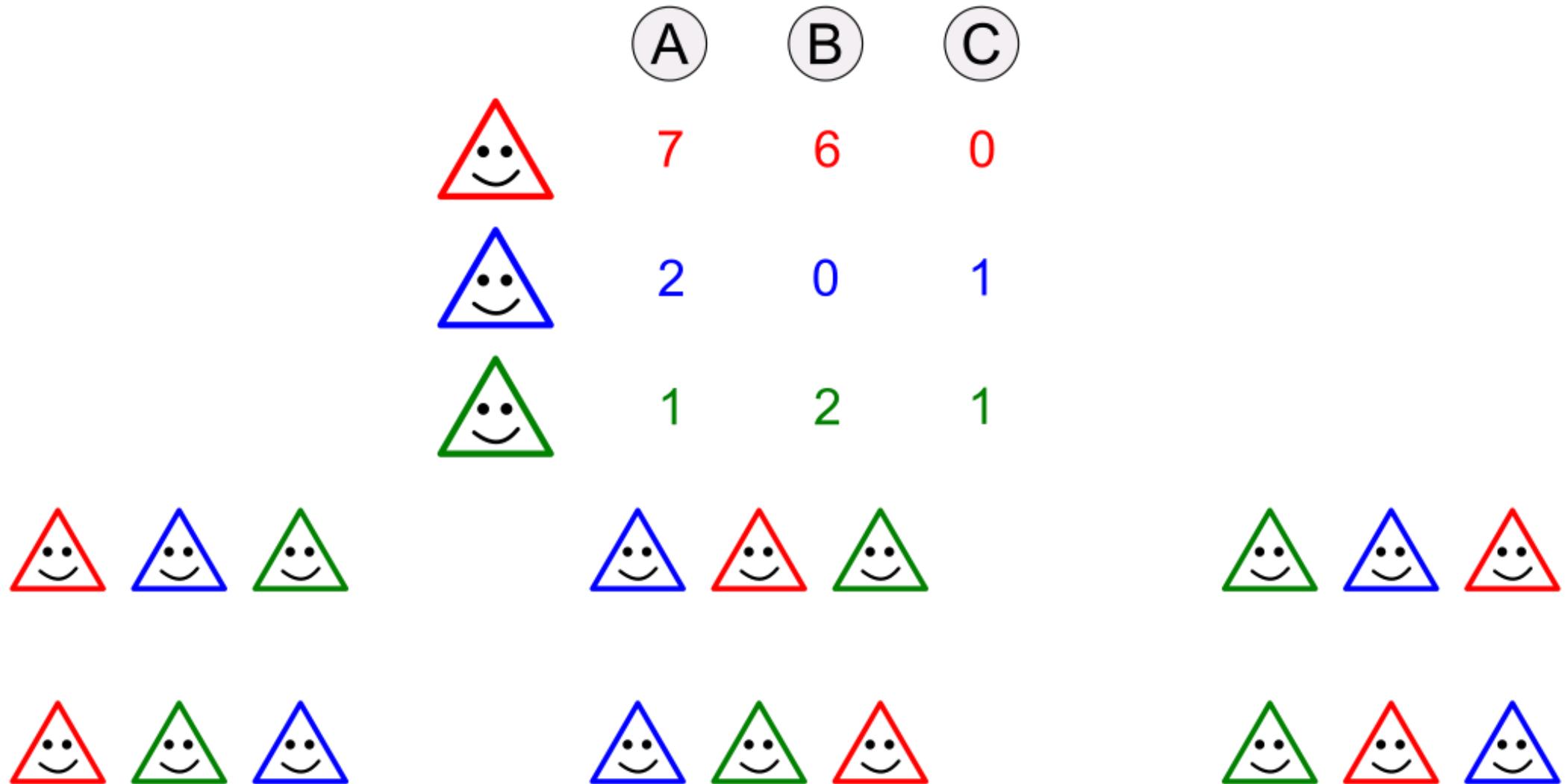
# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)
	7	6	0
	2	0	1
	1	2	1

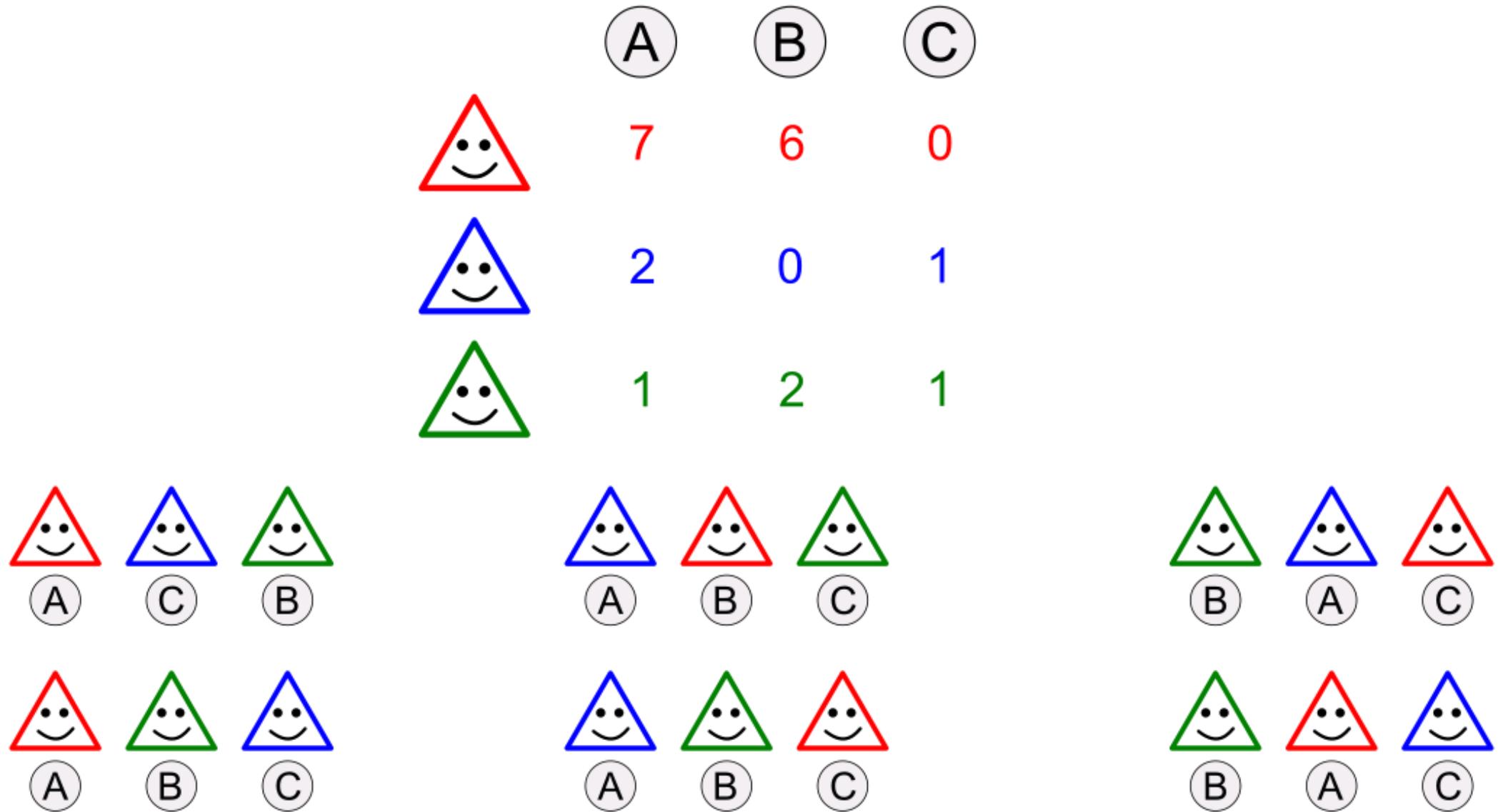
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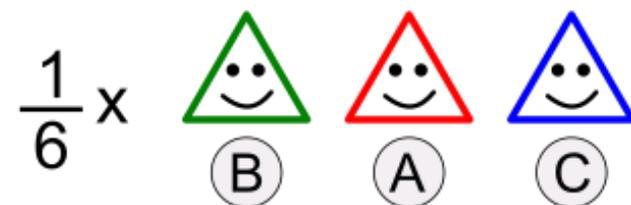
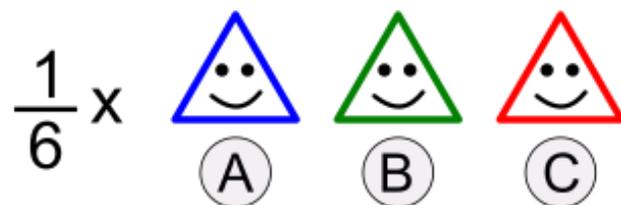
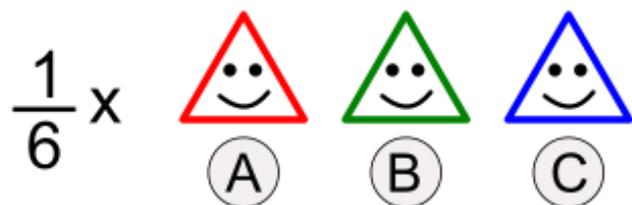
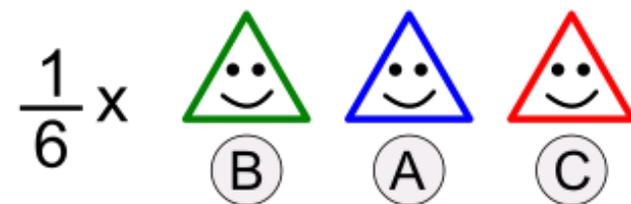
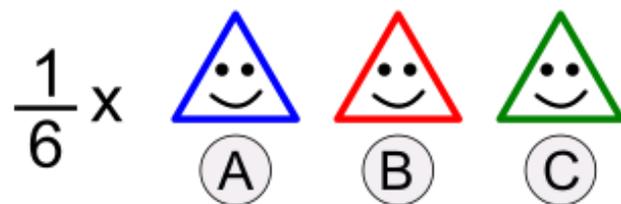
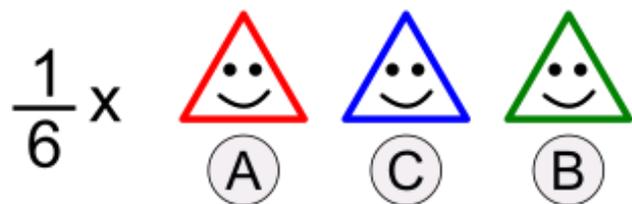
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	(A)	(B)	(C)
	7	6	0
	2	0	1
	1	2	1



# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

(A) (B) (C)

 1/2 1/6 1/3

 1/2 0 1/2

 0 5/6 1/6

$\frac{1}{6} \times$      
(A) (C) (B)

$\frac{1}{6} \times$      
(A) (B) (C)

$\frac{1}{6} \times$      
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$\frac{1}{6} \times$      
(B) (A) (C)

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)
	1/2	1/6	1/3
	1/2	0	1/2
	0	5/6	1/6

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)
	$1/2$	$1/6$	$1/3$
	$1/2$	0	$1/2$
	0	$5/6$	$1/6$

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)		(A)	(B)	(C)
	7	6	0		1/2	1/6	1/3
	2	0	1		1/2	0	1/2
	1	2	1		0	5/6	1/6

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[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)		(A)	(B)	(C)
	7	6	0		1/2	1/6	1/3
	2	0	1		1/2	0	1/2
	1	2	1		0	5/6	1/6

's expected value for its own bundle

's expected value for 's bundle

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	(A)	(B)	(C)		(A)	(B)	(C)
	7	6	0		1/2	1/6	1/3
	2	0	1		1/2	0	1/2
	1	2	1		0	5/6	1/6

's expected value for its own bundle =  $7 \cdot \frac{1}{2} + 6 \cdot \frac{1}{6} + 0 \cdot \frac{1}{3} = 4.5$

's expected value for 's bundle

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

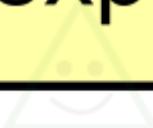
	(A)	(B)	(C)		(A)	(B)	(C)
	7	6	0		1/2	1/6	1/3
	2	0	1		1/2	0	1/2
	1	2	1		0	5/6	1/6

's expected value for its own bundle =  $7 \cdot 1/2 + 6 \cdot 1/6 + 0 \cdot 1/3 = 4.5$

's expected value for 's bundle =  $7 \cdot 0 + 6 \cdot 5/6 + 0 \cdot 1/6 = 5$

# Uniform round-robin is unfair

[Bogomolnaia and Moulin, 2001]

	A	B	C		A	B	C
	7	6	0		1/2	1/6	1/3
	2	0	1		1/2	0	1/2
	1	2	1		0	5/6	1/6

 envies  in expectation

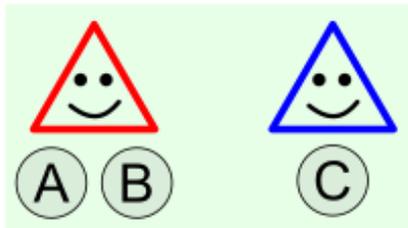
's expected value for its own bundle =  $7 \cdot 1/2 + 6 \cdot 1/6 + 0 \cdot 1/3 = 4.5$

's expected value for 's bundle =  $7 \cdot 0 + 6 \cdot 5/6 + 0 \cdot 1/6 = 5$

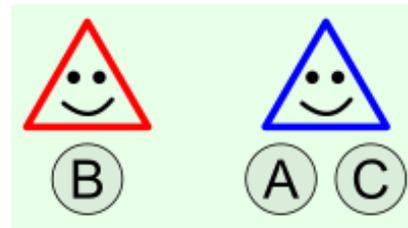
# Fairness of Randomized Allocations

# Fairness of Randomized Allocations

with prob  $1/4$



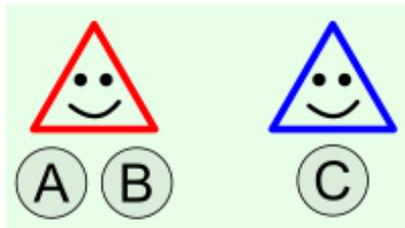
with prob  $3/4$



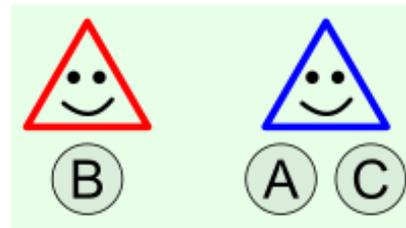
probability distribution over  
deterministic allocations

# Fairness of Randomized Allocations

with prob  $1/4$



with prob  $3/4$



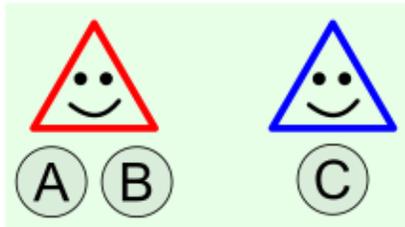
probability distribution over  
deterministic allocations

ex-ante  
fairness

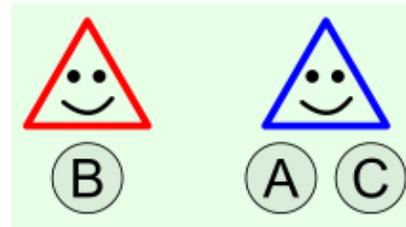
no agent envies another in expectation

# Fairness of Randomized Allocations

with prob  $1/4$



with prob  $3/4$



probability distribution over  
deterministic allocations

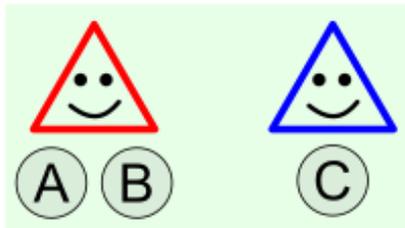
ex-ante  
fairness

no agent envies another in expectation

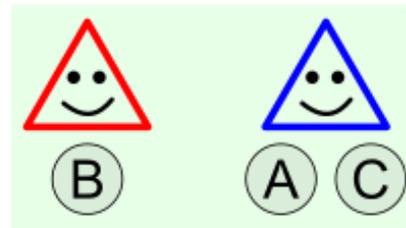
- Uniform round-robin **fails** ex-ante fairness.
- "Bundle everything together and assign uniformly randomly" **is** ex-ante fair.

# Fairness of Randomized Allocations

with prob  $1/4$



with prob  $3/4$



probability distribution over  
deterministic allocations

ex-ante  
fairness

no agent envies another in expectation

ex-post  
fairness

each deterministic allocation in the support is EF1

Does there always exist a randomized allocation that gives  
"best of both worlds", i.e., is ex-ante and ex-post fair?

ex-ante  
fairness

no agent envies another in expectation

ex-post  
fairness

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[Aziz, Freeman, Shah, Vaish, *Operations Research* 2023]

For additive valuations, there always exists a randomized allocation that is ex-ante envy-free and ex-post EF1. Such an allocation can be constructed in polynomial time.

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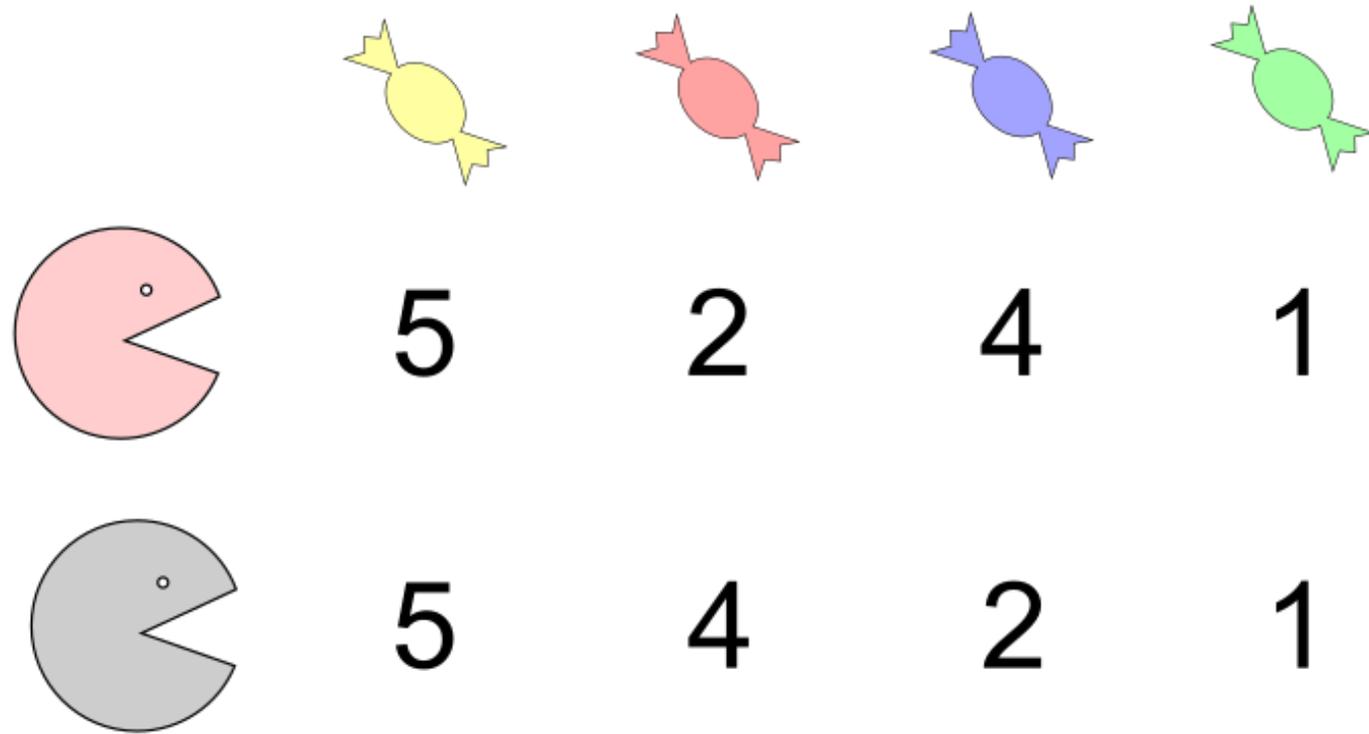
Proof by "eating".

# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]

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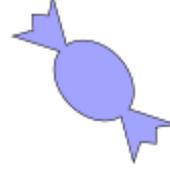
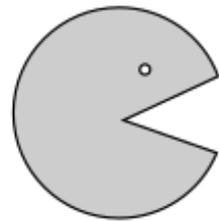
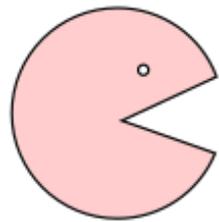
t=0

t=0.5

t=1

t=1.5

t=2



5

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4

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4

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# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



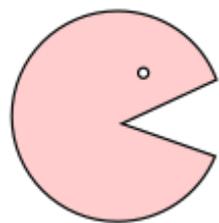
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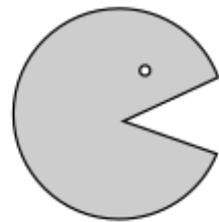


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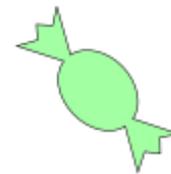
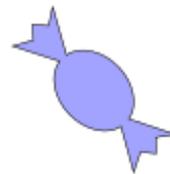


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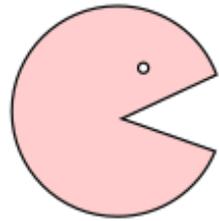
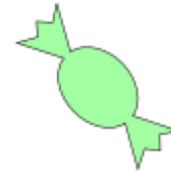
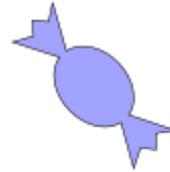
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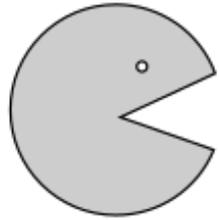


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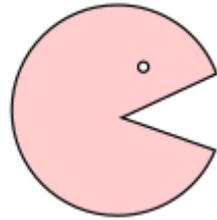
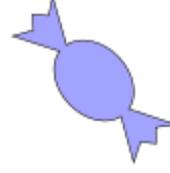
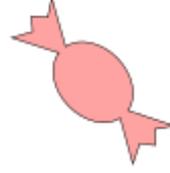
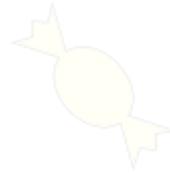
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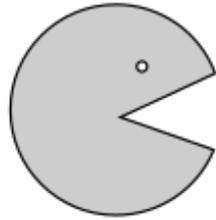


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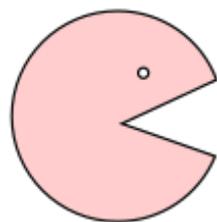
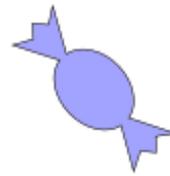
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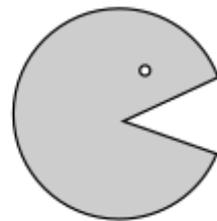


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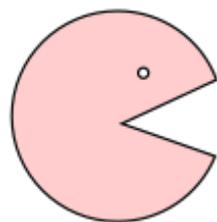
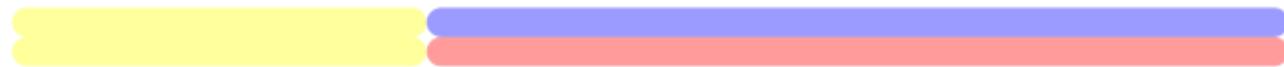
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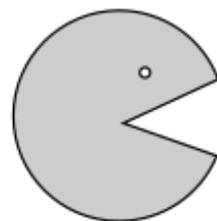


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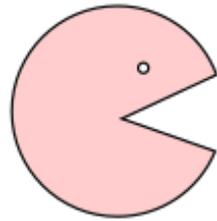
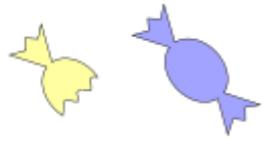
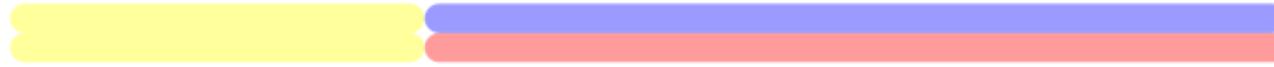
t=0

t=0.5

t=1

t=1.5

t=2



5

2

4

1

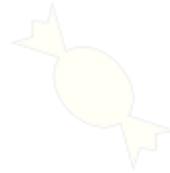


5

4

2

1



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



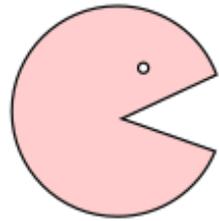
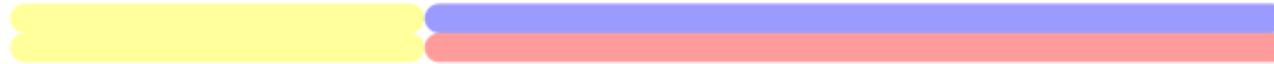
t=0

t=0.5

t=1

t=1.5

t=2

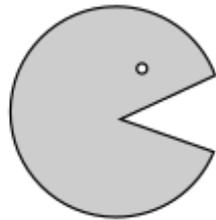


5

2

4

1

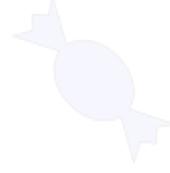
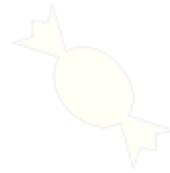


5

4

2

1



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



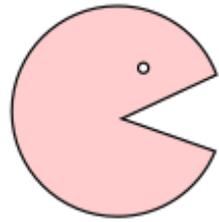
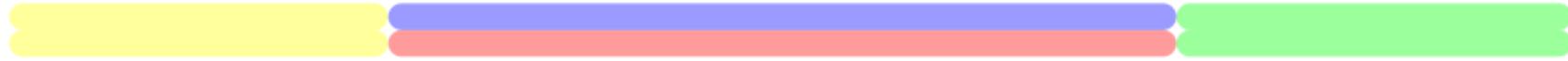
t=0

t=0.5

t=1

t=1.5

t=2

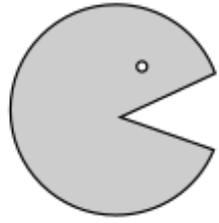


5

2

4

1

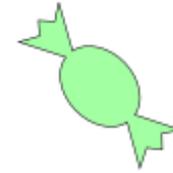
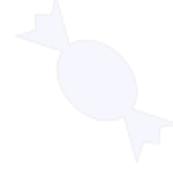


5

4

2

1



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



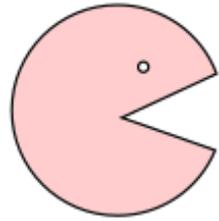
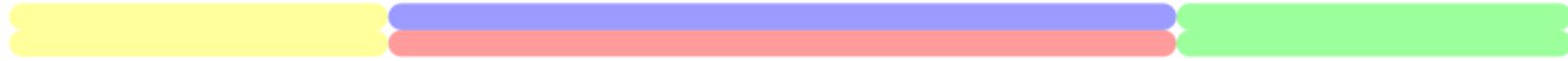
t=0

t=0.5

t=1

t=1.5

t=2



5

2

4

1

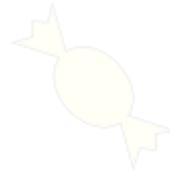


5

4

2

1



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



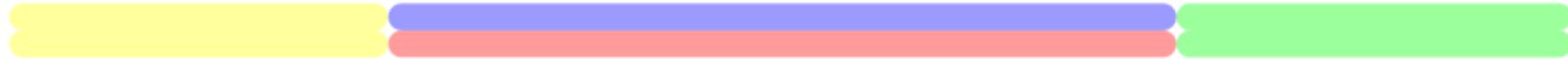
t=0

t=0.5

t=1

t=1.5

t=2



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



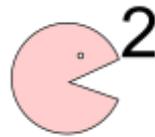
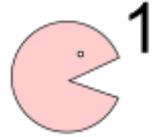
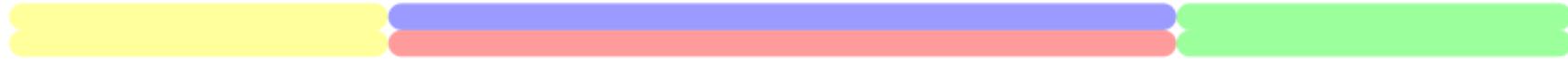
t=0

t=0.5

t=1

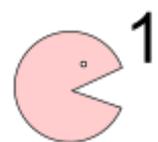
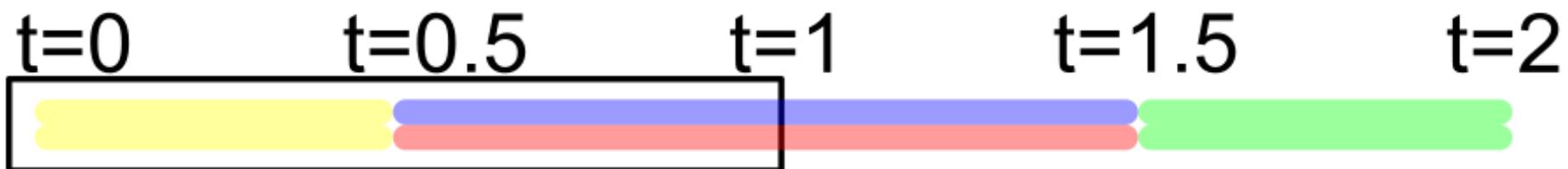
t=1.5

t=2



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



0.5

0

0.5

0

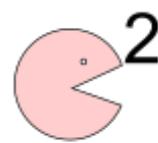


0.5

0.5

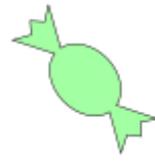
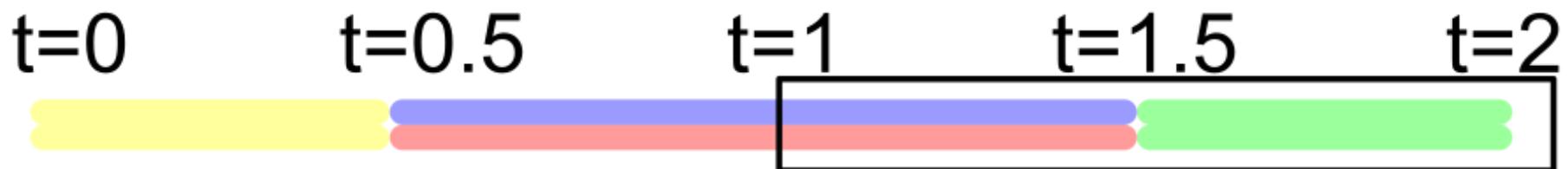
0

0



# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



0.5

0

0.5

0

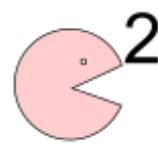


0.5

0.5

0

0



0

0

0.5

0.5



0

0.5

0

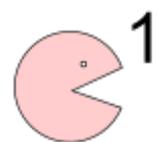
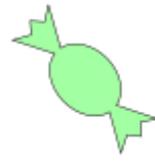
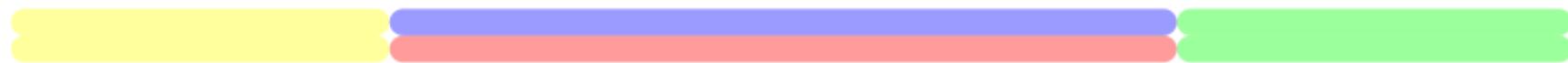
0.5

# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



t=0                      t=0.5                      t=1                      t=1.5                      t=2



0.5

0

0.5

0

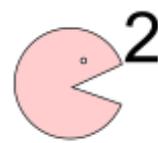


0.5

0.5

0

0



0

0

0.5

0.5



0

0.5

0

0.5

# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



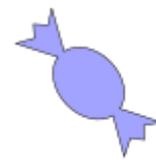
t=0

t=0.5

t=1

t=1.5

t=2



		0.5	0	0.5	0
Doubly		0.5	0.5	0	0
stochastic		0	0	0.5	0.5
		0	0.5	0	0.5

# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



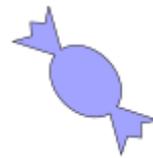
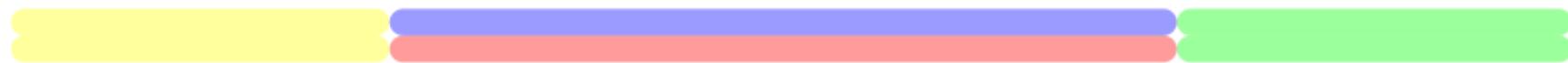
t=0

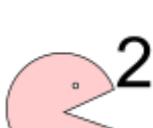
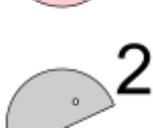
t=0.5

t=1

t=1.5

t=2

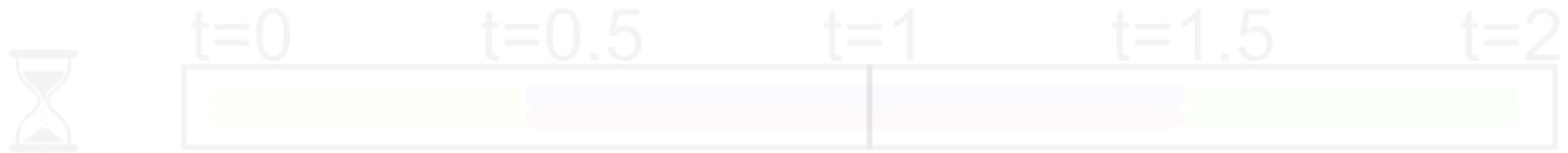


		0.5	0	0.5	0
Doubly		0.5	0.5	0	0
stochastic		0	0	0.5	0.5
		0	0.5	0	0.5

Apply  
Birkhoff-  
von Neumann  
decomposition

# Probabilistic Serial/ "Eating" Algorithm

[Bogomolnaia and Moulin, 2001]



Any doubly stochastic matrix can be expressed as a convex combination of permutation matrices.

Doubly  
stochastic



0.5	0.5	0	0
0	0	0.5	0.5
0	0.5	0	0.5

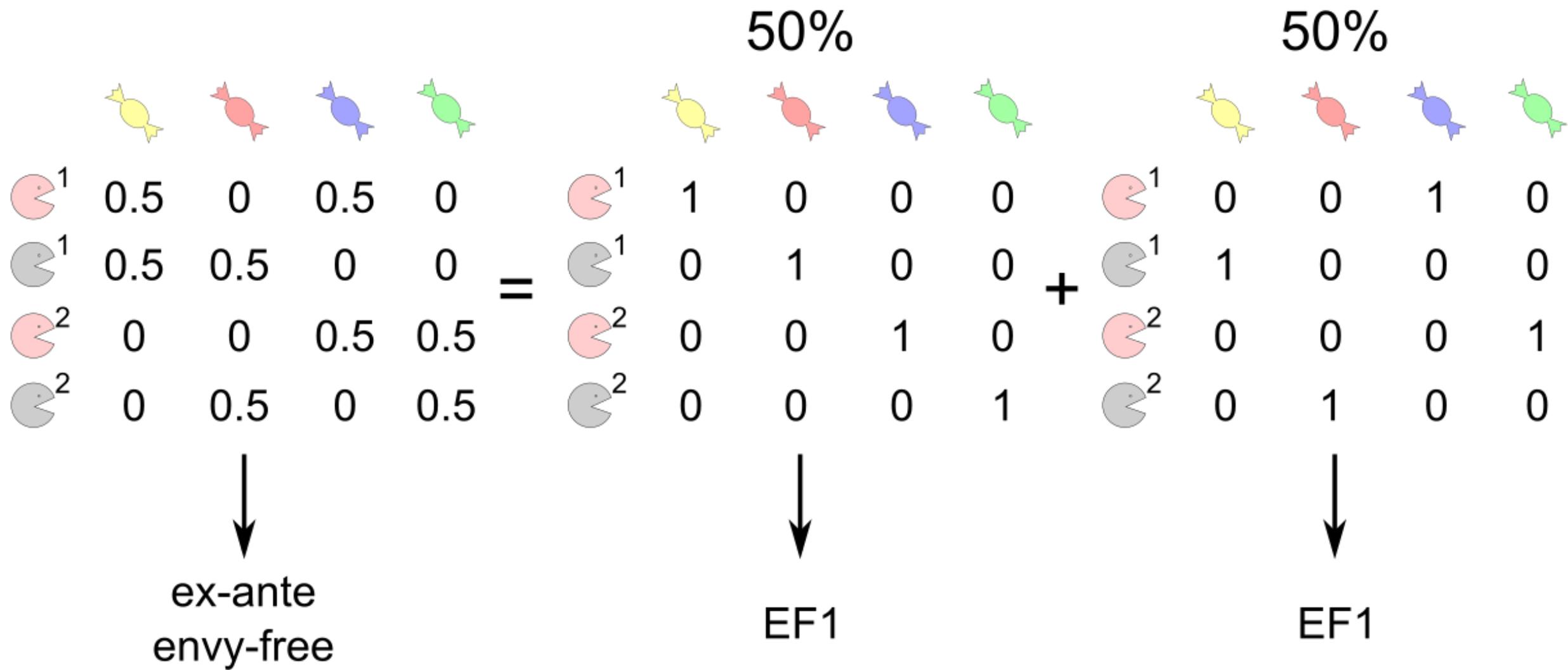
Birkhoff-  
von Neumann  
decomposition

																
 <sup>1</sup>	0.5	0	0.5	0	=	 <sup>1</sup>	1	0	0	0	+	 <sup>1</sup>	0	0	1	0
 <sup>1</sup>	0.5	0.5	0	0		 <sup>1</sup>	0	1	0	0		 <sup>1</sup>	1	0	0	0
 <sup>2</sup>	0	0	0.5	0.5		 <sup>2</sup>	0	0	1	0		 <sup>2</sup>	0	0	0	1
 <sup>2</sup>	0	0.5	0	0.5		 <sup>2</sup>	0	0	0	1		 <sup>2</sup>	0	1	0	0

50%

50%

**Claim:** This probability distribution is **ex-ante** and **ex-post fair**.



**Claim:** This probability distribution is **ex-ante** and **ex-post** fair.

				
 <sup>1</sup>	0.5	0	0.5	0
 <sup>1</sup>	0.5	0.5	0	0
 <sup>2</sup>	0	0	0.5	0.5
 <sup>2</sup>	0	0.5	0	0.5



ex-ante  
envy-free

because each agent eats  
its favorite good  
at each instant of time

**Claim:** This probability distribution is **ex-ante** and **ex-post** fair. ✓

				
 <sup>1</sup>	0.5	0	0.5	0
 <sup>1</sup>	0.5	0.5	0	0
 <sup>2</sup>	0	0	0.5	0.5
 <sup>2</sup>	0	0.5	0	0.5

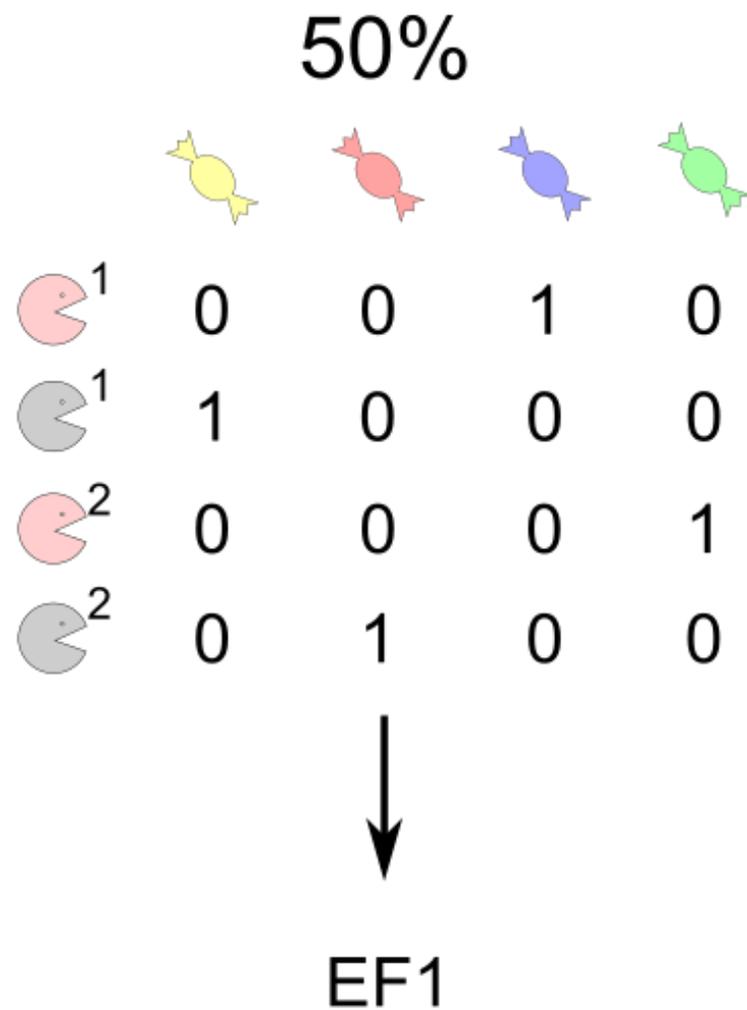
↓  
ex-ante  
envy-free

because each agent eats  
its favorite good  
at each instant of time





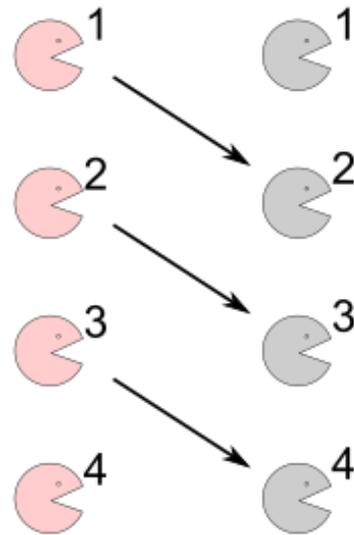
**Claim:** This probability distribution is **ex-ante** and **ex-post fair**.



**Claim:** This probability distribution is **ex-ante** and **ex-post fair**.

for any round  $t$ ,

 prefers the good assigned to  <sup>$t$</sup>   
over one assigned to  <sup>$t+1$</sup>



50%

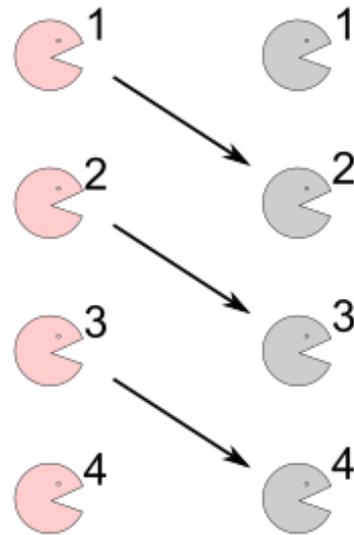
				
 <sup>1</sup>	0	0	1	0
 <sup>1</sup>	1	0	0	0
 <sup>2</sup>	0	0	0	1
 <sup>2</sup>	0	1	0	0

↓  
EF1

**Claim:** This probability distribution is **ex-ante** and **ex-post** fair.

for any round  $t$ ,

 prefers the good assigned to  <sup>$t$</sup>   
over one assigned to  <sup>$t+1$</sup>



50%

				
 <sup>1</sup>	0	0	1	0
 <sup>1</sup>	1	0	0	0
 <sup>2</sup>	0	0	0	1
 <sup>2</sup>	0	1	0	0

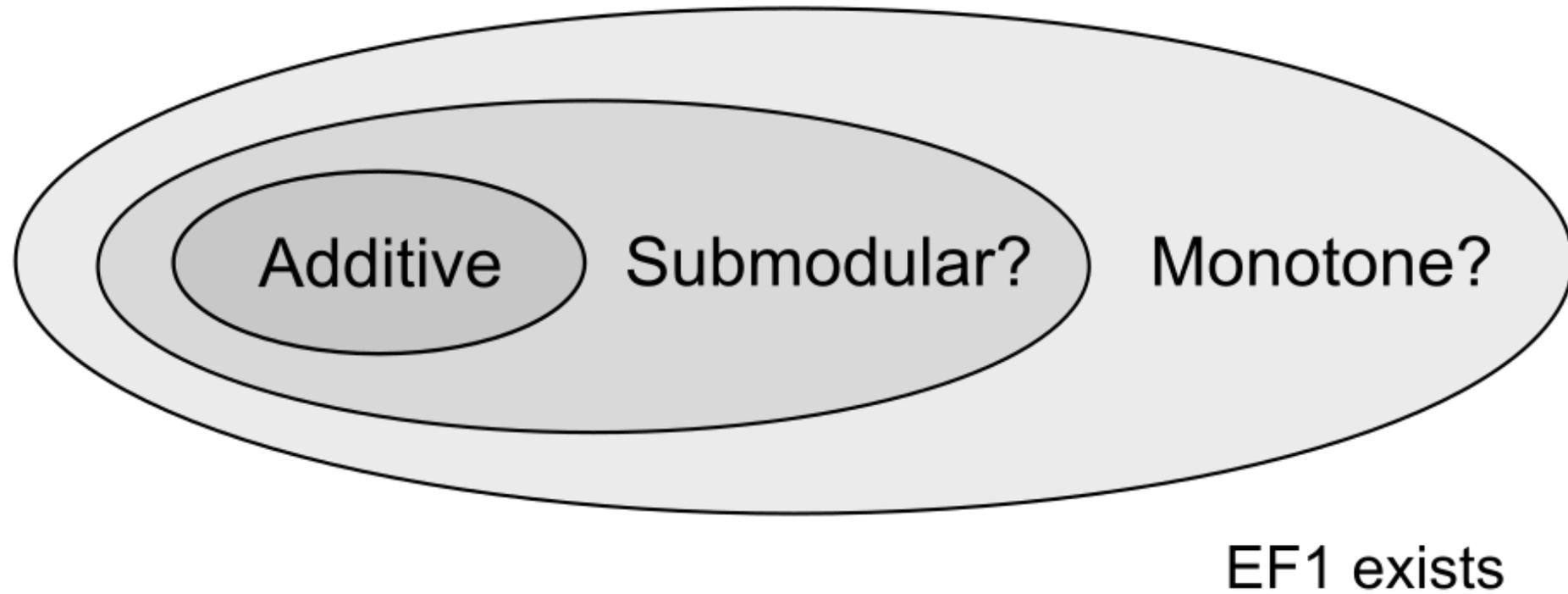
↓  
EF1



Ex-ante EF and ex-post EF1 for additive goods

Ex-ante EF and ex-post EF1 for **additive** goods

Ex-ante EF and ex-post EF1 for **additive** goods



[Lipton, Markakis, Mossel, and Saberi, *EC* 2004]

Ex-ante EF and ex-post **EF1** for additive goods

Envy-free up to *any* good (EFX)?



Ex-ante EF and ex-post **EF1** for additive goods

Envy-free up to *any* good (EFX)?



Ex-post EFX alone unresolved for 4+ agents

Ex-ante EF and ex-post **EF1** for additive goods

and Pareto optimal (PO)?



Ex-ante EF and ex-post **EF1** for additive goods

and Pareto optimal (PO)?

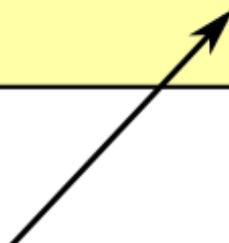


Ex-post EF1 + PO alone always exists for additive valuations

[Caragiannis, Kurokawa, Moulin, Procaccia, Shah, and Wang, *EC* 2016, *TEAC* 2019]

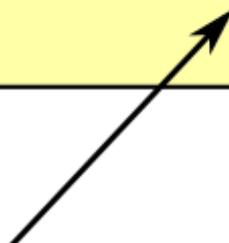
Ex-ante EF and ex-post EF1 for additive **goods**

and chores?



Ex-ante EF and ex-post EF1 for additive **goods**

and chores?



EF1 alone always exists for monotone chores

[Bhaskar, Sricharan, and Vaish, *APPROX* 2021]



# Quiz

# Quiz

Find an ex-ante EF and ex-post EF1 allocation.

	(A)	(B)	(C)	(D)
	4	3	1	1
	5	2	1	1
	0	1	2	3

