

Fair Allocation Problems

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From Welfare Maximisation to Fairness

The goal in designing auctions for markets is optimising a **social welfare** function (e.g., the sum of the agents' valuations)

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Example:

- We have two agents $\{1,2\}$ and two goods $\{g_1, g_2\}$ that can be assigned to the agents, and the valuations of the agents for the goods are:
 $v_1(g_1) = 3, v_1(g_2) = 2, v_2(g_1) = 0.5, v_2(g_2) = 1.$
- If we aim at maximising the total utility, we should assign all the goods to agent 1.
- However, agent 2 will desire at least one piece, that is, agent 2 will be **envious** of agent 1.
- A **fairer** allocation is obtained by assigning good g_1 to agent 1 and good g_2 to agent 2.

From Welfare Maximisation to Fairness

In this lecture, we will focus on fairness aspects of markets/allocations.

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Our new question/problem is the following:

How to fairly allocate (divisible or indivisible) goods among people?

Fair Cake Cutting Problem



- A set of n agents $N = 1, 2, \dots, n$.



- A cake (or divisible good) $C = [0, 1]$

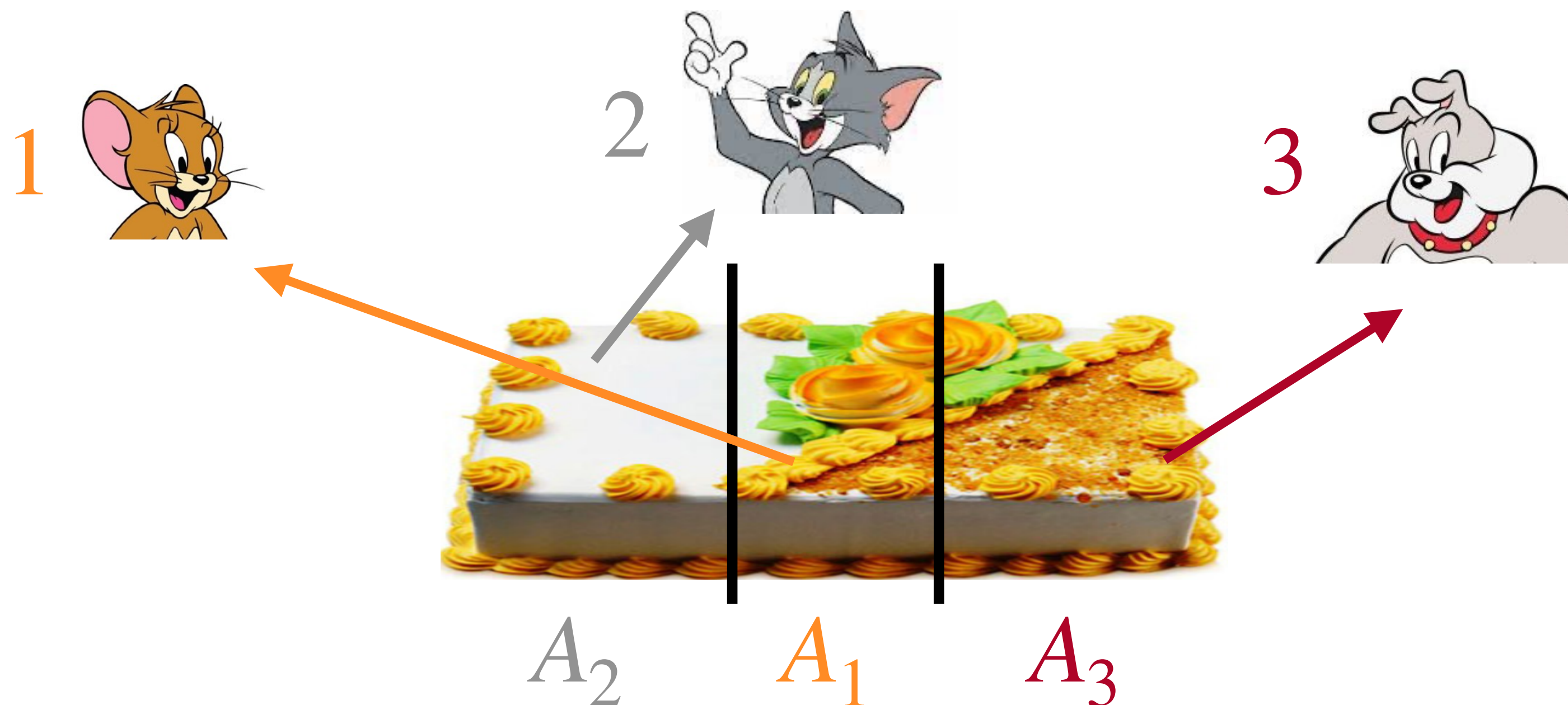


Fair Cake Cutting Problem

- A set of n agents $N = 1, 2, \dots, n$. 
- A cake (or divisible good) $C = [0, 1]$ 
- Each agent has valuation function v_i , that assigns real value $v_i(P)$ to each piece/interval $P = [a, b] \subseteq [0, 1]$.
- Valuations are **monotone**: $0 = v_i(\emptyset) \leq v_i(A) \leq v_i(B)$, if $A \subseteq B$
- Valuations are **continuous**: small variations of the pieces imply small variations of the values.

Fair Cake Cutting Problem

- A **connected division** of the cake is a partition $\mathcal{A} = \{A_1, \dots, A_n\}$ of C in n pieces/intervals, and each piece A_i is assigned to a distinct agent i .
- A connected division \mathcal{A} is **envy-free** if $v_i(A_i) \geq v_i(A_j)$ for any agents i, j .



Real-life Applications

- The "cake" is only a metaphor.
- Procedures for fair cake-cutting can be used to divide various kinds of resources, such as:
 - land estates;
 - advertisement space;
 - broadcast time.

A look into the past

- The prototypical procedure for fair cake-cutting is **Divide (or Cut) and Choose**.
- The Divide and Choose Protocol is mentioned already in the book of **Genesis**.
- During World War II, **Hugo Steinhaus**, who was hiding from the Nazis, occupied himself with the question of how to divide resources fairly.
- Inspired by the “Divide and Choose” procedure for dividing a cake between two brothers, he asked his students (**Stefan Banach** and **Bronisław Knaster**), to find a procedure that can work for any number of people...they obtained the following seminal work:
 - **Steinhaus, Hugo (1948). "The problem of fair division". *Econometrica*. 16 (1): 101–4.**

Envy-free divisions

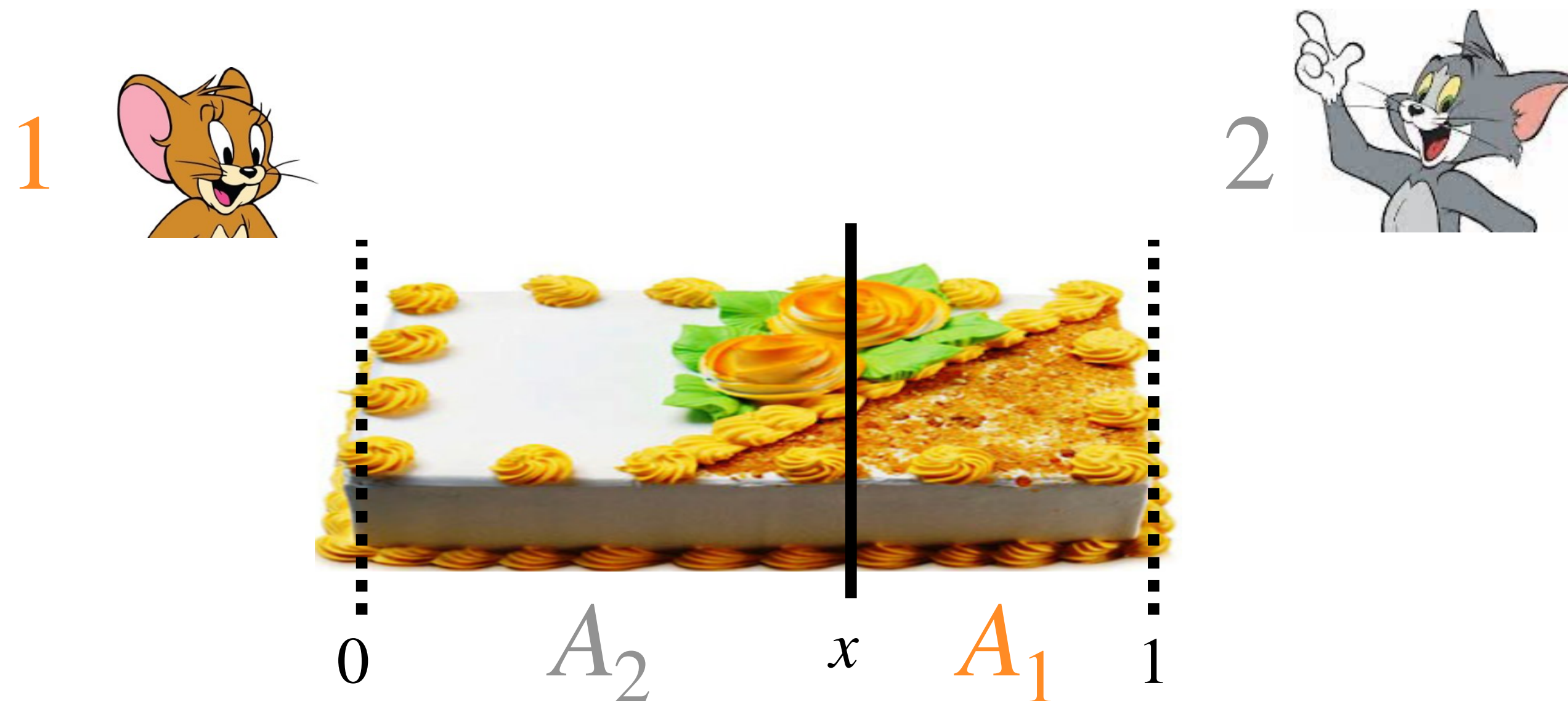
- Does a connected envy-free division always exist?

Envy-free divisions

- Does a connected envy-free division always exist? **YES.**
- For 2 agents case: **Cut and Choose Protocol** (we will see it soon)
- For 3 agents: **Stromquist moving-knives procedure**
 - Brams, Steven J.; Taylor, Alan D. (1996). Fair division: from cake-cutting to dispute resolution. Cambridge University Press.
- For more agents: A non-constructive proof based on **Sperner's Lemma**
 - Francis Edward Su. Rental harmony: Sperner's lemma in fair division. Amer. Math. Monthly, 106(10):930–942, 1999.)
 - Se also the seminal work: Steinhaus, Hugo (1948). "The problem of fair division". Econometrica. 16 (1): 101–4.
- Other protocols if we allow that each agent can get **disconnected pieces**
 - Felix Brandt, Vincent Conitzer, Ulle Endriss, Jérôme Lang, Ariel D. Procaccia: Handbook of Computational Social Choice. Cambridge University Press 2016, ISBN 9781107446984)

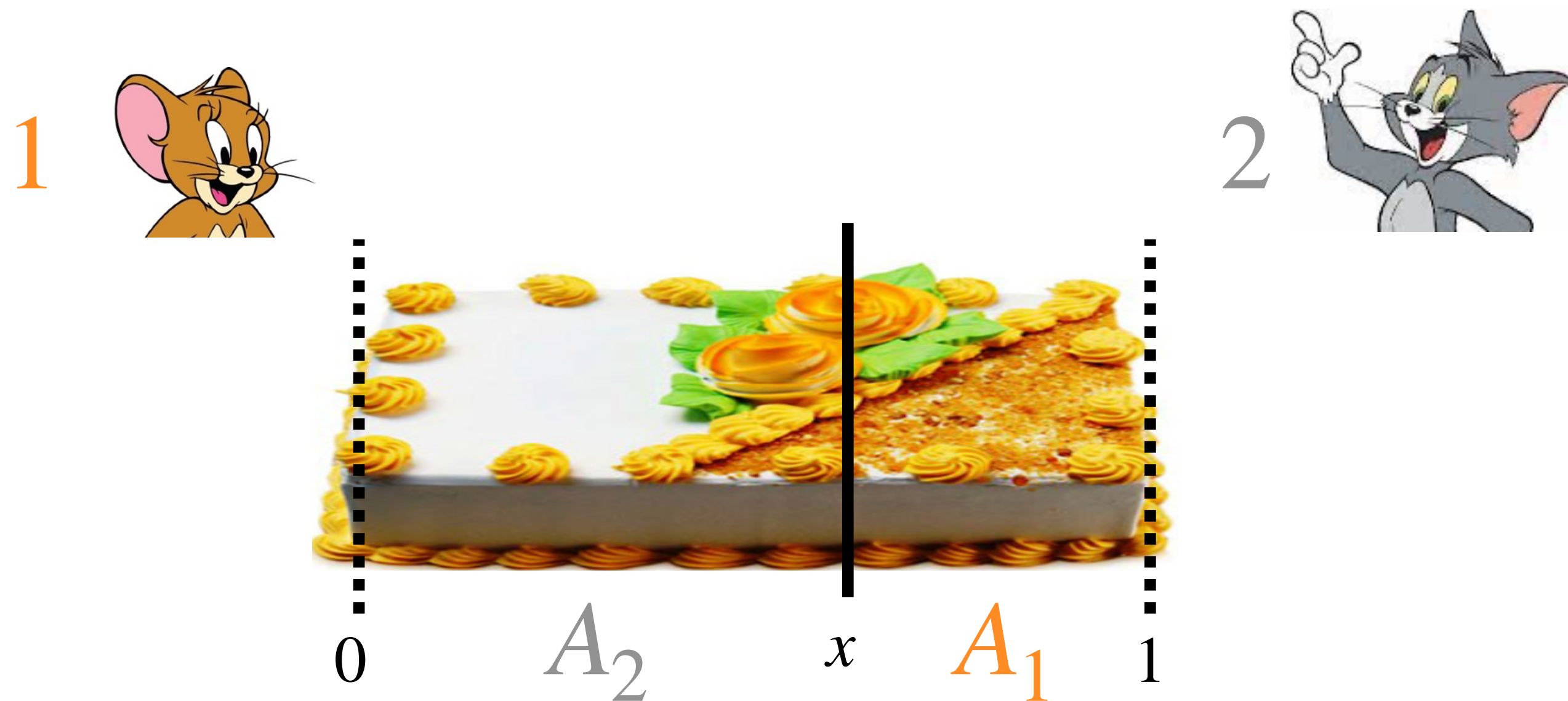
Cut and Choose Protocol (2 agents)

1. **Agent 1 cuts:** Let $x \in [0,1]$ be the first point such that $v_1([0,x]) = v_1([x,1])$
- 2.



Cut and Choose Protocol (2 agents)

1. **Agent 1 cuts:** Let $x \in [0,1]$ be the first point such that $v_1([0,x]) = v_1([x,1])$
2. **Agent 2 chooses:** Agent 2 takes the best pieces among $A = [0,x]$ and $B = [x,1]$ (i.e., the piece $P \in \{A, B\}$ maximising $v_2(P)$).



Cut and Choose Protocol (2 agents)

Example

- $v_1([a, b]) = b^2 - a^2$, $v_2([a, b]) = b - a$
- Agent 1 cuts in $x = 1/\sqrt{2} \approx 0.7$
- Agent 2 chooses piece/interval $A = [0, 1/\sqrt{2}]$

From divisible to indivisible items

- Assume that the cake is made of some indivisible pieces.
- Does a fair allocation always exist?

From divisible to indivisible items

- Assume that the cake is made of some indivisible pieces.
- Does a fair allocation always exist? **NO**

Example (in which envy cannot be avoided)

- We have three (indivisible) cookies C_1, C_2, C_3 .
- Two agents having the following valuations:
$$1 = v_1(A_1) = v_1(A_2) = v_1(A_3) = v_2(A_1) = v_2(A_2) = v_2(A_3)$$

Fair Allocation of Indivisible Goods

- A set of n agents $N = 1, 2, \dots, n$.
- A set of m indivisible goods/items $M = \{g_1, \dots, g_m\}$.
- The valuation of each agent i over bundles/subsets of goods is monotone:
 $0 = v_i(\emptyset) \leq v_i(A) \leq v_i(B)$ if $A \subseteq B$.
- The valuations are additive if $v_i(A) = \sum_{g_h \in A} v_{i,h}$ for any bundle $A \subseteq M$, where
 $v_{i,h} := v_i(g_h)$ is the valuation of agent i for good g_h .
- The valuations are identical if $v_i(A) = v_j(A) =: v(A)$ holds for any $A \subseteq M$ and agents i, j .

Envy-free Allocations

- An **allocation** is a partition $\mathcal{A} = (A_1, \dots, A_n)$ of all the goods, where A_i is the **bundle/subset** assigned to agent i .
- An allocation \mathcal{A} is **envy-free** if $v_i(A_i) \geq v_i(A_j)$ for any agents i, j .

Real-life Applications

- By resorting to the “indivisible setting”, we can model several real-life scenarios. For instance:
 - Several heirs want to divide the inherited property, which contains e.g. a house, a car, a piano and several paintings.
 - Several lecturers want to divide the courses given in their faculty, and each lecturer can teach one or more whole courses.

Envy-free Allocations (cont)

- An **allocation** is a partition $\mathcal{A} = (A_1, \dots, A_n)$ of all the goods, where A_i is the **bundle/subset** assigned to agent i .
- An allocation \mathcal{A} is **envy-free** if $v_i(A_i) \geq v_i(A_j)$ for any agents i, j .
- We already know that envy-free allocations cannot be guaranteed when dealing with indivisible items.
- Can we hope for some weaker notion of fairness?

Relaxed Notions of Envy-Freeness

An allocation $\mathcal{A} = (A_1, \dots, A_n)$ is:

- **Envy-free allocation-up-to-1-good (EF1)** if, for any agents i, j , there exists a good $g_h \in A_j$ such that $v_i(A_i) \geq v_i(A_j \setminus g_h)$.
- **Envy-free allocation-up-to-any-good (EFX)**: if, for any agents i, j , $v_i(A_i) \geq v_i(A_j \setminus g_h)$ holds for any $g_h \in A_j$.
- **α -approximate EFX (α -EFX)**: if, there exists $\alpha \in [0, 1]$ such that, for any agents i, j , $v_i(A_i) \geq \alpha \cdot v_i(A_j \setminus g_h)$ holds for any $g_h \in A_j$.

Relaxed Notions of Envy-Freeness

- Other relaxed notions of fairness are (exact and approximate) proportional allocations, maximin fair allocations, etc...
- Refer to Brandt et al. 2016 for a detailed discussion of fair allocation problems.

Felix Brandt, Vincent Conitzer, Ulle Endriss, Jérôme Lang, Ariel D. Procaccia: Handbook of Computational Social Choice. Cambridge University Press 2016, ISBN 9781107446984

EF1 allocations: existence and computation

- Does an EF1 allocation always exist?

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- Does an EF1 allocation always exist? **YES**
- Can an EF1 allocation be efficiently computed?

EF1 allocations: existence and computation

- Does an EF1 allocation always exist? **YES**
- Can an EF1 allocation be efficiently computed? **YES**

We will consider the following two algorithms to compute EF1 allocations (and implicitly, to show their existence):

- **Round-Robin Algorithm** (for additive valuations)
- **Envy-Cycle Elimination Algorithm** (for monotone valuations)

For further details, see: Richard J. Lipton, Evangelos Markakis, Elchanan Mossel, Amin Saberi: On approximately fair allocations of indivisible goods. EC 2004: 125-131

Round-Robin Algorithm




- Consider an arbitrary order of the agents (we consider w.l.o.g. the increasing order w.r.t. the agents' indexes).
- For $k=1,2,\dots,m$
 - Agent $i=k \bmod n$ takes her most valuable good, among all the unallocated good.
- Return the obtained allocation

Each time agent 1 takes a new good,
a new round starts

(thus, there are $\lceil m/n \rceil$ rounds)

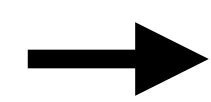
Round-Robin Algorithm (Example)













	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Initial instance

Round-Robin Algorithm (Example)

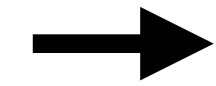












							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Round $s=1$

Iteration $k=1$











Round-Robin Algorithm (Example)



							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

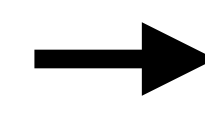
Round $s=1$ Iteration $k=2$











Round-Robin Algorithm (Example)

							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Round $s=1$ Iteration $k=3$

Round-Robin Algorithm (Example)














							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Round $s=2$

Iteration $k=4$

Round-Robin Algorithm (Example)






							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Round $s=2$

Iteration $k=5$

Round-Robin Algorithm (Example)



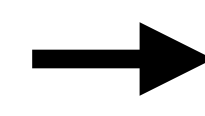
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5













Round $s=2$

Iteration $k=6$

Round-Robin Algorithm (Example)






							
	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	13
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

Round $s=3$

Iteration $k=7$

Round-Robin Algorithm (Example)



	g_1	g_2	g_3	g_4	g_5	g_6	g_7
	3	10	12	1	5	4	20
	40	9	10	30	8	4	50
	1	0	0	3	1	0	5

$A_1 = \{g_3, g_6, g_7\}, v_1(A_1) = 36$

$A_2 = \{g_1, g_2\}, v_2(A_2) = 49$

$A_3 = \{g_4, g_5\}, v_3(A_3) = 4$

Round-Robin Algorithm

Theorem: If valuations are additive, Round-Robin returns an EF1-allocation

Proof:

- For simplicity, we add dummy items of null value so that n divides m .
- Let $\mathcal{A} = \{A_1, \dots, A_n\}$ be the allocation returned by Round-Robin.
- Let $g^{i,s}$ denote the s -th good picked by each agent i (i.e., at round s), so that $A_i = \{g^{i,1}, g^{i,2}, \dots, g^{i,m/n}\}$ for any agent i .

Round-Robin Algorithm

Proof (cont):

For any fixed agent i , we have two cases: $v_i(g^{i,s}) \geq v_i(g^{j,s})$ for any round s (see next table)

1. For $j > i$, we have:
$$v_i(A_i) = \sum_{s=1}^{m/n} v_i(g^{i,s}) \geq \sum_{s=1}^{m/n} v_i(g^{j,s}) = v_i(A_j).$$

2.

Round-Robin Algorithm

A_1	...	A_i	...	A_j	...	A_n
$g^{\{1,1\}}$	$g^{\{i,1\}}$...	$g^{\{j,1\}}$...	$g^{\{n,1\}}$
$g^{\{1,2\}}$...	$g^{\{i,2\}}$...	$g^{\{j,2\}}$...	$g^{\{n,2\}}$
...
$g^{\{1,s-1\}}$...	$g^{\{i,s-1\}}$...	$g^{\{j,s-1\}}$...	$g^{\{n,s-1\}}$
$g^{\{1,s\}}$...	$g^{\{i,s\}}$...	$g^{\{j,s\}}$...	$g^{\{n,s\}}$
$g^{\{1,s+1\}}$...	$g^{\{i,s+1\}}$...	$g^{\{j,s+1\}}$...	$g^{\{n,s+1\}}$
...
$g^{\{1,m/n\}}$...	$g^{\{i,m/n\}}$...	$g^{\{j,m/n\}}$...	$g^{\{n,m/n\}}$

An arrow from a to b means that $a \geq b$

Round-Robin Algorithm

Proof (cont):

For any fixed agent i , we have two cases: $v_i(g^{i,s}) \geq v_i(g^{j,s})$ for any round s (see next table)

1. For $j > i$, we have: $v_i(A_i) = \sum_{s=1}^{m/n} v_i(g^{i,s}) \geq \sum_{s=1}^{m/n} v_i(g^{j,s}) = v_i(A_j).$

2. For $j < i$, we have: $v_i(A_i) \geq \sum_{s=1}^{m/n-1} v_i(g^{i,s}) \geq \sum_{s=2}^{m/n} v_i(g^{j,s}) = v_i(A_j) - v_i(g^{j,1})$

$v_i(g^{i,s}) \geq v_i(g^{j,s+1})$ for any round s (see next table)

Round-Robin Algorithm

A_1	...	A_j	...	A_i	...	A_n
$g^{\{1,1\}}$	$g^{\{j,1\}}$...	$g^{\{i,1\}}$...	$g^{\{n,1\}}$
$g^{\{1,2\}}$...	$g^{\{j,2\}}$...	$g^{\{i,2\}}$...	$g^{\{n,2\}}$
...
$g^{\{1,s-1\}}$...	$g^{\{j,s-1\}}$...	$g^{\{i,s-1\}}$...	$g^{\{n,s-1\}}$
$g^{\{1,s\}}$...	$g^{\{j,s\}}$...	$g^{\{i,s\}}$...	$g^{\{n,s\}}$
$g^{\{1,s+1\}}$...	$g^{\{j,s+1\}}$...	$g^{\{i,s+1\}}$...	$g^{\{n,s+1\}}$
...
$g^{\{1,m/n\}}$...	$g^{\{j,m/n\}}$...	$g^{\{i,m/n\}}$...	$g^{\{n,m/n\}}$

An arrow from a to b means that $a \geq b$

Round-Robin Algorithm

Proof (cont):

For any fixed agent i , we have two cases:

1. For $j > i$, we have:
$$v_i(A_i) = \sum_{s=1}^{m/n} v_i(g^{i,s}) \geq \sum_{s=1}^{m/n} v_i(g^{j,s}) = v_i(A_j).$$
2. For $j < i$, we have:
$$v_i(A_i) \geq \sum_{s=1}^{m/n-1} v_i(g^{i,s}) \geq \sum_{s=2}^{m/n} v_i(g^{j,s}) = v_i(A_j) - v_i(g^{j,1})$$

We conclude that \mathcal{A} is an **EF1 allocation**.

Q.E.D.

Envy-Cycle Elimination Algorithm

Envy-graph and Envy-Cycle-Elimination Subroutine

Given a **partial allocation** $\mathcal{A} = (A_1, \dots, A_n)$ (i.e., there are some unallocated goods), the **envy-graph** associated to \mathcal{A} is a graph $G = (V, E)$ defined as follows:

- V is the set of agents;
- we put a directed edge (u, z) in E iff agent u is envious of z under \mathcal{A} (that is, $v_u(A_u) < v_u(A_z)$).

Envy-Cycle Elimination Algorithm

Envy-graph (cont)

- If there is a cycle $C = (u_1, u_2), (u_2, u_3), \dots, (u_{t-1}, u_1)$ in G , we assign the bundle of u_2 to u_1 , that of u_3 to u_2 ... that of u_1 to u_{t-1} .




Envy-Cycle Elimination Algorithm

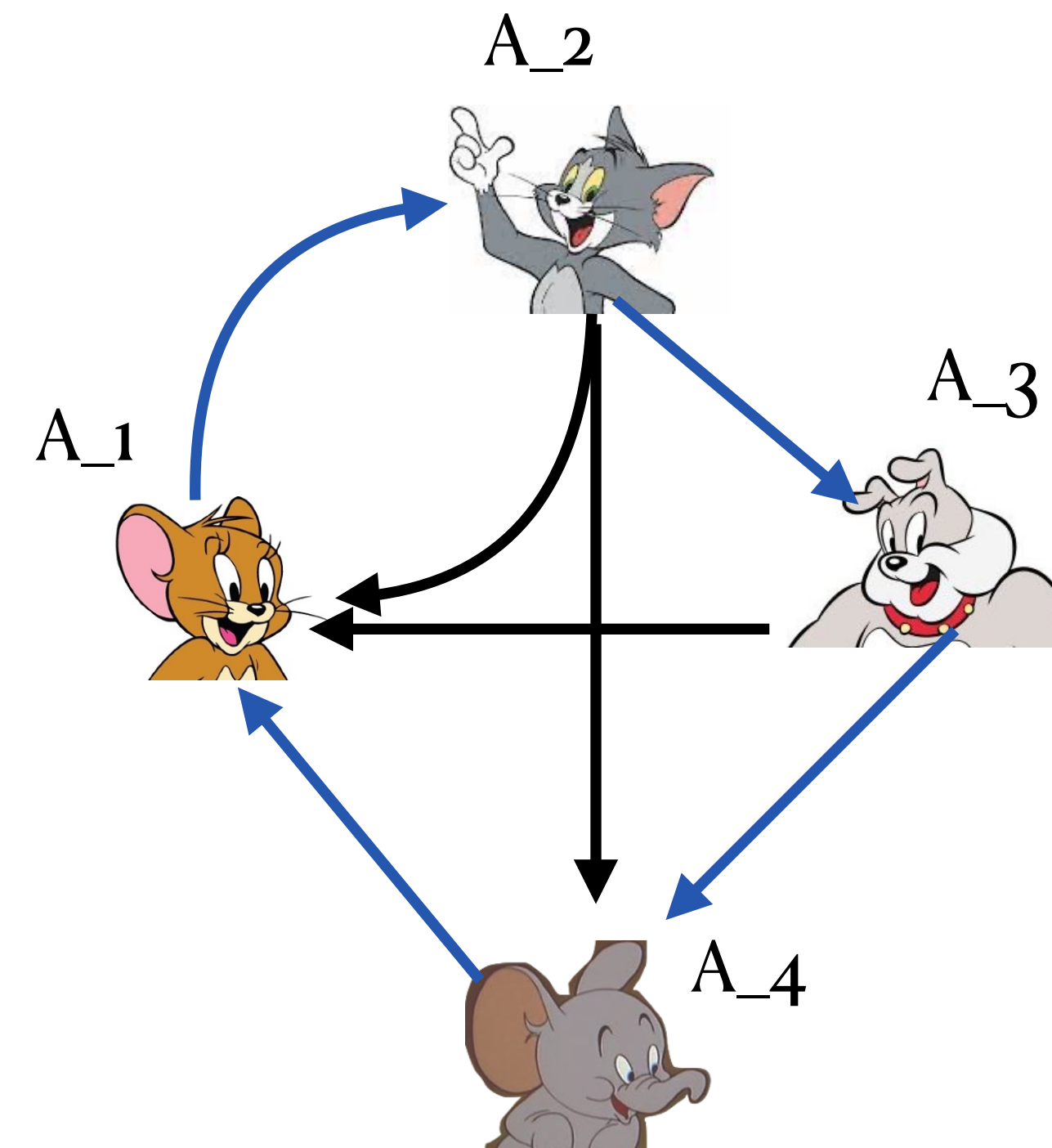
Envy-graph (cont)



- If there is a cycle $C = (u_1, u_2), (u_2, u_3), \dots, (u_{t-1}, u_1)$ in G , we assign the bundle of u_2 to u_1 , that of u_3 to u_2 ... that of u_1 to u_{t-1} . After this reallocation, the following facts hold:
 - the envy-graph $G' = (V, E')$ associated with the new partial allocation \mathcal{A}' does not contain cycle C ;
 - new edges are not included (either we delete edges or “adjust” existing edges).
- We conclude that $|E'| < |E|$. By iterating such process (Q: how many times?) we necessarily reach a partial allocation \mathcal{A}^* whose envy-graph G^* is **acyclic**.
- The above procedure is called **Envy-Cycle-Elimination Subroutine**.

Envy-Cycle Elimination Algorithm

Example (Envy-Cycle-Elimination Subroutine)





	$v(A_1)$	$v(A_2)$	$v(A_3)$	$v(A_4)$
	20	24	5	10
	30	0	35	10
	70	0	40	60
	20	3	10	11

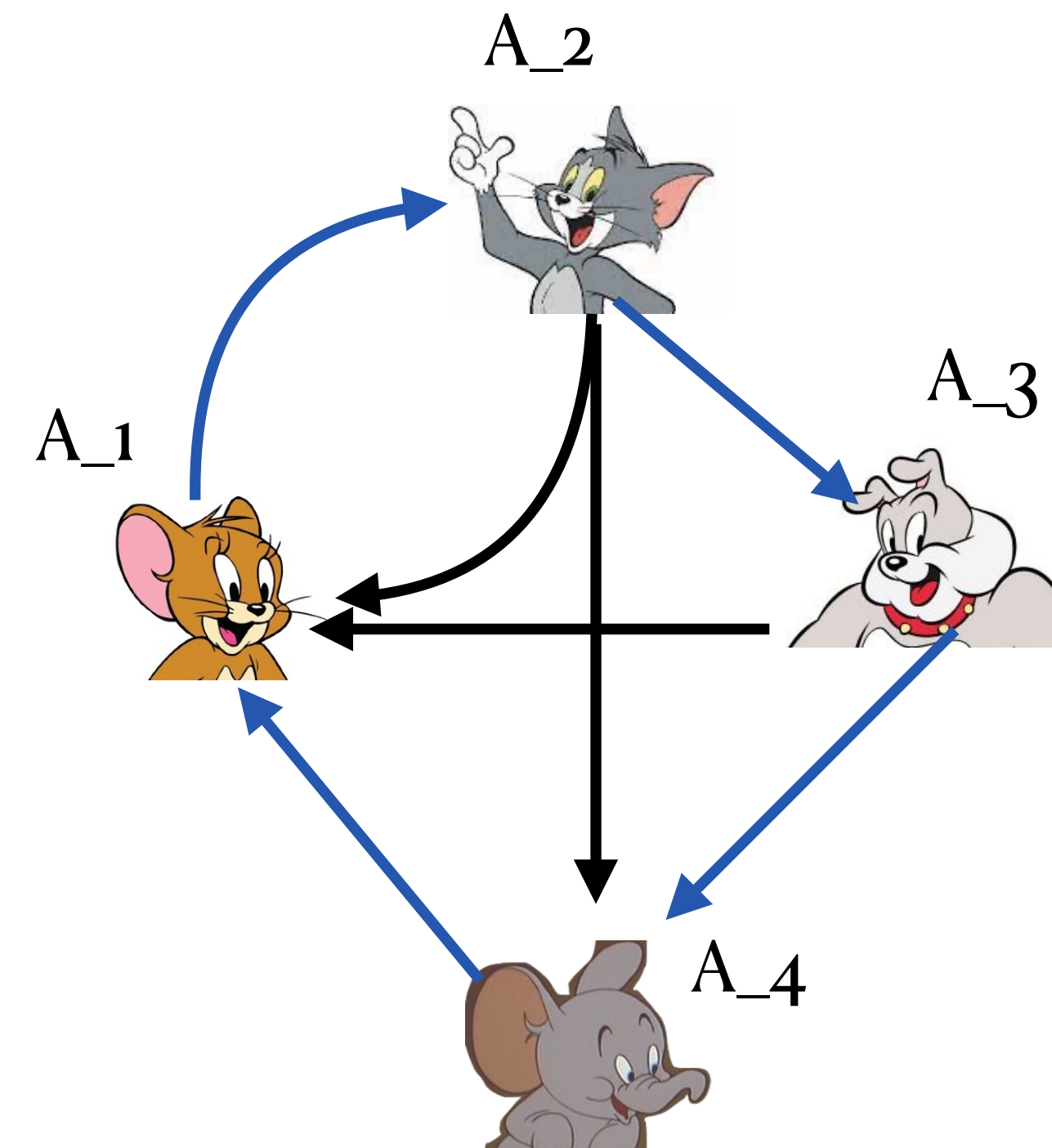


We initially consider the allocation assigning:
 A_1 to , A_2 to , A_3 to , A_4 to 

Envy-Cycle Elimination Algorithm

Example (Envy-Cycle-Elimination Subroutine)





	$v(A_1)$	$v(A_2)$	$v(A_3)$	$v(A_4)$
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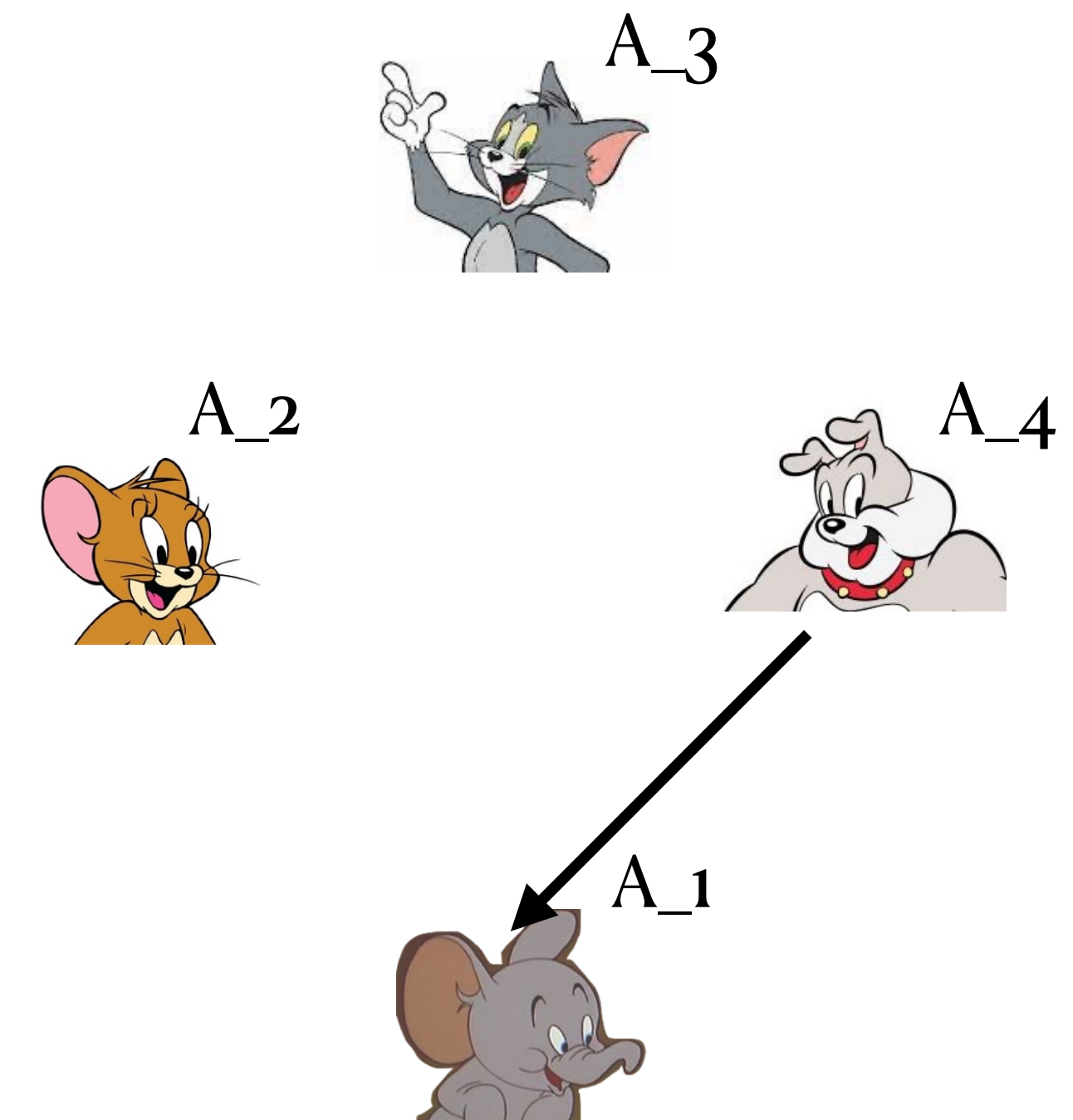


There is (at least) one cycle (the blue one)
in the envy-graph

Envy-Cycle Elimination Algorithm

Example (Envy-Cycle-Elimination Subroutine)





	$v(A_1)$	$v(A_2)$	$v(A_3)$	$v(A_4)$
	20	24	5	10
	30	0	35	10
	70	0	40	60
	20	3	10	11

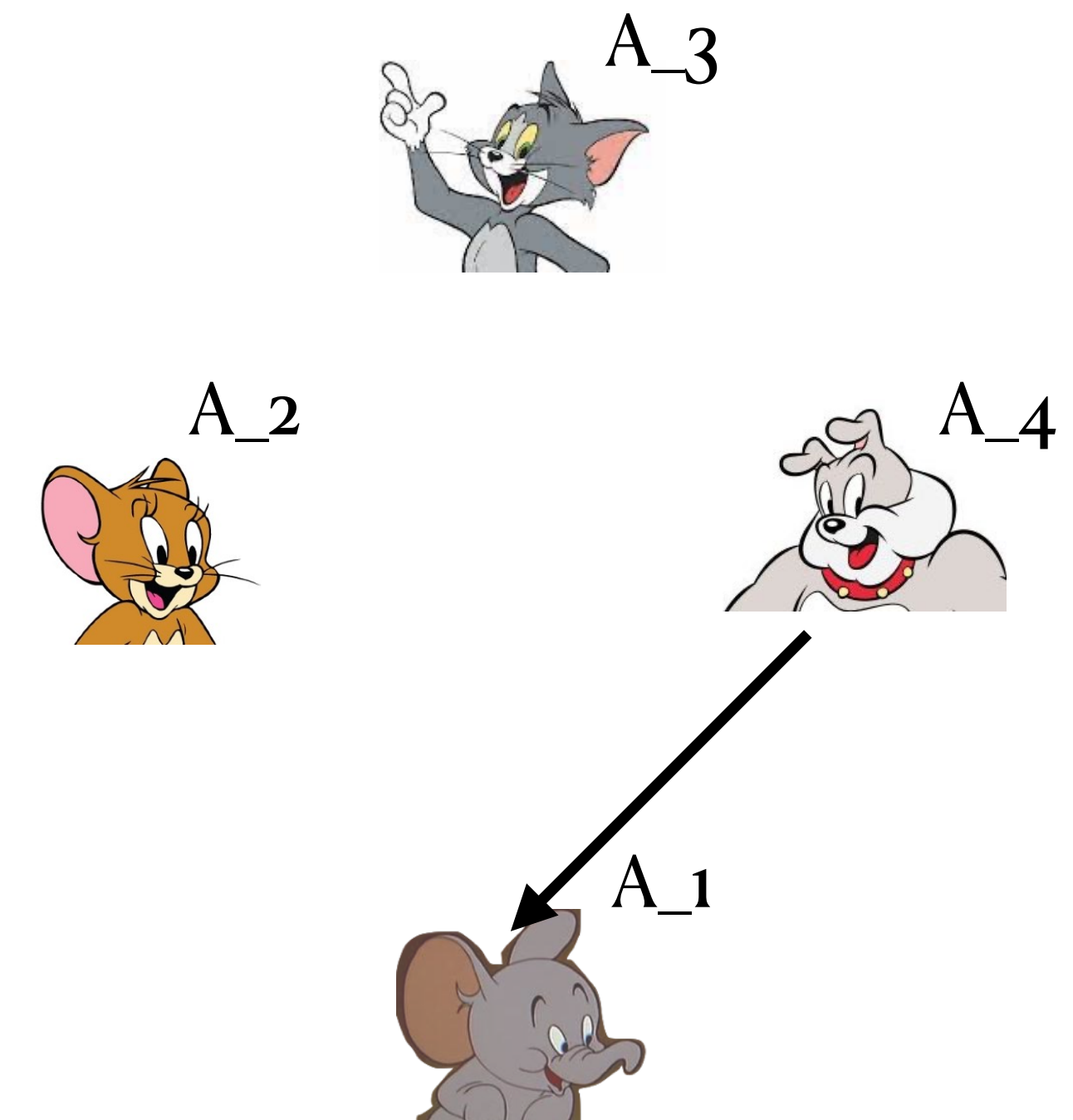


We construct a new envy-graph
(by reallocating the bundles in the cycle)

Envy-Cycle Elimination Algorithm

Example (Envy-Cycle-Elimination Subroutine)

	$v(A_1)$	$v(A_2)$	$v(A_3)$	$v(A_4)$
	20	24	5	10
	30	0	35	10
	70	0	40	60
	20	3	10	11



The new graph is acyclic :)
(otherwise, we would have had to repeat the procedure)

Envy-Cycle Elimination Algorithm

We are ready to present the **Envy-Cycle Elimination Algorithm**:

1. Start with an empty partial allocation $\mathcal{A} = (A_1, \dots, A_n)$ (i.e., $A_i = \emptyset$)
2. Consider an arbitrary order of the goods (e.g., g_1, g_2, \dots, g_m)
3. For $k = 1, 2, \dots, m$ {
 - A. $\mathcal{A} \leftarrow$ Partial all. returned by Envy-cycle Elimination Subroutine applied to \mathcal{A}
 - B. let i be an agent that is not envied by anyone (Q: why does such agent always exist? See the above example for a hint)
 - C. assign g_k to i , and let \mathcal{A} be the resulting partial allocation}
4. Return \mathcal{A}

Envy-Cycle Elimination Algorithm

Theorem: The Envy-Cycle Elimination Algorithm always returns an EF1 allocation (even, for general monotone valuation).

Proof: We will inductively show that the partial allocation \mathcal{A}_k obtained at the end of each iteration $k=0,1,\dots,m$ ($k=0$ refers to the initialisation) is **EF1**.

- **k=0:** Trivial (the empty allocation is trivially EF1).

Envy-Cycle Elimination Algorithm

Proof (cont):

- **k=1:**
 - We know that \mathcal{A}_0 is EF1 (from the previous step).
 - By applying the envy-cycle elimination subroutine to \mathcal{A}_0 we get again an EF1 allocation (indeed, “new envy is not created” and “old envy could be removed”).
 - Item g_1 is assigned to an agent that is not envied (in the strong sense) by anyone. Thus, after receiving g_1 , other agents, in the worst-case, can be “envious of that agent up to item g_1 ”.
 - Thus, the new partial allocation \mathcal{A}_1 is EF1.

Envy-Cycle Elimination Algorithm

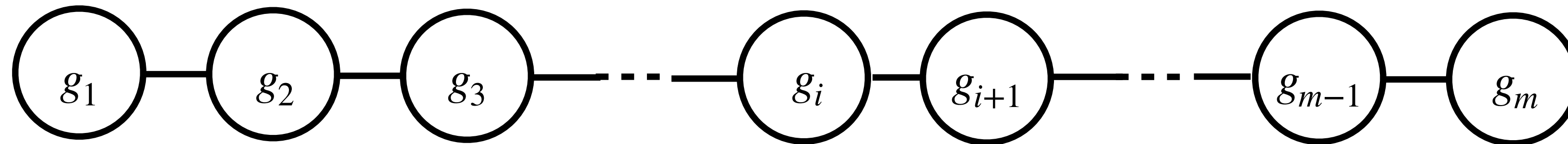
Proof (cont):

- **Any k:**
 - Analogously, by using that \mathcal{A}_1 is EF1, we can show that the partial allocation \mathcal{A}_2 obtained from \mathcal{A}_1 is EF1, and so on...
 - More formally, we can show the claim by **induction**: we assume that \mathcal{A}_{k-1} is EF1 by the inductive hypothesis, and by using the same proof arguments of case $k=1$ we show that partial allocation \mathcal{A}_k is EF1.

We conclude that \mathcal{A}_k is EF1 for any k , and even more so the final allocation \mathcal{A} (i.e., \mathcal{A}_m) is **EF1**. Q.E.D.

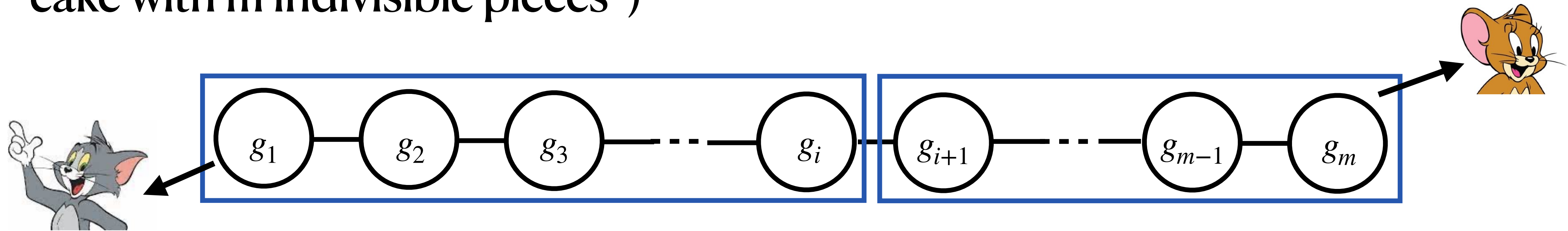
Exercise: Connected EF1 with 2 agents

- We have 2 agents, and assume that the goods are organised as a “path” (that is, a “cake with m indivisible pieces”)



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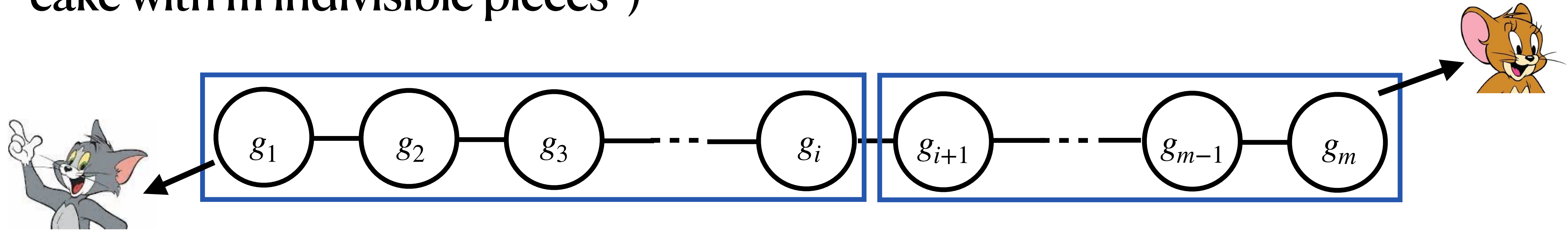
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- You are asked to partition the goods into 2 bundles having the structure of “sub-paths” (that is, each bundle must be of type $\{g_1, \dots, g_i\}$ or $\{g_{i+1}, \dots, g_m\}$ for some i), and assign each bundle to a distinct agent.

Exercise: Connected EF1 with 2 agents

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- You are asked to partition the goods into 2 bundles having the structure of “sub-paths” (that is, each bundle must be of type $\{g_1, \dots, g_i\}$ or $\{g_{i+1}, \dots, g_m\}$ for some i), and assign each bundle to a distinct agent.
- The resulting allocation must be EF1. (Hint: Provide a discrete version of the Cut and Choose protocol)

EFX allocations: existence and computation

- From now on, we implicitly focus on additive valuations.
- Does an EFX allocation always exist?
 - We don't know (the conjecture is YES)
 - This is one of the most important open problems in fair allocation.

For further details, see:

- Ioannis Caragiannis, David Kurokawa, Hervé Moulin, Ariel D. Procaccia, Nisarg Shah, Junxing Wang: The Unreasonable Fairness of Maximum Nash Welfare. ACM Trans. Economics and Comput. 7(3): 12:1-12:32 (2019)
- Benjamin Plaut, Tim Roughgarden: Almost Envy-Freeness with General Valuations. SODA 2018: 2584-2603


EFX allocations: existence and computation

- There are some interesting cases in which they always exist?

EFX allocations: existence and computation

- There are some interesting cases for which EF1 alloc. always exist? **YES.**
 - agents having the same **ranking** on the goods, despite possibly different values: by using a **variant of cycle-elimination** (we will see it)
 - 2 agents: discrete variant of **cut and choose** (see: Benjamin Plaut, Tim Roughgarden: Almost Envy-Freeness with General Valuations. SODA 2018: 2584-2603)
 - 3 agents (see: Bhaskar Ray Chaudhury, Jugal Garg, Kurt Mehlhorn: EFX Exists for Three Agents. EC 2020: 1-19)
 - and so on...

Revisited Envy-cycle Elimination Algorithm

1. Start with an empty partial allocation $\mathcal{A} = (A_1, \dots, A_n)$ (i.e., $A_i = \emptyset$)
2. While there exists an unallocated good {
 - A. $\mathcal{A} \leftarrow$ Envy-cycle Elimination applied to \mathcal{A}
 - B. let i be an agent that is not envied by anyone
 -  Novelty C. let g the most valuable good for i among all the unallocated ones, and assign g to i
 - D. set \mathcal{A} equal to the resulting allocation }
3. Return \mathcal{A}

Revisited Envy-cycle Elimination Algorithm

Theorem: If the agents have the same ranking on the goods, the Revisited Envy-Cycle Elimination Algorithm always returns an EFX allocation.

Proof (sketch): Reconsider the proof for EF1.

- Some agent i becomes “envious because of 1 good” only if she becomes “envious because of the last good g ” added to some bundle A .
- Such good g is necessarily the least valuable among all the goods of bundle A .
- Thus, agent i is “envious because of the least valuable good of A ”, that is, she is “not-envious-up-to-any-good”.

We conclude that, at each iteration, the reached partial allocation is EFX. Q.E.D.

Revisited Envy-cycle Elimination Algorithm

What would we have achieved by applying the revisited Envy-cycle algorithm to general instances?

Theorem: The Revisited Envy-Cycle Elimination Algorithm always returns a $1/2$ -EFX allocation.

The proof follows the same high-level structure of the previous one, but it is more sophisticated.