

# Game Theory

## 8. Mechanism Design

### 8.1. Introduction and Example

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**Objective:** Design the rules of the game such that desirable behavior is dominant behavior.

**Here:** desirable  $\approx$

- truthful about one's own preferences +
- contributing to maximizing social welfare

## Model:

- Strict linear orders  $\prec$  contain no information about “by how much” one alternative is preferred.
- Idea: Instead, use money to measure this.
- Use money also for transfers between players “for compensation”.

## Given:

- set of **alternatives**  $A$
- set of  $n$  **players**  $N$
- **valuation functions**  $v_i : A \rightarrow \mathbb{R}$  such that  $v_i(a)$  denotes the value player  $i$  assigns to alternative  $a$

## Find:

- a **chosen alternative**  $a \in A$ .
- **payments**  $p_i \in \mathbb{R}$  to be paid by player  $i$

**Utility** of player  $i$ :  $u_i(a) = v_i(a) - p_i$ .

## Second price auctions (aka Vickrey auctions):

- There are  $n$  players **bidding** for a single item.
- Player  $i$ 's **private** valuations of item:  $w_i$ .
- **Desired outcome**: Player with highest private valuation wins bid.
- Players should reveal their valuations truthfully.
- Winner  $i$  pays price  $p^*$  and has utility  $w_i - p^*$ .
- Non-winners pay nothing and have utility 0.

# Example: Vickrey Auctions



Formally:

- $A = N$
- $v_i(a) = \begin{cases} w_i & \text{if } a = i \\ 0 & \text{else} \end{cases}$
- What about payments? Say player  $i$  wins:
  - $p^* = 0$  (winner pays nothing): bad idea, players would manipulate and publicly declare values  $w'_i \gg w_i$ .
  - $p^* = w_i$  (winner pays his valuation): bad idea, players would manipulate and publicly declare values  $w'_i = w_i - \varepsilon$ .
  - **better:**  $p^* = \max_{j \neq i} w_j$  (winner pays second highest bid).

## Definition (Vickrey Auction)

The winner of the **Vickrey Auction** (aka second price auction) is the player  $i$  with the highest declared value  $w_i$ . He has to pay the second highest declared bid  $p^* = \max_{j \neq i} w_j$ .

## Proposition (Vickrey)

Let  $i$  be one of the players and  $w_i$  his valuation for the item,  $u_i$  his utility if he truthfully declares  $w_i$  as his valuation of the item, and  $u'_i$  his utility if he falsely declares  $w'_i$  as his valuation of the item. Then  $u_i \geq u'_i$ .

## Proof

See

[http://en.wikipedia.org/wiki/Vickrey\\_auction](http://en.wikipedia.org/wiki/Vickrey_auction). □

- New preference model: with **money**.
- To ensure truthful revelation of preferences, we need the right payment functions.
- Example: Vickrey auctions.



# Game Theory

## 8. Mechanism Design

### 8.2. Incentive Compatible Mechanisms

#### 8.2.1. VCG Mechanisms

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- **Idea:** Generalization of Vickrey auctions.
- **Preferences** modeled as functions  $v_i : A \rightarrow \mathbb{R}$ .
- Let  $V_i$  be the **space of all such functions** for player  $i$ .
- Unlike for social choice functions: Not only decide about **chosen alternative**, but also about **payments**.

## Definition (mechanism)

A **mechanism**  $\langle f, p_1, \dots, p_n \rangle$  consists of

- a **social choice function**  $f : V_1 \times \dots \times V_n \rightarrow A$  and
- for each player  $i$ , a **payment function**  $p_i : V_1 \times \dots \times V_n \rightarrow \mathbb{R}$ .

## Definition (incentive compatibility)

A mechanism  $\langle f, p_1, \dots, p_n \rangle$  is called **incentive compatible** if for each player  $i = 1, \dots, n$ , for all preferences  $v_1 \in V_1, \dots, v_n \in V_n$  and for each preference  $v'_i \in V_i$ ,

$$v_i(f(v_i, v_{-i})) - p_i(v_i, v_{-i}) \geq v_i(f(v'_i, v_{-i})) - p_i(v'_i, v_{-i}).$$

- If  $\langle f, p_1, \dots, p_n \rangle$  is **incentive compatible**, **truthfully** declaring ones preference is **dominant** strategy.
- The **Vickrey-Clarke-Groves mechanism** is an incentive compatible mechanism that maximizes “social welfare”, i.e., the sum over all individual utilities  $\sum_{i=1}^n v_i(a)$ .
- **Idea:** Reflect other players' utilities in payment functions, align all players' incentives with goal of maximizing social welfare.

## Definition (Vickrey-Clarke-Groves mechanism)

A mechanism  $\langle f, p_1, \dots, p_n \rangle$  is called a **Vickrey-Clarke-Groves mechanism (VCG mechanism)** if

- 1  $f(v_1, \dots, v_n) \in \operatorname{argmax}_{a \in A} \sum_{i=1}^n v_i(a)$  for all  $v_1, \dots, v_n$  and
- 2 there are functions  $h_1, \dots, h_n$  with  $h_i : V_{-i} \rightarrow \mathbb{R}$  such that  $p_i(v_1, \dots, v_n) = h_i(v_{-i}) - \sum_{j \neq i} v_j(f(v_1, \dots, v_n))$  for all  $i = 1, \dots, n$  and  $v_1, \dots, v_n$ .

**Note:**  $h_i(v_{-i})$  independent of player  $i$ 's declared preference  $\Rightarrow h_i(v_{-i}) = c$  constant from player  $i$ 's perspective.

**Utility of player  $i$**   $= v_i(f(v_1, \dots, v_n)) + \sum_{j \neq i} v_j(f(v_1, \dots, v_n)) - c = \sum_{j=1}^n v_j(f(v_1, \dots, v_n)) - c = \text{social welfare} - c$ .

## Theorem (Vickrey-Clarke-Groves)

Every VCG mechanism is incentive compatible.

### Proof.

Let  $i$ ,  $v_{-i}$ ,  $v_i$  and  $v'_i$  be given. Show: Declaring true preference  $v_i$  dominates declaring false preference  $v'_i$ .

Let  $a = f(v_i, v_{-i})$  and  $a' = f(v'_i, v_{-i})$ .

$$\text{Utility player } i = \begin{cases} v_i(a) + \sum_{j \neq i} v_j(a) - h_i(v_{-i}) & \text{if declaring } v_i \\ v_i(a') + \sum_{j \neq i} v_j(a') - h_i(v_{-i}) & \text{if declaring } v'_i \end{cases}$$

Alternative  $a = f(v_i, v_{-i})$  maximizes social welfare

$$\Rightarrow v_i(a) + \sum_{j \neq i} v_j(a) \geq v_i(a') + \sum_{j \neq i} v_j(a').$$

$$\Rightarrow v_i(f(v_i, v_{-i})) - p_i(v_i, v_{-i}) \geq v_i(f(v'_i, v_{-i})) - p_i(v'_i, v_{-i}).$$



- New preference model: with **money**.
- VCG mechanisms generalize **Vickrey auctions**.
- **VCG mechanisms** are **incentive compatible** mechanisms maximizing social welfare.

# Game Theory

## 8. Mechanism Design

### 8.2. Incentive Compatible Mechanisms

#### 8.2.2. Clarke Pivot Functions

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- So far: functions  $h_i$  unspecified  
     $\rightsquigarrow$  payment functions  $p_i$  only partially specified
- One possibility:  $h_i(v_{-i}) = 0$  for all  $h_i$  and  $v_{-i}$   
    Drawback: too much money distributed among players  
    (more than necessary)
- Further requirements:
  - Players should pay at most as much as they value the outcome.
  - Players should only pay, never receive money.

## Definition (individual rationality)

A mechanism is **individually rational** if all players always get a nonnegative utility, i.e., if for all  $i = 1, \dots, n$  and all  $v_1, \dots, v_n$ ,

$$v_i(f(v_1, \dots, v_n)) - p_i(v_1, \dots, v_n) \geq 0.$$

## Definition (positive transfers)

A mechanism has **no positive transfers** if no player is ever paid money, i.e., if for all  $i = 1, \dots, n$  and all  $v_1, \dots, v_n$ ,

$$p_i(v_1, \dots, v_n) \geq 0.$$

## Definition (Clarke pivot function)

The **Clarke pivot function** is the function

$$h_i(v_{-i}) = \max_{b \in A} \sum_{j \neq i} v_j(b).$$

- This leads to **payment functions**

$$p_i(v_1, \dots, v_n) = \max_{b \in A} \sum_{j \neq i} v_j(b) - \sum_{j \neq i} v_j(a)$$

for  $a = f(v_1, \dots, v_n)$ .

- Player  $i$  pays the difference between what the other players could achieve without him and what they achieve with him.
- Each player **internalizes the externalities** he causes.

## Example

- Players  $N = \{1, 2\}$ , alternatives  $A = \{a, b\}$ .
- Values:  $v_1(a) = 10$ ,  $v_1(b) = 2$ ,  $v_2(a) = 9$  and  $v_2(b) = 15$ .
- Without player 1:  $b$  best, since  $v_2(b) = 15 > 9 = v_2(a)$ .
- With player 1:  $a$  best, since  
 $v_1(a) + v_2(a) = 10 + 9 = 19 > 17 = 2 + 15 = v_1(b) + v_2(b)$ .
- With player 1, other players (i.e., player 2) lose  
 $v_2(b) - v_2(a) = 6$  units of utility.

⇒ Clarke pivot function  $h_1(v_2) = 15$

⇒ payment function

$$p_1(v_1, \dots, v_n) = \max_{b \in A} \sum_{j \neq 1} v_j(b) - \sum_{j \neq 1} v_j(a) = 15 - 9 = 6.$$

## Lemma (Clarke pivot rule)

A VCG mechanism with Clarke pivot functions has no positive transfers. If  $v_i(a) \geq 0$  for all  $i = 1, \dots, n$ ,  $v_i \in V_i$  and  $a \in A$ , then the mechanism is also individually rational.

## Proof.

Let  $a = f(v_1, \dots, v_n)$  be the alternative maximizing  $\sum_{j=1}^n v_j(a)$ , and  $b$  the alternative maximizing  $\sum_{j \neq i} v_j(b)$ .

Payment function for  $i$ :  $p_i(v_1, \dots, v_n) = \sum_{j \neq i} v_j(b) - \sum_{j \neq i} v_j(a)$ .

Since  $b$  maximizes  $\sum_{j \neq i} v_j(b)$ :  $p_i(v_1, \dots, v_n) \geq 0$   
( $\leadsto$  no positive transfers).

Utility of player  $i$ :  $u_i = v_i(a) + \sum_{j \neq i} v_j(a) - \sum_{j \neq i} v_j(b)$ .

...

## Proof (ctd.)

Individual rationality: Since  $v_i(b) \geq 0$ ,

$$u_i = v_i(a) + \sum_{j \neq i} v_j(a) - \sum_{j \neq i} v_j(b) \geq \sum_{j=1}^n v_j(a) - \sum_{j=1}^n v_j(b).$$

Since  $a$  maximizes  $\sum_{j=1}^n v_j(a)$ ,

$$\sum_{j=1}^n v_j(a) \geq \sum_{j=1}^n v_j(b)$$

and hence  $u_i \geq 0$ .

Therefore, the mechanism is also individually rational. □

- Recall: **VCG mechanisms** are incentive compatible mechanisms maximizing social welfare.
- With **Clarke pivot functions**:
  - **no positive transfers** and
  - **individual rationality** (if nonnegative valuations).

# Game Theory

## 8. Mechanism Design

### 8.2. Incentive Compatible Mechanisms

#### 8.2.3. Examples

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- $A = N$ . Valuations:  $w_i$ .  $v_a(a) = w_a$ ,  $v_i(a) = 0$  ( $i \neq a$ ).
- $a$  maximizes social welfare  $\sum_{i=1}^n v_i(a)$  iff  $a$  maximizes  $w_a$ .
- Let  $a = f(v_1, \dots, v_n) = \operatorname{argmax}_{j \in A} w_j$  be the highest bidder.
- Payments:  $p_i(v_1, \dots, v_n) = \max_{b \in A} \sum_{j \neq i} v_j(b) - \sum_{j \neq i} v_j(a)$ .
- But  $\max_{b \in A} \sum_{j \neq i} v_j(b) = \max_{b \in A \setminus \{i\}} w_b$ .
- Winner pays value of second highest bid:

$$\begin{aligned} p_a(v_1, \dots, v_n) &= \max_{b \in A} \sum_{j \neq a} v_j(b) - \sum_{j \neq a} v_j(a) \\ &= \max_{b \in A \setminus \{a\}} w_b - 0 = \max_{b \in A \setminus \{a\}} w_b. \end{aligned}$$

- Non-winners pay nothing: For  $i \neq a$ ,

$$\begin{aligned} p_i(v_1, \dots, v_n) &= \max_{b \in A} \sum_{j \neq i} v_j(b) - \sum_{j \neq i} v_j(a) \\ &= \max_{b \in A \setminus \{i\}} w_b - w_a = w_a - w_a = 0. \end{aligned}$$

# Example: Bilateral Trade



- **Seller**  $s$  offers item he values with  $0 \leq w_s \leq 1$ .
- Potential **buyer**  $b$  values item with  $0 \leq w_b \leq 1$ .
- Alternatives  $A = \{\text{no-trade}, \text{trade}\}$ .
- Valuations:

$$v_s(\text{no-trade}) = 0, \quad v_s(\text{trade}) = -w_s,$$

$$v_b(\text{no-trade}) = 0, \quad v_b(\text{trade}) = w_b.$$

- VCG mechanism maximizes  $v_s(a) + v_b(a)$ .
- We have

$$v_s(\text{no-trade}) + v_b(\text{no-trade}) = 0,$$

$$v_s(\text{trade}) + v_b(\text{trade}) = w_b - w_s$$

i.e., **trade** maximizes social welfare iff  $w_b \geq w_s$ .

# Example: Bilateral Trade (ctd.)



- **Requirement:** if *no-trade* is chosen, neither player pays anything:

$$p_s(v_s, v_b) = p_b(v_s, v_b) = 0.$$

- To that end, choose Clarke pivot function **for buyer**:

$$h_b(v_s) = \max_{a \in A} v_s(a).$$

- **For seller:** Modify Clarke pivot function by an additive constant and set

$$h_s(v_b) = \max_{a \in A} v_b(a) - w_b.$$

# Example: Bilateral Trade (ctd.)



- For alternative *no-trade*,

$$\begin{aligned} p_s(v_s, v_b) &= \max_{a \in A} v_b(a) - w_b - v_b(\text{no-trade}) \\ &= w_b - w_b - 0 = 0 \quad \text{and} \\ p_b(v_s, v_b) &= \max_{a \in A} v_s(a) - v_s(\text{no-trade}) \\ &= 0 - 0 = 0. \end{aligned}$$

- For alternative *trade*,

$$\begin{aligned} p_s(v_s, v_b) &= \max_{a \in A} v_b(a) - w_b - v_b(\text{trade}) \\ &= w_b - w_b - w_b = -w_b \quad \text{and} \\ p_b(v_s, v_b) &= \max_{a \in A} v_s(a) - v_s(\text{trade}) \\ &= 0 + w_s = w_s. \end{aligned}$$

## Example: Bilateral Trade (ctd.)

- Because  $w_b \geq w_s$ , the seller gets at least as much as the buyer pays, i.e., the mechanism **subsidizes** the trade.
- Without subsidies, no incentive compatible bilateral trade possible.
- **Note:** Buyer and seller can exploit the system by **colluding**.

# Example: Public Project

- Project costs  $C$  units.
- Each citizen  $i$  privately values the project at  $w_i$  units.
- Government will undertake project if  $\sum_i w_i > C$ .
- Alternatives:  $A = \{\text{no-project}, \text{project}\}$ .
- Valuations:

$$\begin{aligned}v_G(\text{no-project}) &= 0, & v_G(\text{project}) &= -C, \\v_i(\text{no-project}) &= 0, & v_i(\text{project}) &= w_i.\end{aligned}$$

- VCG mechanism with Clarke pivot rule: for each citizen  $i$ ,

$$\begin{aligned}h_i(v_{-i}) &= \max_{a \in A} \left( \sum_{j \neq i} v_j(a) + v_G(a) \right) \\&= \begin{cases} \sum_{j \neq i} w_j - C, & \text{if } \sum_{j \neq i} w_j > C \\ 0, & \text{otherwise.} \end{cases}\end{aligned}$$

# Example: Public Project (ctd.)

- Citizen  $i$  **pivotal** if  $\sum_j w_j > C$  and  $\sum_{j \neq i} w_j \leq C$ .
- **Payment function for citizen  $i$ :**

$$p_i(v_{1..n}, v_G) = h_i(v_{-i}) - \left( \sum_{j \neq i} v_j(f(v_{1..n}, v_G)) + v_G(f(v_{1..n}, v_G)) \right)$$

- **Case 1: Project undertaken,  $i$  pivotal:**

$$p_i(v_{1..n}, v_G) = 0 - \left( \sum_{j \neq i} w_j - C \right) = C - \sum_{j \neq i} w_j$$

- **Case 2: Project undertaken,  $i$  not pivotal:**

$$p_i(v_{1..n}, v_G) = \left( \sum_{j \neq i} w_j - C \right) - \left( \sum_{j \neq i} w_j - C \right) = 0$$

- **Case 3: Project not undertaken:**

$$p_i(v_{1..n}, v_G) = 0$$

# Example: Public Project (ctd.)

- I.e., citizen  $i$  pays nonzero amount

$$C - \sum_{j \neq i} w_j$$

only if he is pivotal.

- He pays difference between value of project to fellow citizens and cost  $C$ , in general less than  $w_i$ .
- Generally,

$$\sum_i p_i(\text{project}) \leq C$$

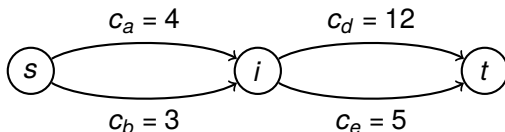
i.e., project has to be subsidized.



# Example: Buying a Path in a Network



- Communication network modeled as  $G = (V, E)$ .
- Each link  $e \in E$  owned by different player  $e$ .
- Each link  $e \in E$  has cost  $c_e$  if used.
- **Objective:** procure communication path from  $s$  to  $t$ .
- **Alternatives:**  $A = \{\pi \mid \pi \text{ path from } s \text{ to } t\}$ .
- **Valuations:**  $v_e(\pi) = -c_e$ , if  $e \in \pi$ , and  $v_e(\pi) = 0$ , if  $e \notin \pi$ .
- **Maximizing social welfare:**  
minimize  $\sum_{e \in \pi} c_e$  over all paths  $\pi$  from  $s$  to  $t$ .
- **Example:**



# Example: Buying a Path in a Network (ctd.)



- For  $G = (V, E)$  and  $e \in E$  let  $G \setminus e = (V, E \setminus \{e\})$ .
- **VCG mechanism:**

$$h_e(v_{-e}) = \max_{\pi' \in G \setminus e} \sum_{e' \in \pi'} -c_{e'}$$

i.e., the cost of the cheapest path from  $s$  to  $t$  in  $G \setminus e$ .  
(Assume that  $G$  is 2-connected, s.t. such  $\pi'$  exists.)

- **Payment functions:** for chosen path  $\pi = f(v_1, \dots, v_n)$ ,

$$p_e(v_1, \dots, v_n) = h_e(v_{-e}) - \sum_{e' \neq e' \in \pi} -c_{e'}.$$

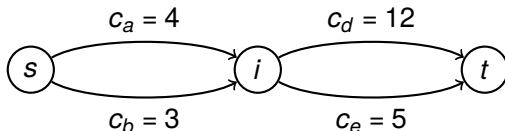
- **Case 1:**  $e \notin \pi$ . Then  $p_e(v_1, \dots, v_n) = 0$ .
- **Case 2:**  $e \in \pi$ . Then

$$p_e(v_1, \dots, v_n) = \max_{\pi' \in G \setminus e} \sum_{e' \in \pi'} -c_{e'} - \sum_{e' \neq e' \in \pi} -c_{e'}.$$

# Example: Buying a Path in a Network (ctd.)



## ■ Example:



- Cost along  $b$  and  $e$ : 8
- Cost without  $e$ : 3
- Cost of cheapest path without  $e$ : 15 (along  $b$  and  $d$ )
- Difference is payment:  $-15 - (-3) = -12$   
I.e., owner of arc  $e$  gets paid 12 for using his arc.
- **Note:** Alternative path after deletion of  $e$  does not necessarily differ from original path at only one position. Could be totally different.

We saw some examples of applications of VCG mechanisms:

- Vickrey Auctions
- Bilateral Trade
- Public Projects
- Buying a Path in a Network

# Game Theory

## 8. Mechanism Design

### 8.3. Mechanisms without Money

#### 8.3.1. House Allocation

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## Motivation 1:

- According to Gibbard-Satterthwaite:  
In general, **nontrivial social choice functions manipulable**.
- **One way out: introduction of money**  
(cf. VCG mechanisms)
- **Other way out: restriction of preferences**  
(cf. single-peaked preferences; this chapter)

## Motivation 2:

- Introduction of central concept from cooperative game theory: **the core**

## Examples:

- **House allocation problem** (Sec. 8.3.1)
- **Stable matchings** (Sec. 8.3.2)

- **Players**  $N = \{1, \dots, n\}$ .
- Each player  $i$  owns house  $i$ .
- Each player  $i$  has **strict linear preference** order  $\triangleleft_i$  over the set of houses.  
**Example:**  $j \triangleleft_i k$  means player  $i$  prefers house  $k$  to house  $j$ .
- **Alternatives**  $A$ : allocations of houses to players (permutations  $\pi \in S_n$  of  $N$ ).  
**Example:**  $\pi(i) = j$  means player  $i$  gets house  $j$ .
- **Objective:** **reallocate the houses** among the agents “appropriately”.

- Note on preference relations:
  - arbitrary (strict linear) preference orders  $\triangleleft_i$  over houses,
  - but **no** arbitrary preference orders  $\preceq_i$  over  $A$ .
- **Rather:** player  $i$  **indifferent** between different allocations  $\pi_1$  and  $\pi_2$  as long as  $\pi_1(i) = \pi_2(i)$ .  
Indifference denoted as  $\pi_1 \approx_i \pi_2$ .
- If player  $i$  is not indifferent:  $\pi_1 \prec_i \pi_2$  iff  $\pi_1(i) \triangleleft_i \pi_2(i)$ .
- **Notation:**  $\pi_1 \preceq_i \pi_2$  iff  $\pi_1 \prec_i \pi_2$  or  $\pi_1 \approx_i \pi_2$ .
- This makes **Gibbard-Satterthwaite inapplicable**.



- Important new aspect of house allocation problem: players control resources to be allocated.
- Allocation can be subverted by subset of agents breaking away and trading among themselves.
- How to avoid such allocations?
- How to make allocation mechanism non-manipulable?

**Notation:** For  $M \subseteq N$ , let

$$A(M) = \{\pi \in A \mid \forall i \in M : \pi(i) \in M\}$$

be the set of allocations that can be achieved by the agents in  $M$  trading among themselves.

## Definition (blocking coalition)

Let  $\pi \in A$  be an allocation. A set  $M \subseteq N$  is called a **blocking coalition** for  $\pi$  if there exists a  $\pi' \in A(M)$  such that

- $\pi \preceq_i \pi'$  for all  $i \in M$  and
- $\pi \prec_i \pi'$  for at least one  $i \in M$ .

## Intuition:

A blocking coalition can receive houses everyone from the coalition likes at least as much as under allocation  $\pi$ , with at least one player being strictly better off, by trading among themselves.

## Definition (core)

The set of allocations that is not blocked by any subset of agents is called the **core**.

**Question:** Is the core nonempty?

- Algorithm to construct allocation
- Let  $G = \langle V, A, c \rangle$  be an arc-colored directed graph where:
  - $V = N$  (i.e., one vertex for each player),
  - $A = V \times V$ , and
  - $c : A \rightarrow N$  such that  $c(i, j) = k$  if house  $j$  is player  $i$ 's  $k$ th ranked choice according to  $\triangleleft_i$ .
- **Note:** Loops  $(i, i)$  are allowed. We treat them as cycles of length 0.

# Top Trading Cycle Algorithm (TTCA)



## Pseudocode:

let  $\pi(i) = i$  for all  $i \in N$ .

**while** players unaccounted for **do**

    consider subgraph  $G'$  of  $G$  where each vertex has  
    only one outgoing arc: the least-colored one from  $G$ .

    identify cycles in  $G'$ .

    add corresponding cyclic permutations to  $\pi$ .

    delete players accounted for and incident edges from  $G$ .

**end while**

output  $\pi$ .

## Notation:

Let  $N_i$  be the set of vertices on cycles identified in iteration  $i$ .

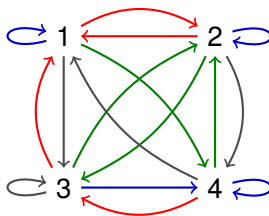
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
- Player 3:  $3 \triangleleft_3 4 \triangleleft_3 2 \triangleleft_3 1$
- Player 4:  $1 \triangleleft_4 4 \triangleleft_4 2 \triangleleft_4 3$

Corresponding graph:



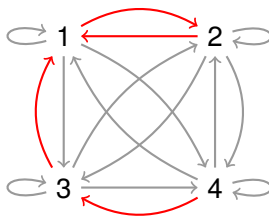
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
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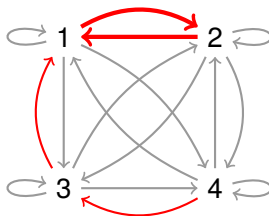
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
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- Player 4:  $1 \triangleleft_4 4 \triangleleft_4 2 \triangleleft_4 3$

Corresponding graph:



- Iteration 1:  $\pi(1) = 2, \pi(2) = 1$ .



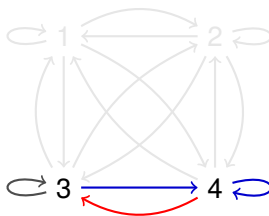
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
- Player 3:  $3 \triangleleft_3 4 \triangleleft_3 2 \triangleleft_3 1$
- Player 4:  $1 \triangleleft_4 4 \triangleleft_4 2 \triangleleft_4 3$

Corresponding graph:



- Iteration 1:  $\pi(1) = 2, \pi(2) = 1$ .

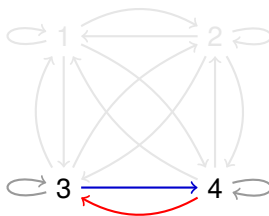
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
- Player 3:  $3 \triangleleft_3 4 \triangleleft_3 2 \triangleleft_3 1$
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Corresponding graph:



- Iteration 1:  $\pi(1) = 2, \pi(2) = 1$ .

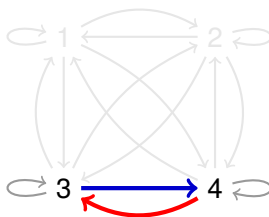
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
- Player 3:  $3 \triangleleft_3 4 \triangleleft_3 2 \triangleleft_3 1$
- Player 4:  $1 \triangleleft_4 4 \triangleleft_4 2 \triangleleft_4 3$

Corresponding graph:



- Iteration 1:  $\pi(1) = 2, \pi(2) = 1$ .
- Iteration 2:  $\pi(3) = 4, \pi(4) = 3$ .

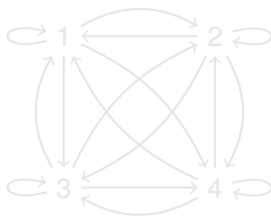
# Top Trading Cycle Algorithm (TTCA)



Example:

- Player 1:  $3 \triangleleft_1 1 \triangleleft_1 4 \triangleleft_1 2$
- Player 2:  $4 \triangleleft_2 2 \triangleleft_2 3 \triangleleft_2 1$
- Player 3:  $3 \triangleleft_3 4 \triangleleft_3 2 \triangleleft_3 1$
- Player 4:  $1 \triangleleft_4 4 \triangleleft_4 2 \triangleleft_4 3$

Corresponding graph:



- Iteration 1:  $\pi(1) = 2, \pi(2) = 1$ .
- Iteration 2:  $\pi(3) = 4, \pi(4) = 3$ .
- Done:  $\pi(1) = 2, \pi(2) = 1, \pi(3) = 4, \pi(4) = 3$ .

## Theorem

*The core of the house allocation problem consists of exactly one matching.*

## Proof sketch

**At most one matching:** Show that if a matching is in the core, it must be the one returned by the TTCA.

In TTCA, each player in  $N_1$  receives his favorite house.

Therefore,  $N_1$  would form a blocking coalition to any allocation that does not assign to all of those players the houses they would receive in TTCA.

...

## Proof sketch (ctd.)

That is, any core allocation must assign  $N_1$  to houses as TTCA assigns them.

Argument can be extended inductively to  $N_k$ ,  $2 \leq k \leq n$ .

**At least one matching:** Show that TTCA allocation is in the core, i.e., that there is no other blocking coalition  $M \subseteq N$ .  
Homework. □

Question: What about manipulability?

## Definition (top trading cycle mechanism)

The **top trading cycle mechanism (TTCM)** is the function that, for each profile of preferences, returns the allocation computed by the TTCA.

## Theorem

*The TTCM cannot be manipulated.*

## Proof

Homework. ☐

- **Avoid Gibbard-Satterthwaite** by restricting domain of preferences.
- **House allocation** problem:
  - Solved using **top trading cycle** algorithm.
  - Algorithm finds **unique solution in the core**, where no **blocking coalition** of players has an incentive to break away.
  - The top trading cycle mechanism **cannot be manipulated**.



# Game Theory

## 8. Mechanism Design

### 8.3. Mechanisms without Money

#### 8.3.2. Stable Matchings

Albert-Ludwigs-Universität Freiburg



**UNI  
FREIBURG**

Bernhard Nebel and Robert Mattmüller

Summer semester 2020

## Motivation 1:

- According to Gibbard-Satterthwaite:  
In general, **nontrivial social choice functions manipulable**.
- **One way out: introduction of money**  
(cf. VCG mechanisms)
- **Other way out: restriction of preferences**  
(cf. single-peaked preferences; this chapter)

## Motivation 2:

- Introduction of central concept from cooperative game theory: **the core**

## Examples:

- House allocation problem (Sec. 8.3.1)
- **Stable matchings** (Sec. 8.3.2)

## Problem statement:

- Given disjoint finite sets  $M$  of men and  $W$  of women.
- Assume WLOG that  $|M| = |W|$   
(introduce dummy-men/dummy-women).
- Each  $m \in M$  has strict preference ordering  $\prec_m$  over  $W$ .
- Each  $w \in W$  has strict preference ordering  $\prec_w$  over  $M$ .
- **Matching:** “appropriate” assignment of men to women such that each man is assigned to at most one woman and vice versa.

**Note:** A group of players can **subvert a matching** by opting out.

## Definition (stability, blocking pair)

A matching is called **unstable** if there are two men  $m, m'$  and two women  $w, w'$  such that

- $m$  is matched to  $w$ ,
- $m'$  is matched to  $w'$ , and
- $w \prec_m w'$  and  $m' \prec_{w'} m$ .

The pair  $\langle m, w' \rangle$  is called a **blocking pair**.

A matching that has no blocking pairs is called **stable**.

## Definition (core)

The **core** of the matching game is the set of all stable matchings.

Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
- Woman 1:  $m_2 \prec_{w_1} m_3 \prec_{w_1} m_1$
- Woman 2:  $m_2 \prec_{w_2} m_1 \prec_{w_2} m_3$
- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

Two matchings:

## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
- Woman 1:  $m_2 \prec_{w_1} m_3 \prec_{w_1} m_1$
- Woman 2:  $m_2 \prec_{w_2} m_1 \prec_{w_2} m_3$
- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

## Two matchings:

- Matching  $\{\langle m_1, w_1 \rangle, \langle m_2, w_2 \rangle, \langle m_3, w_3 \rangle\}$

## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
- Woman 1:  $m_2 \prec_{w_1} m_3 \prec_{w_1} m_1$
- Woman 2:  $m_2 \prec_{w_2} m_1 \prec_{w_2} m_3$
- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

## Two matchings:

- Matching  $\{\langle m_1, w_1 \rangle, \langle m_2, w_2 \rangle, \langle m_3, w_3 \rangle\}$ 
  - unstable ( $\langle m_1, w_2 \rangle$  is a blocking pair)

## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
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## Two matchings:

- Matching  $\{\langle m_1, w_1 \rangle, \langle m_2, w_2 \rangle, \langle m_3, w_3 \rangle\}$ 
  - unstable ( $\langle m_1, w_2 \rangle$  is a blocking pair)
- Matching  $\{\langle m_1, w_1 \rangle, \langle m_3, w_2 \rangle, \langle m_2, w_3 \rangle\}$ 
  - stable



Question: Is there always a stable matching?

Answer: Yes! And it can even be efficiently constructed.

How? Deferred acceptance algorithm!

## Definition (deferred acceptance algorithm, male proposals)

- 1 Each man proposes to his top-ranked choice.
- 2 Each woman who has received at least one proposal (including tentatively kept one from earlier rounds) tentatively keeps top-ranked proposal and rejects rest.
- 3 If no man is left rejected, stop.
- 4 Otherwise, each man who has been rejected proposes to his top-ranked choice among the women who have not rejected him. Then, goto 2.

## Note:

- Algorithm has polynomial runtime.
- No man is assigned to more than one woman.
- No woman is assigned to more than one man.
- $\rightsquigarrow$  matching

# Deferred Acceptance Algorithm



Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
- Woman 1:  $m_2 \prec_{w_1} m_3 \prec_{w_1} m_1$
- Woman 2:  $m_2 \prec_{w_2} m_1 \prec_{w_2} m_3$
- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

Deferred acceptance algorithm:

# Deferred Acceptance Algorithm



## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
- Man 3:  $w_3 \prec_{m_3} w_2 \prec_{m_3} w_1$
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- Woman 2:  $m_2 \prec_{w_2} m_1 \prec_{w_2} m_3$
- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

## Deferred acceptance algorithm:

- 1  $m_1$  proposes to  $w_2$ ,  $m_2$  to  $w_1$ , and  $m_3$  to  $w_1$ .

# Deferred Acceptance Algorithm



## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
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- Woman 3:  $m_2 \prec_{w_3} m_3 \prec_{w_3} m_1$

## Deferred acceptance algorithm:

- 1  $m_1$  proposes to  $w_2$ ,  $m_2$  to  $w_1$ , and  $m_3$  to  $w_1$ .
- 2  $w_1$  keeps  $m_3$  and rejects  $m_2$ ,  $w_2$  keeps  $m_1$ .

# Deferred Acceptance Algorithm



## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
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## Deferred acceptance algorithm:

- 1  $m_1$  proposes to  $w_2$ ,  $m_2$  to  $w_1$ , and  $m_3$  to  $w_1$ .
- 2  $w_1$  keeps  $m_3$  and rejects  $m_2$ ,  $w_2$  keeps  $m_1$ .
- 3  $m_2$  now proposes to  $w_3$ .

# Deferred Acceptance Algorithm



## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
- Man 2:  $w_2 \prec_{m_2} w_3 \prec_{m_2} w_1$
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## Deferred acceptance algorithm:

- 1  $m_1$  proposes to  $w_2$ ,  $m_2$  to  $w_1$ , and  $m_3$  to  $w_1$ .
- 2  $w_1$  keeps  $m_3$  and rejects  $m_2$ ,  $w_2$  keeps  $m_1$ .
- 3  $m_2$  now proposes to  $w_3$ .
- 4  $w_3$  keeps  $m_2$ .



# Deferred Acceptance Algorithm



## Example:

- Man 1:  $w_3 \prec_{m_1} w_1 \prec_{m_1} w_2$
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## Deferred acceptance algorithm:

- 1  $m_1$  proposes to  $w_2$ ,  $m_2$  to  $w_1$ , and  $m_3$  to  $w_1$ .
- 2  $w_1$  keeps  $m_3$  and rejects  $m_2$ ,  $w_2$  keeps  $m_1$ .
- 3  $m_2$  now proposes to  $w_3$ .
- 4  $w_3$  keeps  $m_2$ .

Resulting matching:  $\{\langle m_1, w_2 \rangle, \langle m_2, w_3 \rangle, \langle m_3, w_1 \rangle\}$ .

## Theorem

The deferred acceptance algorithm with male proposals terminates in a stable matching.

## Proof.

Suppose not.

Then there exists a blocking pair  $\langle m_1, w_1 \rangle$  with  $m_1$  matched to some  $w_2$  and  $w_1$  matched to some  $m_2$ .

Since  $\langle m_1, w_1 \rangle$  is blocking and  $w_2 \prec_{m_1} w_1$ , in the proposal algorithm,  $m_1$  would have proposed to  $w_1$  before  $w_2$ .

Since  $m_1$  was not matched with  $w_1$  by the algorithm, it must be because  $w_1$  received a proposal from a man she ranked higher than  $m_1$ . ...

## Proof (ctd.)

Since the algorithm matches her to  $m_2$  it follows that  $m_1 \prec_{w_1} m_2$ .

This contradicts the fact that  $\langle m_1, w_1 \rangle$  is a blocking pair. □

Analogous version where the women propose: outcome would also be a stable matching.

Denote a matching by  $\mu$ . The woman assigned to man  $m$  in  $\mu$  is  $\mu(m)$ , and the man assigned to woman  $w$  is  $\mu(w)$ .

## Definition (optimality)

A matching  $\mu$  is **male-optimal** if there is no stable matching  $\nu$  such that  $\mu(m) \prec_m \nu(m)$  or  $\mu(m) = \nu(m)$  for all  $m \in M$  and  $\mu(m) \prec_m \nu(m)$  for at least one  $m \in M$ . **Female-optimal**: similar.

## Theorem

- *The stable matching produced by the (fe)male-proposal deferred acceptance algorithm is (fe)male-optimal.*
- *In general, there is no stable matching that is male-optimal and female-optimal.*



## Theorem

*The mechanism associated with the (fe)male-proposal algorithm cannot be manipulated by the (fe)males.*



**Note:** The mechanism associated with the male-proposal algorithm **can** be manipulated by the females and vice versa.

**Idea:** strategically reject a proposal who then binds your main competitor for your favorite partner in the next round, freeing up that partner for you  $\rightsquigarrow$  try this out with our running example!)

- **Avoid Gibbard-Satterthwaite** by restricting domain of preferences.
- **Stable matchings:**
  - Solved using **deferred acceptance** algorithm.
  - Algorithm finds **a stable matching in the core**, where no **blocking pair** of players has an incentive to break away.
  - The mechanism associated with the (fe)male-proposal algorithm cannot be manipulated by the (fe)males.

# Game Theory

## 8. Mechanism Design

### 8.4. Combinatorial Auctions

Albert-Ludwigs-Universität Freiburg



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Summer semester 2020

## Motivation:

- **Multiple items** are auctioned concurrently.
- Bidders have preferences for **combinations (bundles)** of items.
- Items can **complement** or **substitute** one another.
  - **complement**: left and right shoe together.
  - **substitute**: two right shoes.
- **Aim**: socially optimal **allocation** of items to bidders.



## Applications:

- Spectrum auctions (with combinations of spectrum bands and geographical areas)
- Procurement of transportation services for multiple routes
- ...

## Notation:

- **Items:**  $G = \{1, \dots, m\}$
- **Bidders:**  $N = \{1, \dots, n\}$

## Definition (valuation)

A **valuation** is a function  $v : 2^G \rightarrow \mathbb{R}^+$  with  $v(\emptyset) = 0$  and  $v(S) \leq v(T)$  for  $S \subseteq T \subseteq G$ .

- Requirement  $v(\emptyset) = 0$  to “normalize” valuations.
- Requirement  $v(S) \leq v(T)$  for  $S \subseteq T \subseteq G$ : monotonicity (or “free disposal”).

Let  $S, T \subseteq G$  be disjoint.

- $S$  and  $T$  are **complements** to each other if  $v(S \cup T) > v(S) + v(T)$ .
- $S$  and  $T$  are **substitutes** if  $v(S \cup T) < v(S) + v(T)$ .

## Definition (allocation)

An **allocation** of the items to the bidders is a tuple  $\langle S_1, \dots, S_n \rangle$  with  $S_i \subseteq G$  for  $i = 1, \dots, n$  and  $S_i \cap S_j = \emptyset$  for  $i \neq j$ .

The **social welfare** obtained by an allocation is  $\sum_{i=1}^n v_i(S_i)$  if  $v_1, \dots, v_n$  are the valuations of the bidders.

An allocation is called **socially efficient** if it maximizes social welfare among all allocations.

Let  $A$  be the set of all allocations.

## Definition (winner determination problem)

Let  $v_i : 2^G \rightarrow \mathbb{R}^+$ ,  $i = 1, \dots, n$ , be the declared valuations of the bidders. The **winner determination problem (WDP)** is the problem of finding a socially efficient allocation  $a \in A$  for these valuations.

**Aim:** Develop **mechanism** for WDP.

## Challenges:

- Incentive compatibility
- Complexity of representation and communication of preferences (exponentially many subsets of items!)
- Computational complexity



## Motivation:

- **Focus on single-minded bidders:** cuts complexity of representation down to polynomial space.
- **Idea: single-minded bidder** focuses on one bundle, has fixed valuation  $v^*$  for that bundle (and its supersets), valuation 0 for all other bundles.

## Definition (single-minded bidder)

A valuation  $v$  is called **single-minded** if there is a bundle  $S^* \subseteq G$  and a value  $v^* \in \mathbb{R}^+$  such that

$$v(S) = \begin{cases} v^* & \text{if } S^* \subseteq S \\ 0 & \text{otherwise} \end{cases}$$

A **single-minded bid** is a pair  $\langle S^*, v^* \rangle$ .

- **Representational** complexity: **solved**.
- **Computational** complexity: **not solved**.

Combinatorial Auctions

Single-Minded Bidders

Definitions

Complexity

Greedy Mechanism for Single-Minded Bidders

Properties of Greedy Mechanism

Summary

# Allocation Problem for Single-Minded Bidders



## Definition (allocation problem for single-minded bidders)

The **allocation problem for single-minded bidders (APSMB)** is defined by the following input and output.

- **INPUT.** Bids  $\langle S_i^*, v_i^* \rangle$  for  $i = 1, \dots, n$
- **OUTPUT.**  $W \subseteq \{1, \dots, n\}$  with  $S_i^* \cap S_j^* = \emptyset$  for  $i, j \in W, i \neq j$  such that  $\sum_{i \in W} v_i^*$  is maximized.

**Claim:** APSMB is NP-complete.

# Allocation Problem for Single-Minded Bidders



Since APSMB is an **optimization problem**, consider the corresponding **decision problem**:

**Definition (allocation problem for single-minded bidders, decision problem)**

The **decision problem version of APSMB (APSMB-D)** is defined by the following input and output.

- **INPUT.** Bids  $\langle S_i^*, v_i^* \rangle$  for  $i = 1, \dots, n$  and  $k \in \mathbb{N}$
- **OUTPUT.** Is there a  $W \subseteq \{1, \dots, n\}$  with  $S_i^* \cap S_j^* = \emptyset$  for  $i, j \in W, i \neq j$  such that  $\sum_{i \in W} v_i^* \geq k$ ?

## Theorem

APSMB-D is NP-complete.

Combinatorial Auctions

Single-Minded Bidders

Definitions

Complexity

Greedy Mechanism for Single-Minded Bidders

Properties of Greedy Mechanism

Summary



# APSMB-D is NP-complete



## Proof

NP-hardness: reduction from INDEPENDENT-SET.

INDEPENDENT-SET instance:

- undirected graph  $\langle V, E \rangle$  and  $k_{IS} \in \mathbb{N}$ .
- **Question:** Is there an independent set of size  $k_{IS}$  in  $\langle V, E \rangle$ ?

Corresponding APSMB-D instance:

- $k = k_{IS}$ , items  $G = E$ , bidders  $N = V$ , and
- for each bidder  $i \in V$  the bid  $\langle S_i^*, v_i^* \rangle$  with  $S_i^* = \{e \in E \mid i \in e\}$  and  $v_i^* = 1$ .
- **Question:** Is there an allocation with social welfare  $\geq k$ ?
- (Intuitively: Vertices bid for their incident edges.)

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## Proof (ctd.)

Since  $S_i^* \cap S_j^* = \emptyset$  for  $i, j \in W, i \neq j$ , the set of winners  $W$  represents an independent set of cardinality

$$|W| = \sum_{i \in W} v_i^*.$$

Therefore, there is an independent set of cardinality at least  $k_{IS}$  iff there is a set of winners  $W$  with  $\sum_{i \in W} v_i^* \geq k$ .

This proves NP-hardness.

**APSMB-D  $\in$  NP:** obvious (guess and verify set of winners).

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# APSMB-D is NP-complete



## Consequences:

- Solving APSMB **optimally**: too costly.
- **Alternatives**:
  - **approximation** algorithm
  - **heuristic** approach
  - **special cases**
- **Here**: approximation algorithm

## Definition (approximation factor)

Let  $c \geq 1$ . An allocation  $\langle S_1, \dots, S_n \rangle$  is a **c-approximation** of an optimal allocation if

$$\sum_{i=1}^n v_i(T_i) \leq c \cdot \sum_{i=1}^n v_i(S_i)$$

for an optimal allocation  $\langle T_1, \dots, T_n \rangle$ .

## Proposition

Approximating APSMB within a factor of  $c \leq m^{1/2-\varepsilon}$  for any  $\varepsilon > 0$  is NP-hard. □

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Best we can still hope for in case of single-minded bidders:

- incentive compatible
- $m^{1/2}$ -approximation algorithm
- with polynomial runtime.

Good news:

- Such an algorithm exists!

## Definition (mechanism for single-minded bidders)

Let  $V_{sm}$  be the set of all single-minded bids and  $A$  the set of all allocations.

A **mechanism for single-minded bidders** is a tuple  $\langle f, p_1, \dots, p_n \rangle$  consisting of

- a **social choice function**  $f : V_{sm}^n \rightarrow A$  and
- **payment functions**  $p_i : V_{sm}^n \rightarrow \mathbb{R}$  for all  $i = 1, \dots, n$ .

# Mechanism for Single-Minded Bidders



## Definition (efficient computability)

A mechanism for single-minded bidders is **efficiently computable** if  $f$  and all  $p_i$  can be computed in polynomial time.

## Definition (incentive compatibility)

A mechanism for single-minded bidders is **incentive compatible** if

$$v_i(f(v_i, v_{-i})) - p_i(v_i, v_{-i}) \geq v_i(f(v'_i, v_{-i})) - p_i(v'_i, v_{-i})$$

for all  $i = 1, \dots, n$  and all  $v_1, \dots, v_n, v'_i \in V_{sm}$ , where  $v_i(a) = v_i^*$  if  $i$  wins in  $a$  (gets the desired bundle), and  $v_i(a) = 0$ , otherwise.

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## How to build such a mechanism?

- In principle: could use a **VCG mechanism**.
- Problem with VCG: incentive compatible, but **not efficiently computable**  
(need to compute social welfare, which is NP-hard)
- Alternative idea: VCG-like mechanism that **approximates social welfare**
- Problem with alternative: efficiently computable, but **not incentive compatible**
- Solution: forget VCG, **use specific mechanism for single-minded bidders**.

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# Greedy Mechanism for Single-Minded Bidders



## Definition (greedy mechanism for single-minded bidders)

The **greedy mechanism for single-minded bidders (GMSMB)** is defined as follows.

Let the bidders  $1, \dots, n$  be ordered such that

$$\frac{v_1^*}{\sqrt{|S_1^*|}} \geq \frac{v_2^*}{\sqrt{|S_2^*|}} \geq \dots \geq \frac{v_n^*}{\sqrt{|S_n^*|}}.$$

...

# Greedy Mechanism for Single-Minded Bidders



Definition (greedy mechanism for single-minded bidders, ctd.)

Let the set  $W \subseteq \{1, \dots, n\}$  be procedurally defined by the following pseudocode:

```
 $W \leftarrow \emptyset$   
for  $i = 1, \dots, n$  do  
  if  $S_i^* \cap \left( \bigcup_{j \in W} S_j^* \right) = \emptyset$  then  
     $W \leftarrow W \cup \{i\}$   
  end if  
end for
```

...

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# Greedy Mechanism for Single-Minded Bidders



Definition (greedy mechanism for single-minded bidders, ctd.)

**Result:** allocation  $a$  where exactly the bidders in  $W$  win.

**Payments:**

- **Case 1:** If  $i \in W$  and there is a smallest index  $j$  such that  $S_i^* \cap S_j^* \neq \emptyset$  and for all  $k < j$ ,  $k \neq i$ ,  $S_k^* \cap S_j^* = \emptyset$ , then

$$p_i(v_1, \dots, v_n) = \frac{v_j^*}{\sqrt{|S_j^*|/|S_i^*|}},$$

- **Case 2:** Otherwise,

$$p_i(v_1, \dots, v_n) = 0.$$

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## Example

Let  $N = \{1, 2, 3, 4\}$  and  $G = \{1, \dots, 13\}$ .

$i$	Package $S_i^*$	Val. $v_i^*$	$v_i^* / \sqrt{ S_i^* }$	Assignm. order
1	$\{1, 2, 3, 4, 5, 6, 7, 8, 9\}$	15		
2	$\{3, 4, 5, 6, 7, 8, 9, 12, 13\}$	3		
3	$\{1, 2, 10, 11\}$	12		
4	$\{10, 11, 12, 13\}$	8		

Positions in assignment order? Winner set? Assignment?  
Social welfare of winner set?

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## Example (ctd.)

### Assignments:

- 1 Bidder 3 gets  $\{1, 2, 10, 11\}$ .
- 2 Bidder 1 gets nothing (obj. 1 and 2 already assigned).
- 3 Bidder 4 gets nothing (obj. 10 and 11 already assigned).
- 4 Bidder 2 gets the remainder, i.e.,  $\{3, 4, 5, 6, 7, 8, 9, 12, 13\}$ .

### Payments:

- 1 Bidder 3 pays

$$\frac{v_1^*}{\sqrt{|S_1^*|/|S_3^*|}} = \frac{15}{\sqrt{9/4}} = \frac{15}{3/2} = 10.$$

- 2 Bidders 1, 4 and 2 pay 0.

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## Example (ctd.)

Therefore:

- Winner set:  $W = \{2, 3\}$ .  
Social welfare:  $U = 12 + 3 = 15$ .
- Optimal winner set:  $W^* = \{1, 4\}$ .  
Optimal social welfare:  $U^* = 15 + 8 = 23$ .
- Approximation ratio:  $23/15 < 2 < 3 < \sqrt{13} = \sqrt{m}$

# Greedy Mechanism for Single-Minded Bidders: Efficient Computability



## Theorem

GMSMB is efficiently computable. ☐

## Open questions:

- What about incentive compatibility?
- What about approximation factor of  $\sqrt{m}$ ?

# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



To prove incentive compatibility:

- Step 1: Show that GMSMB is **monotone**.
- Step 2: Show that GMSMB **uses critical payments**.
- Step 3: Show that in GMSMB **losers pay nothing**.
- Step 4: Show that every mechanism for single-minded bidders that is **monotone**, that **uses critical payments**, and where **losers pay nothing** is **incentive compatible**.



# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Definition (monotonicity)

A mechanism for single-minded bidders is **monotone** if a bidder who wins with bid  $\langle S^*, v^* \rangle$  would also win with any bid  $\langle S', v' \rangle$  where  $S' \subseteq S^*$  and  $v' \geq v^*$  (for fixed bids of the other bidders).

## Definition (critical payments)

A mechanism for single-minded bidders **uses critical payments** if a bidder who wins pays the minimal amount necessary for winning, i.e., the infimum of all  $v'$  such that  $\langle S^*, v' \rangle$  still wins.

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# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Lemma

GMSBM is monotone, uses critical payments, and losers pay nothing.

## Proof

**Monotonicity:** Increasing  $v_i^*$  or decreasing  $S_i^*$  can only move bidder  $i$  up in the greedy order, making it easier to win.

**Critical payments:** Bidder  $i$  wins as long as he is before bidder  $j$  in the greedy order (if such a  $j$  exists). This holds iff

$$\frac{v_i^*}{\sqrt{|S_i^*|}} \geq \frac{v_j^*}{\sqrt{|S_j^*|}} \quad \text{iff} \quad v_i^* \geq \frac{v_j^* \sqrt{|S_i^*|}}{\sqrt{|S_j^*|}} = \frac{v_j^*}{\sqrt{|S_j^*|/|S_i^*|}} = p_i.$$

**Losers pay nothing:** Obvious. □

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# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Lemma

A mechanism for single-minded bidders that is monotone, that uses critical payments, and where losers pay nothing is incentive compatible.

## Proof

(A) Truthful bids never lead to negative utility.

- If the declared bid loses, bidder has utility 0.
- If the declared bid wins, he has utility  $v^* - p^* \geq 0$ , since  $v^* \geq p^*$ , because  $p^*$  is the critical payment, and if the bid wins, the bidder must have (truthfully) bid a value  $v^*$  of at least  $p^*$ .

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# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Proof (ctd.)

(B) Truthful bids never lead to lower utility than untruthful bids. Suppose declaration of untruthful bid  $\langle S', v' \rangle$  deviating from truthful bid  $\langle S^*, v^* \rangle$ .

(B.1) Case 1: untruthful bid is losing or not useful for bidder.

Suppose  $\langle S', v' \rangle$  is losing or  $S^* \not\subseteq S'$  (bidder does not get the bundle he wants). Then utility  $\leq 0$  in  $\langle S', v' \rangle$ , i.e., no improvement over utility when declaring  $\langle S^*, v^* \rangle$  (cf. (A)).

(B.2) Case 2: untruthful bid is winning and useful for bidder.

Assume  $\langle S', v' \rangle$  is winning and  $S^* \subseteq S'$ . To show that  $\langle S^*, v^* \rangle$  is at least as good a bid as  $\langle S', v' \rangle$ , show that  $\langle S^*, v' \rangle$  is at least as good as  $\langle S', v' \rangle$  and that  $\langle S^*, v^* \rangle$  is at least as good as  $\langle S^*, v' \rangle$ .

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# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Proof (ctd.)

- (B.2.a) Lying about desired bundle does not help.

Show that  $\langle S^*, v' \rangle$  is at least as good as  $\langle S', v' \rangle$ .

Let  $p'$  be the payment for bid  $\langle S', v' \rangle$  and  $p$  the payment for bid  $\langle S^*, v' \rangle$ .

For all  $x < p$ ,  $\langle S^*, x \rangle$  is losing, since  $p$  is the critical payment for  $S^*$ .

Due to monotonicity, also  $\langle S', x \rangle$  is losing for all  $x < p$ .

Hence, the critical payment  $p'$  for  $S'$  is at least  $p$ .

Thus,  $\langle S^*, v' \rangle$  is still winning, if  $\langle S', v' \rangle$  was, and leads to the same or even lower payment.

...

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## Proof (ctd.)

- (B.2.b) Lying about valuation does not help.

Show that  $\langle S^*, v^* \rangle$  is at least as good as  $\langle S^*, v' \rangle$ .

- (B.2.b.i) Case 1:  $\langle S^*, v^* \rangle$  is winning with payment  $p^*$ .

If  $v' > p^*$ , then  $\langle S^*, v' \rangle$  is still winning with the same payment, so there is no incentive to deviate to  $\langle S^*, v' \rangle$ .

If  $v' \leq p^*$ , then  $\langle S^*, v' \rangle$  is losing, so there is also no incentive to deviate to  $\langle S^*, v' \rangle$ .

- (B.2.b.ii) Case 2:  $\langle S^*, v^* \rangle$  is losing.

Then  $v^*$  is less than the critical payment, i.e., the payment  $p'$  for a winning bid  $\langle S^*, v' \rangle$  would be greater than  $v^*$ , making a deviation to  $\langle S^*, v' \rangle$  unprofitable.  $\square$

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# Greedy Mechanism for Single-Minded Bidders: Incentive Compatibility



## Corollary

The greedy mechanism for single-minded bidders is incentive compatible. ☐

## Open question:

- What about approximation factor of  $\sqrt{m}$ ?

# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



In the next proof, we will need the **Cauchy-Schwarz inequality**:

## Theorem (Cauchy-Schwarz inequality)

Let  $x_j, y_j \in \mathbb{R}$ . Then

$$\sum_j x_j y_j \leq \sqrt{\sum_j x_j^2} \cdot \sqrt{\sum_j y_j^2}.$$



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## Lemma

GMSBM produces a winner set  $W$  that induces a social welfare that is at most a factor  $\sqrt{m}$  worse than the optimal social welfare.



# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof

- Let  $W^*$  be a set of winning bidders such that  $\sum_{i \in W^*} v_i^*$  is maximal and  $S_i^* \cap S_j^* = \emptyset$  for  $i, j \in W^*, i \neq j$ .
- Let  $W$  be the result of GMSMB.

Show:

$$\sum_{i \in W^*} v_i^* \leq \sqrt{m} \sum_{i \in W} v_i^*.$$

For  $i \in W$  let

$$W_i^* = \{j \in W^* \mid j \geq i \text{ and } S_i^* \cap S_j^* \neq \emptyset\}$$

be the winners in  $W^*$  identical with  $i$  or not contained in  $W$  because of bidder  $i$ . ...

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# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

Since no  $j \in W_i^*$  is before  $i$  in the greedy ordering, for such  $j$ ,

$$v_j^* \leq \frac{v_i^*}{\sqrt{|S_i^*|}} \sqrt{|S_j^*|} \quad \text{and, summing over } j \in W_i^*$$

$$\sum_{j \in W_i^*} v_j^* \leq \frac{v_i^*}{\sqrt{|S_i^*|}} \sum_{j \in W_i^*} \sqrt{|S_j^*|}. \quad (1)$$

With Cauchy-Schwarz for  $x_j = 1$  and  $y_j = \sqrt{|S_j^*|}$ :

$$\sum_{j \in W_i^*} \sqrt{|S_j^*|} \leq \sqrt{\sum_{j \in W_i^*} 1^2} \sqrt{\sum_{j \in W_i^*} |S_j^*|} = \sqrt{|W_i^*|} \sqrt{\sum_{j \in W_i^*} |S_j^*|}. \quad (2)$$

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# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

For all  $j \in W_i^*$ ,  $S_i^* \cap S_j^* \neq \emptyset$ , i.e., there is a  $g(j) \in S_i^* \cap S_j^*$ .

Since  $W^*$  induces an allocation, for all  $j_1, j_2 \in W_i^*$ ,  $j_1 \neq j_2$ ,

$$S_{j_1}^* \cap S_{j_2}^* = \emptyset$$

Hence,

$$(S_i^* \cap S_{j_1}^*) \cap (S_i^* \cap S_{j_2}^*) = \emptyset$$

i.e.,  $g(j_1) \neq g(j_2)$  for  $j_1, j_2 \in W_i^*$  with  $j_1 \neq j_2$ , making  $g$  an injective function from  $W_i^*$  to  $S_i^*$ .

Thus,

$$|W_i^*| \leq |S_i^*|. \quad (3)$$

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# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

Since  $W^*$  induces an allocation and  $W_i^* \subseteq W^*$ ,

$$\sum_{j \in W_i^*} |S_j^*| \leq m. \quad (4)$$

...

# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

Recall inequalities (1), (2), (3), and (4):

$$\sum_{j \in W_i^*} v_j^* \stackrel{(1)}{\leq} \frac{v_i^*}{\sqrt{|S_i^*|}} \sum_{j \in W_i^*} \sqrt{|S_j^*|}, \quad |W_i^*| \stackrel{(3)}{\leq} |S_i^*|,$$

$$\sum_{j \in W_i^*} \sqrt{|S_j^*|} \stackrel{(2)}{\leq} \sqrt{|W_i^*|} \sqrt{\sum_{j \in W_i^*} |S_j^*|}, \quad \sum_{j \in W_i^*} |S_j^*| \stackrel{(4)}{\leq} m.$$

With these, we get (5):

$$\begin{aligned} \sum_{j \in W_i^*} v_j^* &\stackrel{(1)}{\leq} \frac{v_i^*}{\sqrt{|S_i^*|}} \sum_{j \in W_i^*} \sqrt{|S_j^*|} \stackrel{(2)}{\leq} \frac{v_i^*}{\sqrt{|S_i^*|}} \sqrt{|W_i^*|} \sqrt{\sum_{j \in W_i^*} |S_j^*|} \\ &\stackrel{(3)}{\leq} \frac{v_i^*}{\sqrt{|S_i^*|}} \sqrt{|S_i^*|} \sqrt{\sum_{j \in W_i^*} |S_j^*|} \stackrel{(4)}{\leq} \frac{v_i^*}{\sqrt{|S_i^*|}} \sqrt{|S_i^*|} \sqrt{m} = \sqrt{m} v_i^*. \quad \dots \end{aligned}$$

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# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

Recall that for  $i \in W$ ,

$$W_i^* = \{j \in W^* \mid j \geq i \text{ and } S_i^* \cap S_j^* \neq \emptyset\}.$$

Let  $j \in W^*$ .

- If  $j \in W$ : then by definition,  $j \in W_j^*$  (assuming, WLOG,  $S_j^* \neq \emptyset$ ).
- If  $j \notin W$ : then there must be some  $i \in W$  such that  $j \geq i$  and  $S_i^* \cap S_j^* \neq \emptyset$ , i.e.,  $j \in W_i^*$ .

Therefore, for each  $j \in W^*$ , there is an  $i \in W$  such that  $j \in W_i^*$ :

$$W^* \subseteq \bigcup_{i \in W} W_i^*. \quad \dots \quad (6)$$

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# Greedy Mechanism for Single-Minded Bidders: Approximation Factor



## Proof (ctd.)

Recall (5) and (6):

$$\sum_{j \in W_i^*} v_j^* \stackrel{(5)}{\leq} \sqrt{m} v_i^*, \quad W^* \stackrel{(6)}{\subseteq} \bigcup_{i \in W} W_i^*.$$

With these, we finally obtain the desired estimation

$$\sum_{i \in W^*} v_i^* \stackrel{(6)}{\leq} \sum_{i \in W} \sum_{j \in W_i^*} v_j^* \stackrel{(5)}{\leq} \sum_{i \in W} \sqrt{m} v_i^* = \sqrt{m} \sum_{i \in W} v_i^*.$$

Thus, the social welfare of  $W$  differs from the optimal social welfare by a factor of at most  $\sqrt{m}$ . □

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Summary

The following theorem summarizes the results in this chapter:

## Theorem

The greedy mechanism for single-minded bidders is efficiently computable, incentive compatible, and leads to an allocation whose social welfare is a  $\sqrt{m}$ -approximation of the optimal social welfare.







- In **combinatorial auctions**, bidders bid for **bundles of items**.
- **Exponential space** needed just to represent and communicate valuations.
- **Therefore:** Focus on **special case of single-minded bidders** (compact representation of valuations).
- **Unfortunately**, still, **optimal allocation NP-hard**.
- **Solution:** **approximate** optimal allocation.
- Polynomial-time approximation possible for approximation factor no better than  $\sqrt{m}$ .
- **Greedy mechanism for single-minded bidders:**
  - achieves  **$\sqrt{m}$ -approximation** of social welfare,
  - is **efficiently computable**, and
  - is **incentive compatible**.