

# L09 Introduction to Mechanism Design

CS 280 Algorithmic Game Theory  
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Inspired by  
by J. Hartline and T. Roughgarden notes

# Warm-up: Single-item allocation

**Definition (Single-item).** *The single-item allocation problem is given by*

- *a single **indivisible** item,*
- *$n$  agents **competing** for the item,*
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Example:

- Ask the agents/bidders to *report their values (bids)*, each agent reports  $b_i$ .
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Outcome of mechanism is unpredictable, hard to reason about performance

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*Proof.* Consider  $v_1 = 1$  and  $v_i = \epsilon$  for  $i \geq 2$ . Expected surplus is

$$\frac{1}{n} (1 + (n - 1)\epsilon).$$

Thus approximation ratio is  $\frac{n}{1+(n-1)\epsilon} \rightarrow n$  as  $\epsilon \rightarrow 0$ .

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**Charge payments, proportionally to the agents' bids => Discourage low-valued agents from making high bids.**

# Single-item auctions

**Approach 2 (First-price auction).** *The first-price auction is defined:*

- *Agents report their bids  $b_i$ .*
- *Select agent  $i^* = \arg \max_i b_i$  (highest bid).*
- *$i^*$  gets the item and pays the amount of  $b_{i^*}$ .*

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First-price auctions are **hard to reason** about.  
As a participant, it's hard to figure out **how to bid**.  
As an auction designer, it's **hard to predict** what  
will happen.

# Single-item auctions

**Approach 3 (Second-price auction).** *The second-price or Vickrey auction is defined:*

- *Agents report their bids  $b_i$ .*
- *Let agent  $i^* = \arg \max_i b_i$  and let  $j^*$  be the agent with (second) highest bid.*
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**Theorem (Vickrey is truthful).** *In second price auctions, every bidder  $i$  has a dominant strategy: Set her bid  $b_i$  equal to her private  $v_i$  (report truthfully). Dominant means the utility of bidder  $i$  is maximized no matter what other bidders do.*

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**Remark:** Utility of bidder  $i$  is  $u_i := v_i - p_i$  **if he gets the item** and  $u_i := 0$  **otherwise.**



# Second price auctions

*Proof.* Fix an agent  $i$  and set  $B = \max_{j \neq i} b_j$ .

Consider the cases:

- If  $b_i < B$  then  $i$  gets utility 0.
- If  $b_i \geq B$  then  $i$  wins the item and  $u_i = v_i - B$ .

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Assume that  $v_i < B$ .

- If  $i$  reports truthfully, she gets utility 0 (did not win the item).
- Assume not and  $b_i < v_i$  then the utility of  $i$  will still be 0.
- Assume not and  $B > b_i > v_i$  then the utility will still be 0.
- Assume not and  $b_i \geq B > v_i$  then the utility will be negative.

# Second price auctions

*Proof cont.* Fix an agent  $i$  and set  $B = \max_{j \neq i} b_j$ .

Assume that  $v_i \geq B$ .

- If  $i$  reports **truthfully**, she gets utility  $v_i - B \geq 0$  (won the item).
- Assume not and  $b_i > u_i$  then the utility of  $i$  will still be  $v_i - B \geq 0$ .
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No matter what other bidders do, truthtelling is best strategy (**Dominant strategy**).

# A general approach

An auction should satisfy following properties:

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- (Computational Efficiency) The auction can be implemented in **polynomial time**.

# An Example

**Problem:** Consider a society of  $n$  citizens and public good  $G$ .

- Each agent has (**private**) valuation  $v_i$  for the good.
- Cost of building  $G$  is (**publicly known**)  $C$ .
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**Question:** Allocating the cost  $C$  equally works? Why not?

**Answer:** For citizen  $i$ , if  $v_i > \frac{C}{n}$ ,  $i$  should report  $C + \epsilon$  so  $G$  will be built!

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**Answer:**

**No DSIC!**

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**Solution:** Charge citizen  $i$  the amount  $p_i := \max(0, C - \sum_{j \neq i} v_j)$ .  
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**Definition (Quasi-linear environments).** Also known as Vickrey-Groves-Clark (VCG) environments:

- $n$  agents,
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- Each agent  $i$  has a valuation  $v_i : \mathcal{X} \rightarrow \mathbb{R}^+$ ,
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Remark: This framework is called mechanism design **with money**.

# VCG mechanisms

**Definition (VCG mechanism).** *The family of mechanisms is defined as follows:*

- *Agents have valuations  $v_i$  and report their bids  $b_i$ .*
- *Set  $x^* = \arg \max_{x \in \mathcal{X}} \sum_{i=1}^n b_i(x)$ .*
- *Each agent pays  $p_i := \underbrace{h_i(b_{-i})}_{\text{without } i} - \underbrace{\sum_{j \neq i} b_j(x^*)}_{\text{with } i}$*
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**Theorem (VCG is DSIC).** *Every VCG mechanism is DSIC.*

*Proof.* Fix an agent  $i$  and let  $x^* = \arg \max \sum_{i=1}^n v_i(x)$ . Assume that  $i$  reports  $b_i \neq v_i$  and  $x'$  be the maximum if  $i$  reports  $b_i$ .

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By definition of  $x^*$   $u_i \geq u'_i$

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**Theorem (VCG is DSIC).** Every VCG mechanism is DSIC.

*Proof.* Fix an agent  $i$  and let  $x^* = \arg \max \sum_{i=1}^n v_i(x)$ . Assume that  $i$  reports  $b_i \neq v_i$  and  $x'$  be the maximum if  $i$  reports  $b_i$ . Observe that

$$\begin{aligned} u_i &= v_i(x^*) + \sum_{j \neq i} v_j(x^*) - h_i(v_{-i}) \text{ if } i \text{ reports } v_i \text{ and} \\ u'_i &= v_i(x') + \sum_{j \neq i} v_j(x') - h_i(v_{-i}) \text{ if } i \text{ reports } b_i. \end{aligned}$$

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Remark: For single-item, this is the **second-price** auction!

VCG might not be efficiently computable (e.g., **combinatorial auctions**)