# Mechanism design

Lecture 1: Introduction, VCG, single parameter settings

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# A simple mechanism design setting: Single item auction

- A seller has a single item to sell
- There are *n* potential buyers, aka bidders, players, agents
- Buyer i has value  $v_i$  for the item

If bidder *i* acquires the item at a price  $p_i$ , their utility will be  $u_i = v_i - p_i$ . This is called a quasilinear utility setting.

The essential difficulty of auctions and more generally of mechanism design is that **the values**  $v_i$  are **private**.

Thus, a mechanism must elicit these values and compute

- the outcome, i.e., who gets the item
- the payment of each bidder

## **Examples of single item auctions**

- English or Ascending auction
  - price starts at 0 and goes up
  - as the price goes up, bidders drop out
  - last bidder to remain gets the item and pays the current price
- Dutch or Descending auction
  - price starts at infinity and goes down
  - first bidder to accept the price wins the item and pays the price
- Sealed-bid first-price auction
  - bidders submit their bids in sealed envelopes
  - highest bidder gets the item and pays the highest bid
- Sealed-bid second-price auction
  - bidders submit their bids in sealed envelopes
  - highest bidder gets the item and pays the second highest bid

First-price auction  $\leftrightarrow$  Dutch auction Second-price auction  $\leftrightarrow$  English auction

# First-price auction

- v<sub>i</sub> value of bidder i
- $b_i$  bid of bidder i, not necessarily equal to  $v_i$
- p<sub>i</sub> payment of bidder i
- $u_i = v_i p_i$  utility of bidder i

In the **first-price auction**, the item is given to the bidder with the maximum bid, who pays their bid.

An auction induces a game between the bidders. This is usually an incomplete information game.

What do the players know? Two common settings:

- players have complete information; they know the values of all bidders
- Bayesian setting, in which values come from known probability distributions: v<sub>i</sub> ~ F<sub>i</sub>; player i knows v<sub>i</sub> and F<sub>1</sub>,..., F<sub>n</sub>.

# First-price auction – complete information example

Two bidders with values  $v_1 = 4$  and  $v_2 = 7$  participate in a first-price auction.

(Let's assume that the bids must be positive integers and in case of a tie, the item is given to bidder 1.)

This defines a  $4 \times 7$  matrix game. What are its **Nash equilibria**? It has a few of them. For example:

- $(b_1, b_2) = (4, 5)$ , which gives utilities  $(u_1, u_2) = (0, v_2 b_2) = (0, 2)$  or
- $(b_1, b_2) = (3, 4)$ , which gives utilities  $(u_1, u_2) = (0, v_2 b_2) = (0, 3)$

Is there a dominant strategy equilibrium?

# First-price auction – Bayesian example

Two bidders with values drawn independently from the [0,1] uniform distribution.

In the **Bayesian setting**, the appropriate equilibrium concept is **Bayes-Nash equilibrium**, in which deviations do not increase the expected utility.

It can be shown that  $(b_1,b_2)=(\frac{v_1}{2},\frac{v_2}{2})$  is a Bayes-Nash equilibrium.

# Dominant strategies and truthful auctions

In auctions and more generally in mechanism design, it is desirable to move beyond Nash equilibria and consider (weakly) dominant-strategy equilibria.

#### Definition

A mechanism is **truthful** (or incentive compatible) if bidding truthfully is a weakly dominant strategy equilibrium.

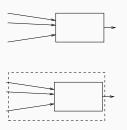
In the jargon of auctions, this is also known as **DSIC** — dominant strategy incentive compatible.

In these lectures, we consider DSIC mechanisms.

# Dominant strategies and truthful auctions

We want **truthful bidding**, i.e.  $b_i = v_i$ , to be a weakly dominant strategy for a few reasons:

- Bidding truthfully remains a dominant strategy even when a bidder has incomplete information about the values of the other bidders
- It makes it easier for bidders to compute their best strategy
- The **revelation principle** works :: every mechanism is equivalent (same allocation, same payments) to a truthful mechanism



# Second-price (Vickrey) auction

Are there any interesting truthful auctions?

The first-price auction is not truthful. In particular, the highest bidder has no reason to bid higher than the second highest bid.

# Second-price (Vickrey) auction

#### Theorem

The second-price auction is truthful.

#### Proof.

The payment for bidder i is  $p_i = \max_{j \neq i} b_j$ .

If  $b_i \ge p_i$  then bidder i wins and gains  $u_i = v_i - p_i$ . Otherwise  $u_i = 0$ .

Bidder i selects  $b_i$  to maximize their utility:

$$u_i = \max(v_i - p_i, 0).$$

So,

- if  $v_i p_i \ge 0$ , the bidder should bid any value greater than  $p_i$ ; in particular  $b_i = v_i$  is optimal
- otherwise the utility will be  $u_i = 0$ , so by bidding  $b_i < p_i$ , the bidder will lose the item and achieve utility 0; in particular  $b_i = v_i$  is optimal.

# Truthfulness in second-price auction

## Why is the second-price auction truthful?

- The payment depends only on the allocation and the values of the other players
- The allocation is monotone: increasing the declared value makes it more likely to get the item

## A look at truthfulness

Consider one bidder with value v for an item. Let

- b be the bid, the value that the bidder declares
- a(b) be the probability or fraction that the bidder gets
- p(b) be the payment
- The utility of the bidder is

$$u(b|v) = a(b) \cdot v - p(b)$$

• For which functions a and p is the mechanism truthful? That is, when

$$u(v|v) = \sup_{b} u(b|v)?$$

## A look at truthfulness

#### Theorem

A mechanism is truthful if and only if

- the utility u(v) = u(v|v) of the bidder is a convex function of the private value v.
- the probability of getting the item is given by

$$a(v) = u'(v)$$

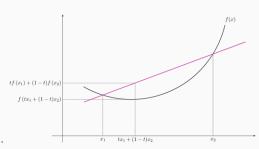
Note: no mention of payments!

A digression: convexity

# Convexity

Definition: A function  $f: \mathbb{R}^n \to \mathbb{R}$  is called convex when

$$\lambda f(x) + (1 - \lambda)f(y) \ge f(\lambda x + (1 - \lambda)y)$$



for every x, y and  $\lambda \in [0, 1]$ .

# The three layers of convexity

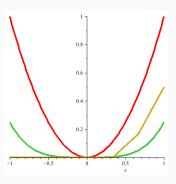
We focus on functions of one variable, but everything generalizes appropriately to many variables.

The following are equivalent (for doubly differentiable functions)

- 1. f(x) is convex
- 2. f'(x) is monotone (nondecreasing)
- 3. f''(x) is nonnegative

# **Examples of convex functions**

- x<sup>2</sup>
  \frac{1}{4}x<sup>4</sup>
- $\max\{0, \frac{x}{2} \frac{1}{6}, x \frac{1}{2}\}$



# Important properties of convex functions

## Proposition

For every function g, the function f defined by

$$f(x) = \sup_{y} \{x \cdot y - g(y)\},\$$

is convex.

## **Proposition**

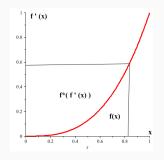
For every convex function f, there exists a function  $f^*$  (called the conjugate of f), such that

$$f(x) = \sup_{y} \{f'(y) \cdot x - f^{*}(f'(y))\}$$

## Conjugate

The conjugate function  $f^*$  of a function f is defined by

$$f^*(y) = \sup_{x} \{x \cdot y - f(x)\}$$



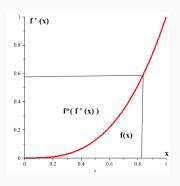
$$y = f'(x)$$

Notice the symmetry

$$f^*(y) = \sup_{x} \{x \cdot y - f(x)\}$$
 
$$x \leftrightarrow y$$
  
$$f(x) = \sup_{y} \{x \cdot y - f^*(x)\}$$
 
$$f \leftrightarrow f^*$$

# **Example**

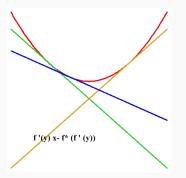
- $f(x) = \frac{1}{4}x^4$
- $f'(x) = x^3$
- $f^{*'}(x) = x^{1/3}$
- $f^*(x) = \frac{3}{4}x^{4/3}$
- $f'(x) \cdot x f^*(f'(x)) = x^3 \cdot x \frac{3}{4}(x^3)^{4/3} = f(x)$



# Supporting hyperplanes

For every convex function f, the conjugate function  $f^*$  defines the supporting hyperplanes

$$f(x) = \sup_{y} \{f'(y) \cdot x - f^{*}(f'(y))\}$$



Example:

$$f(x) = \frac{1}{4}x^4 = \sup_{y} \left\{ y^3 \cdot x - \frac{3}{4}y^4 \right\}.$$

## A look at truthfulness

Putting everything together for truthful mechanisms:

- the **utility** of the agent is  $u(v) = \operatorname{argmax}_b\{a(b) \cdot v p(b)\}$ , which is convex even if the mechanism is not truthful.
- Convexity implies  $u(v) = \sup_{v} \{u'(v) \cdot v u^*(u'(y))\}$
- The allocation is a(v) = u'(v), and
- the **payment** is  $p(v) = u^*(u'(v))$ , where  $u^*$  is the conjugate of u.

We can add a constant to the payment without affecting the argmax:  $p(v) = u^*(u'(v)) + \text{const.}$ 

# Back to truthfulness

# Truthfulness and monotonicity

**Monotonicity for the single-item auction:** For any fixed bids of the other bidders, when bidder i increases their bid  $b_i$ , the chances of getting the item cannot decrease

- The second-price auction is truthful and allocates the item to the agent with the highest value.
- Is there a truthful mechanism to allocate it to the agent with median (or minimum) value?
   No, because the allocation function is not monotone.

# The general mechanism design framework domains and objectives

# Beyond single-item auctions

- k-unit auction: there are k copies of an item and each bidder wants a single copy.
- combinatorial auctions: there are m items and each bidder i has a private valuation  $v_i(S)$  for every subset S of these items.
- general mechanism design setting: there is a set A of possible outcomes and each bidder has a private valuation  $v_i(a)$  for every  $a \in A$ .

# General mechanism design setting

## Definition (Mechanism design setting)

We can define a mechanism design problem by:

- the set of bidders or agents. Wlog we assume it to be  $\{1, 2, ..., n\}$ .
- the set A of possible outcomes
- sets of valuation functions  $V_1 \times \cdots \times V_n$ , one for each bidder. Each element  $v_i \in V_i$  determines the value of bidder i for each outcome:  $v_i : A \to \mathbb{R}$ .

## Example (Two voters, three candidates)

	Alice	Bob	Carol
Voter 1	10	18	20
Voter 2	21	18	12

# Single-item auction is the general mechanism design setting

## **Example (Single-item auction)**

Suppose that  $v_1^*, \dots, v_n^*$  denote the values of the bidders in a single-item auction. Then

- The set of outcomes is  $A = \{1, 2, ..., n\}$ : which bidder gets the item.
- The valuation functions are of the form

$$v_i(a) = \begin{cases} v_i^* & a = i \\ 0 & \text{otherwise} \end{cases}$$

	01	02	03
Player 1	$v_1$	0	0
Player 2	0	<i>V</i> <sub>2</sub>	0
Player 3	0	0	<i>V</i> 3

## Combinatorial auction

## **Example (Combinatorial auction)**

Suppose that  $v_1^*(S), \ldots, v_n^*(S)$  denote the valuation functions of the bidders in a combinatorial auction of m items. Then

- The set of outcomes A contains all allocation functions of m items to n bidders. Such an allocation can be represented by a legal 0-1 matrix  $a_{i,j}$ ,  $i \in [n]$ ,  $j \in [m]$ , where  $a_{i,j} = 1$  if and only if bidder i gets item j.
- The valuation functions are of the form

$$v_i(a) = v_i^*(\{j: a_{i,j} = 1\}).$$

## Example (3 players, 2 items)

	01	02	03	04	O5	O6	07	80	09
Agent 1	$v_1^*(12)$	$v_1^*(1)$	$v_1^*(1)$	$v_1^*(2)$	$v_1^*(2)$	0	0	0	0
Agent 2	Ò	$v_{2}^{*}(2)$	0	$v_{2}^{*}(1)$	0	$v_2^*(12)$	$v_2^*(1)$	$v_2^*(2)$	0
Agent 3	0	0	$v_3^*(2)$	0	$v_3^*(1)$	Ò	$v_3^*(2)$	$v_3^*(1)$	$v_3^*(12)$

# Mechanisms in the general setting

#### **Definition**

Fix a mechanism design setting with n bidders, set of outcomes A, and set of valuation functions  $V = V_1 \times \cdots \times V_n$ .

A (direct revelation) mechanism consists of two parts:

- a social choice function  $f: V \to A$
- a vector of payment functions  $p = (p_1, \dots, p_n)$ , where  $p_i \colon V \to \mathbb{R}$ .
- Each bidder *i* provides its valuation function  $v_i \in V_i$ .
- The outcome is determined by the social choice function f and the payments of the bidders are determined by the payment functions.
- The utility of bidder i is  $u_i(v) = v_i(f(v)) p_i(v)$ .

## **Definition (Truthful)**

A mechanism (f, p) is called **truthful or incentive compatible** if for every player i, every  $v \in V$  and every  $v'_i \in V_i$ :

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

# Single-parameter domains

An important special class of mechanism design is the **single-parameter setting**.

- The private value of a bidder i is a single real value,  $v_i^* \in \mathbb{R}$
- For every outcome a, the value of the bidder is proportional to  $v_i^*$ , i.e.,  $v_i(a) = \lambda_i(a)v_i^*$ , for some  $\lambda_i$ .

## **Example (Examples)**

- Shortest path problem on a graph, where every edge e belongs to some agent who is willing to sell it at a price.
- A multi-unit auction where the value of bidder i for k items is  $(2k^2-1)v_i^*$ .

# Vickrey-Clarke-Groves (VCG) mechanism

For a given outcome, the sum of the values of all bidders is called **social** welfare, i.e., the social welfare for outcome a is  $\sum_{i \in [n]} v_i(a)$ .

The VCG mechanism is a truthful mechanism, which selects the outcome that maximizes the social welfare. For example, for the single-item auction, it allocates the item to the bidder with the highest value.

## Definition (Vickrey-Clarke-Groves (VCG) mechanisms)

The VCG mechanism has

- $f(v) = \operatorname{argmax}_{a \in A} \sum_{i \in [n]} v_i(a)$
- $p_i(v) = -\sum_{j \neq i} v_j(f(v)) + h_i(v_{-i})$ , for some  $h_i : V_{-i} \to \mathbb{R}$ .

# The VCG mechanism

# Vickrey-Clarke-Groves (VCG) mechanism

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- $f(v) = \operatorname{argmax}_{a \in A} \sum_{i \in [n]} v_i(a)$
- $p_i(v) = -\sum_{i \neq i} v_j(f(v)) + h_i(v_{-i})$ , for some  $h_i : V_{-i} \to \mathbb{R}$ .

Note that the payments of VCG are not completely determined.

Choosing  $h_i(v_{-i}) = \max_{a \in A} \sum_{j \neq i} v_j(a)$  is called the **Clarke pivot rule**. With these payments, one can interpret the VCG as the mechanism that each bidder pays their value minus a discount equal to the increase of the social welfare due to their participation in the mechanism.

$$p_i(v) = v_i(f(v)) - \left(\sum_{j \in [n]} v_j(f(v)) - \max_{a \in A} \sum_{j \neq i} v_j(a)\right)$$

# **Examples of the VCG mechanism**

	Alice	Bob	Carol
Bidder 1	10	18	20
Bidder 2	21	18	12

VCG selects Bob (his column has the maximum sum).

Bidder 1 pays 18-(36-21)=3

Bidder 2 pays 18-(36-20)=2

# Examples of the VCG mechanism

## **Example (Second-price auction)**

The second price auction is VCG with Clarke pivot rule.

The social welfare of the VCG outcome is equal to the highest value. For simplicity, assume that  $v_1 \geq v_2 \cdots \geq v_n$ . Then the social welfare is  $v_1$ . If the winner does not participate, the social welfare will drop to  $v_2$ , so the winner pays their value  $(v_1)$  minus a discount  $v_1 - v_2$ ; so the payment is  $v_1 - (v_1 - v_2) = v_2$ .

#### **Example (multi-unit auction)**

There are  $k \ge 1$  identical units of a good and each bidder wants a single one of them.

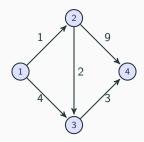
VCG (with Clarke pivot rule) will give the k items to the k highest bids and each one of them will pay the (k+1)-st highest bid.

#### **Example (Buying a shortest path)**

Given a graph in which every edge is controlled by a different seller, we want to buy a shortest path from some vertex s to some other vertex t.

Note that this is an inverse auction (procurement): the bidders are sellers and the auctioneer a buyer.

The VCG mechanism will select the shortest path. Each bidder will get their value plus the increase in the length of the shortest path when we remove their edge.



- VCG selects a shortest path P: P = (1, 2, 3, 4)
- Edges not in *P* are paid nothing
- To compute the payment of an edge e on the path P:
  - We remove e and compute a shortest path  $P_e$
  - The payment for edge e is

$$p_e = v_e + \text{length of } P_e - \text{length of } P$$

For example,

- for edge [1,2],  $P_e = (1,3,4)$ . The payment is 1+7-6=2
- for edge [2,3],  $P_e = (1,3,4)$ . The payment is 2+7-6=3

#### **Example (Public project)**

A city wants to undertake a public project with cost C, for example to build a new road. There are n citizens/bidders. Bidder i will get benefit  $v_i$  from the project.

The city will undertake the project when the sum of reported values exceeds C, i.e.,  $\sum_{i \in [n]} b_i \ge C$ . The social welfare will be  $\sum_{i \in [n]} v_i - C$  if the road is built, and 0 otherwise.

If we use the VCG mechanism, the payment of bidder i will be 0, unless bidder i is critical. A bidder is critical if the total bids of the other bidders is below C, but together with  $b_i$  the sum is above C. In this case, the bidder will pay  $C - \sum_{j \neq i} b_j$ .

Note that this solution has some undesirable properties: it is **not budget-balanced** (i.e., the total of payments is less than C in general), and it is susceptible to collusion (e.g., if two bidders report that their values are C, the project will be built, and they will pay nothing).

#### Truthfulness of the VCG

#### Theorem

VCG is truthful.

#### Proof.

Fix some player i,  $v_{-i}$ , and  $v_i$ . If  $b_i$  is the bid of bidder i, its utility is

$$u_i(b_i, v_{-i}) = v_i(f(b_i, v_{-i})) - p(b_i, v_{-i}).$$

We want to show that  $b_i = v_i$  maximizes this expression. By the definition of VCG,  $p(b_i, v_{-i}) = -\sum_{j \neq i} v_j(f(b_i, v_{-i})) + h_j(v_{-i})$ , so

$$u_i(b_i, v_{-i}) = \sum_{j \in [n]} v_j(f(b_i, v_{-i})) + h_i(v_{-i}).$$

Note that the term with  $h_i$  is not affected by  $b_i$ , so bidder i wants to select  $b_i$  that maximizes the social welfare  $\sum_{j \in [n]} v_j(f(b_i, v_{-i}))$ . But, by the definition of VCG, this is maximized when  $b_i = v_i$ .

It follows from the proof that VCG aligns the interests of all bidders with the objective of the mechanism (i.e., to maximize the social welfare).

### Why VCG is not always the answer

VCG is a truthful mechanism that can be applied to **every mechanism design setting**.

Why do we need to search for other mechanisms?

- The computational or communication complexity of VCG may be prohibitive (e.g. combinatorial auctions).
- VCG optimizes the social welfare. But in many cases, the objective may be different.
  - For example, a usual objective is to maximize revenue. The theory of optimal auctions tries to maximize revenue when the mechanism designer knows the probability distributions of the values of bidders.
  - Another example: in the scheduling problem, there are n machines (bidders) and m tasks, and we want a mechanism to allocate the tasks to minimize the makespan. The objective here (makespan) is different than the social welfare.
- Payments may not be allowed (e.g. in voting), or payments may have to satisfy certain conditions (e.g., in the public project setting, we require **budget balance**).

There are n machines (agents) with speeds  $s_1, \ldots, s_n$  which are private values. There is also a set of tasks T to be scheduled on the machines. The mechanism consists of the allocation function  $a(s_1, \ldots, s_n)$  that allocates the tasks to the machines and the payment functions  $p_1(s_1, \ldots, s_n), \ldots, p_n(s_1, \ldots, s_n)$ .

Greedy algorithm = allocate one-by-one the items to optimize the makespan myopically.

**Greedy is not truthful:** Take two machines with the first slightly faster than the second and jobs  $2, 1+\epsilon, 1+\epsilon$ . The first (fast) machine will get the first job, while the second will get the other two tasks.

#### Two things to note:

- each set of tasks defines a different mechanism design setting
- the objective is to minimize the makespan, i.e., the time when every task has finished.

Therefore VCG, which maximizes the total welfare (= the sum of completion times), may not be optimal.

Actually, VCG has approximation ratio n (equal to the number of machines).

#### Theorem

There exists a truthful mechanism with optimal makespan.

Algorithm: return the **lexicographically minimum** among the optimal allocations.

Let  $w = (w_1, ..., w_n)$  be the load assigned to the machines. w is lexicographically smaller than w' if there is k such that

$$w_i = w_i'$$
 for  $i < k$ , and  $w_k < w_k'$ 

This algorithm achieves optimal makespan, but it is not a polynomial time algorithm. However, there is a monotone PTAS.

# Selling digital goods

# Selling digital goods

Formally, we have n unit-demand bidders with valuations  $v_1, \ldots, v_n$  and n identical items.

Suppose that  $v_1 \ge v_2 \ge \cdots \ge v_n$  and assume we know them. If we want to maximize revenue with the **same price for all bidders**, we should select  $p = v_i$ , where  $v_i$  maximizes  $iv_i$ .

What can we do if we don't know the values?

For every bidder i, a truthful mechanism should make a take-it-or-leave-it offer of some price  $p_i$ .

#### **Example (Random Sampling Optimal Pricing (RSOP))**

The bidders are uniformly partitioned into two parts, and the optimal single price of each part (i.e.,  $\operatorname{argmax}_{v_i}\{i\cdot v_i\}$ ) is offered to the bidders of the other part.

This is a **prior-free mechanism**: the designer does not assume anything about the values.