# 6.254 : Game Theory with Engineering Applications Lecture 19: Mechanism Design I

Asu Ozdaglar MIT

April 29, 2010

# Outline

- Mechanism design
- Revelation principle
  - Incentive compatibility
  - Individual rationality
- "Optimal" mechanisms
- Reading:
- Krishna, Chapter 5
- Myerson, "Optimal Auction Design," *Mathematics of Operations Research*, vol. 6, no. 1, pp. 58-73, 1981.

#### Introduction

- In the next 3 lectures, we will study Mechanism Design, which is an area in economics and game theory that has an engineering perspective.
- The goal is to design economic mechanisms or incentives to implement desired objectives (social or individual) in a strategic setting-assuming that the different members of the society each act rationally in a game theoretic sense.
- Mechanism design has important applications in economics (e.g., design of voting procedures, markets, auctions), and more recently finds applications in networked-systems (e.g., Internet interdomain routing, design of sponsored search auctions).

# Auction Theory Viewpoint

- We first study the mechanism design problem in an auction theory context, i.e., we are interested in allocating a single indivisible object among agents.
- An auction is one of many ways that a seller can use to sell an object to potential buyers with unknown values.
- In an auction, the object is sold at a price determined by competition among buyers according to rules set by the seller (auction format), but the seller can use other methods.
- The question then is: what is the "best" way to allocate the object?
- Here, we consider the underlying allocation problem by abstracting away from the details of the selling format.

#### Model

- We assume a seller has a single indivisible object for sale and there are N potential buyers (or bidders) from the set N = {1,..., N}.
- Buyers have private values X<sub>i</sub> drawn independently from the distribution F<sub>i</sub> with associated density function f<sub>i</sub> and support X<sub>i</sub> = [0, w<sub>i</sub>].
  - Notice that we allow for asymmetries among the buyers, i.e., the distributions of the values need not be the same for all buyers.
- We assume that the value of the object to the seller is 0.
- Let  $\mathcal{X} = \prod_{j=1}^{N} \mathcal{X}_j$  denote the product set of buyers' values and let  $\mathcal{X}_{-i} = \prod_{j \neq i} \mathcal{X}_j$ .
- We define f(x) to be the joint density of  $x = (x_1, \ldots, x_N)$ . Since values are independently distributed, we have  $f(x) = f_1(x_1) \times \cdots \times f_N(x_N)$ . Similarly, we define  $f_{-i}(x_{-i})$  to be the joint density of  $x_{-i} = (x_1, \ldots, x_{i-1}, x_{i+1}, \ldots, x_N)$ .

# Mechanism

- A selling mechanism  $(\mathcal{B}, \pi, \mu)$  has the following components:
  - A set of messages (or bids/strategies)  $\mathcal{B}_i$  for each buyer *i*,
  - An allocation rule  $\pi : \mathcal{B} \to \Delta$ , where  $\Delta$  is the set of probability distributions over the set of buyers  $\mathcal{N}$ ,
  - A payment rule  $\mu : \mathcal{B} \to \mathbb{R}^N$ .
- An allocation rule specifies, as a function of messages  $b = (b_1, \ldots, b_N)$ , the probability  $\pi_i(b)$  that *i* will get the object. Similarly, a payment rule specifies the payment  $\mu_i(b)$  that *i* must make.
- Every mechanism defines a game of incomplete information among the buyers.
  - Strategies:  $\beta_i : [0, w_i] \rightarrow \mathcal{B}_i$
  - Payoffs: Expected payoff for a given strategy profile and selling mechanism
- A strategy profile β(·) is a Bayesian Nash equilibrium of a mechanism if for all *i* and for all x<sub>i</sub>, given the strategies β<sub>-i</sub> of other buyers, β<sub>i</sub>(x<sub>i</sub>) maximizes buyer *i*'s expected payoff.

#### Direct Mechanisms and Revelation Principle

- A mechanism could be very complicated since we made no assumptions on the message sets  $\mathcal{B}_i$ .
- A special class of mechanisms, referred to as direct mechanisms, are those for which the set of messages is the same as the set of types (or values), i.e., B<sub>i</sub> = X<sub>i</sub> for all *i*.
- These mechanisms are called "direct" since every buyer is asked directly to report a value.
- Formally a direct mechanism (Q, M) consists of the following components:
  - A function  $Q: \mathcal{X} \to \Delta$ , where  $Q_i(x)$  is the probability that *i* will get the object,
  - A function  $M : \mathcal{X} \to \mathbb{R}^N$ , where  $M_i(x)$  is the payment by buyer *i*.
- If it is a Bayesian Nash equilibrium for each buyer to report (or reveal) their type x<sub>i</sub> correctly, we say that the direct mechanism has a truthful equilibrium.
- We refer to the pair (Q(x), M(x)) as the outcome of the mechanism.

# **Revelation Principle**

- The following key result, referred to as the revelation principle, allows us to restrict our attention to direct mechanisms.
- More specifically, it shows that the outcomes resulting from any equilibrium of any mechanism can be replicated by a truthful equilibrium of some direct mechanism.

#### Proposition (Revelation Principle)

Given a mechanism  $(\mathcal{B}, \pi, \mu)$  and an equilibrium  $\beta$  of that mechanism, there exists a direct mechanism (Q, M), in which

(i) it is a Bayesian Nash equilibrium for each buyer to report his value truthfully,

(ii) the outcomes are the same as in equilibrium  $\beta$  of the original mechanism.

*Proof:* This follows simply by defining the functions  $Q: \mathcal{X} \to \Delta$  and  $M: \mathcal{X} \to \mathbb{R}^N$  as  $Q(x) = \pi(\beta(x))$ , and  $M(x) = \mu(\beta(x))$ . Instead of buyers submitting message  $b_i = \beta(x_i)$ , the mechanism asks the buyer to report their value and makes sure the outcome is the same as if they had submitted  $\beta_i(x_i)$ .

# **Revelation Principle**

The basic idea behind revelation principle is as follows:

- Suppose that in mechanism (B, π, μ), each agent finds that, when his type is x<sub>i</sub>, choosing β<sub>i</sub>(x<sub>i</sub>) is his best response to others' strategies.
- Then, if we have a mediator who says "Tell me your type  $x_i$  and I will play  $\beta_i(x_i)$  for you," each agent will find truth telling to be an optimal strategy given that all other agents tell the truth.
- In other words, a direct mechanism does the "equilibrium calculations" for the buyers automatically.

# Incentive Compatibility

• For a given direct mechanism (Q, M), we define

$$q_i(z_i) = \int_{\mathcal{X}_{-i}} Q_i(z_i, x_{-i}) f_{-i}(x_{-i}) dx_{-i},$$

to be the probability that *i* will get the object when he reports his value to be  $z_i$  and all other buyers report their values truthfully.

• Similarly, we define

$$m_i(z_i) = \int_{\mathcal{X}_{-i}} M_i(z_i, x_{-i}) f_{-i}(x_{-i}) dx_{-i}$$

to be the expected payment of i when his report is  $z_i$  and all other buyers tell the truth.

• The expected payoff of buyer *i* when his true value is *x<sub>i</sub>* and he reports *z<sub>i</sub>*, assuming all others tell the truth, can be written as

$$q_i(z_i)x_i-m_i(z_i).$$

# Incentive Compatibility

#### Definition

We say that the direct revelation mechanism (Q, M) is incentive compatible (IC) if

$$q_i(x_i)x_i - m_i(x_i) \ge q_i(z_i)x_i - m_i(z_i)$$
 for all  $i, x_i, z_i$ .

We refer to the left-hand side of this relation as the equilibrium payoff function denoted by  $U_i(x_i)$ , i.e.,

$$U_i(x_i) = \max_{z_i \in \mathcal{X}_i} \{q_i(z_i)x_i - m_i(z_i)\}.$$

Properties under IC:

- Since  $U_i$  is a maximum of a family of affine functions, it follows that  $U_i$  is a convex function.
- Moreover, it can be seen that incentive compatibility is equivalent to having for all z<sub>i</sub> and x<sub>i</sub>

$$U_i(z_i) \ge U_i(x_i) + q_i(x_i)(z_i - x_i).$$
 (1)

• This follows by writing for all z<sub>i</sub> and x<sub>i</sub>

$$q_i(x_i)z_i - m_i(x_i) = q_i(x_i)x_i - m_i(x_i) + q_i(x_i)(z_i - x_i) = U_i(x_i) + q_i(x_i)(z_i - x_i).$$

- Eq. (1) implies that for all  $x_i$ ,  $q_i(x_i)$  is a subgradient of the function  $U_i$  at  $x_i$ .
- Thus at every point that U<sub>i</sub> is differentiable,

$$U_i'(x_i) = q_i(x_i).$$

- Since  $U_i$  is convex, this implies that  $q_i$  is a nondecreasing function.
- Moreover, we have

$$U_i(x_i) = U_i(0) + \int_0^{x_i} q_i(t_i) dt_i.$$
 (2)

- This shows that, up to an additive constant, the expected payoff to a buyer in an IC direct mechanism (Q, M) depends only on the allocation rule Q.
- From the preceding relations, one can also infer that incentive compatibility is *equivalent to* the function q<sub>i</sub> being nondecreasing.

# Revenue Equivalence

The payoff equivalence derived in the previous slide leads to the following general revenue equivalence principle.

#### Proposition (Revenue Equivalence)

If the direct mechanism (Q, M) is incentive compatible, then for all i and  $x_i$ , the expected payment is given by

$$m_i(x_i) = m_i(0) + q_i(x_i)x_i - \int_0^{x_i} q_i(t_i)dt_i.$$

Thus the expected payments in any two IC mechanisms with the same allocation rule are equivalent up to a constant.

# Revenue Equivalence

Remarks:

- Given two BNE of two different auctions such that for each *i*:
  - For all  $(x_1, \ldots, x_N)$ , probability of *i* getting the object is the same,
  - They have the same expected payment at 0 value.

These equilibria generate the same expected revenue for the seller.

- This generalizes the result from last time:
  - Revenue equivalence at the symmetric equilibrium of standard auctions (object allocated to buyer with the highest bid).

### Individual Rationality (participation constraints)

- A seller cannot force a bidder to participate in an auction which offers him less expected utility than he could get on his own.
- If he did not participate in the auction, the bidder could not get the object, but also would not pay any money, so his payoff would be zero.
- We say that a direct mechanism (Q, M) is individually rational (IR) if for all *i* and  $x_i$ , the equilibrium expected payoff satisfies  $U_i(x_i) \ge 0$ .
- If the mechanism is IC, then from Eq. (2), individual rationality is equivalent to  $U_i(0) \ge 0$ .
- Since  $U_i(0) = -m_i(0)$ , individual rationality is equivalent to

 $m_i(0) \leq 0.$ 

# **Optimal Mechanisms**

- Our goal is to design the optimal mechanism that maximizes the expected revenue among all mechanisms that are IC and IR.
- Without loss of generality we can focus on direct revelation mechanisms.
- Consider the direct mechanism (Q, M).
- We can write the expected revenue to the seller as:

$$E[R] = \sum_{i \in N} E[m_i(X_i)], \quad \text{where} \\ E[m_i(X_i)] = \int_0^{w_i} m_i(x_i) f_i(x_i) dx_i \\ = m_i(0) + \int_0^{w_i} q_i(x_i) x_i f_i(x_i) dx_i - \int_0^{w_i} \int_0^{x_i} q_i(t_i) dt_i f_i(x_i) dx_i$$

• Changing the order of integration in the third term, we obtain

$$E[m_i(X_i)] = m_i(0) + \int_0^{w_i} (x_i - \frac{1 - F_i(x_i)}{f_i(x_i)}) q_i(x_i) f_i(x_i) dx_i$$
  
=  $m_i(0) + \int_{\mathcal{X}} (x_i - \frac{1 - F_i(x_i)}{f_i(x_i)}) Q_i(x) f(x) dx.$ 

# Optimal Mechanism Design Problem

• The optimal mechanism design problem can be written as

 $\begin{array}{ll} \text{maximize} & E[R] \\ \text{subject to} & IC(\Leftrightarrow q_i \text{nondecreasing}) + IR(\Leftrightarrow m_i(0) \leq 0) \end{array}$ 

• We define the virtual valuation of a buyer with value x<sub>i</sub> as

$$\Psi_i(x_i) = x_i - \frac{1 - F_i(x_i)}{f_i(x_i)}$$

- We say that the design problem is regular when the virtual valuation Ψ<sub>i</sub>(x<sub>i</sub>) is strictly increasing in x<sub>i</sub>.
- We next show that under this regularity assumption, we can without loss of generality neglect the IC and the IR constraints.
- The seller should choose Q and M to maximize

$$\sum_{i\in\mathcal{N}}m_i(0)+\int_{\mathcal{X}}(\sum_{i\in\mathcal{N}}\Psi_i(x_i)Q_i(x))f(x)dx.$$

# **Optimal Mechanism**

The following is an optimal mechanism:

• Allocation Rule:

$$Q_i(x) > 0 \Longleftrightarrow \Psi_i(x_i) = \max_{j \in \mathcal{N}} \Psi_j(x_j) \ge 0.$$

• Payment Rule:

$$M_i(x) = Q_i(x)x_i - \int_0^{x_i} Q_i(z_i, x_{-i})dz_i.$$

We finally show that this mechanism satisfies IC and IR.

- We have M<sub>i</sub>(0, x<sub>-i</sub>) = 0 for all x<sub>-i</sub> implying that m<sub>i</sub>(0) = 0, and therefore satisfying IR.
- By the regularity assumption, for any  $z_i < x_i$ , we have  $\Psi_i(z_i) < \Psi_i(x_i)$ . This implies that  $Q_i(z_i, x_{-i}) \leq Q_i(x_i, x_{-i})$  for all  $x_{-i}$ , and therefore  $q_i(z_i) \leq q_i(x_i)$ , i.e.,  $q_i$  is nondecreasing. Hence, IC is also satisfied.

# **Optimal Mechanism**

• The optimal expected revenue is given by

```
E[\max\{\Psi_1(x_1),\ldots,\Psi_N(x_N),0\}],
```

i.e., it is the expectation of the highest virtual valuation provided it is nonnegative.

• We define

$$y_i(x_{-i}) = \inf\{z_i | \Psi_i(z_i) \ge 0, \Psi_i(z_i) \ge \Psi_j(x_j) \text{ for all } j \ne i\},$$

i.e., it is the smallest value for *i* that wins against  $x_{-i}$ .

• Using this, we can write

$$Q_{i}(z_{i}, x_{-i}) = \begin{cases} 1 & if \quad z_{i} > y_{i}(x_{-i}) \\ 0 & if \quad z_{i} < y_{i}(x_{-i}) \end{cases}$$

# **Optimal Mechanism**

We have

$$\int_{0}^{x_{i}} Q_{i}(z_{i}, x_{-i}) = \begin{cases} x_{i} - y_{i}(x_{-i}) & \text{if } x_{i} > y_{i}(x_{-i}) \\ 0 & \text{if } x_{i} < y_{i}(x_{-i}) \end{cases}$$

implying that

$$M_{i}(x) = \begin{cases} y_{i}(x_{-i}) & \text{if } Q_{i}(x) = 1\\ 0 & \text{if } Q_{i}(x) = 0 \end{cases}$$

- This implies that:
  - Only the winning buyer pays,
  - He pays the smallest value that would result in his winning.

## Optimal Mechanism – Symmetric Case

- Suppose that distributions of values are identical across buyers, i.e., for all *i*, we have f<sub>i</sub> = f. This implies that for all *i*, we have Ψ<sub>i</sub> = Ψ.
- Note that in this case, we have

$$y_i(x_{-i}) = \max\{\Psi^{-1}(0), \max_{j \neq i} x_j\}.$$

#### Proposition

Assume that the design problem is regular and symmetric. Then a second price auction (Vickrey) with reservation price  $r^* = \Psi^{-1}(0)$  is an optimal mechanism.

 Note that, unlike first and second price auctions, the optimal mechanism is not efficient, i.e., object does not necessarily end up with the person who values it most.

# 6.254 Game Theory with Engineering Applications Spring 2010

For information about citing these materials or our Terms of Use, visit: http://ocw.mit.edu/terms.