

Module 17: Mechanism Design & Optimal Auctions

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Examples:

- Auctions
- Bilateral trade
- Production and distribution in society

General Setup

- N agents
- Each agent has private information θ_i ; $\theta = \{\theta_i\}_{i=1}^N$.
- Outcomes $y \in Y$; often allocation plus transfers: $y = \{k, t_1, \dots, t_N\}$.
- Utility $u_i = u_i(y, \theta)$
 - Quasi-linear utility: $u_i = u_i^k(\theta) - t_i$.
- Mechanism designer's objective: "Implement" a choice rule $\psi : \Theta \rightarrow Y$ to maximize objective; *e.g.*,
 - *Efficiency*: maximize $\sum_i u_i^k(\theta)$
 - *Revenue*: maximize $\mathbb{E}_\theta [\sum_i t_i(\theta)]$

Definition. A choice rule $\psi : \Theta \rightarrow Y$ is incentive compatible with respect to an equilibrium concept "X" if each agent revealing his type truthfully (*i.e.*, reporting $\tilde{\theta}_i = \theta_i$) is an "X"-equilibrium.

Equilibrium Concepts

1. Dominant-strategy (strategy-proof) implementation: For all i , θ_i , $\tilde{\theta}_i$, θ_{-i} and $\tilde{\theta}_{-i}$

$$u_i \left(\psi \left(\theta_i, \tilde{\theta}_{-i} \right); \theta \right) \geq u_i \left(\psi \left(\tilde{\theta}_i, \tilde{\theta}_{-i} \right); \theta \right)$$

- Reporting truthfully is an optimal strategy for each agent irrespective of the others' strategies.
- Quite restrictive.

2. Bayesian Nash implementation:

- There is a common prior π over θ , and the agents' beliefs $\pi_i(\cdot|\theta_i)$ over Θ_{-i} are given by Bayesian updating.
- For all i , θ_i and $\tilde{\theta}_i$

$$\mathbb{E}_{\pi_i(\cdot|\theta_i)} u_i \left(\psi \left(\theta_i, \theta_{-i} \right); \theta \right) \geq \mathbb{E}_{\pi_i(\cdot|\theta_i)} u_i \left(\psi \left(\tilde{\theta}_i, \theta_{-i} \right); \theta \right)$$

- Reporting truthfully is an optimal strategy on expectation, given beliefs $\pi_i(\cdot|\theta_i)$.

3. Ex-post implementation: For all i , θ_i , $\tilde{\theta}_i$ and θ_{-i}

$$u_i \left(\psi \left(\theta_i, \theta_{-i} \right); \theta \right) \geq u_i \left(\psi \left(\tilde{\theta}_i, \theta_{-i} \right); \theta \right)$$

- Ea. agent finds it optimal to report truthfully given that others also report truthfully - after others' types are revealed ("no regret").
- Advantage: Robust against different priors and higher order beliefs.

Revelation Principle

- Set of all mechanisms has little structure.
- Focus on a particular class of mechanism: Revelation mechanism $S_i = \Theta_i$; i.e., strategy is to state a type $\tilde{\theta}$.

Theorem. (*Revelation Principle for Bayesian Nash implementation*) A choice rule ψ is (partially) implementable by any mechanism if and only if it is incentive compatible.

- *Proof:* Skipped.

- Very robust result.
 - Holds for all standard implementation concepts.
- If agents control actions a_i on top of common decision ψ , then one can replace any mechanism with a centralized mechanism where
 - Each agent reports his type $\tilde{\theta}_i$; and
 - the mechanism designer recommends actions \tilde{a}_i .
 - In equilibrium, the agents are truthful $\tilde{\theta}_i = \theta_i$ and obedient ($a_i = \tilde{a}_i$).

i.e., Moral hazard together with adverse selection (*Myerson, Ecta '82*)
- If agents can act sequentially and acquire further information, then one can replace any mechanism with a centralized mechanism where
 - Agents report everything they have learned so far ; and
 - the mechanism designer recommends actions \tilde{a}_i .
 - In equilibrium, the agents are truthful and obedient.
- Not robust to:
 - Communication costs
 - Bounded rationality.
- Full vs. Partial implementation:
 - *Partial*: $\psi(\theta)$ is an equilibrium.
 - *Full*: $\psi(\theta)$ is the only equilibrium.

Optimal Auctions

- N bidders.
- $\theta \in [\underline{\theta}, \bar{\theta}]$ with pdf f .
- Mechanism specifies:
 1. Allocation function $p_i : [\underline{\theta}, \bar{\theta}]^N \rightarrow [0, 1]$ for each agent i such that $p_i \geq 0$ and $\sum_i p_i \leq 1$.
 - If the seller has n objects for sale, then $\sum_i p_i \leq n$.
 2. Transfer function $t_i : [\underline{\theta}, \bar{\theta}] \rightarrow \mathbb{R}$ for each agent i .
- Independent private values (IPV) model: $u_i(\theta_i) = \theta_i p_i - t_i$
- Revenue: $\sum_i t_i + (1 - \sum_i p_i) \theta_0$
 - θ_0 : seller's value. Can be shown that the seller can disclose θ_0 wolog.

Examples of Auctions

1. First-Price Auction: $p_i(\theta) = 1$ if $\theta_i > \theta_{-i}$, and $t_i(\theta_i) = p_i(\theta) b(\theta_i)$.
 - $b(\theta_i)$ is the bid of type θ_i .
 - Under symmetry assumptions.
 - Otherwise: Maskin and Riley (REStud, 2000)
2. Second-Price Auction: $p_i(\theta) = 1$ if $\theta_i > \theta_{-i}$, and $t_i(\theta_i) = p_i(\theta) b(\theta_{(2)})$.
 - $b(\theta_{(2)})$ is the second-highest bid.
3. All-pay Auction: $p_i(\theta) = 1$ if $\theta_i > \theta_{-i}$, and $t_i(\theta_i) = b(\theta_i)$.
4. Raffle: $n(\theta_i) = \#$ of tickets, $p(\theta) = \frac{n(\theta_i)}{\sum_j n(\theta_j)}$, and $t_i(\theta_i) = c n(\theta_i)$.

Revenue Maximization

$$\begin{aligned} \max \quad & \mathbb{E}_\theta \left[\sum_i t_i(\theta_i) + \left[1 - \sum_i p_i(\theta) \right] \theta_0 \right] \\ \text{s.t.} \quad & u_i(\theta_i; \theta_i) \geq 0 \\ & u_i(\theta_i; \theta_i) \geq u_i(\theta_i; \tilde{\theta}_i) \end{aligned}$$

where $u_i(\theta_i; \tilde{\theta}_i) = \mathbb{E}_{\theta_{-i}} \left[p_i(\tilde{\theta}_i, \theta_{-i}) \theta_i - t(\tilde{\theta}_i, \theta_{-i}) \right]$.

Proposition. *is IC if and only if*

1. $u_i(\theta_i; \theta_i) = u_i(\underline{\theta}; \underline{\theta}) + \int_{\underline{\theta}}^{\theta_i} \mathbb{E}_{\theta_{-i}} [p_i(s, \theta_{-i})] ds$ (IC-FOC)

2. $\mathbb{E}_{\theta_{-i}} [p_i(\theta_i, \theta_{-i})]$ increases in θ_i (Monotonicity)

(IR) can be replaced by $u(\underline{\theta}; \underline{\theta}) = 0$.

◦ *Proof:* Similar to the single-agent case.

◦ Re-write objective function:

$$\text{Revenue} = \mathbb{E}_\theta \left[\sum_i p_i(\theta) \theta_i + \left[1 - \sum_i p_i(\theta) \right] \theta_0 - \sum_i u_i(\theta_i; \theta_{-i}) \right]$$

◦ Calculate expected rent:

$$\begin{aligned} \mathbb{E}_{\theta_i} [u_i(\theta_i; \theta_{-i})] &= \underbrace{u_i(\underline{\theta}; \underline{\theta})}_{=0 \text{ (IR)}} + \int_{\underline{\theta}}^{\bar{\theta}} \int_{\underline{\theta}}^{\theta_i} \mathbb{E}_{\theta_{-i}} [p_i(s, \theta_{-i})] ds \underbrace{dF(\theta_i)}_{-[1-F(\theta_i)]' d\theta_i} \\ &= - \underbrace{[\mathbb{E}_{\theta_{-i}} [p(\theta_i, \theta_{-i})] [1 - F(\theta_i)]]_{\underline{\theta}}}_{=0}^{\bar{\theta}} + \int_{\underline{\theta}}^{\bar{\theta}} \mathbb{E}_{\theta_{-i}} [p_i(\theta_i, \theta_{-i})] [1 - F(\theta_i)] d\theta_i \\ &= \mathbb{E}_\theta \left[p_i(\theta) \frac{1 - F(\theta_i)}{f(\theta_i)} \right] \end{aligned}$$

◦ Compile:

$$\text{Revenue} = \mathbb{E}_\theta \left[\sum_i p_i(\theta) \left[\underbrace{\theta_i - \frac{1 - F(\theta_i)}{f(\theta_i)}}_{MR(\theta_i)} - \theta_0 \right] \right] + \theta_0$$

Proposition. (Revenue Equivalence): *Any auction that has the same allocation function, generates the same revenue.*

Proof.

- Revenue depends on $p(\cdot)$, but not on $t(\cdot)$.

□

- *Implication:* What matters is allocations; not “how you get there”.
- Optimal Auction:
 - Award good to agent i if $MR(\theta_i) > \max\{\theta_0, MR(\theta_{-i})\}$.
 - If $MR(\theta)$ increases in θ , then (Monotonicity) is satisfied, and we have an optimal auction. Otherwise, we need to “iron it”.

Implementation:

- First-price auction with reserve price $r = MR^{-1}(\theta_0)$.
- Second-price auction with entry fee $e = MR^{-1}(\theta_0) F^{N-1}(MR^{-1}(\theta_0))$.

Example:

- N bidders, $\theta_i \sim U[0, 1]$, $\theta_0 = 0$.
- $MR(\theta) = 2\theta - 1$.
- Award good to agent with highest value if $\theta \geq \frac{1}{2}$; *i.e.*, reserve price $r = \frac{1}{2}$.
- Note: $r > \theta_0$. Why? (By increasing the reserve price, the seller can reduce information rents.)

Deriving bidding strategies:

- Assume that bidding functions are (i) monotone, and (ii) symmetric.
- First-price auction:

$$\begin{aligned}
 u_i(\theta_i, \theta_i) &= \mathbb{E}_{\theta_{-i}} [(\theta_i - b(\theta_i)) p_i(\theta)] = F^{N-1}(\theta_i) [\theta_i - b(\theta_i)] \\
 u_i(\theta_i, \theta_i) &= \int_{\underline{\theta}}^{\theta_i} \mathbb{E}_{\theta_{-i}} [p(s, \theta_{-i})] ds = \int_{\underline{\theta}}^{\theta_i} F^{N-1}(s) ds
 \end{aligned}$$

- Equating the two expressions, we obtain

$$b(\theta) = \theta - \frac{\int_{\theta}^{\theta} F^{N-1}(s) ds}{F^{N-1}(\theta)}$$

Asymmetries:

- Suppose $\theta_i \sim F_i(\cdot)$ (*i.e.*, valuations come from different distributions).
- Define: $MR_i(\theta_i) = \theta_i - \frac{1-F_i(\theta_i)}{f_i(\theta_i)}$
- Revenue = $\mathbb{E}_{\theta} [\sum_i p_i(\theta) [MR(\theta_i) - \theta_0]] + \theta_0$
- If bidder j has ex-ante higher valuation than bidder i (*i.e.*, if $\frac{1-F_j(\theta)}{f_j(\theta)} > \frac{1-F_i(\theta)}{f_i(\theta)}$), then bias auction in favor of θ_i . (Formally, we say that $\theta_j >_{HRO} \theta_i$.)
 - If $\theta_i = \theta_j - \epsilon$, then still allocate good to bidder i .
 - Favor weak bidders to induce the stronger bidders to bid higher.

Welfare Maximization (First Best)

$$\max_{p_i(\cdot)} \left\{ \mathbb{E}_{\theta} \left[\sum_i p_i(\theta) \theta_i + \left[1 - \sum_i p_i(\theta) \right] \theta_0 \right] \right\}$$

- *Solution:* Allocate the good to the agent with the highest valuation (incl. seller)
 - $p_i(\theta) = 1$ if and only if $\theta_i > \theta_j$ for all $j \neq i$ (otherwise 0).
- Implementation:
 1. First-price auction with reserve price θ_0 .
 2. Second-price auction with reserve price θ_0 .
 3. All-pay auction with reserve price θ_0 .

References

Bolton and Dewatripont, (2005), *Contract Theory*, MIT Press.

Ortner J., (2013), Lecture Notes.