

CS599: Algorithm Design in Strategic Settings
Fall 2012
Lecture 3: Mechanism Design Preliminaries

Instructor: Shaddin Dughmi

Administrivia

- HW out soon (monday), due in two weeks
- Office hours next week rescheduled
- Email list
- Announcements on class page

Outline

- 1 Notes Regarding Last Lecture
- 2 Examples of Mechanism Design Problems
- 3 Review: Incomplete Information Games
- 4 The General Mechanism Design Problem
- 5 The Revelation Principle and Incentive Compatibility
- 6 Impossibilities in General Settings
- 7 Mechanisms with Money: The Quasilinear Utility Model

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Rationality

Some of you asked for a formalization of rationality. . .

Definition

A **utility function** on choice set A is a map $u : A \rightarrow \mathbb{R}$.

Definition

When choice set A is a family of lotteries over some other choice set B , a utility function $u : A \rightarrow \mathbb{R}$ is a **Von-Neumann Morgenstern utility function** if there is a utility function $v : B \rightarrow \mathbb{R}$ over B such that $u(a) = \mathbb{E}_{b \sim a}[v(b)]$.

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We assume agents are equipped with VNM utility functions over (distributions over) outcomes of a game / mechanism, and moreover they act to maximize (expected) utility.

Definition

A **rational** agent always chooses the element of his choice set maximizing his (expected) utility.

Arguments in Favor of Nash Equilibrium

- MWG has a nice discussion
- Favorite arguments: self-enforcing agreement, stable social convention

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Single-item Allocation



- n players
- Player i 's private data (type): $v_i \in \mathbb{R}_+$
- Outcome: choice of a winning player, and payment from each player
- Utility of a player for an outcome is his value for the outcome if he wins, less payment

Objectives: Revenue, welfare.

Single-item Allocation



First Price Auction

- 1 Collect bids
- 2 Give to highest bidder
- 3 Charge him his bid

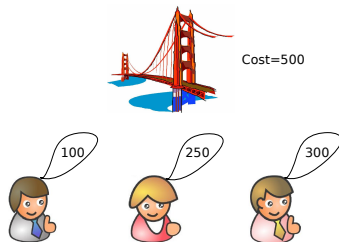
Single-item Allocation



Second-price (Vickrey) Auction

- 1 Collect bids
- 2 Give to highest bidder
- 3 Charge second highest bid

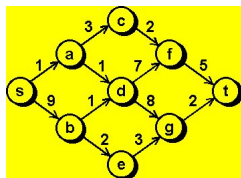
Example: Public Project



- n players
- Player i 's private data (type): $v_i \in \mathbb{R}_+$
- Outcome: choice of whether or not to build, and payment from each player covering the cost of the project if built
- Utility of a player for an outcome is his value for the project if built, less his payment

Goal: Build if sum of values exceeds cost

Shortest Path Procurement



- Players are edges in a network, with designated source/sink
- Player i 's private data (type): cost $c_i \in \mathbb{R}_+$
- Outcome: choice of s-t shortest path to buy, and payment to each player
- Utility of a player for an outcome is his payment, less his cost if chosen.

Goal: buy path with lowest total cost (welfare), or buy a path subject to a known budget, ...

Example: Voting

- n players
- m candidates
- Player i 's private data (type): total preference order on candidates
- Outcome: choice of winning candidate

Goal: ??

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Recall: Incomplete Information Game

- A **game of strict incomplete information** is a tuple (N, A, T, u) , where
- N is a finite set of **players**. Denote $n = |N|$ and $N = \{1, \dots, n\}$.
 - $A = A_1 \times \dots \times A_n$, where A_i is the set of **actions** of player i . Each $\vec{a} = (a_1, \dots, a_n) \in A$ is called an **action profile**.
 - $T = T_1 \times \dots \times T_n$, where T_i is the set of **types** of player i . Each $\vec{t} = (t_1, \dots, t_n) \in T$ is called an **type profile**.
 - $u = (u_1, \dots, u_n)$, where $u_i : T_i \times A \rightarrow \mathbb{R}$ is the **utility function** of player i .

For a Bayesian game, add a common prior \mathcal{D} on types.

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Example: Vickrey Auction

- $A_i = \mathbb{R}$ is the set of possible bids of player i .
- $T_i = \mathbb{R}$ is the set of possible values for the item.
- For $v_i \in T_i$ and $b \in A$, we have $u_i(v_i, b) = v_i - b_{-i}$ if $b_i > b_{-i}$, otherwise 0.

Strategies in Incomplete Information Games

- Strategies of player i
 - **Pure strategy** $s_i : T_i \rightarrow A_i$: a choice of action $a_i \in A_i$ for every type $t_i \in T_i$.
 - Example: Truthtelling is a strategy in the Vickrey Auction
 - Example: Bidding half your value is also a strategy
 - **Mixed strategy**: a choice of distribution over actions A_i for each type $t_i \in T_i$
 - Won't really use... all our applications will involve pure strategies

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Note

In a strategy, player decides how to act based only on his private info (his type), and NOT on others' private info nor their actions.

$s_i : T_i \rightarrow A_i$ is a **dominant strategy** for player i if, for all $t_i \in T_i$ and $a_{-i} \in A_{-i}$ and $a'_i \in A_i$,

$$u_i(t_i, (s_i(t_i), a_{-i})) \geq u_i(t_i, (a'_i, a_{-i}))$$

Equivalently: $s_i(t_i)$ is a best response to $s_{-i}(t_{-i})$ for all t_i, t_{-i} and s_{-i} .

Illustration: Vickrey Auction

Vickrey Auction

Consider a Vickrey Auction with incomplete information.

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Claim

The truth-telling strategy is dominant for each player.

Bayes-Nash Equilibrium

As before, a strategy s_i for player i is a map from T_i to A_i . Now, we define the extension of Nash equilibrium to this setting.

A pure **Bayes-Nash Equilibrium** of a Bayesian Game of incomplete information is a set of strategies s_1, \dots, s_n , where $s_i : T_i \rightarrow A_i$, such that for all $i, t_i \in T_i, a'_i \in A_i$ we have

$$\mathbf{E}_{t_{-i} \sim \mathcal{D}|t_i} u_i(t_i, s(t)) \geq \mathbf{E}_{t_{-i} \sim \mathcal{D}|t_i} u_i(t_i, (a'_i, s_{-i}(t_{-i})))$$

where the expectation is over t_{-i} drawn from p after conditioning on t_i .

- Note: Every dominant strategy equilibrium is also a Bayes-Nash Equilibrium
- But, unlike DSE, BNE is guaranteed to exist.

Example: First Price Auction

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- $A_i = T_i = [0, 1]$
- $u_i(v_i, b) = v_i - b_i$ if $b_i > b_j$ for all $j \neq i$, otherwise 0.
- \mathcal{D} draws each $v_i \in T_i$ independently from $[0, 1]$.

Show that the strategies $b_i(v_i) = v_i/2$ form a Bayes-Nash equilibrium.

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Mechanism Design Setting (Prior-free)

Given by a tuple (N, \mathcal{X}, T, u) , where

- N is a finite set of **players**. Denote $n = |N|$ and $N = \{1, \dots, n\}$.
- \mathcal{X} is a set of **outcomes**.
- $T = T_1 \times \dots \times T_n$, where T_i is the set of **types** of player i . Each $\vec{t} = (t_1, \dots, t_n) \in T$ is called a **type profile**.
- $u = (u_1, \dots, u_n)$, where $u_i : T_i \times \mathcal{X} \rightarrow \mathbb{R}$ is the **utility function** of player i .

General Form

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Example: Single-item Allocation

- Outcome: choice $x \in \{e_1, \dots, e_n\}$ of winning player, and payment p_1, \dots, p_n from each
- Type of player i : value $v_i \in \mathbb{R}_+$.
- $u_i(v_i, x) = v_i x_i - p_i$.

Social Choice Functions

A **principal** wants to communicate with players and aggregate their private data (types) into a choice of outcome. Such aggregation captured by

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Choosing a Social Choice Function

- A particular social choice function in mind (e.g. majority voting, utilitarian allocation of a single item, etc).
- An **objective function** $o : T \times \mathcal{X} \rightarrow \mathbb{R}$, and want $f(T)$ to (approximately) maximize $o(T, f(T))$
 - Either worst case over T (Prior-free) or in expectation (Bayesian)

Example: Single-item Allocation

- Welfare objective: $welfare(v, (x, p)) = \sum_i v_i x_i$
- Revenue objective: $revenue(v, (x, p)) = \sum_i p_i$

Mechanisms

To perform such aggregation, the principal runs a protocol, known as a **mechanism**. Formally,

A **mechanism** is a pair (A, g) , where

- $A = A_1 \times \dots \times A_n$, where A_i is the set of possible **actions** (think messages, or bids) of player i in the protocol. \mathcal{A} is the set of **action profiles**.
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The resulting game of mechanism design is a game of incomplete information where when players play $a \in A$, player i 's utility is $u_i(t_i, g(a))$ when his type is t_i .

Example: First price auction

- $A_i = \mathbb{R}$
- $g(b_1, \dots, b_n) = (x, p)$ where $x_{i^*} = 1$, $p_{i^*} = b_{i^*}$ for $i^* = \operatorname{argmax}_i b_i$, and $x_i = p_i = 0$ for $i \neq i^*$.

Implementation of Social Choice Functions

We say a mechanism (A, g) **implements** social choice function $f : T \rightarrow \mathcal{X}$ in dominant-strategy [Bayes-Nash] equilibrium if there is a strategy profile $s = (s_1, \dots, s_n)$ with $s_i : T_i \rightarrow A_i$ such that

- $s_i : T_i \rightarrow A_i$ is a dominant-strategy [Bayes-Nash] equilibrium in the resulting incomplete information game
- $g(s_1(t_1), s_2(t_2), \dots, s_n(t_n)) = f(t_1, t_2, \dots, t_n)$ for all $t \in T$

Example: First price, two players, i.i.d $U[0, 1]$

Implements in BNE the following social choice function: give the item to the player with the highest value and charges him half his value.

Example: Vickrey Auction

Implements in DSE the following social choice function: give the item to the player with the highest value and charges him the second highest value.

The Task of Mechanism Design

Task of Mechanism Design (Take 1)

Given a notion of a “good” social choice function from T to X , find

- A mechanism
 - An action space $A = (A_1, \dots, A_n)$,
 - an outcome function $g : A \rightarrow \mathcal{X}$,
- an equilibrium (s_1, \dots, s_n) of the resulting game of mechanism design

such that the social choice function $f(t_1, \dots, t_n) = g(s_1(t_1), \dots, s_n(t_n))$ is “good.”

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This seems like a complicated, multivariate search problem.

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Luckily

The **revelation principle** reduces the search space to just $g : T \rightarrow \mathcal{X}$.

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Incentive-Compatibility

Direct Revelation

A mechanism (A, g) is a **direct revelation mechanism** if $A_i = T_i$ for all i .

i.e. in a direct revelation mechanism, players simultaneously report types (not necessarily truthfully) to the mechanism. Such mechanisms can simply be described via the function $g : T \rightarrow \mathcal{X}$.

Incentive-Compatibility

A direct-revelation mechanism is dominant-strategy [Bayesian] **incentive-compatible** (aka **truthful**) if the truth-telling is a dominant-strategy [Bayes-Nash] equilibrium in the resulting incomplete-information game.

Note: A direct revelation incentive-compatible mechanism implements its outcome function $g : T \rightarrow \mathcal{X}$, by definition.

The social choice function IS the mechanism!!

Examples

Vickrey Auction

Direct revelation mechanism, dominant-strategy incentive-compatible.

First Price Auction

Direct revelation mechanism, not Bayesian incentive compatible.

Example: Posted price

The auction that simply posts a fixed price to players in sequence until one accepts is not direct revelation.

Revelation Principle

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If there is a mechanism implementing social choice function f in dominant-strategy [Bayes-Nash] equilibrium, then there is a direct revelation, dominant-strategy [Bayesian] incentive-compatible mechanism implementing f .

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This simplifies the task of mechanism design

Task of Mechanism Design (Take 2)

Given a notion of a “good” social choice function from T to X , find such a function $f : T \rightarrow X$ such that truth-telling is an equilibrium in the following mechanism:

- Solicit reports $\tilde{t}_i \in T_i$ from each player i (simultaneous, sealed bid)
- Choose outcome $f(\tilde{t}_1, \dots, \tilde{t}_n)$

Example

2 players, with values i.i.d uniform from $[0, 1]$, facing the first-price auction.

First-price Auction

- 1 Solicit bids b_1, b_2
- 2 Give item to highest bidder, charging him his bid

Recall

The strategies where each player reports half their value are in BNE. In other words, when player 1 knows his value v_1 , and faces player 2 who is bidding uniformly from $[0, 1/2]$, he maximizes his expected utility $(v_1 - b_1) \cdot 2b_1$ by bidding $b_1 = v_1/2$. And vice versa.

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Therefore . . .

the first price auction implements in BNE the social choice function which gives the item to the highest bidder, and charges him half his bid

Example

Modified First-price Auction

- 1 Solicit bids b_1, b_2
- 2 Give item to highest bidder, charging him half his bid
 - Equivalently, simulate a first price auction where bidders bid $b_1/2, b_2/2$

Claim

Truth-telling is a BNE in the modified first-price auction.

Therefore, the modified auction implements the same social-choice function in equilibrium, but is truthful.

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Proof

Assume player 2 bids truthfully. Player 1 faces a (simulated) first price auction where his own bid is halved before participating, and player 2 bids uniformly from $[0, 1/2]$. To respond optimally in the simulation, he bids $b_1 = v_1$ and lets the mechanism halve his bid on his behalf.

Proof (Bayesian Setting)

Consider mechanism (A, g) , with BNE strategies $s_i : T_i \rightarrow A_i$.

- Implements $f(t_1, \dots, t_n) = g(s_1(t_1), \dots, s_n(t_n))$ in BNE
- For all i and t_i , action $s_i(t_i)$ maximizes player i 's expected utility when other players are playing $s_{-i}(t_{-i})$ for $t_{-i} \sim \mathcal{D}|t_i$.

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Modified Mechanism

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 - Equivalently, simulate (A, g) when players play $s_i(t_i)$
- Assume all players other than i report truthfully
 - When i 's type is t_i , other players playing $s_{-i}(t_{-i})$ for $t_{-i} \sim \mathcal{D}|t_i$ in simulated mechanism
 - As stated above, his best response in simulation is $s_i(t_i)$.
 - Mechanism transforms his bid by applying s_i , so best to bid t_i .

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Mechanism Design Impossibilities

The revelation principle reduces mechanism design to the design of direct-revelation, truthful mechanisms.

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Unfortunately...

Absent structure on the outcome space and utility functions, no reasonably good mechanisms exist even in simple settings.

Examples coming up: single-item allocation without payments, voting

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Luckily

The structure that enables much of mechanism design is assuming that the outcome space incorporates monetary payments, and player utilities are linear in these payments.

Single-item Allocation Without Money

Question

- Consider allocating a single item among n players, with private values (types) $v_1, \dots, v_n \in \mathbb{R}_+$ for the item, without access to monetary payments.
- Restricted to mechanisms that implement their social choice function in dominant strategies.
- What is the smallest worst-case approximation ratio for social welfare of such a mechanism? Prove it.
- WLOG by revelation principle: restrict attention to dominant-strategy truthful mechanisms $f : \mathbb{R}_+^n \rightarrow \{1, \dots, n\}$.

The **worst-case approximation ratio** of mechanism f for social welfare is defined as

$$\max_{v \in \mathbb{R}_+^n} \frac{\max_i v_i}{v_{f(v)}}$$

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The smallest worst-case approximation ratio is n . No mechanism can guarantee better than $1/n$ fraction of the optimal social welfare in dominant strategy equilibrium!

Recall: voting

- n players
- m candidates
- Player i 's private data (type): total preference order on candidates
- Outcome: choice of winning candidate

Theorem (Gibbard-Satterthwaite)

*Assume the number of candidates C is at least 3. Consider a voting mechanism implementing allocation rule $f : \Sigma^n \rightarrow C$ in dominant strategies. Either f is a **dictatorship** or some candidate can never win in f .*

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Incorporating Payments

To make much of modern mechanism design possible, we assume that

- The set of outcomes has a particular structure: every outcome includes a payment to and from each player.
- Player utilities vary linearly with their payment.

Examples: Single-item allocation, public project, shortest path procurement

Non-examples: Single-item allocation without money, voting.

Quasilinear Utilities

The Quasi-linear Setting

Formally, $\mathcal{X} = \Omega \times \mathbb{R}^n$.

- Ω is the set of **allocations**
- For $(\omega, p_1, \dots, p_n) \in \mathcal{X}$, p_i is the payment from (or to) player i . and player i 's utility function $u_i : T_i \times \mathcal{X} \rightarrow \mathbb{R}$ takes the following form

$$u_i(t_i, (\omega, p_1, \dots, p_n)) = v_i(t_i, \omega) - p_i$$

for some **valuation function** $v_i : T_i \times \Omega \rightarrow \mathbb{R}$.

We say players have **quasilinear** utilities.

Example: Single-item Allocation

- $\Omega = \{e_1, \dots, e_n\}$
- $u_i(t_i, (\omega, p_1, \dots, p_n)) = t_i \omega_i - p_i$

Further simplification

Recall that, using the revelation principle, we got

Task of Mechanism Design (Take 2)

Given a notion of a “good” social choice function from T to X , find such a function $f : T \rightarrow X$ such that truth-telling is an equilibrium in the following mechanism:

- Solicit reports $\tilde{t}_i \in T_i$ from each player i (simultaneous, sealed bid)
- Choose outcome $f(\tilde{t}_1, \dots, \tilde{t}_n)$

Further simplification

In quasilinear settings this breaks down further

Task of Mechanism Design in Quasilinear settings

Find a “good” **allocation rule** $f : T \rightarrow \Omega$ and **payment rule** $p : T \rightarrow \mathbb{R}^n$ such that the following mechanism is **incentive-compatible**:

- Solicit reports $\tilde{t}_i \in T_i$ from each player i (simultaneous, sealed bid)
- Choose allocation $f(\tilde{t})$
- Charge player i payment $p_i(\tilde{t})$

We think of the mechanism as the pair (f, p) .

Sometimes, we abuse notation and think of type t_i directly as the valuation $v_i : \Omega \rightarrow \mathbb{R}$.

Incentive-Compatibility

Incentive compatibility can be stated simply now

Incentive-compatibility (Dominant Strategy)

A mechanism (f, p) is dominant-strategy truthful if, for every player i , true type t_i , possible mis-report \tilde{t}_i , and reported types t_{-i} of the others, we have

$$v_i(t_i, f(t)) - p_i(t) \geq v_i(t_i, f(\tilde{t}_i, t_{-i})) - p_i(\tilde{t}_i, t_{-i})$$

If (f, p) randomized, add expectation signs.

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If (f, p) randomized, add expectation signs.

Incentive-compatibility (Bayesian)

A mechanism (f, p) is Bayesian incentive compatible if, for every player i , true type t_i , possible mis-report \tilde{t}_i , the following holds in expectation over $t_{-i} \sim D|t_i$

$$\mathbf{E}[v_i(t_i, f(t)) - p_i(t)] \geq \mathbf{E}[v_i(t_i, f(\tilde{t}_i, t_{-i})) - p_i(\tilde{t}_i, t_{-i})]$$

Vickrey Auction

- Allocation rule maps b_1, \dots, b_n to e_{i^*} for $i^* = \operatorname{argmax}_i b_i$
- Payment rule maps b_1, \dots, b_n to p_1, \dots, p_n where $p_{i^*} = b_{(2)}$, and $p_i = 0$ for $i \neq i^*$.

Dominant-strategy truthful.

First Price Auction

- Allocation rule maps b_1, \dots, b_n to e_{i^*} for $i^* = \operatorname{argmax}_i b_i$
- Payment rule maps b_1, \dots, b_n to p_1, \dots, p_n where $p_{i^*} = b_{(1)}$, and $p_i = 0$ for $i \neq i^*$.

For two players i.i.d $U[0, 1]$, players bidding half their value is a BNE.
Not Bayesian incentive compatible.

Modified First Price Auction

- Allocation rule maps b_1, \dots, b_n to e_{i^*} for $i^* = \operatorname{argmax}_i b_i$
- Payment rule maps b_1, \dots, b_n to p_1, \dots, p_n where $p_{i^*} = b_{(1)}/2$, and $p_i = 0$ for $i \neq i^*$.

For two players i.i.d $U[0, 1]$, Bayesian incentive compatible.