# Auctions \& <br> Mechanism Design Basics 

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- We study about a kind of science of rule-making.
- To make it simple, we first consider single-item auctions.
- We will go over some basics about first-price auctions and second-price auctions.
- Also, we will talk about
- incentive guarantees,
- strong performance guarantees, and
- computational efficiency
in an auction.
- We will end the discussion with Myerson's Lemma.


## Outline

## Single-Item Auctions

Sealed-Bid Auctions
First-Price Auctions
Second-Price Auctions
Case Study: Sponsored Search Auctions
Myerson's Lemma
Single-Parameter Environments
The Lemma
Application to the Sponsored Search Auction

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- Her maximum willingness-to-pay for it.
- $v_{i}$ is private.
- Unknown to the seller and other bidders.


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- Step (ii): The selection rule. We consider giving the item to the highest bidder.


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## Issues of the First-Price Auctions

- For a bidder: Hard to figure how to bid.
- For the seller: Hard to predict what will happen.


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- Your valuation for the item: the number of your birth month + the day of your birth.


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- Your valuation is between 2 and 43.
- Suppose that there is another bidder who has the same valuation like you.
- Would it help to know your opponent's birthday?
- Would your answer change if you knew there were two other bidders rather than one?


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whichever comes first.
- For example, if the highest other bid is $\$ 90$.

You only pay $\$ 90+\epsilon$ for some small increment $\epsilon$.
$\approx$ highest other bid!

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## Second-Price/Vickrey Auction

The highest bidder wins and pays a price equal to the second-highest bid.

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- Is such a strategy a dominant strategy?
- The strategy is guaranteed to maximize a bidder's utility no matter what other bidders do.


## Truthfully Bidding Is Dominant Here

Proposition (Incentives in Second-Price Auctions)
In a second-price auction, every bidder $i$ has a dominant strategy: set the bid $b_{i}=v_{i}$, equal to her private valuation.

## Proof of the Proposition

- Fix a bidder $i$ with valuation $v_{i}$.
- b: the vector of all bids.
- $\boldsymbol{b}_{-i}$ : the vector of $\boldsymbol{b}$ with $b_{i}$ removed.
* Goal: Show that bidder i's utility is maximized by setting $b_{i}=v_{i}$.


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## Second-Price Single-Item Auctions are "ideal"

Definition (Dominant-Strategy Incentive Compatible)
An auction is dominant-strategy incentive compatible (DSIC) if

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The social welfare of an outcome of a single-item auction is

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where $\sum_{i=1}^{n} x_{i} \leq 1 ; x_{i}=1$ if bidder $i$ wins and 0 if she loses.

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- So such an auction is welfare maximizing if bids are truthful.


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## Theorem

A second-price single-item auction satisfies:
(1) DSIC.
(2) Welfare maximizing.
(3) It can be implemented in polynomial time.

In fact, (3) is linear.

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## Background

## The Social Dilemma (2020) - Trailer

- Web search results:
- relevant to your query (by an algorithm, e.g., PageRank).
- pops out a list of sponsored links.
- They are paid by advertisers.
- Every time you give a search query into a search engine, an auction is run in real time to decide
- which advertiser's links are shown,
- how these links are arranged visually,
- what the advertisers are charged.


## Multiple Items for Sponsored Search Auctions

- Let's say the items for sale are $k$ "slots" on a search results page.
- Bidders: the advertisers who have a bid on the keyword that was searched on.
- On the keyword, "university", NTU, NYCU, NCKU, TKU, etc., might be the bidders.


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- On the keyword, "camera", Nikon, Canon, Sony, etc., might be the bidders.
- On the keyword, "SUV", Toyota, Ford, Honda, Porsche, etc., might be the bidders.
- Let's say the items are not identical.
- Higher slots are more valuable. What do you think?


## Multiple Items for Sponsored Search Auctions

- Consider the click-through-rates (CTRs) $\alpha_{j}$ of slot $j$.
- The probability that the user clicks on this slot.
- Assumption: $\alpha_{1} \geq \alpha_{2} \geq \ldots \alpha_{k}$.


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- The CTR of advertiser $i$ in slot $j: \beta_{i} \alpha_{j}$.
- The expected value derived by advertiser $i$ from slot $j: v_{i} \alpha_{j}$
- The social welfare is $\sum_{i=1}^{n} v_{i} x_{i}$.
- $x_{i}$ : the CTR of the slot to which bidder $i$ is assigned.
- $x_{i}=0$ : bidder $i$ is not assigned to a slot.
- Each slot can only be assigned to one bidder; each bidder gets only one slot.


## Our Design Approach

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- Who pays what?
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- If the payments are not just right, then the strategic bidders will game the system.


## Our Design Approach

## Design Steps

(a): Assume that the bidders bid truthfully. Then, how should we assign bidders to slots so that property (2) and (3) holds?
(b): Given the answer of above, how should we set selling prices so that property (1) holds?

## Step (a)

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- You can prove that this assignment achieves the maximum social welfare as an exercise.


## Step (b)

- There is an analog of the second-price rule.
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- There is an analog of the second-price rule.
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* Myerson's lemma.
- A powerful and general tool for implementing this second step.


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## Single-Parameter Environments

Consider a more generalized and abstract setting:
Single-Parameter Environments

- $n$ agents (e.g., bidders).
- A private valuation $v_{i} \geq 0$ for each agent $i$ (per unit of stuff).
- A feasible set $X=\left\{\left(x_{1}, x_{2}, \ldots, x_{n}\right) \mid x_{i} \in \mathbb{R}\right\} \subseteq \mathbb{R}^{n}$.
- $x_{i}$ : amount of stuff given to agent $i$.


## Single-Parameter Environments (Examples)

- Single-item auction:
- $\sum_{i=1}^{n} X_{i} \leq 1$, and $x_{i} \in\{0,1\}$ for each $i$.


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- Single-item auction:
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- k-Unit auction:
- $k$ identical items, $\sum_{i=1}^{n} X_{i} \leq k$, and $x_{i} \in\{0,1\}$ for each $i$.


## Single-Parameter Environments (Examples)

- Single-item auction:
- $\sum_{i=1}^{n} X_{i} \leq 1$, and $x_{i} \in\{0,1\}$ for each $i$.
- k-Unit auction:
- $k$ identical items, $\sum_{i=1}^{n} X_{i} \leq k$, and $x_{i} \in\{0,1\}$ for each $i$.
- Sponsored Search Auction:
- $X$ : the set of $n$-vectors $\Leftrightarrow$ assignments of bidders to slots.
- Each slot (resp., bidder) is assigned to $\leq 1$ bidder (resp., slot).
- The component $x_{i}=\alpha_{j}$ : bidder $i$ is assigned to slot $j$.
- $\alpha_{j}$ : the click-through rate of slot $j$.
- Assume that the quality score $\beta_{i}=1$ for all $i$.


## Allocation and Payment Rules

Choices to make in a sealed-bid auction

- Collect bids $\boldsymbol{b}=\left(b_{1}, \ldots, b_{n}\right)$.
- Allocation Rule: Choose a feasible $\boldsymbol{x}(\boldsymbol{b}) \in X \subseteq \mathbb{R}^{n}$.
- Payment Rule: Choose payments $\boldsymbol{p}(\boldsymbol{b}) \in \mathbb{R}^{n}$.
- A direct-revelation mechanism.


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- A direct-revelation mechanism.
- Example of indirect mechanism: iterative ascending auction.


## Allocation and Payment Rules (contd.)

With allocation rule $\boldsymbol{x}$ and payment rule $\boldsymbol{p}$,

- agent $i$ receives utility $u_{i}(\boldsymbol{b})=v_{i} \cdot x_{i}(\boldsymbol{b})-p_{i}(\boldsymbol{b})$.
- $p_{i}(\boldsymbol{b}) \in\left[0, b_{i} \cdot x_{i}(\boldsymbol{b})\right]$.
- $p_{i}(\boldsymbol{b}) \geq 0$ : prohibiting the seller from paying the agents.
- $p_{i}(\boldsymbol{b}) \leq b_{i} \cdot x_{i}(\boldsymbol{b}):$ a truthful agent receives nonnegative utility.


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- $p_{i}(\boldsymbol{b}) \in\left[0, b_{i} \cdot x_{i}(\boldsymbol{b})\right]$.
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- $p_{i}(\boldsymbol{b}) \leq b_{i} \cdot x_{i}(\boldsymbol{b}):$ a truthful agent receives nonnegative utility. Why?


## The Myerson's Lemma

Definition (Implementable Allocation Rule)
An allocation rule $\boldsymbol{x}$ for a single-parameter environment is implementable if there is a payment rule $\boldsymbol{p}$ such that the direct-revelation mechanism $(\boldsymbol{x}, \boldsymbol{p})$ is DSIC.

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Bidding higher can only get you more stuff!
So, how about awarding the item to the second-highest bidder?
You raise your bid, you might lose the chance of getting it!

## Outline

Sealed-Bid Auctions
First-Price Auctions
Second-Price Auctions Case Study: Sponsored Search Auctions

Myerson's Lemma
Single-Parameter Environments
The Lemma
Application to the Sponsored Search Auction

## The Myerson's Lemma

Theorem (Myerson's Lemma)
Fix a single-parameter environment.
(i) An allocation rule $\boldsymbol{x}$ is implementable if and only if it is monotone.
(ii) If $\boldsymbol{x}$ is monotone, then there is a unique payment rule for which the direct-revelation mechanism ( $\boldsymbol{x}, \boldsymbol{p})$ is DSIC and $p_{i}(\boldsymbol{b})=0$ whenever $b_{i}=0$.
(iii) The payment rule in (ii) is given by an explicit formula.
"Monotone" is more operational.

## Allocation curves: allocation as a function of bids



Figures from Tim Roughgarden's lecture notes.

## Constraints from DSIC

Consider $0 \leq z<y$.
Say agent $i$ has a private valuation $z$ and free to submit a false bid $y$ or agent $i$ has a private valuation $y$ and free to submit a false bid $z$

DSIC: Bidding truthfully brings maximum utility.

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\begin{aligned}
z \cdot x(z)-p(z) & \geq z \cdot x(y)-p(y) \\
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$p(y)-p(z)$ can be bounded below and above.
$\Rightarrow$ every implementable allocation rule is monotone (why?)

Case: $x$ is a piecewise constant function

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- Try: fix $z$ and let $y$ tend to $z$.

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p_{i}\left(b_{i}, \boldsymbol{b}_{-i}\right)=\sum_{j=1}^{\ell} z_{j} \cdot\left[\text { jump in } x_{i}\left(\cdot, \boldsymbol{b}_{-i}\right) \text { at } z_{j}\right],
$$

where $z_{1}, \ldots, z_{\ell}$ are breakpoints of $x_{i}\left(\cdot, \boldsymbol{b}_{-i}\right)$ in the range $\left[0, b_{i}\right]$.

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- Suppose $x$ is differentiable.
- Dividing the inequalities by $y-z$ :


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p_{i}\left(b_{i}, \boldsymbol{b}_{-i}\right)=\int_{0}^{b_{i}} z \cdot \frac{d}{d z} x_{i}\left(z, \boldsymbol{b}_{-i}\right) d z .
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## Apply to Sponsored Search Auction

The allocation rule is piecewise.
re-index the bidders: $b_{1} \geq b_{2} \geq \ldots \geq b_{n}$.


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## Exercise 1 (5\%)

- Recall that in the model of sponsored search auctions:
- There are $k$ slots, the $j$ th slot has a click-through rate (CTR) of $\alpha_{j}$ (nonincreasing in $j$ ).
- The utility of bidder $i$ in slot $j$ is $\alpha_{j}\left(v_{i}-p_{j}\right)$, where $v_{i}$ is the private value-per-click of the bidder and $p_{j}$ is the price charged per-click in slot $j$.
- The Generalized Second Price (GSP) Auction is defined as follows:


## Exercise 1 (5\%) (contd.)

## The Generalized Second Price (GSP) Auction

1. Rank advertisers from highest to lowest bid; assume without loss of generality that $b_{1} \geq b_{2} \geq \cdots \geq b_{n}$.
2. For $i=1,2, \ldots, k$, assign the $i$ th bidder to the $i$ slot.
3. For $i=1,2, \ldots, k$, charge the $i$ th bidder a price of $b_{i+1}$ per click.
(a) Prove that for every $k \geq 2$ and sequence $\alpha_{1} \geq \cdots \geq \alpha_{k}>0$ of CTRs, the GSP auction is NOT DSIC. (Hint: Find out an example.)
(b) A bid profile $\boldsymbol{b}$ with $b_{1} \geq \cdots \geq b_{n}$ is envy-free if for every bidder $i$ and slot $j \neq i$,

$$
\alpha_{i}\left(v_{i}-b_{i+1}\right) \geq \alpha_{j}\left(v_{i}-b_{j+1}\right) .
$$

Please verify that every envy-free bid profile is an equilibrium.

