

Auctions & 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 efficiencyin 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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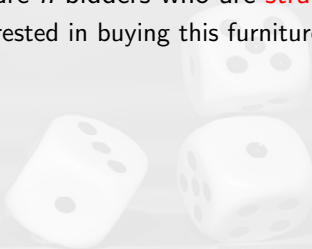
Strategic bidders in an auction

- ▶ Consider a seller with a single item.
 - ▶ For example, an antiquated furniture.



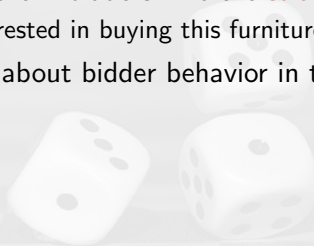
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 - ▶ Her maximum willingness-to-pay for it.
 - ▶ v_i is **private**.
 - ▶ Unknown to the seller and other bidders.

What does a bidder want? What's her utility?

- ▶ Each bidder wants to acquire the item as cheaply as possible.



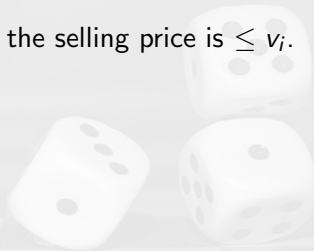
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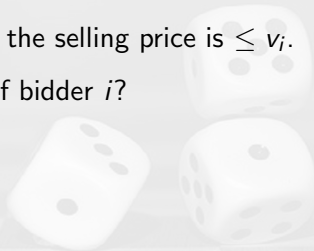
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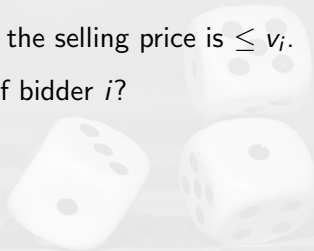
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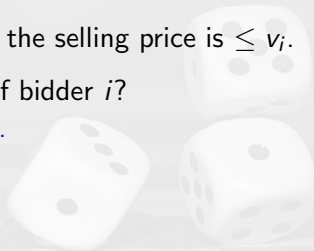
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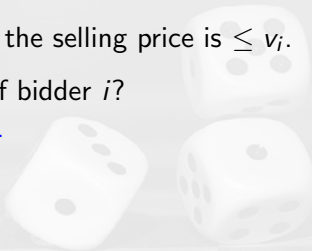
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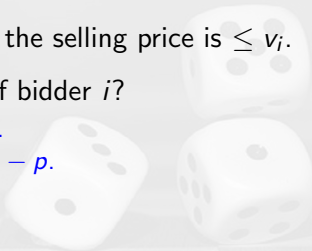
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Second-Price Auctions

Case Study: Sponsored Search Auctions

Myerson's Lemma

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Sealed-Bid Auctions

Sealed-Bid Auction

- (i) Each bidder i privately communicates a bid b_i to the seller—in a sealed envelope.
- (ii) The seller decides who gets the item (if any).
- (iii) The seller decides the selling price.

Sealed-Bid Auctions

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- (i) Each bidder i **privately** communicates a bid b_i to the seller—in a sealed envelope.
 - (ii) The seller **decides who** gets the item (if any).
 - (iii) The seller **decides the selling price**.
- Step (ii): The selection rule. We consider giving the item to the **highest** bidder.

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The winning bidder **pays her bid**.

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- ▶ Why?

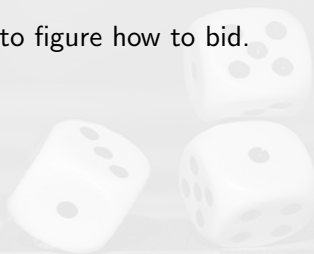
Issues of the First-Price Auctions

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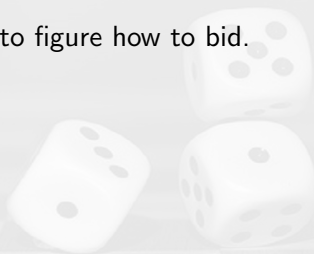
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Issues of the First-Price Auctions

- ▶ For a bidder: Hard to figure how to bid.
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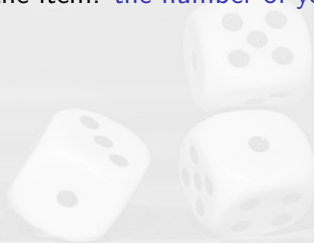


Issues of the First-Price Auctions

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

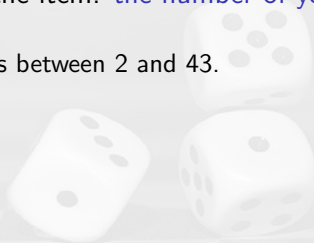
An Example

- ▶ Suppose that you are participating in the first-price auction.
- ▶ Your valuation for the item: the number of your birth month + the day of your birth.



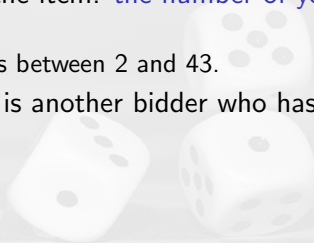
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 - ▶ Would it help to know your opponent's birthday?

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 - ▶ Would your answer change if you knew there were two other bidders rather than one?

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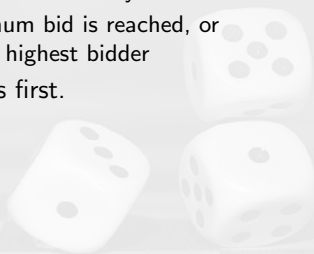
eBay/Yahoo auction

- If you bid \$100 and win, do you pay \$100?



eBay/Yahoo auction

- ▶ If you bid \$100 and win, do you pay \$100?
 - ▶ eBay increases your bid on your behalf until
 - ▶ Your maximum bid is reached, or
 - ▶ You are the highest bidder
- whichever comes first.



eBay/Yahoo auction

- ▶ If you bid \$100 and win, do you pay \$100?
 - ▶ eBay increases your bid on your behalf until
 - ▶ Your maximum bid is reached, or
 - ▶ You are the highest bidderwhichever comes first.
 - ▶ For example, if the highest other bid is \$90.
You only pay \$90 + ϵ for some small increment ϵ .
- ≈ highest other bid!

Second-Price auction

Second-Price/Vickrey Auction

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

- Is such a strategy a **dominant strategy**?

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Second-Price/Vickrey Auction

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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 .
- ▶ \mathbf{b} : the vector of all bids.
- ▶ \mathbf{b}_{-i} : the vector of \mathbf{b} with b_i removed.
- * **Goal:** Show that bidder i 's utility is maximized by setting $b_i = v_i$.

Proof of the Proposition (contd.)

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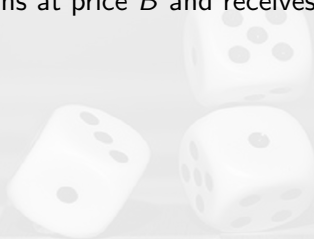
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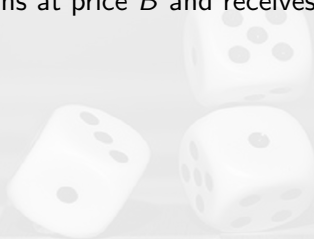
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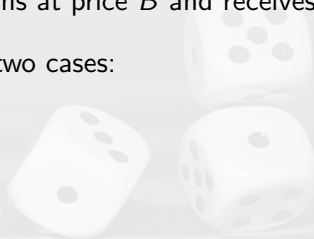
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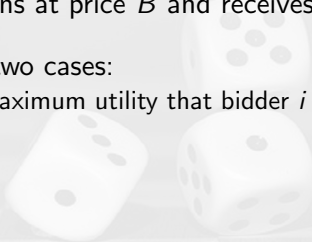
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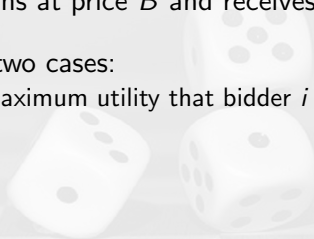
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Proposition 2 (Nonnegative Utility)

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 - ▶ \therefore bidder i wins and bids her true valuation v_i , so $p \leq v_i \Rightarrow v_i - p \geq 0$.

Second-Price Single-Item Auctions are “ideal”

Definition (Dominant-Strategy Incentive Compatible)

An auction is **dominant-strategy incentive compatible (DSIC)** if

- ▶ truthful bidding is a dominant strategy for every bidder, and
- ▶ truthful bidders always obtain nonnegative utility.



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Social Welfare

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

$$\sum_{i=1}^n v_i x_i.$$

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

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

Second-Price Single-Item Auctions are “ideal” (contd.)

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.

Second-Price Single-Item Auctions are “ideal” (contd.)

Theorem

A second-price single-item auction satisfies:

- (1) DSIC. (strong incentive guarantees)
- (2) Welfare maximizing. (strong performance guarantees)
- (3) It can be implemented in polynomial time. (computational efficiency)

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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.

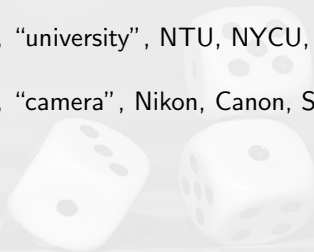
Multiple Items for Sponsored Search Auctions

- ▶ 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, “SUV”, Toyota, Ford, Honda, Porsche, 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?

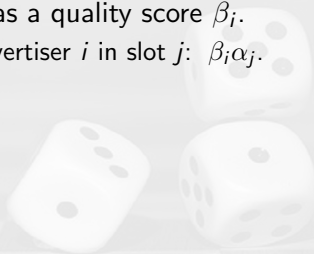
Multiple Items for Sponsored Search Auctions

- ▶ Consider the click-through-rates (CTRs) α_j of slot j .
 - ▶ The probability that the user clicks on this slot.
 - ▶ Assumption: $\alpha_1 \geq \alpha_2 \geq \dots \alpha_k$.



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- ▶ 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.

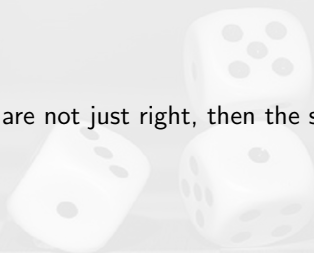
Our Design Approach

- ▶ Who wins what?
- ▶ Who pays what?
- ▶ The payment.



Our Design Approach

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- ▶ The payment.
 - ▶ If the payments are not just right, then the strategic bidders will game the system.



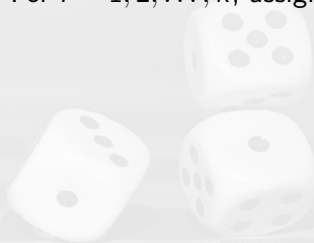
Our Design Approach

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)

- ▶ Given truthful bids. For $i = 1, 2, \dots, k$, assign the i th highest bid to the i th best slot.

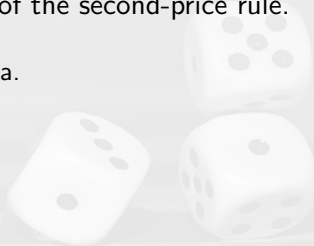


Step (a)

- ▶ Given truthful bids. For $i = 1, 2, \dots, k$, assign the i th highest bid to the i th best slot.
- ▶ 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.
 - ▶ DSIC.
 - ★ Myerson's lemma.



Step (b)

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

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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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 $X = \{(x_1, x_2, \dots, x_n) \mid x_i \in \mathbb{R}\} \subseteq \mathbb{R}^n$.
 - ▶ 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 .



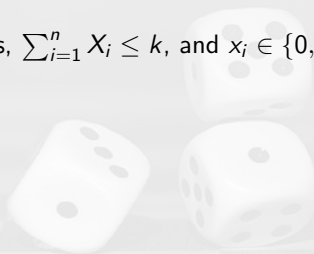
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▶ Sponsored Search Auction:

- ▶ X : the set of n -vectors \Leftrightarrow assignments of bidders to slots.
- ▶ Each slot (resp., bidder) is assigned to ≤ 1 bidder (resp., slot).
- ▶ The component $x_i = \alpha_j$: bidder i is assigned to slot j .
 - ▶ α_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 $\mathbf{b} = (b_1, \dots, b_n)$.
 - ▶ Allocation Rule: Choose a feasible $\mathbf{x}(\mathbf{b}) \in X \subseteq \mathbb{R}^n$.
 - ▶ Payment Rule: Choose payments $\mathbf{p}(\mathbf{b}) \in \mathbb{R}^n$.
- ▶ *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 \mathbf{x} and payment rule \mathbf{p} ,

- ▶ agent i receives utility $u_i(\mathbf{b}) = v_i \cdot x_i(\mathbf{b}) - p_i(\mathbf{b})$.
- ▶ $p_i(\mathbf{b}) \in [0, b_i \cdot x_i(\mathbf{b})]$.
 - ▶ $p_i(\mathbf{b}) \geq 0$: prohibiting the seller from paying the agents.
 - ▶ $p_i(\mathbf{b}) \leq b_i \cdot x_i(\mathbf{b})$: a truthful agent receives nonnegative utility.

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Why?

The Myerson's Lemma

Definition (Implementable Allocation Rule)

An allocation rule \mathbf{x} for a single-parameter environment is **implementable** if there is a payment rule \mathbf{p} such that the direct-revelation mechanism (\mathbf{x}, \mathbf{p}) is **DSIC**.



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You raise your bid, you might lose the chance of getting it!

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The Myerson's Lemma

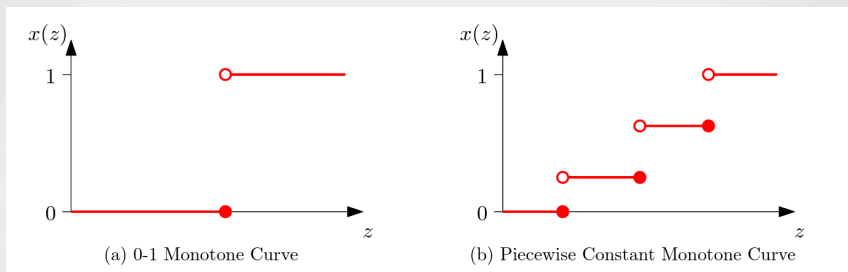
Theorem (Myerson's Lemma)

Fix a single-parameter environment.

- (i) An allocation rule \mathbf{x} is **implementable** if and only if it is **monotone**.
- (ii) If \mathbf{x} is monotone, then there is a unique payment rule for which the direct-revelation mechanism (\mathbf{x}, \mathbf{p}) is DSIC and $p_i(\mathbf{b}) = 0$ 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.

$$z \cdot x(z) - p(z) \geq z \cdot x(y) - p(y)$$

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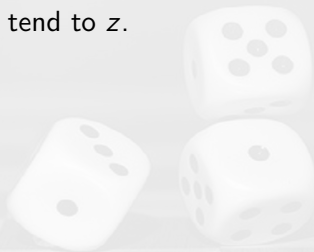
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\Rightarrow every implementable allocation rule is monotone (why?)

Case: x is a piecewise constant function

$$z \cdot (x(y) - x(z)) \leq p(y) - p(z) \leq y \cdot (x(y) - x(z)).$$

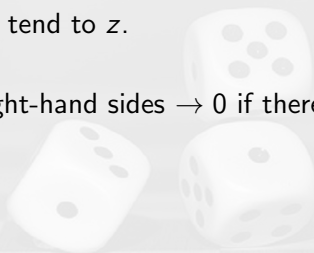
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 \Rightarrow left-hand and right-hand sides $\rightarrow 0$ if there is no jump in x at z .



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$$p_i(b_i, \mathbf{b}_{-i}) = \sum_{j=1}^{\ell} z_j \cdot [\text{jump in } x_i(\cdot, \mathbf{b}_{-i}) \text{ at } z_j],$$

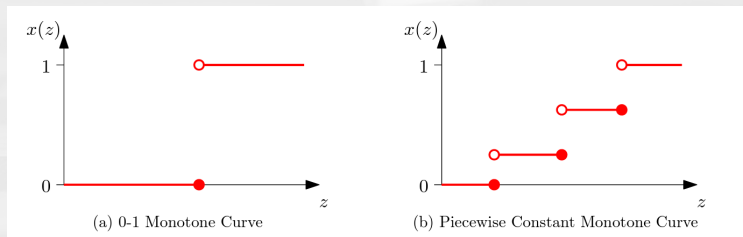
where z_1, \dots, z_{ℓ} are breakpoints of $x_i(\cdot, \mathbf{b}_{-i})$ in the range $[0, b_i]$.

Case: x is a piecewise constant function

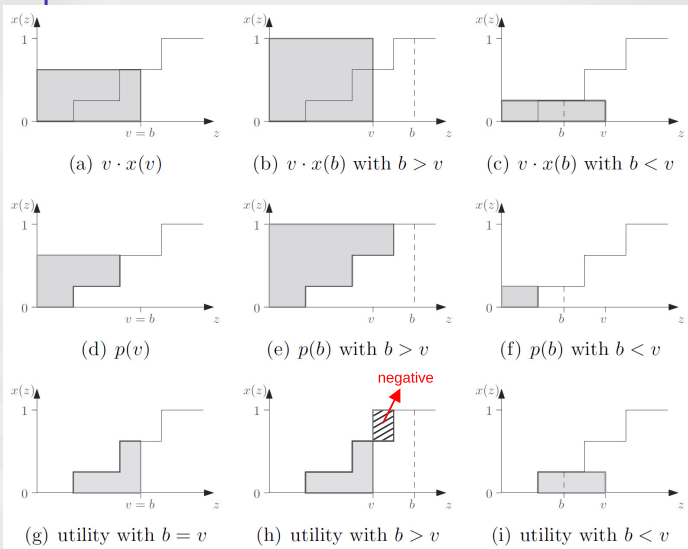
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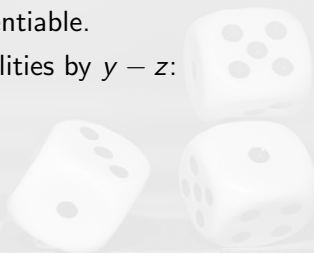
Case: x is a piecewise constant function



Case: x is a monotone function

$$z \cdot (x(y) - x(z)) \leq p(y) - p(z) \leq y \cdot (x(y) - x(z)).$$

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



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$$p'(z) = z \cdot x'(z).$$

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$$p_i(b_i, \mathbf{b}_{-i}) = \int_0^{b_i} z \cdot \frac{d}{dz} x_i(z, \mathbf{b}_{-i}) dz.$$

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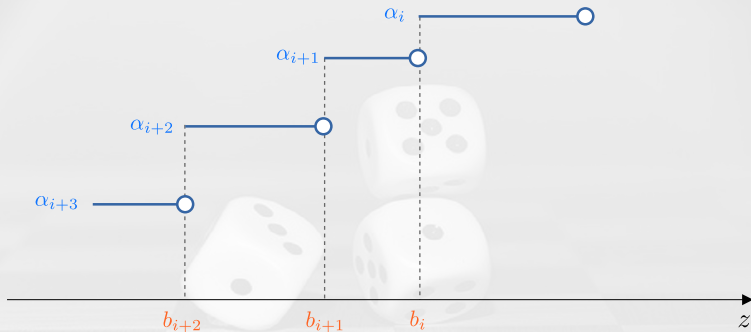
The Lemma

Application to the Sponsored Search Auction

Apply to Sponsored Search Auction

The allocation rule is piecewise.

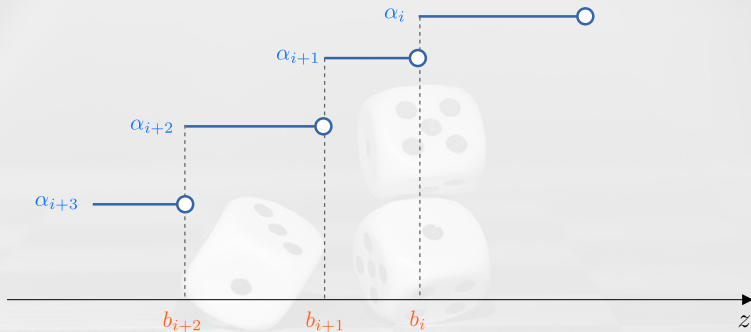
re-index the bidders: $b_1 \geq b_2 \geq \dots \geq b_n$.



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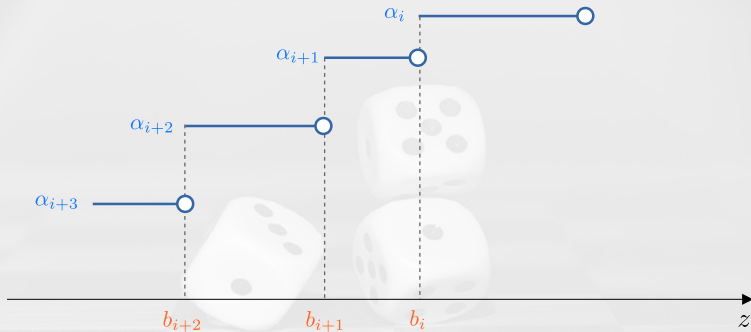


$$p_i(\mathbf{b}) = \sum_{j=i}^k b_{j+1}(\alpha_j - \alpha_{j+1}).$$

Apply to Sponsored Search Auction

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$$p_i(\mathbf{b}) = \sum_{j=i}^k b_{j+1} \frac{\alpha_j - \alpha_{j+1}}{\alpha_i} \text{ (scaled per click).}$$

Exercise 1 (4%)

- ▶ Recall that in the model of sponsored search auctions:
 - ▶ There are k slots, the j th slot has a click-through rate (CTR) of α_j (nonincreasing in j).
 - ▶ The utility of bidder i in slot j is $\alpha_j(v_i - p_j)$, 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 $b_1 \geq b_2 \geq \dots \geq b_n$.
 2. For $i = 1, 2, \dots, k$, assign the i th bidder to the i slot.
 3. For $i = 1, 2, \dots, k$, charge the i th bidder a price of b_{i+1} per click.
- (a) Prove that for every $k \geq 2$ and sequence $\alpha_1 \geq \dots \geq \alpha_k > 0$ of CTRs, the GSP auction is **NOT** DSIC. (*Hint: Find out an example.*)
- (b) A bid profile \mathbf{b} with $b_1 \geq \dots \geq b_n$ is **envy-free** if for every bidder i and slot $j \neq i$,

$$\alpha_i(v_i - b_{i+1}) \geq \alpha_j(v_i - b_{j+1}).$$

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