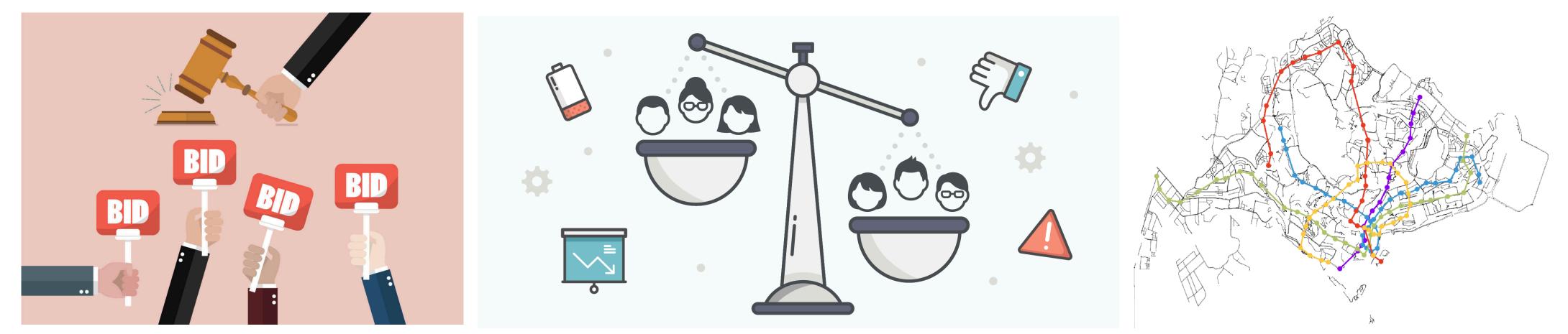
CSCI 357: Algorithmic Game Theory Lecture 4: Myerson's Lemma



Shikha Singh

Announcements and Logistics

- Hand in **Homework 2**
- Paper evaluation 1 (due next Fri): Case study of internet ad auctions
 - Read the research paper
 - **Part A**: Submit a google form individually
 - **Part B**: Work on technical analysis in groups of 4
 - Each group must turn in their write up of at least 3 out of 5 proofs in class and present one of them on the board
- Assignment 2 will be released on Mon and due the following week



Last Time

- Discussed single item (sealed bid) auctions
- Second price (Vickrey auctions) are dominant strategyproof and maximize welfare in linear time
- Ran a first price auction:
 - We will discuss the results next week, stay tuned!

Single-Parameter Mechanism Design

Multiple items but each agent has a single valuation for their allocation

n buyer with private valuations which can be described by a single number v_i



Multiple items



Example: k identical goods

- Simple example of single-parameter setting: we have k copies on an item
- Feasible allocation is then $X = (x_1, \dots, x_n) \subseteq \{0,1\}^n$, where $x_i = 1$ if bidder gets an copy; 0 otherwise and $\sum x_i \leq k$ i=1

n buyers, each has private value v_i for a single copy of the item





Example: Single Subset Case

- Each buyer *i* has value v_i for a certain subset $S_i \subseteq S$, 0 other others
- Feasible allocation is $X = (T_1, ..., T_n)$ where each $T_i \subseteq S$

n buyers but each buyer only wants a certain subset

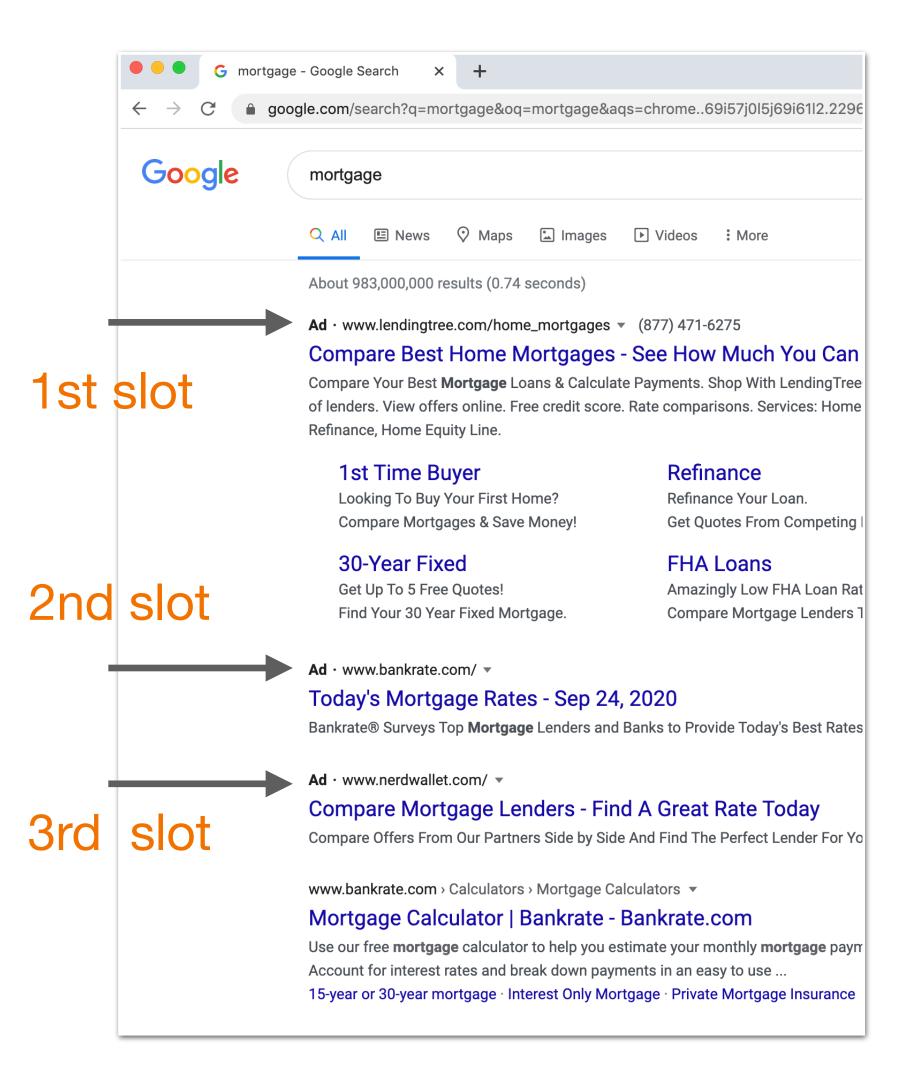


Multiple items



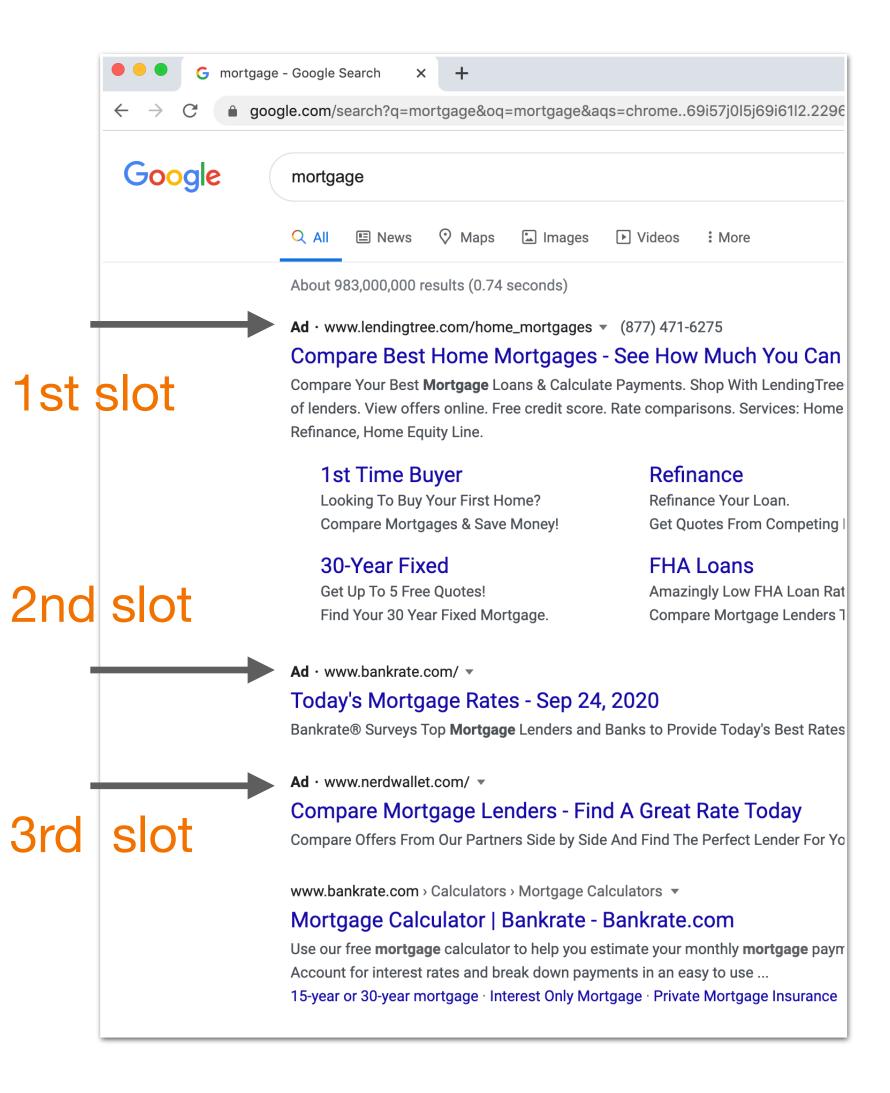
Sponsored Search Model [Edelman & Varian]

- Every time someone searches a query, an auction is run in real time to decide: which advertisers links are shown, in what order, and how they are charged
- We look at a simplified but effective model to study sponsored search auction
- Items for sale are k slots for sponsored links on a page
- Bidders (advertisers) have a standing bid on a keyword that was searched on
- Slots higher up on the page are more valuable than low
 - Users more likely to click on them



Sponsored Search Model [Edelman & Varian]

- Slots higher up on the page more likely to be clicked
 - Quantified through click-through-rates (CTRs)
 - CTR α_i of a slot j is the probability of clicks it is expected to receive
 - Reasonable to assume $\alpha_1 \geq \alpha_2 \geq \ldots \geq \alpha_n$
- **Simplifying assumption.** CTR of a slot is independent of its occupant, that is, doesn't depend on the quality of the ad
- Assume advertisers have a private valuation v_i for each click on its link: value derived from slot j by advertiser i is $v_i \cdot \alpha_i$



Example: Sponsored Search

- A feasible allocation is an assignment of bidders to slots, such that each slot is assigned to at most one bidder and each bidder is assigned at most one slot, that is, $X = (x_1, x_2, ..., x_n)$
 - where $x_i = \alpha_i$, the click through of slot j if bidder i is assigned to it; otherwise $x_i = 0$ if bidder is unassigned

n buyers, each has private value of v_i "per click" they get



k slots, with different click-through rates α_i



Sealed-Bid Mechanism

- We will focus on sealed-bid mechanisms that
 - Collect bids/reports $\mathbf{b} = (b_1, \dots, b_n)$
 - Choose a feasible allocation rule $\mathbf{x}(\mathbf{b}) \in X \subseteq \mathbb{R}^n$
 - Choose payments $\mathbf{p}(\mathbf{b}) \in \mathbb{R}^n$
- Such mechanisms are called direct-revelation mechanism
 - Mechanisms that ask agents to report their private value up front
- Quasilinear utility: $u_i(\mathbf{b}) = v_i \cdot x_i(\mathbf{b}) p_i(\mathbf{b})$ on the bid profile **b**
- We will focus on payment rules that satisfy
 - $p_i(\mathbf{b}) \ge 0$: sellers can't pay the bidders
 - $p_i(0, \mathbf{b}_{-i}) = 0$: a zero bid leads to a zero payment



Design Approach

Our goal is to **maximize surplus argmax**

- Challenge: jointly design two pieces: who gets what, and how much do they pay
 - Not enough to figure out who wins, if don't charge them the right amount
- Usually, the recipe we will follow:
 - Step 1. Assume truthful bids, and decide how to allocate so as to maximize surplus (in polynomial time)
 - **Step 2.** Using the allocation in step 1, decide how to charge payments so as that the mechanism is strategyproof (DSIC)

$$(x_1,\ldots,x_n) \in X \sum_{i=1}^n v_i x_i$$

k identical goods: Allocation

- Collect sealed bids
- Who should we give the k items to maximize surplus (assuming truthful bids)
 - Top k bidders

n buyers, each has private value v_i for a single copy of the item



• **Question**. What should we charge them so that truth telling is dominant strategy?



Sponsored Search: Allocation

How do we do we assign slots to maxim

- Greedy allocation is optimal (can be showed by an exchange argument)
- Recall that CTR rates $\alpha_1 \geq \alpha_2 \geq \cdots \geq \alpha_k$
- Sort and relabel bids $b_1 \ge b_2 \ge \cdots$
- Assign *j*th highest bidder to *j*th highest slot
- Can we create a payment rule (an analog of second-price rule) that makes the greedy allocation incentive compatible?

nize
$$\sum_{i=1}^{n} b_i x_i$$
?

$$\geq b_n$$

Towards a General Characterization

- **Question.** Can any allocation rule be paired with a payment rule such that the mechanism is strategyproof (truthtelling is a dominant strategy)?
 - When is this possible and how should we design the payment rule?
- Myerson's lemma gives a general characterization of allocation rules that can be turned into a truthful mechanism
- And tells us exactly how to design payment rules to achieve that

Myerson's Lemma: Informal

- In a fixed-parameter setting,
 - an allocation rule x can be made dominant-strategy incentive compatible if and only if x is monotone (non decreasing), and
 - if x is monotone, there is a unique payment rule p such that (x, p) is dominant strategyproof.

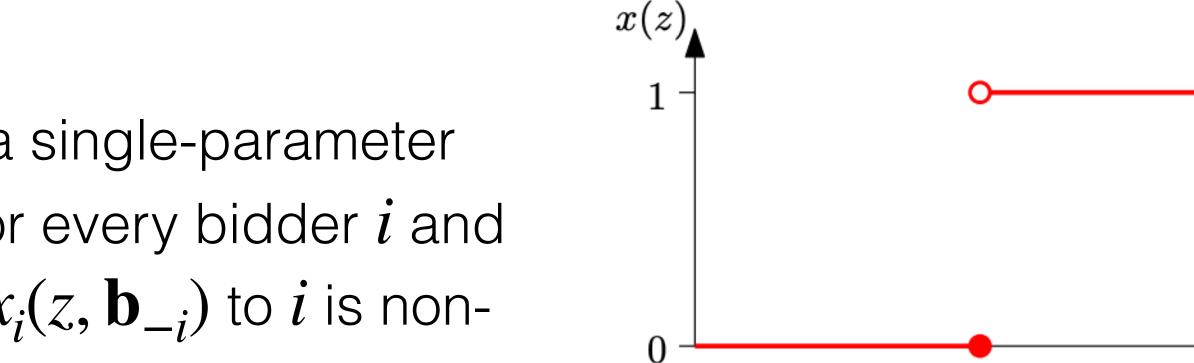
Implications of Myerson's Lemma

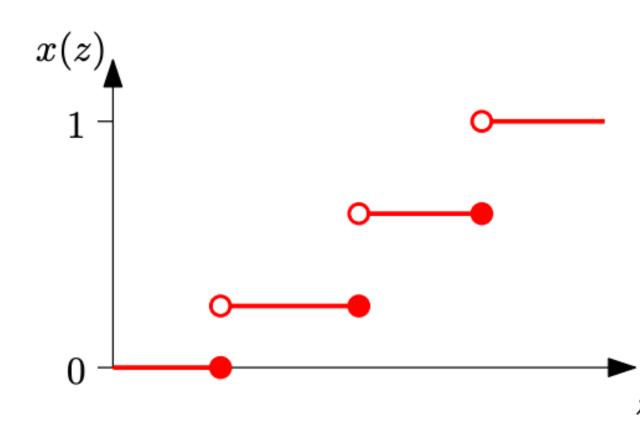
- Very powerful characterization
- Our initial design dilemma: can we make some allocation rule **x** dominant strategyproof by pairing it with an appropriate payment rule?
- Myerson's lemma takes this question and turns into one that is more wieldy and operational: checking if ${\bf x}$ is monotone
- If an allocation rule is monotone, the lemma says there is exactly one way to assign payments to make it dominant strategyproof
 - A direct formula for the payments

Monotone Allocation Rule

Definition.

An allocation rule $\mathbf{x} = (x_1, \dots, x_n)$ for a single-parameter domain is monotone-non-decreasing if for every bidder i and bids \mathbf{b}_{-i} of other bidders, the allocation $x_i(z, \mathbf{b}_{-i})$ to *i* is nondecreasing in its bid z.

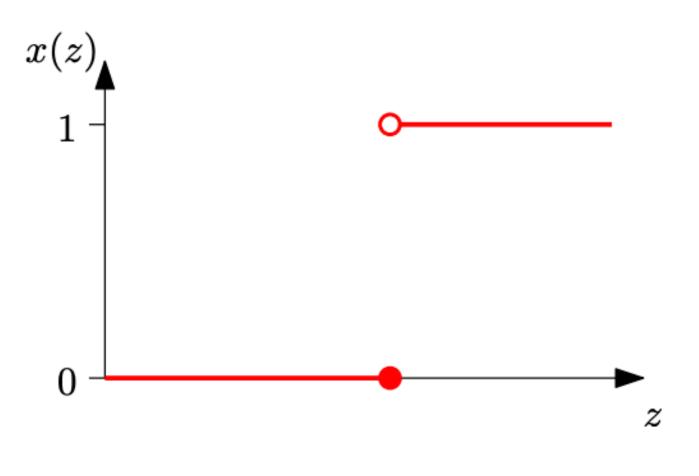


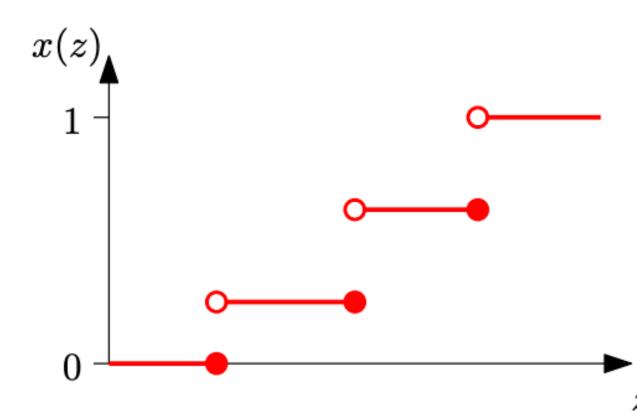




Monotone Allocation Rule

- That is, in a monotone allocation rule, bidding higher can only get you "more" stuff
- Example of a monotone allocation rule?
- Example of a non-monotone allocation rule?





Myerson's Lemma: Proof

- Part 1: An allocation \mathbf{x} rule can be made dominant strategyproof only if \mathbf{x} is monotone
- Part 2: A mechanism (x, p), where is x is monotone, is dominant strategyproof only if
 p is given by the expression in Myerson's lemma
- Part 3: Finally, we show that if the allocation x is monotone and the payment rule p is as given by the expression in the lemma then, (x, p) is dominant strategyproof.

Myerson's Lemma: Proof

- Recall dominant strategyproof condition:
 - for every agent i, every possible private valuation v_i , every set of bids \mathbf{b}_{-i} by the other agents, i's utility is maximized by bidding truthfully
- Fix an arbitrary player i and bid profile of others \mathbf{b}_{-i}
- Let x(z) and p(z) be shorthand for i's allocation $x_i(z, \mathbf{b}_{-i})$ & payment $p_i(z, \mathbf{b}_{-i})$
- Throughout the proof, we will vary the bid z and see how it changes the allocation

- **Part 1.** An allocation rule **x** can be made dominant-strategy incentive compatible only if \mathbf{x} is monotone non-decreasing
- If player i (with value v) deviates and bids as if she has value z, then her utility is $v \cdot x(z) - p(z)$
 - Notice: no control over your value v
- For truth telling to be a (weakly) dominant strategy for all values, must be that †) for all v, v^{\dagger}

•
$$v \cdot x(v) - p(v) \ge v \cdot x(v^{\dagger}) - p(v^{\dagger})$$

- We consider two possible values z_1, z_2 with $z_1 < z_2$
 - Case 1 (Underbidding): $v = z_2$, $v^{\dagger} = z_1$
 - Case 2 (Overbidding): $v = z_1$, $v^{\dagger} = z_2$

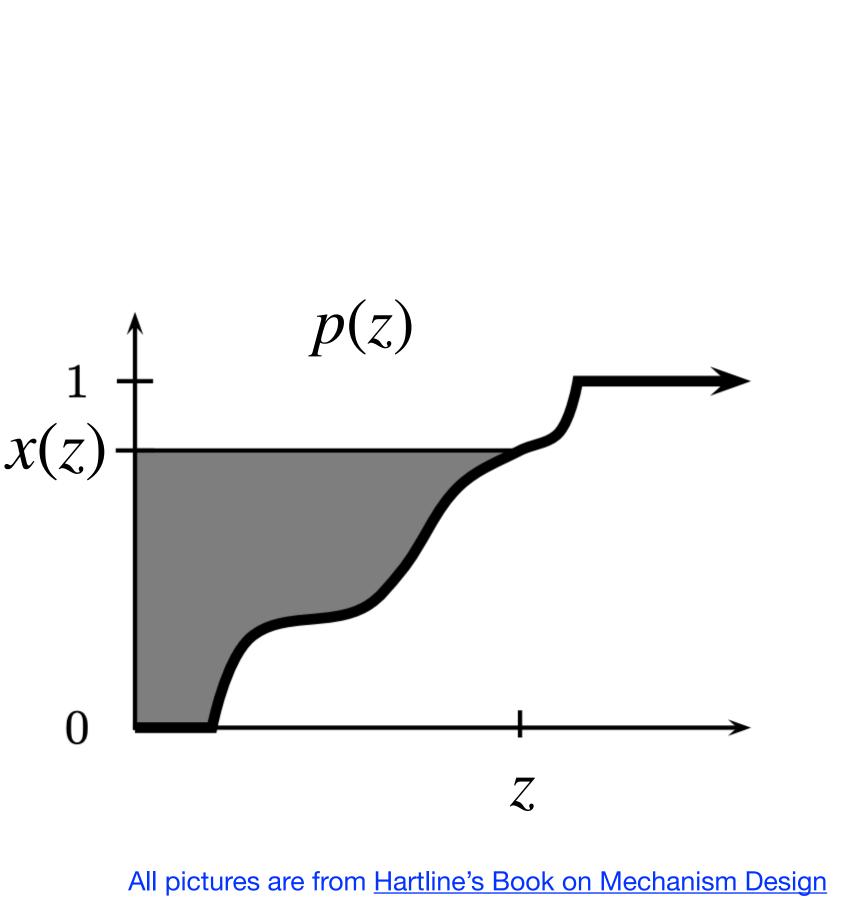
- In case (a), where $v = z_2$ and player underbids z_1 $z_2 \cdot x(z_2) - p(z_2) \ge z_2 \cdot x(z_1) - p(z_1) - (\text{lneq 1})$
- In case (b), where $v = z_1$ and player overbids z_2 $z_1 \cdot x(z_1) - p(z_1) \ge z_1 \cdot x(z_2) - p(z_2) - (\text{lneq 2})$
- Adding both: $z_2 \cdot x(z_2) + z_1 \cdot x(z_1) \ge z_2 \cdot x(z_1) + z_1 \cdot x(z_2)$
- Rearranging: $(z_2 z_1) \cdot (x(z_2) x(z_1)) \ge 0$
 - Does this imply something about the allocation rule \mathbf{x} ?
- Since $z_2 > z_1$, this only holds if $x(z_2) \ge x(z_1)$: thus **x** must be monotone nondecreasing \blacksquare (Part 1)

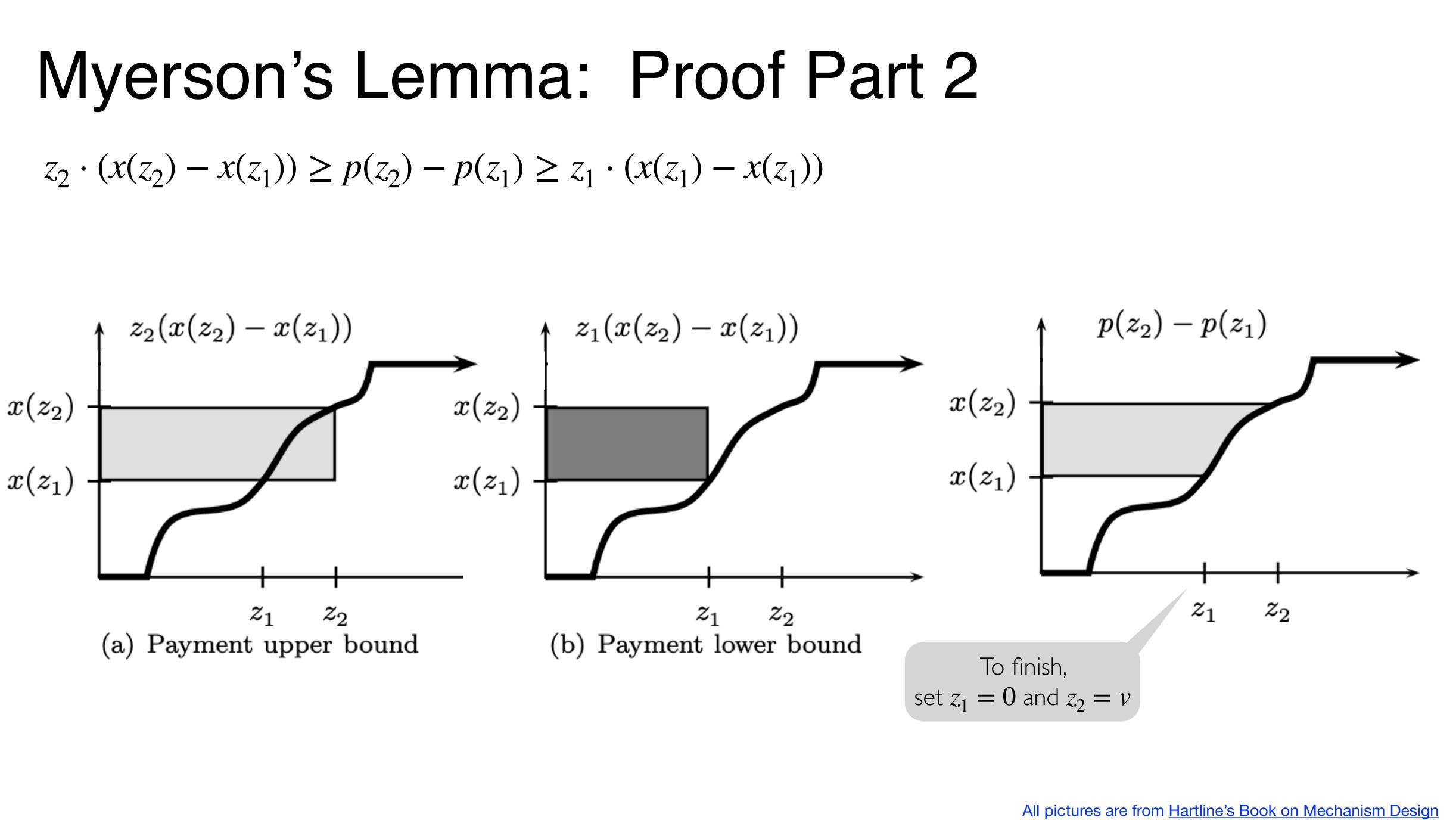
All pictures are from Hartline's Book on Mechanism Design



• **Part 2.** Suppose mechanism (**x**, **p**) dominant-strategyproof, where **x** is monotone, let's derive **p**

- We reuse the inequalities from part 2 of the proof: $z_2 \cdot x(z_2) - p(z_2) \ge z_2 \cdot x(z_1) - p(z_1) - (\text{lneq 1})$ $z_1 \cdot x(z_1) - p(z_1) \ge z_1 \cdot x(z_2) - p(z_2) - (\text{lneq 2})$
- We can upper and lower bound $p(z_2) p(z_1)$ using them as $z_2 \cdot (x(z_2) - x(z_1)) \ge p(z_2) - p(z_1) \ge z_1 \cdot (x(z_2) - x(z_1))$





Myerson Payment Rule

This payment rule is given by the following expression for all i:

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

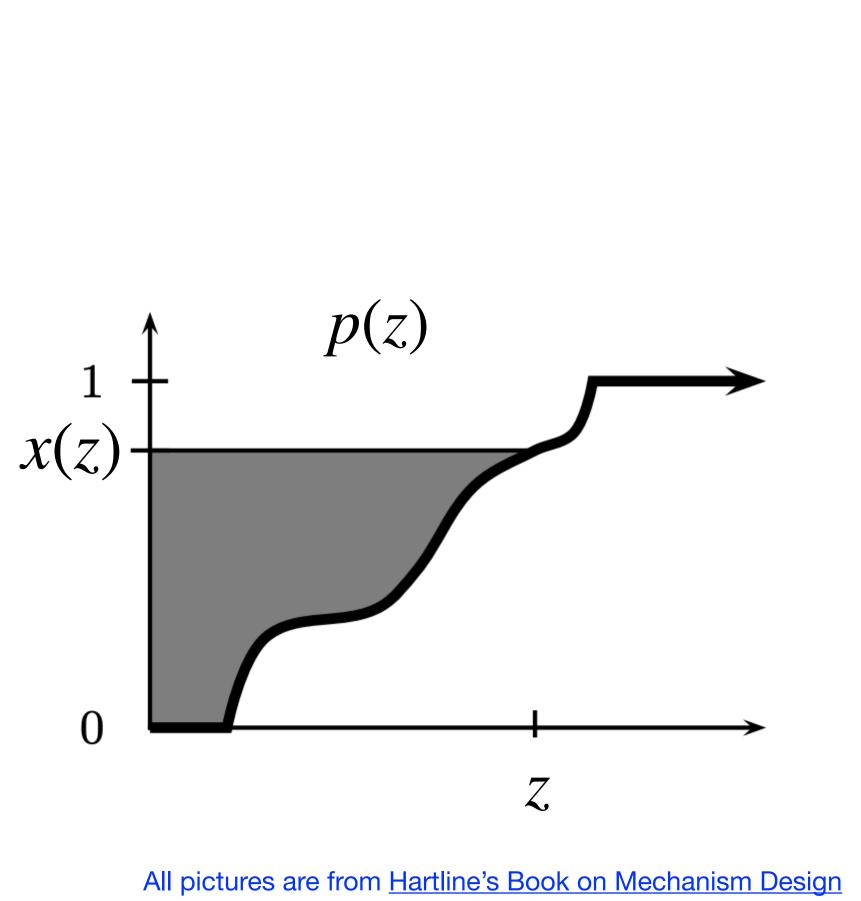
where player i bids z.

Keeping \mathbf{b}_{-i} fixed, we can simplify:

$$p_i(z) = z \cdot x_i(z) - \int_0^z x_i(z) \, dz$$

Assuming that $p_i(0) = 0$.

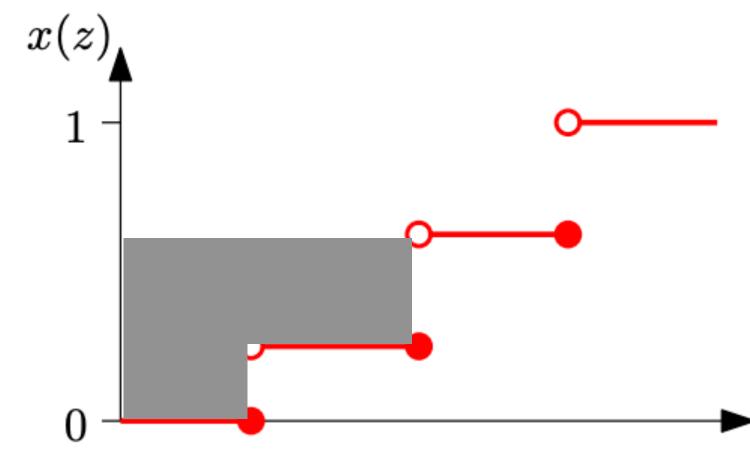
 $\mathbf{b}_{-\mathbf{i}}$) dz



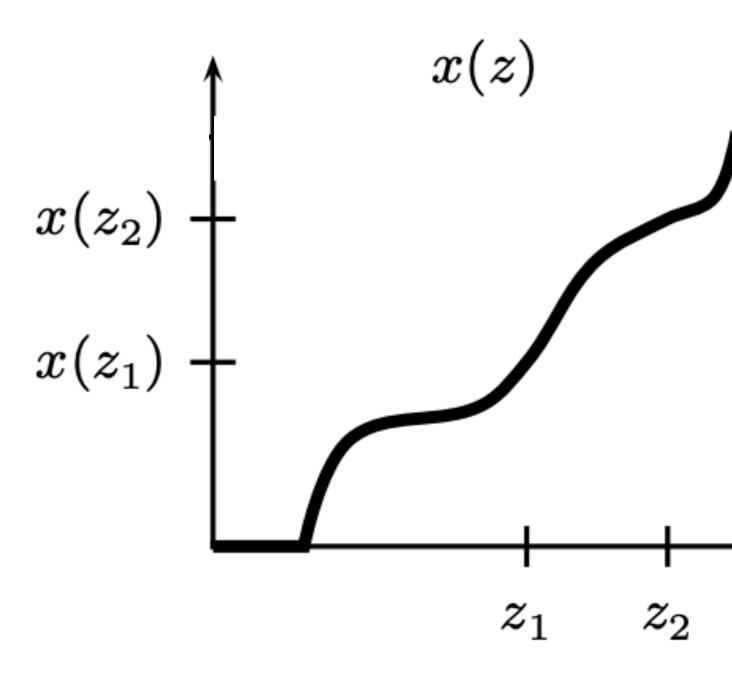
Myerson Payment Rule

- Suppose **x** is piecewise constant
- If there are ℓ points at which the allocation "jumps" before bid z, the payment at bid z

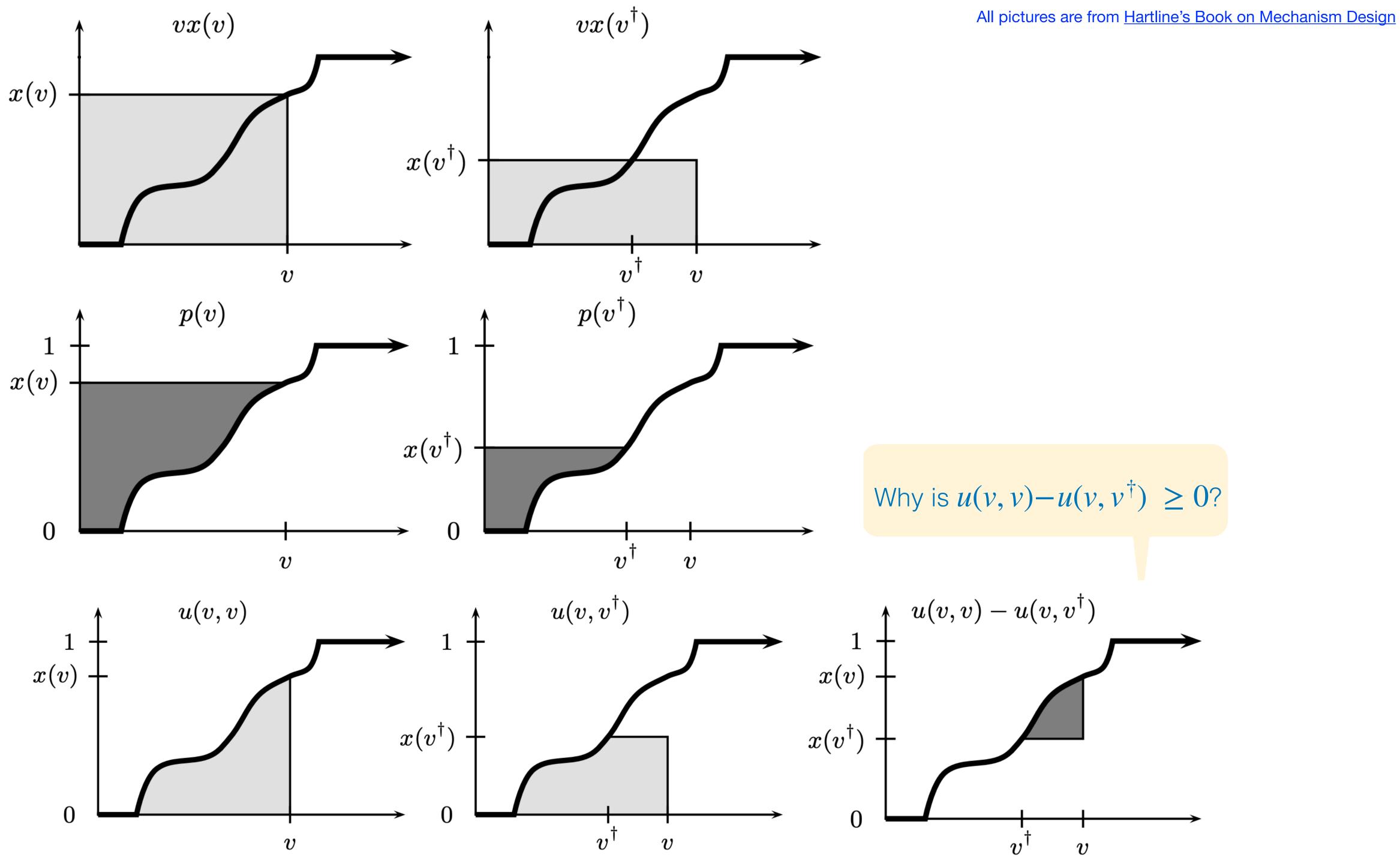
$$p_i(z) = \sum_{j=1}^{\ell} z_j \cdot [\text{jump in } x_i \text{ at } z_j]$$



- Part 3. If the allocation x is monotone and the payment rule p is as given by the expression in the lemma then, (x, p) is dominant strategyproof
- Suppose Alice's value is $v = z_2$, and she **underbids** $v^{\dagger} = z_1$
- We will compare utilities $v \cdot x(v) p(v)$ and $v \cdot x(v^{\dagger}) p(v^{\dagger})$



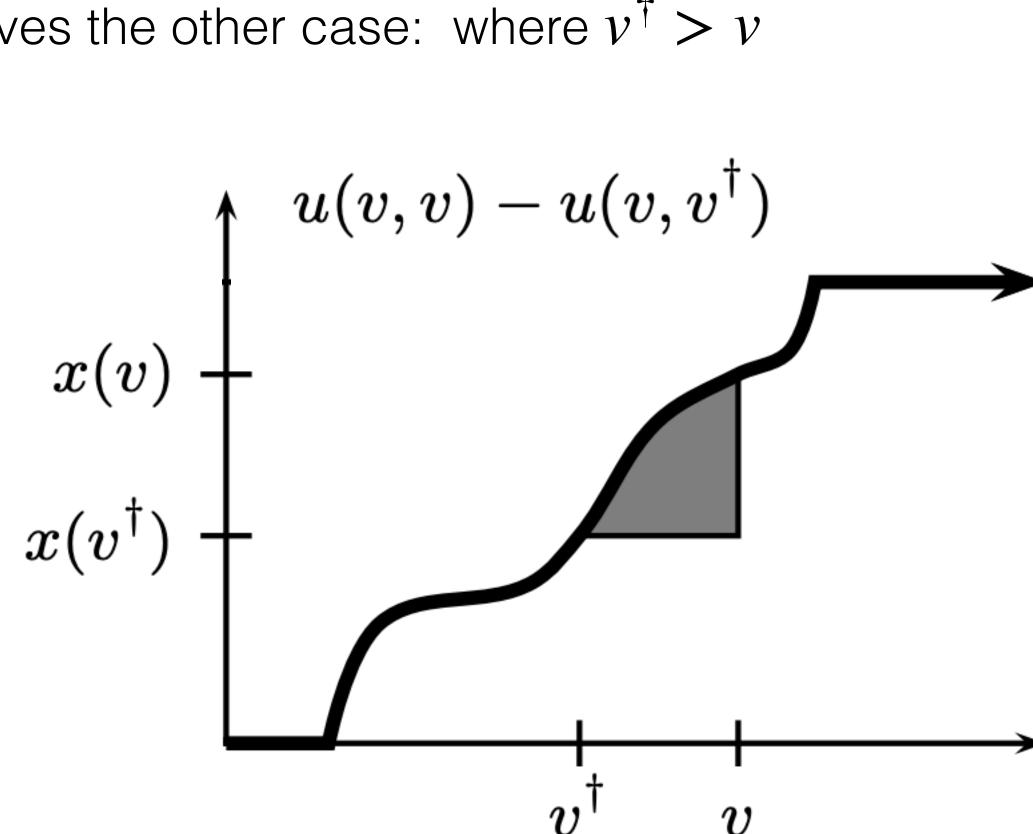






• $u(v, v) - u(v, v^{\dagger}) \ge 0$ because **x** is monotone non-decreasing

- Since $v > v^{\dagger}$, we have $x(v) \ge x(v^{\dagger})$
- A similar argument proves the other case: where $v^{\dagger} > v$





Myerson's Lemma Complete

• Fix an single-parameter domain. We state the result for the continuous case.

(a) An allocation rule \mathbf{x} can be made dominant-strategy incentive compatible if and only if \mathbf{X} is monotone (non decreasing).

(b) If x is monotone, there is a **unique** payment rule p such that (x, p) is dominant strategyproof. This payment rule is given by the following expression for all i:

$$p_i(z, \mathbf{b}_{-\mathbf{i}}) = z \cdot x_i(z, \mathbf{b}_{-\mathbf{i}}) - \int_0^z x_i(z, \mathbf{b}_{-\mathbf{i}})$$

where player *i* bids *z*. Keeping \mathbf{b}_{-i} fixed, we can simplify:

$$p_i(z) = z \cdot x_i(z) - \int_0^z x_i(z) \, dz$$

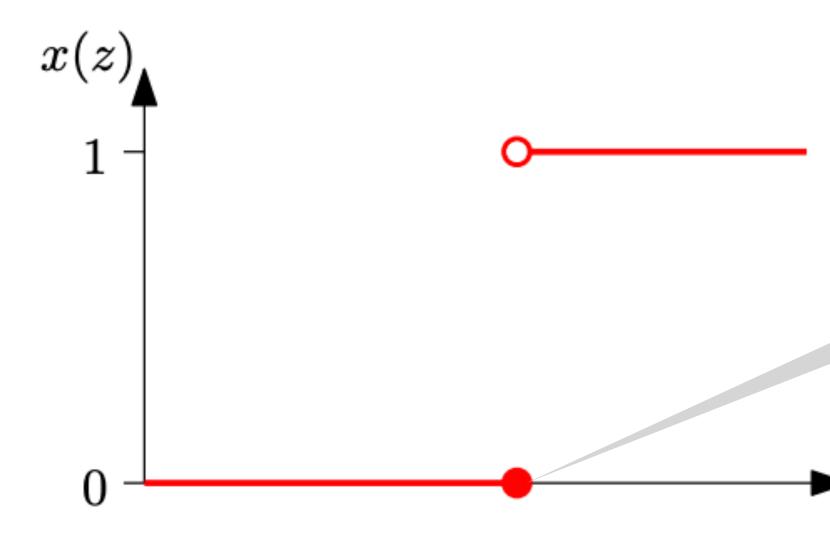
Assuming that $p_i(0) = 0$.

dz

Examples

Single-Item Auction

- Let's apply Myerson's lemma to a single item auction that allocates the item to highest bidder
- This allocation rule is monotone: in fact a 0/1 monotone curve
- Fixing \mathbf{b}_{-i} , we can plot bidder *i* allocation wrt to bid:

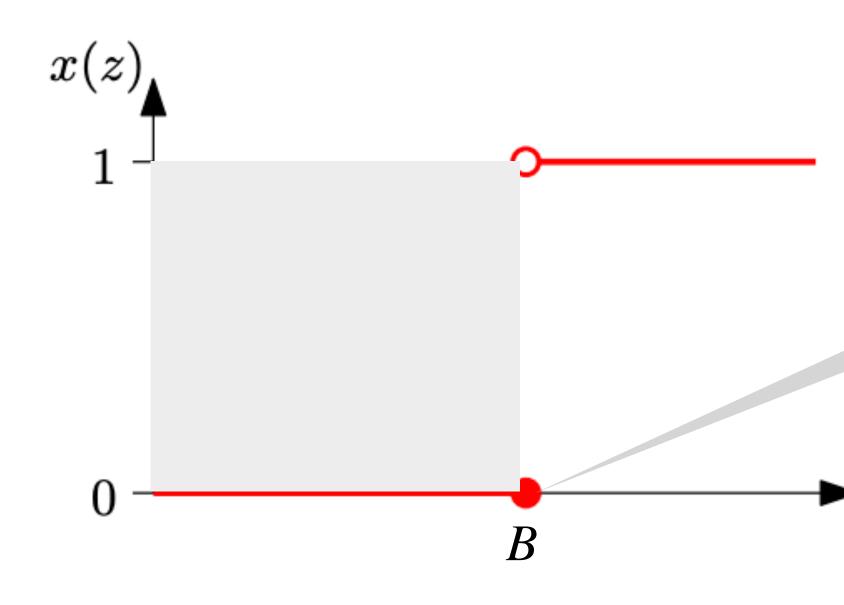


z

This jump occurs exactly at $B = \max \mathbf{b}_{-i}$, called critical bid

Single-Item Auction

- If z < B: payment is 0
- If $z \ge B$: payment is given by shaded region, that is, B
- We have recreated the Vickrey auction from Myerson's lemma
- Moreover, this payment scheme is the only way to make the allocation rule (giving to highest bidder) truthful!

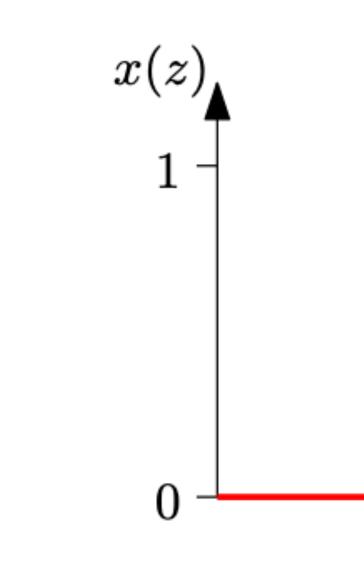


z

This jump occurs exactly at $B = \max \mathbf{b}_{-i}$, called **critical bid**

Any 0/1 Allocation Mechanism

- In a single-parameter environment, let *X* be any 0/1 feasible allocation (each player either wins $x_i = 0$ or loses $x_i = 1$)
 - Example: auctioning k units of the same item to n bidders
- In such auctions, what should the winners pay?



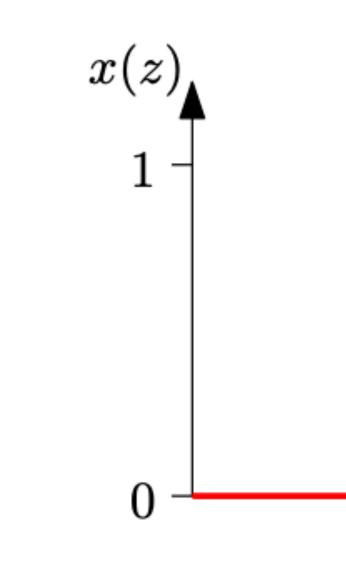
$$p(b_i, \mathbf{b}_{-i}) = \begin{cases} 0 & \text{if } x_i(b_i, \mathbf{b}_{-i}) = 0 \\ b_i^*(\mathbf{b}_{-i}) & \text{if } x_i(b_i, \mathbf{b}_{-i}) = 1 \end{cases}$$

Critical bid: $b_i^*(\mathbf{b}_{-i})$ lowest bid at which *i*'s allocation goes from 0 to 1



Any 0/1 Allocation Mechanism

- In a single-parameter environment, let *X* be any 0/1 feasible allocation (each player either wins $x_i = 0$ or loses $x_i = 1$)
 - Example: auctioning k units of the same item to n bidders
- In such auctions, what should the winners pay?
 - $(k+1)^{st}$ highest bid



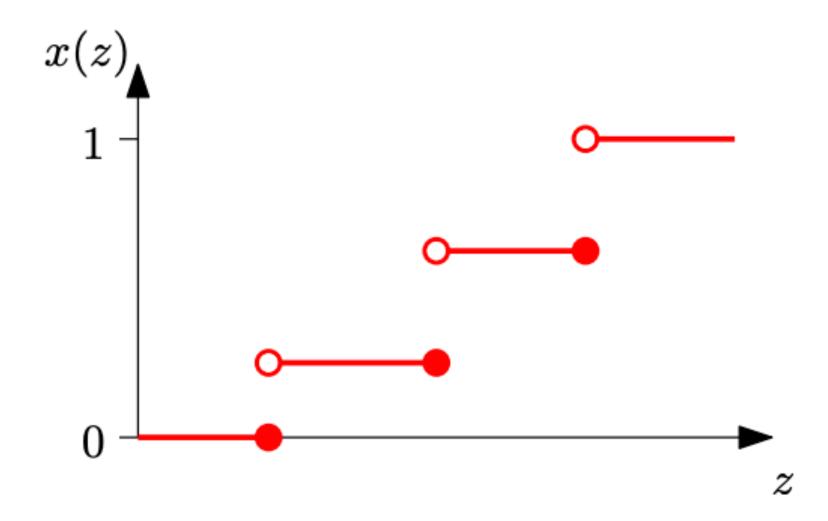
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Critical bid: $b_i^*(\mathbf{b}_{-i})$ lowest bid at which *i*'s allocation goes from 0 to 1

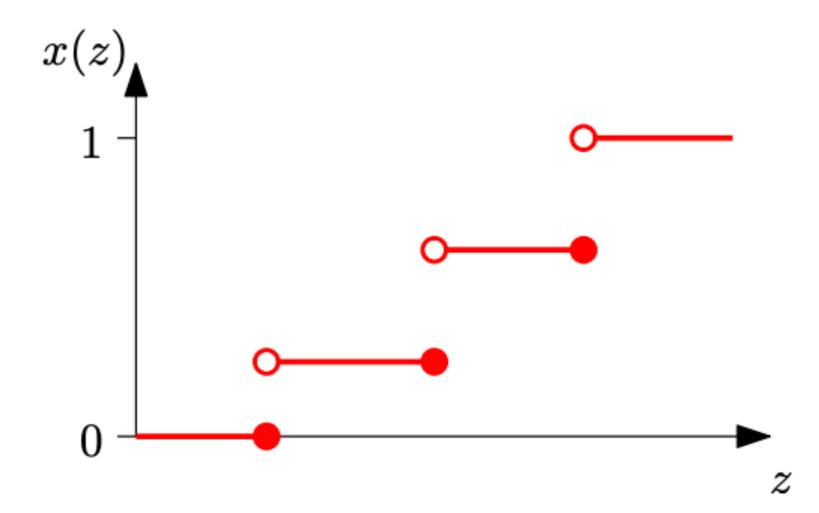


Sponsored-Search Auctions

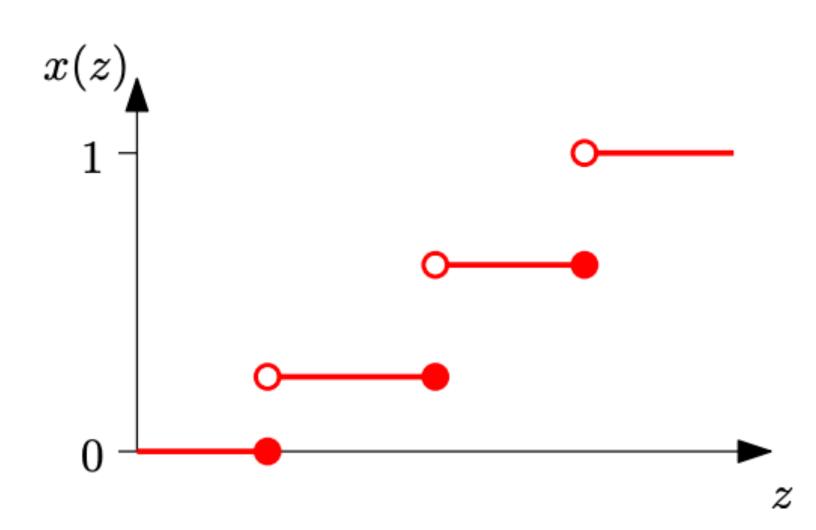
- Sort bids $b_1 \ge b_2 \ge \ldots \ge b_n$ (reorder bidders in this order)
- Assign slot 1 to bidder 1, slot 2 to bidder 2, etc.
- That is, CTR $lpha_i$ of slot j gets assigned to bidder j
- What does the graph of such an allocation rule look like?
 - For intuition fix b_{-i} and think of yourself as bidder 1 slowing raising your value from 0



- If you get no slot, you pay zero
- If you get last slot, you pay the "critical" bid that you beat out to get the slot (the bid of the person just below you in sorted order)
- If you get a lower slot j better than k, what do you pay?
 - **Exercise:** come with the expression for the payment p_j of bidder who wins slot j using Myerson's rule?
- We will come back to this!



- If you get no slot, you pay zero
- If you get last slot, you pay the "critical" bid that you beat out to get the slot (the bid of the person just below you in sorted order)
- If you get slot $1 \le j \le k$, what do you pay?
 - Exercise: come with the expression for the payment of bidder who wins slot j using Myerson's rule?
- We will come back to this!



- Myerson's payment rule of monotone piece-wise constant allocation
- If there are ℓ points at which the allocation "jumps" before bid z, the payment at bid z

$$p_i(z) = \sum_{j=1}^{\ell} z_j \cdot [\text{jump in } x_i \text{ at } z_j]$$

• Using Myerson's lemma, the *i*th highest bidder (who wins slot *i*) should pay:

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

- Payments have a nice interpretation:
 - If you win, you pay a suitable convex combination of lower bids!

 $\alpha_{k+1} = 0$ The "**per click payment**" of bidder *i* who is in slot *i* is $\sum_{j=i}^{\kappa} b_{j+1} \cdot \frac{\alpha_j - \alpha_{j+1}}{\alpha_i}$

Question. Are sponsored-search auctions

in real life based on our (Myerson's) theory?

Generalized Second Price Auctions

- By "historical accident," the sponsored search auctions in real life (called generalized-second price auction or GSP) are not DSIC
- In GSP, the allocation rule is the same
 - Allocate slots to highest bidders
- Payment rule: a bidder wins slot i pays the per-click bid of the winner of slot i 1 or 0 if i = k (rather than a convex combination of lower bids)
 - Some say Google incorrectly implemented Myerson's lemma
 - Most likely reason is that the payment rule of GSP is much easier to explain to advertisers and share-holders
- Which one is better for revenue?
 - We'll explore this question next week