Mechanism design: Single-parameter environments

Alexandros Voudouris

University of Oxford

- A seller with one item for sale
- n agents

- A seller with one item for sale
- *n* agents
- Each agent i has a **private value** v_i for the item
 - This value represents the **willingness-to-pay** of the agent; that is, v_i is the maximum amount of money that agent i is willing to pay in order to buy the item

- A seller with one item for sale
- *n* agents
- Each agent i has a **private value** v_i for the item
 - This value represents the **willingness-to-pay** of the agent; that is, v_i is the maximum amount of money that agent i is willing to pay in order to buy the item
- The utility of each agent is quasilinear in money:
 - If agent i loses the item, then her utility is 0
 - If agent i wins the item at price p, then her utility is $v_i p$

- General structure of an auction:
 - **Input:** every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price

- General structure of an auction:
 - Input: every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price
- Deciding the winner is easy: the highest bidder

- General structure of an auction:
 - Input: every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price
- Deciding the winner is easy: the highest bidder
- Deciding the selling price is more complicated

- General structure of an auction:
 - Input: every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price
- Deciding the winner is easy: the highest bidder
- Deciding the selling price is more complicated
 - A selling price of 0, creates a competition among the bidders as to who can think of the highest number

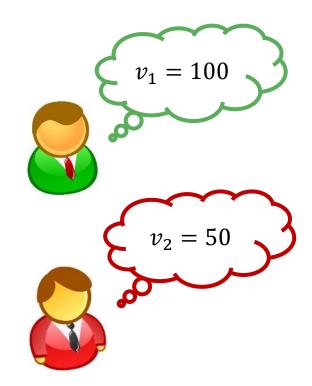
- General structure of an auction:
 - Input: every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price
- Deciding the winner is easy: the highest bidder
- Deciding the selling price is more complicated
 - A selling price of 0, creates a competition among the bidders as to who can think of the highest number
- We are interested in payment rules that incentivize the bidders to bid their true values

- General structure of an auction:
 - Input: every agent i submits a bid b_i (agents = bidders)
 - Allocation rule: decide the winner
 - Payment rule: decide a selling price
- Deciding the winner is easy: the highest bidder
- Deciding the selling price is more complicated
 - A selling price of 0, creates a competition among the bidders as to who can think of the highest number
- We are interested in payment rules that incentivize the bidders to bid their true values
 - Truthful auctions that maximize the social welfare

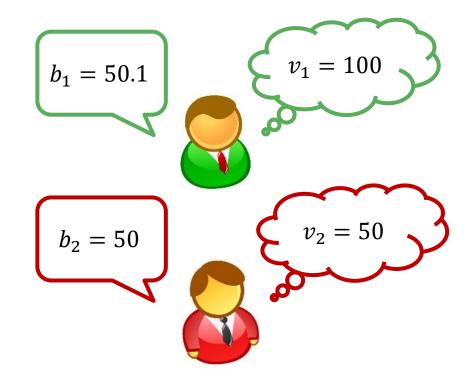
- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays her bid

- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays her bid
- Is this a truthful auction?

- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays her bid
- Is this a truthful auction?



- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays her bid
- Is this a truthful auction?



- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays the second highest bid

- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays the second highest bid

Theorem [Vickrey, 1961]

In a second-price auction

- (a) it is a dominant strategy for every bidder i to bid $b_i = v_i$, and
- (b) every truthtelling bidder gets non-negative utility

- Allocation rule: the winner is the highest bidder
- Payment rule: the winner pays the second highest bid

Theorem [Vickrey, 1961]

In a second-price auction

- (a) it is a dominant strategy for every bidder i to bid $b_i = v_i$, and
- (b) every truthtelling bidder gets non-negative utility
- (b) is obvious:
 - the selling price is at most the winner's bid, and the bid of a truthtelling bidder is equal to her true value

• For (a), our goal is to show that the utility of bidder i is maximized by bidding v_i , no matter what v_i and the bids of the other bidders are

- For (a), our goal is to show that the utility of bidder i is maximized by bidding v_i , no matter what v_i and the bids of the other bidders are
- Second highest bid: $B = \max_{j \neq i} b_j$
- The utility of bidder i is either 0 if $b_i < B$, or $v_i B$ otherwise

- For (a), our goal is to show that the utility of bidder i is maximized by bidding v_i , no matter what v_i and the bids of the other bidders are
- Second highest bid: $B = \max_{j \neq i} b_j$
- The utility of bidder i is either 0 if $b_i < B$, or $v_i B$ otherwise

Case I: $v_i < B$

- Maximum possible utility = 0
- Achieved by setting $b_i = v_i$

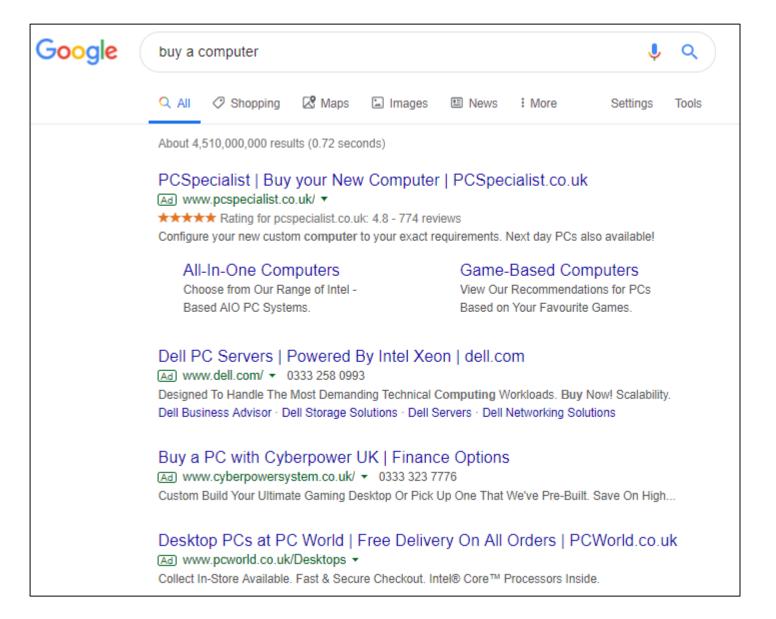
- For (a), our goal is to show that the utility of bidder i is maximized by bidding v_i , no matter what v_i and the bids of the other bidders are
- Second highest bid: $B = \max_{j \neq i} b_j$
- The utility of bidder i is either 0 if $b_i < B$, or $v_i B$ otherwise

Case I: $v_i < B$

- Maximum possible utility = 0
- Achieved by setting $b_i = v_i$

Case II: $v_i \geq B$

- Maximum possible utility = $v_i B$
- Bidder i wins the item by setting $b_i = v_i$



- k advertising slots
- n bidders (advertisers) who aim to occupy a slot

- *k* advertising **slots**
- *n* bidders (advertisers) who aim to occupy a slot
- Slot j has a **click-through-rate** (CTR) a_i
 - The CTR of a slot represents the probability that the ad placed at this slot will be clicked on
 - Assumption: the CTRs are independent of the ads that occupy the slots

- *k* advertising **slots**
- *n* bidders (advertisers) who aim to occupy a slot
- Slot j has a **click-through-rate** (CTR) a_j
 - The CTR of a slot represents the probability that the ad placed at this slot will be clicked on
 - Assumption: the CTRs are independent of the ads that occupy the slots
- The slots are ranked so that $a_1 \ge \cdots \ge a_k$

- k advertising slots
- *n* bidders (advertisers) who aim to occupy a slot
- Slot j has a **click-through-rate** (CTR) a_j
 - The CTR of a slot represents the probability that the ad placed at this slot will be clicked on
 - Assumption: the CTRs are independent of the ads that occupy the slots
- The slots are ranked so that $a_1 \ge \cdots \ge a_k$
- Each bidder i has a **private value** v_i **per click**
 - Bidder *i* derives utility $a_i \cdot v_i$ from slot *j*

Sponsored search auctions: goals

- **Truthfulness:** It is a dominant strategy for each bidder to bid her true value
- Social welfare maximization: $\sum_i v_i \cdot x_i$
 - $-x_i$ is the CTR of the slot that bidder i is assigned to, or 0 otherwise
- Poly-time execution: running the auction should be quick

Sponsored search auctions: goals

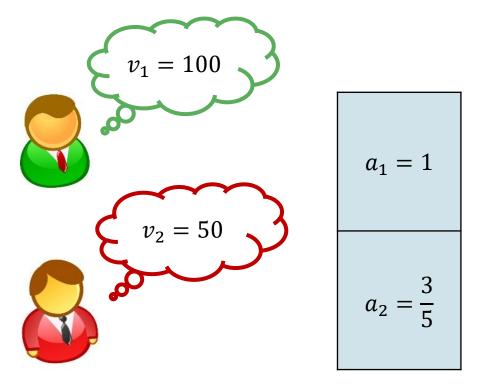
- **Truthfulness:** It is a dominant strategy for each bidder to bid her true value
- Social welfare maximization: $\sum_i v_i \cdot x_i$
 - $-x_i$ is the CTR of the slot that bidder i is assigned to, or 0 otherwise
- Poly-time execution: running the auction should be quick
- If the bidders are truthful, then maximizing the social welfare is easy: sort the bidders in decreasing order of their bids
- So, the problem is to incentivize them to be truthful, again

Sponsored search auctions: goals

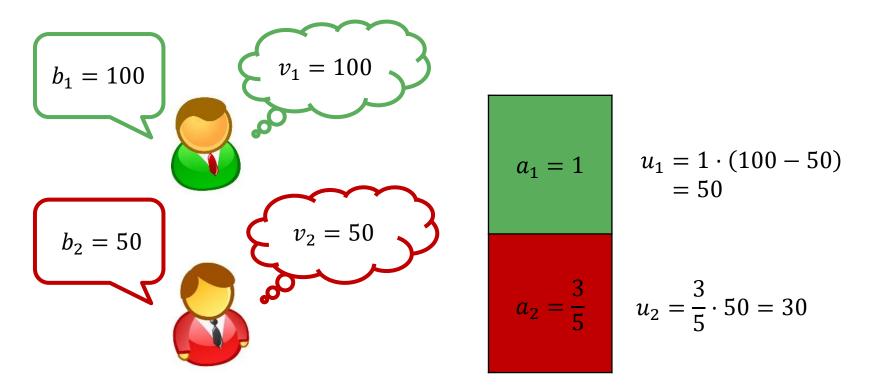
- Truthfulness: It is a dominant strategy for each bidder to bid her true value
- Social welfare maximization: $\sum_i v_i \cdot x_i$
 - $-x_i$ is the CTR of the slot that bidder i is assigned to, or 0 otherwise
- Poly-time execution: running the auction should be quick
- If the bidders are truthful, then maximizing the social welfare is easy: sort the bidders in decreasing order of their bids
- So, the problem is to incentivize them to be truthful, again
- Can we extend the ideas we exploited for single-item auctions?

- Allocation rule: sort the bidders in decreasing order of their bids and rename them so that $b_1 \geq \cdots \geq b_n$
- **Payment rule:** every bidder $i \le k$ (who is assigned at slot i) pays the next highest bid b_{i+1} per click, and every bidder i > k pays 0

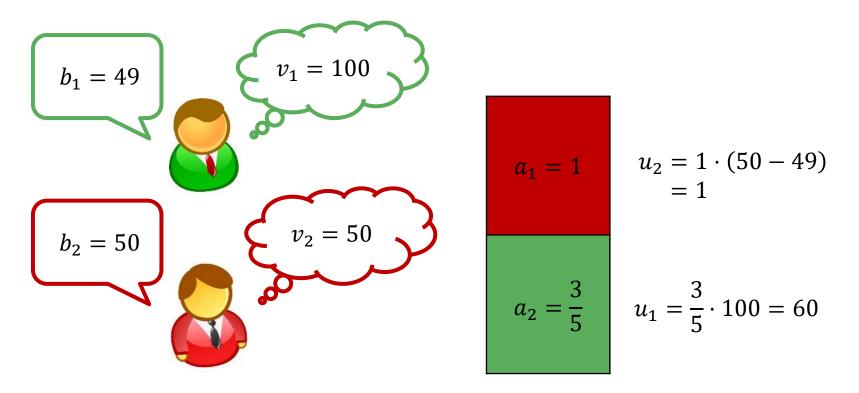
- Allocation rule: sort the bidders in decreasing order of their bids and rename them so that $b_1 \geq \cdots \geq b_n$
- Payment rule: every bidder $i \le k$ (who is assigned at slot i) pays the next highest bid b_{i+1} per click, and every bidder i > k pays 0



- Allocation rule: sort the bidders in decreasing order of their bids and rename them so that $b_1 \geq \cdots \geq b_n$
- **Payment rule:** every bidder $i \le k$ (who is assigned at slot i) pays the next highest bid b_{i+1} per click, and every bidder i > k pays 0



- Allocation rule: sort the bidders in decreasing order of their bids and rename them so that $b_1 \geq \cdots \geq b_n$
- **Payment rule:** every bidder $i \le k$ (who is assigned at slot i) pays the next highest bid b_{i+1} per click, and every bidder i > k pays 0



Myerson's Lemma

That didn't work for sponsored search auctions, so what now?

Myerson's Lemma

- That didn't work for sponsored search auctions, so what now?
- Let's try to see how the optimal truthful auction should look like, for any single parameter environment

Myerson's Lemma

- That didn't work for sponsored search auctions, so what now?
- Let's try to see how the optimal truthful auction should look like, for any single parameter environment
- Input by bidders: $\boldsymbol{b} = (b_1, \dots, b_n)$
- Allocation rule: $x(b) = (x_1(b), ..., x_n(b))$
- Payment rule: $p(b) = (p_1(b), ..., p_n(b))$
- The utility of bidder i is $u_i(\mathbf{b}) = v_i \cdot x_i(\mathbf{b}) p_i(\mathbf{b})$

- That didn't work for sponsored search auctions, so what now?
- Let's try to see how the optimal truthful auction should look like, for any single parameter environment
- Input by bidders: $\boldsymbol{b} = (b_1, \dots, b_n)$
- Allocation rule: $x(b) = (x_1(b), ..., x_n(b))$
- Payment rule: $p(b) = (p_1(b), ..., p_n(b))$
- The utility of bidder i is $u_i(\boldsymbol{b}) = v_i \cdot x_i(\boldsymbol{b}) p_i(\boldsymbol{b})$
- Focus on payment rules such that $p_i(\mathbf{b}) \in [0, b_i \cdot x_i(\mathbf{b})]$
 - $-p_i(\mathbf{b}) \ge 0$ ensures that the seller does not pay the bidders
 - $-p_i(\mathbf{b}) \leq b_i \cdot x_i(\mathbf{b})$ ensures non-negative utility for truthful bidders

• An allocation rule x is **implementable** if there exists a payment rule p such that (x, p) is a truthful auction

- An allocation rule x is **implementable** if there exists a payment rule p such that (x, p) is a truthful auction
- An allocation rule x is **monotone** if for every bidder i and bid vector b_{-i} , the allocation $x_i(z, b_{-i})$ is non-decreasing in the bid z of bidder i

- An allocation rule x is **implementable** if there exists a payment rule p such that (x, p) is a truthful auction
- An allocation rule x is **monotone** if for every bidder i and bid vector b_{-i} , the allocation $x_i(z, b_{-i})$ is non-decreasing in the bid z of bidder i

Lemma [Myerson, 1981]

- (a) An allocation rule x is implementable if and only if it is monotone
- (b) For every allocation rule x, there exists a unique payment rule p such that (x, p) is a truthful auction

- Fix a bidder i, and the bids b_{-i} of the other bidders
- Given that these quantities are now fixed, we simplify our notation:

$$-x(z)=x_i(z,\boldsymbol{b}_{-i})$$

$$-p(z)=p_i(z,\boldsymbol{b}_{-i})$$

$$- u(z) = u_i(z, \boldsymbol{b}_{-i})$$

- Fix a bidder i, and the bids b_{-i} of the other bidders
- Given that these quantities are now fixed, we simplify our notation:

$$-x(z)=x_i(z,\boldsymbol{b}_{-i})$$

$$-p(z)=p_i(z,\boldsymbol{b}_{-i})$$

$$- u(z) = u_i(z, \boldsymbol{b}_{-i})$$

- The idea:
 - assuming (x, p) is a truthful auction, the bidder has no incentive to unilaterally deviate to any other bid
 - This will give us a relation between x and p, which we can use to derive an explicit formula for p as a function of x

• Consider two bids $0 \le z < y$ and assume x is implementable by p

- Consider two bids $0 \le z < y$ and assume x is implementable by p
- True value = z, deviating bid = y:

$$u(z) \ge u(y)$$

- Consider two bids $0 \le z < y$ and assume x is implementable by p
- True value = z, deviating bid = y:

$$u(z) \ge u(y) \iff z \cdot x(z) - p(z) \ge z \cdot x(y) - p(y)$$

- Consider two bids $0 \le z < y$ and assume x is implementable by p
- True value = z, deviating bid = y:

$$u(z) \ge u(y) \iff z \cdot x(z) - p(z) \ge z \cdot x(y) - p(y)$$
$$\iff p(y) - p(z) \ge z \cdot \left(x(y) - x(z)\right)$$

- Consider two bids $0 \le z < y$ and assume x is implementable by p
- True value = z, deviating bid = y:

$$u(z) \ge u(y) \Leftrightarrow z \cdot x(z) - p(z) \ge z \cdot x(y) - p(y)$$

$$\Leftrightarrow p(y) - p(z) \ge z \cdot (x(y) - x(z))$$

• True value = y, deviating bid = z:

$$u(y) \ge u(z) \Leftrightarrow y \cdot x(y) - p(y) \ge y \cdot x(z) - p(z)$$

$$\Leftrightarrow p(y) - p(z) \le y \cdot (x(y) - x(z))$$

Combining these two, we get:

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

Combining these two, we get:

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

This also implies that

$$(y-z)\cdot (x(y)-x(z)) \ge 0$$

Combining these two, we get:

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

This also implies that

$$(y-z)\cdot (x(y)-x(z)) \ge 0$$

• Since $0 \le z < y$, this is possible if and only if x is monotone so that $y-z \le 0$ and $x(y)-x(z) \le 0$

Combining these two, we get:

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

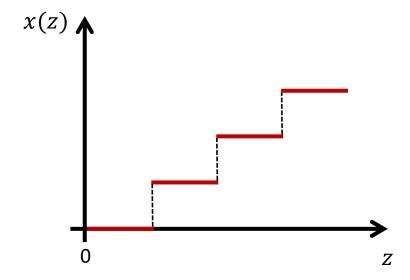
This also implies that

$$(y-z)\cdot (x(y)-x(z)) \ge 0$$

- Since $0 \le z < y$, this is possible if and only if x is monotone so that $y-z \le 0$ and $x(y)-x(z) \le 0$
 - \Rightarrow (a) is now proved

• We can now assume that x is monotone

- We can now assume that x is monotone
- Assume x is piecewise constant, like in sponsored search auctions



The break points are defined by the highest bids of the other bidders

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

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

• By fixing z and taking the limit as y tends to z, we have that

jump of
$$p$$
 at $z = z \cdot (\text{jump of } x \text{ at } z)$

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

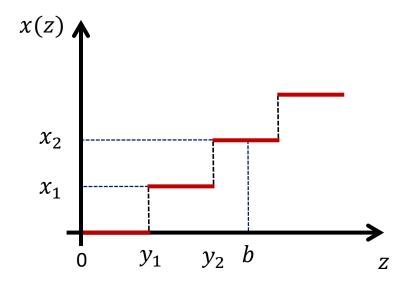
• By fixing z and taking the limit as y tends to z, we have that jump of p at $z=z\cdot (\text{jump of }x \text{ at }z)$

Therefore, we can define the payment of the bidder as

$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

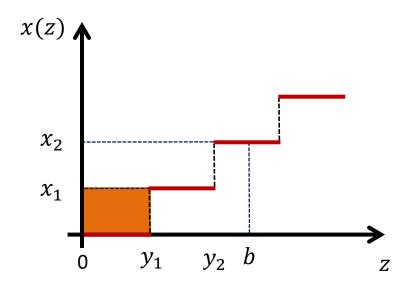
where y enumerates all break points of x in [0, b]

• Example:



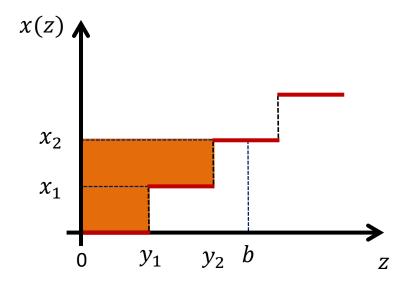
$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

• Example:

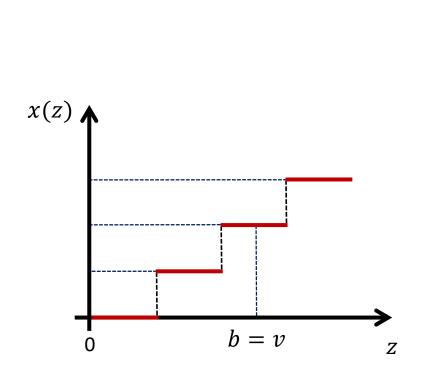


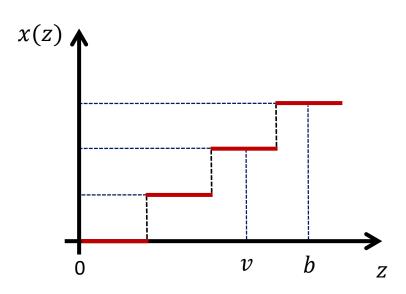
$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y) = y_1 \cdot x_1$$

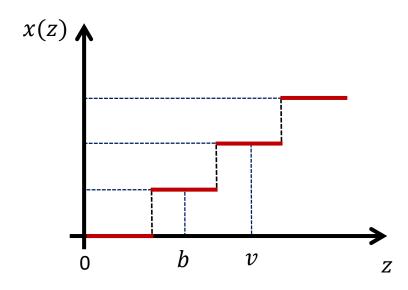
• Example:

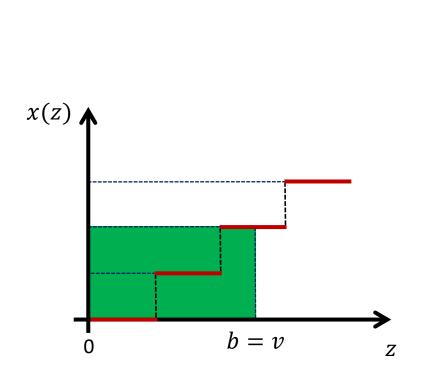


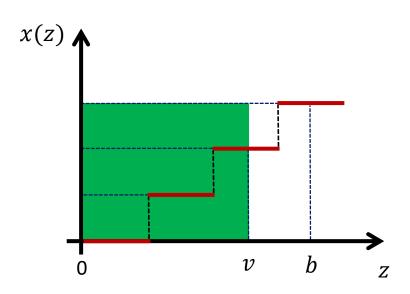
$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y) = y_1 \cdot x_1 + y_2 \cdot (x_2 - x_1)$$

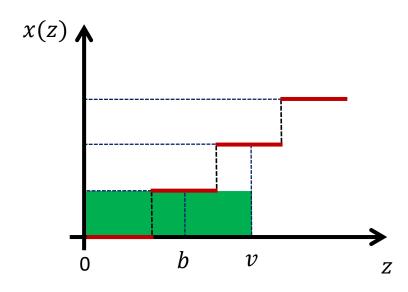


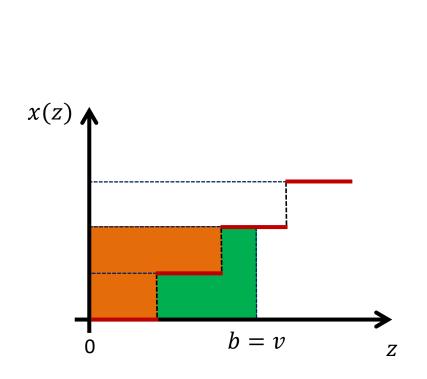


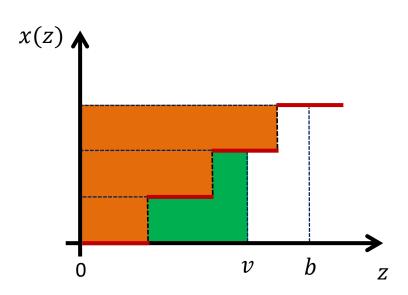


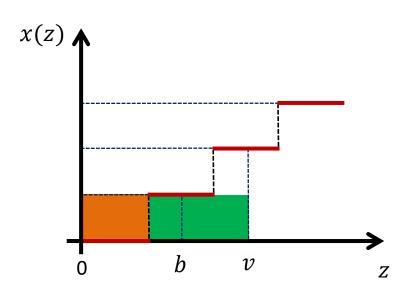












$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

- y enumerates the break points: the bids that are smaller than b
 - In other words, y enumerates the slots from worst to best

$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

- ullet y enumerates the break points: the bids that are smaller than b
 - In other words, y enumerates the slots from worst to best
- jump of x at y: the difference in CTR between two consecutive slots

$$p(b) = \sum_{y \in [0,b]} y \cdot (\text{jump of } x \text{ at } y)$$

- y enumerates the break points: the bids that are smaller than b
 - In other words, y enumerates the slots from worst to best
- jump of x at y: the difference in CTR between two consecutive slots
- The total payment of the *i*-th highest bidder is:

$$p_i(b_i, \boldsymbol{b}_{-i}) = \sum_{j=i}^{\kappa} b_{j+1}(a_j - a_{j+1})$$

• Auctions: allocation rule + payment rule

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction
- An allocation rule is monotone, if larger bids give more stuff

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction
- An allocation rule is monotone, if larger bids give more stuff
- Single-item auctions: first-price is not truthful, second-price is truthful and maximizes the social welfare (sells to the bidder with the highest value)

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction
- An allocation rule is monotone, if larger bids give more stuff
- Single-item auctions: first-price is not truthful, second-price is truthful and maximizes the social welfare (sells to the bidder with the highest value)
- **Sponsored search auctions:** generalized second-price auction is not truthful

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction
- An allocation rule is monotone, if larger bids give more stuff
- Single-item auctions: first-price is not truthful, second-price is truthful and maximizes the social welfare (sells to the bidder with the highest value)
- **Sponsored search auctions:** generalized second-price auction is not truthful
- Myerson's Lemma: a characterization of truthful mechanisms in single-parameter environments

- Auctions: allocation rule + payment rule
- An allocation rule is implementable is there exists a payment rule, so that together they define a truthful auction
- An allocation rule is monotone, if larger bids give more stuff
- Single-item auctions: first-price is not truthful, second-price is truthful and maximizes the social welfare (sells to the bidder with the highest value)
- Sponsored search auctions: generalized second-price auction is not truthful
- Myerson's Lemma: a characterization of truthful mechanisms in single-parameter environments
- Using Myerson's Lemma we can design a truthful sponsored search auction