# Mechanism design: Single-parameter environments 

Alexandros Voudouris
University of Oxford

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- $n$ agents


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- 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 $v_{i}-p$


## Single-item auctions

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- Truthful auctions that maximize the social welfare


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- (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


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Case II: $\boldsymbol{v}_{\boldsymbol{i}} \geq B$

- Maximum possible utility $=v_{i}-B$
- Bidder $i$ wins the item by setting $b_{i}=v_{i}$


## Sponsored search auctions

## Google

```
buy a computer
```



```
About \(4,510,000,000\) results ( 0.72 seconds)
PCSpecialist | Buy your New Computer | PCSpecialist.co.uk [Ad www.pcspecialist.co.uk/
あ丸ᄎᄎᄎ Rating for pcspecialist.co.uk: 4.8-774 reviews
Configure your new custom computer to your exact requirements. Next day PCs also available!
\begin{tabular}{ll} 
All-In-One Computers & Game-Based Computers \\
Choose from Our Range of Intel - & View Our Recommendations for PCs \\
Based AIO PC Systems. & Based on Your Favourite Games.
\end{tabular}
Dell PC Servers | Powered By Intel Xeon | dell.com
Ad www.dell.com/ • 03332580993
Designed To Handle The Most Demanding Technical Computing Workloads. Buy Now! Scalability. Dell Business Advisor - Dell Storage Solutions - Dell Servers - Dell Networking Solutions
Buy a PC with Cyberpower UK | Finance Options
(Ad) www.cyberpowersystem.co.uk) 03333237776
Custom Build Your Ultimate Gaming Desktop Or Pick Up One That We've Pre-Built. Save On High...
Desktop PCs at PC World | Free Delivery On All Orders | PCWorld.co.uk
(Ad) www.pcworld.co.uk/Desktops -
Collect In-Store Available. Fast \& Secure Checkout. Intel® Core \({ }^{\text {TM }}\) Processors Inside.
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- The slots are ranked so that $a_{1} \geq \cdots \geq a_{k}$
- Each bidder $i$ has a private value $v_{i}$ per click
- Bidder $i$ derives utility $a_{j} \cdot v_{i}$ 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


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- Can we extend the ideas we exploited for single-item auctions?


## Generalized second-price auction

- 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 $i \leq k$ (who is assigned at slot $i$ ) pays the next highest bid $b_{i+1}$ per click, and every bidder $i>k$ pays 0


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- The utility of bidder $i$ is $u_{i}(\boldsymbol{b})=v_{i} \cdot x_{i}(\boldsymbol{b})-p_{i}(\boldsymbol{b})$


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- 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 $p_{i}(\boldsymbol{b}) \in\left[0, b_{i} \cdot x_{i}(\boldsymbol{b})\right]$
- $p_{i}(\boldsymbol{b}) \geq 0$ ensures that the seller does not pay the bidders
$-p_{i}(\boldsymbol{b}) \leq b_{i} \cdot x_{i}(\boldsymbol{b})$ ensures non-negative utility for truthful bidders


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Lemma [Myerson, 1981]
(a) An allocation rule $\boldsymbol{x}$ is implementable if and only if it is monotone
(b) For every allocation rule $\boldsymbol{x}$, there exists a unique payment rule $\boldsymbol{p}$ such that $(\boldsymbol{x}, \boldsymbol{p})$ is a truthful auction

## Proof of Myerson's Lemma

- Fix a bidder $i$, and the bids $\boldsymbol{b}_{-i}$ of the other bidders
- Given that these quantities are now fixed, we simplify our notation:
$-x(z)=x_{i}\left(z, \boldsymbol{b}_{-i}\right)$
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- The idea:
- assuming ( $\boldsymbol{x}, \boldsymbol{p}$ ) is a truthful auction, the bidder has no incentive to unilaterally deviate to any other bid
- This will give us a relation between $\boldsymbol{x}$ and $\boldsymbol{p}$, which we can use to derive an explicit formula for $\boldsymbol{p}$ as a function of $\boldsymbol{x}$


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$\Rightarrow(\mathrm{a})$ is now proved


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- Assume $\boldsymbol{x}$ is piecewise constant, like in sponsored search auctions

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


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

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- 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 $[0, b]$

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- 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}\left(b_{i}, \boldsymbol{b}_{-i}\right)=\sum_{j=i}^{k} b_{j+1}\left(a_{j}-a_{j+1}\right)
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- 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

