

COMP/MATH 553 Algorithmic Game Theory Lecture 7: Bulow-Klemperer & VCG Mechanism

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An overview of today's class

Prior-Independent Auctions & Bulow-Klemperer Theorem

General Mechanism Design Problems

Vickrey-Clarke-Groves Mechanism



Another Critique to the Optimal Auction

- □ What if your distributions are *unknown*?
- □ When there are many bidders and enough past data, it is reasonable to assume you know exactly the value distributions.
- But if the market is "thin", you might not be confident or not even know the value distributions.
- □ Can you design an auction that does not use any knowledge about the distributions but performs *almost as well as* if you know *everything* about the distributions.
- Active research agenda, called prior-independent auctions.

Bulow-Klemperer Theorem



Remark:

- Vickrey's auction is prior-independent!
- This means with the same number of bidders, Vickrey Auction achieves at least n-1/n fraction of the optimal revenue. (exercise)
- More competition is better than finding the right auction format.

Proof of Bulow-Klemperer

- Consider another auction *M* with *n*+1 bidders:
 - 1. Run Myerson on the first n bidders.
 - 2. If the item is unallocated, give it to the last bidder for free.
- This is a *DSIC* mechanism. It has the *same* revenue as Myreson's auction with n bidders.
- Notice that it's allocation rule always gives out the item.
- Vickrey Auction also always gives out the item, but always to the bidder who has the highest value (also with the highest virtual value).
- Vickrey Auction has the highest virtual welfare among all DSIC mechanisms that always give out the item!



General Mechanism Design Problem (Multi-Dimensional)

Multi-Dimensional Environment

□ So far, we have focused on single-dimensional environment.

□ In many scenarios, bidders have different value for different items.

- Sotherby's Auction:



□ Multi-Dimensional Environment

- *n* strategic participants/agents,
- a set of possible outcomes Ω .
- agent *i* has a private value $v_i(\omega)$ for each ω in Ω (abstract and could be large).

Examples of Multi-Dimensional Environment

- □ Single-item Auction in the single-dimensional setting:
 - n+1 outcomes in Ω .
 - Bidder *i* only has positive value for the outcome in which he wins, and has value *0* for the rest *n* outcomes
- □ Single-item Auction in the multi-dimensional setting:
 - Imagine you are not selling an item, but auctioning a startup who has a lot of valuable patents.
 - *n* companies are competing for it.
 - Still n+1 outcomes in Ω .
 - But company *i* doesn't win, it might prefer the winner to be someone in a different market other than a direct competitor.
 - So besides the outcome that *i* wins, *i* has different values for the rest *n* outcomes.

How do you optimize Social Welfare (Non-bayesian)?

- □ What do I mean by optimize social welfare (algorithmically)?
 - $\omega^* := \operatorname{argmax}_{\omega} \Sigma_i v_i(\omega)$
- □ How do you design a DSIC mechanism that optimizes social welfare.
 - Take the same two-step approach.
 - Sealed-bid auction. Bidder *i* submits b_i which is indexed by Ω .
 - Allocation rule is clear: assume b_i 's are the true values and choose the outcome that maximizes social welfare.
 - In single-dimensional settings, once the allocation rule is decided, Myerson's lemma tells us the unique payment rule.
 - In multi-dimensional settings, Myerson's lemma doesn't apply ... How can you define monotone allocation rule when bids are multi-dimensional?
 - Similarly, how can we define the payment rule even if we know the allocation rule.



Vickrey-Clarke-Groves (VCG) Mechanism

[The Vickrey-Clarke-Groves (VCG) Mechanism] In every general mechanism design environment, there is a DSIC mechanism that maximizes the social welfare. In particular the allocation rule is

 $X(b) = \operatorname{argmax}_{w} \Sigma_{i} b_{i}(w) \quad (1);$

and the payment rule is $p_i(b) = \max_w \sum_{j \neq i} b_j(w) - \sum_{j \neq i} b_j(w^*)$ (2),

where $w^* = argmax_w \Sigma_i b_i(w)$ is the outcome chosen in (1).

Understand the payment rule

What does the payment rule mean?

- $p_i(b) = max_{\omega} \Sigma_{j\neq i} b_j(\omega) \Sigma_{j\neq i} b_j(\omega^*)$
- $max_{\omega} \sum_{j \neq i} b_j(\omega)$ is the optimal social welfare when *i* is not there.
- ω^* is the optimal social welfare outcome, and $\sum_{j \neq i} b_j(\omega^*)$ is the welfare from all agents except *i*.
- So the difference max_ω Σ_{j≠i} b_j(ω) Σ_{j≠i} b_j(ω*) can be viewed as "the welfare loss inflicted on the other n−1 agents by i's presence". Called "externality" in Economics.
- Example: single-item auction.
 - If *i* is the winner, $\max_{\omega} \sum_{j \neq i} b_j(\omega)$ is the second largest bid.
 - $\Sigma_{j\neq i}b_j(\omega^*)=0.$
 - So exactly second-price.

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where $w^* = argmax_w \Sigma_i b_i(w)$ is the outcome chosen in (1).

Proof: See the board!

Discussion of the VCG mechanism

- DSIC mechanism that *optimizes social welfare* for *any* mechanism design problem !
- □ However, sometimes *impractical*.
- **□** How do you find the allocation that maximizes social welfare. If Ω is really large, what do you do?
 - m items, n bidders, each bidder wants only one item.
 - m items, n bidders, each bidder is single-minded (only like a particular set of items).
 - m items, n bidders, each bidder can take any set of items.

□ Computational intractable.

□ If you use approximation alg., the mechanism is no longer DSIC.