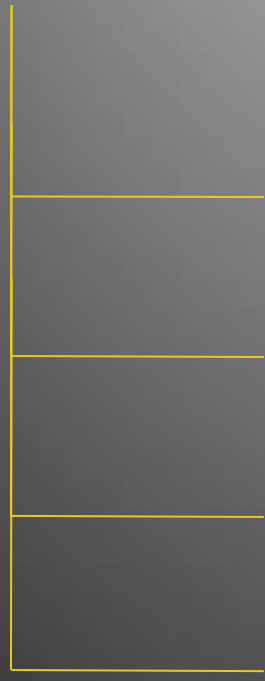


**COMP/MATH 553 Algorithmic
Game Theory
Lecture 4: Myerson's Lemma
(cont'd) and Revenue Optimization**

Sep 15, 2014

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An overview of today's class

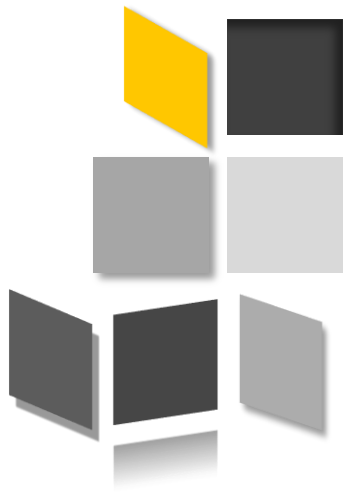
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- Myerson's Lemma (cont'd)*
 - Application of Myerson's Lemma*
 - Revelation Principle*
 - Intro to Revenue Maximization*

Myerson's Lemma



[Myerson '81 ] Fix a single-dimensional environment.

- (a) An allocation rule x is implementable **if and only if** it is **monotone**.
- (b) If x is monotone, then there is a **unique** payment rule such that the sealed-bid mechanism (x, p) is DSIC [assuming the normalization that $b_i = 0$ implies $p_i(b) = 0$].
- (c) The payment rule in (b) is given by an explicit formula.

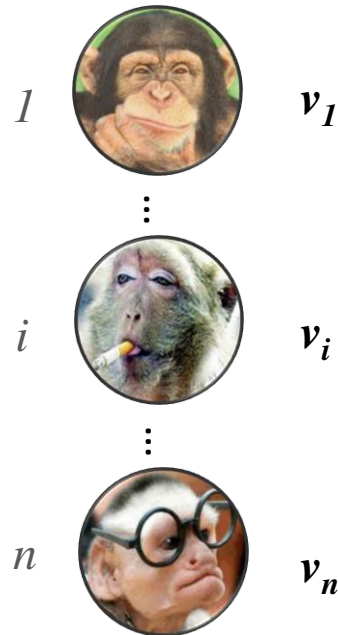


Application of Myerson's Lemma

Single-item Auctions: Set-up



Bidders



Auctioneer



Item



Allocation Rule: give the item to the highest bidder.

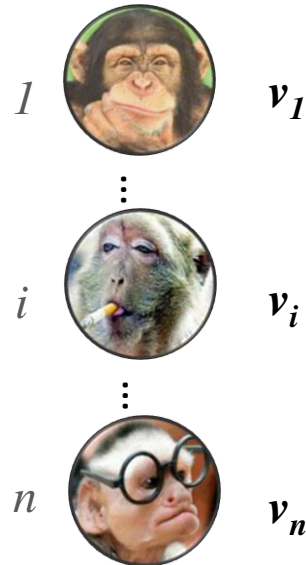


Payment Rule: ?

Sponsored Search Auctions: Set-up



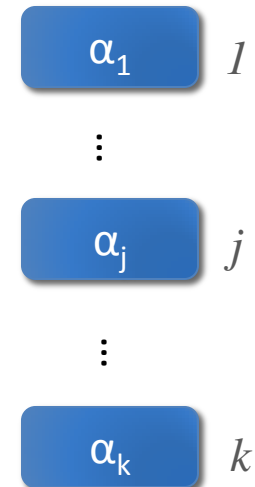
Bidders (advertisers)



Auctioneer/
Google



Slots



- **Allocation Rule:** allocate the slots greedily based on the bidders' bids.



- **Payment Rule:** ?



Revelation Principle

- ❑ It's **easy** for the bidders to play.



- ❑ Designer can predict the outcome with **weak** assumption on bidders' behavior.

Q: Why DSIC?

- ❑ But sometimes first price auctions can be useful in practice.

- ❑ Can **non-DSIC** mechanisms accomplish things that DSIC mechanisms can't?

Two assumptions about DSIC



- ❑ Assumption (1): **Every** participant in the mechanism has a **dominant strategy**, no matter what its private valuation is.

- ❑ Assumption (2): This dominant strategy is **direct revelation**, where the participant truthfully reports all of its private information to the mechanism.

- ❑ There are mechanisms that satisfy (1) but not (2).
 - Run Vickrey on bids $\times 2$...



- Assumption (1): **Every** participant in the mechanism has a **dominant strategy**, no matter what its private valuation is.
 - Can relax (1)? but need stronger assumptions on the bidders' behavior, e.g. Nash eq. or Bayes-Nash eq.
 - Relaxing (1) can give **stronger** results in certain settings.
 - DSIC is **enough** for most of the simple settings in this class.
 - Incomparable: Performance or Robustness?



- Assumption 2: This dominant strategy is **direct revelation**, where the participant truthfully reports all of its private information to the mechanism.
 - Comes for “free”.
 - Proof: Simulation.

Revelation Principle

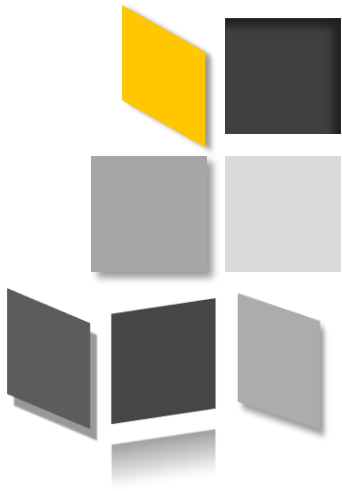


Theorem (Revelation Principle): For every mechanism M in which every participant has a dominant strategy (no matter what its private information), there is an equivalent **direct-revelation DSIC mechanism** M' .

Revelation Principle



- ❑ Same principle can be extended to other solution concept, e.g. Bayes Nash Eq.
- ❑ The requirement of **truthfulness** is **not** what makes mechanism design hard...
- ❑ It's **hard** to find a desired outcome in a certain type of **Equilibrium**.
- ❑ Changing the type of equilibrium leads to different theory of mechanism design.



REVENUE-OPTIMAL AUCTION

Welfare Maximization, Revisited



- ❑ Why did we start with Welfare?
- ❑ Obviously a fundamental objective, and has broad real world applications. (government, highly competitive markets)
- ❑ For welfare, you have DSIC achieving the optimal welfare as if you know the values (single item, sponsored search, and even arbitrary settings (will cover in the future))
- ❑ Not true for many other objectives.

One Bidder + One Item



- ❑ The only DSIC auctions are the “posted prices”.
- ❑ If the seller posts a price of r , then the revenue is either r (if $v \geq r$), or 0 (if $v < r$).
- ❑ If we know v , we will set $r = v$. But v is private...
- ❑ Fundamental issue is that, for revenue, different auctions do better on different inputs.
- ❑ Requires a model to reason about tradeoffs between different inputs.

Bayesian Analysis/Average Case



Classical Model: pose a distribution over the inputs, and compare the expected performance.

- A single-dimensional environment.

- The private valuation v_i of participant i is assumed to be drawn from a distribution F_i with density function f_i with support contained in $[0, v_{max}]$.
 - We assume that the distributions F_1, \dots, F_n are independent (not necessarily identical).
 - In practice, these distributions are typically derived from data, such as bids in past auctions.

- The distributions F_1, \dots, F_n are known in advance to the mechanism designer. The realizations v_1, \dots, v_n of bidders' valuations are private, as usual.

Solution for One Bidder + One Item



- ❑ Expected revenue of a posted price r is $r(1-F(r))$
- ❑ When F is the uniform dist. on $[0,1]$, optimal choice of r is $\frac{1}{2}$ achieving revenue $\frac{1}{4}$.
- ❑ The optimal posted price is also called the *monopoly price*.

Two Bidders + One Item



- ❑ Two bidders' values are drawn i.i.d. from $U[0,1]$.
- ❑ Revenue of Vickrey's Auction is the expectation of the min of the two random variables = $1/3$.
- ❑ What else can you do? Can try reserve price.
- ❑ Vickrey with reserve at $1/2$ gives revenue $5/12 > 1/3$.
- ❑ Can we do better?

Revenue-Optimal Auctions



□ [Myerson '81]

- Single-dimensional settings
- Simple Revenue-Optimal auction