

**COMP/MATH 553 Algorithmic Game Theory Lecture 5: Myerson's Optimal Auction** 

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An overview of today's class

*Expected Revenue* = *Expected Virtual Welfare* 

2 Uniform [0,1] Bidders Example

Optimal Auction

# **Revelation Principle Recap**





# **Bayesian Analysis Model**

#### □ A single-dimensional environment, e.g. single-item

- □ 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  $F_1, \ldots, F_n$  are independent (not necessarily identical).
  - In practice, these distributions are typically derived from data, such as bids in past auctions.
- □ The distributions  $F_1, \ldots, F_n$  are known in advance to the mechanism designer. The realizations  $v_1, \ldots, v_n$  of bidders' valuations are private, as usual.

# **Revenue-Optimal Auctions**





- Single-dimensional settings
- Simple Revenue-Optimal auction

# What do we mean by optimal?

- □ Step 0: What types of mechanism do we need to consider? In other words, optimal amongst what mechanisms?
- □ Consider the set of mechanisms that have a *dominant strategy equilibrium*.
- □ Want to find the one whose revenue at the dominant strategy equilibrium is the *highest*.
- □ A large set of mechanisms. How can we handle it?

Revelation Principle comes to rescue! We only need to consider the directrevelation DSIC mechanisms!



# Expected Revenue = Expected Virtual Welfare

### **Revenue = Virtual Welfare**

[Myerson '81 ] For any single-dimensional environment. Let  $F = F_1 \times F_2 \times ... \times F_n$  be the joint value distribution, and (x,p) be a DSIC mechanism. The expected revenue of this mechanism

### $\mathsf{E}_{\mathsf{v}\sim\mathsf{F}}[\boldsymbol{\Sigma}_i \; \boldsymbol{\mathsf{p}}_i(\mathsf{v})] = \mathsf{E}_{\mathsf{v}\sim\mathsf{F}}[\boldsymbol{\Sigma}_i \; \boldsymbol{\mathsf{x}}_i(\mathsf{v}) \; \boldsymbol{\varphi}_i \; (\mathsf{v}_i)],$

where  $\varphi_i(v_i) := v_i - (1 - F_i(v_i))/f_i(v_i)$  is called bidder i's virtual value ( $f_i$  is the density function for  $F_i$ ).



# Myerson's OPTIMAL AUCTION

□ Two bidders' values are drawn i.i.d. from U[0,1].

 $\Box$  Vickrey with reserve at  $\frac{1}{2}$ 

- If the highest bidder is lower than  $\frac{1}{2}$ , no one wins.
- If the highest bidder is at least ½, he wins the item and pay max {1/2, the other bidder's bid}.

 $\Box$  Revenue 5/12.

□ This is optimal. WHY???

# **Two Bidders + One Item**

□ Virtual value for v:  $\varphi(v) = v - (1 - F(v))/f(v) = v - (1 - v)/1 = 2v - 1$ 

□ Optimize expected revenue = Optimize expected *virtual welfare*!!!

□ Should optimize *virtual welfare* on *every bid profile*.

□ For any bid profile  $(v_1, v_2)$ , what allocation rule optimizes virtual welfare?  $(\phi(v_1), \phi(v_2))=(2v_1-1, 2v_2-1).$ 

- If  $\max\{v_1, v_2\} \ge 1/2$ , give the item to the highest bidder
- Otherwise,  $\varphi(v_1)$ ,  $\varphi(v_2) < 0$ . Should not give it to either of the two.

□ This allocation rule is monotone.

# **Revenue-optimal Single-item Auction**

- □ Find the *monotone* allocation rule that optimizes expected *virtual welfare*.
- □ Forget about *monotonicity* for a while. What allocation rule optimizes expected *virtual welfare*?
- □ Should optimize *virtual welfare* on *every bid profile v*.

 $-\max \Sigma_i x_i(v) \varphi_i(v_i)., \text{ s.t } \Sigma_i x_i(v) \leq 1$ 

□ Call this *Virtual Welfare-Maximizing Rule*.

# **Revenue-optimal Single-item Auction**

#### □ Is the Virtual Welfare-Maximizing Rule *monotone*?

- Depends on the distribution.
- Definition 1 (Regular Distributions): A single-dimensional distribution F is *regular* if the corresponding virtual value function v- (1-F(v))/f(v) is non-decreasing.
- □ **Definition 2 (Monotone Hazard Rate (MHR))**: A single-dimensional distribution F has *Monotone Hazard Rate*, if (1-F(v))/f(v) is non-increasing.

# **Revenue-optimal Single-item Auction**

What distributions are in these classes?

- MHR: uniform, exponential and Gaussian distributions and many more.
- Regular: MHR and Power-law...
- Irregular: Multi-modal or distributions with very heavy tails.

□ When all the F<sub>i</sub>'s are regular, the Virtual Welfare-Maximizing Rule is *monotone*!

# **Two Extensions Myerson did (we won't teach)**

- □ What if the distributions are irregular?
  - Point-wise optimizing virtual welfare is not monotone.
  - Need to find the allocation rule that maximizes expected virtual welfare among all monotone ones. Looks hard...
  - This can be done by "ironing" the virtual value functions to make them monotone, and at the same time preserving the virtual welfare.
  - We restrict ourselves to DSIC mechanisms
    - Myerson's auction is optimal even amongst a much larger set of "Bayesian incentive compatible (BIC)" (essentially the largest set) mechanisms.
    - For example, this means first-price auction (at equilibrium) can't generate more revenue than Myerson's auction.
- □ Won't cover them in class.
  - Section 3.3.5 in "Mechanism Design and Approximation", book draft by Jason Hartline.
  - "Optimal auction design", the original paper by Roger Myerson.