

## CS 396: Online Markets

### Lecture 14: Revenue Maximization and Learning

#### Last Time:

- revenue of auctions (cont).
- virtual values.
- truthfulness and the revelation principle.
- optimization of truthful auctions.

#### Today:

- optimization of truthful auctions (cont).
  - optimal first-price auctions.
  - learning to price.
  - learning to auction.
- 

#### Exercise: Expected Payment

##### Recall:

- allocation rule:  
 $x(v) = \mathbf{Pr}[\text{bidder wins with value } v]$
- can view  $x(\cdot)$  as cumulative distribution function of random price.

##### Setup:

- allocation rule  $x(v) = v$

##### Questions:

- what is expected price offered to the bidder?
  - what is expected payment of bidder with value  $v = 1/2$ ?
- 

## Revenue Maximization

**Recall:** virtual value  $\varphi(v) = v - \frac{1-F(v)}{f(v)}$ .

**Thm:** in multi-bidder mechanisms, expected revenue equals expected virtual welfare.

#### Example 1: symmetric buyers

- two bidders, uniform values
- $\varphi(v) = 2v - 1$
- 1 wins if  $v_1 \geq \max(v_2, \varphi^{-1}(0))$
- optimal auction:  
second price auction with reserve  $\varphi^{-1}(0) = 1/2$

**Cor:** for i.i.d. buyers second-price auction with reserve  $\varphi^{-1}(0)$  is revenue optimal

#### Example 2: asymmetric buyers

- bidder 1:
  - value:  $U[0, 2]$
  - virtual value:  $\varphi_1(v_1) = 2v_1 - 2$
- bidder 2:
  - value:  $U[0, 3]$
  - virtual value:  $\varphi_2(v_2) = 2v_2 - 3$
- bidder 1 wins
  - when  $2v_1 - 2 \geq 0 \Rightarrow v_1 \geq 1$ .
  - when  $2v_1 - 2 \geq 2v_2 - 3 \Rightarrow v_1 \geq v_2 - 1/2$ .

DRAW PICTURE

---

## Revenue Optimal First-price Auction.

“truthful auctions are often impractical”

**Approach:** find first-price auction with same allocation rule as optimal truthful auction.

#### Example:

- two bidders, uniform values

**Q:** what is revenue optimal first-price auction?

**A:** first-price auction with reserve  $1/2$

- winner is bidder with highest value over  $1/2$   
 $\Rightarrow x$  as second-price auction with reserve  $1/2$   
 $\Rightarrow$  optimal expected virtual welfare
-

## Exercise: Selling Introductions

### Setup:

- you are selling introductions
- two bidders, values  $U[0, 1]$
- your mechanism either
  - (a) introduces bidders to each other
  - (b) does not introduce them
- design a truthful mechanism to maximize your revenue.

**Questions:** What is outcome (introduce or not) in the revenue optimal mechanism when

- $v_1 = 0.9$  and  $v_2 = 0.2$ ?
  - $v_1 = 0.8$  and  $v_2 = 0.1$ ?
  - $v_1 = 0.6$  and  $v_2 = 0.6$ ?
- 

## Learning to Price

### Model:

in round  $i$ :

- buyer arrives
- offer price
- learn whether buyer takes-it-or-leaves-it

**Approach:** “learning to price” similar to “learning to bid”.

## Learning to Auction

**Model:** symmetric buyers

- assumption: buyers have i.i.d. values

in round  $i$ :

- bidders arrive
- run truthful single-item auction
- learn values

**Approach:** “learning to reserve price” similar to “learning to price”

**Model:** asymmetric buyers

in round  $i$ :

- bidders arrive
- run truthful single-item auction
- learn values

**Observation:** virtual values only decide order.

### Approach:

- $m$  bidders
- $\ell$  values (discretized)
- virtual welfare optimizing auction is defined by an interleaving of these values and  $\emptyset$   
e.g.,  $a_1, b_1, a_2, a_3, b_2, \emptyset, b_3$
- sell to agent ranked first according to interleaved order (if before  $\emptyset$ )
- at most  $k \leq m^{m\ell}$  interleavings.
  - interleaving determined by:
  - $m\ell$  positions
  - $m$  possible bidders per position
- use online learning with actions = auctions
- recall: regret for Exponential Weights:  $2h\sqrt{\frac{\ln k}{n}}$
- regret for learning auction:  $2h\sqrt{\frac{m\ell}{n} \ln m}$