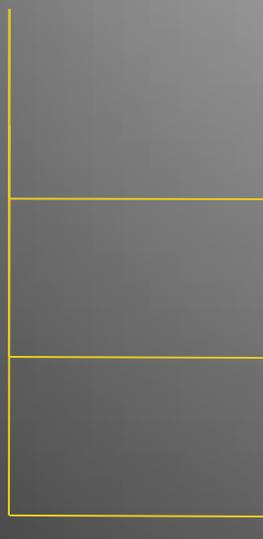


**COMP/MATH 553 Algorithmic
Game Theory
Lecture 9: Revenue Maximization
in Multi-Dimensional Settings**

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An overview of today's class

- 
- Myerson's Auction Recap*
 - Challenge of Multi-Dimensional Settings*
 - Unit-Demand Pricing*

Myerson's Auction Recap



[Myerson '81 ] For any single-dimensional environment.

Let $F = F_1 \times F_2 \times \dots \times F_n$ be the joint value distribution, and (x, p) be a DSIC mechanism. The expected revenue of this mechanism

$$E_{v \sim F}[\sum_i p_i(v)] = E_{v \sim F}[\sum_i x_i(v) \varphi_i(v_i)],$$

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 Auction Recap

- Bidders report their values;
- The reported values are **transformed** into **virtual**-values;
- the virtual-welfare maximizing allocation is chosen.
- Charge the payments according to **Myerson's Lemma**.
- **Transformation** = depends on the distributions; **deterministic** function (the virtual value function);



❖ Myerson's auction looks like the following

Nice Properties of Myerson's Auction



- ❑ **DSIC**, but optimal among all Bayesian Incentive Compatible (BIC) mechanisms!
- ❑ **Deterministic**, but optimal among all possibly randomized mechanisms!
- ❑ **Central open problem** in Mathematical Economics: How can we extend Myerson's result to **Multi-Dimensional Settings**?
- ❑ **Important progress** in the past a few years.
- ❑ See the **Challenges** first!



Challenges in Multi-Dimensional Settings

Example 1:



- ❑ A single buyer, 2 non-identical items
- ❑ Bidder is additive e.g. $v(\{1,2\}) = v_1 + v_2$.
- ❑ Further simplify the setting, assume v_1 and v_2 are drawn **i.i.d.** from distribution $F = U\{1,2\}$ (1 w.p. $\frac{1}{2}$, and 2 w.p. $\frac{1}{2}$).
- ❑ What's the optimal auction here?
- ❑ Natural attempt: How about sell both items using Myerson's auction separately?

Example 1:



- ❑ Selling each item separately with Myerson's auction has expected revenue \$2.
- ❑ Any other mechanism you might want to try?
- ❑ How about bundling the two items and offer it at \$3?
- ❑ What is the expected revenue?
- ❑ Revenue = $3 \times \Pr[v_1 + v_2 \geq 3] = 3 \times \frac{3}{4} = \frac{9}{4} > 2!$
- ❑ *Lesson 1: Bundling Helps!!!*

Example 1:



- ❑ The effect of bundling becomes more **obvious** when the number of items is large.
- ❑ Since they are i.i.d., by the **central limit theorem** (or Chernoff bound) you know the bidder's value for the grand bundle (contains everything) will be a *Gaussian distribution*.
- ❑ The *variance* of this distribution decreases *quickly*.
- ❑ If set the price slightly lower than the expected value, then the bidder will buy the grand bundle w.p. almost 1. Thus, **revenue is almost the expected value!**
- ❑ This is the best you could hope for.

Example 2:



- ❑ Change F to be $U\{0,1,2\}$.

- ❑ Selling the items separately gives $\$4/3$.

- ❑ The best way to sell the Grand bundle is set it at price $\$2$, this again gives $\$4/3$.

- ❑ Any other way to sell the items?

- ❑ Consider the following menu. The bidder picks the best for her.
 - Buy either of the two items for $\$2$
 - Buy both for $\$3$

Example 2:



□ Bidder's choice:

$v_1 \backslash v_2$	0	1	2
0	\$0	\$0	\$2
1	\$0	\$0	\$3
2	\$2	\$3	\$3

□ Expected Revenue = $3 \times 3/9 + 2 \times 2/9 = 13/9 > 4/3!$

Example 3:



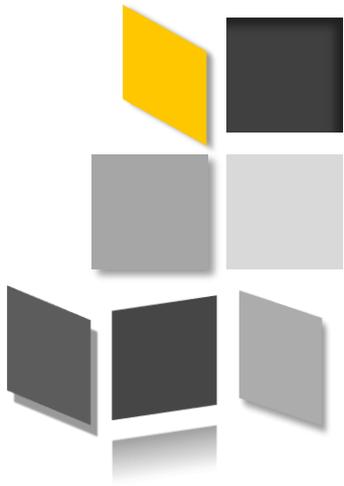
- ❑ Change F_1 to be $U\{1,2\}$, F_2 to be $U\{1,3\}$.

- ❑ Consider the following menu. The bidder picks the best for her.
 - Buy both items with price \$4.
 - *A lottery*: get the first item for sure, and *get the second item with prob. $\frac{1}{2}$* .
pay \$2.50.

- ❑ The expected revenue is \$2.65.

- ❑ Every deterministic auction — where every outcome awards either nothing, the first item, the second item, or both items — has **strictly less expected revenue**.

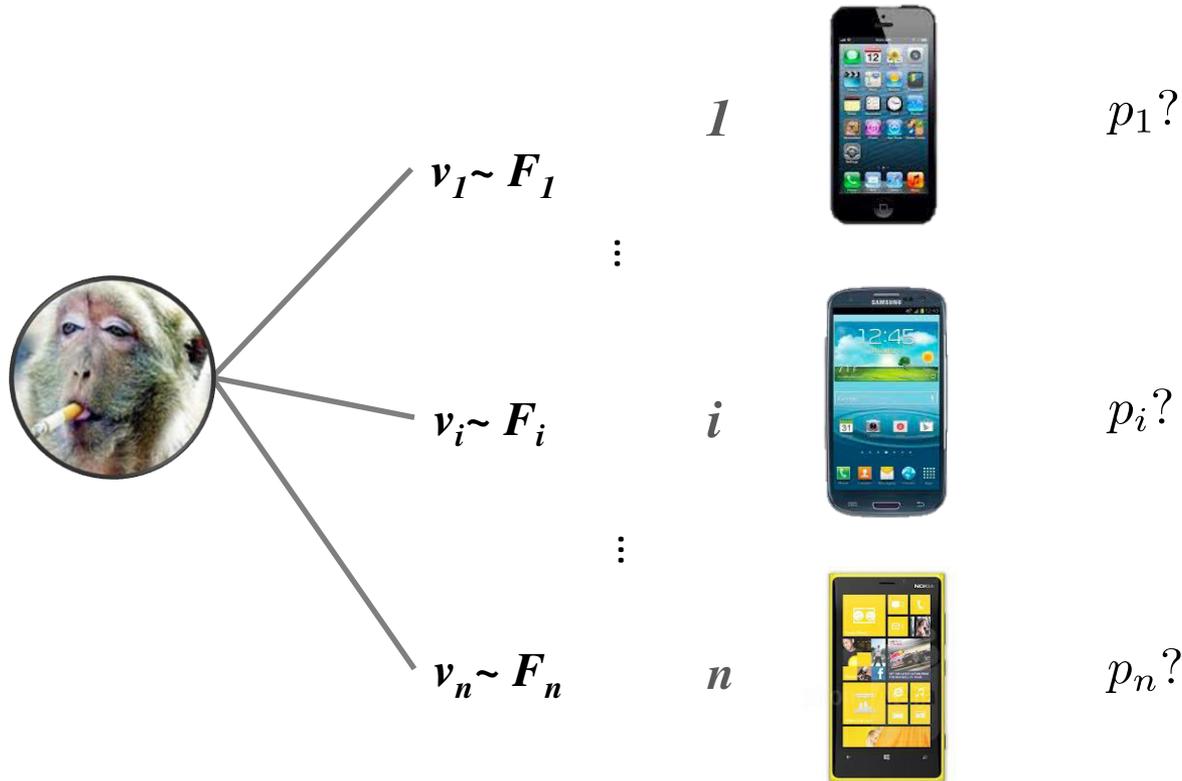
- ❑ *Lesson 2: randomization could help!*



Unit-demand Bidder Pricing Problem

Unit-Demand Bidder Pricing Problem (UPP)

□ A fundamental pricing problem



- Bidder chooses the item that maximizes $v_i - p_i$, if any of them is positive.
- Revenue will be the corresponding p_i .
- Focus on pricing only, not considering randomized ones.
- It's known randomized mechanism can only get a constant factor better than pricing.

Our goal for UPP



- ❑ Goal: design a pricing scheme that achieves a constant fraction of the revenue that is achievable by the optimal pricing scheme.
- ❑ Assumption: F_i 's are regular.

Theorem [CHK '07]: There exists a simple pricing scheme (poly-time computable), that achieves at least $\frac{1}{4}$ of the revenue of the optimal pricing scheme.

Remark: the constant can be improved with a better analysis.

What is the Benchmark???



- When designing simple nearly-optimal auctions. The benchmark is clear.
- Myerson's auction, or the maximum of the virtual welfare.
- In this setting we don't know what the optimal pricing scheme looks like.
- We want to compare to the optimal revenue, but we have no clue what the optimal revenue is?
- Any natural upper bound for the optimal revenue?

Two Scenarios



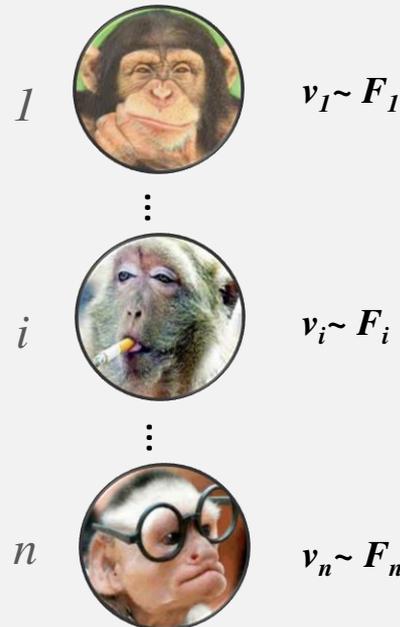
❖ (b) Auction

❑ *n bidders*

❑ *One item*

❑ Bidder I 's value for the item v_i is drawn independently from F_i

Bidders



Item

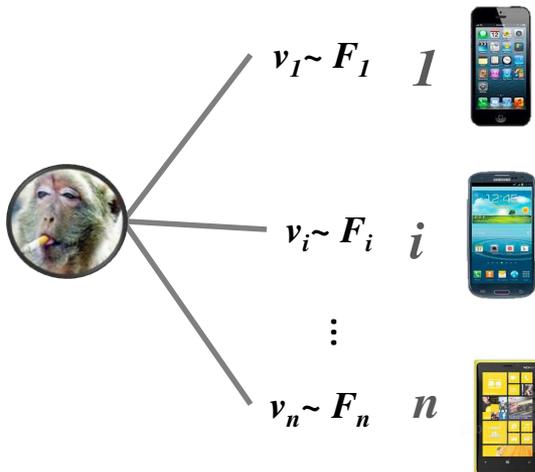


❖ (a) UPP

❑ *One unit-demand bidder*

❑ *n items*

❑ Bidder's value for the i -th item v_i is drawn independently from F_i





Lemma 1: The optimal revenue achievable in scenario (a) is always less than the optimal revenue achievable in scenario (b).

- Proof: See the board.
- Remark: This gives a natural benchmark for the revenue in (a).

An even simpler benchmark



- In a single-item auction, the optimal expected revenue

$$\mathbf{E}_{\mathbf{v} \sim \mathbf{F}} [\max \sum_i x_i(\mathbf{v}) \varphi_i(\mathbf{v}_i)] = \mathbf{E}_{\mathbf{v} \sim \mathbf{F}} [\max_i \varphi_i(\mathbf{v}_i)^+] \text{ (the expected prize of the prophet)}$$

- Remember the following mechanism **RM** we learned in Lecture 6.

1. Choose t such that $\Pr[\max_i \varphi_i(\mathbf{v}_i)^+ \geq t] = 1/2$.
2. Set a reserve price $r_i = \varphi_i^{-1}(t)$ for each bidder i with the t defined above.
3. Give the item to the highest bidder that meets her reserve price (if any).
4. Charge the payments according to Myerson's Lemma.

- By prophet inequality:

$$\mathbf{ARev}(\mathbf{RM}) = \mathbf{E}_{\mathbf{v} \sim \mathbf{F}} [\sum_i x_i(\mathbf{v}) \varphi_i(\mathbf{v}_i)] \geq 1/2 \mathbf{E}_{\mathbf{v} \sim \mathbf{F}} [\max_i \varphi_i(\mathbf{v}_i)^+] = 1/2 \mathbf{ARev}(\mathbf{Myerson})$$

- Let's use the revenue of RM as the benchmark.

Inherent loss of this approach



- ❑ Relaxing the benchmark to be Myerson's revenue in (b)
- ❑ This step might lose a constant factor already.
- ❑ To get real optimal, a different approach is needed.