

Simple vs Optimal Auctions.

$v_1, \dots, v_n \stackrel{iid}{\sim} D$ optimal auction = Vickrey.

$$\phi(v) = v - \frac{1 - F(v)}{f(v)}, \quad F \text{ cdf of } D, \quad f = F' \text{ pdf of } D.$$

1) $v_1, v_2, \dots, v_n \stackrel{iid}{\sim} D$, but D is unknown.

Theorem: [Bulow-Klemperer '96]

Let D be a regular distribution (ϕ is strictly increasing), and n be a positive number. Then:

$$\mathbb{E}[\text{Rev}(VA_{n+1})] \geq \mathbb{E}[\text{Rev}(OPT_{D,n})]$$

$v_1, \dots, v_{n+1} \sim D$ $v_1, \dots, v_n \sim D$

Proof: Let A be a third auction on $n+1$ bidders.

(A):

1) Simulate $OPT_{D,n}$ on the first n bidders.

2) If step 1 does not allocate the item, give it to bidder $n+1$ for free.

• $\mathbb{E}[\text{Rev}(A_{n+1})] = \mathbb{E}[\text{Rev}(OPT_{D,n})]$

• A always allocates the item.

J_1 always allocates the item.

(WTS that VA is optimal among all auctions
which always allocate the item)

By Myerson's Opt auction, $(OPT_{D,n+1})$ allocates the item to
the bidder w/ highest virtual welfare.

VA_{n+1} allocates the item to the bidder w/ highest welfare.

When D is regular $\Rightarrow \phi$ is increasing.

\hookrightarrow Bidder w/ highest welfare = Bidder w/ highest virtual welfare.

$$\begin{aligned} \mathbb{E}[\text{Rev}(VA_{n+1})] &= \mathbb{E}[\text{Rev}(OPT_{D,n+1} \text{ has to allocate})] \geq \\ &\geq \mathbb{E}[\text{Rev}(A_{n+1})] = \mathbb{E}[\text{Rev}(OPT_{D,n})] \end{aligned}$$

Corollary: VA_n w/ no reserve price is a $(1 - 1/n)$ -approximation
to OPT_n .

2) $V_1, \dots, V_n \stackrel{\text{ind.}}{\sim} D_1, \dots, D_n$ and D_1, \dots, D_n known distributions.

• Greco's Auction

→ Set a price p for item.

→ Buyer's arrive*, one after the other and if $V_i \geq p$, buyer i buys the item.

* in adversarial order.
(worst-case)

Properties

- Trivially truthful.
- We don't have to collect bids.
- After price is set, auction runs itself.



Theorem: [Hajagahayi et al '07, Chawla et al '10]

The grocer's auction is equivalent to the prophet inequality.

Prophet Inequality

$$X_1, X_2, \dots, X_n \stackrel{\text{ind.}}{\sim} D_1, D_2, \dots, D_n.$$

At step i , we see realization Z_i of X_i . We decide immediately and irrevocably, whether to accept or reject Z_i .

→ If we accept Z_i , game ends.

→ Else, we proceed to Z_{i+1} .

Objective: Select $\max Z_i$ (maximize the value of the selected item).

Objective: Select $\max_i Z_i$. Compete against prophet. who always gets $\max_i Z_i$.

Prophet's value: $\mathbb{E}[\max_i X_i]$

Theorem: [Krengel, Sucheston & Garling '76, Samuel-Cahn '84]

For the prophet inequality setting, let τ be the median value of the distribution of $\max_i X_i$, i.e. $\Pr[\max_i X_i \geq \tau] = 1/2$. Then, the algorithm ALG that selects the first realization where $X_i \geq \tau$, obtains value:

$$\mathbb{E}[\text{ALG}] \geq \frac{1}{2} \mathbb{E}[\max_i X_i].$$

$$(X_i - \tau)^+ = \max\{X_i - \tau, 0\}$$

Proof | Let $X^* = \max_i X_i$.

$$\mathbb{E}[\text{ALG}] = \tau \cdot \Pr[X^* \geq \tau] + \sum_{i=1}^n \Pr[\text{we "reach" } X_i] \cdot \mathbb{E}[(X_i - \tau)^+]$$

$$= \frac{1}{2} \tau + \sum_{i=1}^n \Pr[\max_{j=1}^{i-1} X_j < \tau] \cdot \mathbb{E}[(X_i - \tau)^+]$$

$$\geq \frac{1}{2} \tau + \sum_{i=1}^n \Pr[\max_{j=1}^n X_j < \tau] \cdot \mathbb{E}[(X_i - \tau)^+]$$

$$= \frac{1}{2} \tau + \sum_{i=1}^n \Pr[X^* < \tau] \cdot \mathbb{E}[(X_i - \tau)^+]$$

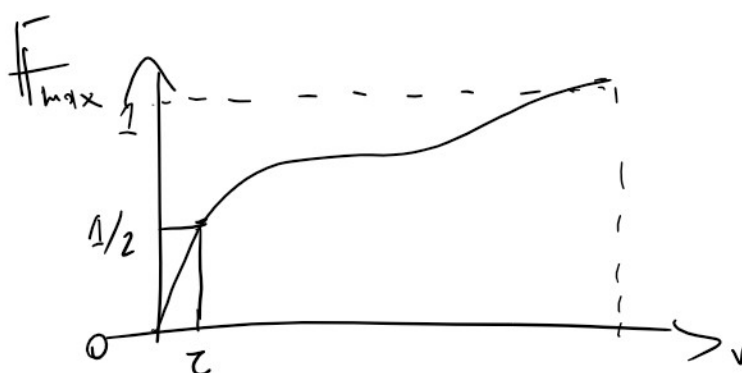
$$= \frac{1}{2} \tau + \sum_{i=1}^n \frac{1}{2} \cdot \mathbb{E}[(X_i - \tau)^+]$$

$$\frac{1}{2} \tau + \frac{1}{2} \mathbb{E}[X^* - \tau]$$

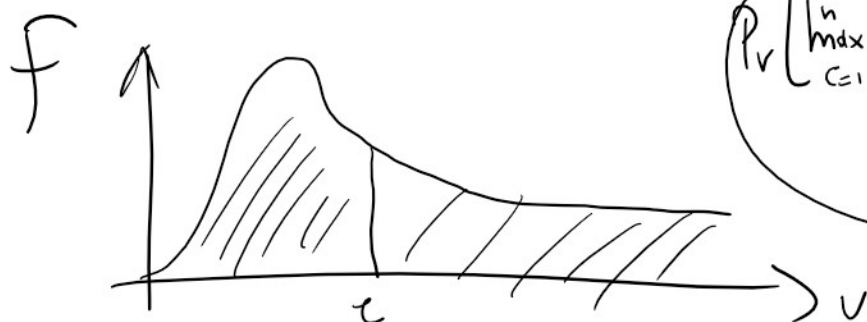
$$\geq \frac{1}{2} \tau + \frac{1}{2} \mathbb{E}[(X^* - \tau)^+]$$

$$\geq \frac{1}{2} \tau + \frac{1}{2} \mathbb{E}[X^* - \tau]$$

$$= \frac{1}{2} (\tau + \mathbb{E}[X^* - \tau]) = \frac{1}{2} \mathbb{E}[X^*] = \frac{1}{2} \mathbb{E}\left[\max_{i=1}^n X_i\right].$$



$$F(v) = P_v[X \leq v].$$



$$P_v\left[\max_{i=1}^n X_i \leq v\right] = \prod_{i=1}^n P_v[X_i \leq v] \downarrow F_i(v)$$

$$= F_1(v) \cdot F_2(v) \cdot \dots \cdot F_n(v).$$

[Robinstein et al '20]

[Kleinberg-Weinberg '12]: Just set $\tau = \frac{1}{2} \mathbb{E}\left[\max_{i=1}^n X_i\right]$