Simple Near-Optimal Auctions

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Outline

- Background & Introduction
- 2 The Prophet Inequality
- Simple Single-Item Auctions
- Prior-Independent Mechanisms



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What we have learned

- For a single-parameter environment where agents' valuations are drawn independently from regular distributions, the corresponding virtual welfare maximizer maximizes the expected revenue over all DSIC mechanisms.
 - The allocation rule:

$$\mathbf{x}(\mathbf{v}) = \arg\max_{\mathbf{X}} \sum_{i=1}^{n} \varphi_i(\mathbf{v}_i) \mathbf{x}_{\mathbf{v}}(\mathbf{v})$$

for each valuation profile \mathbf{v} , where

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 - ullet a second-price auction augmented with the reserve price $arphi^{-1}(0)$



Optimal Auctions Can Be Complex

 What if bidders' valuations are drawn from different regular distributions?



Optimal Auctions Can Be Complex

- What if bidders' valuations are drawn from different regular distributions?
- We would like to know if there is any simple and practical single-item auction formats that are at least approximately optimal.



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- In stage i, we are offered a nonnegative prize π_i , drawn from a distribution G_i .
- We know the *independent* distributions G_1, \ldots, G_n in advance.
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 - discard the prize, and then proceed to the next stage.
- What's the risk and difficulty?



The Prophet Inequality

 It offers a simple strategy that performs almost as well as (approximately) a fully clairvoyant prophet.

Theorem (Prophet Inequality)

For every sequence G_1, \ldots, G_n of *n* independent distributions,

- There is a strategy that guarantees expected reward $\geq \frac{1}{2} \mathbf{E}_{\pi \sim \mathbf{G}}[\max_i \pi_i]$.
- There is such a threshold strategy, which accepts prize i if and only if $\pi_i \geq t$.
- $z^+ := \max\{z, 0\}.$
- $[n] := \{1, 2, \dots, n\}.$



Proof of Prophet Inequality (1/3)

- Compare the expected payoff of a threshold strategy with that of a prophet, through lower and upper bounds.
- q(t): the prob. that the threshold strategy accepts no prize at all.
- First, we want to have a lower bound on

 $\psi := \mathbf{E}_{\pi \sim \mathbf{G}}[\text{payoff of the } t\text{-threshold strategy}].$



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- The payoff: 0 with prob. q(t) and $\geq t$ with prob. 1 q(t).
- Credit the baseline t with "extra credit" of $\pi_i t$ if only one prize satisfies $\pi_i \geq t$.
- Only credit the baseline t for two or more prizes $\geq t$ (: LB).



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Proof of Prophet Inequality (2/3)

$$\psi \geq (1 - q(t))t + \sum_{i=1}^{n} \mathbf{E}_{\pi}[\pi_{i} - t \mid \pi_{i} \geq t, \pi_{j} < t \,\forall j \neq i] \cdot \Pr[\pi_{i} \geq t] \cdot \Pr[\pi_{j} < t \,\forall j \neq i]$$

$$= (1 - q(t))t + \sum_{i=1}^{n} \underbrace{\mathbf{E}_{\pi}[\pi_{i} - t \mid \pi_{i} \geq t] \cdot \Pr[\pi_{i} \geq t]}_{= \mathbf{E}_{\pi}[(\pi_{i} - t)^{+}]} \cdot \underbrace{\Pr[\pi_{j} < t \,\forall j \neq i]}_{\geq q(t) = \Pr[\pi_{j} < t \,\forall j]}$$

$$\geq (1 - q(t))t + q(t) \cdot \sum_{i=1}^{n} \mathbf{E}_{\pi}[(\pi_{i} - t)^{+}]$$

Proof of Prophet Inequality (3/3)

Moreover, as to the upper bound on the prophet's expected payoff:

$$\psi^* := \mathbf{E}_{\pi} \begin{bmatrix} \max_{i \in [n]} \pi_i \end{bmatrix} = \mathbf{E}_{\pi} \begin{bmatrix} t + \max_{i \in [n]} (\pi_i - t) \end{bmatrix}$$

$$\leq t + \mathbf{E}_{\pi} \begin{bmatrix} \max_{i \in [n]} (\pi_i - t)^+ \end{bmatrix}$$

$$\leq t + \sum_{i=1}^n \mathbf{E}_{\pi} [(\pi_i - t)^+].$$

• Set t such that $q(t) = \frac{1}{2}$ we can complete the proof.

$$\frac{1}{2} + \frac{1}{2} \cdot \sum_{i=1}^{n} \mathsf{E}_{\pi}[(\pi_{i} - t)^{+}] \le \psi \le \psi^{*} \le t + \sum_{i=1}^{n} \mathsf{E}_{\pi}[(\pi_{i} - t)^{+}]$$



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- Set t such that $q(t) = \frac{1}{2}$ we can complete the proof.
 - LB:= $\frac{t}{2} + \frac{1}{2} \cdot \sum_{i=1}^{n} \mathbf{E}_{\pi}[(\pi_{i} t)^{+}] \le \psi \le \psi^{*} \le t + \sum_{i=1}^{n} \mathbf{E}_{\pi}[(\pi_{i} t)^{+}] = 2 \cdot 1 \text{ B}.$

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Back to the motivating example

- Single-item auction with n bidders.
- Bidders' valuations are drawn independently from regular distributions F_1, \ldots, F_n that might not be identical.
- Using the prophet inequality:
 - Define the *i*th prize as $\varphi_i(v_i)^+$ of bidder *i*.
 - G_i : the corresponding distribution induced by F_i (independent).
- We have (by Theorem 5.2; with maximizer $\mathbf{x} = (x_i)_{i \in [n]}$)

$$\mathbf{E}_{\boldsymbol{v} \sim \boldsymbol{F}} \left[\sum_{i=1}^{n} \varphi_i(v_i) x_i(\boldsymbol{v}) \right] = \mathbf{E}_{\boldsymbol{v} \sim \boldsymbol{F}} \left[\max_{i \in [n]} \varphi_i(v_i)^+ \right].$$

• The expected revenue of the optimal auction.



The allocation rule

Consider any allocation rule having the following form:

Virtual Threshold Allocation Rule

- Choose t such that $\Pr[\max_i \varphi_i(v_i)^+ \geq t] = \frac{1}{2}$.
- Give the item to a bidder i with $\varphi_i(v_i) \ge t$, if any, breaking ties among multiple candidate winners arbitrarily.
- We immediately have the following corollary:



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Corollary (Virtual Threshold Rules are Near-Optimal)

If x is a virtual threshold allocation rule, then

$$\mathbf{E}_{\boldsymbol{v}}\left[\sum_{i=1}^n \varphi_i(v_i)^+ x_i(\boldsymbol{v})\right] \geq \frac{1}{2} \mathbf{E}_{\boldsymbol{v}}\left[\max_{i \in [n]} \varphi_i(v_i^+)\right].$$

Simple Near-Optimal Auctions

¹What if no such t exists? An exercise!

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- So far, the valuation distributions are assumed to be known to the mechanism designer in advance.
- What if the mechanism designer does NOT know about the valuation distributions in advance?



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- Next, we consider that
 - Bidders' valuations are still drawn from such valuation distributions;
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A different critique so far

- So far, the valuation distributions are assumed to be known to the mechanism designer in advance.
- What if the mechanism designer does NOT know about the valuation distributions in advance?
- Next, we consider that
 - Bidders' valuations are still drawn from such valuation distributions;
 - Yet, these distributions are still unknown to the mechanism designer.
 - * We use the distributions in the analysis, but NOT in the design of mechanisms.
- Goal: design a good prior-independent mechanism.
 - Such a mechanism makes NO reference to a valuation distribution.

A Beautiful Result from Auction Theory

• The expected revenue of an optimal single-item auction is at most that of a second-price auction (with no reserved price) with one extra bidder.

Theorem [Bulow-Klemperer Theorem (1989)]

We have

- F: a regular distribution;
- n: a positive integer.
- p: the payment rule of the second-price auction with n+1 bidders.
- p^* : the payment rule of the optimal auction for F with n bidders.

Then,

$$\mathbf{E}_{\mathbf{v} \sim F^{n+1}} \left[\sum_{i=1}^{n+1} p_i(\mathbf{v}) \right] \geq \mathbf{E}_{\mathbf{v} \sim F^n} \left[\sum_{i=1}^n p_i^*(\mathbf{v}) \right]$$

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Interpretation of the Bulow-Klemperer Theorem

- Extra competition is more important than getting the auction format just right.
- It is better to invest your resources to recruit more serious participants than sharpen your knowledge of their preferences.



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- ① Simulate an optimal n-bidder auction for F on the first n bidders.
- ② If the item was not awarded in the first step, then give the item to the (n+1)th bidder for free.



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- ② If the item was not awarded in the first step, then give the item to the (n+1)th bidder for free.
- The expected revenue of A equals that of an optimal auction with n bidders.
 - The right-hand side of the inequality.



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Goal: The second-price auction maximizes the expected revenue over all DSIC auctions that always allocate the item.



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- \bullet All bidders share the same nondecreasing virtual valuation function $\varphi.$
 - A bidder with highest valuation also has the highest virtual valuation.
- Hence, the second-price auction maximizes expected revenue subject to always awarding the item.

Discussions

