

Algorithmic Mechanism Design

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- single-parameter environment
- Allocation and Payment
- implementable and monotone
- Myerson's Lemma
- Applications: second-price auctions and sponsored search auctions

“Ideal” ^{理想} mechanisms for general (single-parameter) environments?

NP-hard!

Answer: Social Welfare **approximation**

↑
近似

Definition (Knapsack Auction)

Each bidder i has a publicly known size w_i and a private value v_i . The seller has a capacity W . The feasible set X is the 0-1 vectors (x_1, \dots, x_n) such that $\sum_{i=1}^n w_i x_i \leq W$, where $x_i = 1$ means that i is a winner.

- Whenever there is a shared resource with limited capacity, you have a knapsack problem.
- ADs during the Olympic.
- Cloud storage.

Are Knapsack auctions the single-parameter environments? YES!

Next we first assume that the bids are truthful, and then decide the payment.

We define the allocation rule by

$$\mathbf{x}(\mathbf{b}) = \operatorname{argmax}_X \sum_{i=1}^n b_i x_i.$$

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收益 $v_i = b_i$

单调

One can prove that this allocation rule is monotone. By Myerson's Lemma, we can provide the payment rule to make it DSIC. (Why?) But what is the breakpoint? holding other bids \mathbf{b}_{-i} fixed.

The knapsack problem is NP-hard, that is, there is no polynomial-time algorithm for the allocation rule, assuming that $P \neq NP$.¹ What should we do? SW vs. efficiency?

- Relax the efficiency? 有效外生

- Relax the welfare-optimal?

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¹FPTAS (fully polynomial time approximation scheme), i.e., $1 - \epsilon$ -approx in time $\operatorname{poly}(n, 1/\epsilon)$ for any given $\epsilon > 0$.

近似算法

We will show that $x(b)$ is monotone, recall that

$$x(b) = \operatorname{argmax}_{x \in X} \sum_{i=1}^n b_i x_i.$$

Proof.

As the proof of Myerson's Lemma, fix $i \in [n]$, other bids b_{-i} and $y > z > 0$. Let $x^y = x(b_{-i}, y)$ and $x^z = x(b_{-i}, z)$, we show that $x_i^y \geq x_i^z$. By the defs of x (SW maximization), we have

只有第 i 位改成了 y

$$\begin{aligned} x^y \cdot (b_{-i}, y) &\geq x^z \cdot (b_{-i}, y); \\ x^z \cdot (b_{-i}, z) &\geq x^y \cdot (b_{-i}, z). \end{aligned}$$

Simplifying the above, we have

$$(y - z)(x_i^y - x_i^z) \geq 0$$



- One of the first and most well-studied branches of AGT.
- Relax the welfare-optimal condition as little as possible (approximation).
- For single-parameter environments, we only need to design a polynomial-time and monotone allocation rule, due to Myerson's Lemma.
① 多 ② 少 ③ 有解
- Approximation algorithm to NP-hard problems.²
- While the search space of DSIC is smaller than that of approximation algorithms.
- One question: Does *approximate* welfare-optimal also yields a monotone allocation rule?³

²See <https://www.ics.uci.edu/~vazirani/book.pdf> and <https://www.designofapproxalgs.com/book.pdf>.

³Actually, the FPTAS algorithm also yields a monotone allocation... but too hard for our lecture.

A Greedy Knapsack Heuristic

1. Sort and re-index the bidders such that

$$\frac{b_1}{w_1} \geq \frac{b_2}{w_2} \geq \dots \geq \frac{b_n}{w_n}.$$

2. Pick winners in this order until one doesn't fit (say i^* is the max index s.t. $\sum_{j=1}^i w_j \leq W$), and then halt.
3. Return either the solution from the previous step ($x_i(\mathbf{b}) = 1$ if $i \leq i^*$) or the highest bidder ($x_i(\mathbf{b}) = 1$ if $i = \operatorname{argmax}_{j \in [n]} b_j$), whichever has larger social welfare.

We need the second choice...

$b_1 = 2, b_2 = W$ and $w_1 = 1, w_2 = W$ for sufficient large $W > 0$.

Theorem

$\geq \frac{1}{2} \text{opt}$ $\frac{1}{2}$ 近似

This allocation rule is a $1/2$ -approx.

Proof.

每个人 b_i 竞拍后
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Suppose that $\sum_{i=1}^{i^*} b_i < \text{opt}/2$. However, we have $\sum_{i=1}^{i^*+1} b_i \geq \text{opt}$.

Hence, we have $\max_{i \in [n]} b_i \geq b_{i^*+1} \geq \text{opt}/2$.

$$\sum_{i=1}^{i^*} w_i \leq w$$

□

Theorem

$$\sum_{i=1}^{i^*+1} b_i = \text{sum} \quad \sum_{i=1}^{i^*+1} w_i > w$$

Our allocation rule is also monotone.

- If $x_i(\mathbf{b}_{-i}, z) = 0$, we have $x_i(\mathbf{b}_{-i}, y) \geq 0$ where $y > z \geq 0$.
- Otherwise, i should be either the highest bidder or $i \leq i^*$. EASY!

The Revelation Principle

揭示性原理

Until now, we only focus on the DSIC mechanisms.

The reasons for a DSIC mechanism:

- Easy for an agent to operate (bid)
- Predict the outcome, assuming agents are rational ...

How about non-DSIC? like first-price auctions?

The DSIC Condition

- (1) For every valuation ^{估价图后} profile, the mechanism has a *dominant-strategy equilibrium* — an outcome that results from every participant playing a dominant strategy.
- (2) In this dominant-strategy equilibrium, every participant truthfully reports her private information to the mechanism.
(direct revelation) ^{直接披露}

如果不是DSIC

满足(1)不满足(2)

There are mechanisms that satisfy (1) but not (2). Suppose that running the Vickrey auction on the bids ~~25~~.2v

The theorem below state that (1) is more important.

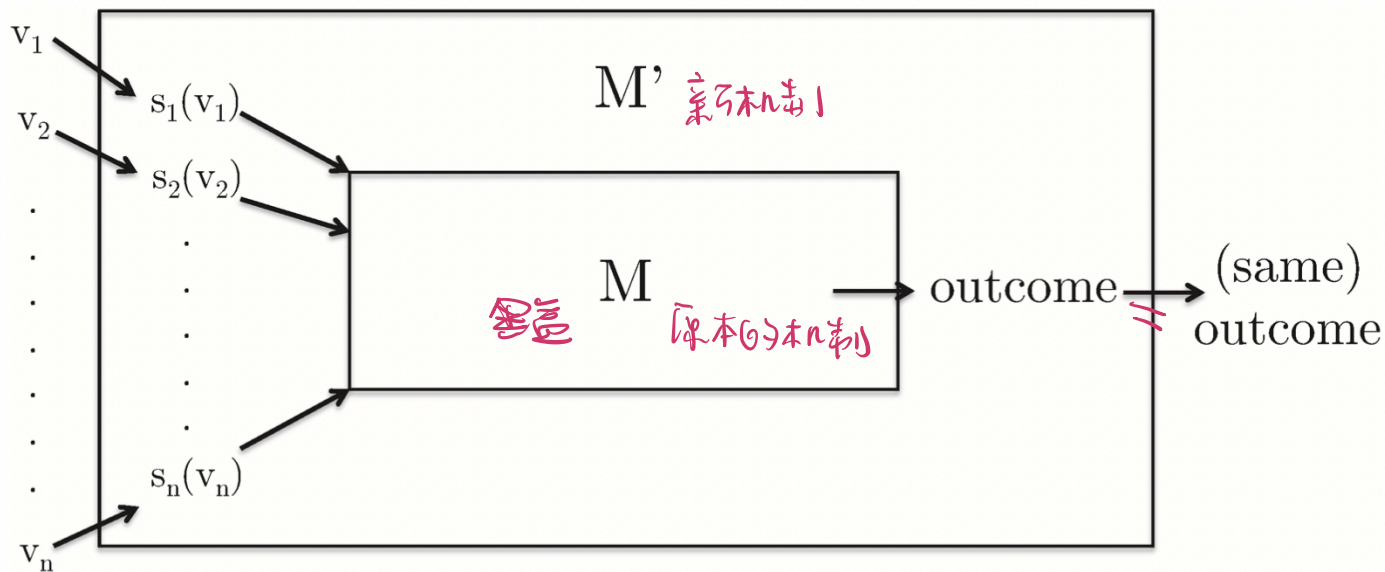
Theorem (Revelation Principle for DSIC Mechanisms)

For every mechanism M in which every participant always has a dominant strategy, there is an equivalent direct-revelation DSIC mechanism M' .

Proof.

We know that i with the private value v_i has a dominant strategy $s_i(v_i)$ in M . We construct a new mechanism M' with the same outcome with M given the bids.

M' accepts bids b_1, \dots, b_n from the agents, and submits $s_i(b_i)$ to M .
 M' is direct-revelation and DSIC. □



背包拍卖的社会福利最大化是一个NP难问题

- Knapsack auctions, SW optimal in this auction is NP-hard
- ✓ *useful* • State-of-the-art approximation algorithms for the welfare maximization problem may or may not induce monotone allocation rules. *不能直接引用
需要论证*
- The revelation principle

揭示性原理

激励机制 \rightarrow DSIC

激励是更好的

每个人报真实价格

Q&A?