

# Algorithmic Game Theory

## Truthful Mechanisms for Welfare Maximization

Vangelis Markakis  
[markakis@gmail.com](mailto:markakis@gmail.com)

# **Designing welfare maximizing truthful auctions for single parameter environments**

# Single parameter auctions

- For the single-item case, we saw that the Vickrey auction is ideal
- We would like to achieve the same properties for any other type of auction
  - truthfulness and individual rationality [incentive guarantees]
  - welfare maximization [economic performance guarantees]
  - implementation in polynomial time [computational performance guarantees]
- Can we achieve all 3 properties for any single-parameter environment?

# Knapsack auctions

- We will see an illustration for knapsack auctions
- $k$  identical items for sale
- Each bidder  $i$  has a **publicly known** demand for  $w_i$  items
  - Inelastic demand
  - The mechanism should either give  $w_i$  items to the bidder or should not give him anything
- Each bidder  $i$  submits a bid  $b_i$  for his value per unit
- Real value per unit =  $v_i$
- Assume the **demands**  $(w_1, w_2, \dots, w_n)$  are **known** to the mechanism
  - Say bidders have no incentive to lie about them
- Only **private information** to bidder  $i$  is  $v_i$

# Knapsack auctions

## Alternative view of knapsack auctions

- The auctioneer has a resource of total capacity  $k$  (a knapsack)
- Each bidder requires **size  $w_i$** , if he is served
- Each bidder has a **value  $v_i w_i$** , if he is served
- The auctioneer needs to select a **subset of bidders to serve** so as not to exceed the capacity  $k$

## Feasible allocations:

- $(x_1, x_2, \dots, x_n)$  with  $x_i \in \{0, 1\}$ , and  $\sum_i w_i x_i \leq k$
- Just like the feasible solutions of a knapsack problem

# Knapsack auctions

## Example

- Resource = the half-time break in the Champions League final
- Capacity  $k$  = total length of the break
- Each bidder corresponds to a company who wants to be advertised during the break
- The size  $w_i$  is the duration of the ad of bidder  $i$
- The auctioneer needs to select a subset of bidders as winners and present their ads without exceeding the time capacity  $k$

# Knapsack auctions

- Let  $\mathbf{b} = (b_1, b_2, \dots, b_n)$  be the bidding vector
- Need to decide the allocation and payment rule
- For the allocation rule:
  - Think of maximizing the social welfare
  - Then we have precisely the 0-1 Knapsack problem!

$$\max \sum_i b_i x_i$$

s.t.

$$\sum_i w_i x_i \leq k$$

$$x_i \in \{0, 1\}, \text{ for } i = 1, \dots, n$$

# Knapsack auctions

**Claim:** The allocation rule that maximizes the social welfare is monotone

- Consider a winner and see what can happen if he increases his bid

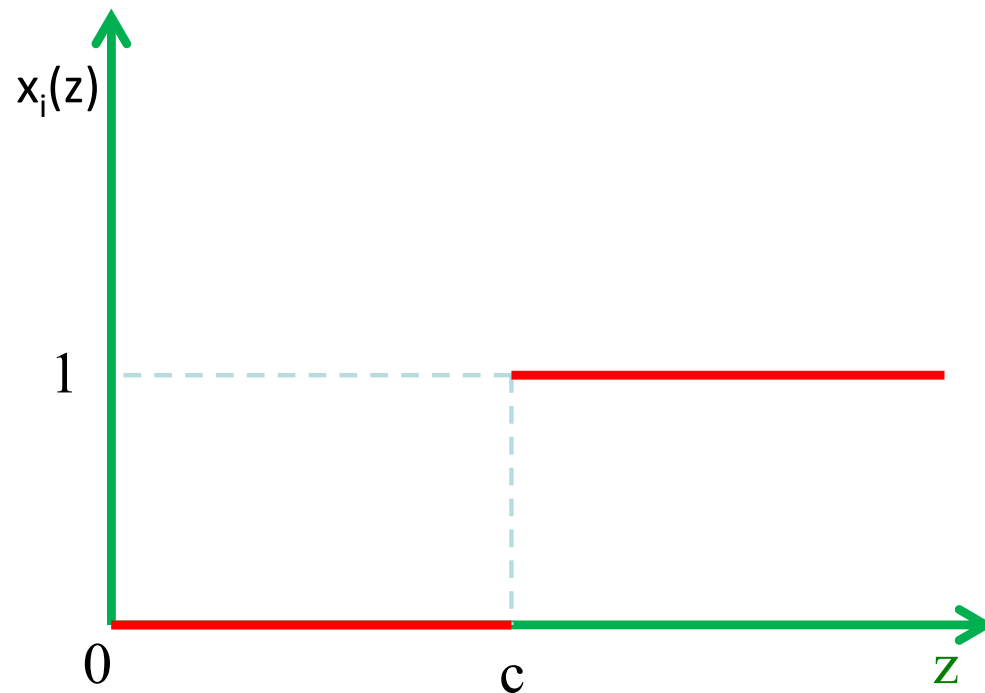
Hence, we can apply Myerson's lemma

How many jumps can we have for the allocation of a single player?

- At most one, a player can jump from being a loser ( $x_i = 0$ ) to being a winner ( $x_i = 1$ )

# Myerson's lemma and knapsack auctions

- The jump for a winner  $i$  happens at  $i$ 's *critical bid*: the minimum he could bid and still be a winner, also known as *threshold bid*
- Generalization of the payment in Vickrey auction



Final mechanism:

- Solve the knapsack problem and find an optimal solution
- Give to each winner  $i$ , the requested number of items  $w_i$
- Charge the winners their **critical bid**

# Myerson's lemma and knapsack auctions

Does this mechanism achieve the desirable properties we wanted?

- truthfulness [YES]
- welfare maximization [YES]
- implementation in polynomial time [?]
- Knapsack is an NP-complete problem
- The properties can be enforced only for special cases where Knapsack is easy
  - If highest bid or highest demand is polynomial in  $n$  (by dynamic programming)
  - If weights form a super-increasing sequence

# Algorithmic Mechanism Design

- The requirement for low complexity usually comes in conflict with the other criteria
- Goal of algorithmic mechanism design: explore the trade-offs between the 3 main properties (or any other properties that we may require in a given setting)
  - Truthfulness
  - welfare maximization
  - implementation in polynomial time
- **Approach:** relax one of the criteria and see if we can achieve the others
- For Knapsack and in general whenever welfare maximization is NP-complete: resort to approximation algorithms

# Knapsack auctions

Goal for Knapsack:

- Find an approximation algorithm for the social welfare
- Prove that it is **monotone**

**Recall:**

Definition: An algorithm  $A$ , for a maximization problem, achieves an approximation factor of  $\gamma$  ( $\gamma \leq 1$ ), if for every instance  $I$  of the problem, the solution returned by  $A$  satisfies:

$$\text{SOL}(I) \geq \gamma \text{ OPT}(I)$$

Where  $\text{OPT}(I)$  is the value of the optimal solution for instance  $I$

# Knapsack auctions

- There are several heuristics and approximation algorithms for Knapsack, but not all of them are monotone
- A greedy  $\frac{1}{2}$ -approximation:
  - For each bidder  $i$ , we care to evaluate the quantity  $b_i/w_i$
  - Intuitively, we prefer bidders with small size/demand and large value

- **Step 1:** Sort and re-index the bidders so that

$$b_1/w_1 \geq b_2/w_2 \geq \dots \geq b_n/w_n$$

- **Step 2:** Pick bidders in that order until the first time that adding someone exceeds the knapsack capacity
- **Step 3:** Return either the previous solution, or just the highest bidder if he achieves higher social welfare on his own

# Knapsack auctions

- Why do we need the last step?
- Maybe there is a bidder with a very high value, but with a large demand as well
- The algorithm may not select this bidder in the first steps
- Step 3 ensures we do not miss out such highly-valued bidders
- **Claim:** This algorithm is monotone
- **Theorem:** Using Myerson's lemma, we can have a truthful polynomial time mechanism, that produces at least 50% of the optimal social welfare

# Knapsack auctions

Going further

- Knapsack also admits an FPTAS (Fully Polynomial Time Approximation Scheme)
  - We can have a  $(1 - \varepsilon)$ -approximation for any constant  $\varepsilon > 0$   
[Ibarra, Kim '75]
  - But this is not a monotone algorithm
- [Briest, Krysta, Voecking '05]: A truthful FPTAS for Knapsack
- **Conclusion:** For a knapsack auction and any  $\varepsilon > 0$ , we have a truthful mechanism that produces at least  $(1 - \varepsilon)$ -fraction of the optimal social welfare and runs in time polynomial in  $n$  and  $1/\varepsilon$

# General Approach

Suppose we have a single-parameter auction where the social welfare maximization problem is NP-hard

- Check if any of the known approximation algorithms for the problem is monotone (usually not)
- If not, then try to tweak it so as to make it monotone (sometimes feasible)
- Or design a new approximation algorithm that is monotone (hopefully without worsening the approximation guarantee)

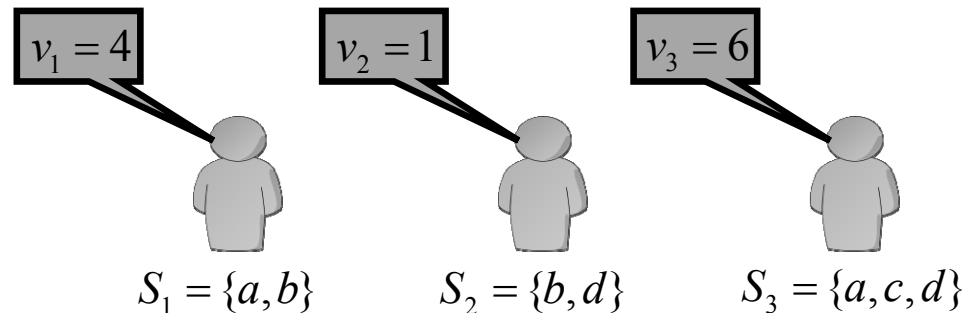
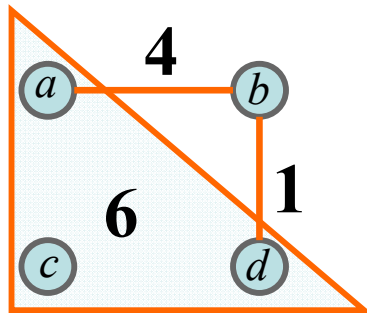
# Single-minded bidders

A single-parameter auction with non-identical items

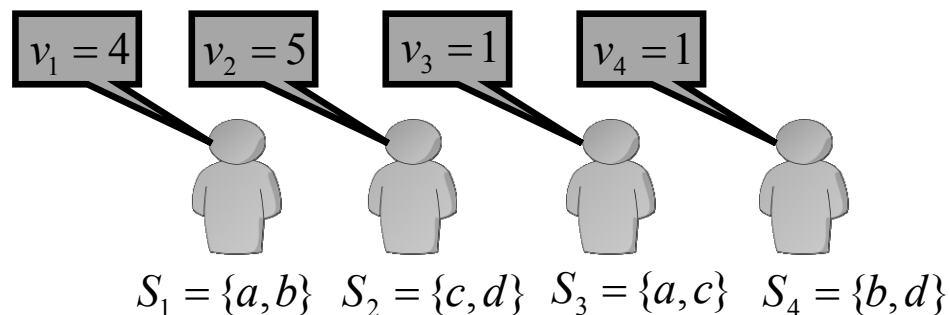
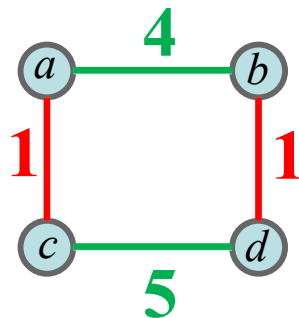
- The auctioneer has a set  $M$  of items for sale
- Each bidder  $i$  is interested in acquiring a **specific subset** of items,  $S_i \subseteq M$  **(known to the mechanism)**
  - If the bidder does not obtain  $S_i$  (or a superset of it), his value is 0
- Each bidder submits a bid  $b_i$  for his value if he obtains the set
- Motivated by certain spectrum auctions
- Feasible allocations: the auctioneer needs to select winners who do not have overlapping sets

# Single-minded bidders

## Examples



- In the example above, the auctioneer can accept only 1 bidder as a winner
- In the example below, the auctioneer can accept up to 2 bidders as winners



# Single-minded bidders

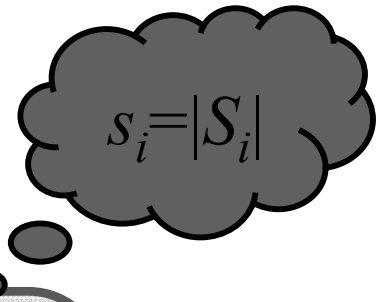
Social welfare maximization:

- Given the bids of the players, select a set of bidders with non-overlapping subsets, so as to maximize the sum of their bids
- It contains the **SET PACKING** problem, hence **NP-hard**
- Actually it gets even worse w.r.t. approximation

**Theorem [Sandholm '99]:** Under certain complexity theory assumptions, we cannot have an algorithm with approximation factor **better than  $1/\sqrt{m}$**

Q: Can we have a  $1/\sqrt{m}$ -approximation?

# Single-minded bidders



[Lehmann, O' Callaghan, Shoham '01]:

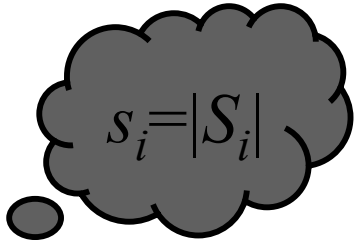
- Order the bidders in decreasing order of  $b_i/\sqrt{s_i}$
- Accept each bidder in this order unless overlapping with previously accepted bidders
- Payment  $i$ : **largest bid  $b_j$**  for set  $S_j$  with **nonempty intersection** with  $S_i$ .

• This algorithm achieves

- $1/\sqrt{m}$ -approximation, where  $m = |M|$
- $1/d$ -approximation, where  $d = \max_i s_i$
- Monotonicity and truthfulness.

**Final conclusion:** truthful polynomial time mechanism with the best possible approximation to the social welfare

# Single-minded bidders


$$s_i = |S_i|$$

- Order the bidders in decreasing order of  $b_i/\sqrt{s_i}$
- Accept each bidder in this order unless overlapping with previously accepted bidders

- A algorithm's solution (set of indices accepted by Greedy)
- O optimal solution (set of indices accepted by OPT)

Wlog. assume that  $O \cap A = \emptyset$ .

Partition  $O$  into  $O_i$ ,  $i \in A$ , s.t.  $j \in O_i$  if  $j \in O$  and  $S_i \cap S_j \neq \emptyset$ .

$$\sum_{j \in O_i} v_j \leq \frac{v_i}{\sqrt{s_i}} \sum_{j \in O_i} \sqrt{s_j}$$

Greedy property

$$\leq \frac{v_i}{\sqrt{s_i}} \sqrt{\sum_{j \in O_i} s_j} \sqrt{|O_i|}$$

Cauchy-Schwarz ineq.

$$\leq \frac{v_i}{\sqrt{s_i}} \sqrt{m} \sqrt{s_i}$$

$$|O_i| \leq s_i \text{ and } \sum_{j \in O_i} s_j \leq m$$

$$\leq v_i \sqrt{m}$$