

## 7 (Backtrack) Auction mechanism design

### 7.2 Myerson's lemma

**Theorem 15.** (Myerson's Lemma) For a single parameter environment, an allocation rule  $x : B \rightarrow X$  is implementable if and only if  $x$  is monotone. Moreover, given a monotone allocation rule  $x : B \rightarrow X$ , there exists a unique payment rule  $p : B \rightarrow \mathbb{R}^N$  such that  $(x, p)$  is DSIC, and  $p_i(b) = 0$  whenever  $b_i = 0$ .

For the converse, we will prove the case where the allocation rule  $x$  is piecewise constant.

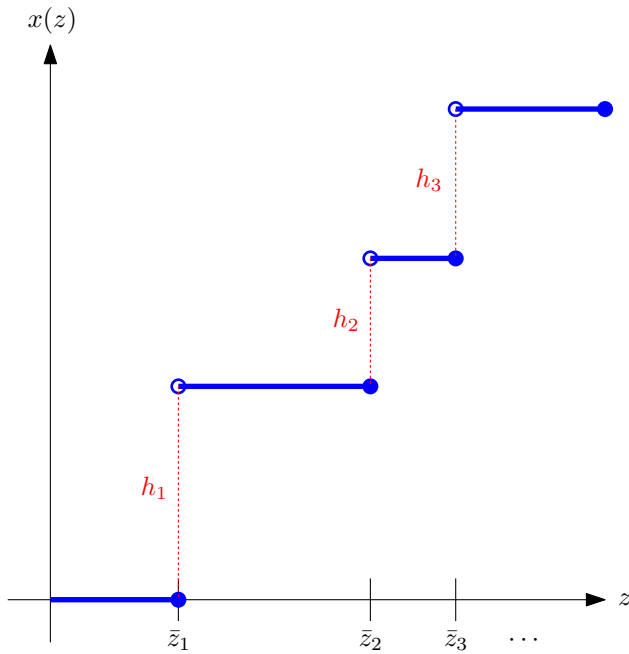
**Approach.** Assume  $x$  is monotone and  $(x, p)$  is DSIC. Derive  $p$ .

Then prove that using this particular  $p$ ,  $(x, p)$  is indeed DSIC.

**Monotone piecewise constant  $x$ .** Jump points  $\bar{z}_1 < \bar{z}_2 < \dots < \bar{z}_q$ .

$x$  is constant in the intervals  $[0, \bar{z}_1], (\bar{z}_1, \bar{z}_2], \dots, (\bar{z}_{q-1}, \bar{z}_q], (\bar{z}_q, \infty)$ .

Define  $h_j$  to be the jump at  $\bar{z}_j$ .



**Lemma 15.1.** If  $x$  is monotone and piecewise constant, and  $(x, p)$  is DSIC with  $p_i(b) = 0$  whenever  $b_i = 0$ , then  $p$  is unique.

*Proof.* We first show that  $p$  is also piecewise constant.

Assume  $y > z$ . Suppose  $z \in (\bar{z}_j, \bar{z}_{j+1})$  for some  $j$ .

We assume DSIC, so the payment sandwich applies.

$$z[x(y) - x(z)] \leq p(y) - p(z) \leq y[x(y) - x(z)].$$

Divide every term by  $y - z$  to get

$$z \cdot \frac{x(y) - x(z)}{y - z} \leq \frac{p(y) - p(z)}{y - z} \leq y \cdot \frac{x(y) - x(z)}{y - z}.$$

Take the limit as  $y \rightarrow z^+$  to get

$$\lim_{y \rightarrow z^+} \frac{x(y) - x(z)}{y - z} = x'(z) = 0 \quad \text{since } x \text{ is constant at } z.$$

Both ends of the payment sandwich approach 0. By Squeeze Theorem,

$$\lim_{y \rightarrow z^+} \frac{p(y) - p(z)}{y - z} = p'(z) = 0.$$

So  $p$  is piecewise constant with jumps at  $\bar{z}_1, \dots, \bar{z}_q$ . What is this jump?

Let  $z = \bar{z}_j$  for some  $j$ . From the payment sandwich, take the limit as  $y \rightarrow z^+$  to get

$$\lim_{y \rightarrow z^+} y(x(y) - x(z)) = \lim_{y \rightarrow z^+} z(x(y) - x(z)) = z \cdot h_j.$$

By Squeeze Theorem,

$$\lim_{y \rightarrow z^+} p(y) - p(z) = z \cdot h_j = \bar{z}_j h_j.$$

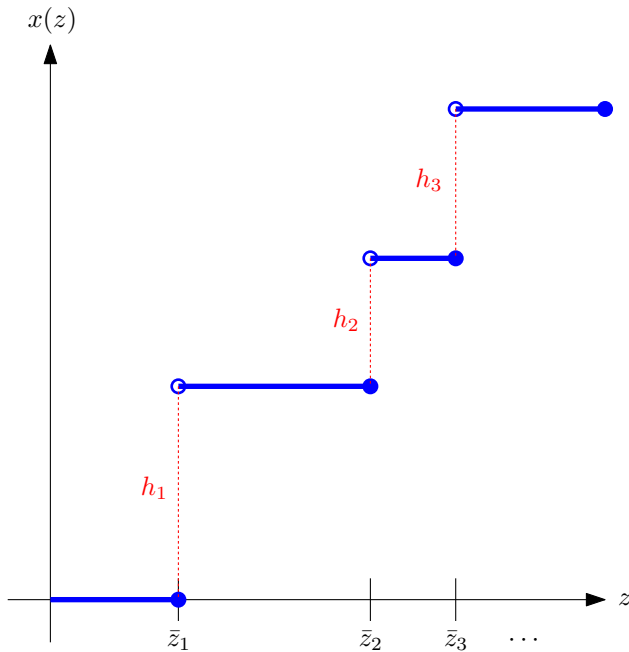
So the payment jump at  $\bar{z}_j$  is  $\bar{z}_j h_j$ .

By assumption,  $p(0) = 0$ , so this uniquely defines  $p$ : If  $j$  is the largest index where  $z \geq \bar{z}_j$ , then

$$p(z) = p_i(z, b_{-i}) = \sum_{k=1}^j \bar{z}_k h_k.$$

□

**Illustration.**



**Graphical interpretation of  $p$ .** The payment is the area to the left of  $z$  in the graph of  $x$ . This can be generalized to any monotone allocation rule.

Lemma 15.1 only says that if  $(x, p)$  is DSIC, then there is only one possible choice of  $p$ . We still need to prove that  $(x, p)$  is DSIC.

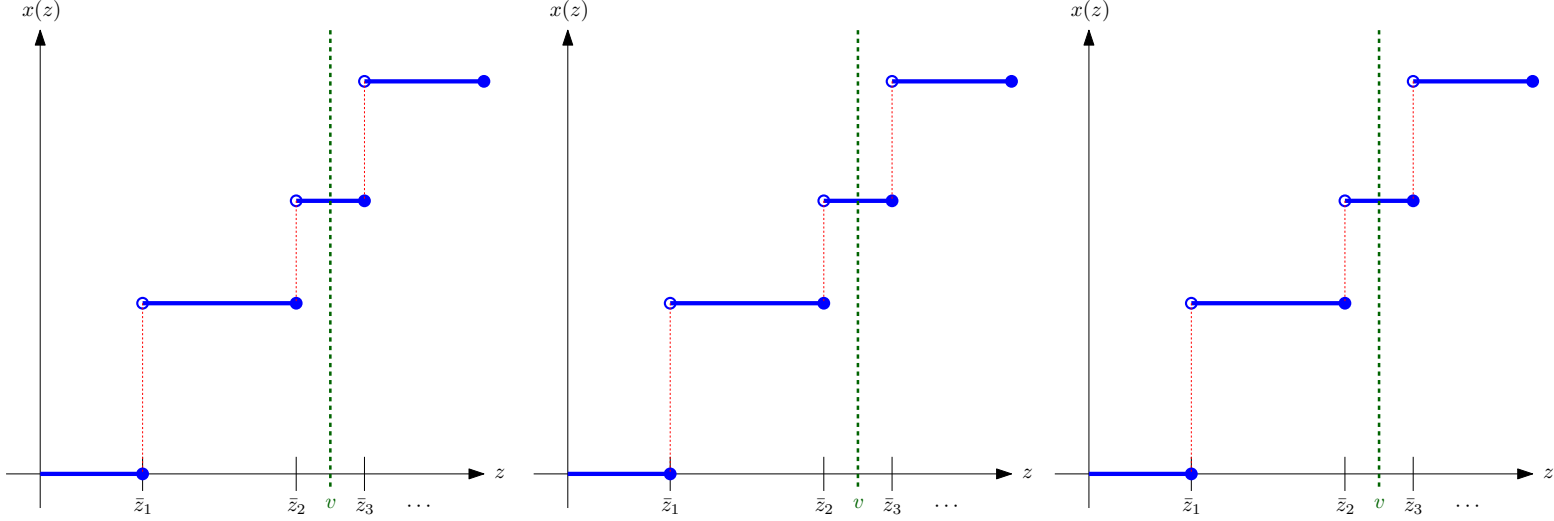
**Lemma 15.2.** The payment rule  $p$  from Lemma 15.1 is DSIC.

*Proof by picture.*

(1) Bidding their valuation.

(2) Underbid.

(3) Overbid.

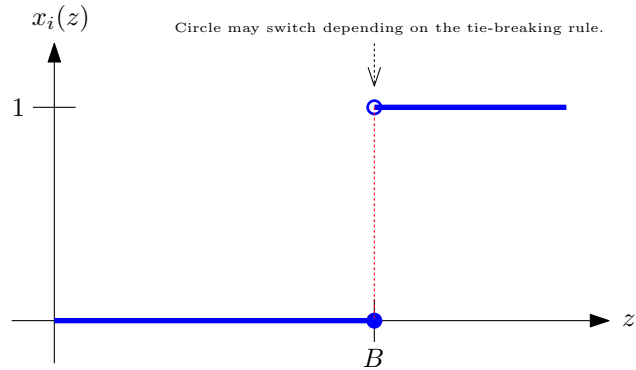


Utilities for (2) and (3) are at most the utility of (1). Utility of (1) is non-negative. Hence DSIC.  $\square$

### 7.3 Applying Myerson's Lemma

**Single item auction.** Let  $x$  be the allocation rule that gives the item to the highest bidder. This is monotone.

Consider a player  $i$  and bids  $b_{-i} \in B_{-i}$ . The allocation for  $i$  has a jump point, at  $B = \max_{j \neq i} b_j$ .

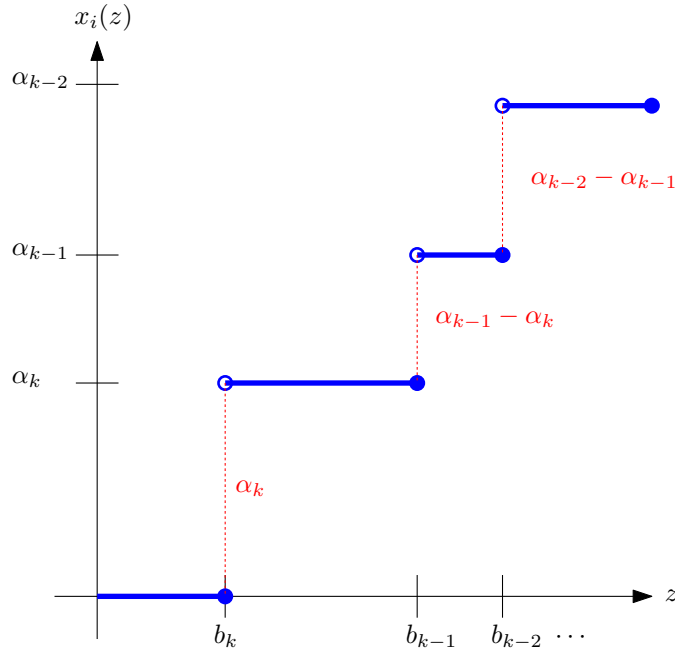


When  $b_i > B$ , player  $i$  wins. Payment is  $B \cdot 1 = B$ . This is the second-price payment rule, it is unique for  $x$ .

**Sponsored search auction.**  $k$  slots with CTR  $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_k$ . Let  $x$  be the allocation rule that assigns  $\alpha_j$  to the  $j$ -th highest bidder. This is monotone.

Consider a player  $i$  and bids  $b_{-i} \in B_{-i}$ . Assume  $b_1 \geq b_2 \geq \dots \geq b_k$  in  $b_{-i}$ .

There are jump points at  $b_k, b_{k-1}, \dots$ . Player  $i$  gets  $\alpha_k$  if  $b_k < z \leq b_{k-1}$ ,  $\alpha_{k-1}$  if  $b_{k-1} < z \leq b_{k-2}$ ,  $\dots$ . Each jump is  $\alpha_{j-1} - \alpha_j$ .



The payment is

$$p_i(z) = \sum_{\{j : b_j < z\}} b_j(\alpha_j - \alpha_{j+1}).$$

## 7.4 Auctions that are almost ideal

Recall: An ideal auction is (1) DSIC; (2) welfare-maximizing; and (3) efficient.

**Example.** We are managing advertising slots for a TV station. We have  $T$  seconds to fill. There are potential advertisers  $N$ . Advertiser  $i$  has an ad of length  $t_i \leq T$  and gets a value of  $v_i$ . Run an auction to determine which ads to air.

**One instance.**

$$T = 10 \quad \boxed{\phantom{0000000000}}$$

$$\boxed{t_1 = 2}$$

$$v_1 = 20$$

$$\boxed{t_2 = 7}$$

$$v_2 = 56$$

$$\boxed{t_3 = 3}$$

$$v_3 = 21$$

$$\boxed{t_4 = 5}$$

$$v_4 = 25$$

Feasible allocations:  $X = \{x \in \{0, 1\}^N : \sum_{i \in N} t_i x_i \leq T\}$ .

Social welfare:  $\sum_{i \in N} v_i x_i$

To maximize social welfare, we need to find  $\max\{v^T x : x \in X\}$ .

**Recall.** We first assume players bid truthfully and find an allocation that is welfare maximizing. Myerson's Lemma gives a payment rule that is DSIC.

**Problem.** Maximizing social welfare is equivalent to the knapsack problem. This is NP-hard.

We cannot fulfill (2) and (3) simultaneously.

**Solution.** Relax (2) or (3). We cannot give up (3) in practice, so we need to give up (2).

Instead of finding a welfare-maximizing allocation, we find a close approximation that is fast to run.

## 7.5 Approximating maximum welfare

**Starting idea.** We want to fill each unit of space with as much value as possible. Calculate value density of the items, and sort them in non-increasing order  $\frac{v_1}{t_1} \geq \frac{v_2}{t_2} \geq \dots \geq \frac{v_n}{t_n}$ .

**Example.**  $T = 10$

$i$	1	2	3	4
$t_i$	2	7	3	5
$v_i$	20	56	21	25
$v_i/t_i$	10	8	7	5

**First try.** Pick items 1, 2, ... until we cannot put the next one in the knapsack.

In the example above, pick 1 and 2.  $t_1 + t_2 = 9 \leq 10$ , no room for 3. Total value 76.

This can be very bad. Say  $v_1 = 1, t_1 = 1$  and  $v_2 = T - 1, t_2 = T$ . 1 has higher density than 2. If we pick 1, then we do not have room for 2. We get a value of 1. Obvious optimal solution is to pick 2 instead.

**Better idea.** Suppose we stop at item  $i$ . Check the value of item  $i + 1$ . If  $v_{i+1} > v_1 + \dots + v_i$ , then  $i + 1$  is the only winner.

**Approximation algorithm.** Assume  $N = \{1, \dots, n\}$  with  $\frac{v_1}{t_1} \geq \frac{v_2}{t_2} \geq \dots \geq \frac{v_n}{t_n}$ .

- Find  $i$  such that it is the largest index with  $t_1 + \dots + t_i \leq T$ .
- If  $v_1 + \dots + v_i \geq v_{i+1}$ , then  $\{1, \dots, i\}$  are the winners. Otherwise,  $\{i + 1\}$  is the only winner.

Suppose  $APX$  is the value of the winners produced by the approximation algorithm, and  $OPT$  is the actual optimal value. How good is  $APX$ ?

**Theorem 15.3.**  $APX \geq \frac{1}{2}OPT$ .

*Proof.*  $OPT$  is the optimal solution of the integer program (IP):  $\max\{v^T x : t^T x \leq T, x_i \in \{0, 1\} \forall i\}$ .

Its LP relaxation is (LP):  $\max\{v^T x : t^T x \leq T, 0 \leq x_i \leq 1 \forall i\}$ . (Solutions can be fractional.)

Let  $i$  be the largest index with  $t_1 + \dots + t_i \leq T$ . Consider the solution  $x$  where

$$x_1 = \dots = x_i = 1, \quad x_{i+1} = \frac{T - t_1 - \dots - t_i}{t_{i+1}}, \quad x_{i+2} = \dots = x_n = 0.$$

(We are filling the remaining space of the knapsack with a fraction of item  $i + 1$ .)

As we have completely filled in the knapsack with the highest-density items,  $x$  is optimal. (Alternative: Prove this using duality.)

Suppose optimal value of (LP) is  $v^*$  (achieved by  $x$ ). Since (LP) is a relaxation of (IP),  $OPT \leq v^*$ . We want to prove that  $APX \geq v^*/2$ .

Let  $S = \{1, \dots, i\}$ . We see that  $v^* \leq v_1 + \dots + v_i + v_{i+1} = v(S) + v_{i+1}$ . Two cases:

- If  $v(S) \geq v_{i+1}$ , then the algorithm picks  $S$ , so  $APX = v(S)$ . Then

$$v^* \leq v(S) + v_{i+1} \leq v(S) + v(S) = 2 \cdot APX.$$

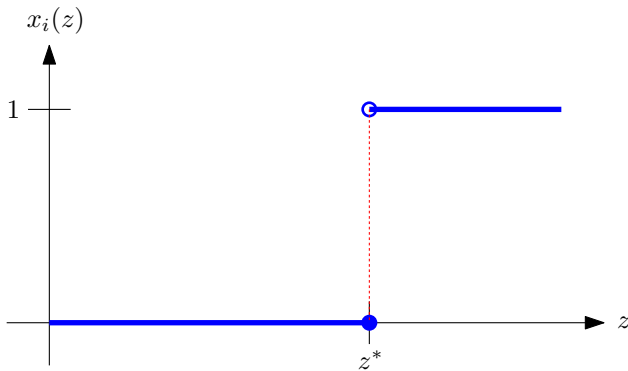
- If  $v(S) < v_{i+1}$ , then the algorithm picks  $\{i+1\}$ , so  $APX = v_{i+1}$ . Then

$$v^* \leq v(S) + v_{i+1} \leq v_{i+1} + v_{i+1} = 2 \cdot APX.$$

So  $APX \geq \frac{v^*}{2} \geq \frac{1}{2}OPT$ . □

**Exercise?** The allocation rule derived from the approximation algorithm is monotone.

Myerson's Lemma gives the payment rule. A player's allocation has one jump point at some point  $z^*$ . The payment is  $z^*$ .



**How to find  $z^*$ ?** Figure out the point at which the approximation algorithm will switch between picking the item and not picking the item.

**Example.**  $T = 10$

$i$	1	2	3	4
$t_i$	2	7	3	5
$v_i$	20	56	21	25
$v_i/t_i$	10	8	7	5

The algorithm picks 1 and 2. What is the payment for player 2? Lower  $v_2$  until the algorithm does not pick  $v_2$ .

As long as  $v_2/t_2 \geq 7$ , the algorithm will pick 2. So player 2 can go down to 49 and still be picked.

What happens if  $v_2 < 49$ ? Say  $v_2/t_2 \geq 5$ . The algorithm picks 1 and 3 with combined value 41.

If  $v_2 > 41$ , then the algorithm will still pick 2.

If  $v_2 \leq 41$ , then the algorithm will stick with 1 and 3.

The jump point is at 41, so the payment is 41.