

SOCIAL WELFARE MAXIMIZATION

a.k.a. Economic efficiency.

- n agents ; m outcomes
- $V_i(j)$: value of agent i for outcome j .
- Goal : Find $j^* = \operatorname{argmax}_j \sum_i V_i(j) =: SW_{j^*}$.

Combinatorial auctions : Outcomes are partitions of objects across agents.

Vickrey
Clark
Groves

VCG Mechanism

- Ask agents to report value functions.
- Compute $j^* = \operatorname{argmax}_j \sum_i V_i(j)$
- Charge agents their "critical value".

$$p_j^{(i)} = \underbrace{\max_{j'} V_{-i}(j')}_{\substack{\text{Value of others} \\ \text{optimized in absence of } i \\ \text{AGENT } i\text{'s EXTERNALITY}}} - \underbrace{V_{-i}(j^*)}_{\substack{\text{val. of others in} \\ \text{presence of } i}}$$

Independent of i 's value once j^* is fixed

Agent i 's alternate view:

- Ask other agents to report values

- Compute

$$p_j^{(i)} = \max_{j'} V_{-i}(j') - V_{-i}(j)$$

CRITICAL PRICES

- Allow agent i to select outcome j^* .

Agent i will choose

$$\tilde{j}^* = \operatorname{argmax}_j V_i(j) - p_j^{(i)}$$

Saw in prev. lec : $j^* = \tilde{j}^*$

Claim : VCG is Dominant Strategy Incentive Compatible

Corollary : VCG implements the efficient outcome.

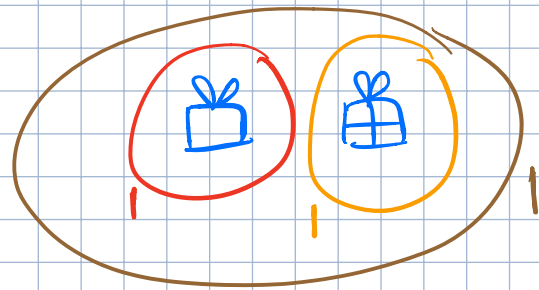
Properties of VCG

- Efficient
- Dominant Strategy Incentive Compatible. (DSIC).

Problems with VCG.

- Needs exact computation of optimum — often computationally hard.
- Needs too much communication.
- Revenue non-monotonicity.
- Collusion ; false name bids.
- Credibility on part of the auctioneer.

Spectrum Auctions



Red & Brown agent

- Red wins & pays \$1.

R & B & O participate

- R & O win
- R pays 0
- O pays 0.

Special simpler case: Single-minded agents.

- n agents; m items

- Agent i 's value is given by (S_i, v_i)

set of items
 i wants

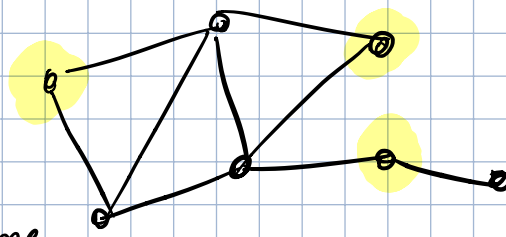
value for
that set

$$\text{For } T \subseteq [m], V_i(T) = \begin{cases} 0 & \text{if } T \not\subseteq S_i \\ v_i & \text{if } S_i \subseteq T \end{cases}$$

Theorem: SW maximization is NP-hard for single minded agents.

Independent Set

- NP-hard to approximate better than $n^{1-\epsilon}$ for some $\epsilon > 0$



$n = \#$ vertices.

Reduction

- Agents are vertices.

- Items are edges. $m \approx n^2$.

- Each vertex will want all edges incident on it.

- Each vertex has value 1.

SW max. is hard to approx. better than $\frac{1}{m}^{1-\epsilon}$ factor.

What about truthful approximations?

Possible "greedy" mechanism:

- Ask agents to report (S_i, v_i)
- Order agents in decreasing order of value $v_i/|S_i|$

$(\tilde{S}_i, \tilde{v}_i)$

$$v_1 \geq v_2 \geq \dots \geq v_n.$$

$$\frac{v_1}{|S_1|} \geq \frac{v_2}{|S_2|} \geq \frac{v_3}{|S_3|} \geq \dots$$

- $S = \emptyset$ (set of items allocated so far)

- For $i = 1$ to n ,

If $S_i \cap S = \emptyset$

then allocate S_i to i and set $S = S \cup S_i$

- Charge i their critical price $\theta_i = \text{min value at which } i \text{ wins } S_i$.

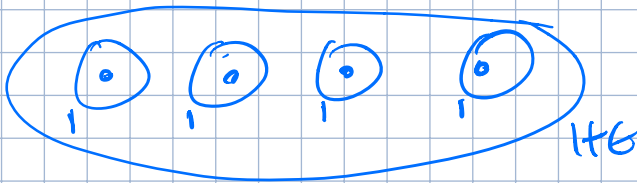
Red variation is DSIC.

Claim: Greedy alg. is DSIC. ✓

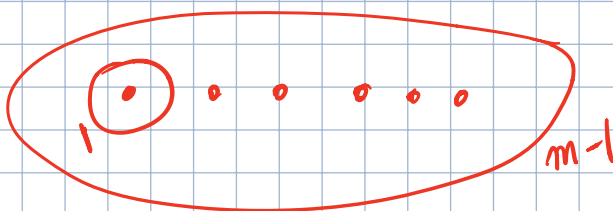
$\tilde{\theta}_i$

i wins if $\tilde{v}_i \geq \tilde{\theta}_i$
 util = $v_i - \tilde{\theta}_i$
 loses if $\tilde{v}_i < \tilde{\theta}_i$
 util = 0

Bad example: n agents; $n-1$ items.



Bad example for red variation: 2 agents, m items



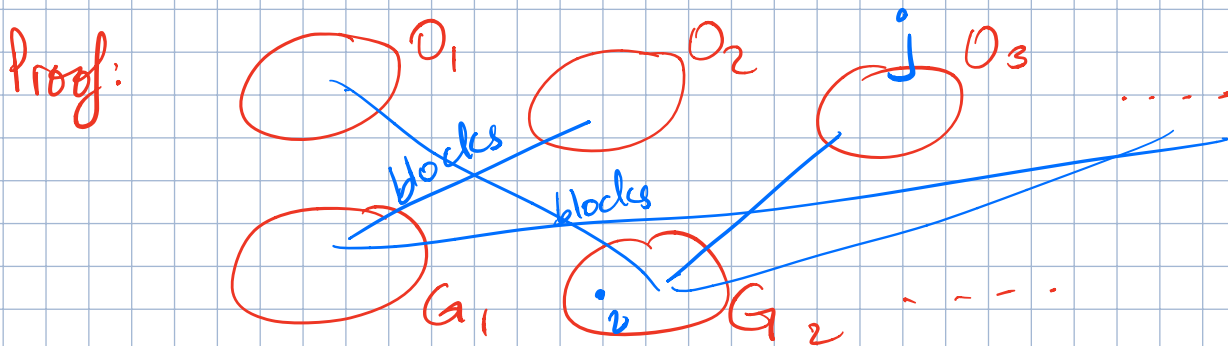
The "LOS" mechanism. Lehman O'Callaghan Shoham '02

- Ask agents to report $(\tilde{S}_i, \tilde{V}_i)$
- Order agents in decreasing order of $\frac{\tilde{V}_i}{\sqrt{|\tilde{S}_i|}}$ $f_i(\tilde{V}_i, \tilde{S}_i)$
- Greedily choose agents to serve in this order.
- Charge critical payments.

Claim: LOS is DSIC.

As long as $f_i(\tilde{V}_i, \tilde{S}_i)$ is non-decreasing in \tilde{V}_i .
and is non-increasing in \tilde{S}_i .
then the greedy mechanism is DSIC.

Claim: LOS achieves an $O(\sqrt{m})$ -approximation.



Every opt. set is blocked by some greedy set.

Consider i allocated by greedy. let T_i = agents allocated in opt. blocked by i .

$$\text{let } j \in T_i \Rightarrow \frac{V_i}{\sqrt{|\tilde{S}_i|}} \geq \frac{V_j}{\sqrt{|\tilde{S}_j|}}$$
$$\Rightarrow V_j \leq V_i \frac{\sqrt{|\tilde{S}_j|}}{\sqrt{|\tilde{S}_i|}}$$

Total optimal value blocked by i

$$= \sum_{j \in T_i} v_j \leq \frac{v_i}{\sqrt{|S_i|}} \sum_{j \in T_i} \sqrt{|S_j|}$$

$$|T_i| \leq |S_i|$$

$$\leq \frac{v_i}{\sqrt{|S_i|}} \sqrt{|T_i|} \sqrt{\sum_{j \in T_i} |S_j|}$$

$$\leq v_i \frac{\sqrt{|S_i|}}{\sqrt{|S_i|}} \cdot \sqrt{m}$$

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

$\Rightarrow \text{OPT} \leq \sqrt{m}$ Value obtained by greedy.

A class of mechs. based on LOS

- Ask agents to report $(\tilde{S}_i, \tilde{v}_i)$
- Order agents in decreasing order of $f_i(v_i, S_i)$
- Greedily choose agents to serve in this order.
- Charge critical payments.

Claim: If $\forall i$, $f_i(v_i, S_i)$ is non-decreasing in v_i and non-increasing in S_i , then the above mechanism is DSIC.

What about other kinds of value functions?

A greedy algo by Borodin & Lucier: ^{'2011}

- Ask agents to report $\tilde{v}_i(\cdot)$.
- While items remain unallocated,
 - let $(i, X_i) = \underset{(i, S) : S \text{ is unallocated}}{\operatorname{argmax}} f_i(\tilde{v}_i, S)$
 - Allocate X_i to i .
- Charge each agent its critical price.
 $\theta_i = \min$ reported value $\tilde{v}_i(X_i)$
at which i would win X_i .

No longer DSIC.

Assumption: Underlying greedy alg. is C-approx
 f_i is non-dec. in \tilde{v}_i and non-inc. in S

NASH EQUILIBRIUM

A tuple of strategies $(\sigma_1, \dots, \sigma_n)$ forms a Nash Equil if for all agents $i \in [n]$, fixing σ_{-i} , agent i 's best response is to play σ_i .

Claim: Suppose that the functions f_i are monotone non-decreasing in v_i and the above greedy algorithm is a c -approximation.

Then, in any NE of the greedy mechanism, the SW obtained is a $(c+1)$ -approx. to the optimal SW.

$$\text{PoA} \leq c+1.$$

1. Observation: bids \leq values.

$$\sum_{(i, x_i) \text{ allocated by greedy}} v_i(x_i) \geq \sum_{(i, x_i) \text{ alloc. by greedy}} \tilde{v}_i(x_i)$$

$$2. \sum_{(i, x_i)} \tilde{v}_i(x_i) \geq \frac{1}{c} \sum_{(i, s_i) \in \text{OPT}} \theta_i(s_i)$$

3. Apply NE property.

$$\sum_{(i, s_i) \in \text{OPT}} v_i(s_i)$$

↑

PRICE OF ANARCHY (maximization objective)

$$\max_{\text{Nash Equil.}} \frac{\text{Optimal obj.}}{\text{Objective at NE.}}$$

Ratio. (like approx ratio)

Worst case over all possible equilibria.