

CS4261 Algorithmic Mechanism Design

→ Players $N = \{1, 2, \dots, n\}$
 Game → Actions
 Preferences over outcomes
 general framework for strategic interaction.

socially optimal: maximize the total benefit
 envy free: no one prefers another bundle more than his own

Socially optimal: maximise $\sum u_i(o)$
 Pareto optimal: $\nexists o'$ s.t. $(u_i(o') \geq u_i(o) \forall i \in N)$
 and $(u_i(o') > u_i(o) \exists i \in N)$
 outcome
 everybody won't be worse off with swapping
 someone gets strictly better-off with swapping

Normal form game → set of players $N = \{1, 2, \dots, n\}$
 → set of actions for each player A_i
 → action profile $\vec{a} \in A_1 \times A_2 \times \dots \times A_n$
 → utility of each \vec{a} for each player $u_i(\vec{a})$.

2,1	1,2
1,2	2,1

Dominant strategy: A best option for a player, regardless of what other players choose
 (may or may not exist)
 (weak) Domination: $\vec{p} \in \Delta(A_i)$ dominates $\vec{q} \in \Delta(A_i)$: $\forall \vec{p}_{-i} \in \Delta(A_{-i})$: $u_i(\vec{p}_{-i}, \vec{p}) \geq u_i(\vec{p}_{-i}, \vec{q})$
 Best response set (Strong) Domination: strict ineq. for all.

$u_1(T, q) = \dots$
 $BR_1(q) = \begin{cases} T (p=1) & \text{if } q > \frac{1}{2} \\ B (p=0) & \text{if } q < \frac{1}{2} \\ p \in [0, 1] & \text{if } q = \frac{1}{2} \end{cases}$
 if player 2 plays L with prob q

$BR_i(\vec{a}_{-i}) := \{b \in A_i \mid b \in \text{argmax } u_i(\vec{a}_{-i}, b)\}$
 the set of options that yield the best outcome for me, given what everyone else is going to play.

A	5, 4	4, 3
B	2, 5	3, 4

Nash equilibrium: $\forall i \in N, a_i \in BR_i(\vec{a}_{-i})$
 i.e. it is in everyone's best response set.

$A > B$ (dominates)
 because $5 > 2$
 $\{4 > 3\}$

Mixed strategies

- Player utility: expected value (we assume players are risk-neutral)

- Mixed Nash equilibrium: $\forall i \in N$:
 always exists!

- Best response
 similarly defined.

$\forall \vec{q}_i \in \Delta(A_i), u_i(\vec{p}) \geq u_i(\vec{p}_{-i}, \vec{q}_i)$
 support of a distribution:

It is never better (in terms of utility) EV to switch to a different mixed strategy.

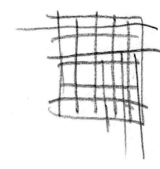
5, 4	4, 4
5, 5	4, 5

Nash equilibrium

intersections are Nash equilibriums.

set of options taken with nonzero probability.

Theorem: If action $a \in A_i$ is strictly dominated by some $\vec{p} \in \Delta(A_i)$ then a is never played (with any positive probability) in a Nash eqm.



iterated removal of dominated strategies



A mixed strategy can dominate another strategy too!

in a Nash eqm.

Zero-sum game: \rightarrow 2-player game
 $\rightarrow \forall a_i \in A_1, \forall b_j \in A_2 : u_1(a_i, b_j) = -u_2(a_i, b_j)$

Mixed Nash eqm is poly-time computible via simplex.

Duality of linear optimization (simplex) problem:

Primal
 minimize $\vec{c}^T \vec{x}$
 s.t. $A\vec{x} \geq \vec{b}$
 and $x_i \geq 0 \forall i \in N$

Dual
 maximize $\vec{b}^T \vec{y}$
 s.t. $A^T \vec{y} \leq \vec{c}$
 $y_j \geq 0 \forall j \in M$

Theorems: If \vec{x}^* and \vec{y}^* are the optimal solutions, then:

- ① Optimality: $\vec{c}^T \vec{x}^* = \vec{b}^T \vec{y}^*$
- ② Complementary Slackness:
 If $x_i^* > 0$ then $(A^T \vec{y}^*)_i = c_i$
 If $y_j^* > 0$ then $(A \vec{x}^*)_j = b_j$

Given A, \vec{b}, \vec{c} , the optimal for both problems are the same.

von Neumann Minimax thm: For any payoff matrix A ,

$$\max_{\vec{p} \in \Delta(A_1)} \min_{\vec{q} \in \Delta(A_2)} \vec{p} A \vec{q}^T =: v_+ = v_- =: \min_{\vec{q} \in \Delta(A_2)} \max_{\vec{p} \in \Delta(A_1)} \vec{p} A \vec{q}^T$$

get expected payoff for strategy
minimax

To prove: ① show that it is (at least) as good if player 2 uses a pure strategy,

$$i.e. \max_{\vec{p} \in \Delta(A_1)} \min_{\vec{q} \in \Delta(A_2)} \vec{p} A \vec{q}^T = \max_{\vec{p} \in \Delta(A_1)} \min_j (\vec{p} A)_j$$

min over all mixed strategies
min over all pure strategies

② Use LP duality:

$$\begin{aligned} \max v_+ & \\ \text{s.t. } v_+ & \leq \sum_{i=1}^n a_{ij} x_i \text{ for all } j \in M \\ \text{and } \sum_{i=1}^n x_i & = 1 \end{aligned} \quad = \quad \begin{aligned} \min v_- & \\ \text{s.t. } v_- & \geq \sum_{j=1}^m a_{ij} y_j \text{ for all } i \in N \\ \text{and } \sum_{j=1}^m y_j & = 1 \end{aligned}$$

all pure strategies

Support of a Nash eqm: support of \vec{p} : $\{a : p(a) > 0\}$

all those options that have +ve contribution to \vec{p} .

Thm: If (\vec{p}, \vec{q}) is a Nash eqm and $a \in \text{supp}(\vec{p})$ then:
 $u_1(a, \vec{q}) \geq u_1(a', \vec{q}) \forall a' \in A_1$

Solving Nash eqm for two players directly cannot be done with LP, but with thm it can be done:

Find \vec{p}, \vec{q} s.t. $\sum_{a \in A_1} p(a) = 1, \sum_{b \in A_2} q(b) = 1$

$\forall \vec{p}' \in \Delta(A_1) : u_1(\vec{p}, \vec{q}) \geq u_1(\vec{p}', \vec{q})$

$\forall \vec{q}' \in \Delta(A_2) : u_2(\vec{p}, \vec{q}) \geq u_2(\vec{p}, \vec{q}')$

(there are infinitely many constraints)

Find \vec{p}, \vec{q} s.t. $\sum_{a \in A_1} p(a) = 1, \sum_{b \in A_2} q(b) = 1$
 $\forall a \notin A_1, p(a) = 0; \forall a \in A_1, p(a) > 0$
 $\forall b \notin A_2, q(b) = 0; \forall b \in A_2, q(b) > 0$
 $\forall b \in B_2, \forall b' \in A_2, u_2(\vec{p}, b) \geq u_2(\vec{p}, b')$
 $\forall a \in B_1, \forall a' \in A_1, u_1(a, \vec{q}) \geq u_1(a', \vec{q})$

exponential time alg. by trying all subsets.

Sperner's Lemma

Given any n -simplex (e.g. triangle, tetrahedron, ...) with corners coloured in distinct colours:

- Triangulate it in any way, where colours of vertices on each face must come from a colour at any of its corners.
- Then there exists an odd number of simplices whose vertices use all $(n+1)$ colours

Proof for $n=2$:

- Q : = num. of triangles with (G, B, B) or (G, G, B) colours of vertices
- R : = num. of triangles with (R, G, B)
- X : = num. of edges with (G, B) on boundary
- Y : = num. of edges with (G, B) on interior

$\therefore 2Q + R = \text{number of directed GB or BG edges} = 2Y + X$

$\therefore X$ is odd, $\therefore R$ is odd ↑ can be shown to be odd.

Brouwer fixed point thm: Given $f: A \rightarrow B$ continuous and K is compact convex set then:

Nash theorem: A Nash equilibrium always exists. $\exists x \in K$ s.t. $f(x) = x$

Regret Minimisation:

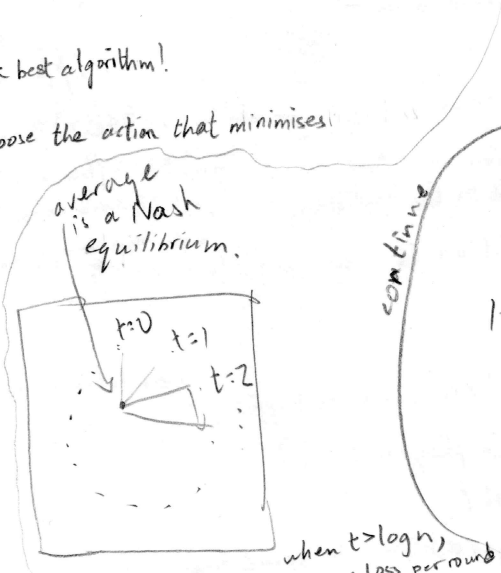
n : = number of actions
 L_{π}^t : = expected loss of algorithm π at time t
 might play mixed strategies
 $L_{\pi}^t = \sum_{i=1}^t L_{\pi}^i$: = total loss of algorithm π up to time t .

Regret: $L_{\pi}^t - L_{\text{best}}^t$
 ↑ best action, not best algorithm!

Greedy algorithm: At time t , choose the action that minimises L_i^t .
 (tie-break by lower index)

Regret of $O(n)$
 i.e. $\frac{L_{\text{greedy}}^t}{L_{\text{best}}^t} \in O(n)$

Randomised greedy alg.:
 if there are ties for the best action, then choose uniform probability.



Multiplicative weight updates
 $W^{t+1} = \sum_{i=1}^n w_i^{t+1} \geq W_{\text{best}}^{t+1}$

$w_i^{t+1} = e^{-\epsilon L_i^t}$

Formula: $\forall x \in [-1, 1]: e^x \leq 1 + tx + \frac{t^2 x^2}{2}$

$\therefore W^{t+1} \leq \sum_{i=1}^n w_i^t (1 + (-\epsilon L_i^t) + \frac{(-\epsilon L_i^t)^2}{2})$

$\leq (1 + \epsilon^2) (\sum w_i^t) - \epsilon (\sum w_i^t L_i^t)$
 $= W_t (1 + \epsilon^2 - \epsilon \sum_{i=1}^n L_i^t p_i^t)$

$1 + x \leq e^x \rightarrow$

$\leq W_t e^{\epsilon^2 - \epsilon L_{\pi}^t}$

when $t > \log n$, average loss per round for alg. is $O(1)$ to average loss per round for best.

$L_{\pi}^t \leq L_{\text{best}}^t + \frac{\log n}{\epsilon} + \epsilon t$

$L \leq W e^{\epsilon^2 - \epsilon (\sum_{i=1}^t L_{\pi}^i)}$
 $= n e^{\epsilon^2 - \epsilon L_{\pi}^t}$

$\therefore \frac{e^{-\epsilon L_{\text{best}}^t}}{e^{\epsilon L_{\pi}^t}} \leq W_{t+1} \leq n \cdot e^{\epsilon^2 - \epsilon L_{\pi}^t}$
 $\Rightarrow e^{\epsilon L_{\pi}^t} \leq e^{\epsilon L_{\text{best}}^t} \cdot n \cdot e^{\epsilon^2 t}$

Multiplicative weight updates (MWU)

Initially: $\{w_i^1 = 1, p_i^1 = \frac{1}{n}\}$
 At t , $\{w_i^t = w_i^{t-1} e^{-\epsilon L_i^{t-1}}, p_i^t = \frac{w_i^t}{W_t}, W_t = \sum w_i^t\}$

when $\epsilon = 0 \Rightarrow$ pure random play
 $\Leftrightarrow +\infty \Rightarrow$ randomised greedy.

Multiplicative weight update & minimax thm

If player 2 uses an online algorithm A with regret R, then

average loss is at most $v_- + \frac{R}{T}$ after T rounds:

Pf: $L_A^T \leq L_{best}^T + R \leq T v_- + R \Rightarrow \frac{L_A^T}{T} \leq v_- + \frac{R}{T}$

there is some pure strategy that guarantees P2 a loss of at most v_- against any \bar{p} at every round.

⇒ If both players use MWU to pick a strategy at every round, then their average strategies are a Nash eqm of the underlying minimax game.

Routing games

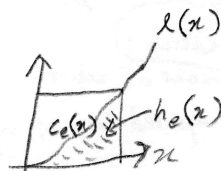


$l_e: \mathbb{R} \rightarrow \mathbb{R}^+$
(congestion function of edge e)

-atomic when edges must have integer flow

Equilibrium: switching to a different path never yields a better result.

Braess' Paradox: Adding an edge increases the social cost.



Price of Anarchy: $\frac{\text{Worst Nash}(G)}{\text{OPT}(G)}$

In non-atomic version, pure Nash eqm always exists!

Let $c_e(x) := x \cdot l_e(x)$ (total social cost of congestion at e)
 Let $c'_e(x) := \frac{d}{dx} c_e(x)$
 Lemma: $(\forall P, P' \in \mathcal{P}_i, f_P^* > 0 \Rightarrow c'_P(f^*) \leq c'_{P'}(f^*))$
 iff a flow is optimal. (i.e. marginal social cost is no more than any other paths.)

$l_e^*(x) := c'_e(x) = l_e(x) + x \cdot l'_e(x); l_P^*(x) = \sum_{e \in P} l_e^*(x)$
 Corollary: a flow is optimal for $\langle G, r, l \rangle \Leftrightarrow$ it is a NE for $\langle G, r, l^* \rangle$
 Thm: IF there exists $\alpha > 1$ s.t. $c_e(x) \leq \alpha x l_e(x) \forall e \in E$, then $\text{PoA} \leq \alpha$.

Cooperative games

Induced subgraph (undirected, weighted): value of a coalition is the sum of edge weights in the coalition.

Network flow (directed, weighted): value of a coalition is the maxflow using those edges only.

Weighted voting games: $(w_1, \dots, w_n; q)$
 a cutoff each player (n players total) has a weight

coalition value = 1 if sum of member weights $\geq q$, 0 otherwise

Bankruptcy problem: split debt amongst creditors in some way.

$N = \{1, 2, \dots, n\}$
 characteristic f^o: $v: 2^N \rightarrow \mathbb{R}$ (value function) mapping from power set to \mathbb{R}
 coalition structure (cs) = some partition of N; $\text{OPT}(G) = \max_{CS} \sum_{S \in CS} v(S)$
 imputation: $\forall S \in CS, \sum_{i \in S} x_i = v(S)$ (the amount to pay each player)
 simple game: $v(S) \in \{0, 1\}$
 monotone: $\forall S \subseteq T \subseteq N \Rightarrow v(S) \leq v(T)$ (i.e. adding people to a group is not worse.)
 convex game: $S \subseteq T \subseteq N$ and $i \in N \setminus T \Rightarrow v(S \cup \{i\}) - v(S) \leq v(T \cup \{i\}) - v(T)$ where S, T are disjoint
 super additive game: $\forall S, T \subseteq N: v(S) + v(T) \leq v(S \cup T)$
 i.e. joining the big group is always better.

Core of a cooperative game:

An imputation \vec{x} is in the core if $\sum_{i \in S} x_i =: x(S) \geq v(S)$, $\forall S \subseteq N$

(i.e. no subset of the coalition will want to break off.)

$$* \Rightarrow \underbrace{v(S)}_{\substack{\text{amount that} \\ S \text{ can get on their own.}}} \leq \sum_{i \in S} x_i \leq \underbrace{v(N) - v(N \setminus S)}_{\substack{\text{marginal utility} \\ \text{of the coalition by} \\ \text{adding } S \text{ to them}}}$$

Proof that core is empty is NP-hard.

Simple game: $\forall S \subseteq N, v(S) \in \{0, 1\}$

winning coalition: those with value 1

losing coalition: otherwise

veto player: a player that is a member of every winning coalition (can't win without them)

Thm: Let $G = \langle N, v \rangle$ be a simple game, then $(\text{Core}(G) \neq \emptyset \Leftrightarrow G \text{ has veto players})$ and $v(N) = 1$

Lemma: Core of induced subgraph game is not empty \Leftrightarrow graph has no negative cut.

Shapley value: $\frac{1}{n!} \sum_{\sigma \in \Pi(N)} m_i(\sigma)$
marginal contribution of i in σ .

Satisfies:

- Efficiency ($\sum_{i \in N} \phi_i = v(N)$): all money are distributed

- Symmetry ($\forall S \subseteq N \setminus \{i, j\}: v(S \cup \{i\}) = v(S \cup \{j\}) \Rightarrow \phi_i = \phi_j$): equal players are paid equally

- Dummy/Null player ($\forall S \subseteq N \setminus \{i\}: v(S \cup \{i\}) = v(S) \Rightarrow \phi_i = 0$): those who don't contribute are not paid anything

- Additivity/Linearity ($\phi_i(v_1) + \phi_i(v_2) = \phi_i(v_1 + v_2)$): combined game combines the payment

Thm: Shapley value is the only division that satisfies all four properties

value function 1 value function 2

Nash bargaining solution

$$= \max_{(v_1, v_2) \in S} (v_1 - d_1)(v_2 - d_2)$$

↑
assuming they are in first quadrant of (v_1, v_2) .

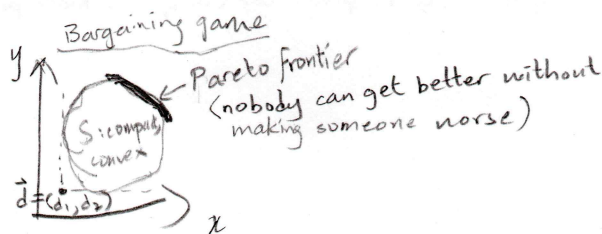
Satisfies:

- Efficiency: no outcome Pareto-dominates $(f_1(S, \vec{d}), f_2(S, \vec{d}))$

- Symmetry: reflecting S and \vec{d} on $y=x$ should reflect the chosen point

- Independence of Irrelevant Alternatives (IIA): Removing portions of S that do not contain the chosen point should not change the chosen point.

- Invariance under Equivalent Representations (IER): Translation and scaling on either axis should translate and scale the chosen point in the same way.



Good markets

- Thick (lots of buyers & sellers, everyone is aware of their options)
- Timely (not too fast or too slow)
- Safe (people cannot be hurt by revealing preferences, fair outcomes, people are better off by participating)

Unravelling

- Matching is done earlier & earlier (bad)

Matching scenario

- Set of students $S = \{s_1, \dots, s_n\}$
- Set of hospitals $H = \{h_1, \dots, h_m\}$
- Each student ranks hospitals with a total ordering: \succ_s
- Each hospital ranks students with a total ordering: \succ_h

Outcome: A one-to-one matching $M: S \rightarrow H$

Blocking pair of a matching M : A pair (s, h) where $h \succ_s M(s)$ and $s \succ_h M^{-1}(h)$ (i.e. both of them prefer each other to their assigned counterpart)

Gale-Shapley Algorithm \rightarrow polytime & returns a stable matching

- Start with all students unassigned
 - While there are unassigned students:
 - Each unassigned student proposes to their favourite not-yet-proposed-to hospital
 - Each hospital looks at the list of students that proposed to it at this round and whoever is assigned to it already (if exists) and picks the most preferred one. All others remain/become unassigned.
- must always exist.

Matching with complex \rightarrow stable matching might not exist (but in practice usually exists)

Stable if: No (s, h) prefers each other

No $((s_1, s_2), (h_1, h_2))$ prefers each other.

Thm: Gale-Shapley assigns each student to their most preferred hospital in which some stable matching exists.

\Rightarrow It is better for students than hospitals

\Rightarrow Students cannot game the system, but hospitals can (to get a more preferred student)

Allocation of Indivisible Goods

$\pi(i)$: set of goods allocated to player i under π

$v_i(S) := \sum_{g \in S} v_i(g)$ (additive valuations)
 value of good g to player i .

Assume that $\forall i, j \in N, v_i(N) = v_j(N)$.

Desirable properties

• Optimality: π is optimal $:= \pi \in \arg \max_{\pi'} \sum_i v_i(\pi'(i))$

• Pareto optimality: no allocation dominates π . (somebody will get strictly worse off in any other outcome)

• Envy-freeness: no player wants another's bundle: $\forall i, j \in N, v_i(\pi(i)) \geq v_i(\pi(j))$

• Maximin share: if I get to partition the items, what is the maximum value (to me) of the worst bundle?

(it is independent of the goods given to other players). i.e. $MMS_i := \max_{\pi} \min_{S \in \pi} v_i(S)$

• Maximin share requirement: Each player gets at least their maximin share, i.e. $\forall i \in N, v_i(\pi(i)) \geq MMS_i$

Approximate solutions

• EF-1: no player wants another's bundle with the best good removed: $\forall i, j \in N, \exists g \in \pi(j) \text{ s.t. } v_i(\pi(i)) \geq v_i(\pi(j) \setminus \{g\})$

• α -EF (for some $0 < \alpha < 1$): $\forall i, j \in N, v_i(\pi(i)) \geq \alpha \cdot v_i(\pi(j))$

• α -MMS (for some $0 < \alpha < 1$): $\forall i \in N, v_i(\pi(i)) \geq \alpha \cdot MMS_i$

Thms for indivisible goods with additive valuations

- EF allocation does not always exist (consider a single good divided amongst two players)
- EF-1 allocation always exists
- MMS allocation does not always exist
- $\frac{2}{3}$ -MMS allocation always exists
- Deciding if an EF allocation exists is NP-complete (by reduction from PARTITION problem, where every item is valued the same by all players)

Algorithm for finding an EF-1 allocation in $O(mn^3)$ time :

- While there is an unallocated good g :
 - Give g to a player that nobody envies (which must always exist because the envy graph is a DAG)
 - Decycle the envy graph
 - For each cycle, swap the bundles amongst them. The cycle disappears, and other people that envy player j will now envy the player that j 's bundle was given to. So the number of edges decreases. Continue until there are no cycles left.
- End.

Rent Division

$N := \{1, \dots, n\}$ players
 $G := \{g_1, \dots, g_n\}$ rooms
 $V_{ij} :=$ player i 's valuation of room j .
 Assume that $\forall i, j \in N, \sum_k V_{ik} = \sum_k V_{jk} =: R$ ← total rent that needs to be paid.

Output: a room allocation $\sigma: N \rightarrow N$ and a rent division \vec{p} where $\sum_j p_j = R$

EF outcome: $\langle \sigma, \vec{p} \rangle$ s.t. $\forall i: \sigma(i) - p_{\sigma(i)} \geq V_{ij} - p_j \quad \forall i, j \in N$.
 i.e. "I don't prefer your room for the price you're being charged"

↑ payment for room j

Thms for rent division:

- EF outcome always exists, and can be computed efficiently
- In any EF outcome, room allocation (i.e. σ) is optimal (1st welfare thm)
- If outcome $\langle \sigma, \vec{p} \rangle$ is EF, then so is $\langle \sigma', \vec{p} \rangle$ for any optimal room allocation σ' . (2nd welfare thm)

General algorithmic framework

• Compute a socially optimal allocation (max weighted matching) → Find an EF price vector (linear programming) → This price vector will work with any optimal allocation (2nd welfare thm)

Ways to choose amongst EF outcomes

- Equitability: minimise the disparity between the highest and lowest utilities: $\arg \min_{\vec{p} \in EF} \max_{i \in N} (u_i(\vec{p}) - u_j(\vec{p}))$
- Maximin: maximise the minimum utility: $\arg \max_{\vec{p} \in EF} \min_{i \in N} u_i(\vec{p})$

Thms:

• There is a unique maximin EF price vector, and this vector is also equitable (however, there might exist equitable price vectors that are not maximin)

Single item auctions

- English auction: auctioneer sets a starting price, bidders take turns raising their bids, last bidder wins and pays his bid.
 - It is rational to bid iff $p + \delta \leq v$, and I should bid $p + \delta$. Winner will eventually pay second highest value or second highest value + δ .
- Japanese auction: auctioneer sets a starting price then keeps raising it until all but one bidder drops out, last bidder pays the current price.

- Dutch auction: auctioneer sets a high starting price and then starts lowering it until a bidder accepts the price.
- Sealed-bid auction: all bidders simultaneously submit their bids, the higher bidder gets the item and pays
 - his bid (first-price auction), or
 - 2nd highest bid (second-price, or Vickrey, auction)

Bidders must trust the auctioneer in a sealed-bid auction.

Thm: In a Vickrey auction, truthful bidding is a dominant strategy (i.e. truthful bidding will never make a person worse off regardless of the actions of everyone else.)

Note: There are many NEs in a Vickrey auction.
 e.g. $\vec{v} = (50, 30, 70)$ (valuations)
 possible NE: $(50, 30, 70), (0, 0, 70), (70, 0, 0)$

↑
 third player wins and pays nothing
 ↑
 first player wins and pays nothing.

Multi unit auctions (identical items)

- There are multiple, identical items
- Each bidder wants one item, and values it at v_i
- How should the items be distributed, and how much should they pay?

Multi unit auctions (different items)

- There are multiple, different items.
- Each player wants one item, values item j at v_{ij} .
- How should the items be distributed, and how much should they pay?

(General) Mechanism Design

- Players $N = \{1, \dots, n\}$
 - Outcomes $O = \{o_1, \dots, o_m\}$
 - Each player has a utility function $u_i : O \rightarrow \mathbb{R}$
 - Centre chooses an outcome to maximise some function (e.g. $\sum_i u_i(o^*)$)
- ↑ takes into consideration payments if any

Incentive Compatibility: If everyone else reports their true valuations, I should report truthfully too. (i.e. reporting true valuations is a NE)

Dominant Strategy Incentive Compatibility (i.e. strategyproofness): I should report truthfully regardless of everyone else. (i.e. reporting true valuations is a (weakly) dominant strategy.)

Revelation Principle: Given any mechanism M , there exists a mechanism M' whose inputs are users' valuations, and whose outputs are exactly like those of M . (i.e. there is always an incentive-compatible mechanism)

Problems: computational burden is pushed to the center
 • the direct mechanism might have additional bad equilibria.

Vickrey-Clarke-Groves Mechanisms

1. Choose an outcome that maximises $\sum_i v_i(o^*)$ (i.e. socially optimal outcome)
2. To determine the payment that player j must make:
 - pretend j does not exist, and choose o_{-j}^* that maximises $\sum_{i \neq j} v_i(o_{-j}^*)$
 - j pays $\sum_{i \neq j} v_i(o_{-j}^*) - \sum_{i \neq j} v_i(o^*)$ (i.e. the negative externality that player j imposes on others)

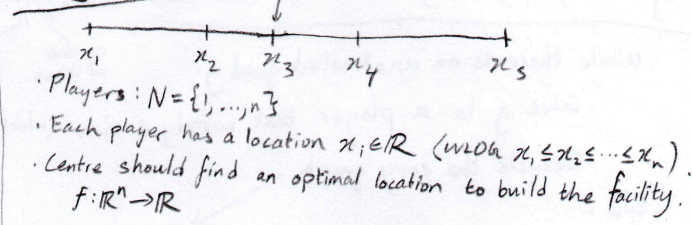
Thm: In a VCG mechanism, truthful reporting is a dominant strategy.

Proof: $u_i(o^*) = v_i(o^*) - \left(\sum_{i \neq j} v_i(o_{-j}^*) - \sum_{i \neq j} v_i(o^*) \right)$

$$= \underbrace{\sum_i v_i(o^*)}_{\text{total social welfare (any other outcome } o' \text{ must have lower total social welfare than } o^* \text{ (which is socially optimal))}} - \underbrace{\sum_{i \neq j} v_i(o_{-j}^*)}_{\text{does not depend on player } j \text{'s reporting}}$$

- Assumptions:
- Choice set monotonicity: $O_{-i} \in O$
 - No negative externalities: $\forall o_{-i} \in O_{-i}, v_i(o_{-i}) \geq 0$

Facility Location

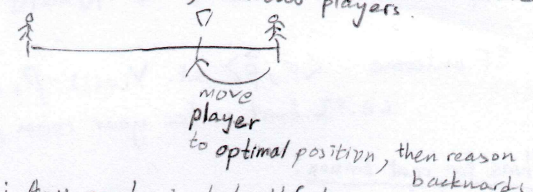


- Measures:
 To minimise...
 • Total cost: $\sum_i |f(\vec{x}) - x_i|$
 • Max cost: $\max_i |f(\vec{x}) - x_i|$

Thm: Choosing the median position has the minimum total cost. It is also strategyproof and group-strategyproof. (tiebreak towards larger player?)
 regardless of what others do, it is best to play truthfully.
 no subset of players can collude to make everybody in the subset strictly better off.

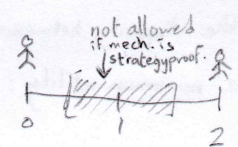
Thm: Any deterministic truthful mechanism has worst-case approximation of at least 2 to the maximum cost.

Proof by contradiction, with two players. (i.e. we are using the max cost measure.)



Thm: Any randomised truthful mechanism has worst-case expected approximation of at least $\frac{3}{2}$ to the maximum cost.

Proof: wlog let $E[f] \leq \frac{1}{2}$. Then the person at 1 has expected cost of $\geq \frac{1}{2}$. Then if he lies and claims that he is at 2 instead, the mechanism has to pick a location either $\leq \frac{1}{2}$ or $\geq \frac{3}{2}$.



Now, if instead we have one person at 0 and one person at 2 playing truthfully, due to the above reasoning, the mechanism has to pick a location either $\leq \frac{1}{2}$ or $\geq \frac{3}{2}$, hence the maximum cost is $\geq \frac{3}{2}$.