Mechanism Design with Monetary Transfers

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The Setup

- Unrestricted Preferences
- Restricted Preferences
- 2 Mechanisms in Quasi-linear Domain
 - Structure of a Mechanism
 - Some Definitions

Results

- Groves Class of Mechanisms
- What Other Mechanisms are Incentive Compatible
- Revenue Equivalence
- Uniqueness of Groves for Efficiency
- Budget Balance
- Bayesian Incentive Compatibility

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The Gibbard-Satterthwaite Setting

Voters can have arbitrary *strict ordinal* preferences over the set of alternatives
Set of alternatives X = {a, b, c, d}

Voter 1	Voter 2	Voter 3	Voter 4
a	d	c	d
b	b	b	b
с	a	a	с
d	c	d	a

The Gibbard-Satterthwaite Setting

- Voters can have arbitrary strict ordinal preferences over the set of alternatives
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- Goal: elicit the preferences truthfully from the agents

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Theorem (Gibbard (1973), Satterthwaite (1975))

If $|X| \ge 3$, an onto social choice function is strategyproof if and only if it is dictatorial.

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- Payments p belong to \mathbb{R}^n
- Each agent i has a valuation function $v_i : A \to \mathbb{R}$ belonging to the set V_i
- Agents' utilities are given by

$$u_i(x) = u_i(a, p) = v_i(a) - p_i$$

Example: Public Good

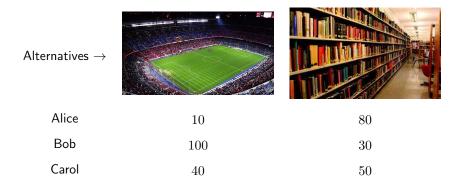
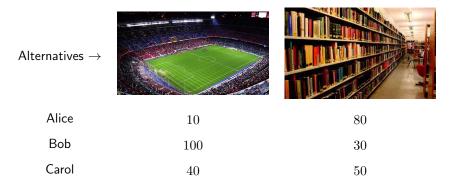


Photo courtesy: wikimedia.org and nimsuniversity.org

Example: Public Good



- Valuations: $v_A(F) = 10, v_A(L) = 80$
- Social planner takes the decision of building F or L
- Can tax people differently depending on their preferences

Example: Resource Allocation

$\stackrel{Commodities}{\to}$	IBMSmartCloud	/ Cloud	cisco
Alice	0.2	0.8	0.5
Bob	0.3	0.1	0.2
Carol	0.5	0.1	0.3

Photo courtesy: individual organizations

Example: Resource Allocation

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- Set of allocations $A = \{x \in [0,1]^{n \times m} : \sum_{j=1}^m x_{i,j} = 1\}$
- Items are divisible among the agents
- Agents' valuations reflect their preferences over different allocations
- They are charged monetary transfers for every allocation

Example: Resource Allocation

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Selfish valuations

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 - Consider two alternatives $x_1 = (a, p_1)$ and $x_2 = (a, p_2)$, where $p_1 < p_2$
 - For all agents, $x_1 \succ x_2$ for any valuation profile
 - There is no preference profile where $x_2 \succ x_1$

An Example of a Truthful Mechanism







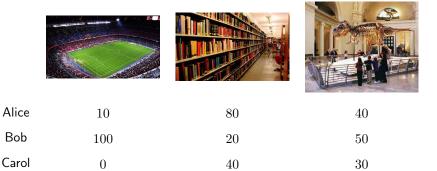
Alice	10	80	40
Bob	100	20	50
Carol	0	40	30

An Example of a Truthful Mechanism



- Consider the mechanism:
 - pick the alternative a^* that maximizes the sum of the valuations (with arbitrary tie-breaking rule)
 - ▶ pay every agent i an amount $\sum_{j \neq i} v_j(a^*)$

An Example of a Truthful Mechanism



- Consider the mechanism:
 - pick the alternative a^* that maximizes the sum of the valuations (with arbitrary tie-breaking rule)
 - ▶ pay every agent i an amount $\sum_{j \neq i} v_j(a^*)$
- The mechanism is truthful, even though not a dictatorship

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Structure of a Mechanism

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- Set of allocations A, finite (for this tutorial)
- Valuation of agent $i, v_i : A \to \mathbb{R}$, the set of valuations is denoted by V_i

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- Valuation of agent $i, v_i : A \to \mathbb{R}$, the set of valuations is denoted by V_i
- A mechanism in quasi-linear (QL) domain is a pair of functions:
 - allocation function, $a: \prod_{j} V_{j} \to A$
 - ▶ payment function, $p_i : \prod_j V_j \to \mathbb{R}$, for all $i \in N$
- Agent *i*'s payoff is given by:

 $v_i(a(v)) - p_i(v)$

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• Only direct revelation mechanisms (DRM) (this talk)

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Social Choice Function

Definition (Social Choice Function)

A social choice function (SCF) f is a mapping from the set of valuation profiles to the set of allocations, i.e., $f: V \to A$, where $V = \prod_{j} V_{j}$.

• Note that the outcome is only the allocations

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- Note that the outcome is only the allocations
- In QL domain:

A mechanism M = (a, p) implements a SCF f if:

- $a(v) = f(v), \forall v \in V \text{ and},$
- ▶ for every agent $i \in N$, reporting v_i truthfully is an *equilibrium*
- Even though the SCF is concerned with only allocations, payments can also be characterized by *revenue equivalence* (defined later)

Incentive Compatibility

Definition (Dominant Strategy Incentive Compatibility (DSIC))

A mechanism (f, p_1, \ldots, p_n) is dominant strategy incentive compatible if for all $i \in N$ and for all $v_{-i} \in V_{-i} := \prod_{j \neq i} V_j$,

 $v_i(f(v_i, v_{-i})) - p_i(v_i, v_{-i}) \ge v_i(f(v'_i, v_{-i})) - p_i(v'_i, v_{-i}), \ \forall v_i, v'_i \in V_i.$

In this case, payments $p_i, i \in N$ implement f in dominant strategies

Incentive Compatibility (Contd.)

- ${ullet}$ In a Bayesian game, the valuations v are generated through a prior P
- Each agent i knows her own realized valuation v_i and P
- Her belief on the valuations of other agents v_{-i} is given by $P(v_{-i}|v_i)$ derived by Baye's rule

Incentive Compatibility (Contd.)

- ${\ensuremath{\, \bullet }}$ In a Bayesian game, the valuations v are generated through a prior P
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- Her belief on the valuations of other agents v_{-i} is given by $P(v_{-i} \vert v_i)$ derived by Baye's rule

Definition (Bayesian Incentive Compatibility (BIC))

A mechanism (f, p_1, \ldots, p_n) is Bayesian incentive compatible for a prior P if for all $i \in N$,

$$\mathbb{E}_{v_{-i}|v_i}[v_i(f(v_i, v_{-i})) - p_i(v_i, v_{-i})] \ge \mathbb{E}_{v_{-i}|v_i}[v_i(f(v'_i, v_{-i})) - p_i(v'_i, v_{-i})]$$

$$\forall v_i, v'_i \in V_i.$$

In this case, payments $p_i, i \in N$ implement f in a Bayesian Nash equilibrium

Observations on IC

• A DSIC mechanism is always BIC

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For a DSIC mechanism $(f,p_1,\ldots,p_n),$ let valuations of agents other than i is fixed at v_{-i}

• If v_i, v_i' be such that $f(v_i, v_{-i}) = f(v_i', v_{-i})$, then $p_i(v_i, v_{-i}) = p_i(v_i', v_{-i})$

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- If v_i, v'_i be such that $f(v_i, v_{-i}) = f(v'_i, v_{-i})$, then $p_i(v_i, v_{-i}) = p_i(v'_i, v_{-i})$
- Consider another payment $q_i(v_i, v_{-i}) = p_i(v_i, v_{-i}) + h_i(v_{-i})$,

 $v_i(f(v_i, v_{-i})) - q_i(v_i, v_{-i}) \ge v_i(f(v'_i, v_{-i})) - q_i(v'_i, v_{-i}), \ \forall v_i, v'_i \in V_i.$

Efficiency

Definition (Efficiency)

An SCF f is *efficient* if for all $v \in V$,

$$f(v) \in \underset{a \in A}{\operatorname{argmax}} \sum_{i \in N} v_i(a).$$

An efficient SCF ensures that the 'social welfare' is maximized

Revenue Equivalence

• This property characterizes the payment functions

Definition (Revenue Equivalence)

An SCF f satisfies *revenue equivalence* if for any two payment rules p and p' that implement f, there exist functions $\alpha_i : V_{-i} \to \mathbb{R}$, such that,

 $p_i(v_i, v_{-i}) = p'_i(v_i, v_{-i}) + \alpha_i(v_{-i}), \ \forall v_i \in V_i, \forall v_{-i} \in V_{-i}, \forall i \in N.$

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- Saw an example of a payment of agent *i* being different by a factor not dependent on *i*'s valuation
- This property says more: pick *any* two payments that implement *f* they must be different by a similar factor

Budget Balance

Definition (Budget Balance) A set of payments $p_i: V \to \mathbb{R}, i \in N$ is budget balanced if, $\sum_{i \in N} p_i(v) = 0, \forall v \in V.$

- This property ensures that the mechanism does not produce any monetary surplus
- Hard to satisfy with incentive compatibility

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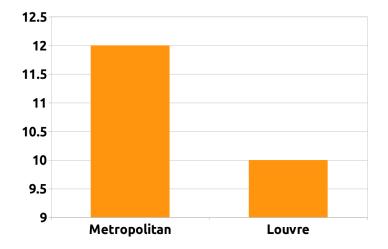
Summary

Single Indivisible Item Auction



Buyer 1 Metropolitan Museum of Arts Buyer 2 Louvre

Second Price Auction



- Metropolitan wins, but pays second highest bid
- The mechanism is DSIC (why?)

Groves Class of Mechanisms

• Allocation rule is efficient:

$$a^*(v) \in \operatorname*{argmax}_{a \in A} \sum_{i \in N} v_i(a)$$

• Payment rule is given by:

$$p_i^*(v_i, v_{-i}) = h_i(v_{-i}) - \sum_{j \in N \setminus \{i\}} v_j(a^*(v)),$$

where $h_i: V_{-i} \to \mathbb{R}$ is any arbitrary function that does not depend on v_i

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Claim

Groves class of mechanisms are DSIC

$$u_i^{(a^*,p^*)}(v_i, v_{-i}) = v_i(a^*(v_i, v_{-i})) - p_i^*(v_i, v_{-i})$$

$$u_i^{(a^*,p^*)}(v_i, v_{-i})$$

= $v_i(a^*(v_i, v_{-i})) - p_i^*(v_i, v_{-i})$
= $v_i(a^*(v_i, v_{-i})) - h_i(v_{-i}) + \sum_{j \in N \setminus \{i\}} v_j(a^*(v_i, v_{-i}))$

$$u_i^{(a^*,p^*)}(v_i, v_{-i}) = v_i(a^*(v_i, v_{-i})) - p_i^*(v_i, v_{-i}) = v_i(a^*(v_i, v_{-i})) - h_i(v_{-i}) + \sum_{j \in N \setminus \{i\}} v_j(a^*(v_i, v_{-i})) = \sum_{i \in N} v_j(a^*(v_i, v_{-i})) - h_i(v_{-i})$$

$$\begin{split} u_i^{(a^*,p^*)}(v_i, v_{-i}) &= v_i(a^*(v_i, v_{-i})) - p_i^*(v_i, v_{-i}) \\ &= v_i(a^*(v_i, v_{-i})) - h_i(v_{-i}) + \sum_{j \in N \setminus \{i\}} v_j(a^*(v_i, v_{-i})) \\ &= \sum_{j \in N} v_j(a^*(v_i, v_{-i})) - h_i(v_{-i}) \\ &\ge \sum_{j \in N} v_j(a^*(v_i', v_{-i})) - h_i(v_{-i}) \text{ (by definition of } a^*) \end{split}$$

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Pivot Mechanism

• A special case of Groves class when the payment is given by:

$$h_i(v_{-i}) = \sum_{j \in N \setminus \{i\}} v_j(a_{-i}^*(v_{-i})),$$

where the allocation $a_{-i}^*(v_{-i})$ is given by:

$$a_{-i}^*(v_{-i}) \in \operatorname*{argmax}_{a \in A} \sum_{j \in N \setminus \{i\}} v_j(a)$$

 $\bullet\,$ The allocation a_{-i}^* maximizes the sum of valuations in the absence of agent i

- The function h_i is the maximum value of this sum
- The payment is therefore:

$$p_i(v_i, v_{-i}) = \max_{a \in A} \sum_{j \in N \setminus \{i\}} v_j(a) - \sum_{j \in N \setminus \{i\}} v_j(a^*(v))$$

Interpretations of the Pivot Mechanism

$$p_i(v_i, v_{-i}) = \max_{a \in A} \sum_{j \in N \setminus \{i\}} v_j(a) - \sum_{j \in N \setminus \{i\}} v_j(a^*(v))$$

Two Interpretations:

- 1. Externality:
 - $\max_{a \in A} \sum_{j \in N \setminus \{i\}} v_j(a)$ is what the agents $N \setminus \{i\}$ can achieve
 - $\sum_{j \in N \setminus \{i\}} v_j(a^*(v))$ is what they achieve under the efficient rule
 - The mechanism asks agent i to pay the difference

Interpretations of the Pivot Mechanism

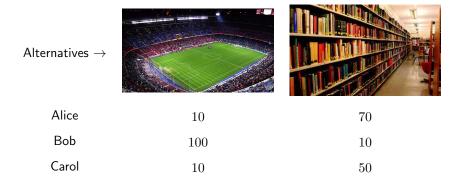
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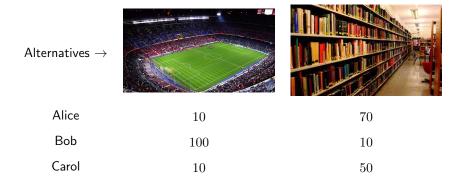
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 - The mechanism asks agent i to pay the difference
- 2. Marginal contribution:
 - Net utility of agent i in pivot mechanism:

$$u_i(v_i, v_{-i}) = \sum_{j \in N} v_j(a^*(v_i, v_{-i})) - \max_{a \in A} \sum_{j \in N \setminus \{i\}} v_j(a)$$

i.e., the difference in sum valuation in presence of agent i and in her absence \blacktriangleright Net utility is agent i's marginal contribution



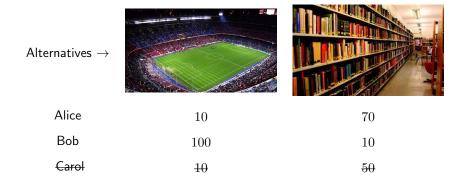


Outcome: L



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• Alice pays (100 + 10) - (10 + 50) = 50



- Outcome: L
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• Carol pays
$$(10 + 100) - (70 + 10) = 30$$



- Outcome: L
- Alice pays (100 + 10) (10 + 50) = 50
- Bob pays (70+50) (70+50) = 0
- Carol pays (10 + 100) (70 + 10) = 30



- Outcome: L
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Summary

Affine Maximizers

An important class of SCFs is that of affine maximizers

Definition (Affine Maximizer)

An SCF $f: V \to A$ is an *affine maximizer* if there exists $w_i \ge 0, i \in N$, not all zero, and a function $\kappa: A \to \mathbb{R}$ such that,

$$f(v) \in \operatorname*{argmax}_{a \in A} \left(\sum_{i \in N} w_i v_i(a) + \kappa(a) \right).$$

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An SCF $f: V \to A$ is an *affine maximizer* if there exists $w_i \ge 0, i \in N$, not all zero, and a function $\kappa: A \to \mathbb{R}$ such that,

$$f(v) \in \underset{a \in A}{\operatorname{argmax}} \left(\sum_{i \in N} w_i v_i(a) + \kappa(a) \right).$$

Special cases:

• $w_i = 1, \forall i \text{ and } \kappa \equiv 0$: efficient SCF

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• $w_i = 1, \forall i \text{ and } \kappa \equiv 0$: efficient SCF

• $w_d = 1$, for some d, $w_i = 0, \forall i \neq d$ and $\kappa \equiv 0$: dictatorial SCF

Affine Maximizers (Contd.)

 An affine maximizer f satisfies independence of irrelevant agents (IIA) if for every i with w_i = 0 and for every v_{-i} ∈ V_{-i},

$$f(v_i, v_{-i}) = f(v'_i, v_{-i}), \forall v_i, v'_i \in V_i$$

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- Every affine maximizer satisfying IIA is implementable
- In particular, payments are of the following form: for all $i \in N$

$$p_i(v_i, v_{-i}) = \begin{cases} \frac{1}{w_i} \left(\sum_{j \neq i} w_j v_j(f(v)) + \kappa(f(v)) + h_i(v_{-i}) \right), & w_i > 0\\ 0 & w_i = 0 \end{cases}$$

f is an affine maximizer

Theorem (Roberts 1979)

Let the allocation space A be finite with $|A| \ge 3$. If the space of valuations V is unrestricted, then an onto and dominant strategy implementable SCF $f: V \to A$ is an affine maximizer.

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Understanding Roberts' Theorem:

- Groves' or pivotal mechanisms are implementable, but this result is giving a necessary condition for implementability
- Moreover, it provides a functional form characterization of the DSIC mechanisms (as opposed to Myerson's monotonicity characterization)
- If payments are enforced to be zero for every valuation profile v, then the only implementable mechanism is dictatorial GS theorem is a corollary of this result

• If an SCF f is implementable in a valuation space V, it is implementable in every valuation space $V' \subseteq V$ - same payments implement them and the number of incentive compatibility constraints reduce

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- Efficient SCF is implementable in any valuation space
- Unrestricted valuation space is crucial for Roberts' theorem some recent results show that the affine maximizer characterization is true even for certain restricted valuation spaces
- Characterization of implementability in restricted domains is an active research area

[A proof by pictures]

Outline of the Talk

The Setup

- Unrestricted Preferences
- Restricted Preferences
- 2 Mechanisms in Quasi-linear Domain
 - Structure of a Mechanism
 - Some Definitions

Results

- Groves Class of Mechanisms
- What Other Mechanisms are Incentive Compatible

• Revenue Equivalence

- Uniqueness of Groves for Efficiency
- Budget Balance
- Bayesian Incentive Compatibility

Summary

Revenue Equivalence

 $\bullet~$ If p~ and p'~ implement f~ in dominant strategies, then

$$p_i(v) = p'_i(v) + \alpha_i(v_{-i}), \forall v \in V$$

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If the type space is convex and the valuations are linear in type, then an SCF, implementable in dominant strategies, satisfies revenue equivalence.

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Theorem (Chung and Olszewski (2007))

Suppose the type space $T \subseteq \mathbb{R}^n$ is a connected set, A is finite and the valuations are continuous in type. If an SCF is implementable in dominant strategies, then it satisfies revenue equivalence.

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Green-Laffont-Holmström Characterization

• An efficient SCF f chooses an alternative in $\operatorname{argmax}_{a \in A} \sum_{j \in N} v_j(a)$

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• An efficient SCF f chooses an alternative in $\operatorname{argmax}_{a \in A} \sum_{j \in N} v_j(a)$

Theorem (Green and Laffont (1979), Holmström (1979))

If the valuation space is convex and smoothly connected, every efficient and DSIC mechanism is a Groves mechanism.

• Shows uniqueness of Groves class in the space of efficient, DSIC mechanisms

[A proof outline]

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Green-Laffont Impossibility

Theorem (Green and Laffont (1979))

No Groves mechanism is budget balanced (BB), i.e., $\nexists p_i^{\text{Groves}} s.t. \sum_{i \in N} p_i^{\text{Groves}}(v) = 0, \forall v \in V.$

• This leads to the following corollary

Corollary

If the valuation space is convex and smoothly connected, no efficient mechanism can be both DSIC and BB.

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Summary

AGV Mechanism

- If the equilibrium condition is relaxed to BIC, we have a positive result
- Payment is defined via a function $\delta_i, i \in N$:

$$\delta_i(v_i) = \mathbb{E}_{v_{-i}|v_i} \left(\sum_{j \in N \setminus \{i\}} v_j(a^*(v)) \right),$$

where a^{\ast} is an efficient allocation

Payment is:

$$p_i^{\mathsf{AGV}}(v) = \sum_{j \in N \setminus \{i\}} \delta_j(v_j) - \delta_i(v_i)$$

Theorem (d'Aspremont and Gerard-Varet (1979), Arrow (1979)) The AGV mechanism is BIC, efficient, and budget-balanced

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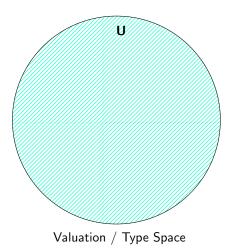
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4 Summary

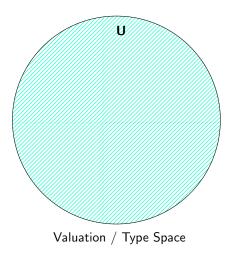
DSIC mechanisms

Valuation / Type Space

DSIC mechanisms

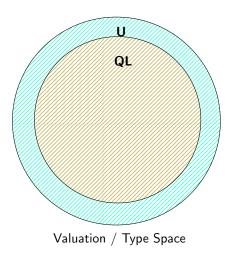


DSIC mechanisms



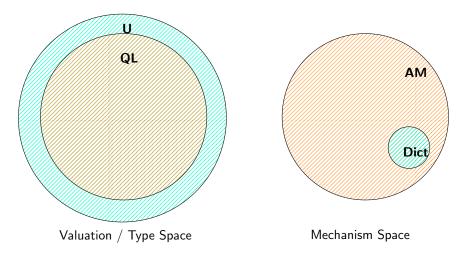


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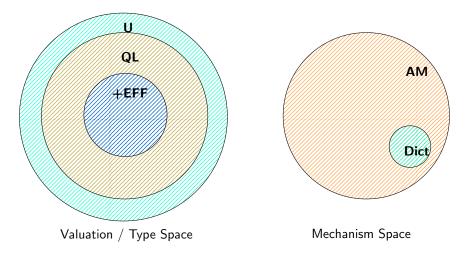




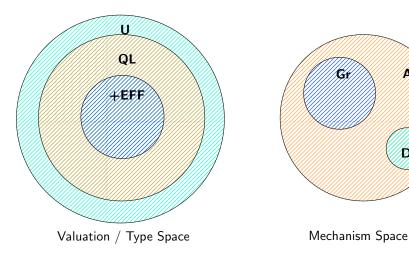
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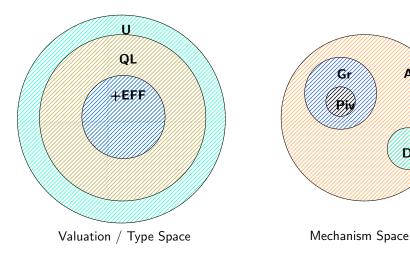
DSIC mechanisms



AM

Dict

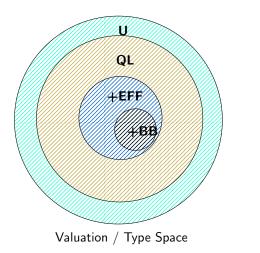
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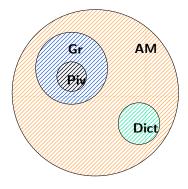


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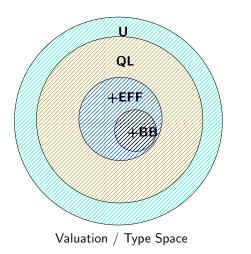
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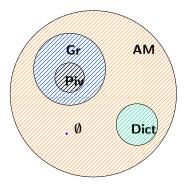
DSIC mechanisms





DSIC mechanisms





Thank you!

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Q: What does affine maximizer mean?

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• Define the value difference set for any pair of distinct alternatives $y, z \in A$.

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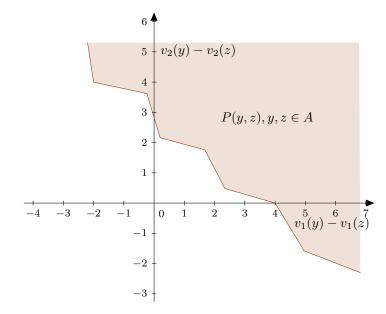
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Claim

If $\alpha \in P(y, z)$, and $\delta > \mathbf{0} \in \mathbb{R}^n$, then $\alpha + \delta \in P(y, z)$, for all distinct $y, z \in A$.

Graphical Illustration for Two Players

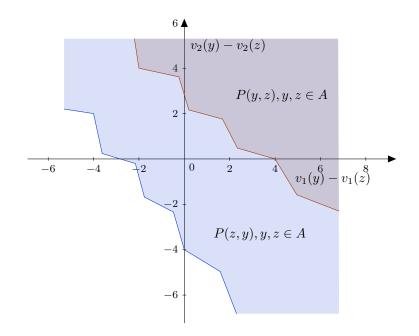


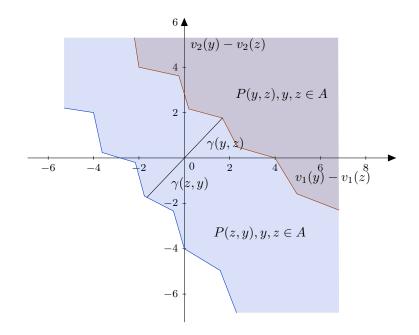
Complementary Structures of P(y, z) and P(z, y)

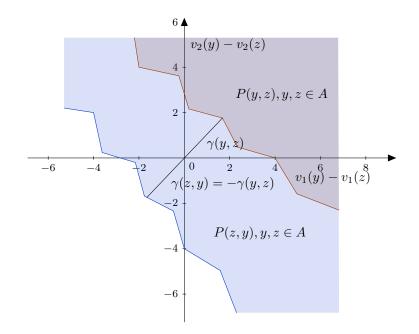
Claim

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For every $\alpha, \epsilon \in \mathbb{R}^n$, $\epsilon > 0$, and for all $y, z \in A$, (a) $\alpha - \epsilon \in P(y, z) \Rightarrow -\alpha \notin P(z, y)$. (b) $\alpha \notin P(y, z) \Rightarrow -\alpha \in P(z, y)$.







Independence of \mathring{C} from the Alternatives in A

- Define the translated set $C(y,z) = P(y,z) \gamma(y,z)\mathbf{1}$
- Denote the 'interior' of C(y,z) by $\mathring{C}(y,z)$

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, for any $y,z,w,l \in A$, $y \neq z$ and $w \neq l$.

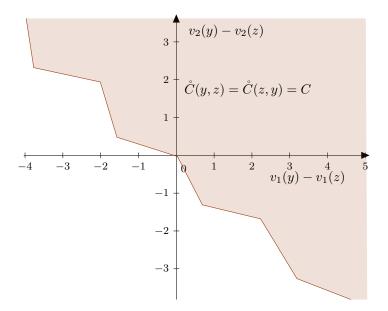
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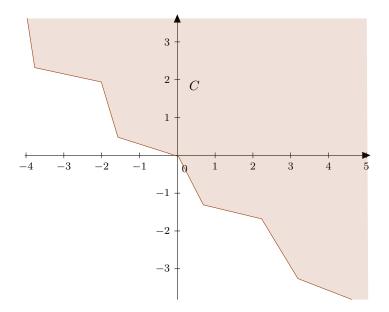
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Remark: Note that this result, in particular, includes the cases, $\mathring{C}(y,z) = \mathring{C}(l,z) = \mathring{C}(l,y) = \mathring{C}(z,y)$. Therefore, the claim holds even without y, z, w, l being all distinct.

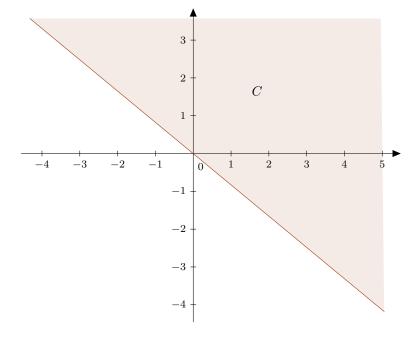




Convexity of \boldsymbol{C}

Claim

The set C is convex.



[BACK]

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- Consider $v_i(a) = v_i^*(a) + \epsilon$, $\epsilon > 0$, and write the DSIC constraint:

$$v_i^*(a) + \epsilon - p_{i,a} \ge v_i(b) - p_{i,b} \tag{1}$$

outcome does not change \Rightarrow payment does not change

• Consider $v_i(a) = v_i^*(a) - \delta$, $\delta > 0$, and similarly:

$$v_i(b) - p_{i,b} \ge v_i^*(a) - \delta - p_{i,a}$$
 (2)

• Combining Equations (1) and (2) and taking limits $\epsilon, \delta \rightarrow 0$, we get,

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• Substituting:

$$p_{i,a} - p_{i,b} = -\left(\sum_{j \in N \setminus \{i\}} v_j(b) - \sum_{j \in N \setminus \{i\}} v_j(a)\right)$$

[BACK]