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Mechanisms with Verification and Fair Allocation Problems

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Based on:

- Mechanisms for Fair Allocation Problems. JAIR 2014
- Structural Tractability of Shapley and Banzhaf Values in Allocation Games. IJCAI 2015
- Fair division rules for funds distribution. Intelligenza Artificiale 2013

See also:

- The Complexity of the Nucleolus in Compact Games. TOCT 2014
- Hypertree Decompositions: Questions and Answers. PODS 2016



Background on Mechanism Design

Mechanisms for Allocation Problems

Complexity Analysis

Case Study

Social Choice Functions



Social Choice Functions



Mechanism Design

- Social Choice Theory is non-strategic
- In practice, agents declare their preferences
 - They are self interested
 - They might not reveal their true preferences
- We want to find optimal outcomes w.r.t. true preferences
- Optimizing w.r.t. the declared preferences might not achieve the goal

How to build a mechanism where agents find convenient to report their true preferences?

Basic Concepts (1/2)

• Each agent i is associated with a **type** $\theta_i \in \Theta_i$

private knowledge, preferences,...



Basic Concepts (2/2)

• Consider the vector of the joint strategies $s = (s_1, \ldots, s_I)$

(A, B, C) ♦ A

• A strategy s_i is **dominant** for agent i, if for every $s'_i \neq s_i$

and for every s_{-i} ,

$$u_i(s_i, s_{-i}, \theta_i) \ge u_i(s'_i, s_{-i}, \theta_i)$$

Independently on the other agents...

Mechanism Design



Mechanism Design



The ideal goal is to build an outcome rule such that truth-telling is a dominant strategy

Impossibility Result

A social choice function is **dictatorial** if one agent always receives one of its most preferred alternatives



Which functions can be implemented in dominant strategies?

Impossibility Result

A social choice function is dictatorial if one agent always receives one of its most preferred alternatives

THEOREM. Assume general preferences, at least two agents, and at least three optimal outcomes. A social choice function can be **implemented in dominant strategies** if, and only if, it is **dictatorial**.

- Very bad news...
 - Gibbard, 1973] and [Satterthwaite, 1975]
- ..., but must be interpreted with care



Which functions can be implemented in dominant strategies?

Payments



Monetary compensation to induce **truthfulness**

A utility is quasi-linear if it has the following form

$$u_i(o, \theta_i) = v_i(o, \theta_i) - p_i$$

valuation function cardinal preferences

payment by the agent

Payments



Monetary compensation to induce **truthfulness**

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valuation function
cardinal preferences

Payments are defined by the mechanism

Payments and Desiderata



Monetary compensation to induce truthfulness

see, e.g., [Shoham, Leyton-Brown; 2009]

Payments and Desiderata



Monetary compensation to induce truthfulness



 $\checkmark\,$ The algebraic sum of the monetary transfers is zero

✓ In particular, mechanisms cannot run into deficit

see, e.g., [Shoham, Leyton-Brown; 2009]

Payments and Desiderata



Monetary compensation to induce truthfulness



✓ The algebraic sum of the monetary transfers is zero
✓ In particular, mechanisms cannot run into deficit





- Monetary compensation to induce fairness
 - ✓ For instance, it is desirable that *no agent envies* the allocation of any another agent, or that
 - ✓ The outcome is *Pareto efficient*, i.e., there is no different allocation such that every agent gets at least the same utility and one of them improves.

(A Few...) Impossibility Results

Efficiency + Truthfulness + Budget Balance

[Green, Laffont; 1977] [Hurwicz; 1975]



Fairness + Truthfulness + Budget Balance

[Tadenuma, Thomson;1995] [Alcalde, Barberà; 1994] [Andersson, Svensson, Ehlers; 2010]



(A Few...) Impossibility Results

Efficiency + Truthfulness + Budget Balance



Fairness + Truthfulness + Budget Balance

Verification on «selected» declarations



(1) Partial Verification

(2) **Probabilistic Verification**

(1) Partial Verification

[Green, Laffont; 1986] [Nisan, Ronen; 2001]

(2) Probabilistic Verification

(1) Partial Verification

[Auletta, De Prisco, Ferrante, Krysta, Parlato, Penna, Persiano, Sorrentino, Ventre]

(2) Probabilistic Verification

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(2) Probabilistic Verification

[Caragiannis, Elkind, Szegedy, Yu; 2012]

(1) Partial Verification

(2) **Probabilistic Verification**

Punishments are used to enforce truthfulness

(1) Partial Verification

(2) Probabilistic Verification

Punishments are used to enforce truthfulness



Verification is performed via sensing

- Hence, it is subject to errors; for instance, because of the limited precision of the measurement instruments.
- It might be problematic to decide whether an observed discrepancy between verified values and declared ones is due to a strategic behavior or to such sensing errors.

[Greco, Scarcello; 2014]





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Approaches to Verification (bis)





- Agents might be uncertain of their private features; for instance, due to limited computational resources
 - There might be no strategic issues

Approaches to Verification (ter)





Punishments enforce truthfulness

- They might be disproportional to the harm done by misreporting
- Inappropriate in real life situations in which uncertainty is inherent due to measurements errors or uncertain inputs.

[Feige, Tennenholtz; 2011]

(1) Partial Verification(2) Probabilistic Verification

Punishments are used to enforce truthfulness

(3) Full Verification

The verifier returns a value.

(1) Partial Verification

(2) **Probabilistic Verification**

Punishments are used to enforce truthfulness

(3) Full Verification

The verifier returns a value. But,...

no punishment

 payments are always computed under the presumption of innocence, where incorrect declared values do not mean manipulation attempts by the agents

error tolerance

 the consequences of errors in the declarations produce a linear "distorting effect" on the various properties of the mechanism



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Case Study



- Goods are indivisible and non-sharable
- Constraints on the maximum number of goods to be allocated to each agent
- Cardinal preferences: Utility functions



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Different agents might have different valuations for the same good



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- Cardinal preferences: *Utility functions*

GOAL: Optimal Allocations

- ✓ Social Welfare
- ✓ Efficiency



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GOAL: Optimal Allocations



✓ Social Welfare✓ Efficiency






Consider an optimal allocation (w.r.t. some declared types)





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The allocation is also optimal for that coalition, even if all goods were actually available

Input: Assumption:	An allocation π for $\langle \mathcal{A}, G, \omega \rangle$, and a vector $\mathbf{w} \in \mathbf{D}$; A verifier \mathbf{v} is available. Let $\mathbf{v}(\pi) = (v_1,, v_n)$;
1. Let \mathbb{C} denote the set of all possible subsets of \mathcal{A} ;	
2. For each set $\mathcal{C} \in \mathbb{C}$,	
3. Compute an optimal allocation $\pi_{\mathcal{C}}$ for $\langle \mathcal{C}, \operatorname{img}(\pi), \omega \rangle$ w.r.t. w;	
4. For each agent $i \in \mathcal{A}$,	
5. For each set $\mathcal{C} \in \mathbb{C}$,	
6. Let	t $\Delta^1_{\mathcal{C},i}(\pi, \mathbf{w}) := \operatorname{val}(\pi_{\mathcal{C}}, (v_i, \mathbf{w}_{-i})); \qquad (=v_i(\pi_{\mathcal{C}}) + \sum_{j \in \mathcal{C} \setminus \{i\}} w_j(\pi_{\mathcal{C}}));$
7. L Le	t $\Delta^2_{\mathcal{C},i}(\pi, \mathbf{w}) := \operatorname{val}(\pi_{\mathcal{C}\setminus\{i\}}, \mathbf{w}); \qquad (=\sum_{j\in\mathcal{C}\setminus\{i\}} w_j(\pi_{\mathcal{C}\setminus\{i\}}));$
8. Let ξ_i	$(\pi, \mathbf{w}) := \sum_{\mathcal{C} \in \mathbb{C}} \frac{(\mathcal{A} - \mathcal{C})! (\mathcal{C} - 1)!}{ \mathcal{A} !} (\Delta^1_{\mathcal{C}, i}(\pi, \mathbf{w}) - \Delta^2_{\mathcal{C}, i}(\pi, \mathbf{w}));$
9. L Define	$e p_i^{\xi}(\pi, \mathbf{w}) := \xi_i(\pi, \mathbf{w}) - v_i(\pi);$







By the previous lemma, this is without loss of generality. In fact, allocated goods are the only ones that we verify.



«Bonus and Compensation», by Nisan and Ronen (2001)



«Bonus and Compensation», by Nisan and Ronen (2001)



No punishments!



«Bonus and Compensation», by Nisan and Ronen (2001)

Truth-telling is a dominant strategy for each agent



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«Bonus and Compensation», by Nisan and Ronen (2001)

Truth-telling is a dominant strategy for each agent

Coalitional Games

- Players form coalitions
- Each coalition is associated with a worth
- A *total worth* has to be distributed

$$\mathcal{G} = \langle \mathbf{N}, \varphi \rangle, \ \varphi \colon \mathbf{2}^{\mathbf{N}} \mapsto \mathbb{R}$$



Solution Concepts characterize outcomes in terms of

- Fairness
- Stability

Coalitional Games: Shapley Value

$$\phi_i(\mathcal{G}) = \sum_{C \subseteq N} \frac{(|N| - |C|)!(|C| - 1)!}{|N|!} (\varphi(C) - \varphi(C \setminus \{i\}))$$

Solution Concepts characterize outcomes in terms of

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Relevant Properties of the Shapley Value

(I) $\sum_{i \in N} \phi_i(\mathcal{G}) = \varphi(N);$

(II) If φ is supermodular (resp., submodular), then $\sum_{i \in R} \phi_i(\mathcal{G}) \geq \varphi(R)$ (resp., $\sum_{i \in R} \phi_i(\mathcal{G}) \leq \varphi(R)$), for each coalition $R \subseteq N$.

(III) If $\mathcal{G}' = \langle N, \varphi' \rangle$ is a game such that $\varphi'(R) \ge \varphi(R)$, for each $R \subseteq l$ then $\phi_i(\mathcal{G}') \ge \phi_i(\mathcal{G})$, for each agent $i \in N$.

Core Allocation

 $\varphi(R \cup T) + \varphi(R \cap T) \ge \varphi(R) + \varphi(T) \text{ (resp., } \varphi(R \cup T) + \varphi(R \cap T) \le \varphi(R) + \varphi(T))$

$$\mathcal{G} = \langle \mathbf{N}, \varphi \rangle, \ \varphi \colon \mathbf{2}^{\mathbf{N}} \mapsto \mathbb{R}$$

• $\varphi(C)$ is the *contribution* of the coalition **w.r.t.**

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• $\varphi(C)$ is the *contribution* of the coalition w.r.t.
$$\begin{cases} \text{selected products} \\ and \\ verified values \end{cases}$$

Best possible allocation, assuming that agents in C are the only ones in the game

$$\mathcal{G} = \langle \mathbf{N}, \varphi \rangle, \ \varphi \colon \mathbf{2}^{\mathbf{N}} \mapsto \mathbb{R}$$

• $\varphi(C)$ is the *contribution* of the coalition **w.r.t.**

selected products and verified values (π)

Each agent gets the Shapley value

 $\phi_i(\mathcal{G})$





The resulting mechanism is «fair» and «buget balanced»

Properties



The resulting mechanism is «fair» and «buget balanced»

 $\sum_{i \in N} \phi_i(\mathcal{G}) = \varphi(N)$

Properties



The resulting mechanism is «fair» and «buget balanced»

The game is supermodular; so the Shapley value is stable

- Let π be an optimal allocation
- Let π' be an allocation

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(best allocation for the coalition with products in π)

As π is optimal, then $\varphi(C)$ is in fact optimal even by considering all possible products as available



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 $\pi \ge \pi'$

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 There is no difference between two different optimal allocations

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Complexity Issues

- For many classes of «compact games» (e.g., graph games), the Shapley-value can be efficiently calculated
- Here, the problem emerges to be #P-complete

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- #P is the class the class of all functions that can be computed by counting Turing machines in polynomial time.
- A counting Turing machine is a standard nondeterministic Turing machine with an auxiliary output device that prints in binary notation the number of accepting computations induced by the input.
- Prototypical problem: to count the number of truth variable assignments that satisfy a Boolean formula.

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Reduction from the problem of counting the number of perfect matchings in certain bipartite graphs [Valiant, 1979]

- #P is the class the class of all functions that can be computed by *counting Turing machines* in polynomial time.
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Complexity Issues

- #P-complete
- However...



Probabilistic Computation

- #P-complete
- However...



Fully Polynomial-Time Randomized Approximation Scheme

- Always Efficient and Budget Balanced
- All other properties in expectation (with high probability)



Coupling of the algorithm with a sampling strategy for the coalitions by [Liben-Nowell,Sharp, Wexler, Woods; 2012]

Probabilistic Computation

Input:An allocation π for $\langle \mathcal{A}, G, \omega \rangle$, and a vector $\mathbf{w} \in \mathbf{D}$;Assumption:A verifier \mathbf{v} is available. Let $\mathbf{v}(\pi) = (v_1,, v_n)$;	
1. Le	et \mathbb{C} denote the set of all possible subsets of \mathcal{A} ;
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7.	Let $\Delta_{\mathcal{C},i}^{2}(\pi, \mathbf{w}) := \operatorname{val}(\pi_{\mathcal{C}\setminus\{i\}}, \mathbf{w}); \qquad (=\sum_{j\in\mathcal{C}\setminus\{i\}} w_j(\pi_{\mathcal{C}\setminus\{i\}}));$
8.	Let $\xi_i(\pi, \mathbf{w}) := \sum_{\mathcal{C} \in \mathbb{C}} \frac{(\mathcal{A} - \mathcal{C})! (\mathcal{C} - 1)!}{ \mathcal{A} !} (\Delta^1_{\mathcal{C}, i}(\pi, \mathbf{w}) - \Delta^2_{\mathcal{C}, i}(\pi, \mathbf{w}));$
9. L	Define $p_i^{\xi}(\pi, \mathbf{w}) := \xi_i(\pi, \mathbf{w}) - v_i(\pi);$

Use sampling, rather than exaustive search.



Coupling of the algorithm with a sampling strategy for the coalitions by [Liben-Nowell,Sharp, Wexler, Woods; 2012]

Back to Exact Computation: Islands of Tractability



Can we find classes of instances for «allocation games» over which the Shapley value can be efficiently computed?

Bounded Sharing Degree



- Sharing degree
 - Maximum number of agents competing for the same good

Bounded Sharing Degree



- Sharing degree
 - Maximum number of agents competing for the same good

The Shapley value can be computed in polynomial time whenever the sharing degree is 2 at most.



Bounded Interactions

Bounded Interactions



- Interaction graph
 - There is an edge between any pair of agents competing for the same good

Bounded Interactions



- Interaction graph
 - There is an edge between any pair of agents competing for the same good

The Shapley value can be computed in polynomial time whenever the interaction graph is a tree.

or, more generally, if it has bounded treewidth





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Case Study: Italian Research Assessment Program

- VQR: ANVUR should evaluate the quality of research of all Italian research structures
- Funds for the structures in the next years depend on the outcome of this evaluation
- Substructures will be also evaluated (departments)







Structures are in charge of selecting the products to submit

Constraints (2004-2010)

- Every researcher has to submit 3 publications
- A publication cannot be allocated to two researchers





Co-Autorships at University of Calabria



Number of publications

Co-Autorships at University of Calabria



Components at University of Calabria



Elements in each component

An Example Component









ssues

- Allocation Problem
- Valuations are declared (punishments?)
- The program is meant to evaluate the structures...
 - ...but outcomes are used to evaluate researchers, too

Global Evaluation



A Closer Look



A Closer Look



Unless it is clear that no penalization will occur, Subscription will act «strategically»



Distribution at University of Calabria



The Story....

- ANVUR did not specify a division rule
- Reserchers considered projas «the rule»
- Researchers submitted (rated) only the minimum number of publications required (by default 3), thus implicitly under-estimating all their other products
- To avoid overlapping submissions, «agreements» have been made



Conflicts resolved «strategically», «hierarchically», …

The optimum has been missed! No fairness at all!

Side Results

- University of Rome uses (parts of) our findings
- University of Calabria uses (parts of) our findings
- Head of the «Presidio della Qualità» at University of Calabria
- Still trying to generalize at national level....

