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Implementation For Perfectly Informed Players**  
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# Perfect and General Virtual Implementation For Perfectly Informed Players

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## Abstract

We show that, when the players are perfectly informed about each other, essentially all social-choice functions can be rationally robustly implemented via an extensive-form public-action mechanism that (1) is perfectly robust against collusion, (2) requires only a linear number of computation steps and communication bits, and (3) preserves the privacy of the players' types to a very high extent.

## 1 Notations

- The set  $N = \{1, \dots, n\}$  is the set of players;
- For each player  $i$ ,  $\Gamma_i$  is the space of types of  $i$ , and  $\Gamma = \prod_i \Gamma_i$  is the space of type profiles;
- The outcome space is  $\Omega$ ;
- A perfect-knowledge context is a 4-tuple  $C = (T, \Omega, u, \mathbb{C})$ , where
  - $T \in \Gamma$  is the profile of the players true types;
  - each  $u_i$  is the utility function for player  $i$ , mapping player  $i$ 's true type  $T_i$  and a state  $\omega$  to a real number;
  - $\mathbb{C}$  is the collusion structure, namely, a partition of  $N$ .

Each subset  $S \in \mathbb{C}$  of players is called an agent.  $S$ , as well as each player in  $S$ , is collusive if  $|S| > 1$ , or independent otherwise.  $S$  is also called a collusive set if  $|S| > 1$ .  $C$  is called an independent context if every player is independent, or a collusive context otherwise. When  $C$  is an independent context,  $\mathbb{C}$  is typically omitted for simplicity.

$\Omega$  is publicly known to everybody, including the mechanism designer.  $T$ ,  $u$ , and  $\mathbb{C}$  are publicly known to the players, and we emphasize that the mechanism designer knows nothing about them.

- A function  $f : \Gamma \times \Omega \rightarrow \mathbb{R}$  is called a (deterministic) numerical social choice function. Note a (deterministic) traditional social-choice function  $F$  maps type profiles to outcomes,  $F : \Gamma \rightarrow \Delta(\Omega)$ , and, given a context  $C$  with true-type profile  $T$ , the goal of a mechanism is to produce the outcome  $F(T)$ . But it is clear that this goal can be achieved by considering the numerical social-choice function  $f$  that, on input  $(\gamma, \omega)$  returns 1 if  $\omega = F(\gamma)$  and 0 otherwise, and maximizing  $f(T, \cdot)$ .

Our assumptions about contexts  $C$  and social choice functions  $f$  are the same as [5], namely:

- (1) For each player  $i$ ,  $u_i$  is not a constant function, and for any two different types  $\gamma_i, \gamma'_i \in \Gamma_i$ ,  $u_i(\gamma_i, \cdot)$  is not a linear transformation of  $u_i(\gamma'_i, \cdot)$ .
- (2) For each player  $i$ , there exists a function  $g_i : \Gamma_i \rightarrow \Omega$  such that for any two different types  $\gamma_i, \gamma'_i \in \Gamma_i$ ,  $u_i(\gamma_i, f_i(\gamma_i)) > u_i(\gamma_i, f_i(\gamma'_i))$ .

(3) A mechanism  $M$  specifies the players' strategies, and the outcome function, denoted also by  $M$ . Given a strategy profile  $\sigma$ ,  $M(\sigma)$  consists of an outcome  $\omega \in \Omega$  and a price profile  $P \in \mathbb{R}^n$  — that is,  $M$  can impose a price to each player. Given  $C$  and  $(\omega, P)$ , each player  $i$ 's utility is  $U_i(T_i, \omega, P_i) = u_i(T_i, \omega) - P_i$ . We use the solution concept proposed in [3, 4], namely, rationally robust implementation. Our mechanism is based on that in [2], in particular, it is an extensive-form mechanism with public actions. The setting considered is finitely, in particular: the type space is finite, the outcome space is finite, and the image spaces of the social choice function  $f$  and  $g_i$ 's for each  $i$  are all finite.

## 2 Our Mechanism

**Notation** In the mechanism below,

- $\epsilon$  and  $\epsilon_j^i$ , for  $i \in \{2, \dots, n\}$  and  $j \in \{1, \dots, n\}$ , are constants such that  $\frac{1}{5n} > \epsilon > \epsilon_1^2 > \dots > \epsilon_n^2 > \epsilon_1^3 > \dots > \epsilon_n^3 > \dots > \epsilon_1^n > \dots > \epsilon_n^n > 0$ .
- Numbered steps are taken by the players, while steps marked by letters are taken by the mechanism.
- Sentences between quotation marks are comments, and could be excised if no clarification is needed.
- We denote by  $n_r$  the number of outcomes  $(\omega, P)$  with revenue  $r$ . For all such outcomes, we denote by  $0 \leq f_r(\omega, P) < n_r$  the rank of the outcome  $(\omega, P)$  in the lexicographic order that first considers the state and then the price profile (where  $P_1, \dots, P_n$  precedes  $P'_1 \dots P'_n$  whenever  $P_1 > P'_1$ , etc.).
- $U = \max_{T \in \Gamma, \omega \in \Omega} f(T, \omega)$ , where  $f$  is the given numerical social-choice function.
- $L = \min_i \min_{T_i, T'_i \in \Gamma_i} u_i(T_i, g_i(T_i)) - u_i(T_i, g_i(T'_i))$ .
- $\epsilon'$  is a positive constant such that  $\epsilon' < \frac{L}{2nU}$ .

### Mechanism $\mathcal{M}$

(1) *Player 1 announces a state  $\omega^*$  and a profile  $K^1$  of natural numbers.*

“( $\omega^*, K^1$ ) is player 1’s proposed outcome, allegedly an outcome of maximum revenue.”

(a) *Set  $\omega = \perp$ , and  $P_i = 0 \forall i$ . If  $\sum_i K_i^1 = 0$ , the mechanism ends right now. Otherwise, proceed to Step 2.*

“Whenever the mechanism ends,  $\omega$  and  $P$  will be, respectively, the final state and price profile.”

(2, ..., n) *In Step  $i$ ,  $2 \leq i \leq n$ , player  $i$  publicly announces a profile  $\Delta^i$  of natural numbers such that  $\Delta_i^i = 0$ .*

“By so doing  $i$  suggests to raise the current price of  $j$ , that is  $K_j^1 + \sum_{\ell=2}^{i-1} \Delta_j^\ell$ , by the amount  $\Delta_j^i$ .”

(b) *For each player  $i$ , publicly select  $bip_i$  and  $P_i^*$  as follows. Let  $R_i = \{j : \Delta_i^j > 0\}$ .*

*If  $R_i \neq \emptyset$ , then  $bip_i$  is highest player in  $R_i$ , and  $P_i^* = K_i^1 + \sum_{\ell=2}^{bip_i} \Delta_i^\ell$ . Else,  $bip_i = 1$  and  $P_i^* = K_i^1$ .*

“We refer to  $bip_i$  as the best informed player about  $i$ , and to  $P_i^*$  as the provisional price of  $i$ .”

(n + 1) *Each player  $i$  such that  $P_i^* > 0$  simultaneously announces YES or NO.*

*By default, each player  $i$  such that  $P_i^* = 0$  announces YES, and player 1 announces YES if  $bip_1 = 1$ .*

“Each player  $i$  announces YES or NO to  $\omega^*$  as the final state and to  $P_i^* - \epsilon$  as his own price. (By default player 1 accepts his own price if no one raises it.) At this point the players are done playing, and the mechanism proceeds as follows. If all say YES, the updated proposal  $(\omega^*, P^*)$  is implemented with probability 1. Else:

- With very high probability the null outcome is chosen, except that the best-informed players of those saying NO are punished.

- With small probability the null outcome is chosen
- With very small probability, proportional to the number of players saying YES, we implement  $(\omega^*, P^*)$  as if all said YES.

Importantly, as we shall see, all get a small reward at the end for their knowledge.”

- (b') *Publicly flip a biased coin  $c_0$  such that  $\Pr[c_0 = \text{Heads}] = 1 - \epsilon$ . If  $c_0 = \text{Heads}$  then proceed to Step c. Otherwise do the following:*
- (n+2) *Each player  $i$  simultaneously announces a type  $\gamma_i \in \Gamma_i$ .*
- (c') *If  $\omega^* = \operatorname{argmax}_{\omega' \in \Omega} f(\gamma, \omega')$ , then let  $v = f(\gamma, \omega)$  and reset  $P_1$  to  $P_1 - \epsilon'v$ ; otherwise reset  $P_1$  to  $\frac{L}{2n}$ .*
- (d') *Choose a random player  $i$  uniformly, reset  $\omega$  to be  $g_i(\gamma_i)$ , and HALT.*
- (c) *Let  $Y$  be the number of players announcing YES. If  $Y = n$ , then reset  $\omega$  to  $\omega^*$  and each  $P_i$  to  $P_i^* - \epsilon$ , and go to Step g. If  $Y < n$ , proceed to Step d.*
- (d) *Publicly flip a biased coin  $c_1$  such that  $\Pr[c_1 = \text{Heads}] = 1 - \epsilon$ .*
- (e) *If  $c_1 = \text{Heads}$ , reset  $P_{\text{bip}_i}$  to  $P_{\text{bip}_i} + 2P_i^*$  for each player  $i$  announcing NO.*
- (f) *If  $c_1 = \text{Tails}$ , letting  $B = \sum_{i \text{ announces NO}} P_i^*$ , flip a biased coin  $c_2$  such that  $\Pr[c_2 = \text{Heads}] = \frac{Y}{nB}$ .  
If  $c_2 = \text{Heads}$ , reset  $\omega$  to  $\omega^*$  and each  $P_i$  to  $P_i^* - \epsilon$ .  
If  $c_2 = \text{Tails}$ ,  $\omega$  and  $P$  continue to be  $\perp$  and  $0^n$ .*
- (g) *Reset  $P_1$  to  $P_1 - \epsilon - 2\epsilon \sum_j K_j^1 + \epsilon \frac{f_r(\omega^*)}{n_r}$  and each other  $P_i$  to  $P_i - \epsilon - \sum_j \epsilon_j^i \Delta_j^i$ .*

“Although players’ prices may be negative, we prove that the mechanism never loses money, and that in the unique rational play the utility of every player is non-negative. For clarity, our rewards are proportional to prices and raises.”

### 3 Statement of Our Theorem

**Theorem 1.** *For every numerical social choice function  $f$ , every context  $C = (T, \Omega, u, \mathbb{C})$ , and every rationally robust play  $\sigma$  of  $(C, \mathcal{M})$ , letting  $(\omega, P) = \mathcal{M}(\sigma)$ , then, with probability  $\geq 1 - \epsilon$ :*

$$\omega = \operatorname{argmax}_{\omega' \in \Omega} f(T, \omega').$$

In a forthcoming paper we shall extend the above mechanism in a simple way (e.g., by introducing envelopes to the mechanism) to (virtually) implement also all probabilistic numerical (and thus traditional) social choice functions.

### References

- [1] D. Abreu and H. Matsushima. Virtual Implementation in Iteratively Undominated Strategies: Complete Information. *Econometrica*, Vol. 60, No. 5, pages 993-1008, Sep., 1992.
- [2] J. Chen, A. Hassidim, and S. Micali. Robust Perfect Revenue From Perfectly Informed Players. To appear at ICS'10.
- [3] J. Chen and S. Micali. A New Approach to Auctions and Resilient Mechanism Design. *Symposium on Theory of Computing*, pages 503-512, 2009. Full version available at [http://people.csail.mit.edu/silvio/Selected\\_Scientific\\_Papers/Mechanism\\_Design/](http://people.csail.mit.edu/silvio/Selected_Scientific_Papers/Mechanism_Design/).

- [4] J. Chen and S. Micali. Rational Robustness for Mechanism Design. Submitted to STOC'10, full version available at [http://people.csail.mit.edu/silvio/Selected Scientific Papers/Mechanism Design/](http://people.csail.mit.edu/silvio/Selected_Scientific_Papers/Mechanism_Design/).
- [5] J. Glazer and M. Perry. Virtual Implementation in Backwards Induction. *Games and Economic Behavior*, Vo.15, pages 27-32, 1996.

