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11 September 2016

Online at <https://mpra.ub.uni-muenchen.de/73660/>  
MPRA Paper No. 73660, posted 12 Sep 2016 11:13 UTC

# A Bayesian implementable social choice function cannot be implemented by a direct mechanism

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## Abstract

The revelation principle is a fundamental theorem in many economics fields such as game theory, mechanism design and auction theory etc. In this paper, I construct an example to show that a social choice function which can be implemented in Bayesian Nash equilibrium by an indirect mechanism cannot be implemented by a direct mechanism. The key point is that agents pay cost in the indirect mechanism, but pay nothing in the direct mechanism. As a result, the revelation principle does not hold at all.

JEL codes: D70

*Key words:* Revelation principle; Game theory; Mechanism design; Auction theory.

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## 1 Introduction

The revelation principle plays an important role in microeconomics theory and has been applied to many other fields such as auction theory, game theory etc. According to the wide-spread textbook given by Mas-Colell, Whinston and Green (Page 884, Line 24 [1]): “*The implication of the revelation principle is ... to identify the set of implementable social choice functions, we need only identify those that are truthfully implementable.*” Related definitions about the revelation principle can be seen in Appendix, which are cited from Section 23.B and 23.D of MWG’s textbook[1].

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However, in this paper, I will construct a simple labor model to show the revelation principle does not hold at all. Section 2 is the main part of this paper, and Section 3 draws conclusions.

## 2 A labor model

Here we consider a simple labor model which uses some ideas from the first-price sealed auction model in Example 23.B.5 [1] and the signaling model in Section 13.C [1]. There are one firm and two workers. The firm wants to hire a worker, and two workers compete for this job offer. Worker 1 and Worker 2 differ in the number of units of output they produce if hired by the firm, which is denoted by productivity type.

For simplicity, we make the following assumptions:

- 1) The possible productivity types of two workers are:  $\theta_1, \theta_2 \in \{\theta_L, \theta_H\}$ , where  $\theta_H > \theta_L > 0$ . Each worker's productivity is a random variable chosen independently, and is private information for each worker.
- 2) Before confronting the firm, each worker  $i = 1, 2$  can get some education. The possible levels of education are:  $e_L$  and  $e_H$ , where  $e_H > e_L > 0$ . Each worker's education is observable to the firm. Education does nothing for a worker's productivity.
- 3) The cost of obtaining education level  $e$  for a worker of some type  $\theta$  is given by a function  $c(e, \theta) = e/\theta$ . That is, the cost of education is lower for a high-productivity worker.

The model's outcome can be represented by a vector  $(y_1, y_2)$ , where  $y_i$  denotes the probability that worker  $i$  gets the job offer. Recall that the firm does not know the exact productivity types of two workers, but its aim is to hire a worker with productivity as high as possible. This aim can be represented by a social choice function  $f(\vec{\theta}) = (y_1(\vec{\theta}), y_2(\vec{\theta}))$ , in which  $\vec{\theta} = (\theta_1, \theta_2)$ ,

$$y_1(\vec{\theta}) = \begin{cases} 1, & \text{if } \theta_1 > \theta_2 \\ 0.5, & \text{if } \theta_1 = \theta_2 \\ 0, & \text{if } \theta_1 < \theta_2 \end{cases}, y_2(\vec{\theta}) = \begin{cases} 1, & \text{if } \theta_1 < \theta_2 \\ 0.5, & \text{if } \theta_1 = \theta_2 \\ 0, & \text{if } \theta_1 > \theta_2 \end{cases} \quad (1)$$

In order to implement the above  $f(\vec{\theta})$ , the firm designs an indirect mechanism  $\Gamma = (S_1, S_2, g)$  as follows:

- 1) A random move of nature determines the productivity of workers:  $\theta_1, \theta_2 \in \{\theta_L, \theta_H\}$ .
- 2) Conditional on his type  $\theta_i$ , each worker  $i = 1, 2$  chooses his education level as a bid  $b_i : \{\theta_L, \theta_H\} \rightarrow \{e_L, e_H\}$ . The strategy set  $S_i$  is the set of all possible

bids  $b_i(\theta_i)$ , and the outcome function  $g$  is defined as:

$$g(b_1, b_2) = (p_1, p_2) = \begin{cases} (1, 0), & \text{if } b_1 > b_2 \\ (0.5, 0.5), & \text{if } b_1 = b_2 \\ (0, 1), & \text{if } b_1 < b_2 \end{cases} \quad (2)$$

where  $(1, 0)$  means worker 1 gets the offer,  $(0, 1)$  means worker 2 gets the offer;  $(0.5, 0.5)$  means that each worker has the same probability 0.5 to be the winner.

Let  $u_0$  be the utility of the firm, and  $u_1, u_2$  be the utilities of worker 1, 2 respectively, then  $u_0(b_1, b_2) = p_1\theta_1 + p_2\theta_2 - w$ , and for  $i, j = 1, 2, i \neq j$ ,

$$u_i(b_i, b_j; \theta_i) = \begin{cases} w - b_i/\theta_i, & \text{if } b_i > b_j \\ 0.5w - b_i/\theta_i, & \text{if } b_i = b_j \\ -b_i/\theta_i, & \text{if } b_i < b_j \end{cases} \quad (3)$$

**Proposition 1:** The social choice function  $f(\vec{\theta})$  can be implemented by the indirect mechanism  $\Gamma$  in Bayesian Nash equilibrium.

**Proof:** Consider a separating strategy, *i.e.*, workers with different productivity types choose different education levels,

$$b_1(\theta_1) = \begin{cases} e_H, & \text{if } \theta_1 = \theta_H \\ e_L, & \text{if } \theta_1 = \theta_L \end{cases}, b_2(\theta_2) = \begin{cases} e_H, & \text{if } \theta_2 = \theta_H \\ e_L, & \text{if } \theta_2 = \theta_L \end{cases}. \quad (4)$$

Now let us check whether this separating strategy yields a Bayesian Nash equilibrium. Assume  $b_j^*(\theta_j)$  takes this form, *i.e.*,

$$b_j^*(\theta_j) = \begin{cases} e_H, & \text{if } \theta_j = \theta_H \\ e_L, & \text{if } \theta_j = \theta_L \end{cases}, \quad (5)$$

then consider worker  $i$ 's problem ( $i \neq j$ ). For each  $\theta_i \in \{\theta_L, \theta_H\}$ , worker  $i$  solves the maximization problem  $\max_{b_i} h(b_i, \theta_i)$ , where by Eq (3) the object function is

$$h(b_i, \theta_i) = (w - b_i/\theta_i)P(b_i > b_j^*(\theta_j)) + (0.5w - b_i/\theta_i)P(b_i = b_j^*(\theta_j)) - (b_i/\theta_i)P(b_i < b_j^*(\theta_j)) \quad (6)$$

We discuss this maximization problem in four different cases:

1) Suppose  $\theta_i = \theta_j = \theta_L$ , then  $b_j^*(\theta_j) = e_L$  by Eq (5).

$$\begin{aligned} h(b_i, \theta_i) &= (w - b_i/\theta_L)P(b_i > e_L) + (0.5w - b_i/\theta_L)P(b_i = e_L) - (b_i/\theta_L)P(b_i < e_L) \\ &= \begin{cases} w - e_H/\theta_L, & \text{if } b_i = e_H \\ 0.5w - e_L/\theta_L, & \text{if } b_i = e_L \end{cases} \end{aligned}$$

Thus, if  $w < 2(e_H - e_L)/\theta_L$ , then  $h(e_H, \theta_i) < h(e_L, \theta_i)$ , which means the optimal value of  $b_i(\theta_i)$  is  $e_L$ . In this case,  $b_i^*(\theta_L) = e_L$ .

2) Suppose  $\theta_i = \theta_L, \theta_j = \theta_H$ , then  $b_j^*(\theta_j) = e_H$  by Eq (5).

$$\begin{aligned} h(b_i, \theta_i) &= (w - b_i/\theta_L)P(b_i > e_H) + (0.5w - b_i/\theta_L)P(b_i = e_H) - (b_i/\theta_L)P(b_i < e_H) \\ &= \begin{cases} 0.5w - e_H/\theta_L, & \text{if } b_i = e_H \\ -e_L/\theta_L, & \text{if } b_i = e_L \end{cases} \end{aligned}$$

Thus, if  $w < 2(e_H - e_L)/\theta_L$ , then  $h(e_H, \theta_i) < h(e_L, \theta_i)$ , which means the optimal value of  $b_i(\theta_i)$  is  $e_L$ . In this case,  $b_i^*(\theta_L) = e_L$ .

3) Suppose  $\theta_i = \theta_H, \theta_j = \theta_L$ , then  $b_j^*(\theta_j) = e_L$  by Eq (5).

$$\begin{aligned} h(b_i, \theta_i) &= (w - b_i/\theta_H)P(b_i > e_L) + (0.5w - b_i/\theta_H)P(b_i = e_L) - (b_i/\theta_H)P(b_i < e_L) \\ &= \begin{cases} w - e_H/\theta_H, & \text{if } b_i = e_H \\ 0.5w - e_L/\theta_H, & \text{if } b_i = e_L \end{cases} \end{aligned}$$

Thus, if  $w > 2(e_H - e_L)/\theta_H$ , then  $h(e_H, \theta_i) > h(e_L, \theta_i)$ , which means the optimal value of  $b_i(\theta_i)$  is  $e_H$ . In this case,  $b_i^*(\theta_H) = e_H$ .

4) Suppose  $\theta_i = \theta_j = \theta_H$ , then  $b_j^*(\theta_j) = e_H$  by Eq (5).

$$\begin{aligned} h(b_i, \theta_i) &= (w - b_i/\theta_H)P(b_i > e_H) + (0.5w - b_i/\theta_H)P(b_i = e_H) - (b_i/\theta_H)P(b_i < e_H) \\ &= \begin{cases} 0.5w - e_H/\theta_H, & \text{if } b_i = e_H \\ -e_L/\theta_H, & \text{if } b_i = e_L \end{cases} \end{aligned}$$

Thus, if  $w > 2(e_H - e_L)/\theta_H$ , then  $h(e_H, \theta_i) > h(e_L, \theta_i)$ , which means the optimal value of  $b_i(\theta_i)$  is  $e_H$ . In this case,  $b_i^*(\theta_H) = e_H$ .

In summary, if  $w \in [2(e_H - e_L)/\theta_H, 2(e_H - e_L)/\theta_L]$ , then the the strategy  $b_i^*(\theta_i)$  of worker  $i$

$$b_i^*(\theta_i) = \begin{cases} e_H, & \text{if } \theta_i = \theta_H \\ e_L, & \text{if } \theta_i = \theta_L \end{cases} \quad (7)$$

is the optimal response to the strategy  $b_j^*(\theta_j)$  of worker  $j$  ( $j \neq i$ ) given in Eq (5). Therefore, the strategy combination  $(b_1^*(\theta_1), b_2^*(\theta_2))$  is a Bayesian Nash equilibrium of the game induced by  $\Gamma$ .

By Eq(2) and Eq(7), for any  $\vec{\theta} = (\theta_1, \theta_2)$ ,  $\theta_1, \theta_2 \in \{\theta_L, \theta_H\}$ ,

$$g(b_1^*(\theta_1), b_2^*(\theta_2)) = \begin{cases} (1, 0), & \text{if } \theta_1 > \theta_2 \\ (0.5, 0.5), & \text{if } \theta_1 = \theta_2 \\ (0, 1), & \text{if } \theta_1 < \theta_2 \end{cases} \quad (8)$$

which implements the social choice function  $f(\vec{\theta})$ . **Q.E.D**

**Proposition 2:** The social choice function  $f(\vec{\theta})$  cannot be implemented truthfully in a direct mechanism.

**Proof:** Consider a direct mechanism  $\Gamma_{direct}$ , the timing steps of which is as follows:

- 1) Each worker  $i = 1, 2$  announces his productivity type  $\hat{\theta}_i \in \{\theta_L, \theta_H\}$  to a virtual mediator.
- 2) The mediator submits  $b_i^*(\hat{\theta}_i)$  ( $i = 1, 2$ ) to the firm:

$$b_i^*(\hat{\theta}_i) = \begin{cases} e_H, & \text{if } \hat{\theta}_i = \theta_H \\ e_L, & \text{if } \hat{\theta}_i = \theta_L \end{cases}$$

- 3) The firm performs the outcome function  $g(b_1, b_2)$ , and hires the winner.

Since each worker  $i$  does not need to pay the cost  $b_i/\theta_i$  when playing in the direct mechanism, the utility function of each worker  $i = 1, 2$  is changed from Eq (3) to the follows:

$$u_i(\hat{\theta}_i, \hat{\theta}_j; \theta_i) = \begin{cases} w, & \text{if } \hat{\theta}_i > \hat{\theta}_j \\ 0.5w, & \text{if } \hat{\theta}_i = \hat{\theta}_j \\ 0, & \text{if } \hat{\theta}_i < \hat{\theta}_j \end{cases} \quad (9)$$

The utility matrix is as follows.

$\hat{\theta}_i \backslash \hat{\theta}_j$	$\theta_L$	$\theta_H$
$\theta_L$	$[0.5w, 0.5w]$	$[0, w]$
$\theta_H$	$[w, 0]$	$[0.5w, 0.5w]$

Obviously, the dominant strategy for each worker  $i = 1, 2$  is to definitely announce  $\hat{\theta}_i = \theta_H$ , no matter what his true productivity type is. Consequently, the social choice function  $f(\vec{\theta})$  cannot be implemented truthfully in a direct mechanism. **Q.E.D**

### 3 Conclusions

From Proposition 1 and 2, we can see that:

- 1) In the indirect mechanism  $\Gamma$ , the utility function of each worker  $i = 1, 2$  is given by Eq (3). It is the cost  $b_i/\theta_i$  of a type  $\theta_i$  worker choosing the bid  $b_i$  that makes the strategy combination  $(b_1^*(\theta_1), b_2^*(\theta_2))$  become a Bayesian Nash equilibrium. The cost of a worker is irrevocable even if he loses the job competition.

2) In the direct mechanism  $\Gamma_{direct}$ , the utility function of each worker  $i = 1, 2$  is given by Eq (9), where the cost disappears. Thus, each worker is free to pretend to be a high-productivity worker without any suffer. There is no way for the direct mechanism to discriminate this counterfeit.

In summary, the revelation principle does not hold at all.

## Appendix: Definitions in Section 23.B and 23.D [1]

**Definition 23.B.1:** A social choice function is a function  $f : \Theta_1 \times \dots \times \Theta_I \rightarrow X$  that, for each possible profile of the agents' types  $(\theta_1, \dots, \theta_I)$ , assigns a collective choice  $f(\theta_1, \dots, \theta_I) \in X$ .

**Definition 23.B.3:** A mechanism  $\Gamma = (S_1, \dots, S_I, g(\cdot))$  is a collection of  $I$  strategy sets  $S_1, \dots, S_I$  and an outcome function  $g : S_1 \times \dots \times S_I \rightarrow X$ .

**Definition 23.B.5:** A direct revelation mechanism is a mechanism in which  $S_i = \Theta_i$  for all  $i$  and  $g(\theta) = f(\theta)$  for all  $\theta \in \Theta_1 \times \dots \times \Theta_I$ .

**Definition 23.D.1:** The strategy profile  $s^*(\cdot) = (s_1^*(\cdot), \dots, s_I^*(\cdot))$  is a *Bayesian Nash equilibrium* of mechanism  $\Gamma = (S_1, \dots, S_I, g(\cdot))$  if, for all  $i$  and all  $\theta_i \in \Theta_i$ ,

$$E_{\theta_{-i}}[u_i(g(s_i^*(\theta_i), s_{-i}^*(\theta_{-i})), \theta_i) | \theta_i] \geq E_{\theta_{-i}}[u_i(g(\hat{s}_i, s_{-i}^*(\theta_{-i})), \theta_i) | \theta_i]$$

for all  $\hat{s}_i \in S_i$ .

**Definition 23.D.2:** The mechanism  $\Gamma = (S_1, \dots, S_I, g(\cdot))$  implements the social choice function  $f(\cdot)$  in Bayesian Nash equilibrium if there is a Bayesian Nash equilibrium of  $\Gamma$ ,  $s^*(\cdot) = (s_1^*(\cdot), \dots, s_I^*(\cdot))$ , such that  $g(s^*(\theta)) = f(\theta)$  for all  $\theta \in \Theta$ .

**Definition 23.D.3:** The social choice function  $f(\cdot)$  is truthfully implementable in Bayesian Nash equilibrium if  $s_i^*(\theta_i) = \theta_i$  (for all  $\theta_i \in \Theta_i$  and  $i = 1, \dots, I$ ) is a Bayesian Nash equilibrium of the direct revelation mechanism  $\Gamma = (\Theta_1, \dots, \Theta_I, f(\cdot))$ . That is, if for all  $i = 1, \dots, I$  and all  $\theta_i \in \Theta_i$ ,

$$E_{\theta_{-i}}[u_i(f(\theta_i, \theta_{-i}), \theta_i) | \theta_i] \geq E_{\theta_{-i}}[u_i(f(\hat{\theta}_i, \theta_{-i}), \theta_i) | \theta_i], \quad (23.D.1)$$

for all  $\hat{\theta}_i \in \Theta_i$ .

## **Acknowledgments**

The author is very grateful to Fang Chen, Hanyue, Hanxing and Hanchen for their great support.

## **References**

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