About scarce resources allocation in conditions of incomplete information

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Abstract

The article examines the problem of the efficient allocation of resources in conditions of incomplete information concerning the parameters of agents' utility functions. Through business game the results are modeled and compared in conditions of incomplete information concerning the agents' utility functions. We experimentally prove the inexpediency of information distortion of the agents' effectiveness when using a non-manipulative distribution mechanism in a multi-step game.

Keywords: game theory; reflexive games; incomplete information; information structure; information management; distribution mechanisms; behavior models; utility functions; nontransferable utility; fuzzy logic

1. Introduction

The problem of effective resource allocation occurs in various applied problems [4]. If the resource value is limited and the participants interests do not coincide, a conflict situation arises. Interaction of participants in this case can be considered as a game. The description of several agents interaction includes the following parameters:

- the multitude of agents;
- agent preferences (he/she is assumed that each agent is interested in maximizing his profits);
- set of permissible actions;
- awareness of agents (at the time of making decisions about the chosen action);
- the order of functioning (the sequence of actions).

These parameters set the game. The game purpose is to define the multitude of active agents' actions. It means finding an equilibrium situation.

Decision-making models, behavior models, the equilibrium concept have been studied in game theory for more than 100 years. The review of the results is given, for example, in [3].

Basically, it is assumed that participants have the same information about the parameters of the game. A class of reflexive games in which agent awareness is not a common knowledge and agents make decisions according to their perceptions of opponents' preferences. Their permissible actions are described in [8].

Obviously, in the situation of incomplete information, participants' behavior patterns. Indeed, if the agent assumes that his rivals are "strong" players, he/she will stick to one behavior pattern; If he/she thinks that opponents are "weak" players, then the behavior pattern can change.

The process and result of the agent's thinking about the values of uncertain parameters, and what about these competitors think about these parameters, are called information reflection [8]. The players' perception hierarchy is represented by the form of information structuretree. The research and analysis of the game information structure allows to determine the conditions for the information equilibrium, as well as to set the information management task-to create the information structure creation that implements the equilibrium situation that is most beneficial to the Resources Allocation Center.

This paper is devoted to the investigation of the appropriateness of information distortion about the agents' effectivenessin different behavior models in the situation of incomplete awareness of the participants about the parameters of the game.

It is assumed that the game participants can distort information about their target functions parameters, posing as "strong" or "weak" players. A hypothesis that the information distortion about the values effectiveness, with the possibility of requests further distortion, does not have a significant effect on the limited resource distribution between the players, is confirmed experimentally.

The study was carried out using an original Fuzzy Logic Model (FLM) [6] and the Best Response Model (BRM) [1].

2. Basic concepts and parameters

Let us consider the problem of distributing the resource R between n players. R be a distributable resource; N is the number of players;

 $u(x_i) = bx_i - a_i x_i^2$, a > 0, b > 0 is the utility function of the *i*-th player.

Obviously, the player will get the maximum profit at the point $x_i^* = \frac{b}{2a_i}$.

In the case $\sum_{i=1}^{n} x_i^* > R$, the conflict situation develops and players are forced to fight for the resource.

If $S(x_1, x_2, ..., x_n) = \sum_{i=1}^n (bx_i - a_i x_i^2)$ is the players total profit and there exists a restriction $\sum_{i=1}^n x_i = R$, then it is easy to show that $S(x_1, x_2, ..., x_n)$ reaches a maximum at the point $(x_1^0, x_2^0, ..., x_n^0)$, where

$$x_i^0 = \frac{a_1 a_2 \dots a_{i-1} a_{i+1} \dots a_n}{\sum_{i \neq j} a_i a_j}, i = \overline{1, n}, j = \overline{1, n}$$

If $x_i^* \neq x_i^0$ then the *i*-th player will be interested in increasing his profit.

As a mathematical model of the described interest conflict situation, we will use the business game for resource allocation R between n players with a reverse priority mechanism. At each step of the game the participant makes an request s_i to the resource. The request is satisfied by the Resource Allocation Center in the volume

$$x(s_i) = \frac{\frac{A_i}{s_i}}{\sum_{j=1}^n \frac{A_j}{s_j}} R, i = \overline{1, n}, \text{ where } A_i = u(x_i^*).$$

The winning is determined by the player's profit from the resource obtained in the last step.

It should be note that the resource distribution is based on the knowledge of the values $A_i = u(x_i^*)$ of each of the participants. In a sense, A_i can be interpreted as the utility limit of the *i*-th player. Let us suppose that the true values of A_i are not known to the Center (the cost factor a_i is known only to the player) and for the distribution of the resource the players themselves inform the Center of the value A_i . In this case, the player has the opportunity to exaggerate, downplay the limit of its usefulness or to convey its true meaning. Also, at each step, players report the value of the required resource, which is adjusted by the players in order to obtain the desired amount.

How will the distribution of the resource change in conditions of incomplete information of the Center about the usefulness of the players? Is it possible in such conditions to maximize the profit of an individual player and the total utility of the players? Is it profitable for participants to hide the true meaning of the limit of their usefulness?

Purpose of the study

The purpose of this study is to compare the participants profit size in a business game on the resource distribution in different information levels of the players parameters conditions.

We tasks:

- to conduct a computational experiment in incomplete information conditions about the needs of players in resources, using different participants' behavior models;

- to conduct a computational experiment in incomplete information conditions about the players target functions and their resource needs, using different participants' behavior models;

- to conduct a comparative analysis of the results.

3. Description of the experiment

For carrying out the computing experiment two models will be used:

- Best Response Model (BRM);

- Fuzzy Logic Model (FLM).

The BRM [7] assumes that at the k+1 step of the game, the bid value s_i^{k+1} must be such that $x(s_i^{k+1}) = x_i^*$. If the remaining players do not change their bids, then the volume of the requestion be calculated from the condition

$$x(s_i^{k+1}) = \frac{\frac{A_i}{s_i^{k+1}}}{\sum_{j=1}^n \frac{A_j}{s_i^k} - \frac{A_i}{s_i^{k+1}}} R = x_i^*$$

then

$$s_{i}^{k+1} = \frac{\frac{A_{i}}{x_{i}^{*}}}{\sum_{j=1}^{n} \frac{A_{j}}{s_{j}^{k}} - \frac{A_{i}}{s_{i}^{k}}} (R - x_{i}^{*})$$

The FLM [6] uses the following input data:

$$\alpha_i = \frac{x(s_i)}{x_i^*}$$

 α_i is the degree of satisfaction of the request;

N is the proportion of players with $\alpha_i \ge 1$.

The rules base, which gives an assessment of the attractiveness of the player's actions, consists of the possible actions:

- to increase the request,

- to lower the request,

- not to change the request.

The rules base has the form:

R1. If the degree of the request satisfaction α_i is small and the players share N is low, then the declining of the request attractiveness is great.

R2. If the degree of the request satisfaction α_i is small and the players proportion N is high, then the attractiveness not to change the request is great.

R3. If the degree of the requests satisfaction α_i is close to 1 and the players share N is low, then the declining of the request attractiveness is great.

R4. If the degree of the requests satisfaction α_i is close to 1 and the share of players N is high, then the attractiveness of the bid increase is great.

R5. If the degree of the request satisfaction α_i is large and the players share N is low, then the attractiveness not to change the request is great.

R6. If the degree of the requests satisfaction α_i is large and the share of players N is high, then the attractiveness of the bid increase is great.

As a result of FLM, the evaluation $\lambda \in [0,1]$ of the attractiveness of player actions is given. The player may increase the bid $(P\uparrow)$, lower the bid $(P\downarrow)$ or not to change the request (P0).

Special software was developed for the experiment in the program environment O-Tree [9].

In the course of study, various combinations of the input parameters considered in Table 1 were considered. In each experiment, a series of 10 stepswas conducted.

Table 1. Experiments input parameters combinations. Deficiency of resorce Utility function Relative location of x_i^* Behavior model № experiment $u(x_i) = bx_i - a_i x_i^2$ BRM 1 the same small the same 2 the same large the same BRM 3 different small narrow spread BRM 4 different BRM small wide spread different 5 BRM large narrow spread BRM 6 different large wide spread 7 small FLM the same the same large 8 the same the same FLM 9 different small narrow spread FLM 10 different small wide spread FLM 11 different large narrow spread FLM 12 different large wide spread FLM

The form of utility function determines whether the player should maximize the amount of the resource he receives, or optimize it.

Deficiency of resource implies various tensions in the game and level request distortion.

Relative location of x_i^* means that the optimal resource values in different functions have the same deviation from equal distribution. In this caseplayers have the same chance to be winner.

Behavior model means that players use special rules for their actions.

4. Results and Discussion





Fig. 2. The first agent overestimates the importance of its effectiveness.

The dynamics of resource allocation is presented on Figures 1, 2, 3.

Here x_1 is the value of the resource allocated to the first player, x_2 is the value of the resourceallocated to the second player, x_1 is the value of the resource allocated to the third player.



Fig. 3. The first agent underestimates the importance of its effectiveness.

In all cases, the deviation of the obtained resource from the optimal individual indicator is approximately the same. Table 2 shows the relative deviations of the resource obtained by agents in cases of reliable reporting of information on effectiveness, overestimation of the first agent effectiveness, underestimation of the first step.

Table 2. The relative deviations of the resource obtained by agents in the first step.

| | Nº1 | N <u>∘</u> 2 | Nº3 |
|----|-------|--------------|-------|
| x1 | 0,33 | 0,25 | 0,43 |
| x2 | 0,07 | 0,12 | 0,00 |
| x3 | -0,07 | 0,00 | -0,15 |

 \mathbb{N} 1. In the first step all players provide reliable information about their own effectiveness and the amount of the required resource. In the next steps distorting the value of the resource request is distorted in accordance with the chosen behavior model.

 \mathbb{N} 2. In the first step the player 1 overstates the information on its own efficiency by 20%, other players provide reliable information about their own effectiveness and all players report reliable information about the amount of the required resource. In the next steps distorting the value of the resource request in accordance with the chosen behavior model.

№ 3. In the first step the player 1 understates information about its own efficiency by 20% other players provide reliable information about their own effectiveness and all players report reliable information about the amount of the required resource. In the next steps distorting the value of the resource request in accordance with the chosen behavior model.

Table 3 shows the resources relative deviations obtained by agents in cases of reliable reporting of information on effectiveness, overestimation of the first agent effectiveness, underestimation of the first agent effectiveness in the tenth step.

| | N <u>∘</u> 1 | № 2 | N <u>∘</u> 3 |
|----|--------------|------------|--------------|
| x1 | 0,12 | 0,12 | 0,12 |
| x2 | 0,16 | 0,16 | 0,16 |
| x3 | 0,17 | 0,17 | 0,16 |

Table 3. The relative deviations of the resource obtained by agents in the tenth step.

The results of the calculations presented in the tables 2, 3. The information distortion about efficiency leads to a significant change in the distribution results in the first step. As we can see from the results of the calculations presented in the Table 3, the information distortion about efficiency does not lead to a change in the distribution results in the tenth step.

Figure 4 presents the averaged values of the relative deviations from the optimal resource values in games with BRM (exact information about the effectiveness of players, distorted information about the effectiveness of players).



a is the players provide reliable information about the maximum of their profits;

b is the players overestimate the value of their maximum profit;

c is the players underestimate the value of their maximum profit;

d is the players distort information about the maximum of their profits.

P1 is the first player, P2 is the second player, P1 is the third player.

Figure 5 shows the averaged values of the total utility of participants in games with BRM and FLM (exact information about the effectiveness of players, distorted information about the effectiveness of players)



Fig. 5. The averaged values of the total utility.

Table 4 presents numerical data on the resource distribution among participants, the magnitude of individual and total profits.

Table 4.The resource distribution among participants.

| Player | The value of the resource distribution | | | Profit of player | | | | |
|--------------|--|----------|----------|------------------|----------|----------|----------|----------|
| | a | b | c | d | а | b | с | d |
| P1 | 453,57 | 455,4782 | 460,1372 | 455,9209 | 289962,8 | 288575,2 | 285095,7 | 288250,2 |
| P2 | 355,1272 | 353,4823 | 349,4531 | 353,096 | 79677,78 | 82215,97 | 88318,84 | 82808,03 |
| P3 | 191,3028 | 191,0396 | 190,4097 | 190,9831 | 89831,48 | 90110,22 | 90772,57 | 90169,83 |
| Total profit | | | | 459472 | 460901,4 | 464187,1 | 461228,1 | |

You can see that difference between total profit in the different experiments consists less of than 1%. We consider this deviation to be insignificant.

5. Conclusion

The article discusses the effectiveness of distorting information about the agents' effectiveness in incomplete information conditions in the limited resource distribution problem.

Experiments were performed by robots with various input parameters combinations. A comparative analysis of the games results with reliable and inaccurate information about the players effectiveness was carried out. The agents' profit and the system total profit are calculated.

The conducted experiments showed that a single distortion of information about the effectiveness of players, with constant distortion of players 'requests, does not affect the distribution of players' profits.

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