

# USING FUZZY SET APPROACH IN MULTI-ATTRIBUTE AUTOMATED AUCTIONS

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Abstract: This paper designs a novel fuzzy attributes and competition based bidding strategy (FAC-Bid), in which the final best bid is calculated on the basis of the assessment of multiple attributes of the goods and the competition for the goods in the market. The assessment of attributes adapts the fuzzy sets technique to handle uncertainty of the bidding process. The bidding strategy also uses and determines competition in the market (based on the two factors i.e. no. of the bidders participating and the total time elapsed for an auction) using Mamdani's Direct Method. Then the final price of the best bid will be determined based on the assessed attributes and the competition in the market using fuzzy reasoning technique.

## 1 INTRODUCTION

Online auctions (e.g. eBay and Amazon) are popular market institutions for conducting transactions between the buyers and the sellers. The software agent technology is well known paradigm in on-line auctions for buying and selling the goods. Software agents provide automated assistance for trading through the knowledge of the market and the requirements of the market (Rahman and Robert 2001). The agents can use different auction mechanisms (e.g. English, Dutch, Vickery etc.) for procurement of goods or reaching agreement between agents. The agent makes decisions on behalf of consumer and endeavours to guarantee the delivery of item according to the buyer's preferences. In these auctions buyers are faced with difficult task of deciding amount to bid in order to get the desired item matching their preferences. The bidding strategies for the software agents can be static or it may be dynamic. The static agents may not be appropriate for the negotiating market situations like extent of competition may vary as traders leave or enter into the market, deadlines and new opportunities may increase the pressure. The dynamic or we can say flexible negotiation

capabilities for software agents in the online auctions have become a central concern (Murugesan 2000). Agents need to be able to prepare bids and evaluate offers on behalf of the users they represent with the aim of obtaining the maximum benefit (Ma and Leung 2007) for their users according to the changing market situation. Much research has already been done by the researchers to formulate different bidding strategies according to the changing market situations.

Strategies based on flexible negotiation agents perform better as compared to the strategies based on fixed negotiation agents (Kwang and Wang 2004). Faratin *et al* (Faratin, Sierra and Jennings 1998) developed strategies based on time, attitude, resources, but many more factors such as competition, trading alternatives are not considered. In this paper we focus on the design of a novel bidding strategy based on the above mentioned factors to be used by the software agent in a online auction. A fuzzy attributes and competition based bidding strategy (FAC-Bid) is designed, in which the final best bid is calculated on the basis of the assessment of the attributes of the goods as well as the competition for these goods in the market.

## 2 A FUZZY ATTRIBUTE AND COMPETITION BASED BIDDING STRATEGY (FAC-BID)

In fuzzy attribute and competition based bidding strategy (FAC-Bid) (Fig. 1), the factors which are focused are assessing attributes of the goods and competition for the goods in the market. For estimation of the price for a bid for winning an auction, the agent must have a balanced behaviour between these factors i.e. the assessment of the good's attributes and finding the competition for the goods in the market. The bidding price of the quality goods is more as compared to the less quality goods. Also the increasing competition for the goods in the market increases the bidding price for that good. The competition in turn depends on the number of bidders and the time elapsed for the auction. As the number of bidders increases, the competition among them also increases, resulting in a higher price. In the beginning of the auction the competition is less and it increases as time elapses and it is at the peak when time approaches approximately in the middle of the auction period. At the end of the auction period the competition among the bidders decreases.

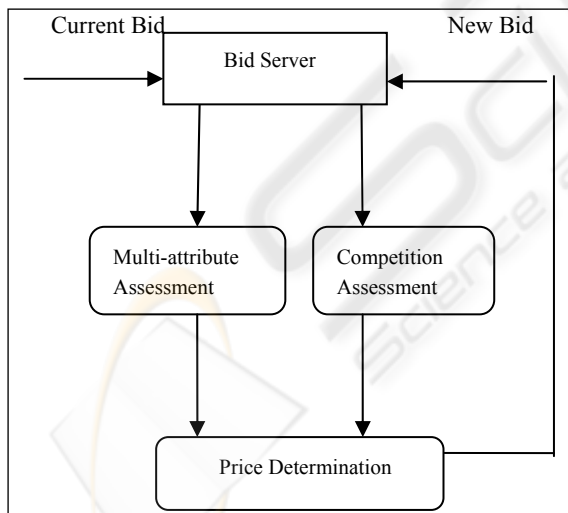


Figure1: A Fuzzy Bidding Strategy (FAC-Bid) Mode.

The steps of the design of fuzzy attribute and competition based bidding strategy (FAC-Bid) are as follows:

- First, each attribute is evaluated

- Then the assessment of all these attributes will be aggregated to form an overall assessment of the goods.
- next the level of competition as the function of no. of bidders and time elapsed for the auction will be found
- Finally the best bid is calculated on the basis of the overall assessment of the good's attributes and the competition for the goods in the market.

In this paper we have used fuzzy set methods to deal with the uncertainty, which exists during the determination of overall assessment of the goods for their attributes, the level of competition in the market and to find the final bid price. First of all, this paper uses a satisfactory degree measure as the common universe of assessment, i.e., an assessment is treated as a fuzzy set on the satisfactory degree. Secondly, competition is expressed as a fuzzy set on the fuzzy sets of the no. of bidders and the time elapsed of the auction. Thirdly, bid displacement factor for finding the final bid is expressed as a fuzzy set on the fuzzy set of assessment of the attributes and the competition for the goods in the market.

### 2.1 Attribute Evaluation

The attribute evaluation is done in two parts (Goyal and Ma 2009). The first one is individual attribute assessment and the second one is assessment aggregation. To implement attribute evaluation, three issues are concerned i.e. attribute weights (relative importance) adjustment, assessment expression and assessment aggregation.

**Weight Adjustment.** Weight adjustment implements dynamically change relative importance of multiple criteria. In a real situation an agent's personal preference on the attributes seldom has quickly fluctuation, i.e., the weights for criteria is relatively stable in a long run. The adjustment of weights resulted from the price should be limited to a rational range. Moreover, the adjustment shouldn't change the relative significance among criteria other than the price because raising price alters the relative significance of it to other criteria. In the following, the agent's preference is treated as an initial weight vector which is the basis of the adjustment. To construe an initial weight vector, the Analytic Hierarchy Process (AHP) method is applied because it is proved validate in practice although it may induce inner inconsistency. Suppose obtained initial weight vector is  $W^{(0)}$ .

Suppose the current bid  $p_c$  belongs to  $[p_l, p_u]$  where  $p_l$  and  $p_u$  are the lower and upper boundaries

of possible bids respectively which are determined by the auction. Let  $C = \{c_0, c_1, \dots, c_K\}$  be the set of  $K+1$  attributes and  $W = \{w_0, w_1, \dots, w_K\}$  is the set of weights for attributes in  $C$ . Because except the price agent's assessments on other criteria do not change, the adjustment of weight for price should be determined first. Suppose  $[-\delta, \delta]$  is the adjustable range of the weight for price and the current net increasing of weight for price is  $\Delta w_0$ , then the current weight vector is determined by

$$w'_0 = w_0 + \Delta w_0 \quad (1)$$

$$w'_k = w_k \frac{1 - w'_0}{1 - w_0} \quad (2)$$

where  $w_k$  ( $k = 0, 1, \dots, K$ ) is the component of  $W^{(0)}$ . Obviously,

$$\sum w_k = 1 \quad \text{for } k=0 \text{ to } 1 \quad (3)$$

and the relative significance of the criteria except for the price will not change after this adjustment.

**Assessment Expression.** Since uncertain expressions are often used in a real situation, this paper uses linguistic terms to express assessments. These linguistic terms are illustrated by fuzzy set. Moreover, the universe of these fuzzy set are unified to real interval  $[0,1]$  which means the satisfactory degree of the agent to a particular attribute.

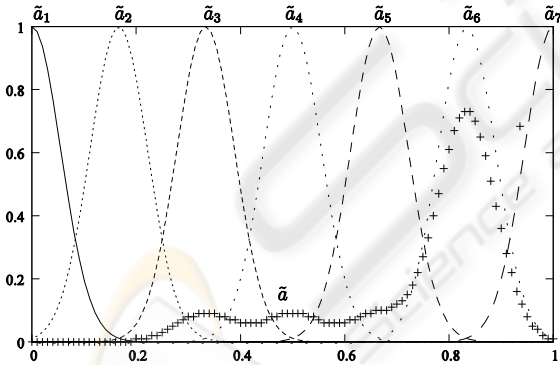


Figure 2: Obtain overall assessment.

Therefore, all fuzzy sets have same universe which is convenient for aggregating assessments. Let  $g_k$  ( $k = 0, 1, \dots, K$ ) is the satisfactory degree measure for attribute  $c_k$ . Then an agent's opinion on the goods in terms of attribute  $c_k$  is denoted by  $g_k(u)$  where  $u \in U_k$  is the real attribute value of attribute  $c_k$  and  $U_k$  is the real universe for attribute  $c_k$ . For instance, departing time is an attribute for a flight ticket. The possible departing time in a day is from 0:00 to 23:59. For any time slot  $u$ , a client may present a

satisfactory degree such as departing at 7:30 is with satisfactory degree 0.9 and departing at 3:00 is with 0.3. In the following,  $A = \{a_1, \dots, a_n\}$  be the set of used assessment terms which are fuzzy sets on satisfactory degree  $[0,1]$ . Then a numeric satisfactory degree is transformed to a linguistic term. In the above example, suppose the assessment set is as shown in figure 2. Notice that  $a_7$  is with the biggest membership degree for 0.9, the assessment for departing at 7:30 is  $a_6$  by the maximum membership degree principle. Similarly, the assessment for 0.3 is  $a_2$ .

**Assessments Aggregation.** An aggregated assessment is the agent's overall opinion/preference on the goods in terms of multiple attributes. Take booking a flight ticket for example, an assessment is made on a ticket usually based on the airlines, flight departure and arrival time, flight type, aircraft types, seat positions, as well as price. The change of an attribute's value may leads to the alternation of an assessment. Instinct natures of different attributes increase the difficulty and uncertainty for obtaining an overall assessment. Notice that an agent's preference on an individual attribute can be expressed through the agent's satisfactory degree on that attribute. This paper uses an satisfactory degree measure as the common universe of assessment. Based on assessment on each individual attribute, an overall assessment can be obtained as follows. Suppose the individual assessments of all attributes are  $v_0, v_1, \dots, v_k$  and the weights of them are  $w_0, w_1, \dots, w_k$  respectively. Then an overall assessment is obtained by

$$a = \text{Agg}\{(v_0, w_0), (v_1, w_1), \dots, (v_k, w_k)\} \quad (4)$$

where  $\text{Agg}$  is a selected aggregation method,  $v_k \in A$  ( $k = 0, 1, \dots, K$ ) is the linguistic assessment on attribute  $c_k$ . To get an overall assessment in terms of a set of criteria, an aggregation method  $\text{Agg}$  is applied. Here we use the weighted-sum-based method to obtain an overall assessment as follows. First, we construct a fuzzy set  $\tilde{a}$  on  $[0,1]$  through

$$\tilde{a}(u) = \sum w_k \cdot v_k(u), \quad u \in [0,1] \text{ for } k=0 \text{ to } 1 \quad (5)$$

where  $v_k(u)$  is the membership degree of  $u$  in  $v_k$ . Next, we calculate the distance between  $\tilde{a}$  and  $a_i \in A$ , by

$$d(\tilde{a}, a_i) = \int |\tilde{a} - a_i| d\lambda. \quad (6)$$

Finally, we select the nearest term(s)  $a$  to  $\tilde{a}$  as the overall assessment. For example,  $A$  has seven terms, namely,  $a_1, a_2, \dots, a_7$  as shown in figure 2. Suppose  $\tilde{a}$  is the obtained fuzzy set. By comparing the distances between  $\tilde{a}$  and each element in  $A$ , we know  $a_6$  is the

nearest item to  $\tilde{a}$ . Hence,  $a_6$  will be taken as the overall assessment.

### 2.2 Competition Assessment

The level of competition in an auction may be captured by the number of bidders and the time elapsed. Competition among bidders plays an integral role in price formation (Reddy and Dass 2006). As the number of bidders increases, the competition among them also increases, (Fig. 3) resulting in a higher price. Bapna, Jank and Shmueli found the number of bidders to be positively associated with the current price of the item. Furthermore, it is observed that, typically, the middle of the auction experiences a smaller amount of bidder participation as compared to the early and later stages of the auction. Bidders generally utilize this time to scrutinize the auctioned item or just simply wait to see how other bidders behave. Therefore, it would be interesting to see how this competition characteristic affects the on-line auction's price formation. We anticipate that the number of bidders has a significant positive relationship with price levels. In the beginning of the auction the competition is less and it increases as time elapses and it is at the peak when time approaches approximately in the middle of the auction period. At the end of the auction period the competition among the bidders decreases (Fig. 4).

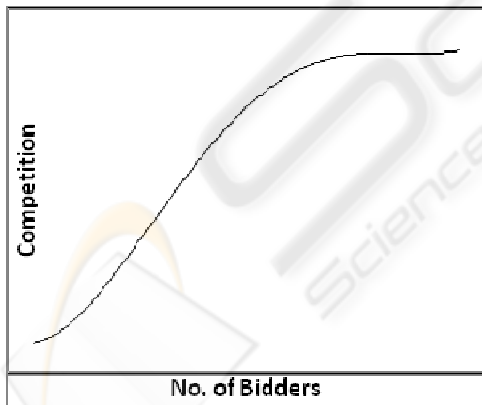


Figure 3: Competition versus Number of Bidders.

Here we will describe the competition factor in terms of no. of bidders ( $b$ ) and the total time elapsed ( $t$ ) for the auction of items. We will consider the competition as a set fuzzy set of values  $c_1, c_2, \dots, c_n$ , no. of bidders  $B$  as a fuzzy set of values  $y_1, y_2, \dots, y_n$ . And the time elapsed as another fuzzy set  $T$  of values  $x_1, x_2, \dots, x_n$ . According to Mamdani's Direct

Method (Tanaka 1991) we can find adaptability  $n$  no. of rules  $w_1, w_2, \dots, w_n$  as

$$w_i = \mu_{x_i}(T) \vee \mu_{y_i}(B) \text{ for } i=1 \text{ to } n$$

Then the conclusion of each rule can be found as

$$\mu_{c'_i}(C) = w_i \vee \mu_{c_i} \text{ for } i=1 \text{ to } n$$

These conclusions can be aggregated to find the final conclusion

$$\mu_c(C) = \mu_{c'_1}(C) \wedge \mu_{c'_2}(C) \wedge \dots \wedge \mu_{c'_n}(C)$$

To find the definite value for the conclusion, here centre of gravity of the fuzzy set has been applied as follows

$$C = \frac{\int \mu_z(C) C dC}{\int \mu_z(C) dC} \quad (7)$$

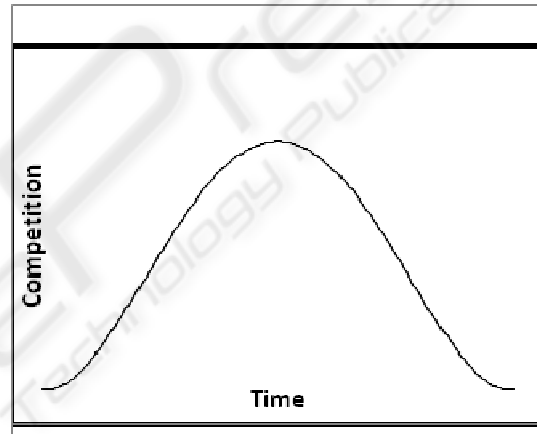


Figure 4: Competition versus Time Elapsed.

### 2.3 Agent Price Determination

Price of the goods depends on the assessed attributes of the goods and the competition in the market for that good. If the assessment of the attributes of the goods is good and also competition for that product in the market is high then the price of the item is high. If the assessment of the attributes of the goods bad and competition is also low then the price for that item is obviously be low. If the attribute's assessment is good and the competition is low then the price is going to be medium and so on.

We can calculate the price of the good based on the assessed attributes and the competition determined which is based on the no. of bidders and time elapsed for the auction by applying Mamdani's Method for fuzzy relations and compositional rule of inference (Tanaka 1991). Here we will describe the price of goods in terms of assessment of the attributes of the goods and competition in the market

for that good. We will consider bid displacement factor  $\Delta P$  as a fuzzy set of values  $p_1, p_2, \dots, p_n$ , assessment of the attributes  $A$  as a fuzzy set of values  $a_1, a_2, \dots, a_n$  and competition  $C$  as a fuzzy set of values  $c_1, c_2, \dots, c_n$ . According to Mamdani's Method for fuzzy relations and compositional rule of inference the rule  $a_i$  and  $c_j \rightarrow p_k$  can be described by

$$\mu R(A, C, \Delta P) = \mu a_i(A) \wedge \mu c_j(C) \wedge \mu p_k(\Delta P) \quad (8)$$

For  $n$  no. of rules, the compiled fuzzy relation  $R$  is given as

$$R = R_1 \cup R_2 \cup \dots \cup R_n$$

For the input of fuzzy set  $A'$  on  $A$  and fuzzy set  $C'$  on  $C$ , the output fuzzy set  $\Delta P'$  on  $\Delta P$  can be obtained as follows

$$\Delta P' = (A' \text{ and } C') \circ R = A' \circ (C' \circ R) = C' \circ (A' \circ R) \quad (9)$$

And then the final price for the bid will be Final bid = Current bid +  $\Delta P'$

### 3 CONCLUSIONS

In this paper we have designed a fuzzy attribute and competition based bidding strategy (FAC-Bid), which uses a soft computing method i.e. fuzzy logic technique to compute the final bid price based on the assessment of the attributes and the competition in the market. Another unique idea presented in this paper is that to deal quantitatively the imprecision or uncertainty of multiple attributes of items to acquire in auctions, fuzzy set technique is used. The bidding strategy also allows for flexible heuristics both for the overall gain and for individual attribute evaluations. Specifically, the bidding strategy is adaptive to the environment as the agent can change the bid amount based its assessment of the various attributes of item, and the competition in the auction. The competition is calculated based on the number of bidders and the time elapsed for the auction. It was noticed that the strategies in which agent's behaviour depends on attributes and competition, are easily adaptable to the dynamic situations of the market. In future we will investigate about the development of the bidding strategies for multiple auctions. We will also compare our bidding techniques with the other strategies to find out the relative strengths and the weaknesses.

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