# DESIGN OF CONTINUOUS CALL MARKET WITH ASSIGNMENT CONSTRAINTS

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Abstract: Today's companies increasingly use Internet as the common communication medium for commercial transactions. Global connectivity and reach of Internet means that companies face increasing competition from various quarters. This requires that companies optimize the way they do business, change their business processes and introduce new business processes. This has opened up new research issues and electronic or automated negotiation is one such area. Few companies have tried to introduce electronic auctions for procurement and for trade negotiations. In the present paper, we propose a design of continuous call market, which can help enterprises in electronic procurement as well as selling items electronically. The design is based on double sided auctions, where multiple buyers and sellers submit their respective bids and asks. Buyers and sellers can also specify assignment constraints. The main feature of our work is an algorithm, which generates optimum matching with polynomial time complexity under assignment constraints.

### **1 INTRODUCTION**

Auction based protocols are widely used in electronic commerce. An application of auction theory in electronic procurement has also been studied in (Eso, 2001), and it gives near optimal solution to bid evaluation problem of the buyer. A procurement process, which minimizes the cost using auctions, has been proposed in (Bichler, 1999). Auctions provide an efficient price discovery mechanism to sellers. Double-sided auctions provide mechanism for clearing markets with multiple buyers and sellers. Two main institutions for double auctions are continuous double auction and a clearing house or continuous call double auction. In double auction, markets buyers submit their bids and sellers submit asks. A transaction occurs if the buyer's bid price exceeds seller's ask price. A continuous double auction is one in which many individual transactions are carried out and trading does not stop. Call markets on the other hand are periodic versions of continuous double auctions, where bids from buyers and asks from sellers are collected over a specified interval of time

and the market is cleared at the end of the interval. The continuous call double auction is the oldest practiced type of market for exchange of stocks, where buyers and sellers post their respective bids and asks continuously. In last few years, there has been growing interest in Internet-based market places.

In continuous call market, buyers submit bids and sellers submit asks continuously. These bids and asks are matched periodically and the market clears them. While clearing the call market, we need to determine optimal matching of asks with bids. In case, there are no assignment constraints, efficient algorithms exists to determine the optimum allocation. This problem of matching asks and bids can become complex in case there are assignment The simplest type of assignment constraints. constraints can be of the types where a bid can be matched with some of the asks. Other types of constraints are the ones where the buyer specifies a bid to be all or nothing type. In such cases, the buyer's demand should be completely satisfied. In some cases, the demand can be indivisible, meaning that a demand from an ask cannot be partly matched

182 R. Dani A., P. Gulati V. and K. Pujari A. (2005). DESIGN OF CONTINUOUS CALL MARKET WITH ASSIGNMENT CONSTRAINTS. In Proceedings of the Seventh International Conference on Enterprise Information Systems, pages 182-187 DOI: 10.5220/0002550001820187 Copyright © SciTePress with supply from an ask. Such constraints (known as indivisible demand constraints (Kalagnanam, 2001)) are not considered here. In this paper, we consider matching problem with multiple assignment constraints.

The main contribution of this paper is the development of an algorithm to match asks and bids under assignment constraints. This algorithm is fundamental to the design of continuous call market presented here. This mechanism helps enterprises in electronic procurement as well as to sell items and also specify constraints on different attributes. The main example presented here is about paper exchange, however the formulation is a general one and can be used in other similar cases. One of the important features of our algorithm is that it takes into account other attributes (like width) apart from common attributes like price and quantity. Another important feature is that it obtains optimum matching under different types of assignment constraints. While submitting bids and asks, buyers and sellers can specify assignment constraints. Apart from this, it also handles implicit assignment constraints. It is also shown that our algorithm can solve the problem with worst case time complexity of  $O(n^2)$ . The algorithm can work in cases where there are no assignment constraints and price and quantity are the only decision attributes. Thus, the problem is addressed in much more general settings. The rest of the paper is organized as follows, in section 2, we discuss how implicit and explicit assignment constraints can arise in different situations. The related work is also discussed in this In section 3, we first present the section. formulation of the problem. The problem has been formulated as 0-1 programming problem. We have discussed some of the related issues here. Our algorithm is presented in section 4. We also discuss related issues are discussed. Experimental results and implementation details like different types of Unified Modeling Language (UML) diagrams and modules are discussed in section 5 and we conclude in section 6.

## 2 EXAMPLE AND RELATED WORK

Most of the stock exchanges like New York, Tokyo (McCabe, 1992), (Arizona Stock Exchange) and (Optimark) have implemented online trading systems based on double auction mechanism. Apart from financial markets, double-auction-based mechanisms have been widely used in multi commodity stock exchanges like National Stock

Exchange of India. In all these cases, the commodity is substitutable and buyers do not have any preference for a particular seller, much in the same way as goods like stocks, do not have any differentiating features apart from price and quantity. In all these cases, asks and bids can be matched without any types of assignment constraints. Only constraints that are imposed by buyers or sellers can be either or nothing type of bid (i.e. fulfilling complete demand), or minimum required demand (minimum quantity to buy or sell). In these cases, price and quantity become the only differentiating factor and matching is done based on them. However, in some cases, as indicated in the following example from the process industry, the price and quantity need not be the only differentiating factor and there can be different types of assignment constraints. Suppose that in a paper exchange following asks (Table 1) from different sellers, supplying paper of different grades, widths (in cm) and quantities (tons) are received. The bids received from different buyers indicating the price they are willing to pay, quantities required, grades of paper required and widths required, are listed in Table 2. Suppose that the sellers are supplying papers of different grades as indicated by positive integer 1,2,3 etc. The buyer also indicates the type of paper required by specifying its grade. Suppose that higher-grade paper is represented by higher integer value. It can be seen that there are certain assignment constraints in the above example, which must be considered while matching. The assignment constraints for matching are:

A buyer's bid requiring paper of grade 3, cannot be matched with an ask supplying paper of grade 1 or 2, as he requires paper of grade 3 only. In some cases, the buyer may specify that the paper of higher grade can be accepted. In this case, the demand for paper of grade 3 can be satisfied by supply with grade 3 or above. It may be possible to match a bid with an ask of higher grade.

Another constraint can be the width of the paper required. It can be seen from the above example that the bid from second buyer requiring paper of width 1200 cm can be matched with only those asks,

Table 1 : Asks for different sellers in Paper Exchange

Seller	Price	Width (cm)	Supply (Quantity)	Grade
1	100	1000	6	1
2	105	1000	2	2
3	110	2000	5	3
4	114	1200	5	4
5	119	1600	3	1

Table 2 : Bids from different buyers in Paper Exchange

Buyer	Price	Width	Demand	Grade	
1	175	2000	5	3	
2	170	1200	5	4	
3	165	800	3	1	
4	163	800	3	1	

which supply paper of width 1200 cm or more. It can also be seen that the supply from ask can be used to satisfy the demand of buyers 3 and 4 by cutting the paper of width 1600 cm into two rolls of 800 cm each.

An algorithm to match these bids and asks and provide optimum matching, based price discovery mechanism proposed in (Gjerstad 1998), when there are no assignment constraints, is presented in (Kalagnanam, 2001). In this case, the price  $p^*$  is determined by constructing the aggregate supply and This price is the point of demand curves. intersection of these two curves and is used as the clearing price. All the bids having prices above it and asks having lower prices are considered for matching. The highest price bid is matched with the lowest price ask. The matching continues sequentially till all selected bids and asks are matched. In some cases, as shown in the example of paper exchange, there can be assignment constraints on different attributes (like grade, width etc.). If there are no indivisible demand bid constraints, the matching assignment problem can be formulated as network flow problem (Kalagnanam, 2001), (Ahuja, 1993). This problem is converted into a network flow problem such that there is a starting node known as source node, intermediate node consisting of all bids and asks and ending node known as sink node. The arcs between these nodes are constructed using the difference between the prices as cost and quantity as weight. The cost is set to 0 for the arcs from starting node and arcs ending in sink node. This problem is then solved as the network flow optimization problem formulating either as maximization of matching of demand and supply or as profit of the auction problem. There are many algorithms available to solve this problem. The complexity of maximum flow problem i.e. matching of demand and supply is O  $(nm + n^2 \log n)$ ]. In this case, n represents number of nodes and m number of edges. In cases where demand is indivisible, finding optimal matching requires solving of NP-hard optimization problems like multiple knapsack problem, bin packing problem or generalized assignment problems (Martello, 1989), (Garey,

1979), (Sandholm, 1999), Kalagnanam, 2001) (Cheriyan, 1979).

To the best of our knowledge, no work related to matching in the presence of assignment constraints has been reported. In the present work, an algorithm to obtain optimum matching in case of assignment constraint has been developed.

### **3 PROBLEM FORMULATION**

One of the important feature of our design of Continuous Call Market with a assignment constraint is an algorithm to find out the optimum assignment of bids and asks. It always generates optimum assignment in case of assignment constraints as well as in unconstrained case. In case of assignment constraints, one approach can be to group set of n asks and m bids into different sets of asks A<sub>1</sub>,A<sub>2</sub>,A<sub>3</sub>,..A<sub>n</sub> and bids, B<sub>1</sub>,B<sub>2</sub>,B<sub>3</sub>,..B<sub>m</sub>. Each set is formed by asks and bids which satisfy a particular constraint (e.g. Grade constraint) and are disjoint. Let  $A_1$  be set of asks for which Grade attribute has value 1, A<sub>2</sub> be the set of asks for which grade attribute has value 2 and so on. Once the asks and bids can be grouped in this way, the matching can be done for each subset of asks and bids separately. However, this approach can be used efficiently when the constraints are of equality type. It will be difficult to group bids exactly where required paper grade is greater than 3 or width required is 800 cm. In the second case, the bid can possibly be matched with any ask where width is 800 cm or more. There can also be another problem. Suppose that a bid, which requires paper width of 800 cm (quantity 5 tons) is matched with an ask of 1200 cm (quantity 5 tons). Then remaining part of the ask i.e. 400cm of width (quantity 5 tons) will either be wasted or should be moved to another set. The algorithms presented in (Kalagnanam, 2001) also describe matching on two attributes, namely, price and quantity. We present an algorithm, which takes into account the size factor, apart from price and quantity. We have formulated the problem as 0-1 programming problem. The details of our formulation can be seen in (Dani 2005).

### **4 ALGORITHM FOR ASSIGNMENT**

Suppose that there are n bids and m asks. In our algorithm, assignment of the asks to bids is done as follows:

1. The assignment starts from the highest price bid. In the first stage, the set of asks with the highest unit contribution are determined. Then we obtain maximum quantity that can be assigned. Then assignment is carried out. We combine asks with the same unit contribution and always determine maximum possible quantity that can be assigned. If ask quantity is less than the bid quantity, then the ask is marked as temporarily assigned, while bid is marked as partly assigned and assignment continues from the next ask onwards till complete demand is fulfilled. The bid and/or all asks are marked as assigned. If ask has more quantity, then bid is marked as assigned and ask is marked as partly assigned. Then assignment is continued in decreasing order of price for bids.

2. Initially a table indicating demand for different width is constructed. If the ask size is multiple of bid size (e.g. bid width 800 cm and ask width 1600 cm), then this table is used to see whether there is demand for the remaining width (i.e. 800 cm). If there is demand, then wastage parameter is set to 0. This is helpful in situations to decide, whether bid is to be matched with ask of 1600 cm or 1000 cm. Here matching of current bid with ask width of 1000 cm will show lesser wastage than that of 1600 cm. However, if there is demand for 800 cm, then it can be matched with the ask of 1600 cm width so that wastage is minimized.

The assignment is continued till one of the three conditions holds (i) no ask is left (ii) no bid is left (iii) when ask price exceeds that of bid price. In our solution, assignment is carried out if bid price is either more than ask price or both are same. This assumption is reasonable in the sense that in most exchanges asks are cleared with bids of same or higher price. It can also be seen in (Kalagnanam, 2001), that equilibrium price is first obtained. This price is used to determine the asks and bids which can be cleared (asks below this price and bids above it). Let A be the list of asks and B be the list of bids.

#### Algorithm findopasg (A,B) /\* Main Algorithm \*/

Call Create\_Size\_Demand\_Table(bids); While ( there is unassigned\_bid in B ){ Call get\_next\_unassigned\_bid(bids); Call get\_opt\_ask(bid\_quantity,bid\_size,); Call assignment(bid,ask,bid\_quantity) } return ; } /\* End of main algorithm \*\*/ /\*\* Function to create Size Demand Table \*\*/ create\_Size\_Demand\_Table(bids) { while ( there is next bid) { get bid\_size, bid\_quantity ; call search\_table(bid\_size,table\_size) ; if ( .not., found )then { increase current\_index by 1 ;

store bid\_size to search\_table(current\_index,1);

store bid\_quantity to search\_table(current\_index,2);
 set table\_size to current\_index ;}
else(edd bid\_suppritive search\_table(suppritie))

else{ add bid\_quantity to search\_table(current\_index,2);}
read next\_bid; } return; }

/\*\* Function to search table \*\*/

Search\_Table(bid\_size,table\_size) {

while (  $i <= table\_size$  ) {

If (search\_table(I,1) = bid\_size) then {set current\_index to i; return .true; }

else {increase i by 1 ;} return false; }

/\*\* This function gets next highest price bid \*\*/

get\_next\_unassigned \_bid(bids) {

while ( there is unassigned bid) {

if (bid\_price > max\_price) then { set max\_price to bid\_price; set bid to current\_bid; } } return; }

/\*\* Function to get ask which brings maximum improvement \*\*/

get\_opt\_ask(bid\_quantity,bid\_size) {

while ( there is unassigned ask) { call get\_next\_unassigned\_ask(asks,bid);

if ( ask\_quantity >= bid\_quantity and ask\_size >= bid\_size) then qty\_asg = bid\_quantity end if;

if ( bid\_quantity < ask \_quantity ) then qty\_asg = qty\_asg\_+ bid\_quantity ; mark ask temp\_assigned ; call cal\_obv(bid,ask,qty\_asg) ;

if ( ov > max\_imp ) then { ask\_ret = current\_ask ;
max\_imp = ov ; } return ; }

/\*\* This function gets next lowest price ask \*\*/ get\_next\_unassigned \_ask(asks,bid) {

while ( there is unassigned and unmarked ask } {

if ( (bid\_price < min\_price) and can be assigned to bid )
then { set min\_price to ask\_price;
set ask to current\_ask; } } return; }</pre>

/ \*\* This function calculates the contribution \*\*/

cal\_obv(bid,ask,qty\_asg) { net\_surplus = ( bid\_price ask\_prce ) \* bid\_quantity;

wastage = ( ask\_size - bid\_size ) \* bid\_quantity ; call search\_table(wastage,table\_size); if found then set wastage = 0; ov = net\_surplus + wastage; return }

/\*\* This function does the assignment \*\*/
assignment(bid,ask,qty\_asg) {

assign ask to bid; quantity\_assigned = qty\_asg;

if ( qty\_asg = bid\_quantity ) then mark bid as assigned; if ( qty\_asg = ask\_quantity and ( ask\_size - bid\_size = 0 )

) then mark ask as assigned ; if ( qty\_asg < bid\_quantity ) then bid\_quantity =

bid\_quantity – qty\_asg;

 $if (qty\_asg < ask\_quantity) then ask\_quantity =$ 

ask\_quantity - qty\_asg ;

if ( ( ask\_size - bid\_size) > 0 ) then ask\_size = ask\_sizebid\_size; return ; }

**Example**: The working of the algorithm is illustrated in the following example with five asks and five bids. The assignment constraints are basically size constraints i.e. a bid can be matched with an ask of same or higher size. The wastage is 8 here. The output can be seen in Table 3.

Bid	Price	Qty	Size	Ask	Price	Qty	Size	Bid	Ask	Spread	Qty	Net Surplus	Wastage
1	187	11	8	1	101	23	8	1	1	86	11	946	0
2	181	12	4	2	109	8	12	2	1	80	12	960	4
3	173	23	8	3	121	6	12	3	4	38	4	152	0
4	161	10	12	4	135	4	8	3	5	22	19	418	0
5	157	8	8	5	151	23	8	4	2	52	8	416	0
		64				64		4	3	40	2	80	0
								5	3	36	4	144	4
								5	5	6	4	24	0
											64	3140	8

Table 3 : Bids & Asks - Optimum Solution

## 5 EXPERIMENTAL RESULTS AND DISCUSSION

The algorithm has been implemented in C++. The data sets of different sizes, with each data set consisting of ask price, quantity, ask size, bid size, bid price and bid quantity, were generated randomly. The number of elements in each data set varied from 5 to 100. The results were compared with unconditional optimum solution and some solutions obtained with the help of MATLAB package. It has also been seen that if size of ask is constant and bid sizes are variable but take few values (as in most of the practical cases), then we can ignore the wastage factor. The approach can be to obtain maximum surplus and then readjust assignment without affecting value of objective function. It can also be seen that time complexity of our algorithm is always polynomial. In this algorithm, a matching ask which generates maximum improvement can be obtained by scanning unassigned asks at any point of time. In the first instance, there will be n unassigned asks, in the next instance, there will be (n-1) unassigned asks and so on. So, in all the solution will require to scan n(n-1)/2 asks. So the time complexity will be polynomial and of order  $O(n^2)$  or O(mn). Apart from this, the demand size table and getting minimum or maximum price asks/bids can be obtained with linear time complexity. So overall the time complexity will always be polynomial. This compares quite favorably with the complexity  $O(nm + n^2 \log n)$  of algorithm presented in (Kalagnanam, 2001) apart from its simplicity for large instance of n and m. The comparison can be seen in Figure 1.

## 5.1 Continuous Call Market System Design and Implementation

The different entities in our continuous call market are buyers, sellers and auctioneers. The bids and asks are continuously submitted by buyers and sellers. At periodic intervals, asks and bids are matched to find optimum assignment of asks and bids, which is worked out by our algorithm and market clears. The payoff is computed and the process is repeated. The UML State Chart Diagram captures different states of our Continuous Call Market in Figure 2. The UML activity diagram is shown in Figure 3. The different modules in the system are:

**User Registration Module**: This module helps buyers and sellers to register with the system.

**Notification Module**: It notifies users about different activities. Once auctions are notified, buyers and sellers can submit respective bids and asks. It also notifies the result of the clearing process.

Web Interface: This interface helps buyers and



Figure 1: Comparison of Results



Figure 2: State Transition diagram of Continuous Call Market

sellers to submit bids and asks respectively. It also helps buyers and sellers to specify the attributes and constraints.

**Validation Module**: This module validates the data submitted. It returns invalid bids and asks.

**Scheduler Module**: This module schedules different activities like notification, closure of bid and ask acceptance, clearing of market, etc.

**Clearing Module**: It first formulates the matching problem depending upon the attributes and constraints specified. It then implements the matching algorithm described in section 4 of the paper. It finds out optimum assignment of asks and bids.

**Payoff Module:** It computes the payoff of buyers and sellers.

### 6 CONCLUSION AND FUTURE WORK

In this paper, we have presented design of Continuous Call Market, which can handle assignment constraints. Its main component is the market clearing algorithm, which generates optimum solution to the problem of matching of asks and bids in case of assignment constraints. This can help enterprises to procure items required as well as sell them. They can also specify the assignment constraints on different attributes, so the problem is handled on more general settings. The algorithm can also handle unconstrained cases. The future work includes extending this work to solve the problem with indivisible demand bid constraints and security component.



Figure 3: Activity diagram of Continuous Call Market

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