

Formalizing double sided auctions in Coq

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Abstract

In this paper we introduce a formal framework for analysing double sided auction mechanisms in a theorem prover. In a double sided auction multiple buyers and sellers participate for trade. Any mechanism for double sided auctions to match buyers and sellers should satisfy certain properties. For example, fairness, perceived-fairness, and individual rationality are some of the important properties and to verify them we need a formal setting. We formally define all these notions in a theorem prover. This provides us a formal setting in which we prove some useful results on matching in a double sided auction. Finally, we use this framework to verify properties of two important class of double sided auction mechanism. All the properties we discuss in this paper are completely formalized in the Coq proof assistant without adding any axiom to it.

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1 Introduction

Trading is a principal component of all modern economy. Over the century more and more complex instruments (for example, index, future, options etc.) are being introduced to trade in the financial markets. With the arrival of computer assisted trading, the volume and liquidity in the markets have improved significantly. Today all big stock exchanges use computer algorithms (matching algorithms) to match buy requests (demands) with sell requests (supplies) of traders. Computer algorithms are also used by many traders to place orders in the markets. This is known as algorithmic trading. As a result of all this the markets has become complex and large. Hence, the analysis of markets is no more feasible without the help of computers.

A potential trader (buyer or seller) places orders in the markets through a broker. These orders are matched by the stock exchange to execute trades. Most stock exchanges divide the trading activity into three main sessions known as pre-markets, continuous markets and post markets. While in the pre-markets session an opening price of a product is discovered through double sided auction. In the continuous markets session the incoming buyers and sellers are continuously matched against each other on a priority basis. In the post-markets session clearing of the remaining orders is done and a closing price is discovered.

A double sided auction mechanism allows multiple buyers and sellers to trade simultaneously [5]. In double sided auctions, auctioneer (e.g. stock exchange) collects buy and sell requests over a period. Each potential trader places the orders with a *limit price*: below which a seller will not sell and above which a buyer will not buy. The exchange at the end of this period matches these orders based on their limit prices. This entire process is completed



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46 using a matching algorithm for double sided auctions.

47 Designing algorithms for double sided auctions is a well studied topic [6, 9, 7, 10]. A major
 48 emphasis of many of these algorithms is to maximize the number of matches or maximize
 49 the profit of the auctioneer. Note that an increase in the number of matches increases the
 50 liquidity in the markets. A matching algorithm can produce a matching with a uniform price
 51 or a matching with dynamic prices. While an algorithm which clears each matched bid-ask
 52 pair at a single price is referred as uniform price algorithm. An algorithm which may clear
 53 each matched bid-ask pair at different prices is referred as dynamic price algorithm. There
 54 are other important properties besides the number of matches which are considered while
 55 evaluating the effectiveness of a matching algorithm. For example, fairness, uniform pricing,
 56 and individual rationality are some of the relevant features used to compare these matching
 57 algorithms. However, it is known that no single algorithm can possess all of these properties
 58 [9, 6].

59 In this paper, we present a formal framework to analyse double sided auctions using a
 60 theorem prover. For this work, we assume that each trader wishes to trade a single unit of
 61 product and the product is indivisible. In Section 2 we formally define the theory of double
 62 sided auctions in the Coq proof assistant. In Section 3 we define and prove some important
 63 properties of matching algorithms in double sided auctions. In particular we present a
 64 dynamic price matching algorithm which produces a maximum as well as fair matching.
 65 In Section 4 we describe a uniform price matching algorithm used for price discovery in
 66 financial markets. Moreover, we prove that it produces a matching which is maximum among
 67 all possible uniform matchings. We summarise the work in Section 5 with an overview of
 68 possible future works. The Coq formalization for this paper is available at [1].

69 2 Modeling double sided auctions

70 To formalize the notion of matching in a double sided auction we use list data structure. List
 71 is also used to define various processes that operate on a matching. However, to express the
 72 properties of these processes we need some relations on lists which are analogous to relations
 73 on multisets. In this section we formally define these relations which are further used for
 74 stating some important results on matching in a double sided auction.

75 2.1 Bid, Ask and limit price

76 An auction is a competitive event, where goods and services are sold to the highest bidders.
 77 In any double sided auction multiple buyers and sellers place their orders to buy or sell a
 78 unit of underlying product. The auctioneer then matches these buy-sell requests based on
 79 their *limit prices*. While the limit price for a buy order (i.e. *bid*), is the price above which
 80 the buyer doesn't want to buy one quantity of the item. The limit price of a sell order (i.e.
 81 *ask*), is the price below which the seller doesn't want to sell one quantity of the item. In this
 82 work we assume that each bid is a buy request for one unit of item. Similarly each ask is
 83 a sell request for one unit of item. If a trader wishes to buy or sell multiple units, he can
 84 create multiple bids or asks with different *ids*.

85 We can express bid as well ask using records containing two fields.

```
86 Record Bid: Type := Mk_bid { bp:> nat;    idb: nat }.
87 Record Ask: Type := Mk_ask { sp:> nat;    ida: nat }.
```

88 For a bid b , $(bp\ b)$ is the limit price and $(idb\ b)$ is its unique identity. Similarly for an ask
 89 a , $(sp\ a)$ is the limit price and $(ida\ a)$ is the unique identity of a . Note the use of coercion

symbol `>` in the first field of *Bid*. It declares *bp* as an implicit function which is applied to any term of type *Bid* appearing in a context where a natural number is expected. Hence, now onward we can simply use *b* instead of (*bp b*) to express the limit price of *b*. Similarly we can use *a* for the limit price of an ask *a*.

Since equality for both the fields of *Bid* as well as *Ask* is decidable (i.e. `nat: eqType`), the equality on *Bid* as well as *Ask* can also be proved decidable. This is achieved by declaring two canonical instances `bid_eqType` and `ask_eqType` which connect *Bid* and *Ask* to the `eqType`.

2.2 Matching in Double Sided Auctions

In a double sided auction (DSA), the auctioneer collects all the buy and sell requests for a fixed duration. All the buy requests can be assumed to be present in a list *B*. Similarly, all the sell requests can be assumed to be present in a list *A*. At the time of auction, the auctioneer matches bids in *B* against asks in *A*. We say a bid-ask pair (*b*, *a*) is *matchable* if $b \geq a$ (i.e. $bp\ b \geq sp\ a$). Furthermore, the auctioneer assigns a trade price to each matched bid-ask pair. This process results in a matching *M*, which consists of all the matched bid-ask pairs together with their trade prices. We define matching as a list whose entries are of type `fill_type`.

```
Record fill_type: Type:= Mk_fill {bid_of: Bid; ask_of: Ask; tp: nat}
```

In a matching *M*, a bid or an ask appears at most once. Note that there might be some bids in *B* which are not matched to any ask in *M*. Similarly there might be some asks in *A* which are not matched to any bid in *M*. The list of bids present in *M* is denoted as B_M and the list of asks present in *M* is denoted as A_M . For example, Fig. 1 shows a matching *M* between list of bids *B* and list of asks *A*. While the asks present in *A* is shown using left brackets and corresponding limit prices. The bids present in *B* is represented using right brackets and corresponding limit prices. All the matched bid-ask pair in *M* is then represented using brackets of same colors. Note that the bid with limit price 37 is not present in B_M since it is not matched to any ask in *M*.

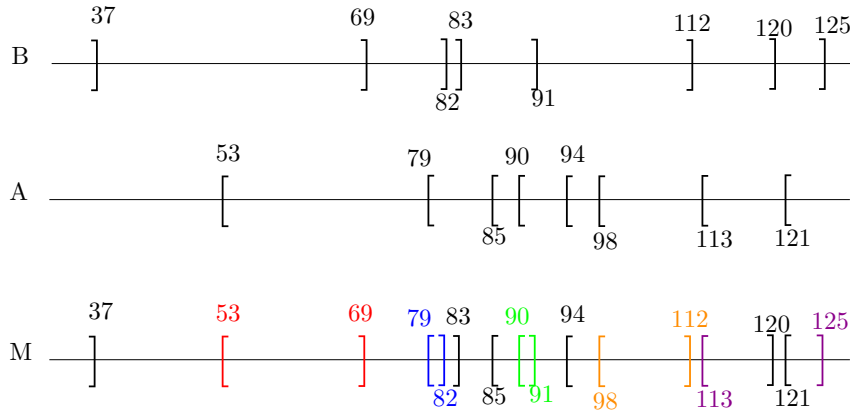


Figure 1 Bids in *B* and asks in *A* are represented using right and left brackets respectively. Every matched bid-ask pair in *M* is shown using brackets of same colors. Bids with limit prices 37, 83 and 120 are not matched to any ask in the matching *M*.

More precisely, for a given list of bids *B* and list of asks *A*, *M* is a matching iff, (1) All the bid-ask pairs in *M* are matchable, (2) B_M is duplicate-free, (3) A_M is duplicate-free, (4)

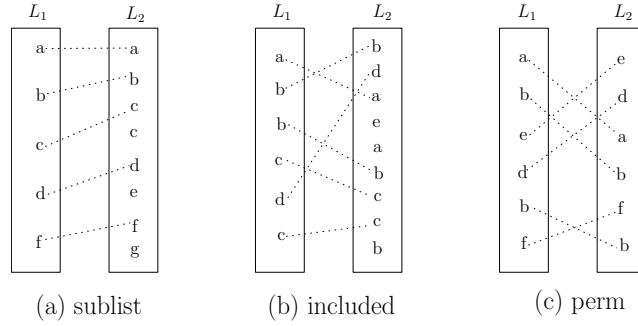
119 $B_M \subseteq B$, and (5) $A_M \subseteq A$.

120 ► **Definition 1.** $\text{matching_in } B \ A \ M := \text{All_matchable } M \wedge \text{NoDup } B_M \wedge \text{NoDup } A_M$
 121 $\wedge B_M \subseteq B \wedge A_M \subseteq A$.

122 While the term $\text{NoDup } B_M$ in above definition indicates that each bid is a request to trade
 123 one unit of item and the items are indivisible. We use the expression $B_M \subseteq B$ to denote
 124 the term $(\text{Subset } B_M \ B)$. It expresses the fact that each element in the list B_M is also
 125 present in the list B .

126 2.3 Lists, sublist and permutation

127 While the predicates NoDup and Subset are sufficient to express the notion of a matching. We
 128 need more definitions to describe the properties of matching in double sided auctions. In the
 129 following paragraphs we describe three binary relations on lists namely **sublist**, **included**
 130 and **perm** which are then used for stating important results on matching in a double sided
 131 auction.



■ **Figure 2** The dotted lines between the entries of lists confirm the presence of these entries in both the lists. (a) If L_1 is **sublist** of L_2 then no two dotted lines can intersect. (b) A list L_1 is **included** in L_2 if every entry in L_1 is also present in L_2 . (c) Two lists L_1 and L_2 are permutation of each other if each entry has same number of occurrences in both L_1 and L_2 .

132 **sublist** $L_1 \ L_2$: The notion of **sublist** is analogous to the subsequence relation on sequences.
 133 For the given lists L_1 and L_2 the term $(\text{sublist } L_1 \ L_2)$ evaluates to **true** if every entry of
 134 L_1 is also present in L_2 and they appear in the same succession. For example, in Fig. 2(a)
 135 the list L_1 is a **sublist** of L_2 since there is a line incident on each entry of L_1 and no two
 136 lines intersect each other.

137 More precisely, for any two lists l and s whose elements are of type T we have following
 138 lemmas specifying the **sublist** relation.

139 ► **Lemma 2.** $\text{sublist_intro1 } (a:T): \text{sublist } l \ s \rightarrow \text{sublist } l \ (a::s)$.

140 ► **Lemma 3.** $\text{sublist_elim3a } (a \ e:T): \text{sublist } (a::l)(e::s) \rightarrow \text{sublist } l \ s$.

141 ► **Lemma 4.** $\text{sublist_elim4}: \text{sublist } l \ s \rightarrow (\forall \ a, \text{count } a \ l \leq \text{count } a \ s)$.

142 The term $(\text{count } a \ l)$ in Lemma 4 represents the number of occurrences of element a in
 143 the list l . Note the recursive nature of **sublist** as evident in Lemma 3. It makes inductive
 144 reasoning easier for the statements which contain **sublist** in the antecedent. However, this
 145 is not true for the other relations (i.e. **included** and **perm**).

146 **included** $L_1 L_2$: A list L_1 is included in list L_2 if every entry of L_1 is also present in L_2 .
 147 The notion of **included** is analogous to the subset relation on multisets. In Fig 2(b) the list
 148 L_1 is **included** in L_2 since there is a line incident on each entry of L_1 . More precisely, we
 149 have following lemmas specifying the **included** relation.

150 ► **Lemma 5.** *included_intro: $(\forall a, \text{count } a \ l \leq \text{count } a \ s) \rightarrow \text{included } l \ s$.*

151 ► **Lemma 6.** *included_elim: $\text{included } l \ s \rightarrow (\forall a, \text{count } a \ l \leq \text{count } a \ s)$.*

152 ► **Lemma 7.** *included_intro3: $\text{sublist } l \ s \rightarrow \text{included } l \ s$.*

153 Note that if l is **sublist** of s then l is also **included** in s but not the vice versa. However,
 154 if both the lists l and s are sorted based on some ordering on type T then l is **sublist** of s
 155 whenever l is **included** in s .

156 ► **Lemma 8.** *sorted_included_sublist: $\text{Sorted } l \rightarrow \text{Sorted } s \rightarrow \text{included } l \ s \rightarrow$
 157 $\text{sublist } l \ s$.*

158 **perm** $L_1 L_2$: A list L_1 is permutation of list L_2 iff L_1 is included in L_2 and L_2 is included
 159 in L_1 . The notion of permutation for lists is similar to the equality in multisets. In Fig 2(c)
 160 the list L_1 is perm of list L_2 . We have following lemmas specifying the essential properties
 161 of the **perm** relation.

162 ► **Lemma 9.** *perm_intro: $(\forall a, \text{count } a \ l = \text{count } a \ s) \rightarrow \text{perm } l \ s$.*

163 ► **Lemma 10.** *perm_elim: $\text{perm } l \ s \rightarrow (\forall a, \text{count } a \ l = \text{count } a \ s)$.*

164 ► **Lemma 11.** *perm_sort($e: T \rightarrow T \rightarrow \text{bool}$): $\text{perm } l \ s \rightarrow \text{perm } l \ (\text{sort } e \ s)$.*

165 The term (**sort** s) in Lemma 11 represents the list s sorted using an ordering relation
 166 e . The definition of matching as a list is necessary for describing processes that operate on
 167 it. However, while describing various properties of a matching we can always consider it as
 168 a collection. For example, consider the following lemma which states that the property of
 169 being a matching is invariant over permutation.

170 ► **Lemma 12.** *match_inv: $\text{perm } M \ M' \rightarrow \text{perm } B \ B' \rightarrow \text{perm } A \ A' \rightarrow \text{matching_in}$
 171 $B \ A \ M \rightarrow \text{matching_in } B' \ A' \ M'$.*

172 We use B_M to represent the list of bids from B that are matched in M . Similarly we use
 173 notation P_M to represent a list containing trade prices of matched bid-ask pair in M . While
 174 proving various properties of a matching M we very often base our arguments solely on the
 175 information present in B_M , A_M and P_M . Therefore it is useful to have lemmas establishing
 176 the interaction of B_M , A_M and P_M with above mentioned relations on lists.

177 ► **Lemma 13.** *included_M_imp_included_bids: $\text{included } M \ M' \rightarrow \text{included } B_M \ B_{M'}$
 178 .*

179 ► **Lemma 14.** *included_M_imp_included_asks: $\text{included } M \ M' \rightarrow \text{included } A_M \ A_{M'}$*

180 ► **Lemma 15.** *included_M_imp_included_tps: $\text{included } M \ M' \rightarrow \text{included } P_M \ P_{M'}$*

181 ► **Lemma 16.** *sorted_nodup_is_sublistB: $\text{Sorted } B \rightarrow \text{Sorted } B' \rightarrow \text{NoDup } B \rightarrow$
 182 $\text{NoDup } B' \rightarrow B \subseteq B' \rightarrow \text{sublist } P_B \ P_{B'}$.*

3 Analysis of Double sided auctions

Usually in a double sided auctions mechanism, the profit of an auctioneer is the difference between the limit prices of matched bid-ask pair. In this work we do not consider analysis of profit for the auctioneer. Therefore the buyer of matched bid-ask pair pays the same amount which the seller receives. This price for a matched bid-ask pair is called the trade price for that pair. Since the limit price for a buyer is the price above which she doesn't want to buy, the trade price for this buyer is expected to be below its limit price. Similarly the limit price for a seller is the price below which he doesn't want to sell, hence the trade price for this seller is expected to be below its limit price. Therefore it is desired that in any matching the trade price of a bid-ask pair lies between their limit prices. A matching which has this property is called an *indivial rational (IR)* matching. Note that any matching can be converted to an IR matching without altering it's bid-ask pair (See Fig 3).

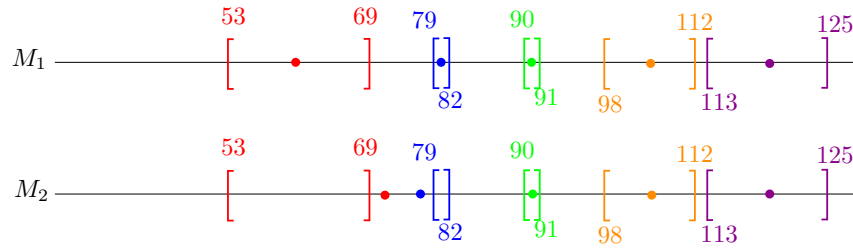


Figure 3 The colored dots represent the trade prices at which the corresponding matched bid-ask pairs are traded. While the matching M_2 is not IR since some dots lie outside the corresponding matched bid ask-pair. The matching M_1 is IR because trade prices for every matched bid-ask pair lie inside the interval. Note that the matching M_1 and M_2 contains exactly same bid-ask pairs.

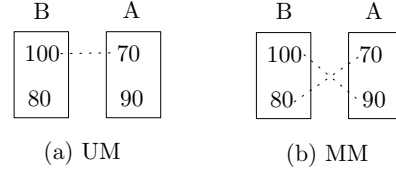
The number of matched bid-ask pairs produced by any matching algorithm is crucial in the design of a double sided auction mechanism. Increasing the number of matched bid-ask pairs increases the liquidity in market. Therefore, producing a maximum matching is an important aspect of double sided auction mechanism design. For a given list of bids B and list of asks A we say a matching M is a maximum matching if no other matching M' on the same B and A contains more number of matched bid-ask pairs than M .

Definition 17. $Is_MM\ M\ B\ A := (matching_in\ B\ A\ M) \wedge (\forall\ M',\ matching_in\ B\ A\ M' \rightarrow |M'| \leq |M|).$

In certain situations, to produce a maximum matching, different bid-ask pair must be assigned different trade prices. However, different prices for the same product in the same market simultaneously leads to dissatisfaction amongst some of the traders. A mechanism which clears all the matched bid-ask pairs at same trade price is called a *uniform matching*. It is also known as perceived-fairness (cite). In many situation it is not possible to produce an IR matching which is maximum and uniform at the same time. For example in Fig 4 a maximum matching of size two is possible but any uniform matching of size more than one is not possible.

3.1 Fairness

A bid with higher limit price is more competitive compared to bids with lower limit prices. Similarly an ask with lower limit price is more competitive compared to asks with higher limit prices. In a competitive market, like double sided auction, it is necessary to prioritise



■ **Figure 4** In this figure two bids with limit prices 100 and 80 respectively are matched against two asks of limit price 70 and 90. There is only one matching M_2 of size two possible and it is not uniform.

more competitive traders for matching. A matching which prioritise competitive traders is a fair matching. Consider the following predicates `fair_on_bids` and `fair_on_asks` which can be used to describe a fair matching.

► **Definition 18.** $\text{fair_on_bids } M B := \forall b b', \text{In } b B \wedge \text{In } b' B \rightarrow b > b' \rightarrow \text{In } b' B_M \rightarrow \text{In } b B_M.$

► **Definition 19.** $\text{fair_on_asks } M A := \forall s s', \text{In } s A \wedge \text{In } s' A \rightarrow s < s' \rightarrow \text{In } s' A_M \rightarrow \text{In } s A_M.$

► **Definition 20.** $\text{Is_fair } M B A := \text{fair_on_asks } M A \wedge \text{fair_on_bids } M B.$

Here, the predicate `fair_on_bids M B` denotes that the matching M is fair for the list of buyers B . Similarly, the predicate `fair_on_asks M A` assures that the matching M is fair for the list of sellers A . A matching which is fair on both the traders (i.e. B and A) is expressed using the predicate `Is_fair M B A`.

Unlike the uniform matching a fair matching can always be achieved without compromising the size the matching. We can accomplish this by converting any matching into a fair matching without changing its size. For example, consider the following function `make_FOB`.

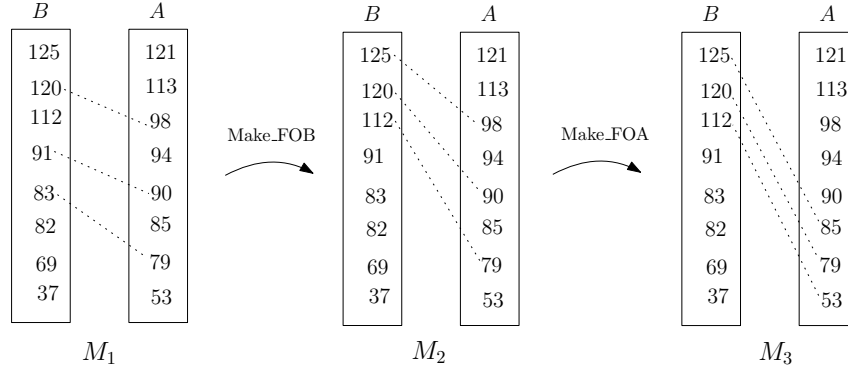
```

230 Fixpoint Make_FOB (M:list fill_type) (B: list Bid):=
231   match (M,B) with
232   | (nil,_) => nil
233   | (m:M',nil) => nil
234   | (m:M',b::B') => (Mk_fill b (ask_of m) (tp m))::(Make_FOB M' B')
235 end.
```

The function `make_FOB` produces `fair_on_bids` matching from a given matching M and a list of bids B , both sorted in decreasing order bid prices (See Fig 5). The function `make_FOB` is a recursive function and it replaces the largest bid in M with the largest bid in B . Since at any moment the largest bid in B is bigger than the largest bid in M , the new bid-ask pair is still matchable. Note that `make_FOB` doesn't change any of the ask in M and due to recursive nature of `make_FOB` on B , a bid is not repeated in the process of replacement. This ensure that the new B_M is duplicate-free. Once a matching is modified to a fair matching on bids, we use similar function `make_FOA` on this matching to produce a fair on ask matching. Hence the final result is a fair matching.

For the function `make_FOB` we have following lemma proving it fair on bids.

► **Lemma 21.** $\text{mfob_fair_on_bid } M B: (\text{Sorted } M) \rightarrow (\text{Sorted } B) \rightarrow \text{sublist } P_{B_M} P_B \rightarrow \text{fair_on_bids } (\text{Make_FOB } M B) B.$



■ **Figure 5** The dotted lines in this figure represent matched bid-ask pairs in matching M_1 , M_2 and M_3 . In the first step function `make_FOB` operates on M_1 recursively. At each step it picks the top bid-ask pair, say (b, a) in M_1 and replaces the bid b with most competitive bid available in B . The result of this process is a `fair_on_bids` matching M_2 . In a similar way the function `make_FOA` changes M_2 into a fair on ask matching M_3 .

248 ► **Lemma 22.** `mfob_fair_on_ask M A: (Sorted M) -> (Sorted A) -> sublist PAM`
 249 `PA -> fair_on_asks (Make_FOA M A) A.`

250 ► **Theorem 23.** `exists_fair_matching (Nb: NoDup B)(Na: NoDup A): matching_in`
 251 `B A M -> (∃ M', matching_in B A M' ∧ Is_fair M' B A ∧ |M| = |M'|).`

252 Proof of Theorem 23 depends on Lemma 22 and Lemma 21. Furthermore, Lemma 22
 253 and Lemma 21 can be proved using induction on M .

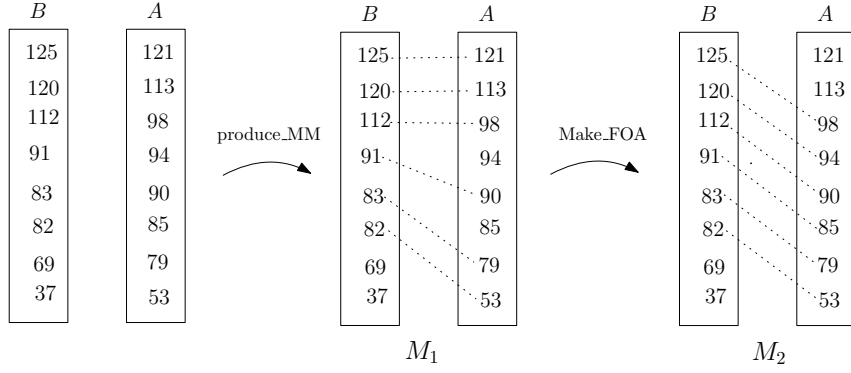
254 3.2 Maximum Matching

255 The liquidity in any market is a measure of how quickly one can trade in the market without
 256 much cost. A highly liquid market boosts the investor's confidence in the market. One way
 257 to increase the liquidity in a double sided auction is to maximize the number of matched
 258 bid-ask pair. In the previous section we have seen that any matching can be changed to a
 259 fair matching without altering its size. Therefore, we can have a maximum matching without
 260 compromising on the fairness of the matching. In this section we describe a matching which
 261 fair as well as maximum. For a given bid B and ask A , a maximum and fair matching can
 262 achieved in two steps. In first we have function `produce_MM` which produce a matching
 263 which is maximum and fair on bids. In the next step we apply `make_FOA` to this maximum
 264 matching to produce a fair on ask matching (See Fig 6).

```

265 Fixpoint produce_MM (B:list Bid) (A: list Ask): (list fill_type) :=
266   match (B, A) with
267   | (nil, _) => nil
268   | (b::B', nil) => nil
269   | (b::B', a::A') => match (a <= b) with
270     | true => ({|bid_of:= b; ask_of:= a; tp:=(bp b)|})::(produce_MM B' A')
271     | false => produce_MM B A'
272   end
273 end.
```

274 At each iteration the above function generates a matchable bid-ask pair (See Fig 6). Due
 275 to the recursive nature of function `produce_MM` on both B and A , it never pair any bid with



■ **Figure 6** In the first step, the function `produce_MM` operates recursively on the list of bids B and list of asks A . At each step the function `produce_MM` selects the most competitive available bid and then pairs it with the largest matchable ask. Note that the output of this function is fair on bid since it doesn't leave any bid from top. In the second step, the function `make_FOA` converts the M_1 into fair matching M_2 .

more than two asks. This ensures that the list of bids in matching B_M is duplicate-free. Note that the function `produce_MM` tries to match a bid until it finds a matchable ask. The function terminates when either all the bids are matched or it encounters a bid for which no matchable ask is available. Therefore, the function `produce_MM` produces a matching from a given lists of bids B and a list of asks A , both sorted in decreasing order by limit prices.

The following theorem states that when the function `produce_MM` is given a list of bids B and list of asks A , both sorted in decreasing order by limit prices, then it produces a maximum matching.

► **Theorem 24.** $\text{produce_MM_is_MM}(\text{Nb:NoDup } B)(\text{Na:NoDup } A): \text{Sorted } B \rightarrow \text{Sorted } A \rightarrow \text{Is_MM } (\text{produce_MM } B \ A) \ B \ A.$

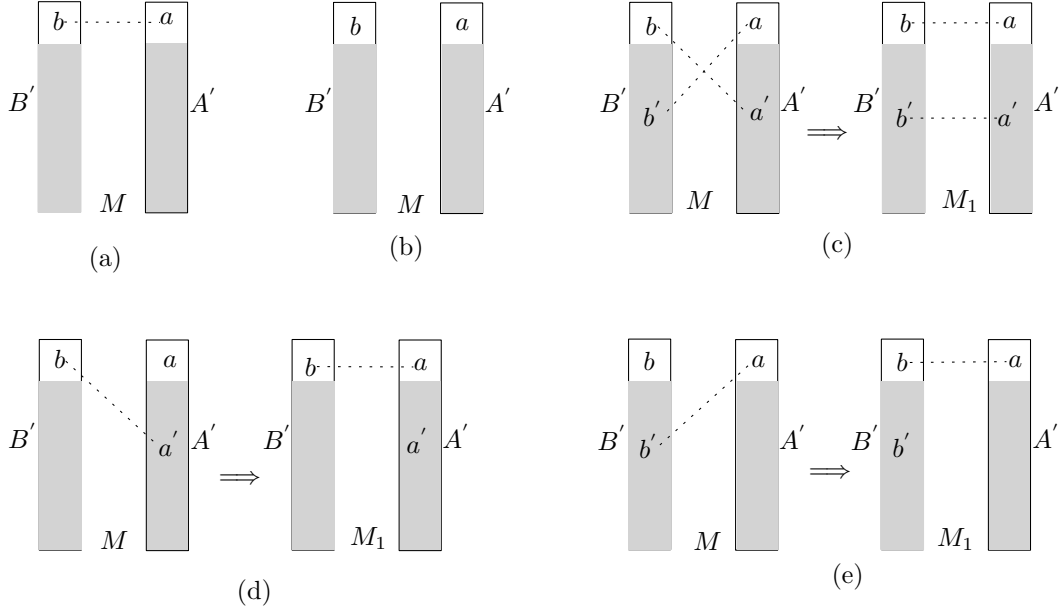
Proof: We prove this result using induction on the size of list A .

■ Induction hypothesis (IH): $\forall A', |A'| < |A| \rightarrow \forall B, \text{Sorted } B \rightarrow \text{Sorted } A' \rightarrow \text{Is_MM } (\text{produce_MM } B \ A') \ B \ A'.$

Let M be an arbitrary matching on the list of bids B and list of asks A . Moreover, assume that b and a are the topmost bid and ask present in B and A respectively (i.e. $A = (a :: A')$ and $B = (b :: B')$). We need to prove that $|M| \leq |\text{produce_MM } B \ A|$. We prove this inequality in the following two cases.

■ **Case-1** ($b < a$): In this case the limit price of a is strictly more than the limit price of b . In this case the function `produce_MM` computes a matching on B and A' . Note that due to the induction hypotheses (i.e. IH) this is a maximum matching for B and A' . Since the limit price of ask a is more than the most competitive bid b in B it cannot be present in any matching of B and A . Therefore a maximum matching on B and A' is also a maximum matching on B and A . Hence we have $|M| \leq |\text{produce_MM } B \ A|$.

■ **Case-2** ($a \leq b$): In this case the function `produce_MM` produces a matching of size $m + 1$ where m is the size of matching `produce_MM` $B' \ A'$. Hence we need to prove that $|M| \leq m + 1$. Note that due to induction hypothesis (i.e. IH) the matching `produce_MM` $B' \ A'$ is a maximum matching on B' and A' . Hence no matching on B' and A' can have size bigger than m . Without loss of generality we can assume that M is also sorted in decreasing order of bid prices. Now we further split this case into the following five sub cases (see Fig 7).



■ **Figure 7** This figure shows all the five sub cases of Case-2 (i.e. when $b \geq a$). The dotted line shows presence of the connected pair in matching M . Both the list of bids B and list of asks A are sorted in decreasing order of their limit prices. Moreover, we assume $B = b :: B'$ and $A = a :: A'$.

- 306 ■ **Case-2A** ($M = (b, a) :: M'$) : In this case bid b is matched to ask a in the matching
307 M (see Fig 7 (a)). Note that M' is a matching on B' and A' . Since $M' \leq m$ we have
308 $|M| = |M'| + 1 \leq m + 1$.
- 309 ■ **Case-2B** ($b \notin B_M \wedge a \notin A_M$) : In this case neither bid b nor ask a is present in
310 matching M (see Fig 7 (b)). Therefore M is a matching on B' and A' . Hence we have
311 $|M| \leq m < m + 1$.
- 312 ■ **Case-2C** $(b, a') \in M \wedge (b', a) \in M$: In this case we have $(b, a') \in M$ and $(b', a) \in M$
313 where $a' \in A'$ and $b' \in B'$. We can obtain another matching M_1 of same size as M
314 (see Fig 7 (c)) where $(b, a) \in M_1$ and $(b', a') \in M_1$. Note that all other entries of M_1
315 is same as M . Therefore we have $M_1 = (b, a) :: M'$ where M' is a matching on B' and
316 A' . Since $|M'| \leq m$ we have $|M| = |M_1| \leq m + 1$.
- 317 ■ **Case-2D**: $(b, a') \in M \wedge a \notin A_M$: In this case we have $(b, a') \in M$ and $a \notin A_M$ where
318 $a' \in A'$. We can obtain another matching M_1 of same size as M (see Fig 7 (d)) where
319 $(b, a) \in M_1$. Therefore we have $M_1 = (b, a) :: M'$ where M' is a matching on B' and
320 A' . Since $|M'| \leq m$ we have $|M| = |M_1| \leq m + 1$.
- 321 ■ **Case-2E**: $(b', a) \in M \wedge b \notin B_M$: In this case we have $(b', a) \in M$ and $b \notin B_M$ where
322 $b' \in B'$. We can obtain another matching M_1 of same size as M (see Fig 7 (e)) where
323 $(b, a) \in M_1$. Therefore we have $M_1 = (b, a) :: M'$ where M' is a matching on B' and
324 A' . Since $|M'| \leq m$ we have $|M| = |M_1| \leq m + 1$.

325 Note that all the cases in the above proof correspond to predicates which can be expressed
326 using membership predicate on lists. Since we have decidable equality on the elements of
327 list all these predicates are also decidable. Hence, we can do case analysis on them without
328 assuming any axiom. \square

329 Now that we proved maximality property of `produce_MM` one can produce a fair as well
330 as maximum matching by applying the functions `Make_FOA` and `Make_FOB` to the output of

331 `produce_MM`. More precisely, for a given list of bids B and list of asks A , we have following
 332 result stating that there exists a matching which is both maximum and fair.

333 ► **Theorem 25.** *exists_fair_maximum* (B : list Bid) (A : list Ask): $\exists M, (Is_fair$
 334 $M B A \wedge Is_MM M B A)$.

335 4 Matching in financial markets

336 An exchange is an organised financial market. There are various types of exchanges for
 337 example stock exchange, commodity exchange, foreign exchange etc. The job of an exchange
 338 is to facilitate trading between buyers and sellers for the products which are registered in
 339 the exchange. Many exchanges operate during a fix duration in the day. Some exchanges
 340 divide the trading activities into multiple sessions for various reasons. Most stock exchanges
 341 hold trading into two main session; pre-market (or auction session) and continuous market
 342 (or regular trading session). During the pre-market session an exchange collects all the bids
 343 and asks for fix duration and then apply double sided auction mechanism to these orders.
 344 At the end of the pre-market session an opening price for the product is discovered. During
 345 the regular sessions a bid (ask) is matched against the existing asks (bids) immediately. If
 346 the bid (ask) is not matchable it is placed in a priority queue based on its limit price.

347 The pre-market session reduces uncertainty and volatility for the regular sessions of
 348 trading. To avoid failure on behalf of the traders the exchange must match all the bids
 349 and asks within their limit prices. So any matching produced during this session must be
 350 individual rational. One of the most important aspects of the pre-market session is to discover
 351 the price of underlying product based on the total demand and supply. Furthermore, the
 352 exchange must be fair towards all the traders. So the matching produced during this session
 353 should be a fair matching. This means discovering a unique price at which maximum number
 354 of bid-ask pairs can be traded. This price is also known as equilibrium price.

355 Most exchanges matches the bids and asks during the pre-market session at equilibrium
 356 price. We describe an algorithm which produces an equilibrium price. The algorithm `UM`
 357 produces a individual rational matching which is fair and maximum amongst all uniform
 358 matchings.

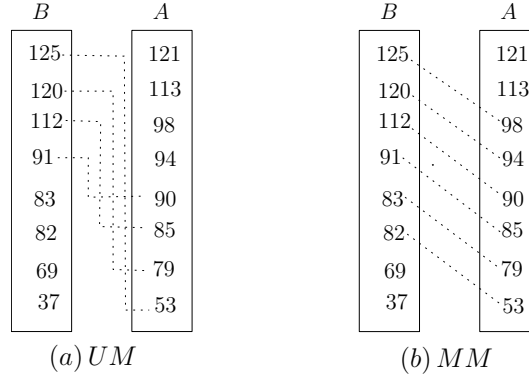
```

359 Fixpoint produce_UM (B:list Bid) (A:list Ask) :=
360   match (B,A) with
361   |(nil, _) => nil
362   |(_,nil)=> nil
363   |(b::B',a::A') => match (a <= b) with
364   |false =>nil
365   |true  => ({|bid_of:= b; ask_of:= a; tp:=(bp b)|}):produce_UM B' A'
366   end
367 end.
368 Definition uniform_price B A := bp (bid_of (last (UM_matching B A))).
369 Definition UM B A:= replace_column (UM_matching B A) (uniform_price B A).
```

370 The function `produce_UM` produces bid-ask pairs, `uniform_price` discover the uniform
 371 price and finally `UM` produces a uniform matching. The function `produce_UM` is a recursive
 372 function which matches the largest available bid in B with the smallest available ask in A at
 373 each iteration (See Fig 8). The function `produce_UM` terminates at a step when the most
 374 competitive bid available in B is not matchable with any available ask in A . The following
 375 theorem states that when the function `produce_UM` is given a list of bids B and list of asks

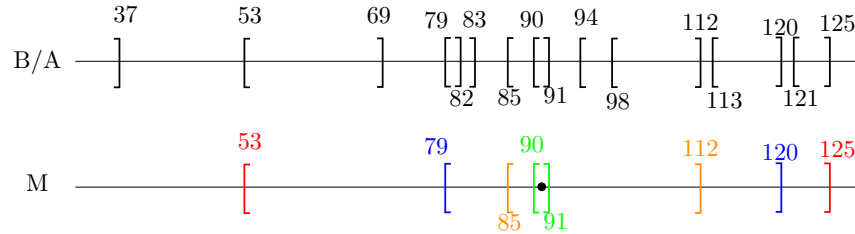
376 A , where B is sorted in decreasing order by limit prices and A is sorted in increasing order
 377 by limit prices, it produces a maximum matching among all uniform matchings.

378 ► **Theorem 26.** $UM_is_maximal_Uniform (B: list\ Bid) (A: list\ Ask): Sorted\ B \rightarrow$
 379 $Sorted\ A \rightarrow \forall M: list\ fill_type, Is_uniform\ M \rightarrow |M| \leq | (UM\ B\ A) |$



■ **Figure 8** (a) The dotted lines indicate all the bid-ask pair produced by function `produce_UM`. In each iteration function `produce_UM` matches the largest available bid in B with the smallest available ask in A . (b) The dotted lines here indicate a maximum matching for the list of bids B and list of asks A . Note that in this case the matching produced by UM is not a maximum matching.

380 **Proof:** Let M be any arbitrary IR and uniform matching on the list of bids B and list
 381 of asks A where each matched bid-ask pair is traded at price t . We need to prove that $m \leq$
 382 $| (UM\ B\ A) |$ where m is the number of matched bid-ask pairs in the matching M . Observe
 383 that in any individual rational and uniform matching the number of bids above trade price
 384 is same as the number of asks below the trade price (See Fig 9). Therefore, there are at least
 385 m bids above t and m asks below t in B and A respectively.



■ **Figure 9** Trade price p for the matching M is shown using a dot that lies between the ask with limit price 90 and bid with limit price 91. Note that since M is individual rational the number of matched asks below the trade price p is same as number of matched bids above the trade price p .

386 Since at each step the function `produce_UM` pairs the largest bid available in B with the
 387 smallest ask available in A it must produce at least m bid-ask pairs. Hence for the list of
 388 bids B and list of asks A the function UM produces a uniform matching which of size at least
 389 m . \square

390 5 Conclusion

391 There have been many attempts in past to design algorithms for double sided auctions [4].
 392 Fairness, individual rationality and uniformity are some of the essential properties on which

effectiveness of these algorithms are evaluated. Theorem provers can be a useful tool in the analysis of these properties [3, 2]. Formalizing the double sided auction in a theorem prover increases the reliability of the mechanism. There is an attempt to formalize the financial markets in theorem prover [8]. Our work is the first attempt to formalize double sided auction in a theorem prover.

We formally define various notions of double sided auctions in the Coq Proof Assistant. Moreover, we prove some general results on matching in a double sided auction. We use lists to define the notion of matching in a double sided auction. To express the properties of various processes operating on a matching we need some relations on lists. We define these relations and develop a library of facts on these relations which is further used to prove important results on matching in a double sided auction. Finally, we use this formal setting to verify properties of two important class of matching algorithms known as dynamic price algorithm and uniform price algorithm. In this work, we assume that each trader wishes to trade a single unit of product and the product is indivisible. In future this work can be extended to accomodate trades involving multiple units of item. Another direction of work could be extending this framework to include the analysis of continuous markets as well.

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