# **Bidding Languages**

Noam Nissan

October 18, 2004

Presenter:

Shahram Esmaeilsabzali

• The Problem

- The Problem
- Some Bidding Languages(OR, XOR, and etc)

- The Problem
- Some Bidding Languages(OR, XOR, and etc)
- Extension to Bidding Languages

- The Problem
- Some Bidding Languages(OR, XOR, and etc)
- Extension to Bidding Languages
- Complexity of handling bids

- The Problem
- Some Bidding Languages(OR, XOR, and etc)
- Extension to Bidding Languages
- Complexity of handling bids
- Conclusion

"Theoretically speaking, bids are simply abstract elements drawn from some space of strategies defined by the auction." [Nis04]

• Main challenge arises when dealing with combinational auctions.

- Main challenge arises when dealing with combinational auctions.
- For m items, there are  $2^m 1$  possible bids to specify.

- Main challenge arises when dealing with combinational auctions.
- For m items, there are  $2^m 1$  possible bids to specify.
- Bidding languages are meant to provide the *syntax* for encoding bids' information in a succinct, simple manner.

- Main challenge arises when dealing with combinational auctions.
- For m items, there are  $2^m 1$  possible bids to specify.
- Bidding languages are meant to provide the *syntax* for encoding bids' information in a succinct, simple manner.
- Needless to say, similar to any language, there is a trade of between expressiveness and simplicity.

• We denote the number of items in an auction by m.

- We denote the number of items in an auction by m.
- v is the valuation function. v(S) is the valuation of items in S by a bidder.

- We denote the number of items in an auction by m.
- v is the valuation function. v(S) is the valuation of items in S by a bidder.
- Notice that the value for S is assumed to be equal to v(S). In this sense, the bidding languages can be considered *valuation languages*.

- We denote the number of items in an auction by m.
- v is the valuation function. v(S) is the valuation of items in S by a bidder.
- Notice that the value for S is assumed to be equal to v(S). In this sense, the bidding languages can be considered *valuation languages*.
- Complementary: if  $v(S \cup T) > v(S) + v(T)$ .

- We denote the number of items in an auction by m.
- v is the valuation function. v(S) is the valuation of items in S by a bidder.
- Notice that the value for S is assumed to be equal to v(S). In this sense, the bidding languages can be considered *valuation languages*.
- **Complementary:** if  $v(S \cup T) > v(S) + v(T)$ .
- Substitutes: if  $v(S \cup T) < v(S) + v(T)$ .

Bidding languages basically try to efficiently model different *patterns* for bids. We are interested in bidding languages that can present useful patterns efficiently.

Bidding languages basically try to efficiently model different *patterns* for bids. We are interested in bidding languages that can present useful patterns efficiently.

A natural taxonomy for valuation functions is:

Bidding languages basically try to efficiently model different *patterns* for bids. We are interested in bidding languages that can present useful patterns efficiently.

A natural taxonomy for valuation functions is:

1. Symmetric Valuations: The valuation function v(S) depends only on |S|. All items are similar from bidders' point of view.

Bidding languages basically try to efficiently model different *patterns* for bids. We are interested in bidding languages that can present useful patterns efficiently.

A natural taxonomy for valuation functions is:

- 1. Symmetric Valuations: The valuation function v(S) depends only on |S|. All items are similar from bidders' point of view.
- 2. Asymmetric Valuations: The valuation function distinguishes between different items in the auction.

• The simple additive valuation: v(S) = |S|

- The simple additive valuation:  $\boldsymbol{v}(S) = |S|$
- The simple unit demand valuation: v(S) = 1 for all  $S \neq \emptyset$ .

- The simple additive valuation:  $\boldsymbol{v}(S) = |S|$
- The simple unit demand valuation: v(S) = 1 for all  $S \neq \emptyset$ .
- The simple K-Budget valuation:  $v(S) = \min(K, |S|)$ .

- The simple additive valuation:  $\boldsymbol{v}(S) = |S|$
- The simple unit demand valuation: v(S) = 1 for all  $S \neq \emptyset$ .
- The simple K-Budget valuation:  $v(S) = \min(K, |S|)$ .
- The Majority-valuation: v(S) = 1 if  $|S| \ge m/2$  else v(S) = 0.

- The simple additive valuation:  $\boldsymbol{v}(S) = |S|$
- The simple unit demand valuation: v(S) = 1 for all  $S \neq \emptyset$ .
- The simple K-Budget valuation:  $v(S) = \min(K, |S|)$ .
- The Majority-valuation: v(S) = 1 if  $|S| \ge m/2$  else v(S) = 0.
- The General Symmetric valuation: The price  $p_j$  specifies the price for the j'th item won and  $v(S) = \sum_{j=1}^{|S|} p_j$ .

- $\bullet$  The simple additive valuation:  $v(S) = \left|S\right|$
- The simple unit demand valuation: v(S) = 1 for all  $S \neq \emptyset$ .
- The simple K-Budget valuation:  $v(S) = \min(K, |S|)$ .
- The Majority-valuation: v(S) = 1 if  $|S| \ge m/2$  else v(S) = 0.
- The General Symmetric valuation: The price  $p_j$  specifies the price for the j'th item won and  $v(S) = \sum_{j=1}^{|S|} p_j$ .
- A Downward Symmetric valuation: A symmetric valuation where  $p_1 \ge p_2 \ge \ldots \ge p_m$ .

• An additive valuation: Value of item j is  $v^j$  and  $v(S) = \sum_{j \in S} v^j$ .

- An additive valuation: Value of item j is  $v^j$  and  $v(S) = \sum_{j \in S} v^j$ .
- The unit demand valuation: The value for item j is  $v^j$  and the bidder wants only a single item,  $v(S) = \max_{j \in S} v^j$ .

- An additive valuation: Value of item j is  $v^j$  and  $v(S) = \sum_{j \in S} v^j$ .
- The unit demand valuation: The value for item j is  $v^j$  and the bidder wants only a single item,  $v(S) = \max_{j \in S} v^j$ .
- The monochromatic valuation: There are colors blue and red, and the bidder requires items from the "same" color. Given the value for each item is 1, the valuation for any set of k blue and l red items is max(k, l).

- An additive valuation: Value of item j is  $v^j$  and  $v(S) = \sum_{j \in S} v^j$ .
- The unit demand valuation: The value for item j is  $v^j$  and the bidder wants only a single item,  $v(S) = \max_{j \in S} v^j$ .
- The monochromatic valuation: There are colors blue and red, and the bidder requires items from the "same" color. Given the value for each item is 1, the valuation for any set of k blue and l red items is max(k, l).
- The one-of-each-kind valuation: There are k pairs and l singletons (|S| = 2k + l). the bidder desires one item from each type at value 1. The valuation would be then (k + l).

• Having recognized some natural valuation patterns, we would like to have *bidding languages* that can specify such patterns efficiently and succinctly.

- Having recognized some natural valuation patterns, we would like to have *bidding languages* that can specify such patterns efficiently and succinctly.
- Often each language is good in expressing some patterns and weak or unable in expressing some other patterns.

- Having recognized some natural valuation patterns, we would like to have *bidding languages* that can specify such patterns efficiently and succinctly.
- Often each language is good in expressing some patterns and weak or unable in expressing some other patterns.
- While some classes of languages can be compared based on their expressivity, it is not always possible to accurately compare two bidding languages.

• Also called *Single Minded Bids*. [LOS99]

- Also called *Single Minded Bids*. [LOS99]
- (S, P) says that the bidder chooses S, a subset of available items, for price P:

$$v(S) = P$$

- Also called *Single Minded Bids*. [LOS99]
- (S, P) says that the bidder chooses S, a subset of available items, for price P:

$$v(S) = P$$

• For all  $S' \neq S$ , v(S') = 0.

- Also called *Single Minded Bids*. [LOS99]
- (S, P) says that the bidder chooses S, a subset of available items, for price P:

$$v(S) = P$$

- For all  $S' \neq S$ , v(S') = 0.
- Obviously, atomic bids are not expressive enough to express majority of symmetric/assymetric valuation patterns, e.g. they can not specify "simple additive valuation".

• Bidder chooses multiple bids. Bids are not necessarily disjoint.

 $S = (S_1, p_1) \operatorname{OR} (S_2, p_2) \operatorname{OR} \ldots \operatorname{OR} (S_k, p_k)$ 

• Bidder chooses multiple bids. Bids are not necessarily disjoint.

$$S = (S_1, p_1) \operatorname{OR} (S_2, p_2) \operatorname{OR} \ldots \operatorname{OR} (S_k, p_k)$$

• The valuation of bids for a certain bidder consists of only disjoint bids.

$$v(S) = \operatorname{Max}_{\mathsf{W}}(\sum_{i \in W} p_i), \quad S_i, S_j \in W \Rightarrow S_i \cap S_j = \emptyset$$

• Bidder chooses multiple bids. Bids are not necessarily disjoint.

$$S = (S_1, p_1) \operatorname{OR} (S_2, p_2) \operatorname{OR} \ldots \operatorname{OR} (S_k, p_k)$$

• The valuation of bids for a certain bidder consists of only disjoint bids.

$$v(S) = \operatorname{Max}_{\mathsf{W}}(\sum_{i \in W} \, p_i), \quad S_i, S_j \in W \Rightarrow S_i \cap S_j = \emptyset$$

• The real problem with OR bids is that they don't support substitutability.

• It is not to possible to express *simple unit demand valuation*.

- It is not to possible to express *simple unit demand valuation*.
- Consider the following OR bid:

$$S = \{ \, (\{1\}, 5\$) \, \operatorname{OR}\, (\{2\}, 4\$) \, \operatorname{OR}\, (\{1, 2\}, 7\$)) \, \}$$

- It is not to possible to express *simple unit demand valuation*.
- Consider the following OR bid:

$$S = \{ \, (\{1\}, 5\$) \, \operatorname{OR}\, (\{2\}, 4\$) \, \operatorname{OR}\, (\{1, 2\}, 7\$)) \, \}$$

The bidder is interested in having both items 1 and 2, only if she pays 7\$.

- It is not to possible to express *simple unit demand valuation*.
- Consider the following OR bid:

 $S = \{ \, (\{1\}, 5\$) \, \operatorname{OR}\, (\{2\}, 4\$) \, \operatorname{OR}\, (\{1, 2\}, 7\$)) \, \}$ 

The bidder is interested in having both items 1 and 2, only if she pays 7\$.

We can't express such constraints with OR bids.

XOR is expressive enough to express all valuation variations. Given a XOR bid:

$$S = (S_1, p_1)$$
 XOR  $(S_2, p_2)$  XOR  $\dots$  XOR  $(S_k, p_k)$ 

XOR is expressive enough to express all valuation variations. Given a XOR bid:

$$S = (S_1, p_1)$$
 XOR  $(S_2, p_2)$  XOR  $\dots$  XOR  $(S_k, p_k)$ 

the valuation function is as follows:

$$v(S) = \max_{i \mid s_i \subseteq S} p_i$$

• While XOR is more expressive than OR, there are valuations that can be specified more succinctly by OR.

- While XOR is more expressive than OR, there are valuations that can be specified more succinctly by OR.
- Consider the simple additive valuation. OR can specify that in size m while XOR require  $2^m$  clauses.

- While XOR is more expressive than OR, there are valuations that can be specified more succinctly by OR.
- Consider the simple additive valuation. OR can specify that in size m while XOR require  $2^m$  clauses.
- This motivates to combine OR and XOR. The result introduces two languages: OR-of-XORs and XOR-of-ORs.

#### **OR-of-XORs**

## **OR-of-XORs**

• Bids consist of clauses that entirely consist of XOR bids and such clauses are connected by ORs.

 $(\ldots XOR \ldots) OR (\ldots XOR \ldots) OR \ldots OR (\ldots XOR \ldots)$ 

## **OR-of-XORs**

• Bids consist of clauses that entirely consist of XOR bids and such clauses are connected by ORs.

 $(\dots XOR \dots) OR (\dots XOR \dots) OR \dots OR (\dots XOR \dots)$ 

• A downward slopping symmetric evaluation can be expressed in size  $m^2$ .

$$\begin{aligned} p_1 &\geq p_2 \\ v(s) &= \left( \left( (s, p_1) \operatorname{XOR}\left( s, p_1 \right) \right) \operatorname{OR}\left( (s, p_2) \operatorname{XOR}\left( s, p_2 \right) \right) \right) \end{aligned}$$

### **OR-of-XORs Cont'd**

### **OR-of-XORs Cont'd**

• The monochromatic valuation of m items, on the other hand, is exponential,  $2.2^{m/2}$ .

$$S = \{(s_{1r}, p_{1r}), (s_{2r}, p_{2r}), (s_{1b}, p_{1b}), (s_{2b}, p_{2b})\}$$

$$v(S) = ((s_{1r}, p_{1r}) \operatorname{XOR} (s_{2r}, p_{2r}) \operatorname{XOR} ((s_{1r} \cup s_{2r}), p_{1r} + p_{2r})$$
  
XOR ... XOR  $((s_{1b} \cup s_{2b}), p_{1b} + p_{2b})$ 

### **OR-of-XORs Cont'd**

• The monochromatic valuation of m items, on the other hand, is exponential,  $2.2^{m/2}$ .

$$S = \{(s_{1r}, p_{1r}), (s_{2r}, p_{2r}), (s_{1b}, p_{1b}), (s_{2b}, p_{2b})\}$$

$$v(S) = ((s_{1r}, p_{1r}) \operatorname{XOR} (s_{2r}, p_{2r}) \operatorname{XOR} ((s_{1r} \cup s_{2r}), p_{1r} + p_{2r})$$
  
XOR ... XOR  $((s_{1b} \cup s_{2b}), p_{1b} + p_{2b})$ 

• This motivates the bidding language XOR-of-ORs.

• Bids consist of clauses that entirely consist of OR bids and such clauses are connected by XORs.

 $(\dots OR \dots) XOR (\dots OR \dots) XOR \dots XOR (\dots OR \dots)$ 

• Bids consist of clauses that entirely consist of OR bids and such clauses are connected by XORs.

 $(\dots OR \dots) XOR (\dots OR \dots) XOR \dots XOR (\dots OR \dots)$ 

• Consider the problem of specifying monochromatic valuation of m items. This can be achieved in size m with XOR-of-ORs.

$$S = \{(s_{1r}, p_{1r}), (s_{2r}, p_{2r}), (s_{1b}, p_{1b}), (s_{2b}, p_{2b})\}$$
$$v(S) = (((s_{1r}, p_{1r}) \operatorname{OR} (s_{2r}, p_{2r})) \operatorname{XOR} ((s_{1b}, p_{1b}) \operatorname{OR} (s_{2b}, p_{2b})))$$

• Bids consist of clauses that entirely consist of OR bids and such clauses are connected by XORs.

 $(\dots OR \dots) XOR (\dots OR \dots) XOR \dots XOR (\dots OR \dots)$ 

• Consider the problem of specifying monochromatic valuation of m items. This can be achieved in size m with XOR-of-ORs.

$$S = \{(s_{1r}, p_{1r}), (s_{2r}, p_{2r}), (s_{1b}, p_{1b}), (s_{2b}, p_{2b})\}$$
$$v(S) = (((s_{1r}, p_{1r}) \operatorname{OR} (s_{2r}, p_{2r})) \operatorname{XOR} ((s_{1b}, p_{1b}) \operatorname{OR} (s_{2b}, p_{2b})))$$

• However, The K-budget valuation requires exponential time to be expressed.

# **OR/XOR Formulas**

• Alternatively, it is possible to specify the bids by applying OR and XOR on the valuation function.

- Alternatively, it is possible to specify the bids by applying OR and XOR on the valuation function.
- This approach is also capable of simulating XOR-of-ORs and OR-of-XORs, and all other bidding languages discussed so far.

- Alternatively, it is possible to specify the bids by applying OR and XOR on the valuation function.
- This approach is also capable of simulating XOR-of-ORs and OR-of-XORs, and all other bidding languages discussed so far.

$$\begin{aligned} &-(v\operatorname{XOR} u)(S) = \max(v(S), u(S)) \\ &-(v\operatorname{OR} u)(S) = \max_{R,T\subseteq S, R\cap T = \emptyset}(v(R) + u(T)) \end{aligned}$$

- Alternatively, it is possible to specify the bids by applying OR and XOR on the valuation function.
- This approach is also capable of simulating XOR-of-ORs and OR-of-XORs, and all other bidding languages discussed so far.

$$\begin{array}{l} - (v \operatorname{XOR} u)(S) = \max(v(S), u(S)) \\ - (v \operatorname{OR} u)(S) = \max_{R, T \subseteq S, R \cap T = \emptyset} (v(R) + u(T)) \end{array}$$

• OR/XOR formulas are defined recursively and provide succinct presentation of bids.

• The idea is to express XOR with OR.

- The idea is to express XOR with OR.
- This can be achieved by using dummy items.

- The idea is to express XOR with OR.
- This can be achieved by using dummy items.
- Dummy items don't have any values but can constrain bids to represent XOR.

 $\left(S_{1},p_{1}
ight)$  XOR  $\left(S_{2},p_{2}
ight)$ 

- The idea is to express XOR with OR.
- This can be achieved by using dummy items.
- Dummy items don't have any values but can constrain bids to represent XOR.

$$(S_1,p_1)$$
 XOR  $(S_2,p_2)$ 

Can be shown as:

 $(S_1 \cup \{dummy\}, p_1) \operatorname{OR} (S_2 \cup \{dummy\}, p_2)$ 

- The idea is to express XOR with OR.
- This can be achieved by using dummy items.
- Dummy items don't have any values but can constrain bids to represent XOR.

$$\left(S_{1},p_{1}
ight)$$
 XOR  $\left(S_{2},p_{2}
ight)$ 

Can be shown as:

 $(S_1 \cup \{dummy\}, p_1) \operatorname{OR} (S_2 \cup \{dummy\}, p_2)$ 

• This language can simulate all bidding languages discussed so far.

#### **OR**<sup>\*</sup> **Bids Cont'd**

# $\mathbf{OR}^* \ \textbf{Bids} \ \textbf{Cont'd}$

• It can be shown that any OR/XOR bid with s clause can be converted to an **OR**<sup>\*</sup> bid with s clause and  $s^2$  dummy bids.

# $\mathbf{OR}^*$ Bids Cont'd

- It can be shown that any OR/XOR bid with s clause can be converted to an **OR**<sup>\*</sup> bid with s clause and  $s^2$  dummy bids.
- This can be observed by considering that there are at most  $\begin{pmatrix} s \\ 2 \end{pmatrix}$  mutually exclusive bid pairs.

# **OR**<sup>\*</sup> **Bids Cont'd**

- It can be shown that any OR/XOR bid with s clause can be converted to an OR\* bid with s clause and s<sup>2</sup> dummy bids.
- This can be observed by considering that there are at most  $\begin{pmatrix} s \\ 2 \end{pmatrix}$  mutually exclusive bid pairs.
- The good thing is that we can use the systems compatible with OR bids to work with  $\mathbf{OR}^*$ .

• Natural extension of OR and XOR would be a bidding language that supports more logical operations.

- Natural extension of OR and XOR would be a bidding language that supports more logical operations.
- In [HB00], authors use CNF-like formulas (AND-of-ORs) to express bids.

- Natural extension of OR and XOR would be a bidding language that supports more logical operations.
- In [HB00], authors use CNF-like formulas (AND-of-ORs) to express bids.
- A more general approach would use all possible logical connectors.

• Another approach, in [ZBS03], uses AND on valuation function rather than on bids themselves. Authors call such ANDs, ALL.

$$v = v_1 \operatorname{ALL} v_2$$
  

$$\begin{cases} v(S) = 0, & \text{if } (v_1(S) = 0) \text{ or } (v_1(S) = 0) \\ v(S) = v_1(S) + v_2(S), & \text{otherwise} \end{cases}$$

• Another approach, in [ZBS03], uses AND on valuation function rather than on bids themselves. Authors call such ANDs, ALL.

$$v = v_1 \operatorname{ALL} v_2$$
  

$$\begin{cases} v(S) = 0, & \text{if } (v_1(S) = 0) \text{ or } (v_1(S) = 0) \\ v(S) = v_1(S) + v_2(S), & \text{otherwise} \end{cases}$$

• They use ALL along with SUM and MAX operators for expressing bids.

• Another approach, in [ZBS03], uses AND on valuation function rather than on bids themselves. Authors call such ANDs, ALL.

$$v = v_1 \operatorname{ALL} v_2 \begin{cases} v(S) = 0, & \text{if } (v_1(S) = 0) \text{ or } (v_1(S) = 0) \\ v(S) = v_1(S) + v_2(S), & \text{otherwise} \end{cases}$$

- They use ALL along with SUM and MAX operators for expressing bids.
- This approach can solve some special cases of *preference elicitation* in polynomial time.

#### **Extensions to Bidding Languages:** *k*-OR

# **Extensions to Bidding Languages:** *k*-OR

• k-OR is the generalization of OR and XOR.

#### **Extensions to Bidding Languages:** *k*-OR

- k-OR is the generalization of OR and XOR.
- An evaluation on k-OR lets at most k bids to be true.

$$v = OR_k(v_1 \dots v_t)$$
  
$$v(S) = \max_{S_1 \dots S_k} \sum_{j=1}^k v_{i_j}(S_j)$$

 $S_1 \dots S_k$  form a partition of the items and  $i_1 < i_2 \dots < i_k$ .

# **Extensions to Bidding Languages: associated prices**

# **Extensions to Bidding Languages: associated prices**

• Instead of specifying the bids' values explicitly the bidder factors out the base price (*associated prices*).

## **Extensions to Bidding Languages: associated prices**

- Instead of specifying the bids' values explicitly the bidder factors out the base price (*associated prices*).
- Let item A's value to be 101 and item B's to be, 102. They can be considered to provide at least benefit of 100 and thus can be shown as:

([(A, 1) OR (B, 2)], 100).

• Symmetric valuations often only deal with the number of items won.

- Symmetric valuations often only deal with the number of items won.
- The information for valuation  $v_1, v_2, \ldots, v_m$  can be stored by a simple vector.

- Symmetric valuations often only deal with the number of items won.
- The information for valuation  $v_1, v_2, \ldots, v_m$  can be stored by a simple vector.
- Alternatively, it is possible to store marginal values,  $p_i = v_i v_{i-1}$ .

- Symmetric valuations often only deal with the number of items won.
- The information for valuation  $v_1, v_2, \ldots, v_m$  can be stored by a simple vector.
- Alternatively, it is possible to store marginal values,  $p_i = v_i v_{i-1}$ .

• Also, it is possible to model the information by a *demand curve d*.

#### **Special Cases: Network Valuations**

### **Special Cases: Network Valuations**

• Often, the items for sale are network resources. The resources can be considered as edges of a graph where nodes represent locations of the network.

#### **Special Cases: Network Valuations**

- Often, the items for sale are network resources. The resources can be considered as edges of a graph where nodes represent locations of the network.
- Consider, as an example, the case where a bidder is interested to transfer information from node s to t. The bid then should consist of (s,t) and the proposed price p for network resource(s) for information transformation between s and t.

Complexity of bidding languages can be studied in three areas:

Complexity of bidding languages can be studied in three areas:

• **Expressions:** How efficient different bidding languages can be translated.

Complexity of bidding languages can be studied in three areas:

- **Expressions:** How efficient different bidding languages can be translated.
- Winner Determination: In general, it is independent of the bidding language and leads to NP-Complete problems.

Complexity of bidding languages can be studied in three areas:

- **Expressions:** How efficient different bidding languages can be translated.
- Winner Determination: In general, it is independent of the bidding language and leads to NP-Complete problems.
- Evaluation: It is basically about how we can extract information from bids.

Two basic types of evaluation, are:

• value query: Given a set S what is v(S)?

Two basic types of evaluation, are:

- value query: Given a set S what is v(S)?
- demand query: Given a set of items  $\{p_i\}$ , find the set that maximizes  $v(S) \sum_{i \in S} p_i$ .

Two basic types of evaluation, are:

- value query: Given a set S what is v(S)?
- demand query: Given a set of items  $\{p_i\}$ , find the set that maximizes  $v(S) \sum_{i \in S} p_i$ .

A value query can be reduced to a demand query via polynomial-time Turing reduction.

Two basic types of evaluation, are:

- value query: Given a set S what is v(S)?
- demand query: Given a set of items  $\{p_i\}$ , find the set that maximizes  $v(S) \sum_{i \in S} p_i$ .

A value query can be reduced to a demand query via polynomial-time Turing reduction.

While, for Simple and XOR bids, it is possible to provide efficient algorithm for evaluation, for other bidding languages, evaluation usually leads to the problem of item allocation which is in general NP-Complete.

• The idea is to design a language that is capable of simulating all bidding languages.

- The idea is to design a language that is capable of simulating all bidding languages.
- An idea is to submit program as valuations for the bids.

- The idea is to design a language that is capable of simulating all bidding languages.
- An idea is to submit program as valuations for the bids.
- The challenge is then what should such *programs* provide.

- The idea is to design a language that is capable of simulating all bidding languages.
- An idea is to submit program as valuations for the bids.
- The challenge is then what should such *programs* provide.
- It follows that such *program valuations* still hold the NP-Completeness property of other bidding languages.

#### Conclusion

# Conclusion

- Differen Bidding Languages provide different levels of expressivity.
- Regardless of the type of bidding language, we are dealing with NP-Complete problems of allocation and winner determination.
- In this way, it seems to me that expressivity should be the main concern for the bidding languages.
- While there seems to exist a hierarchy of languages based on their expressiveness, there is no formal specification of ideal bidding language.

# Reference

### Literatur

[HB00] Holger H. Hoos and Craig Boutilier. Solving combinatorial auctions using stochastic local search. In *Proceedings of the Seventeenth National Conference on Artificial Intelligence and Twelfth Conference on Innovative Applications of Artificial Intelligence*, pages 22–29. AAAI Press / The MIT Press, 2000.

[LOS99] Daniel J. Lehmann, Liaden Ita O'Callaghan, and Yoav Shoham. Truth revelation in approximately efficient combinatorial auctions. In *ACM Conference on Electronic Commerce*, pages 96–102, 1999. [Nis04] Noam Nisan. *Bidding Languages*. 2004.

[ZBS03] Martin A. Zinkevich, Avrim Blum, and Tuomas Sandholm. On polynomial-time preference elicitation with value queries. In *Proceedings of the 4th ACM conference on Electronic commerce*, pages 176–185. ACM Press, 2003.