# CPS 590.4 Auctions \& Combinatorial Auctions 

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## A few different 1-item auction mechanisms

- English auction:
- Each bid must be higher than previous bid
- Last bidder wins, pays last bid

Japanese auction:

- Price rises, bidders drop out when price is too high
- Last bidder wins at price of last dropout


## Dutch auction:

- Price drops until someone takes the item at that price
- Sealed-bid auctions (direct-revelation mechanisms):
- Each bidder submits a bid in an envelope
- Auctioneer opens the envelopes, highest bid wins
$\longrightarrow$ • First-price sealed-bid auction: winner pays own bid
$\rightarrow$. Second-price sealed bid (or Vickrey) auction: winner pays secondhighest bid


## Complementarity and substitutability

- How valuable one item is to a bidder may depend on whether the bidder possesses another item
- Items $a$ and $b$ are complementary if $v(\{a, b\})>$ $v(\{a\})+v(\{b\})$
- E.g.

- Items a and b are substitutes if $v(\{a, b\})<$ $v(\{a\})+v(\{b\})$
- E.g.



## Inefficiency of sequential auctions

- Suppose your valuation function is $v(\square)=$ $\$ 200, \mathrm{v}$ (ID) $=\$ 100, \mathrm{v}$ (
- Now suppose that there are two (say, Vickrey) auctions, the first one for $\square$ and the second one for
- What should you bid in the first auction (for )?
- If you bid $\$ 200$, you may lose to a bidder who bids $\$ 250$, only to find out that you could have won for $\$ 200$
- If you bid anything higher, you may pay more than \$200, only to find out that sells for \$1000
- Sequential (and parallel) auctions are inefficient


## Combinatorial auctions

Simultaneously for sale:

used in truckload transportation, industrial procurement, radio spectrum allocation, ...

## The winner determination problem (WDP)

- Choose a subset A (the accepted bids) of the bids B,
- to maximize $\Sigma_{b \text { in }} \mathrm{V}_{\mathrm{b}}$,
- under the constraint that every item occurs at most once in A
- This is assuming free disposal, i.e., not everything needs to be allocated


## WDP example

- Items A, B, C, D, E
- Bids:
- (\{A, C, D\}, 7)
- (\{B, E\}, 7)
- (\{C\}, 3)
- (\{A, B, C, E\}, 9)
- (\{D\}, 4)
- (\{A, B, C\}, 5)
- (\{B, D\}, 5)
- What's an optimal solution?
- How can we prove it is optimal?


## Price-based argument for optimality

- Items A, B, C, D, E
- Bids:
- (\{A, C, D\}, 7)
- (\{B, E\}, 7)
- (\{C\}, 3)
- (\{A, B, C, E\}, 9)
- (\{D\}, 4)
- (\{A, B, C\}, 5)
- (\{B, D\}, 5)
- Suppose we create the following "prices" for the items:
- $p(A)=0, p(B)=7$, $p(C)=3, p(D)=4$, $p(E)=0$
- Every bid bids at most the sum of the prices of its items, so we can't expect to get more than 14.


## Price-based argument does not

 always give matching upper bound - Clearly can get at most 2- Items A, B, C - If we want to set prices that
- Bids:
- (\{A, B\}, 2) sum to 2 , there must exist two items whose prices sum to <2
- (\{B, C\}, 2)
- (\{A, C\}, 2)
- But then there is a bid on those two items of value 2
- (Can set prices that sum to 3 , so that's an upper bound)

Should not be surprising, since it's an NPhard problem and we don't expect short proofs for negative answers to NP-hard problems (we don't expect NP = coNP)

## An integer program formulation

- $x_{b}$ equals 1 if bid $b$ is accepted, 0 if it is not
- maximize $\Sigma_{b} v_{b} x_{b}$
- subject to
- for each item $\mathrm{j}, \Sigma_{\mathrm{b}: \mathrm{jin} \mathrm{b}} \mathrm{x}_{\mathrm{b}} \leq 1$
- If each $x_{b}$ can take any value in [0, 1], we say that bids can be partially accepted
- In this case, this is a linear program that can be solved in polynomial time
- This requires that
- each item can be divided into fractions
- if a bidder gets a fraction $f$ of each of the items in his bundle, then this is worth the same fraction $f$ of his value $v_{b}$ for the bundle


## Price-based argument does always

 work for partially acceptable bids- Items A, B, C
- Bids:
- (\{A, B\}, 2)
- (\{B, C\}, 2)
- (\{A, C\}, 2)
- Now can get 3, by accepting half of each bid
- Put a price of 1 on each item

General proof that with partially acceptable bids, prices always exist to give a matching upper bound is based on linear programming duality

## Weighted independent set



- Choose subset of the vertices with maximum total weight,
- Constraint: no two vertices can have an edge between them
- NP-hard (generalizes regular independent set)


## The winner determination problem as a weighted independent set problem

- Each bid is a vertex
- Draw an edge between two vertices if they share an item

- Optimal allocation = maximum weight independent set
- Can model any weighted independent set instance as a CA winner determination problem ( 1 item per edge (or clique))
- Weighted independent set is NP-hard, even to solve approximately [Håstad 96] - hence, so is WDP
- [Sandholm 02] noted that this inapproximability applies to the WDP


## Dynamic programming approach to WDP [Rothkopf et al. 98]

- For every subset S of I, compute $w(S)=$ the maximum total value that can be obtained when allocating only items in $S$
- Then, $w(S)=\max \left\{\max _{i} \mathrm{v}_{\mathrm{i}}(\mathrm{S}), \max _{S^{\prime} ;} \mathrm{S}^{\prime}\right.$ is a subset of s , and there exists a bid on $\left.\mathrm{s}^{\prime} \mathrm{w}\left(\mathrm{S}^{\prime}\right)+\mathrm{w}\left(\mathrm{S} \backslash \mathrm{S}^{\prime}\right)\right\}$
- Requires exponential time


## Bids on connected sets of items in a tree

- Suppose items are organized in a tree

- Suppose each bid is on a connected set of items
- E.g. \{A, B, C, G\}, but not $\{A, B, G\}$
- Then the WDP can be solved in polynomial time (using dynamic programming) [Sandholm \& Suri 03]
- Tree does not need to be given: can be constructed from the bids in polynomial time if it exists [Conitzer, Derryberry, Sandholm 04]
- More generally, WDP can also be solved in polynomial time for graphs of bounded treewidth [Conitzer, Derryberry, Sandholm 04]
- Even further generalization given by [Gottlob, Greco 07]


## Maximum weighted matching (not necessarily on bipartite graphs)



- Choose subset of the edges with maximum total weight,
- Constraint: no two edges can share a vertex
- Still solvable in polynomial time


## Bids with few items [Rothkopf et al. 98]

- If each bid is on a bundle of at most two items,
- then the winner determination problem can be solved in polynomial time as a maximum weighted matching problem
- 3-item example:

- If each bid is on a bundle of three items, then the winner determination problem is NP-hard again


# Variants [Sandholm et al. 2002]: combinatorial reverse auction 

- In a combinatorial reverse auction (CRA), the auctioneer seeks to buy a set of items, and bidders have values for the different bundles that they may sell the auctioneer
- minimize $\Sigma_{b} \mathrm{~V}_{\mathrm{b}} \mathrm{x}_{\mathrm{b}}$
- subject to
- for each item $\mathrm{j}, \Sigma_{\mathrm{b} \text { : } \text { in } \mathrm{b}} \mathrm{x}_{\mathrm{b}} \geq 1$


## WDP example (as CRA)

- Items A, B, C, D, E
- Bids:
- (\{A, C, D\}, 7)
- (\{B, E\}, 7)
- (\{C\}, 3)
- (\{A, B, C, E\}, 9)
- (\{D\}, 4)
- (\{A, B, C\}, 5)
- (\{B, D\}, 5)


## Variants:

## multi-unit CAs/CRAs

- Multi-unit variants of CAs and CRAs: multiple units of the same item are for sale/to be bought, bidders can bid for multiple units
- Let $q_{b j}$ be number of units of item $j$ in bid $b, q_{j}$ total number of units of $j$ available/demanded
- maximize $\Sigma_{b} \mathrm{v}_{\mathrm{b}} \mathrm{x}_{\mathrm{b}}$
- subject to
- for each item $j, \Sigma_{b} q_{b j} x_{b} \leq q_{j}$
- minimize $\Sigma_{b} \mathrm{~V}_{\mathrm{b}} \mathrm{x}_{\mathrm{b}}$
- subject to
- for each item $j, \Sigma_{b} q_{b j} x_{b} \geq q_{j}$


## Multi-unit WDP example (as CA/CRA)

- Items: 3A, 2B, 4C, 1D, 3E
- Bids:
- (\{1A, 1C, 1D\}, 7)
- (\{2B, 1E\}, 7)
- (\{2C\}, 3)
- (\{2A, 1B, 2C, 2E\}, 9)
- (\{2D\}, 4)
- (\{3A, 1B, 2C\}, 5)
- (\{2B, 2D\}, 5)


## Variants: (multi-unit) combinatorial exchanges

- Combinatorial exchange (CE): bidders can simultaneously be buyers and sellers
- Example bid: "If I receive 3 units of $A$ and -5 units of $B$ (i.e., I have to give up 5 units of $B$ ), that is worth \$100 to me."
- maximize $\Sigma_{b} \mathrm{~V}_{\mathrm{b}} \mathrm{x}_{\mathrm{b}}$
- subject to
- for each item $\mathrm{j}, \Sigma_{\mathrm{b}} \mathrm{q}_{\mathrm{b}, \mathrm{j}} \mathrm{x}_{\mathrm{b}} \leq 0$


## CE WDP example

- Bids:
- (\{-1A, -1C, -1D\}, -7)
- (\{2B, 1E\}, 7)
- (\{2C\}, 3)
- (\{-2A, 1B, 2C, -2E\}, 9)
- (\{-2D\}, -4)
- (\{3A, -1B, -2C\}, 5)
- (\{-2B, 2D\}, 0)


## Variants: no free disposal

- Change all inequalities to equalities


# （back to 1－unit CAs）Expressing valuation functions using bundle bids 

－A bidder is single－minded if she only wants to win one particular bundle
－Usually not the case
－But：one bidder may submit multiple bundle bids
－Consider again valuation function $v(\mathbb{0})=$ $\$ 200, v($ 目 $)=\$ 100, v(\square)$ 回 $=\$ 500$
－What bundle bids should one place？
－What about： $\mathrm{v}($ ）$)=\$ 300, \mathrm{v}(\mathbb{\square})=\$ 200$ ， v （ 『 ）$=\$ 400$ ？

## Alternative approach: <br> report entire valuation function

- I.e., every bidder i reports $v_{i}(S)$ for every subset $S$ of I (the items)
- Winner determination problem:
- Allocate a subset $S_{i}$ of I to each bidder $i$ to maximize $\Sigma_{i} v_{i}\left(\mathrm{~S}_{\mathrm{i}}\right)$ (under the constraint that for $\mathrm{i} \neq \mathrm{j}, \mathrm{S}_{\mathrm{i}} \cap \mathrm{S}_{\mathrm{j}}=\varnothing$ )
- This is assuming free disposal, i.e., not everything needs to be allocated


## Exponentially many bundles

- In general, in a combinatorial auction with set of items I $(\|\|=m)$ for sale, a bidder could have a different valuation for every subset $S$ of $I$
- Implicit assumption: no externalities (bidder does not care what the other bidders win)
- Must a bidder communicate $2^{\mathrm{m}}$ values?
- Impractical
- Also difficult for the bidder to evaluate every bundle
- Could require $v_{i}(\varnothing)=0$
- Does not help much
- Could require: if $S$ is a superset of $S^{\prime}, v(S) \geq$ $\mathrm{v}\left(\mathrm{S}^{\prime}\right)$ (free disposal)
- Does not help in terms of number of values


## Bidding languages

- Bidding language $=$ a language for expressing valuation functions
- A good bidding language allows bidders to concisely express natural valuation functions
- Example: the OR bidding language [Rothkopf et al. 98, DeMartini et al. 99]
- Bundle-value pairs are ORed together, auctioneer may accept any number of these pairs (assuming no overlap in items)
- E.g. (\{a\}, 3) OR (\{b, c\}, 4) OR (\{c, d\}, 4) implies
- A value of 3 for $\{a\}$
- A value of 4 for $\{b, c, d\}$
- A value of 7 for $\{a, b, c\}$
- Can we express the valuation function $v(\{a, b\})=v(\{a\})=v(\{b\})$ = 1 using the OR bidding language?
- OR language is good for expressing complementarity, bad for expressing substitutability


## XORs

- If we use XOR instead of OR, that means that only one of the bundle-value pairs can be accepted
- Can express any valuation function (simply XOR together all bundles)
- E.g. (\{a\}, 3) XOR (\{b, c\}, 4) XOR (\{c, d\}, 4) implies
- A value of 3 for $\{a\}$
- A value of 4 for $\{b, c, d\}$
- A value of 4 for $\{a, b, c\}$
- Sometimes not very concise
- E.g. suppose that for any $S, v(S)=\Sigma_{\text {sin }} v(\{s\})$
- How can this be expressed in the OR language?
- What about the XOR language?
- Can also combine ORs and XORs to get benefits of both [Nisan 00, Sandholm 02]
- E.g. ((\{a\}, 3) XOR (\{b, c\}, 4)) OR (\{c, d\}, 4) implies
- A value of 4 for $\{a, b, c\}$
- A value of 4 for $\{b, c, d\}$
- A value of 7 for $\{a, c, d\}$


## WDP and bidding languages

- Single-minded bidders bid on only one bundle
- Valuation is $v$ for any subset including that bundle, 0 otherwise
- If we can solve the WDP for single-minded bidders, we can also solve it for the OR language
- Simply pretend that each bundle-value pair comes from a different bidder
- We can even use the same algorithm when XORs are added, using the following trick:
- For bundle-value pairs that are XORed together, add a dummy item to them [Fujishima et al 99, Nisan 00]
- E.g. (\{a\}, 3) XOR (\{b, c\}, 4) becomes (\{a, dummy $\left.\left.{ }_{1}\right\}, 3\right) \mathrm{OR}$ (\{b, c, dummy $\}, 4$ )
- So, we can focus on single-minded bids

