# Combinatorial Auctions 

Lecture 22

## Lecture Overview

## (1) Recap

## (2) Combinatorial Auctions

(3) Bidding Languages

## Multiunit Revenue Equivalence

## Theorem (Revenue equivalence theorem, multiunit version)

Assume that each of $n$ risk-neutral agents has an independent private valuation for a single unit of $k$ identical goods at auction, drawn from a common cumulative distribution $F(v)$ that is strictly increasing and atomless on $[\underline{v}, \bar{v}]$. Then any efficient auction mechanism in which any agent with valuation $\underline{v}$ has an expected utility of zero yields the same expected revenue, and hence results in any bidder with valuation $v_{i}$ making the same expected payment.

## Random Sampling Auction

## Definition (Random sampling optimal price auction)

The random sampling optimal price auction is defined as follows.
(1) Randomly partition the set of bidders $N$ into two sets, $N_{1}$ and $N_{2}$ (i.e., $N=N_{1} \cup N_{2} ; N_{1} \cap N_{2}=\emptyset$; each bidder has probability 0.5 of being assigned to each set).
(2) Using the procedure above find $p_{1}$ and $p_{2}$, where $p_{i}$ is the optimal single price to charge the set of bidders $N_{i}$.
(3) Then set the allocation and payment rules as follows:

- For each bidder $i \in N_{1}$, award a unit of the good if and only if $b_{i} \geq p_{2}$, and charge the bidder $p_{2}$;
- For each bidder $j \in N_{2}$, award a unit of the good if and only if $b_{j} \geq p_{1}$, and charge the bidder $p_{1}$.


## Results

## Theorem

Random sampling optimal price auctions are dominant-strategy truthful, weakly budget balanced and ex post individually rational.

## Theorem

The random sampling optimal price auction always yields expected revenue that is at least a $\left(\frac{1}{4.68}\right)$ constant fraction of the revenue that would be achieved by charging bidders the optimal single price, subject to the constraint that at least two units of the good must be sold.

## Multiunit Demand

How does VCG behave when (some) bidders may want more than a single unit of the good?

## Multiunit Demand

How does VCG behave when (some) bidders may want more than a single unit of the good?

- no longer a $k+1$ st-price auction
- instead, all winning bidders who won the same number of units will pay the same amount as each other.
- the change in social welfare from dropping any of these bidders is the same.
- Bidders who win different numbers of units will not necessarily pay the same per unit prices.
- However, bidders who win larger numbers of units will pay at least as much in total (not necessarily per unit) as bidders who won smaller numbers of units
- their impact on social welfare will always be at least as great


## Winner Determination for Multiunit Demand

- Let $m$ be the number of units available, and let $\hat{v}_{i}(k)$ denote bidder $i$ 's declared valuation for being awarded $k$ units.
- It's no longer computationally easy to identify the winners-now it's a (NP-complete) weighted knapsack problem:

$$
\begin{array}{rlr}
\text { maximize } & \sum_{i \in N} \sum_{1 \leq k \leq m} \hat{v}_{i}(k) x_{k, i} & \\
\text { subject to } & \sum_{i \in N} \sum_{1 \leq k \leq m} k \cdot x_{k, i} \leq m & \\
& \sum_{1 \leq k \leq m} x_{k, i} \leq 1 & \forall i \in N \\
& x_{k, i}=\{0,1\} \quad \forall 1 \leq k \leq m, i \in N \tag{4}
\end{array}
$$

## Winner Determination for Multiunit Demand

$$
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\end{array}
$$

- $x_{k, i}$ indicates whether bidder $i$ is allocated exactly $k$ units
- maximize: sum of agents' valuations for the chosen allocation
- (2): number of units allocated does not exceed number available
- (3): no more than one $x_{\cdot, i}$ is nonzero for any $i$
- (4): all $x$ 's must be integers


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## Valuations for heterogeneous goods

- now consider a case where multiple, heterogeneous goods are being sold.
- consider the sorts of valuations that agents could have in this case:
- complementarity: for sets $S$ and $T, v(S \cup T)>v(S)+v(T)$
- e.g., a left shoe and a right shoe
- substitutability: $v(S \cup T)<v(S)+v(T)$
- e.g., two tickets to different movies playing at the same time
- substitutability is relatively easy to deal with
- e.g., just sell the goods sequentially, or allow bid withdrawal
- complementarity is trickier...


## Fun Game

| 1 | 2 | 3 |
| :--- | :--- | :--- |
| 4 | 5 | 6 |
| 7 | 8 | 9 |

- 9 plots of land for sale, geographically related as shown
- IPV, normally distributed with mean 50 , stdev 5
- payoff:
- if you get one good other than $\# 5: v_{i}$
- any two goods: $3 v_{i}$
- any three (or more) goods: $5 v_{i}$
- Rules:
- auctioneer moves from one good to the next sequentially, holding an English auction for each good.
- bidding stops on a good: move on to the next good
- no bids for any of the 9 goods: end the auction


## Combinatorial auctions

- running a simultaneous ascending auction is inefficient
- exposure problem
- inefficiency due to fear of exposure
- if we want an efficient outcome, why not just run VCG?
- unfortunately, it again requires solving an NP-complete problem
- let there be $n$ goods, $m$ bids, sets $C_{j}$ of XOR bids
- weighted set packing problem:

$$
\begin{array}{cc}
\max & \sum_{i=1}^{m} x_{i} p_{i} \\
\text { subject to } \sum_{i \mid g \in S_{i}} x_{i} \leq 1 & \forall g \\
x_{i} \in\{0,1\} & \forall i \\
\sum_{k \in C_{j}} x_{k} \leq 1 & \forall j
\end{array}
$$

## Combinatorial auctions

$$
\begin{aligned}
\max & \sum_{i=1}^{m} x_{i} p_{i} \\
\text { subject to } & \sum_{i \mid g \in S_{i}} x_{i} \leq 1 \\
& x_{i} \in\{0,1\} \\
& \sum_{k \in C_{j}} x_{k} \leq 1
\end{aligned}
$$

- we don't need the XOR constraints
- instead, we can introduce "dummy goods" that don't correspond to goods in the auction, but that enforce XOR constraints.
- amounts to exactly the same thing: the first constraint has the same form as the third


## Winner determination problem

How do we deal with the computational complexity of the winner determination problem?

- Require bids to come from a restricted set, guaranteeing that the WDP can be solved in polynomial time
- problem: these restricted sets are very restricted...
- Use heuristic methods to solve the problem
- this works pretty well in practice, making it possible to solve WDPs with many hundreds of goods and thousands of bids.


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(1) Recap
(2) Combinatorial Auctions

3 Bidding Languages

## Expressing a bid in combinatorial auctions: OR bidding

- Atomic bid: $(S, p)$ means $v(S)=p$
- implicitly, an "AND" of the singletons in $S$
- OR bid: combine atomic bids
- let $v_{1}, v_{2}$ be arbitrary valuations

$$
\begin{aligned}
\left(v_{1} \vee v_{2}\right)(S)= & \max ^{R, T \subseteq S}\left[v_{1}(R)+v_{2}(S)\right] \\
& R \cup T=\emptyset
\end{aligned}
$$

## Theorem

OR bids can express all valuations that do not have any substitutability, and only these valuations.

## XOR Bids

- XOR bidding: allow substitutabilities
- $\left(v_{1} X O R v_{2}\right)(S)=\max \left(v_{1}(S), v_{2}(S)\right)$


## Theorem

XOR bids can represent any valuation

- this isn't really surprising, since we can enumerate valuations
- however, this implies that they don't represent everything efficiently


## Theorem

Additive valuations require linear space with $O R$, exponential space with XOR

- likewise with many other valuations: any in which the price is different for every bundle


## Composite Bidding Languages

- OR-of-XOR
- sets of XOR bids, where the bidder is willing to get either one or zero from each set
- (...XOR ...XOR ...)OR(...)OR(...)


## Theorem

Any downward sloping valuation can be represented using the OR-of-XOR language using at most $m^{2}$ atomic bids.

- XOR-of-OR
- a set of OR atomic bids, where the bidder is willing to select from only one of these sets
- generalized OR/XOR
- arbitrary nesting of OR and XOR


## The OR* Language

- OR*
- OR, but uses dummy goods to simulate XOR constraints


## Theorem

OR-of-XOR size $k \Rightarrow O R^{*}$ size $k, \leq k$ dummy goods

## Theorem

Generalized $O R / X O R$ size $k \Rightarrow O R^{*}$ size $k, \leq k^{2}$ dummy goods

## Corollary

$X O R$-of-OR size $k \Rightarrow O R^{*}$ size $k, \leq k^{2}$ dummy goods

## Advanced topics in combinatorial auctions

- iterative combinatorial auction mechanisms
- reduce the amount bidders have to disclose / communication complexity
- allow bidders to learn about each others' valuations: e.g., affiliated values
- non-VCG mechanisms for restricted valuation classes
- these can rely on polynomial-time winner determination algorithms

