

# Auctions 2

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### Efficiency of Single-Item Auctions?

**Efficiency** in single-item auctions: the item allocated to the agent who values it the most.

With independent private values (IPV):

Auction	Efficient
English (without reserve price)	yes
Japanese	yes
Dutch	no
Sealed bid second price	yes
Sealed bid first price	no

Note: Efficiency (often) lost in the correlated value setting.

# **Optimal Auctions**

### **Optimal Auction Design**

The seller's problem is to **design an auction mechanism** which has a Nash equilibrium giving him/her the **highest possible expected utility**.

assuming individual rationality

Second-prize sealed bid auction **does not maximize** expected revenue  $\rightarrow$  not the best choice if profit maximization is important (in the short term).

### Designing an Optimum Auction

We assume the IPV setting and risk-neutral bidders.

Each bidder *i*'s valuation is drawn from some **strictly increasing** cumulative density function  $F_i(v)$ , having probability density function  $f_i(v)$  that is continuous and bounded below.

• Allow  $F_i(v) \neq F_j(v)$ : **asymmetric** valuations

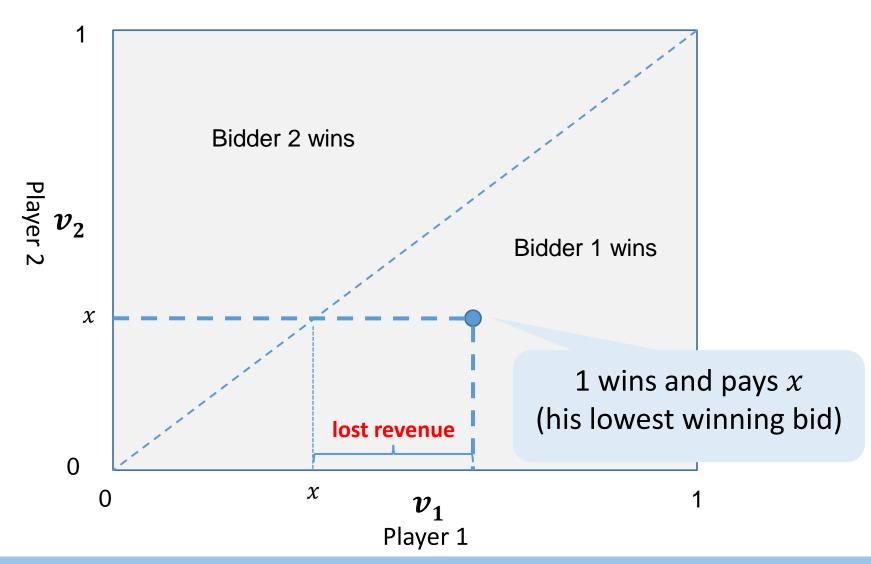
The **risk neutral** seller knows each  $F_j$  and has **zero value** for the object.

The auction that maximizes the **seller's expected revenue** subject to **individual rationality** and **Bayesian incentive-compatibility** for the buyers is an **optimal auction**.



2 bidders,  $v_i$  uniformly distributed on [0,1]. Second-price sealed bid auction.

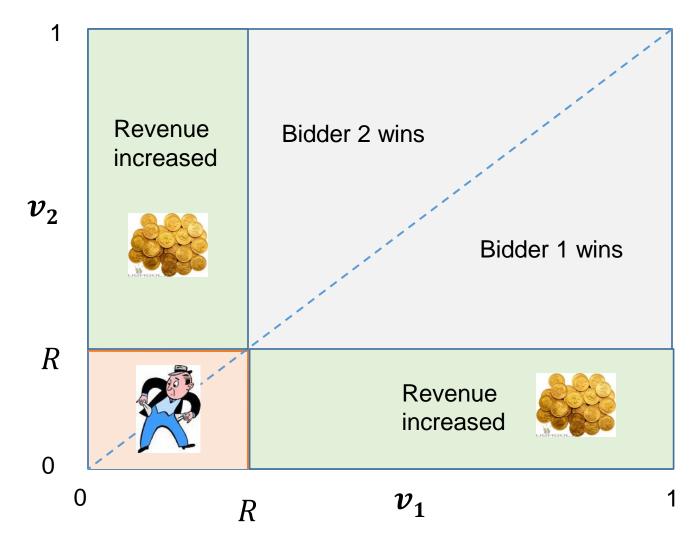
### Outcome without reserve price



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### Outcome with reserve price

#### Some reserve price improves revenue.



### Outcome with reserve price

Bidding true value is still the dominant strategy, so:

- 1. [Both bides below R]: No sale. This happens with probability  $R^2$  and then **revenue**=0
- 2. [One bid above the reserve and the other below]: Sale at **reserve price** RThis happens with probability 2(1 - R)R and the **revenue**= R
- 3. [Both bids above the reserve]: Sale at the **second highest bid**. This happens with probability  $(1 - R)^2$  and the **revenue** =  $E[\min v_i | \min v_i \ge R] = \frac{1+2R}{3}$

Expected **revenue** =  $2(1-R)R^2 + (1-R)^2 \frac{1+2R}{3}$ =  $\frac{1+3R^2-4R^3}{3}$ Maximizing:  $0 = 2R - 4R^2$ , i.e.,  $R = \frac{1}{2}$ 

### Outcome with reserve price

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Reserve price of 1/2: revenue = 5/12
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Reserve price of 0: **revenue** = 1/3 = 4/12

Tradeoffs:

- Lose the sale when both bids below 1/2: but low revenue then in any case and probability 1/4 of happening.
- Increase the sale price when one bidder has low valuation and the other high: happens with probability 1/2.

Setting a reserve price is like **adding another bidder**: it increases competition in the auction.

### **Optimal Single Item Auction**

#### **Definition (Virtual valuations)**

Consider an **IPV setting** where bidders are **risk neutral** and each bidder *i*'s valuation is drawn from some **strictly increasing** cumulative density function  $F_i(v)$ , having probability density function  $f_i(v)$ . We then define: where

- Bidder *i*'s virtual valuation is  $\psi_i(v_i) = v_i \frac{1 F_i(v_i)}{f_i(v_i)}$
- Bidder *i*'s **bidder-specific reserve price**  $r_i^*$  is the value for which  $\psi_i(r_i^*) = 0$

Example: uniform distribution over [0,1]:  $\psi(v) = 2v - 1$ 

# Example virtual valuation functions virtual valuation $\phi_1$ ф, $\phi_1(v_2)$ $\phi_1(v_1)$ valuation $v_2$ $v_1$

### **Optimal Single Item Auction**

#### **Theorem (Optimal Single-item Auction)**

The **optimal (single-good) auction** is a sealed-bid auction in which every agent is asked to **declare his valuation**. The good is sold to the agent  $i = \operatorname{argmax}_i \psi_i(\widehat{v}_i)$ , as long as  $\widehat{v}_i > r_i^*$ . If the good is sold, the winning agent i is charged the smallest valuation that it could have declared while still remaining the winner:

$$\inf\{v_i^*:\psi_i(v_i^*)\geq 0 \land \forall j\neq i, \psi_i(v_i^*)\geq \psi_j(\widehat{v}_j)\}$$

Can be understood as a second-price auction with a reserve price, held **in virtual valuation space** rather than in the space of actual valuations.

Remains dominant-strategy truthful.

### Second-Price Auction with Reservation Price

**Symmetric case**: second-price auction with reserve price  $r^*$ satisfying:  $\psi(r^*) = r^* - \frac{1-F(r^*)}{f(r^*)} = 0$ 

- Truthful mechanism when  $\psi(v)$  is non-decreasing.
- Uniform distribution over [0, p]: optimum reserve price = p/2.

Second-price sealed bid auction with Reserve Price is **not efficient**!

### Second-Price Auction with Reservation Price

Why does this increase revenue?

- Reservation prices are like competitors: increase the payments of winning bidders.
- The virtual valuation can increase the impact of weak bidders' bids, making the more competitive.
- Bidders with higher expected valuations bid more aggressively.

### **Optimal Auctions: Remarks**

#### For **optimal revenue** one needs to **sacrifice** some **efficiency**.

#### Optimal auctions are not **detail-free:**

- they require the seller to incorporate information about the bidders' valuation distributions into the mechanism
- $\rightarrow$  rarely used in practice

Theorem (Bulow and Klemperer): *revenue* of an efficiencymaximizing auction with *k*+1 bidder is at least as high as that of the revenue-maximizing one with *k* bidders.

→ better to spend energy on attracting more bidders

### **Combinatorial Auctions**

Auctions for **bundles of goods.** 

Let  $\mathcal{G} = \{g_1, \dots, g_n\}$  be a set of items (goods) to be auctioned

A valuation function  $v_i: 2^{\mathcal{G}} \mapsto \mathbb{R}$  indicates how much a bundle  $G \subseteq \mathcal{G}$  is worth to agent *i*.

We typically assume the following properties:

- normalization:  $v(\emptyset) = 0$
- free disposal:  $G_1 \subseteq G_2$  implies  $v(G_1) \leq v(G_2)$

### Non-Additive Valuations

Combinatorial auctions are interesting when the valuation function is **not additive.** 

Two main types on non-additivity.

#### **Substitutability**

The valuation function v exhibits **substitutability** if there exist two sets of goods  $G_1, G_2 \subseteq G$  such that  $G_1 \cap G_2 = \emptyset$  and  $v(G_1 \cup G_2) < v(G_1) + v(G_2)$ . Then this condition holds, we say that the valuation function v is **subadditive**.

Ex: Two different brands of TVs.

#### Complementarity

The valuation function v exhibits **complementarity** if there exist two sets of goods  $G_1, G_2 \subseteq G$  such that  $G_1 \cap G_2 = \emptyset$  and  $v(G_1 \cup G_2) >$   $v(G_1) + v(G_2)$ . Then this condition holds, we say that the valuation function v is **superadditive**.

Ex: Left and right shoe.

# How to Sell Goods with Non-Additive Valuations?

1. Ignore valuations dependencies and sell sequentially via a sequence of **independent single-item** auctions.

→ Exposure problem: A bidder may bid aggressively for a set of goods in the hope of winning a bundle but, only succeed in winning a subset (a thus paying to much)

- 2. Run separate but **connected single-item** auctions **simultaneously**.
  - a bidder bids in one auction he has a reasonably good indication of what is transpiring in the other auctions of interest.

3. Combinatorial auction: bid directly on a bundle of goods.,

### Allocation in Combinatorial Auction

**Allocation** is a list of sets  $G_1, ..., G_n \subseteq G$ , one for each agent i such that  $G_i \cap G_j = \emptyset$  for all  $i \neq j$  (i.e. not good allocated to more than one agent)

Which way to choose an allocation for a combinatorial auction?

→ The simples is to maximize **social welfare (efficient allocation)**:  $U(G_1, ..., G_n, v_1, ..., v_n) = \sum_{i=1}^n v_i(G_i)$ 

### Simple Combinatorial Auction Mechanism

The mechanism determines the **social welfare maximizing allocation** and then **charges** the winners their **bid** (for the bundle they have won), i.e.,  $\rho_i = \hat{v}_i$ .

Example:

Bidder 1	Bidder 2	Bidder 3
$v_1(x,y) = 100$	$v_2(x) = 75$	$v_3(y) = 40$
$v_1(x) = v_1(y) = 0$	$v_2(x,y) = v_2(y) = 0$	$v_3(x,y) = v_3(x) = 0$

Is this incentive-compatible? No.

A Vickrey–Clarke–Groves (VCG) auction is a type of sealed-bid auction of multiple items. Bidders submit bids that report their valuations for the items, without knowing the bids of the other bidders. The auction system assigns the items in a <u>socially</u> <u>optimal</u> manner: it charges each individual the harm they cause to other bidders.<sup>[1]</sup>

Vickrey–Clarke–Groves (VCG) auction, an analogy to secondprice sealed bid single-unit auctions, exists for the combinatorial setting and it is dominant-strategy truthful and efficient.

### VCG example

Suppose two apples are being auctioned among three bidders.

- Bidder A wants one apple and is willing to pay \$5 for that apple.
- Bidder B wants one apple and is willing to pay \$2 for it.
- Bidder C wants two apples and is willing to pay \$6 to have both of them but is uninterested in buying only one without the other.

First, the outcome of the auction is determined by maximizing social welfare:

- the apples go to bidder A and bidder B, since their combined bid of \$5 + \$2 = \$7 is greater than the bid for two apples by bidder C who is willing to pay only \$6.
- Thus, after the auction, the value achieved by bidder A is \$5, by bidder B is \$2, and by bidder C is \$0 (since bidder C gets nothing).

Payment of bidder A:

- an auction that excludes bidder A, the social-welfare maximizing outcome would assign both apples to bidder C for a total social value of \$6.
- the total social value of the original auction *excluding A's value* is computed as \$7 \$5 = \$2.
- Finally, subtract the second value from the first value. Thus, the payment required of A is \$6 \$2 = \$4.

Payment of bidder **B**:

- the best outcome for an auction that excludes bidder B assigns both apples to bidder C for \$6.
- The total social value of the original auction *minus B's portion* is \$5. Thus, the payment required of B is \$6 \$5 = \$1.

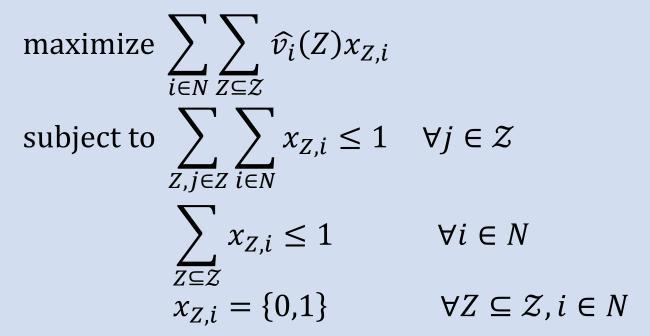
Finally, the payment for bidder C is ((\$5 + \$2) - (\$5 + \$2)) = \$0.

After the auction, A is \$1 better off than before (paying \$4 to gain \$5 of utility), B is \$1 better off than before (paying \$1 to gain \$2 of utility), and C is neutral (having not won anything).

### Winner Determination Problem

#### Definition

The **winner determination problem** for a combinatorial auctions, given the agents' declared valuations  $\hat{v}_i$  is to find the social**welfare-maximizing allocation** of goods to agents. This problem can be expressed as the following integer program



### Complexity of the Winner Determination Problem

# Equivalent to a **set packing problem** (SSP) which is known to be **NP-complete**.

Worse: SSP cannot be **approximated uniformly** to a fixed constant.

#### Two possible solutions:

- Limit to instance where polynomial-time solutions exist.
- Heuristic methods that drop the guarantee of polynomial runtime, optimality or both.

### **Bid Representation**

The problem: How to **encode** the bid (i.e. the valuation function) in a **succinct** (polynomial-size) form?

**Expressivity vs. conciseness.** 

Bidding for just one particular subset of goods.

An **atomic bid** is a pair (S, p) indicating the agent is willing to pay **price** p for the subset of **goods** S.

Example: The agents wants to pay \$100 for a bundle of a TV and a gaming console.

Very **limited expressive power**: not even the basic **additive valuation** function can be represented.

### OR bids

More expressive than atomic bids.

**OR bid** is a **disjunction** of atomic bids

$$(S_1, p_1) \lor (S_2, p_2) \lor \cdots \lor (S_k, p_k)$$

that indicates that the agent is willing to pay a price of  $p_1$  for the subset of goods  $S_1$ , or a price of  $p_2$  for the subset of goods S2, etc.

We interpret OR as an **operator for combining valuation functions**. Let V be the space of possible valuation functions, and  $v_1, v_2 \in V$  be arbitrary valuation functions. Then we have that

$$(v_1 \lor v_2)(S) = \max_{R,T \subseteq S, R \cap T = \emptyset} (v_1(R) + v_2(T)).$$

OR bid can express additive valuations but still quite limited.

**Theorem**: OR bids can express all valuation functions that exhibit **no substitutability**, and only these.

Example:

- Let's have two goods x and y and a valuation function v(x) = v(y) = 10and v(x, y) = 15.
- This valuation function cannot be expressed an OR bid because max(v (x) + v(y), v(x, y)) = 20, i.e., the interpretation of the OR bid would ascribe the valuation of 20 to the (x, y) bundle.

### XOR bids

XOR bids are more powerful.

XOR bid is an **exclusive OR** of atomic bids  $(S_1, p_1) \oplus (S_2, p_2) \oplus \cdots \oplus (S_k, p_k)$  that indicates that the agent is willing to accept one but no more than one of the atomic bids.

The XOR operator is defined on the space of valuation functions. Let V be the space of possible valuation functions, and  $v_1, v_2 \in V$  be arbitrary valuation functions. The we have that

$$(v_1 \oplus v_2)(S) = \max(v_1(S), v_2(S))$$

Example:  $({TV, DVD}, 100) \oplus ({TV, Dish}, 150)$ .

### XOR bids expressivity

**Theorem:** XOR bids can represent **all** possible valuation functions.

But: Not every valuation function can be represented efficiently by XOR bids.

In fact, the simple **additive** valuations can be represented by short OR bids but require XOR bids of **exponential size**.

### The OR\* bidding language

We can simulate the effect of an XOR by allowing bids to include **dummy** (or **phantom**) items.

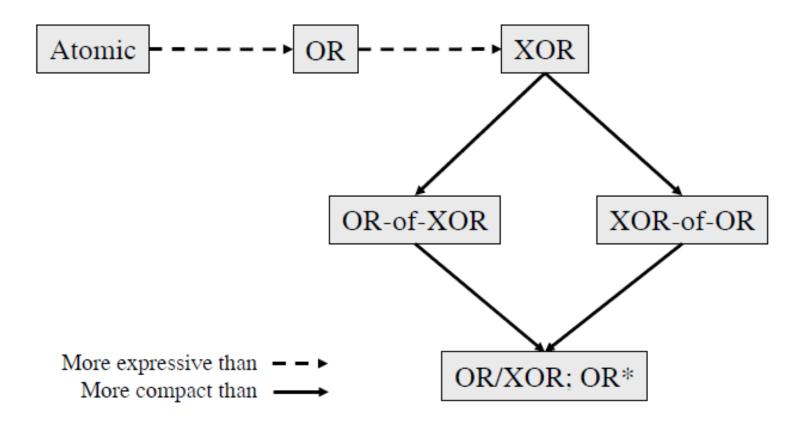
**Definition (OR\* bid)** Given a set of dummy items  $G_i$  for each agent  $i \in N$ , an OR\* bid is a **disjunction of atomic bids**  $(S_1, p_1) \lor (S_2, p_2) \lor \cdots \lor (S_k, p_k)$ , where for each l = 1, ..., k, the agent is willing to pay a price of  $p_l$  for the set of items  $S_l \subseteq G \cup G_i$ .

Example:

 $({TV, D}, 100) \lor ({DVD, D}, 100) \lor ({TV, DVD, D}, 150)$ 

OR\* can express all bids and is more succinct than OR, XOR languages and their combinations.

### Relationships between Bid Languages



However, interpretation complexity can be non-polynomial.

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### **Auctions Summary**

# Auctions are mechanisms for allocating scarce resource among self-interested agent

Mechanism-design and game-theoretic perspective

Many auction mechanisms: English, Dutch, Japanese, First-price sealed bid, Second-price sealed bid

**Desirable** properties: truthfulness, efficiency, optimality, ...

Rapidly expanding list of **applications** worth billions of dollars

Reading:

[Shoham] – Chapter 11