

LECTURE 8: Multiagent Decision Making (II)

Introduction to Multi-Agent Systems (MESIIA, MIA)

URV

Types of Agreement

- Multiagent encounters (game-like character)
- Voting.
- Coalition forming.
- Auctions (Allocating Scarce Resources)



Overview

- Allocation of scarce resources amongst a number of agents is central to multiagent systems.
- A resource might be:
 - a physical object
 - the right to use land
 - computational resources (processor, memory, . . .)
 - ..., etc.
- It is a question of supply vs demand
 - If the *resource isn't scarce*..., or if there is *no competition* for the resource...
 - Then there is no trouble allocating it
 - If there is a *greater demand than supply*
 - Then we need to determine how to allocate it



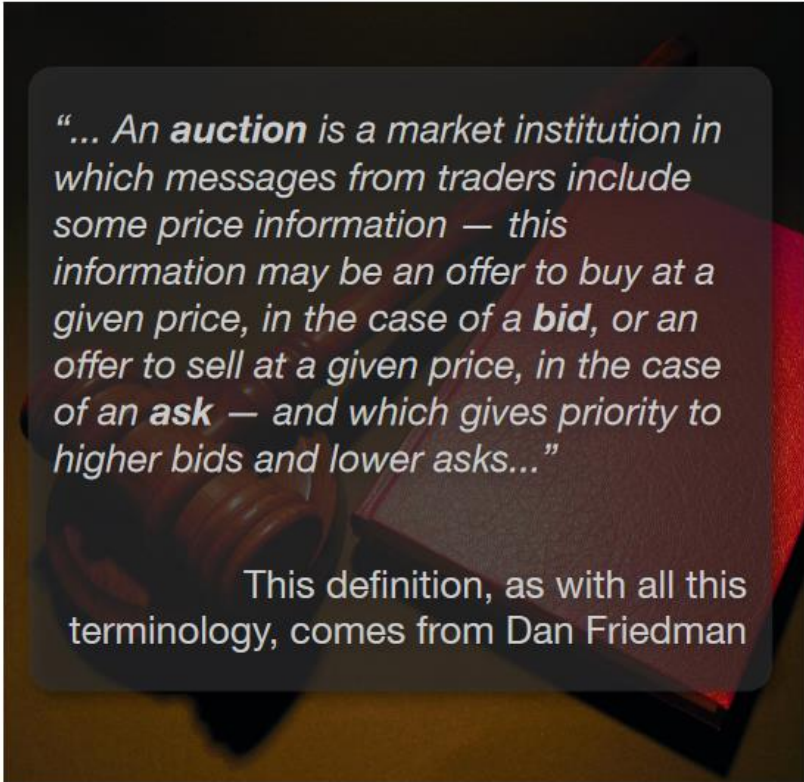
Overview

- In practice, this means we will be talking about auctions.
 - These used to be rare (and not so long ago)
 - However, auctions have grown massively with the Web/Internet
- Now feasible to auction things that weren't previously profitable:
 - eBay
 - Adword auctions



What is an auction

- Auctions are effective in allocating resources efficiently
 - They also can be used to reveal truths about bidders
- Concerned with *traders* and their allocations of:
 - Units of an indivisible *good*; and
 - Money, which is divisible.
- Assume some initial allocation.
- *Exchange* is the free alteration of allocations of goods and money between traders



*“... An **auction** is a market institution in which messages from traders include some price information — this information may be an offer to buy at a given price, in the case of a **bid**, or an offer to sell at a given price, in the case of an **ask** — and which gives priority to higher bids and lower asks...”*

This definition, as with all this terminology, comes from Dan Friedman

Types of value

- There are several models, embodying different assumptions about the nature of the good.
 - Private Value / Common Value / Correlated Value
 - With a common value, there is a risk that the winner will suffer from the *winner's curse*, where the winning bid in an auction exceeds the intrinsic value or true worth of an item
- Each trader has a value or *limit price* that they place on the good.
 - Limit prices have an effect on the behaviour of traders

Private Value

Good has an value to me that is independent of what it is worth to you.

- *For example: John Lennon's last dollar bill.*

Common Value

The good has the same value to all of us, but we have differing estimates of what it is.

- *Winner's curse.*

Correlated Value

Our values are related.

- *The more you're prepared to pay, the more I should be prepared to pay.*

Auction Protocol Dimensions

- **Winner Determination**

- Who gets the good, and what do they pay?
 - e.g. first vs second price auctions

- **Open Cry vs Sealed-bid**

- Are the bids public knowledge?
 - Can agents exploit this public knowledge in future bids?

- **One-shot vs Iterated Bids**

- Is there a single bid (i.e. one-shot), after which the good is allocated?
- If multiple bids are permitted, then:
 - Does the price ascend, or descend?
 - What is the terminating condition?



English Auction

- This is the kind of auction everyone knows.
 - Typical example is sell-side.
- Buyers call out bids, bids increase in price.
 - In some instances the auctioneer may call out prices with buyers indicating they agree to such a price.
- The seller may set a **reserve price**, the lowest acceptable price.
- Auction ends:
 - at a fixed time (internet auctions); or when there is no more bidding activity.
 - The “last man standing” pays their bid.

English Auction



Classified in the terms we used above:

- ***First-price***
- ***Open-cry***
- ***Ascending***

Around 95% of internet auctions are of this kind.
The classic use is the sale of antiques and artwork.

Susceptible to:

- ***Winner's curse***
- ***Shills***

Dutch Auction

- Also called a “descending clock” auction
 - Some auctions use a clock to display the prices.
- Starts at a **high price**, and the auctioneer calls out **descending prices**.
 - One bidder claims the good by indicating the current price is acceptable.
 - **Ties are broken** by restarting the descent from a slightly higher price than the tie occurred at.
- The winner pays the price at which they “**stop the clock**”.

Dutch Auction



Classified in the terms we used above:

- **First-price**
- **Open-cry**
- **Descending**

High volume (since auction proceeds swiftly). Often used to sell perishable goods:

- *Flowers in the Netherlands (eg. Aalsmeer)*
- *Fish in Spain and Israel.*
- *Tobacco in Canada.*

First-Price Sealed-Bid Auction

- In an English auction, you get information about how much a good is worth
 - Other people's bids tell you things about the market.
- In a **sealed bid auction**, none of that happens
 - at most you know the winning price after the auction.
- In the First-Price Sealed-Bid (FPSB) auction the **highest bid wins as always**.
 - As its name suggests, the winner pays that highest price (which is what they bid).

FPSB



Classified in the terms we used above:

- **First-price**
- **Sealed Bid**
- **One-shot**

Governments often use this mechanism to sell treasury bonds (the UK still does, although the US recently changed to Second-Price sealed Bids).

Property can also be sold this way (as in Scotland).

Vickrey Auction

- The Vickrey auction is a sealed bid auction.
 - The winning bid is the **highest bid**, but the winning bidder **pays the amount of the second highest bid**.
- This sounds odd, but it is actually a very smart design.
 - Will talk about why it works later.
- It **is in the bidders' interest to bid their true value**.
 - **incentive compatible** in the usual terminology.
- However, it is not a panacea, as the New Zealand government found out in selling radio spectrum rights
 - Due to interdependencies in the rights, that led to strategic bidding
 - one firm bid NZ\$100,000 for a license, and paid the second-highest price of only NZ\$6.

Vickrey Auction



Classified in the terms we used above:

- **Second-price**
- **Sealed Bid**
- **One-shot**

Historically used in the sale of stamps and other paper collectibles.

Why does the Vickrey auction work?

- Suppose you bid more than your valuation.
 - You may win the good.
 - If you do, you may end up paying more than you think the good is worth.
 - Not so smart.
- Suppose you bid less than your valuation.
 - You stand less chance of winning the good.
 - However, even if you do win it, you will end up paying the same.
 - Not so smart.

Proof of dominance of truthful bidding

- Let v_i be the bidding agent i 's value for an item, and b_i be the agent's bid.

- The payoff for bidder i is:

$$p_i = \begin{cases} v_i - \max_{j \neq i} b_j & \text{if } b_i > \max_{j \neq i} b_j \\ 0 & \text{otherwise} \end{cases}$$

- Assume bidder i bids $b_i > v_i$ (i.e. **overbids**)
 - If $\max_{j \neq i} b_j < v_i$, then the bidder would win whether the bid was truthful. Therefore, the strategies of bidding truthfully and overbidding have equal payoffs
 - If $\max_{j \neq i} b_j > b_i$, then the bidder would lose whether the bid was truthful. Again, both strategies have equal payoffs
 - If $v_i < \max_{j \neq i} b_j < b_i$, then the strategy of overbidding would win the action, but the payoff would be negative (as the bidder will have overpaid). A truthful strategy would pay zero.

Proof of dominance of truthful bidding

- Let v_i be the bidding agent i 's value for an item, and b_i be the agent's bid.
 - The payoff for bidder i is:
$$p_i = \begin{cases} v_i - \max_{j \neq i} b_j & \text{if } b_i > \max_{j \neq i} b_j \\ 0 & \text{otherwise} \end{cases}$$
- Assume bidder i bids $b_i < v_i$ (i.e. **underbids**)
 - If $\max_{j \neq i} b_j > v_i$, then the bidder would lose whether the bid was truthful. Therefore, the strategies of bidding truthfully and underbidding have equal payoffs
 - If $\max_{j \neq i} b_j < b_i$, then the bidder would win whether the bid was truthful. Again, both strategies have equal payoffs
 - If $b_i < \max_{j \neq i} b_j < v_i$, then only the strategy of truth-telling would win the action, with a positive payoff (as the bidder would have). An underbidding strategy would pay zero.

Collusion

- None of the auction types discussed so far are immune to collusion
 - A **grand coalition of bidders** can agree beforehand to collude
 - Propose to artificially lower bids for a good
 - Obtain true value for good
 - Share the profit
 - An auctioneer could employ bogus bidders
 - **Shills** could artificially increase bids in open cry auctions
 - Can result in **winner's curse**

Combinatorial Auctions

- A combinatorial auction is an *auction for bundles of goods*.
 - A good example of bundles of goods are spectrum licences.
 - For the 1.7 to 1.72 GHz band for Brooklyn to be useful, you need a license for Manhattan, Queens, Staten Island.
 - Most valuable are the licenses for the same bandwidth.
 - But a different bandwidth license is more valuable than no license.
 - a phone will work, but will be more expensive.



Combinatorial Auctions

- Define a set of items to be auctioned as: $\mathcal{Z} = \{z_1, \dots, z_m\}$
- Given a set of agents $Ag = \{1, \dots, n\}$ the preferences of agent i are given with the **valuation function**: $v_i: 2^{\mathcal{Z}} \rightarrow \mathbb{R}$, meaning that for every possible bundle of goods $\mathcal{Z}' \subseteq \mathcal{Z}$, $v_i(\mathcal{Z}')$ says how much \mathcal{Z}' is worth to i .
 - If that sounds to you like it would place a big requirement on agents to specify all those preferences, you would be right.
 - If $v_i(\phi) = 0$, then we say that the valuation function for i is **normalized**.
 - i.e. Agent i does not get any value from an empty allocation
- Another useful idea is free disposal, $\mathcal{Z}_1 \subseteq \mathcal{Z}_2 \Rightarrow v_i(\mathcal{Z}_1) \leq v_i(\mathcal{Z}_2)$
 - In other words, an agent is never worse off having more stuff

Allocation of Goods

- An outcome is an allocation of goods to the agents.
 - Note that we don't require all off the goods to be allocated
 - Formally an allocation is a list of sets $\mathcal{Z}_1, \dots, \mathcal{Z}_n$ one for each agent i such that $\mathcal{Z}_i \subseteq \mathcal{Z}$
 - and for all $i, j \in Ag$ such that $i \neq j$, we have $\mathcal{Z}_i \cap \mathcal{Z}_j = \phi$
 - Thus, no good is allocated to more than one agent
- The set of all allocations of \mathcal{Z} to agents Ag is: $alloc(\mathcal{Z}, Ag)$

Maximising Social Welfare

- If we design the auction, we get to say how the allocation is determined.
 - Combinatorial auctions can be viewed as different social choice functions, with different outcomes relating to different allocations of goods
 - A desirable property would be to maximize social welfare
 - i.e. maximise the sum of the utilities of all the agents.
- We can define a social welfare function:

$$sw(\underbrace{Z_1 \dots, Z_n}_{\text{allocations}}, \underbrace{v_1, \dots, v_n}_{\text{valuations}}) = \sum_{i=1}^n v_i(Z_i)$$

Defining a Combinatorial Auction

- Given this, we can define a combinatorial auction.
 - Given a set of goods \mathcal{Z} and a collection of valuation functions v_1, \dots, v_n one for each agent $i \in Ag$, the goal is to find allocation $\mathcal{Z}_1^*, \dots, \mathcal{Z}_n^*$ that maximises sw :
$$\mathcal{Z}_1^*, \dots, \mathcal{Z}_n^* = \arg \max_{(\mathcal{Z}_1, \dots, \mathcal{Z}_n) \in alloc(\mathcal{Z}, Ag)} sw(\mathcal{Z}_1, \dots, \mathcal{Z}_n, v_1, \dots, v_n)$$
- Figuring this out, i.e. solving this optimization problem, is called the **winner determination problem**

Winner Determination

- How do we do this?
- Well, we could get every agent i to declare their valuation: \hat{v}_i
 - The hat denotes that this is what the agent says, not what it necessarily is.
 - Remember that the agent may lie! 😊
- Then we just look at all the possible allocations and figure out what the best one is.
- One problem here is representation, valuations are exponential in terms of the number of items: $v_i: 2^Z \rightarrow \mathbb{R}$
 - A naive representation is impractical.
 - In a bandwidth auction with 1122 licenses we would have to specify 2^{1122} values for each bidder.
- Searching through them is computationally intractable

Bidding Languages

- Rather than exhaustive evaluations, allow bidders to construct valuations from the bits they want to mention.
 - An atomic bid β is a pair (Z', p) where $Z' \subseteq Z$
 - A bundle Z^* **satisfies** a bid (Z', p) if $Z' \subseteq Z^*$.
- In other words a bundle **satisfies** a bid if it contains at least the things in the bid.
- Atomic bids **define valuations**
$$v_{\beta}(Z^*) = \begin{cases} p & \text{if } Z^* \text{ satisfies } (Z', p) \\ 0 & \text{otherwise} \end{cases}$$
- Atomic bids alone don't allow us to construct very interesting valuations.

XOR Bids

- With XOR bids, **we pay for at most one**
 - A bid $\beta = (Z_1, p_1) XOR \dots XOR (Z_k, p_k)$ defines a valuation function v_β as follows
$$v_\beta(Z^*) = \begin{cases} 0 & \text{if } Z^* \text{ does not satisfy any } (Z_i, p_i) \\ \max\{p_i \mid Z_i \subseteq Z^*\} & \text{otherwise} \end{cases}$$
 - I pay nothing if your allocation Z^* doesn't satisfy any of my bids
 - Otherwise, I will pay the largest price of any of the satisfied bids.
- XOR bids are **fully expressive**, that is they can express any valuation function over a set of goods.
 - To do that, we may need an exponentially large number of atomic bids
 - However, the valuation of a bundle can be computed in polynomial time.

$$B_1 = (\{a, b\}, 3) XOR (\{c, d\}, 5)$$

“...I would pay 3 for a bundle that contains a and b but not c and d. I will pay 5 for a bundle that contains c and d but not a and b, and I will pay 5 for a bundle that contains a, b, c and d...”

From this we can construct the valuation:

$$\begin{aligned} v_{\beta_1}(\{a\}) &= 0 \\ v_{\beta_1}(\{b\}) &= 0 \\ v_{\beta_1}(\{a, b\}) &= 3 \\ v_{\beta_1}(\{c, d\}) &= 5 \\ v_{\beta_1}(\{a, b, c, d\}) &= 5 \end{aligned}$$

OR Bids

- With OR bids, *we are prepared to*
- *pay for more than one bundle*
 - A bid $\beta = (\mathcal{Z}_1, p_1) OR \dots OR (\mathcal{Z}_k, p_k)$ defines k valuations for different bundles
 - An allocation of goods \mathcal{Z}' is assigned given a set \mathcal{W} of atomic bids such that:
 - Every bid in \mathcal{W} is satisfied by \mathcal{Z}'
 - No goods appear in more than one bundle; i.e. $\mathcal{Z}_i \cap \mathcal{Z}_j = \emptyset$ for all i, j where $i \neq j$
 - No other subset \mathcal{W}' satisfying the above condition is better:

$$\sum_{(\mathcal{Z}_i, p_i) \in \mathcal{W}'} p_i > \sum_{(\mathcal{Z}_j, p_j) \in \mathcal{W}'} p_j$$

$$B_1 = (\{a, b\}, 3) OR (\{c, d\}, 5)$$

“...I would pay 3 for a bundle that contains a and b but not c and d. I will pay 5 for a bundle that contains c and d but not a and b, and I will pay 8 for both bundles that contain a combination of a, b, c and d...”

From this we can construct the valuation:

$$v_{\beta_1}(\{a\}) = 0$$

$$v_{\beta_1}(\{b\}) = 0$$

$$v_{\beta_1}(\{a, b\}) = 3$$

$$v_{\beta_1}(\{c, d\}) = 5$$

$$v_{\beta_1}(\{a, b, c, d\}) = 8$$

Note that the **cost of the last bundle is different to that when the XOR bid** was used

OR Bids

- Here is another example!

- $\beta_3 = OR((\{e, f, g\}, 4), (\{f, g\}, 1), (\{e\}, 3), (\{c, d\}, 4))$
- This gives us:

$$v_{\beta_3}(\{e\}) = 3$$

$$v_{\beta_3}(\{e, f\}) = 3$$

$$v_{\beta_3}(\{e, f, g\}) = 4$$

$$v_{\beta_3}(\{b, c, d, f, g\}) = 4 + 1 = 5$$

$$v_{\beta_3}(\{a, b, c, d, e, f, g\}) = 4 + 4 = 8$$

$$v_{\beta_3}(\{c, d, e\}) = 4 + 3 = 7$$

- Remember that if more than one bundle is satisfied, then you pay for each of the bundles satisfied.
- Also remember free disposal, which is why the bundle $\{e, f\}$ satisfies the bid $(\{e\}, 3)$ as the agent doesn't pay extra for f .

OR Bids

- OR bids are strictly less expressive than XOR bids
 - Some valuation functions cannot be expressed
 - E.g., $v(\{a\}) = 1, v(\{b\}) = 1, v(\{a, b\}) = 1$
- OR bids also suffer from computational complexity
 - Given an OR bid β and a bundle \mathcal{Z} , computing $v_\beta(\mathcal{Z})$ is NP-hard.

Winner Determination

- Determining the winner is a combinatorial optimisation problem (NP-hard)
 - But this is a worst case result, so it may be possible to develop approaches that are either **optimal** and run well in many cases, or **approximate** (within some bounds).
- Typical approach is to code the problem as an **integer linear program** and use a standard solver.
 - This is NP-hard in principle, but often provides solutions in reasonable time.
 - Several algorithms exist that are efficient in most cases
- Approximate algorithms have been explored
 - Few solutions have been found with reasonable bounds
- Heuristic solutions based on **greedy algorithms** have also been investigated
 - e.g. that try to find the largest bid to satisfy, then the next etc



The VCG Mechanism

- Auctions are easy to strategically manipulate
 - In general ***we don't know*** whether the agents valuations ***are true valuations***.
 - Life would be easier if they were...
 - ... so can we make them true valuations?
- Yes!
 - In a generalization of the Vickrey auction.
 - Vickrey/Clarke/Groves Mechanism
- Mechanism is incentive compatible: ***telling the truth is a dominant strategy.***

Recall that we could get every agent i to declare their valuation:

$$\hat{v}_i$$

where the hat denotes that this is what the agent says, not what it necessarily is.

- *The agent may lie!*

The VCG Mechanism

- Need some more notation.
 - **Indifferent valuation** function: $v^0(\mathcal{Z}') = 0$ for all \mathcal{Z}'
 - i.e. the value for a bid that does not care about the goods
 - sw_{-i} is the **social welfare function without i** :
$$sw_{-i}(\mathcal{Z}_1, \dots, \mathcal{Z}_n, v_1, \dots, v_n) = \sum_{j \in Ag, j \neq i} v_j(\mathcal{Z}_j)$$
 - This is how well everyone **except agent i** does from $\mathcal{Z}_1, \dots, \mathcal{Z}_n$
- And we can then define the VCG mechanism.

The VCG Mechanism

- Every agent simultaneously declares a valuation \hat{v}_i
 - Remember that this not be the actual valuation
- The mechanism computes the allocation Z_1^*, \dots, Z_n^* :

$$Z_1^*, \dots, Z_n^* = \arg \max_{(Z_1, \dots, Z_n) \in \text{alloc}(Z, Ag)} sw(Z_1, \dots, Z_n, v_1, \dots, v_n)$$

- Each agent i pays p_i
 - This is effectively a **compensation** to the other agents for their loss in utility due to i winning an allocation
 - This is the difference in social welfare to agents other than i
 - Between the outcome Z'_1, \dots, Z'_n when i does not participate
 - And the outcome Z_1^*, \dots, Z_n^* when i does participate
 - Formally: $p_i = sw_{-i}(Z'_1, \dots, Z'_n, \hat{v}_1, \dots, v^0, \dots, \hat{v}_n) - sw_{-i}(Z_1^*, \dots, Z_n^*, \hat{v}_1, \dots, \hat{v}_i, \dots, \hat{v}_n)$
 - Therefore the mechanism computes, for each agent i the allocation that maximises social welfare were that agent to have declared v^0 to be its valuation:

$$Z_1^*, \dots, Z_n^* = \arg \max_{(Z_1, \dots, Z_n) \in \text{alloc}(Z, Ag)} sw(Z_1, \dots, Z_n, v_1, \dots, v^0, \dots, v_n)$$

The VCG Mechanism

- With the VCG, each agent pays out the cost (to the other agents) of it having participated in the auction.
 - It is incentive compatible for exactly the same reason as the Vickrey auction was before.
 - No agent can benefit by declaring anything other than its true valuation
 - To understand this, think about VCG with a singleton bundle
 - The only agent that pays anything will be the agent i that has the highest bid
 - But if it had not participated, then the agent with the second highest bid would have won
 - Therefore agent i “**compensates**” the other agents by paying this second highest bid
- Therefore we get a dominant strategy for each agent that guarantees to maximise social welfare.
 - i.e. ***social welfare maximisation can be implemented in dominant strategies***

Summary

- Allocating scarce resources comes down to auctions
- We looked at a range of different simple auction mechanisms.
 - English auction
 - Dutch auction
 - First price sealed bid
 - Vickrey auction
- We looked at the popular field of combinatorial auctions
 - We discussed some of the problems in implementing combinatorial auctions.
- And we talked about the Vickrey/Clarke/Groves mechanism, a rare ray of sunshine on the problems of multiagent interaction

Readings for this week

- Chapters 14 of the book by M.Wooldridge “An introduction to Multi-Agent Systems” (2nd edition).