

# Multiagent Systems: Intro to Mechanism Design

CS 486/686: Introduction to Artificial Intelligence

# Introduction

---

---

- So far almost everything we have looked at has been in a single-agent setting
  - Today - Multiagent Decision Making!
- For participants to act optimally, they must account for how others are going to act
- We want to
  - Understand the ways in which agents interact and behave
  - **Design systems so that agents behave the way we would like them to**

**Hint for the final exam:** MAS is my main research area. I like MAS problems. I even enjoy marking MAS questions. Two of the TAs for this course do MAS research. They also like marking MAS questions. There *will* be an MAS question on the final exam.

# Mechanism Design

---

---

- Game Theory asks
  - Given a game, what should rational agents do?
- Mechanism Design asks
  - Given rational agents, what sort of games should we design?
  - Can we guarantee that agents will reach an outcome with properties **we** want

# Fundamentals

---

---

- Set of possible outcomes:  $O$
- Set of agents:  $N$ ,  $|N|=n$ 
  - Each agent has a type  $\theta_i$  from  $\Theta_i$
  - The type captures all private information relevant to the agent's decision making
- Utility functions:  $u_i(o, \theta_i)$
- Social choice function:  $f: \Theta_1 \times \dots \times \Theta_n \rightarrow O$

# Examples of Social Choice Functions

---

---

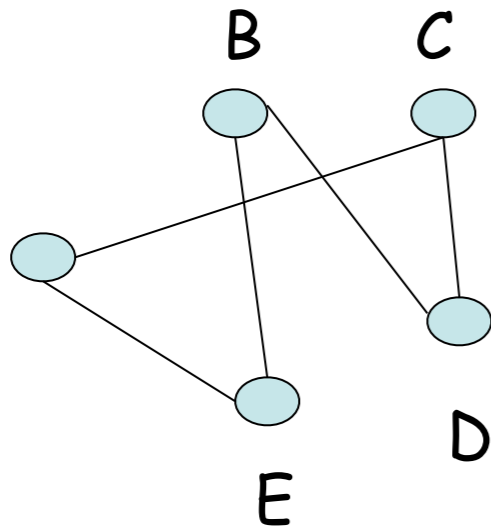
- Voting
  - Choose a candidate from amongst a group
- Public Project
  - Decide whether to build a road whose cost must be funded by the agents themselves
- Allocation
  - Allocate an item or resource to one agent in the group

# Scenario

---

---

- Network routing problem to allocate resources to minimize the total cost of delay over all agents



My unit cost of delay for sending messages from A to D is \$1



My unit cost of delay for sending messages between E and D is \$5

# A Potential Problem

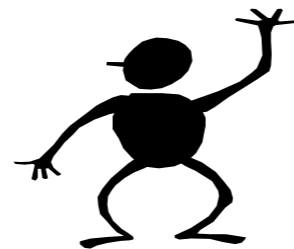
---

---

- Agents' types are not public, and agents are acting in their own self-interest



I like the  
bear the  
most!



No, I do!

# Mechanism Design Problem

---

---

- By having agents interact through an “institution” we might be able to solve this problem
- Mechanism

$$M = (S_1, \dots, S_n, g(\cdot))$$

- $S_i$  is the strategy space of agent  $i$
- $g: S_1 \times \dots \times S_n \rightarrow O$  is the outcome function



# Implementation

---

---

- A mechanism  $M=(S_1, \dots, S_n, g())$  implements social choice function  $f(\theta)$  if there is an equilibrium  $s^*$

$$s^* = (s_1^*(\theta_1), \dots, s_n^*(\theta_n))$$

such that

$$g(s_1^*(\theta_1), \dots, s_n^*(\theta_n)) = f(\theta_1, \dots, \theta_n)$$

for all

$$(\theta_1, \dots, \theta_n) \in \Theta_1 \times \Theta_n$$

# Direct Mechanisms

---

---

- A direct mechanism is a mechanism where

$$S_i = \Theta_i \text{ for all } i$$

and

$$g(\theta) = f(\theta) \text{ for all } \theta \in \Theta_1 \times \Theta_n$$

# Incentive Compatibility

---

---

- A direct mechanism is incentive compatible if it has an equilibrium  $s^*$  where

$$s_i^*(\theta_i) = \theta_i$$

for all  $\theta_i$  in  $\Theta_i$  and for all  $i$ .

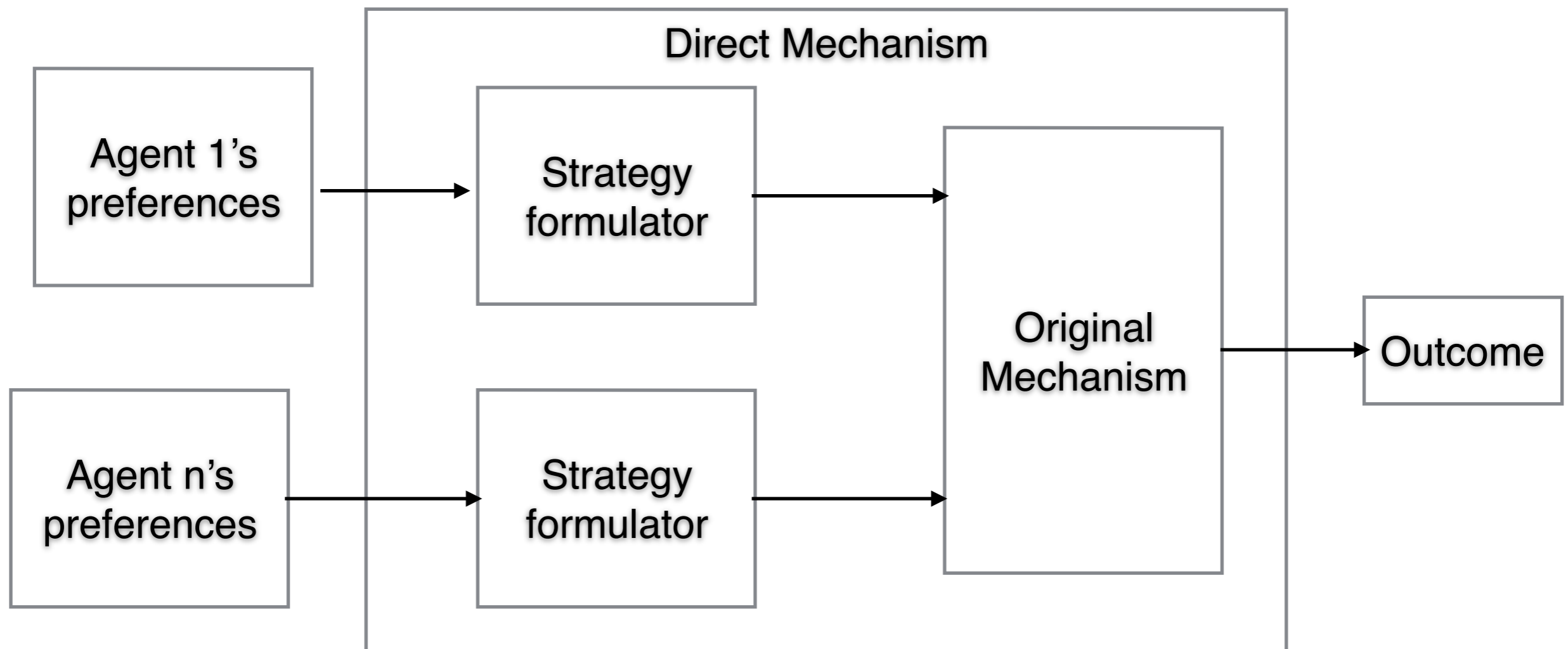
- A direct mechanism is strategy proof if the equilibrium above is a dominant strategy equilibrium

# Revelation Principle

---

---

- **Theorem:** Suppose there exists a mechanism  $M$  that implements social choice function  $f$  in dominant strategies. Then there is a direct strategy-proof mechanism  $M'$  which also implements  $f$ .



# Quick Review

---

---

- We know
  - What a mechanism is
  - What it means for a SCF to be (dominant-strategy) implementable
  - Revelation Principle
- We do not yet know
  - What types of SCF are dominant-strategy implementable

# Gibbard-Satterthwaite Theorem

---

---

- **Theorem:** Assume that
  - $O$  is finite and  $|O| > 2$
  - Each  $o$  in  $O$  can be achieved by SCF  $f$  for some  $\theta$
  - $\theta$  includes all possible strict orderings over  $O$

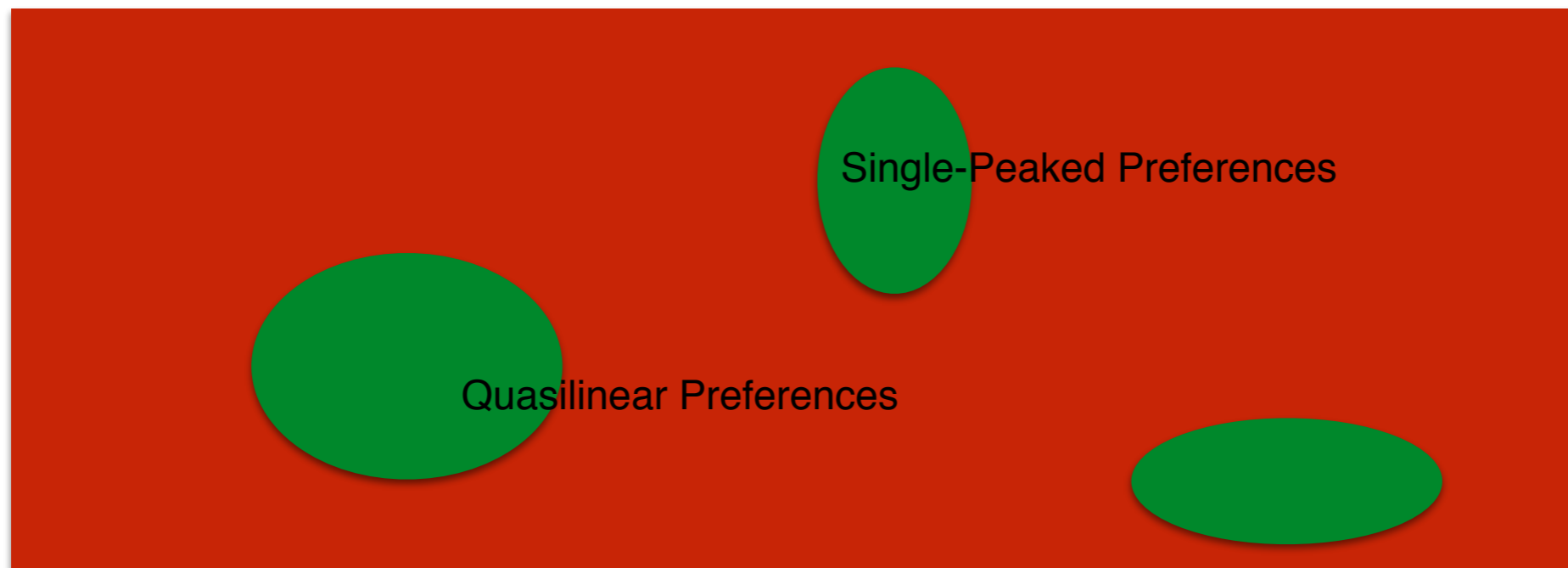
Then  $f$  is implementable in dominant strategies if and only if  $f$  is dictatorial.

# Circumventing Gibbard-Satterthwaite

---

---

- Use a weaker equilibrium concept
- Design mechanisms where computing manipulations is computationally hard
- Restrict the structure of agents' preferences



# Single-Peaked Preferences

---

---

- Median-Voter rule is strategy proof for single-peaked preferences



# Quasilinear Preferences

---

---

- Outcome  $o=(x,t_1,\dots,t_n)$ 
  - $x$  is a “project choice”
  - $t_i$  in  $\mathbb{R}$  are transfers (“money”)
- Utility functions:  $u_i(o,\theta_i)=v_i(x,\theta_i)-t_i$
- Quasilinear mechanism  $M=(S_1,\dots,S_n,g())$   
where
  - $g()=(x(),t_1,\dots,t_n)$

# Groves Mechanisms

---

---

- Choice rule

$$x^*(\theta) = \arg \max_x \sum_i v_i(x, \theta_i)$$

- Transfer rules

$$t_i(\theta) = h_i(\theta_{-i}) - \sum_{j \neq i} v_j(x^*(\theta), \theta_j)$$

# Groves Mechanisms

---

---

- **Theorem:** Groves mechanisms are strategy-proof and efficient.
- **Theorem:** Groves mechanisms are unique (up to  $h_i(\theta_{-i})$ )

# Vickrey-Clarke-Groves Mechanism

---

---

- Outcome

$$x^* = \arg \max_x \sum_i v_i(x, \theta_i)$$

- Transfers

$$t_i(\theta) = \sum_{j \neq i} v_j(x^{-i}, \theta_j) - \sum_{j \neq i} v_j(x^*(\theta), \theta_j)$$

- VCG is an example of a Groves mechanism
  - Efficient and strategy-proof
  - Agents' equilibrium utility is their marginal contribution to the welfare of the system

# Example: Allocation Problem

---

---

- Social choice function
  - Maximize social welfare (i.e. give item to agent who values it the most)
- Utility functions:  $u_i = v_i(o) - t_i$
- Mechanism (Vickrey Auction)
  - $S_i$ : a bid of any non-negative number
  - Outcome function  $g$ :
    - Give item to agent who submits highest bid
    - Highest bidder pays amount of second highest bid, all else pay nothing

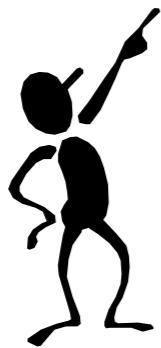
# Vickrey Auction

---

---



$V_1 = \$6$



$V_2 = \$5$



$V_3 = \$2$



If agents bid truthfully then  
Agent 1 wins  
Pays \$5

$$U_1 = \$6 - \$5 = \$1$$

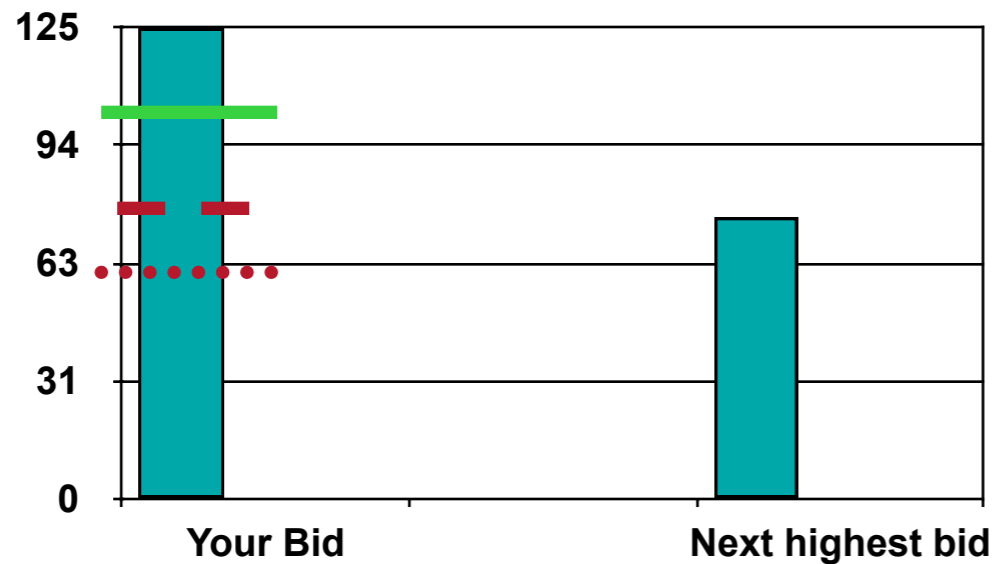
$$U_2 = U_3 = 0$$

# Vickrey Auction

---

---

- Case 1: Bidding truthfully and you are the highest bidder



Bid more:

No difference  
Still pay the same

Bid less:

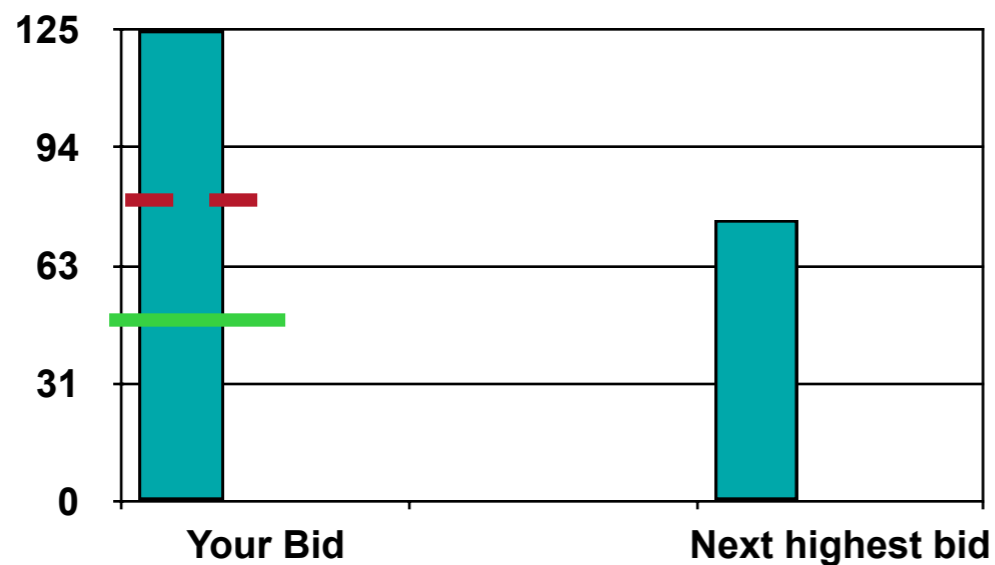
No difference  
Lose the auction

# Vickrey Auction

---

---

- Case 2: Bidding truthfully and you are not the highest bidder



Bid less:

No difference

Bid more:

No difference

Win the auction

and pay too much



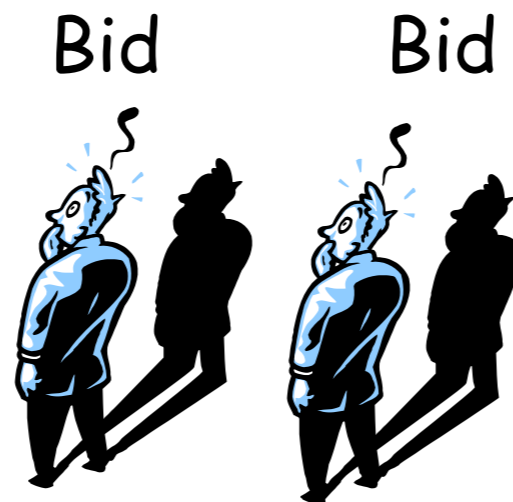
# Another Application: Sponsored Search

---

---

Slot 1
Slot 2
Slot 3
Slot 4
Slot 5

< Keyword >



1. Advertisers are ranked and assigned slots based on the ranking.
2. If an ad is clicked on, only then does the advertiser pay.

# Ranking

---

---

- Rank-by-relevance
  - Assign slots of order of  $(\text{quality score}) * (\text{bid})$

Bidder	Bid	Quality Score
A	1.50	0.5
B	1.00	0.9
C	0.75	1.5



Ranking
C (1.25)
B (0.9)
A (0.75)

# Pricing

---

---

- An advertiser only pays when its ad is clicked on
- How much does it pay?
  - The lowest price it could have bid and still been in the same position

# Example

---

---

Bidder	Bid	Quality Score
A	1.50	0.5
B	1.00	0.9
C	0.75	1.5



Ranking
C (1.25)
B (0.9)
A (0.75)

C will pay  $p=0.9/1.5=0.6$   
B will pay  $p=0.75/0.9 = 0.83$

How much will A pay?

# Sponsored Search

---

---

- How would you design a bidding agent for sponsored search?
- Different from the Vickrey auction
  - There is no single best strategy
  - It depends on the strategies of others

# Summary

---

---

- Definition of a mechanism
- What it means for a mechanism to implement a social choice function
- Revelation Principle
- Gibbard-Satterthwaite Theorem
- Possibility results
  - Groves mechanisms