

# Distributed Optimization and Games

## **Auctions**

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# Our starting problem

- We want to give an object to the person who values it the most, i.e.

$$\begin{aligned} & \text{maximize} && \sum_{i=1}^N x_i v_i \\ & \text{subject to} && \sum_{i=1}^N x_i = 1 \\ & \text{over} && x_i \in \{0,1\} \end{aligned}$$

- Difficulty: we do not know values  $v_i$  ...
- and we cannot ask to people (they would lie)
- Solution: auctions, but we need to introduce money

# Types of auctions

- ❑ 1<sup>st</sup> price & descending bids (Dutch auctions)
- ❑ 2<sup>nd</sup> price & ascending bids (English auctions)

# Google



digital photo camera



Giovanni Neglia

0

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## Search

About 426,000,000 results (0.25 seconds)

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### Ads ⓘ

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Très grande sélection de  
**digital photo cameras** à petits prix

# How it works

- ❑ Companies bid for keywords
- ❑ On the basis of the bids Google puts their link on a given position (first ads get more clicks)
- ❑ Companies are charged a given cost for each click (the cost depends on all the bids)
- ❑ Why Google adopted this solution:
  - It has no idea about the value of a click...
  - It lets the companies reveal it

# Some numbers

- ❑  $\approx$  90% of Google revenues from ads
  - 2014 out of 66 billions\$
  - 2016 out of 89 billions\$ [abc.xyz/investor/](#)
- ❑ Costs
  - "calligraphy pens" \$1.70
  - "Loan consolidation" \$50
  - "mesothelioma" \$50 per click
- ❑ Click fraud problem

# Outline

## □ Preliminaries

- Auctions
- Matching markets

## □ Possible approaches to ads pricing

## □ Google mechanism

## □ References

- Easley, Kleinberg, "Networks, Crowds and Markets", ch.9,10,15

# Game Theoretic Model

- ❑ N players (the bidders)
- ❑ Strategies/actions:  $b_i$  is player  $i$ 's bid
- ❑ For player  $i$  the good has value  $v_i$
- ❑  $p_i$  is player  $i$ 's payment if he gets the good
- ❑ Utility:
  - $v_i - p_i$  if player  $i$  gets the good
  - 0 otherwise
- ❑ Assumption here: values  $v_i$  are *independent* and *private*
  - i.e. very particular goods for which there is not a reference price



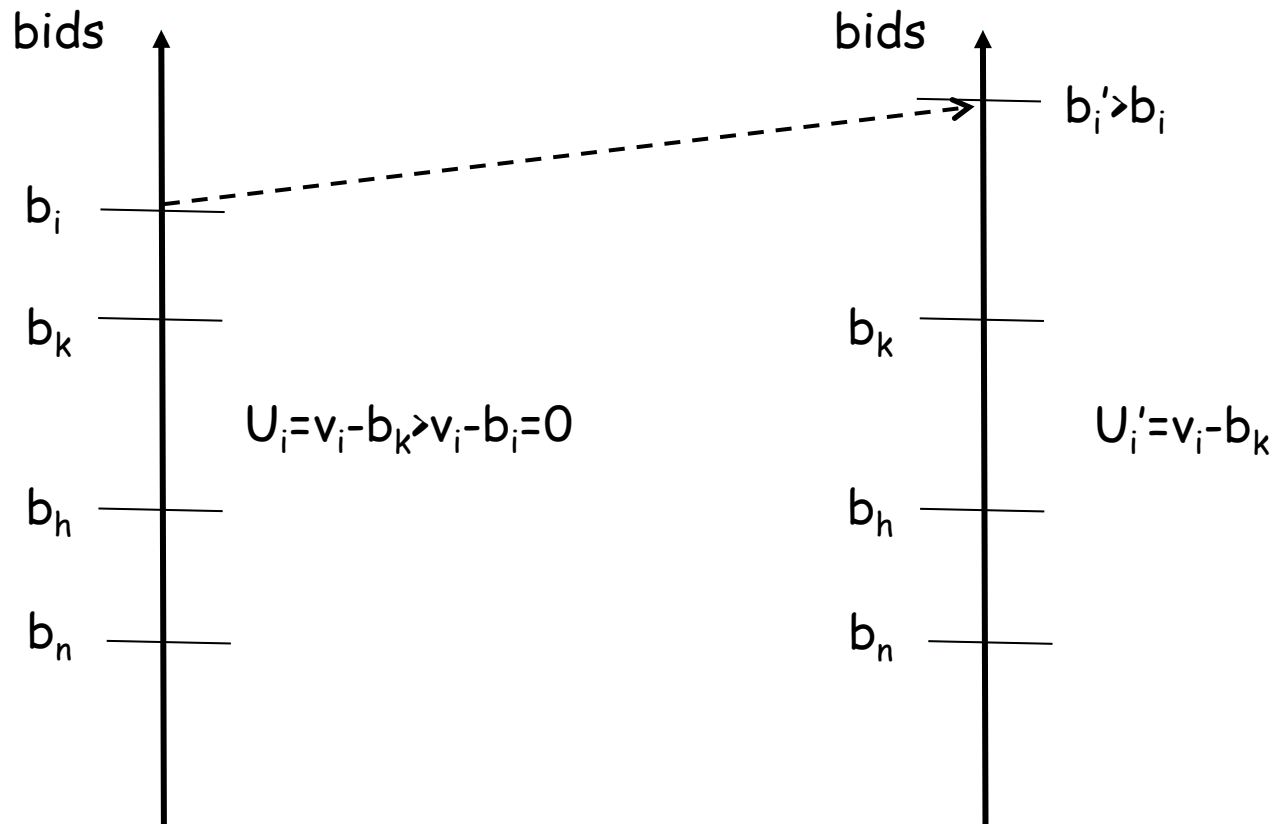
# Game Theoretic Model

- ❑ N players (the bidders)
- ❑ Strategies:  $b_i$  is player  $i$ 's bid
- ❑ Utility:
  - $v_i - b_i$  if player  $i$  gets the good
  - 0 otherwise
- ❑ Difficulties:
  - Utilities of other players are unknown!
  - Better to model the strategy space as continuous (differently from the games we looked at)

## 2<sup>nd</sup> price auction

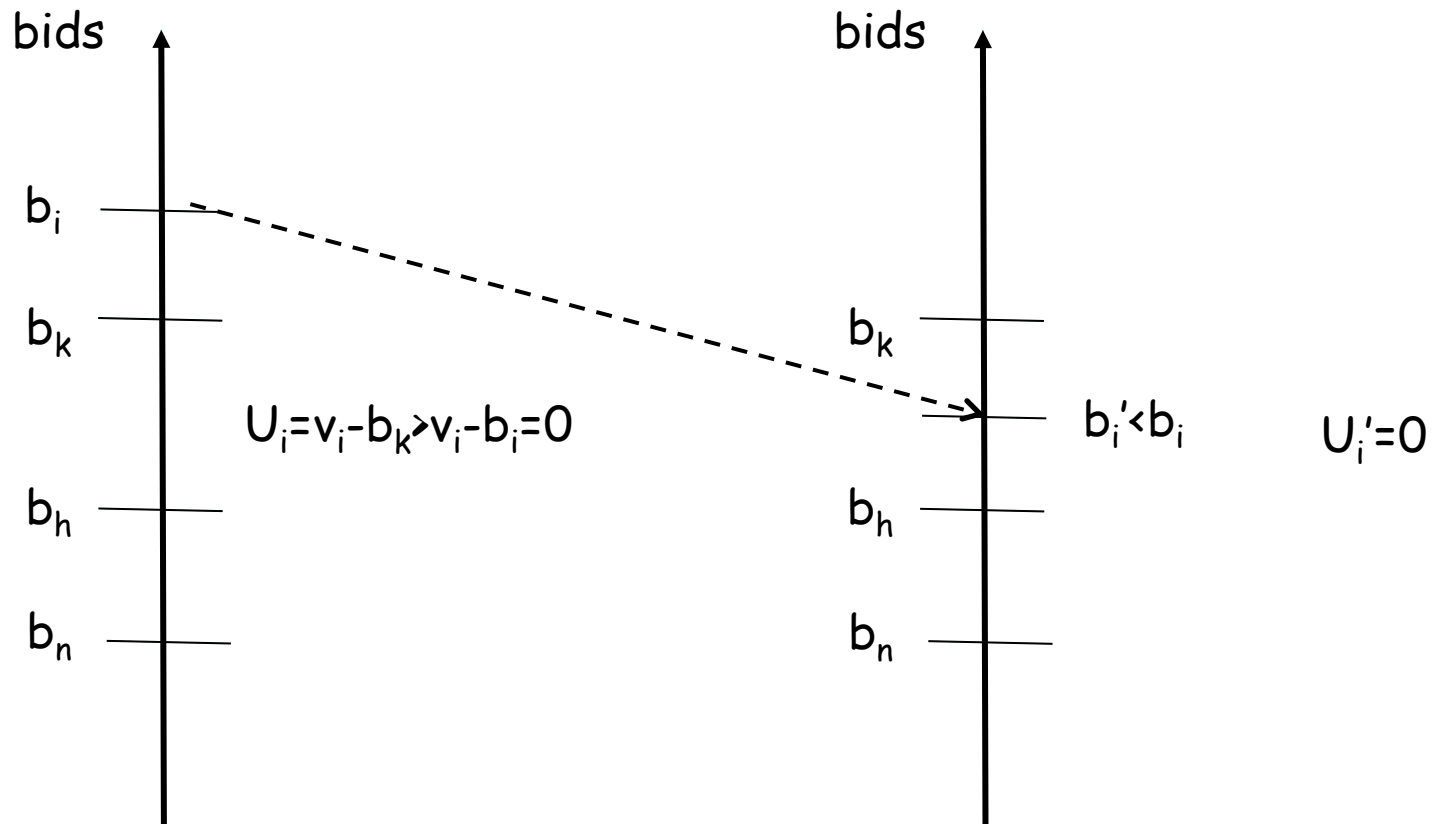
- ❑ Player with the highest bid gets the good and pays a price equal to the 2<sup>nd</sup> highest bid
- ❑ There is a dominant strategies
  - I.e. a strategy that is more convenient independently from what the other players do
  - **Be truthful**, i.e. bid how much you evaluate the good ( $b_i = v_i$ )
  - Social optimality: the bidder who value the good the most gets it!

$b_i = v_i$  is the highest bid



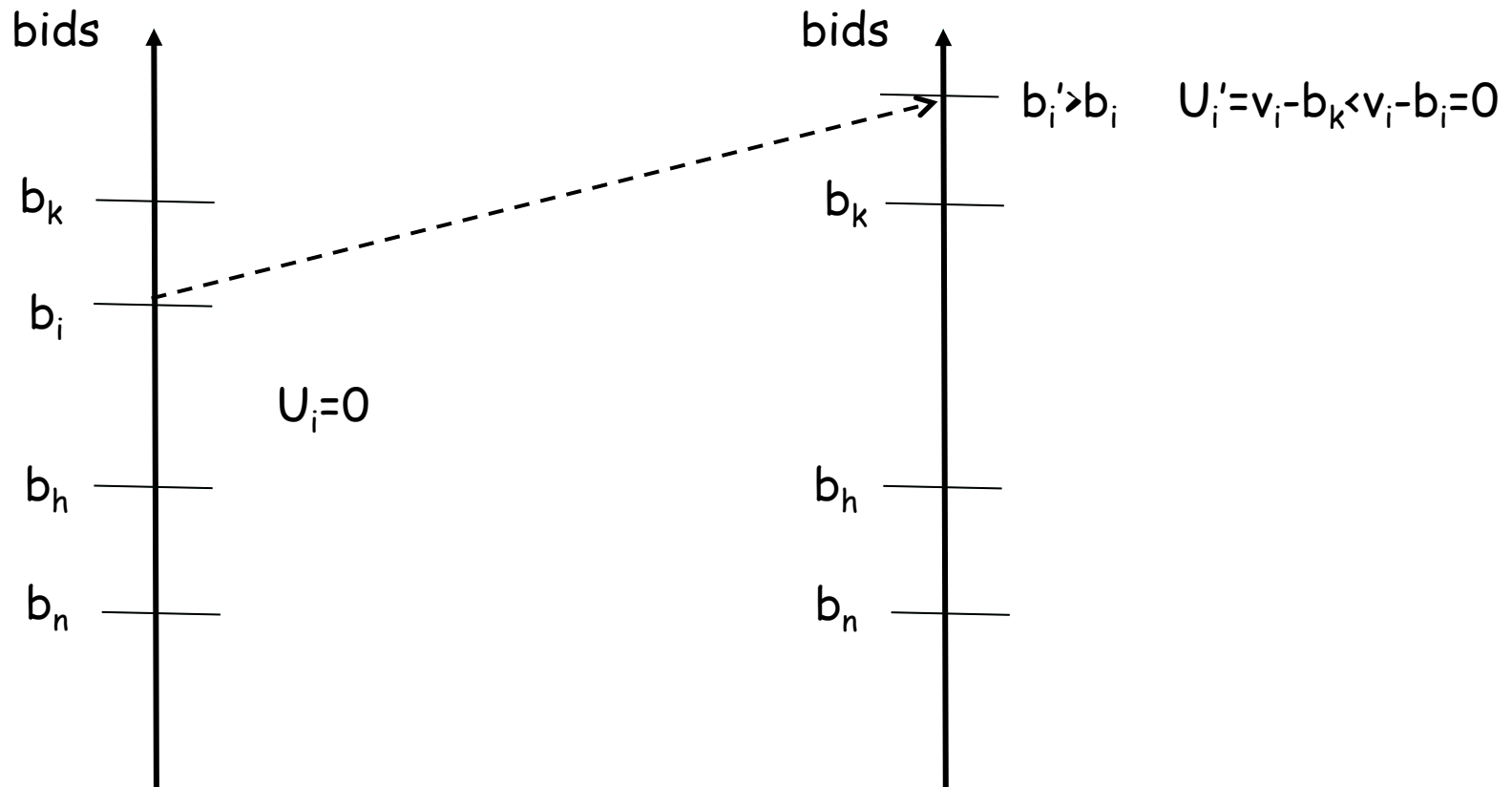
Bidding more than  $v_i$  is not convenient

$b_i = v_i$  is the highest bid



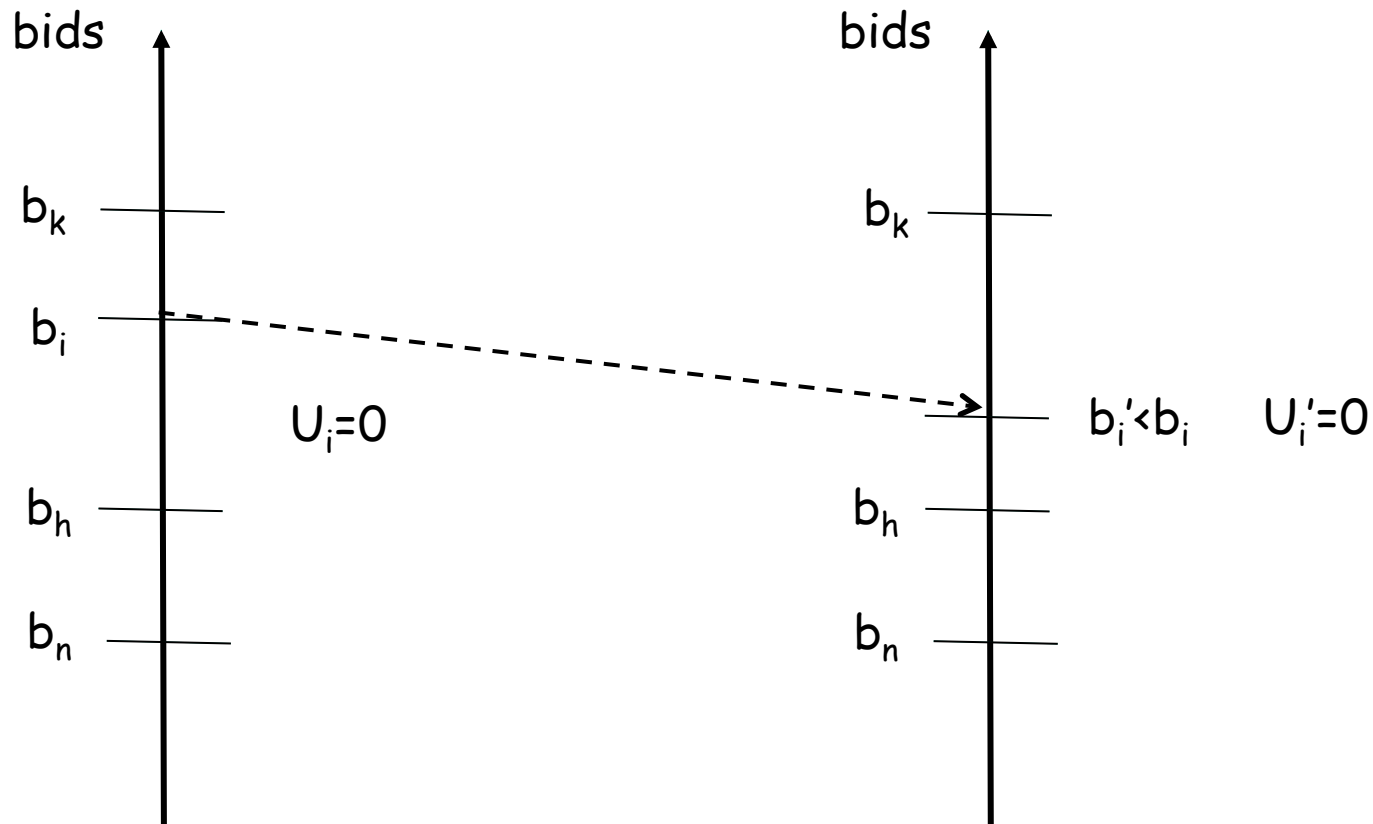
Bidding less than  $v_i$  is not convenient (may be inconvenient)

$b_i = v_i$  is not the highest bid



Bidding more than  $v_i$  is not convenient (may be inconvenient)

$b_i = v_i$  is not the highest bid



Bidding less than  $v_i$  is not convenient

# Seller revenue

- N bidders
- Values are independent random values between 0 and 1
- Expected  $i^{\text{th}}$  largest utility is  $(N+1-i)/(N+1)$
- Expected seller revenue is  $(N-1)/(N+1)$

# 1<sup>st</sup> price auction

- ❑ Player with the highest bid gets the good and pays a price equal to her/his bid
- ❑ Being truthful is not a dominant strategy anymore!
  - Consider for example if I knew other players' utilities
- ❑ How to study it?



# 1<sup>st</sup> price auction

- ❑ Assumption: for each player the other values are i.i.d. random variables between 0 and 1
  - to overcome the fact that utilities are unknown
- ❑ Player  $i$ 's strategy is a function  $s()$  mapping value  $v_i$  to a bid  $b_i$ 
  - $s()$  strictly increasing, differentiable function
  - $0 \leq s(v) \leq v \rightarrow s(0) = 0$
- ❑ We investigate if there is a strategy  $s()$  common to all the players that leads to a Nash equilibrium

# 1<sup>st</sup> price auction

- Assumption: for each player the other values are i.i.d. random variables between 0 and 1
- Player  $i$ 's strategy is a function  $s()$  mapping value  $v_i$  to a bid  $b_i$
- Expected payoff of player  $i$  if all the players plays  $s()$ :

$$U_i(s(v_1), \dots, s(v_i), \dots, s(v_N)) = v_i^{N-1} (v_i - s(v_i))$$

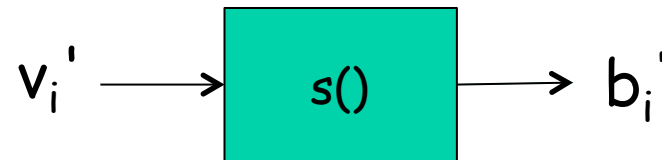
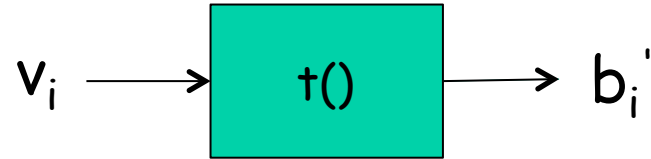
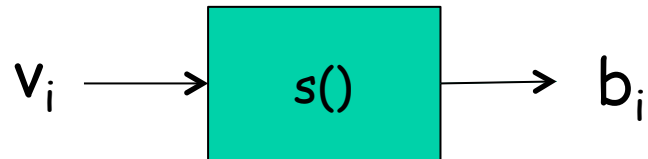
prob.  $i$  wins

$i$ 's payoff if he/she wins

# 1<sup>st</sup> price auction

- ❑ Expected payoff of player  $i$  if all the players play  $s()$ :
  - $U_i(s(v_1), \dots, s(v_i), \dots, s(v_N)) = v_i^{N-1} (v_i - s(v_i))$
- ❑ What if  $i$  plays a different strategy  $t()$ ?
  - If all players playing  $s()$  is a NE, then :
  - $U_i(s(v_1), \dots, s(v_i), \dots, s(v_N)) = v_i^{N-1} (v_i - s(v_i))$   
 $\geq s^{-1}(t(v_i))^{N-1} (v_i - t(v_i)) = U_i(s(v_1), \dots, t(v_i), \dots, s(v_N))$
- ❑ Difficult to check for all the possible functions  $t()$  different from  $s()$
- ❑ Help from the **revelation principle**

# The Revelation Principle



- All the strategies are equivalent to bidder  $i$  supplying to  $s()$  a different value of  $v_i$

# 1<sup>st</sup> price auction

- Expected payoff of player  $i$  if all the players plays  $s()$ :
  - $U_i(s(v_1), \dots, s(v_i), \dots, s(v_N)) = v_i^{N-1} (v_i - s(v_i))$
- What if  $i$  plays a different strategy  $t()$ ?
- By the revelation principle:
  - $U_i(s(v_1), \dots, t(v_i), \dots, s(v_N)) =_{eq} U_i(s(v_1), \dots, s(v), \dots, s(v_N)) = v_i^{N-1} (v_i - s(v))$
- If  $v_i^{N-1} (v_i - s(v_i)) \geq v_i^{N-1} (v_i - s(v))$  for each  $v$  (and for each  $v_i$ )
  - Then all players playing  $s()$  is a NE

# 1<sup>st</sup> price auction

- If  $v_i^{N-1} (v_i - s(v_i)) \geq v^{N-1} (v_i - s(v))$  for each  $v$  (and for each  $v_i$ )
  - Then all players playing  $s()$  is a NE
- $f(v) = v_i^{N-1} (v_i - s(v_i)) - v^{N-1} (v_i - s(v))$  is minimized for  $v = v_i$
- $f'(v) = 0$  for  $v = v_i$ ,
  - i.e.  $(N-1) v_i^{N-2} (v_i - s(v_i)) - v_i^{N-1} s'(v_i) = 0$  for each  $v_i$
  - $s'(v_i) = (N-1)(1 - s(v_i)/v_i)$ ,  $s(0) = 0$
  - Solution:  $s(v_i) = (N-1)/N v_i$

# 1<sup>st</sup> price auction

- ❑ All players bidding according to  $s(v) = (N-1)/N v$  is a NE
- ❑ Remarks
  - They are not truthful
  - The more they are, the higher they should bid
- ❑ Expected seller revenue
  - $((N-1)/N) E[v_{\max}] = ((N-1)/N) (N/(N+1)) = (N-1)/(N+1)$
  - Identical to 2<sup>nd</sup> price auction!
  - A general revenue equivalence principle

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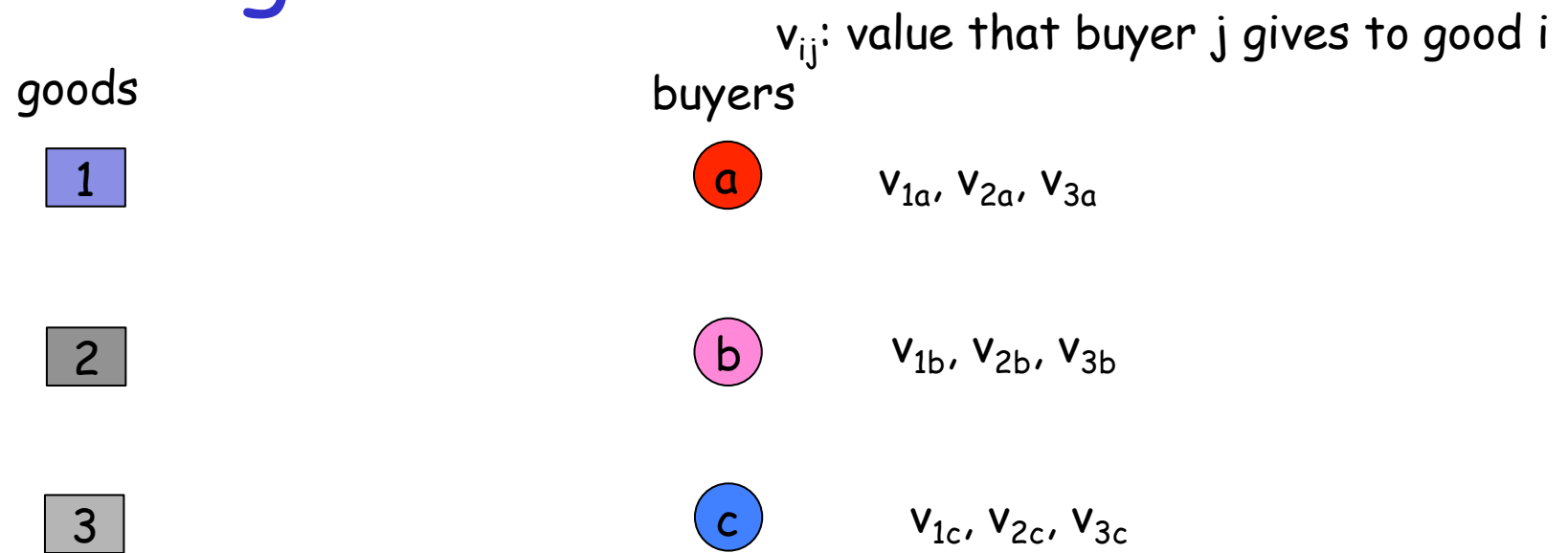
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# Matching Markets



How to match a set of  
different goods to  
a set of buyers with  
different evaluations

# Matching Markets

goods

1

2

3

buyers

a

$v_{1a}, v_{2a}, v_{3a}$

b

$v_{1b}, v_{2b}, v_{3b}$

c

$v_{1c}, v_{2c}, v_{3c}$

$v_{ij}$ : value that buyer  $j$  gives to good  $i$

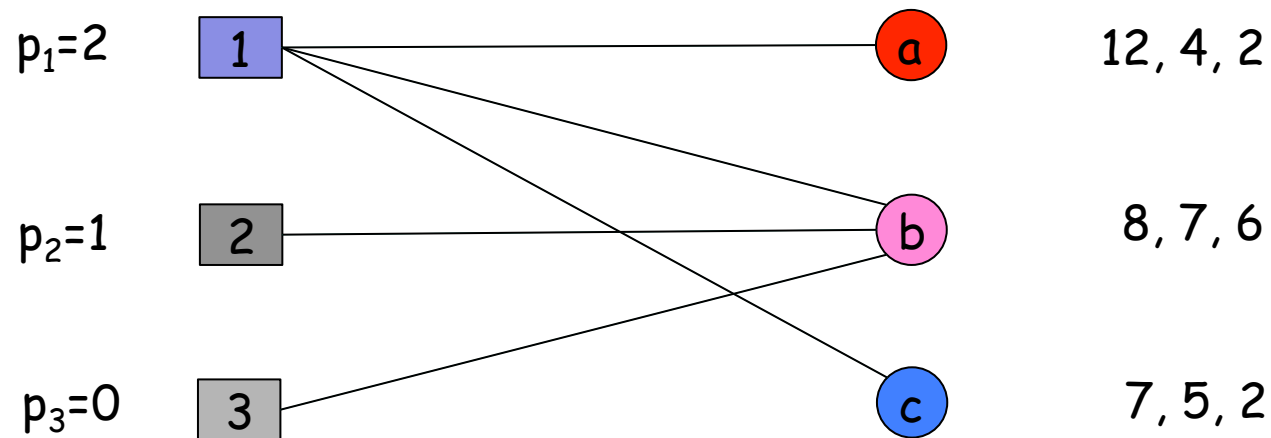
How to match a set of different goods to a set of buyers with different evaluations

$$\text{maximize } \sum_{i,j=1}^N x_{ij} v_{ij}$$

$$\text{subject to } \sum_{j=1}^N x_{ij} = 1, \quad \sum_{i=1}^N x_{ij} = 1,$$

$$\text{over } x_{ij} \in \{0,1\}$$

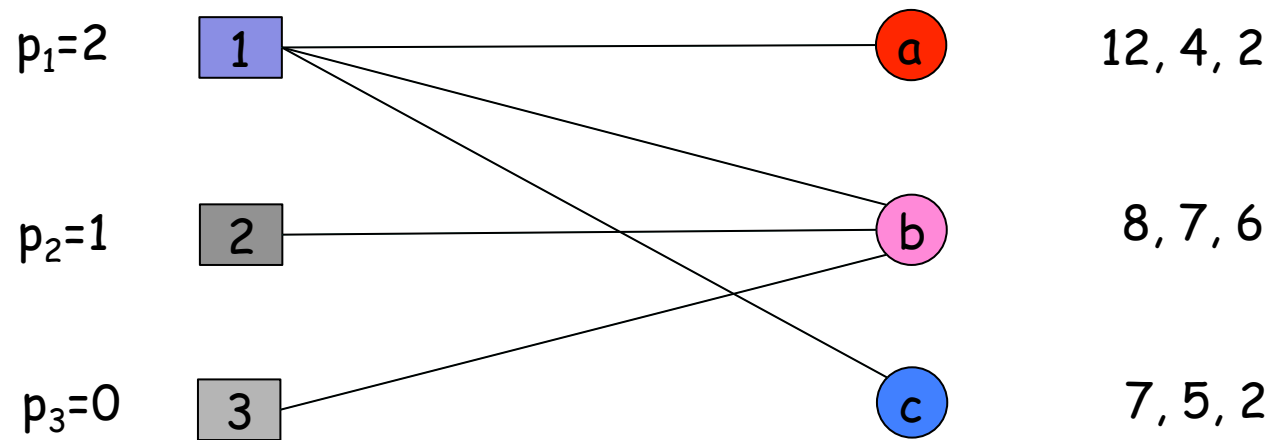
# Matching Markets



Which goods buyers like most? Preferred seller graph

How to match a set of different goods to a set of buyers with different evaluations

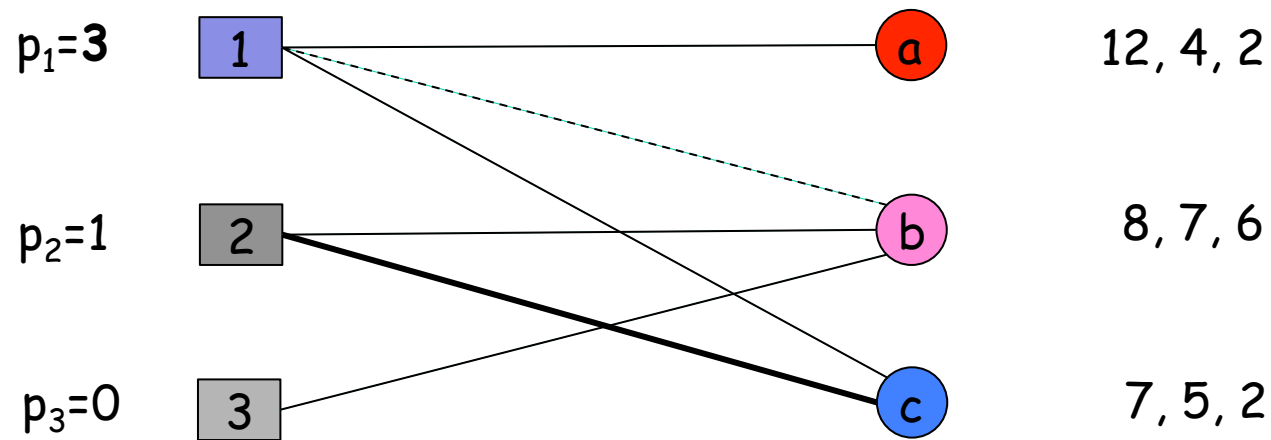
# Matching Markets



Which goods buyers like most? Preferred seller graph

- ❑ Given the prices, look for a perfect matching on the preferred seller graph
- ❑ There is no such matching for this graph

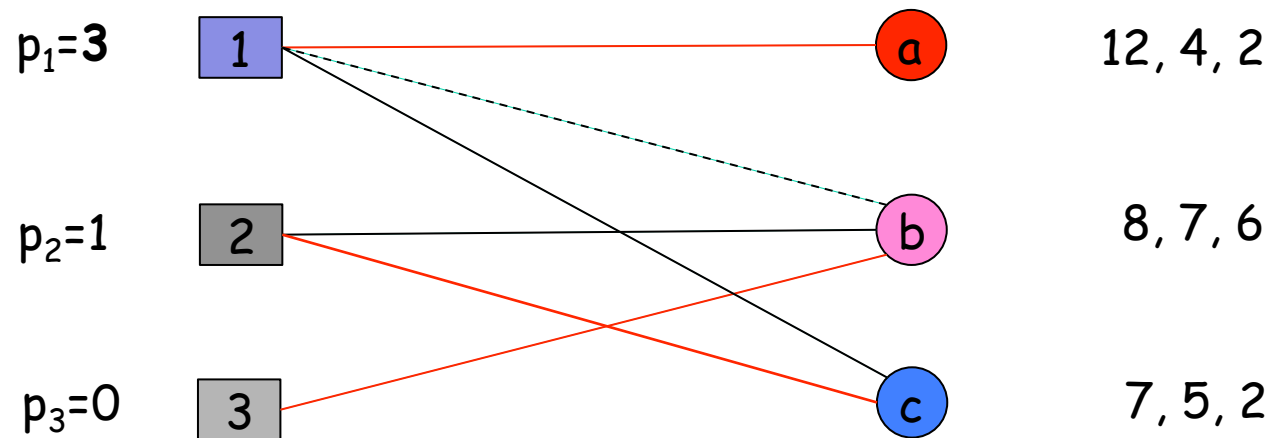
# Matching Markets



Which goods buyers like most? Preferred seller graph

□ But with different prices, there is

# Matching Markets



Which goods buyers like most? Preferred seller graph

- But with different prices, there is
- Such prices are **market clearing prices**

# Market Clearing Prices

- ❑ They always exist
  - And can be easily calculated if valuations are known
- ❑ They are socially optimal in the sense that
  - they achieve the maximum total valuation of any assignment of sellers to buyers
  - Or, equivalently, they maximize the sum of all the payoffs in the network (both sellers and buyers)