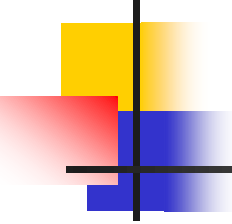


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SECOND PART on  
Non-cooperative Networks:  
Algorithmic Mechanism Design



# Game Theory vs Implementation Theory (in poor words)

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**Game Theory** aims to investigate rational decision-making in conflicting situations, whereas **Implementation Theory** just concerns the **reverse question**: given some desirable outcome, can we design a game that produces it (in equilibrium)?

# The implementation problem (informally)

## ■ Given:

- An economic system comprising of self-interested, rational players, which hold some **secret** information about their preference
- A system-wide goal (**social-choice function (SCF)**, i.e., an aggregation of players' preferences)

## ■ Question:

- Does there exist a **mechanism** that can enforce (through suitable *economic incentives*) the players to reveal their secret information so that the desired **goal** is implemented optimally (w.r.t. the true preferences)?





# Designing a Mechanism

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- Informally, designing a mechanism means to define a **game** in which a desired outcome must be reached (in equilibrium)
- However, games induced by mechanisms are different from games seen so far:
  - Players hold independent **private values**, called **types**
  - The payoffs are a function of these types

⇒ each player does not really know about the other players' payoffs, but only about her one!
- ⇒ Games with **incomplete information**

# An example: sealed-bid auctions

$t_1=10$

$r_1=11$

$t_2=12$

$r_2=10$

$t_3=7$

$r_3=7$

$t_i$  is the **maximum** amount of money player  $i$  is willing to pay for the painting, i.e., her **valuation** of the painting in case she will get it

If player  $i$  **wins** and has to pay  $p$  then her **utility** is  $u_i=t_i-p$ , otherwise it is 0

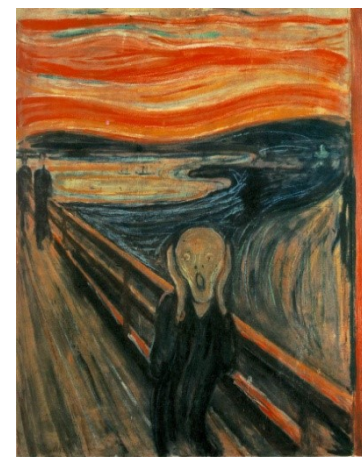
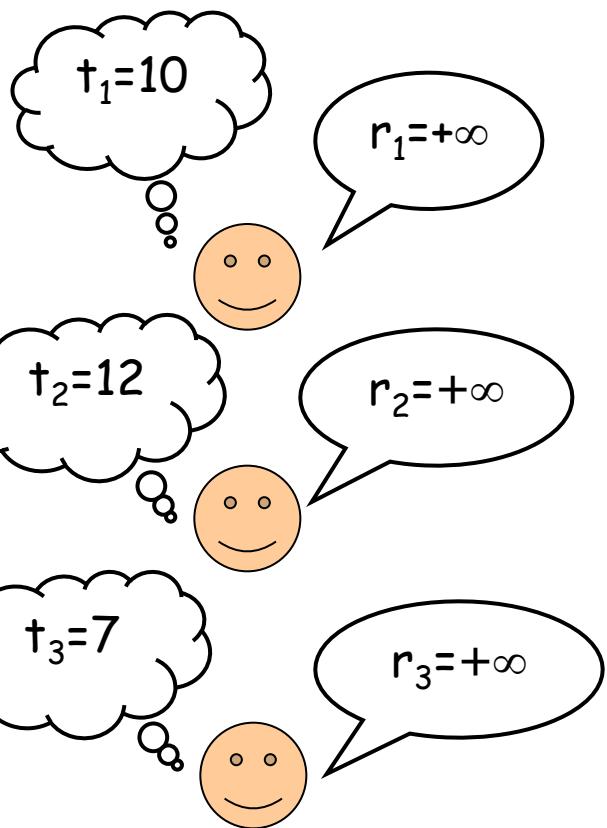
$r_i$  is the amount of money player  $i$  bids (in a sealed envelope) for the painting

SCF: the winner should be the guy **having in mind** the highest value for the painting



- The mechanism tells to players:
- (1) How the item will be allocated (i.e., who will be the **winner**), depending on the received bids
  - (2) The payment the winner has to return, as a function of the received bids

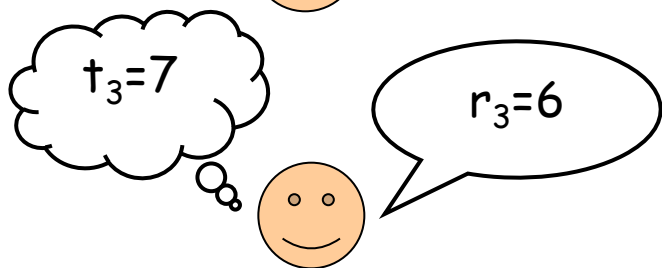
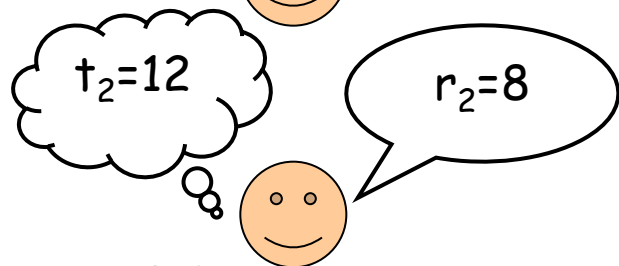
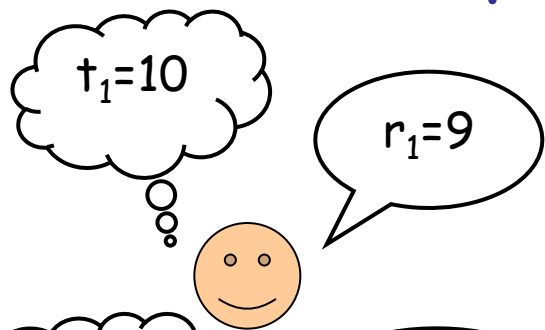
# A simple mechanism: no payment



**Mechanism:** The highest bid wins  
and the price of the item  
is 0

...it doesn't work...

# Another simple mechanism: pay your bid (a.k.a. first-price auction)



Player  $i$  may bid  $r_i < t_i$  (in this way she is guaranteed not to incur a negative utility)

...and so the winner could be the wrong one...

...it doesn't work...

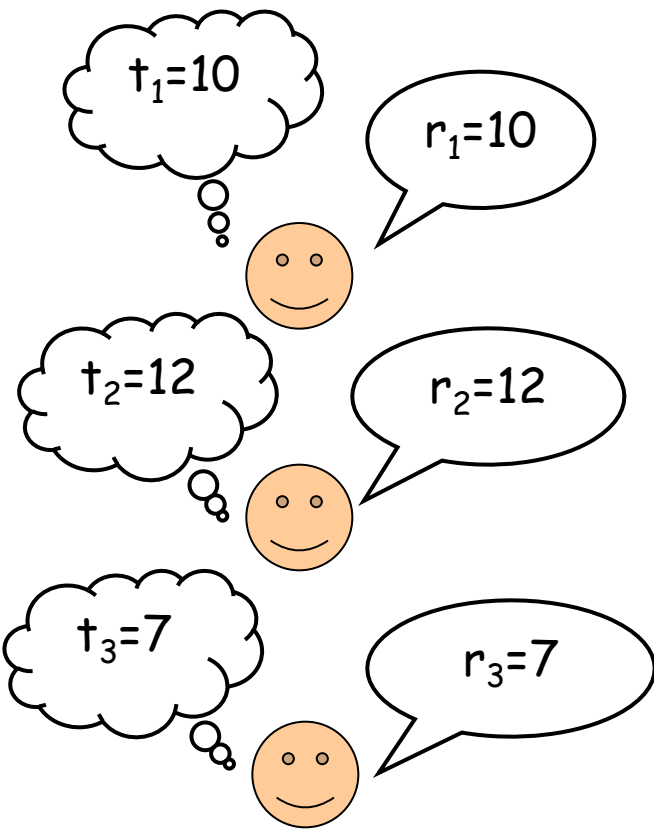
The winner is player 1 and she will pay 9

Is it the right choice?



**Mechanism:** The highest bid wins and the winner will pay her bid

# An elegant solution: Vickrey's second price auction



The winner is player 2 and she will pay 10

I know they are not lying



**Mechanism:** The highest bid wins (ties are broken arbitrarily) and the winner will pay the second highest bid

every player has convenience to declare the truth!  
(we prove it in the next slide)



# Theorem

In the Vickrey auction, for every player  $i$ ,  $r_i = t_i$  is a dominant strategy

**proof** Fix  $i$  and  $t_i$ , and look at strategies for player  $i$ . Let  $R = \max_{j \neq i} \{r_j\}$ .

Case  $t_i > R$  (observe that  $R$  is unknown to player  $i$ )

1. declaring  $r_i = t_i$  gives utility  $u_i = t_i - R > 0$  (player **wins**)
2. declaring any  $r_i > R$ ,  $r_i \neq t_i$ , yields again utility  $u_i = t_i - R > 0$  (player **wins**)
3. declaring  $r_i = R$  yields a utility depending on the tie-breaking rule: if player  $i$  **wins**, she has again utility  $u_i = t_i - R > 0$ , while if she loses, then  $u_i = 0$
4. declaring any  $r_i < R$  yields  $u_i = 0$  (player **loses**)

$\Rightarrow$  In any case, the best utility is  $u_i = t_i - R$ , which is obtained when declaring  $r_i = t_i$

Case  $t_i < R$

1. declaring  $r_i = t_i$  yields utility  $u_i = 0$  (player **loses**)
2. declaring any  $r_i < R$ ,  $r_i \neq t_i$ , yields again utility  $u_i = 0$  (player **loses**)
3. declaring  $r_i = R$  yields a utility depending on the tie-breaking rule: if player  $i$  **wins**, she has utility  $u_i = t_i - R < 0$ , while if she loses, then she has again utility  $u_i = 0$
4. declaring any  $r_i > R$  yields  $u_i = t_i - R < 0$  (player **wins**)

$\Rightarrow$  In any case, the best utility is  $u_i = 0$ , which is obtained when declaring  $r_i = t_i$

# Proof (cont'd)

Case  $t_i = R$

1. declaring  $r_i = t_i$  yields utility  $u_i = t_i - R = 0$  (player **wins/loses** depending on the tie-breaking rule, but her utility in this case is always 0)
2. declaring any  $r_i < R$  yields again utility  $u_i = 0$  (player **loses**)
3. declaring any  $r_i > R$  yields  $u_i = t_i - R = 0$  (player **wins**)

⇒ In any case, the best utility is  $u_i = 0$ , which is obtained when declaring  $r_i = t_i$

⇒ In all the cases, reporting a **false** type produces a not better **utility**, and so telling the truth is a **dominant strategy**!



# Mechanism Design Problem: ingredients



- **N** players; each player  $i$ ,  $i=1,\dots,N$ , has some **private** information  $t_i \in T_i$  (actually, this is the **only private** information of the game, all the other functions provided in the following are **public**) called **type**
  - **Vickrey's auction**: the type is the value of the painting that a player has in mind, and so  $T_i$  is the set of positive real numbers
- A set of **feasible outcomes**  $X$  (i.e., the result of the interaction of the players with the mechanism)
  - **Vickrey's auction**:  $X$  is the set of players (indeed an outcome of the auction is a **winner** of it, i.e., a player)

# Mechanism Design Problem: ingredients (2)

- For each vector of types  $t=(t_1, t_2, \dots, t_N)$ , and for each feasible outcome  $x \in X$ , a SCF  $f(t,x)$  that measures the quality of  $x$  as a function of  $t$ . This is the function that the mechanism aims to **implement** (i.e., it aims to select an outcome  $x^*$  that minimizes/maximizes it, but the problem is that types are unknown!)
  - **Vickrey's auction**:  $f(t,x)$  is the type associated with a feasible winner  $x$  (i.e., any of the players), and the objective is to maximize  $f$ , i.e., to allocate the painting to the bidder with **highest type**
- Each player  $i$  selects a strategic action taken from a **strategy space**  $S_i$ ; we restrict ourselves to *direct revelation mechanisms*, in which the action is **reporting a value**  $r_i$  from the type space (with possibly  $r_i \neq t_i$ ), i.e.,  $S_i = T_i$ 
  - **Vickrey's auction**: the action is to bid a value  $r_i$

# Mechanism Design Problem: ingredients (3)

- For each feasible outcome  $x \in X$ , each player  $i$  makes a **valuation**  $v_i(t_i, x)$  (in terms of some common **currency**), expressing her preference about that output  $x$ 
  - **Vickrey's auction**: if player  $i$  wins the auction then her valuation is equal to her type  $t_i$ , otherwise it is 0
- For each feasible outcome  $x \in X$ , each player  $i$  receives a **payment**  $p_i(x)$  by the system in terms of the common currency (a negative payment means that the player makes a payment to the system); payments are used by the system to incentive players to be collaborative.
  - **Vickrey's auction**: if player  $i$  wins the auction then she "receives" a payment equal to  $-r_j$ , where  $r_j$  is the second highest bid, otherwise it is 0
- Then, for each feasible outcome  $x \in X$ , the **utility** of player  $i$  (in terms of the common currency) coming from outcome  $x$  will be:
$$u_i(t_i, x) = p_i(x) + v_i(t_i, x)$$
  - **Vickrey's auction**: if player  $i$  wins the auction then her utility is equal to  $u_i = -r_j + t_i \geq 0$ , where  $r_j$  is the second highest bid, otherwise it is  $u_i = 0 + 0 = 0$



# Our focus: Truthful (or Strategy-proof) Mechanism Design

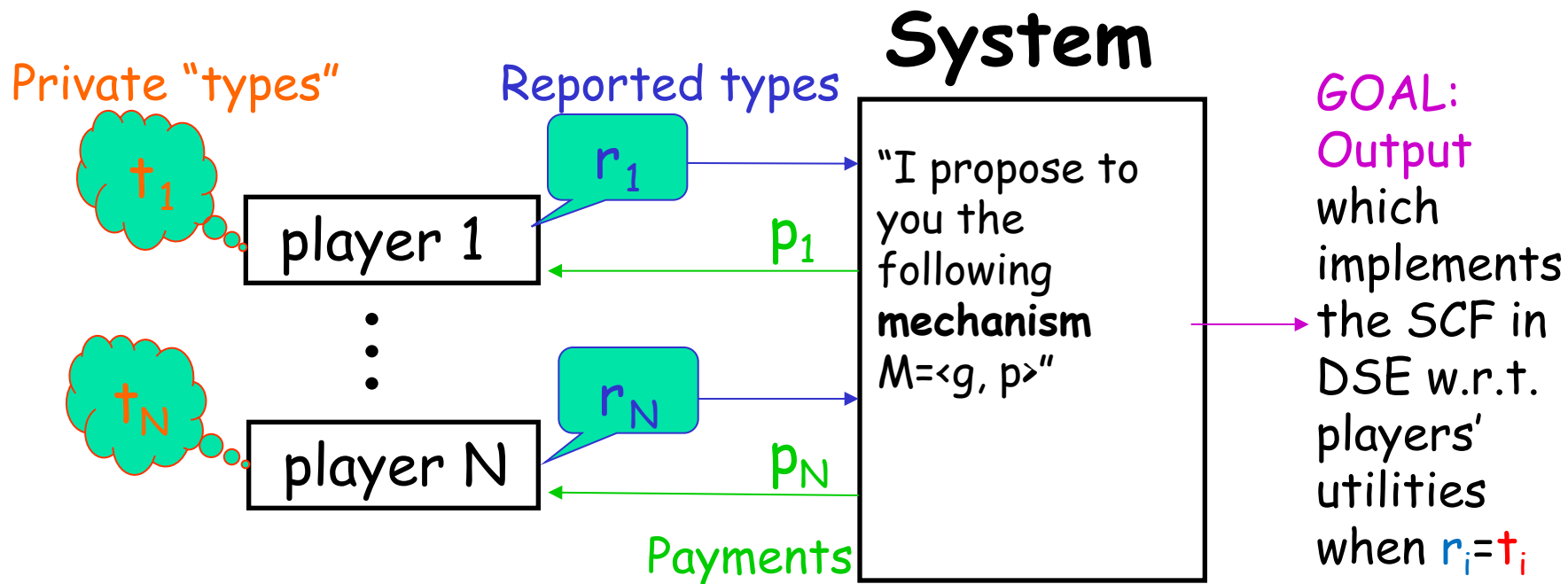
Given all the above ingredients, design a mechanism  $M = \langle g, p \rangle$ , where:

- $g: S_1 \times \dots \times S_N \rightarrow X$  is an algorithm which computes an outcome  $g(r) \in X$  as a function of the reported types  $r$
- $p(g(r)) = (p_1(g(r)), \dots, p_N(g(r))) \in \mathbb{R}^N$  is a payment scheme w.r.t. outcome  $g(r)$  that specifies a payment for each player

which implements (i.e., optimize) the SCF  $f(t, x)$  in dominant strategy equilibrium w.r.t. players' utilities whenever players report their true types. Such a mechanism is called a truthful (or strategy-proof) mechanism.

(In other words, with the reported type vector  $r = t$  the mechanism provides a solution  $g(t)$  and a payment scheme  $p(g(t))$  such that players' utilities  $u_i(t_i, g(t)) = p_i(g(t)) + v_i(t_i, g(t))$  are maximized in DSE and  $f(t, g(t))$  is optimal (either minimum or maximum)).

# Truthful Mechanism Design: a picture



Each player reports strategically to maximize her utility...  
...in response to a payment which is a function of the output!



# Truthful Mechanism Design in DSE: Economics Issues

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**QUESTION:** How to design a truthful mechanism? Or, in other words:

1. How to design the **algorithm  $g$** , and
2. How to define the **payment scheme  $p$**

in such a way that the underlying SCF is implemented truthfully in DSE? Under which conditions can this be done?





# Truthful Mechanism Design in DSE: Computational Issues

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**QUESTION:** What is the **time complexity** of the mechanism? Or, in other words:

- What is the time complexity of computing  $g(r)$ ?
- What is the time complexity to calculate the  $N$  **payment functions**?
- What does it happen if it is **NP-hard** to implement the underlying SCF?

**Question:** What is the **time complexity** of the Vickrey auction?

**Answer:**  $\Theta(N)$ , where  $N$  is the number of players. Indeed, it suffices to check all the offers, by keeping track of the largest one and of the second largest one.



# Algorithmic Mechanism Design

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*Algorithmic  
Mechanism  
Design* = *Theory of  
Algorithms* + *Implementation  
Theory*



# A prominent class of problems

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- **Utilitarian problems:** A problem is *utilitarian* if its SCF is such that  $f(t, x) = \sum_i v_i(t_i, x)$ , i.e., the SCF is *separately-additive* w.r.t. players' valuations.

**Remark 1:** the auction problem is utilitarian, in that  $f(t, x)$  is the *type* associated with the winner  $x$ , and the valuation of a player is either her *type* or 0, depending on whether she wins or not. Then,  $f(t, x) = \sum_i v_i(t_i, x) = \text{type of the winner}$

**Remark 2:** in many *network optimization problems* (which are of our special interest) the SCF is separately-additive

**Good news:** for utilitarian problems there exists a class of truthful mechanisms 😊

# Vickrey-Clarke-Groves (VCG) Mechanisms

- A VCG-mechanism is (the only) strategy-proof mechanism for **utilitarian** problems:
  - Algorithm  $g$  computes:

$$g(r) = \arg \max_{y \in X} \sum_i v_i(r_i, y)$$

- Payment function for player  $i$ :

$$p_i(g(r)) = h_i(r_{-i}) + \sum_{j \neq i} v_j(r_j, g(r))$$

where  $h_i(r_{-i}) = h(r_1, r_2, \dots, r_{i-1}, r_{i+1}, \dots, r_N)$  is an arbitrary function of the types reported by players other than player  $i$ .

- What about **non-utilitarian** problems? Strategy-proof mechanisms are known only when the type is a **single** parameter.

# Theorem

VCG-mechanisms are truthful for utilitarian problems

**Proof:** We show that a player has no interest in lying.

Fix  $i, r_{-i}, t_i$ . Let  $\check{r} = (r_{-i}, t_i)$  and consider a strategy  $r_i \neq t_i$

$$x = g(r_{-i}, t_i) = g(\check{r}) \quad x' = g(r_{-i}, r_i) \quad \check{r}_j = r_j \text{ if } j \neq i, \text{ and } \check{r}_i = t_i$$

$$u_i(t_i, x) = [h_i(r_{-i}) + \sum_{j \neq i} v_j(r_j, x)] + v_i(t_i, x) = h_i(r_{-i}) + \sum_j v_j(\check{r}_j, x)$$

$$u_i(t_i, x') = [h_i(r_{-i}) + \sum_{j \neq i} v_j(r_j, x')] + v_i(t_i, x') = h_i(r_{-i}) + \sum_j v_j(\check{r}_j, x')$$

but  $x$  is an optimal solution w.r.t.  $\check{r} = (r_{-i}, t_i)$ , i.e.,

$$x = \arg \max_{y \in X} \sum_j v_j(\check{r}_j, y)$$

$$\Rightarrow \sum_j v_j(\check{r}_j, x) \geq \sum_j v_j(\check{r}_j, x')$$

$$\Rightarrow u_i(t_i, x) \geq u_i(t_i, x').$$





# How to define $h_i(r_{-i})$ ?

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**Remark:** not all functions make sense. For instance, what does it happen in our Vickrey's auction if we set for every player  $h_i(r_{-i}) = -1000$  (notice this is independent of reported value  $r_i$  of player  $i$ , and so it obeys to the definition)?

**Answer:** It happens that players' utility become **negative**; more precisely, in our example the winner's type/valuation was 12, and so her utility becomes

$$u_i(t_i, x) = p_i(x) + v_i(t_i, x) = h_i(r_{-i}) + \sum_{j \neq i} v_j(r_j, x) + v_i(t_i, x) = -1000 + 0 + 12 = -988$$

while utility of losers becomes

$$u_i(t_i, x) = p_i(x) + v_i(t_i, x) = h_i(r_{-i}) + \sum_{j \neq i} v_j(r_j, x) + v_i(t_i, x) = -1000 + 12 + 0 = -988$$

⇒ This is undesirable in reality, since with such perspective players would not participate to the auction!



# Voluntary participation

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A mechanism satisfies the **voluntary participation** condition if players who reports truthfully never incur a net loss, i.e., for every player  $i$ , type  $t_i$ , and other players' bids  $r_{-i}$

$$u_i(t_i, g(r_{-i}, t_i)) \geq 0.$$

# The Clarke payments

solution maximizing the sum  
of valuations when **player i** doesn't play

- This is a special VCG-mechanism in which

$$h_i(\mathbf{r}_{-i}) = -\sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}_{-i}))$$

$$\Rightarrow p_i(g(\mathbf{r})) = -\sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}_{-i})) + \sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}))$$

- With Clarke payments, it can be shown that players' utility are always non-negative; indeed:

$$u_i(t_i, g(\mathbf{r})) = p_i(g(\mathbf{r})) + v_i(t_i, g(\mathbf{r})) = -\sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}_{-i})) + \sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r})) +$$

$$v_i(t_i, g(\mathbf{r})) = -\sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}_{-i})) + \sum_j v_j(\mathbf{r}_j, g(\mathbf{r})) \geq 0$$

since the first term is never larger (in absolute value) than the second one (intuitively, in utilitarian problems, adding one more player will never decrease the social welfare)

$\Rightarrow$  players are interested in playing the game



# The Vickrey's auction is a VCG mechanism with Clarke payments

- Recall that auctions are **utilitarian** problems. Then, the VCG-mechanism associated with the Vickrey's auction is:
  - $g(\mathbf{r}) = \arg \max_{\mathbf{y} \in \mathcal{X}} \sum_i v_i(\mathbf{r}_i, \mathbf{y})$   
...this is equivalent to allocate to the bidder with **highest reported type** (in the end, the **highest type**, since it is strategy-proof)
  - $p_i(g(\mathbf{r})) = -\sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}_{-i})) + \sum_{j \neq i} v_j(\mathbf{r}_j, g(\mathbf{r}))$   
...this is equivalent to say that the winner pays the **second highest reported type** (in the end, the **second highest type**, since it is strategy-proof), and the losers pay **0**, respectively

**Remark:** the difference between the second highest offer and the highest offer is **unbounded** (frugality issue)



# VCG-Mechanisms: Advantages

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- For System Designer:
  - The goal, i.e., the optimization of the SCF, is achieved with certainty
- For players:
  - players have truth telling as the dominant strategy, so they need not any computational systems to deliberate about other players strategies



# VCG-Mechanisms: Disadvantages

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- For System Designer:

- The payments may be sub-optimal (frugality)
- Apparently, with Clarke payments, the system may need to run the mechanism's algorithm  $N+1$  times: once with all players (for computing the outcome  $g(r)$ ), and once for every player (indeed, for computing the payment  $p_i$  associated with player  $i$ , we need to know  $g(r_{-i})$ )
  - ⇒ If the problem is hard to solve then the computational cost may be very heavy

- For players:

- players may not like to tell the truth to the system designer as it can be used in other ways



# Algorithmic mechanism design and network protocols

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- Large networks (e.g., Internet) are built and controlled by diverse and competitive entities:
    - Entities own different components of the network and hold private information
    - Entities are selfish and have different preferences
- ⇒ Mechanism design is a useful tool to design protocols working in such an environment, but time complexity is an important issue due to the massive network size

# Algorithmic mechanism design for network optimization problems

- Simplifying the Internet model, we assume that each player owns a **single edge** of a graph  $G=(V,E)$ , and privately knows the **cost** for using it
- ⇒ Classic optimization problems on  $G$  become **private-edge mechanism design optimization problems**, in which the player's type is the **weight** of the edge!
- Many basic network design problems have been studied in this framework: shortest path (SP), single-source shortest-path tree (SPT), minimum spanning tree (MST), and many others
- **Remark:** Quite naturally, SP and MST are **utilitarian** problems: indeed the cost of a solution (social-choice function) is simply the **sum** of the edge costs
- On the other hand, the SPT is not! Can you see why?



# Some remarks

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- In general, network optimization problems are **minimization** problems (the Vickrey's auction was instead a maximization problem)
- Accordingly, we have:
  - for each  $x \in X$ , the **valuation function**  $v_i(t_i, x)$  represents a **cost** incurred by player  $i$  in the solution  $x$  (and so it is a **negative** function of her type)
  - the **social-choice function**  $f(t, x)$  is negative (since it is an "aggregation" of negative valuation functions), and so its **maximization** corresponds to a **minimization** of the costs incurred by the players
  - **payments** are now from the mechanism to players (i.e., they are positive)



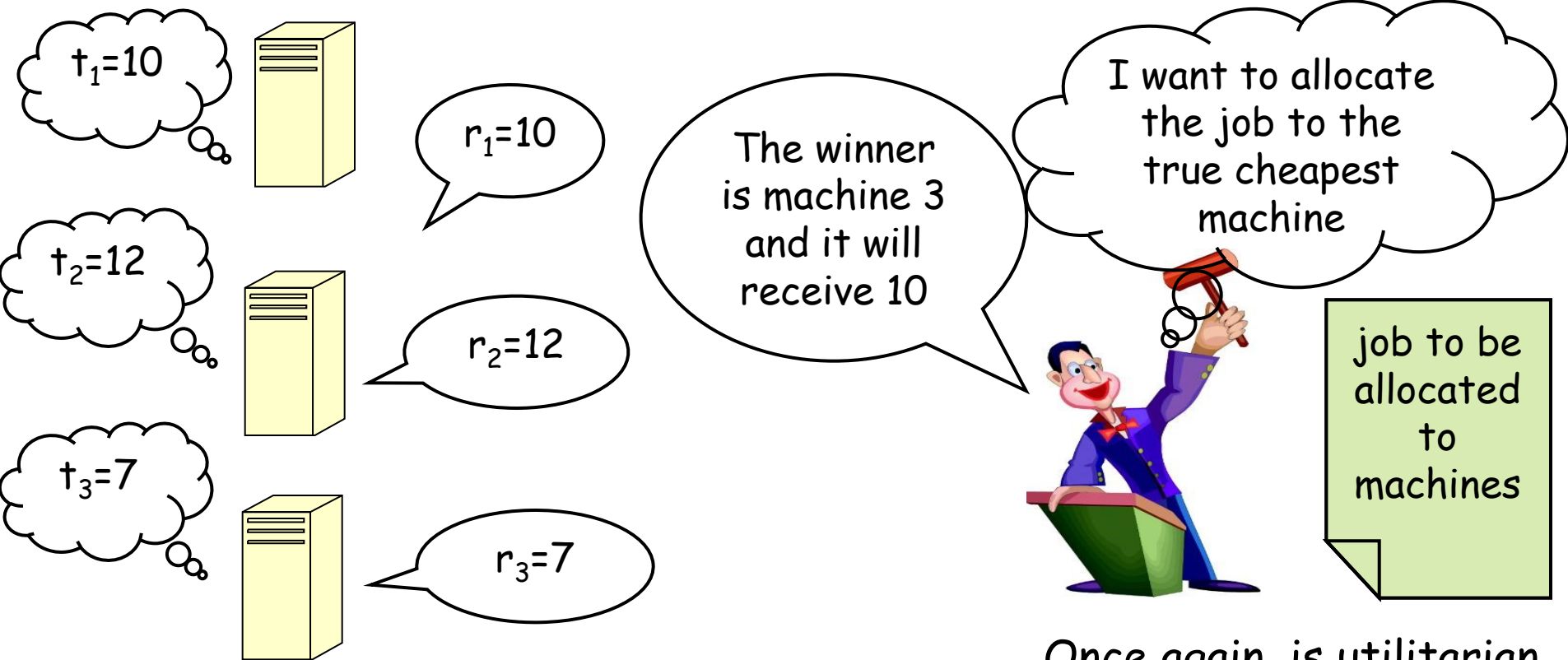
# Summary of forthcoming results

	Centralized algorithm	Private-edge mechanism
SP	$O(m+n \log n)$	$O(m+n \log n)$ (VCG)
MST	$O(m \alpha(m,n))^*$	$O(m \alpha(m,n))$ (VCG)
SPT	$O(m+n \log n)$	$O(m+n \log n)$ (single-parameter)

$\alpha(m,n)$  is the extremely slow-growing inverse of the Ackermann function

⇒ For all these basic problems, the time complexity of the mechanism equals that of the canonical centralized algorithm!

# Exercise: redefine the Vickrey auction in the minimization version (so-called procurement auction)



$t_i$ : cost incurred by  $i$  if she does the job  
 $v_i$ : is equal to  $-t_i$  if  $i$  is the winner, and 0 otherwise  
 $p_i$ : is equal to the second highest type if  $i$  is the winner, and 0 otherwise

Once again, is utilitarian, and so the the second price auction (VCG mechanism) is truthful: the cheapest bid wins and the winner will get the second cheapest bid