

CDS270: Optimization, Game and Layering in Communication Networks

Lecture 1: Static Games and Classical  
Mechanism Design

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# Agenda

- Strategic games and their solution concepts
  - Strategic form games and dominated strategies
  - Nash equilibrium and correlated equilibrium
- Classical mechanism design
  - Incomplete information games
  - Incentive-compatible mechanism
  - VCG mechanism

# Strategic game

□ Def: a game in strategic form is a triple

$$G = \{N, S_{i \in N}, u_{i \in N}\}$$

- $N$  is the set of players (agents)
- $S_i$  is the player  $i$  strategy space
- $u_i : S \rightarrow R$  is the player  $i$  payoff function

## □ Notations

- $S = S_1 \times S_2 \times \dots \times S_N$  : the set of all profiles of player strategies
- $s = (s_1, s_2, \dots, s_N)$  : profile of strategies
- $s_{-i} = (s_1, s_2, \dots, s_{i-1}, \dots, s_{i+1}, \dots, s_N)$  : the profile of strategies other than player  $i$

- Implicitly assume that players have preferences over different outcomes, which can be captured by assigning payoffs to the outcomes
- The basic model of **rationality** is that of a payoff maximizer
- First consider pure strategy, will consider mixed strategy later

# Example: finite game

		column		
		L	M	R
row	U	4,3	5,1	6,2
	M	2,1	8,4	3,6
	D	3,0	9,6	2,8

# Example: Continuous strategy game

## □ Cournot competition

- Two players: firm 1 and firm 2
- Strategy  $s_i \in [0, \infty]$  : the amount of widget that firm  $i$  produces
- The payoff for each firm is the net revenue

$$u_i(s_1, s_2) = s_i p(s_1 + s_2) - c_i s_i$$

where  $p$  is the price,  $c_i$  is the unit cost for firm  $i$

# Dominated strategies

- How to predict the outcome of a game?
- Prisoner's Dilemma

	D	C
D	-2,-2	-5,-1
C	-1,-5	-4,-4

- Two prisoners will play (C,C)
- Def: a strategy  $s_i$  is (weakly) dominated for player  $i$  if there exists  $s'_i \in S_i$  such that

$$u_i(s'_i, s_{-i}) \geq u_i(s_i, s_{-i}) \text{ for all } s_{-i} \in S_{-i}$$

## Iterated elimination of dominated strategies

- Iterated elimination of dominated strategies

	L	M	R
U	4,3	5,1	6,2
M	2,1	8,4	3,6
D	3,0	9,6	2,8

- However, most of games are not solvable by iterated elimination of dominated strategies

# Nash equilibrium

- Def: a strategy profile  $s^*$  is a **Nash equilibrium**, if for all  $i$ ,

$$u_i(s_i^*, s_{-i}^*) \geq u_i(s_i, s_{-i}^*) \text{ for all } s_i \in S_i$$

- For any  $s_{-i} \in S_{-i}$ , define best response function

$$B_i(s_{-i}) = \{s_i \in S_i \mid u_i(s_i, s_{-i}) \geq u_i(s'_i, s_{-i}) \quad \forall s'_i \in S_i\}.$$

Then a strategy profile  $s^*$  is a Nash equilibrium iff  $s_i^* \in B_i(s_{-i}^*)$ .

# Examples

## □ Battle of the Sexes

	Ballet	Soccer
Ballet	2,1	0,0
Soccer	0,0	1,2

Two Nash equilibria (Ballet, Ballet) and (Soccer, Soccer)

# Cournot Competition

- Suppose a price function  $p(s_1 + s_2) = \max\{0, 1 - (s_1 + s_2)\}$
- Suppose cost  $0 \leq c_1 = c_2 = c \leq 1$
- Then, the best response function

$$B_1(s_2) = (1 - s_2 - c) / 2$$

$$B_2(s_1) = (1 - s_1 - c) / 2$$

- Nash equilibrium satisfies  $s_1 = B_1(s_2)$ ,  $s_2 = B_2(s_1)$ , i.e.,  $s_1 = (1 - c) / 3$   
 $s_2 = (1 - c) / 3$

# Second price auction

- An object to be sold to a player in  $N$
- Each player  $i$  has a valuation  $v_i$  of the object. We further assume  $v_1 > v_2 > \dots > v_N > 0$
- The players simultaneously submit bids,  $b_1, \dots, b_N$
- The object is given to the player with highest bid. The winner pays the second highest bid.
- The payoff of the winner is his valuation of the object minus the price he pays. All other players' payoff is zero.

- $(b_1, \dots, b_N) = (v_1, \dots, v_N)$  is Nash equilibrium
  - Player 1 receives the object and pay  $v_2$ , and has payoff  $v_1 - v_2 > 0$ . Player 1 has no incentive to deviate, since his payoff can only decrease.
  - For other players, the payoff is zero. In order to change his payoff, he needs to bid more than  $v_1$ , but that will result in negative payoff. So, no player has incentive to change.
- Question: are there more Nash equilibria?

- ❑ Not all games have (pure) Nash equilibrium
- ❑ Matching Pennies

	Heads	Tails
Heads	1,-1	-1,1
Tails	-1,1	1,-1

# Mixed strategies

- Let  $\Sigma_i$  denote the set of probability distribution over player  $i$  strategy space  $S_i$
- A mixed strategy  $\sigma_i \in \Sigma_i$  is a probability mass function over pure strategies  $s_i \in S_i$
- The payoff of a mixed strategy is the expected value of the pure strategy profiles

$$u_i = \sum_{s \in S} \left( \prod_{j \in N} \sigma_j(s_j) \right) u_i(s)$$

# Mixed strategy Nash equilibrium

- Def: a mixed strategy profile  $\sigma^*$  is a (mixed strategy) Nash equilibrium if for all  $i$

$$u_i(\sigma_i^*, \sigma_{-i}^*) \geq u_i(\sigma_i, \sigma_{-i}^*) \text{ for all } \sigma_i \in \Sigma_i$$

- A mixed strategy profile  $\sigma^*$  is a (mixed strategy) Nash equilibrium if for all  $i$

$$u_i(\sigma_i^*, \sigma_{-i}^*) \geq u_i(s_i, \sigma_{-i}^*) \text{ for all } s_i \in S_i$$

- The payoff  $u_i(s_i, \sigma_{-i}^*)$  is the same for all  $s_i \in \text{supp}(\sigma_i^*)$
- The payoff  $u_i(s_i, \sigma_{-i}^*)$  for each  $s_i \notin \text{supp}(\sigma_i^*)$  is not larger

# Example

	Ballet	Soccer
Ballet	2,1	0,0
Soccer	0,0	1,2

- Assume row (column) player choose "ballet" with probability  $p$  ( $q$ ) and "soccer" with probability  $1-p$  ( $1-q$ )

$$2 \times q + 0 \times (1 - q) = 0 \times q + 1 \times (1 - q)$$

$$1 \times p + 0 \times (1 - p) = 0 \times p + 2 \times (1 - p)$$

- Mixed strategy Nash equilibrium is  $\begin{cases} p = 2/3 \\ q = 1/3 \end{cases}$

# Existence of Nash equilibrium

- Theorem (Nash '50): Every finite strategic game has a mixed strategy Nash equilibrium.
- Example: Matching Pennies game has a mixed strategy Nash equilibrium  $(\frac{1}{2}, \frac{1}{2}; \frac{1}{2}, \frac{1}{2})$

	Heads	Tails
Heads	1,-1	-1,1
Tails	-1,1	1,-1

- Proof: using Kakutani's fixed point theorem. See section 1.3.1 of the handout for details

# Continuous strategy game

- Theorem (Debreu '52; Glicksberg '52; Fan '52): Consider a strategic game  $\{N, S_{i \in N}, u_{i \in N}\}$  with continuous strategy space. A pure strategy Nash equilibrium exists if
  - $S_i$  is nonempty compact convex set
  - $u_i$  is continuous in  $S$  and quasi-concave in  $S_i$
- Theorem (Glicksberg '52): Consider a strategic game  $\{N, S_{i \in N}, u_{i \in N}\}$  with nonempty compact strategy space. A mixed strategy Nash equilibrium exists if  $u_i$  is continuous.

# Correlated equilibrium

- In Nash equilibrium, players choose strategies **independently**. How about players observing some common signals?
- Traffic intersection game

	Stop	Go
Stop	2,2	1,3
Go	3,1	0,0

- Two pure Nash equilibria: (stop, go) and (go, stop)
- One mixed strategy equilibrium:  $(\frac{1}{2}, \frac{1}{2}; \frac{1}{2}, \frac{1}{2})$
- If there is a traffic signal such that with probability  $\frac{1}{2}$  (red light) players play (stop, go) and with probability  $\frac{1}{2}$  (green light) players play (go, stop). This is a correlated equilibrium.

- Def: **correlated equilibrium** is a probability distribution  $p(\cdot)$  over the pure strategy space such that for all  $i$

$$\sum_{s_{-i}} p(s_i, s_{-i}) [u_i(s_i, s_{-i}) - u_i(t_i, s_{-i})] \geq 0 \text{ for all } s_i, t_i \in S_i$$

- A mixed strategy Nash equilibrium is a correlated equilibrium
- The set of correlated equilibria is convex and contains the convex hull of mixed strategy Nash equilibria

# Dynamics in games

- ❑ Nash equilibrium is a very strong concept. It assumes player strategies, payoffs and rationality are "common knowledge".
- ❑ "Game theory lacks a general and convincing argument that a Nash outcome will occur".
- ❑ One justification is that equilibria arise as a result of adaptation (learning).
  - ❑ Consider repeated play of the strategic game
  - ❑ Players are myopic, and adjust their strategies based on the strategies of other players in previous rounds.

- Best response

$$s_i(t+1) = B_i(s_{-i}(t))$$

- Fictitious play, regret-based heuristics, etc

- Many if not most network algorithms are repeated and adaptive, and achieving some equilibria. Will discuss these and networking games later in this course.

# Classical mechanism design (MD)

- Mechanisms: Protocols to implement an outcome (equilibrium) with desired system-wide properties despite the self-interest and private information of agents.
- Mechanism design: the design of such mechanisms.
- Provide an introduction to game theoretic approach to classical mechanism design

# Game theoretic approach to MD

- Start with a strategic model of agent behavior
- Design rules of a game, so that when agents play as assumed the outcomes with desired properties happen

# Incomplete information games

- Players have private type  $(\theta_1, \theta_2, \dots, \theta_N) \in \Theta$
- Strategy  $s_i(\theta_i) \in S_i$  is a function of a player's type
- Payoff  $u_i(s(\theta), \theta_i) \in R$  is a function of player's type
- Assume types are drawn from some objective distribution  $p(\theta_1, \theta_2, \dots, \theta_N)$
- Def: a strategy profile  $s^*$  is a **Bayesian-Nash equilibrium** if every players  $i$  plays a best response to maximize expected payoff given its belief about distribution  $p(\theta_{-i} | \theta_i)$ , i.e.,

$$s^*_i(\theta_i) \in \arg \max_{s_i} \sum_{\theta_{-i}} p(\theta_{-i} | \theta_i) u_i(s_i, s^*_{-i}(\theta_{-i}), \theta_i)$$

# Example: variant of Battle of the Sexes

- Two types: either wants to meet the other or does not
- Assume row player wants to meet column player, but not sure if column player want to meet her or not (assign  $\frac{1}{2}$  probability to each case); and column player knows row player's type
- If column player want to meet row player, the payoffs are

	Ballet	Soccer
Ballet	2,1	0,0
Soccer	0,0	1,2

- If column player does not want to meet row player, the payoffs are

	Ballet	Soccer
Ballet	2,0	0,2
Soccer	0,1	1,0

- The Bayesian-Nash equilibrium is (Ballet, (Ballet, Soccer))
  - $E[\text{Ballet}, (\text{Ballet}, \text{Soccer})] = \frac{1}{2} \times 2 + \frac{1}{2} \times 0 = 1$
  - $E[\text{Soccer}, (\text{Ballet}, \text{Soccer})] = \frac{1}{2} \times 0 + \frac{1}{2} \times 1 = \frac{1}{2}$

# Stronger solution concepts

- Def: a strategy profile  $s^*$  is **Ex post Nash equilibrium** if every player  $i$ 's strategy is best response whatever the type of others

$$s_i^*(\theta_i) \in \arg \max_{s_i} u_i(s_i, s_{-i}^*(\theta_{-i}), \theta_i) \text{ for all } \theta_{-i}$$

- Def: a strategy profile  $s^*$  is **dominant strategy equilibrium** if every player  $i$ 's strategy is best response whatever the type and whatever strategy of others

$$s_i^*(\theta_i) \in \arg \max_{s_i} u_i(s_i, s_{-i}(\theta_{-i}), \theta_i) \text{ for all } s_{-i}, \theta_{-i}$$

# Example: second price auction

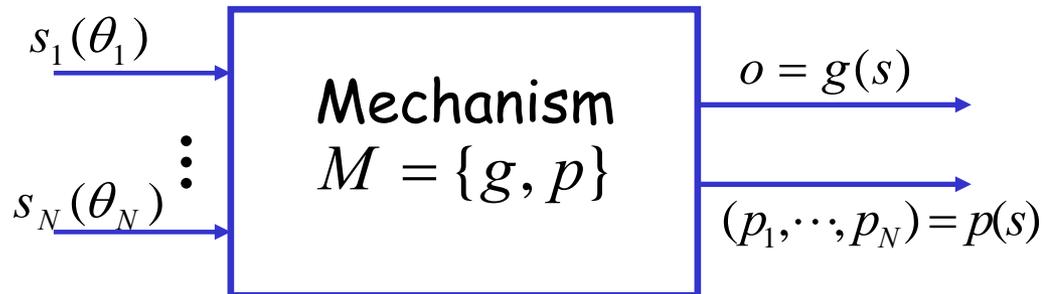
- The type is player valuation  $v_i$
- Each player submit bid  $b_i(v_i)$
- A dominant strategy is to bid  $b_i^*(v_i) = v_i$
- Players don't need to know valuations (types), or strategies of others.

# Model of MD

- Set of alternative outcomes  $O$
- Player  $i$  has private information (type)  $\theta_i$
- Type defines a value function  $v_i(o; \theta_i) \in R$  for outcome  $o \in O$  for each player  $i$
- Player payoff  $u_i(o; \theta_i) = v_i(o; \theta_i) - p_i$  for outcome  $o$  and payments  $p_i$
- The desired properties are encapsulated in the social choice function  $f : \Theta \rightarrow O$ 
  - e.g., choose  $o$  to maximize social welfare, i.e.,

$$f(\theta) = \arg \max_{o \in O} \sum_i u_i(o; \theta)$$

- The goal is to implement social choice function  $f(\theta)$



- A mechanism is defined by an outcome rule  $g : S \rightarrow O$  and a payment rule  $p : S \rightarrow R^n$
- A mechanism  $M$  implements social choice function  $f(\theta)$  if  $g(s^*_1(\theta_1), \dots, s^*_N(\theta_N)) = f(\theta)$ , where the strategy profile  $(s^*_1, \dots, s^*_N)$  is an equilibrium solution of the game induced by  $M$ .

# Properties of social choice functions and mechanisms

□ Pareto optimal:

if for every  $a \neq f(\theta)$ ,  $u_i(a, \theta) > u_i(scf(\theta), \theta) \Rightarrow \exists j u_j(a, \theta) < u_j(scf(\theta), \theta)$

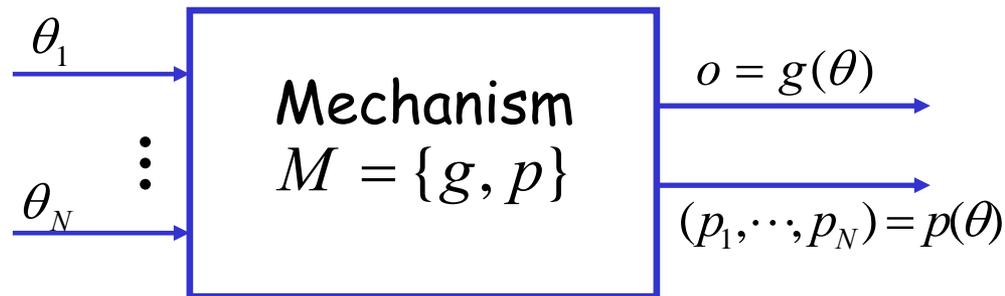
□ Efficient: if  $f(\theta) \in \arg \max_a \sum_i v_i(a, \theta_i)$

□ Budget-balance: if  $\sum_i p_i(\theta) = 0$

□ A mechanism that implements the corresponding social choice functions is called Pareto optimal, efficient or budget-balanced mechanisms, respectively.

# Incentive-compatible mechanism

- Revelation principle: any mechanism can be transformed into an incentive compatible, direct-revelation mechanism that implements the same social choice function
- Direct-revelation mechanism is a mechanism in which player strategy space is restricted to their types



- Incentive-compatible means the equilibrium strategy is to report truthful information about their types (truth-revelation).
  - First price auction is not incentive-compatible. In first price auction, the buyer with highest bid gets the object and pays his bid.
  - The second price auction is incentive compatible, direct-revelation mechanism.

# Truthful mechanism

- Truthful (aka "strategy-proof") mechanism: truth-revelation is a dominant strategy equilibrium.
  - Very robust to assumption about agent rationality and information about each other
  - An agent can compute its optimal strategy without modeling the types and strategies of others

# Vickrey-Clarke-Groves mechanisms

## □ VCG mechanism:

□ Collect  $\theta = (\theta_1, \theta_2, \dots, \theta_N)$  from agents

□  $g(\theta)$ : select an outcome  $o^* \in \arg \max_{o \in O} \sum_i v_i(o; \theta_i)$

□  $p(\theta)$ : agent  $i$  pays  $\sum_{j \neq i} v_j(o^{-i}; \theta_j) - \sum_{j \neq i} v_j(o^*; \theta_j)$ ,  
where  $o^{-i} \in \arg \max_{o \in O} \sum_{j \neq i} v_j(o; \theta_j)$

- Theorem: VCG mechanism is efficient and truthful.
- Proof:  $u_i(\theta_i, \theta_{-i}) = v_i(o^*; \theta_i) + \sum_{j \neq i} v_j(o^*; \theta_j) - \sum_{j \neq i} v_j(o^{-i}; \theta_j)$
- VCG mechanism is the only mechanism that is efficient and strategy-proof amongst direct-revelation mechanisms.

# Combinatorial auction

- Goods  $P$
- Outcomes: allocations  $A = (A_1, \dots, A_N)$ , where  $A_i \subseteq P$  and  $A_i$ s are not overlapped.
- Agent valuation  $v_i(A_i; \theta_i)$  for  $A_i \subseteq P$
- Goal: allocate goods to maximize  $\sum_i v_i(A_i; \theta_i)$
- Applications: wireless spectrum auction, course scheduling, ...

- Two items  $A$  and  $B$ ; 3 agents (taken from Parkes)
- Valuation

	A	B	AB
1	5	0	5
2	0	5	5
3	0	0	12

- Agent 3 wins  $AB$  and pays  $10-0=10$ .

□ Another valuation

	A	B	AB
1	5	0	5
2	0	5	5
3	0	0	7

□ Agents 1 and 2 win and each pays  $7-5=2$

# Remarks

- ❑ Only consider the incentive issue: to overcome the self-interest of agents
- ❑ Not discuss computational and informational issues. Will discuss these in distributed mechanism design and its applications in networking.