

Welcome to

**CMU 15-896**  
**Algorithms, Games, and Networks**

Spring 2013

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Overall

Theory and algorithms for systems of interacting agents, each with their own interests in mind.

Topics

- Basics of game theory, equilibria
- Quality of equilibria: price of anarchy
- Social choice: voting, manipulation
- Mechanism design: designing rules of the game to achieve desired outcome, auctions.
- Kidney exchange, matching markets
- Social networks
- Fair division

Admin

Free online book: "Algorithmic game theory"

Course requirements:

- 4 homeworks
- Final project
- Scribing one lecture
- Helping grade one homework
- Participation in class

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01/15/13

**A Basic Introduction to Game Theory**

Avrim Blum

[Readings: Ch. 1.1-1.3 of AGT book]

Game theory

- Field developed by economists to study social & economic interactions.
  - Wanted to understand why people behave the way they do in different economic situations. Effects of incentives. Rational explanation of behavior.
- "Game" = interaction between parties with their own interests. Could be called "interaction theory".
- Big in CS for understanding large systems:
  - Internet routing, social networks, e-commerce
  - Problems like spam etc.

## Setting

- Have a collection of participants, or *players*.
- Each has a set of choices, or *strategies* for how to play/behave.
- Combined behavior results in *payoffs* (satisfaction level) for each player.

Most examples today will involve just 2 players (which will make them easier to picture, as will become clear in a moment...)

## Example: walking on the sidewalk

- What side of sidewalk should I walk on?
- Two options for you (left or right). Same for person walking towards you.
- Can describe payoffs in matrix:

		Left	Right
Left	(1,1)	(-1,-1)	
	(-1,-1)	(1,1)	

Your payoff for RR      His payoff for RR

## Key notion

- "Nash Equilibrium": pair of strategies such that each player is playing a best-response to the other. Neither has an incentive to change.

		Left	Right
Left	(1,1)	(-1,-1)	
	(-1,-1)	(1,1)	

Your payoff for RR      His payoff for RR

## Example: prisoner's dilemma

- Consider two companies deciding whether to install pollution controls.
- Imagine pollution controls cost \$4 but improve everyone's environment by \$3

		control	don't control
control	(2,2)	(-1,3)	
	(3,-1)	(0,0)	

What to do? Both companies have incentives to defect, but together they can get good overall behavior.

## Example: matching pennies / penalty shot

- Shooter can choose to shoot left or shoot right.
- Goalie can choose to dive left or dive right.
- If goalie guesses correctly, (s)he saves the day. If not, it's a **GOAALLL!!** Vice-versa for shooter.

		Left	Right
Left	(0,0)	(1,-1)	
	(1,-1)	(0,0)	

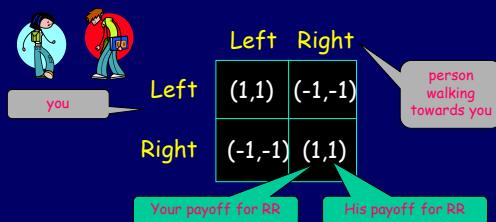
Each playing 50/50 is Nash equilibrium

## Nash (1950)

- Proved that if you allow **randomized** (mixed) strategies then every game has at least one equilibrium.
- I.e., a **pair of (randomized) strategies** that is **stable** in the sense that each is a best response to the other in terms of expected payoff.
- For this, and its implications, Nash received the Nobel prize.

## Game theory terminology

- Rows and columns called pure strategies.
- Randomized algs called mixed strategies.



## Game theory terminology

- Rows and columns called pure strategies.
- Randomized algs called mixed strategies.
- Often describe in terms of 2 matrices  $R$ ,  $C$ .

$R$	1	-1
	-1	1

$C$	1	-1
	-1	1

$(p, q)$  is Nash equilib if  $p^T R q \geq e_i^T R q \forall i$  and  $p^T C q \geq p^T C e_j \forall j$ .

## Basic facts

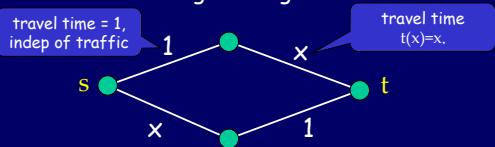
- $(p, q)$  is NashEq if  $p^T R q \geq e_i^T R q \forall i$ ,  $p^T C q \geq p^T C e_j \forall j$ .
- $\Rightarrow$  for all  $i$  s.t.  $p_i > 0$  we have  $e_i^T R q = \max_j e_j^T R q$
- $\Rightarrow$  for all  $j$  s.t.  $q_j > 0$  we have  $p^T C e_j = \max_i p^T C e_j$

1	-1
-1	1

1	-1
-1	1

## NE can do strange things

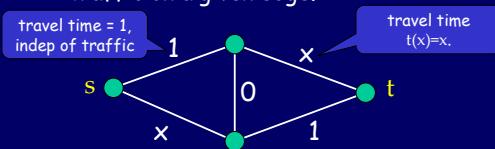
- Braess paradox:
  - Road network, traffic going from  $s$  to  $t$ .
  - travel time as function of fraction  $x$  of traffic on a given edge.



Fine. NE is 50/50. Travel time = 1.5

## NE can do strange things

- Braess paradox:
  - Road network, traffic going from  $s$  to  $t$ .
  - travel time as function of fraction  $x$  of traffic on a given edge.



Add new superhighway. NE: everyone uses zig-zag path. Travel time = 2.

## 2-Player Zero-Sum games

- Zero-sum games are the special case of purely-competitive 2-player games.

• Recall: an entry  $(x, y)$  means:  $x$  = payoff to row player,  $y$  = payoff to column player. "Zero sum" means that  $y = -x$ .

- E.g., matching pennies / penalty shot:

$shooter$	Left	Right	$goalie$
	(0,0)	(1,-1)	
	(1,-1)	(0,0)	<b>GOALLL!!</b>

## Minimax-optimal strategies

- Minimax optimal strategy is a (randomized) strategy that has the best guarantee on its expected gain, over choices of the opponent. [maximizes the minimum]
- I.e., the thing to play if your opponent knows you well.

	Left	Right	shooter	goalie
Left	(0,0)	(1,-1)		GOAALLL!!
Right	(1,-1)	(0,0)		

Minimax optimal for both players is 50/50. Gives expected gain of  $\frac{1}{2}$  for shooter,  $-\frac{1}{2}$  for goalie. Any other is worse.

## Minimax-optimal strategies

- How about penalty shot with goalie who's weaker on the left?

Say shooter uses  $(p, 1-p)$ .

- If goalie dives left, gets  $p/2 + 1-p = 1 - p/2$ .
- If goalie dives right, gets  $p$ .
- Maximize minimum by setting equal.
- Gives  $p = 2/3$ .

	Left	Right	shooter	goalie
Left	$(\frac{1}{2}, -\frac{1}{2})$	(1,-1)		GOAALLL!!
Right	(1,-1)	(0,0)		50/50

## Minimax-optimal strategies

- How about penalty shot with goalie who's weaker on the left?

Minimax optimal for shooter is  $(2/3, 1/3)$ .

Guarantees expected gain at least  $2/3$ .

Minimax optimal for goalie is also  $(2/3, 1/3)$ .

Guarantees expected loss at most  $2/3$ .

	Left	Right	shooter	goalie
Left	$(\frac{1}{2}, -\frac{1}{2})$	(1,-1)		GOAALLL!!
Right	(1,-1)	(0,0)		50/50

## Minimax-optimal strategies

- Can solve for minimax optimal strategy using Linear Programming:

Variables  $p, v$ .

Maximize  $v$  subject to:

- $p \cdot M_j \geq v$ , for all  $j$ .
- $p$  is legal prob dist ( $p_i \geq 0, \sum_i p_i = 1$ ).

	Left	Right	shooter	goalie
Left	$(\frac{1}{2}, -\frac{1}{2})$	(1,-1)		GOAALLL!!
Right	(1,-1)	(0,0)		50/50

## Minimax Theorem (von Neumann 1928)

- Every 2-player zero-sum game has a unique value  $V$ .
- Minimax optimal strategy for  $R$  guarantees  $R$ 's expected gain at least  $V$ .
- Minimax optimal strategy for  $C$  guarantees  $C$ 's expected loss at most  $V$ .

**Counterintuitive:** Means it doesn't hurt to publish your strategy if both players are optimal. (Borel had proved for symmetric 5x5 but thought was false for larger games)

## Nash $\Rightarrow$ Minimax

- Nash's theorem actually gives minimax thm as a corollary.
  - Pick some NE and let  $V =$  value to row player in that equilibrium.
  - Since it's a NE, neither player can do better even knowing the (randomized) strategy their opponent is playing.
  - So, they're each playing minimax optimal.

## Nash $\Rightarrow$ Minimax

- On the other hand, for minimax, also have very constructive, algorithmic arguments:
  - Can solve for minimax optimum using linear programming in time  $\text{poly}(n)$  ( $n = \text{size of game}$ )
  - Have adaptive procedures that in repeated play guarantee to approach/beat best fixed strategy in hindsight
- But for Nash, no efficient procedures to find: NP-hard to find equilib with special properties, PPAD-hard just to find one.

Can use notion of minimax optimality to explain bluffing in poker

## Simplified Poker (Kuhn 1950)

- Two players **A** and **B**.
- Deck of 3 cards: 1,2,3.
- Players ante \$1.
- Each player gets one card.
- **A** goes first. Can bet \$1 or pass.
  - If **A** bets, **B** can call or fold.
  - If **A** passes, **B** can bet \$1 or pass.
    - If **B** bets, **A** can call or fold.
- High card wins (if no folding). Max pot \$2.

- Two players **A** and **B**. 3 cards: 1,2,3.
- Players ante \$1. Each player gets one card.
- **A** goes first. Can bet \$1 or pass.
  - If **A** bets, **B** can call or fold.
  - If **A** passes, **B** can bet \$1 or pass.
    - If **B** bets, **A** can call or fold.

## Writing as a Matrix Game

- For a given card, **A** can decide to
  - Pass but fold if **B** bets. [PassFold]
  - Pass but call if **B** bets. [PassCall]
  - Bet. [Bet]
- Similar set of choices for **B**.

## Can look at all strategies as a big matrix...

	[FP,FP,CB]	[FP,CP,CB]	[FB,FP,CB]	[FB,CP,CB]
[PF,PF,PC]	0	0	-1/6	-1/6
[PF,PF,B]	0	1/6	-1/3	-1/6
[PF,PC,PC]	-1/6	0	0	1/6
[PF,PC,B]	-1/6	-1/6	1/6	1/6
[B,PF,PC]	-1/6	0	0	1/6
[B,PF,B]	1/6	-1/3	0	-1/2
[B,PC,PC]	1/6	-1/6	-1/6	-1/2
[B,PC,B]	0	-1/2	1/3	-1/6
	0	-1/3	1/6	-1/6

## And the minimax optimal strategies are...

- **A**:
  - If hold 1, then 5/6 PassFold and 1/6 Bet.
  - If hold 2, then  $\frac{1}{2}$  PassFold and  $\frac{1}{2}$  PassCall.
  - If hold 3, then  $\frac{1}{2}$  PassCall and  $\frac{1}{2}$  Bet.

Has both bluffing and underbidding...
- **B**:
  - If hold 1, then 2/3 FoldPass and 1/3 FoldBet.
  - If hold 2, then 2/3 FoldPass and 1/3 CallPass.
  - If hold 3, then CallBet

Minimax value of game is -1/18 to **A**.

## One more interesting game

“Ultimatum game”:

- Two players “Splitter” and “Chooser”
- 3<sup>rd</sup> party puts \$10 on table.
- **Splitter** gets to decide how to split between himself and Chooser.
- **Chooser** can accept or reject.
- If reject, money is burned.

## One more interesting game

“Ultimatum game”: E.g., with \$4

			1	2	3
			Chooser: how much to accept	Splitter: how much to offer chooser	
1	1	(1,3)	(2,2)	(3,1)	
	2	(0,0)	(2,2)	(3,1)	
	3	(0,0)	(0,0)	(3,1)	

## How to prove existence of NE

- Proof will be non-constructive.
- Notation:
  - Assume an  $n \times n$  matrix.
  - Use  $(p_1, \dots, p_n)$  to denote mixed strategy for row player, and  $(q_1, \dots, q_n)$  to denote mixed strategy for column player.

## Proof

- We'll start with Brouwer's fixed point theorem.
  - Let  $S$  be a bounded convex region in  $\mathbb{R}^n$  and let  $f: S \rightarrow S$  be a continuous function.
  - Then there must exist  $x \in S$  such that  $f(x) = x$ .
  - $x$  is called a “fixed point” of  $f$ .
- Simple case:  $S$  is the interval  $[0,1]$ .
- We will care about:
  - $S = \{(p,q): p,q \text{ are legal probability distributions on } 1, \dots, n\}$ . I.e.,  $S = \text{simplex}_n \times \text{simplex}_n$

## Proof (cont)

- $S = \{(p,q): p,q \text{ are mixed strategies}\}$ .
- Want to define  $f(p,q) = (p',q')$  such that:
  - $f$  is continuous. This means that changing  $p$  or  $q$  a little bit shouldn't cause  $p'$  or  $q'$  to change a lot.
  - Any fixed point of  $f$  is a Nash Equilibrium.
- Then Brouwer will imply existence of NE.

## Try #1

- What about  $f(p,q) = (p',q')$  where  $p'$  is best response to  $q$ , and  $q'$  is best response to  $p$ ?
- Problem: not continuous:
  - E.g., penalty shot: If  $p = (0.51, 0.49)$  then  $q' = (1,0)$ . If  $p = (0.49, 0.51)$  then  $q' = (0,1)$ .

		Left	Right
Left	Left	(0,0)	(1,-1)
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$$R = \begin{array}{|c|c|} \hline 0 & 1 \\ \hline 1 & 0 \\ \hline \end{array}$$

$$C = \begin{array}{|c|c|} \hline 0 & -1 \\ \hline -1 & 0 \\ \hline \end{array}$$

### Try #1

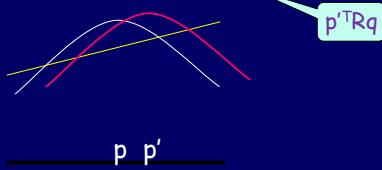
- What about  $f(p,q) = (p',q')$  where  $p'$  is best response to  $q$ , and  $q'$  is best response to  $p$ ?
- Problem: also not necessarily well-defined:
  - E.g., if  $p = (0.5, 0.5)$  then  $q'$  could be anything.

$$R = \begin{array}{|c|c|} \hline 0 & 1 \\ \hline 1 & 0 \\ \hline \end{array}$$

$$C = \begin{array}{|c|c|} \hline 0 & -1 \\ \hline -1 & 0 \\ \hline \end{array}$$

### Instead we will use...

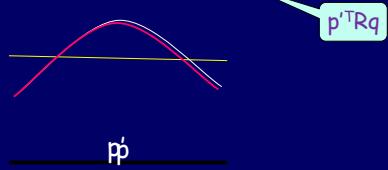
- $f(p,q) = (p',q')$  such that:
  - $q'$  maximizes  $[(\text{expected gain wrt } p) - \|q-q'\|^2]$
  - $p'$  maximizes  $[(\text{expected gain wrt } q) - \|p-p'\|^2]$



Note: quadratic + linear = quadratic.

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  - $p'$  maximizes  $[(\text{expected gain wrt } q) - \|p-p'\|^2]$
- $f$  is well-defined and continuous since quadratic has unique maximum and small change to  $p,q$  only moves this a little.
- Also fixed point = NE. (even if tiny incentive to move, will move little bit).
- So, that's it!

### Algorithmic Game Theory

Algorithmic issues in game theory:

- Computing equilibria / approximate equilibria in different kinds of games
- Understanding **quality** of equilibria in **load-balancing, network-design, routing, machine scheduling...**
- Analyzing **dynamics** of simple behaviors or adaptive (learning) algorithms: **quality guarantees, convergence,...**
- Design issues: constructing rules so that game will (ideally) have dominant-strategy equilibria with good properties.