Game Theory

15-451

12/05/06

- Zero-sum games
- General-sum games

Shall we play a game?

Game Theory and Computer
Science

Plan for Today

- · 2-Player Zero-Sum Games (matrix games)
 - Minimax optimal strategies
 - Minimax theorem and proof

test material not test material

GOAALLL!!!

No goal

- General-Sum Games (bimatrix games)
 - notion of Nash Equilibrium
- · Proof of existence of Nash Equilibria
 - using Brouwer's fixed-point theorem

Consider the following scenario...

- Shooter has a penalty shot. 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 goooooaaaaall!
- · Vice-versa for shooter.

2-Player Zero-Sum games

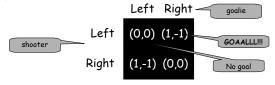
- Two players R and C. Zero-sum means that what's good for one is bad for the other.
- Game defined by matrix with a row for each of R's options and a column for each of C's options.
 Matrix tells who wins how much.
 - an entry (x,y) means: x = payoff to row player, y = payoff to column player. "Zero sum" means that y = -x.
- E.g., penalty shot: Left Right =

shooter

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

<u>Minimax-optimal strategies</u>

- 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.



Minimax-optimal strategies

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- I.e., the thing to play if your opponent knows you well.
- In class on Linear Programming, we saw how to solve for this using LP.
 - polynomial time in size of matrix if use poly-time LP alg.

Minimax-optimal strategies

• E.g., penalty shot:

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

Minimax optimal strategy 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

• E.g., penalty shot with goalie who's weaker on the left.

> Left Right $(\frac{1}{2}, -\frac{1}{2})(1, -1)$ (1,-1) (0,0)

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.

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)

Matrix games and Algorithms

- · Gives a useful way of thinking about guarantees on algorithms for a given problem.
- · Think of rows as different algorithms, columns as different possible inputs.
- M(i,j) = cost of algorithm i on input j.

· Algorithm design goal: good strategy for row player. Lower bound: good strategy for adversary.

One way to think of upper-bounds/lower-bounds: on value of this game

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E.g., sorting

· Algorithm design goal: good strategy for row player. Lower bound: good strategy for adversary.

Of course matrix may be HUGE. But helpful conceptually.

Matrix games and Algs

- ·What is a deterministic alg with a good worst-case quarantee?
 - A row that does well against all columns.
- ·What is a lower bound for deterministic algorithms?
 - Showing that for each row i there exists a column j such that M(i,j) is bad.
- ·How to give lower bound for randomized algs?
 - Give randomized strategy for adversary that is bad for all i. Must also be bad for all distributions over i.

E.g., hashing



- ·Rows are different hash functions.
- ·Cols are different sets of n items to hash.
- ·M(i,j) = #collisions incurred by alg i on set j.

We saw:

- ·For any row, can reverse-engineer a bad column.
- •Universal hashing is a randomized strategy for row player that has good behavior for every column.
- For any set of inputs, if you randomly construct hash function in this way, you won't get many collisions in expectation.

Nice proof of minimax thm (sketch)

- Suppose for contradiction it was false.
- This means some game G has $V_C > V_R$:
 - If Column player commits first, there exists a row that gets at least $\rm V_{\it C}$.
 - But if Row player has to commit first, the Column player can make him get only V_R.
- Scale matrix so payoffs to row are in [0,1]. Say $V_R = V_C \delta$.



Proof sketch, contd

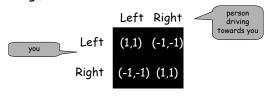
- Consider exponential weighting alg from Nov16 lecture as Row, against opponent who always plays best response to Row's distrib.
- · In T steps,
 - Alg gets $\geq (1-\epsilon/2)OPT$ $\log(n)/\epsilon$ [use $\epsilon=\delta$]
 - OPT \geq T·V_C [Best against opponent's empirical distribution]
 - Alg $\leq \text{T-V}_R$ [Each time, opponent knows your randomized strategy]
 - Gap is δT . Contradicts assumption once δT > $(\epsilon/2)OPT + log(n)/\epsilon$.

General-Sum Games

- Zero-sum games are good formalism for design/analysis of algorithms.
- General-sum games are good models for systems with many participants whose behavior affects each other's interests
 - E.g., routing on the internet
 - E.g., online auctions

General-sum games

- In general-sum games, can get win-win and lose-lose situations.
- E.g., "what side of road to drive on?":



<u>General-sum games</u>

- In general-sum games, can get win-win and lose-lose situations.
- E.g., "which movie should we go to?":

Borat Happy-feet
(8,2) (0,0)

Happy-feet (0,0) (2,8)

No longer a unique "value" to the game.

Nash Equilibrium

- A Nash Equilibrium is a stable pair of strategies (could be randomized).
- Stable means that neither player has incentive to deviate on their own.
- E.g., "what side of road to drive on":

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

NE are: both left, both right, or both 50/50.

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NE are: both B, both HF, or (80/20,20/80)

Uses

- Economists use games and equilibria as models of interaction.
- E.g., pollution / prisoner's dilemma:
- (imagine pollution controls cost \$4 but improve everyone's environment by \$3)

don't pollute pollute

don't pollute

(2,2) (-1,3)

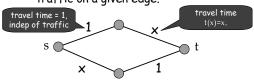
pollute

(3,-1) (0,0)

Need to add extra incentives to get good overall behavior.

NE can do strange things

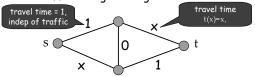
- · Braess paradox:
 - Road network, traffic going from s to t.
 - travel time as function of fraction x of traffic on a given edge.



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.

Existence of NE

- Nash (1950) proved: any general-sum game must have at least one such equilibrium.
 - Might require randomized strategies (called "mixed strategies")
- This also yields 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.

Existence of NE

- · Proof will be non-constructive.
- Unlike case of zero-sum games, we do not know any polynomial-time algorithm for finding Nash Equilibria in n × n general-sum games. [known to be "PPAD-hard"]
- · Notation:
 - Assume an nxn matrix.
 - Use $(p_1,...,p_n)$ to denote mixed strategy for row player, and $(q_1,...,q_n)$ to denote mixed strategy for column player.

Proof

- We'll start with Brouwer's fixed point theorem.
 - Let S be a compact convex region in R^n and let $f:S \to S$ 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 are legal probability distributions on 1,...,n}. I.e., $S = simplex_n \times simplex_n$

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 necessarily well-defined:
 - E.g., penalty shot: if p = (0.5,0.5) then q' could be anything.

Left Right

Left

(0,0) (1,-1)

Right

(1,-1) (0,0)

<u>Try #1</u>

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

Left Right

Left

(0,0) (1,-1)

Right

(1,-1) (0,0)

Instead we will use...

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

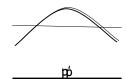


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Note: quadratic + linear = quadratic.

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 - q' maximizes [(expected gain wrt p) $||q-q'||^2$]
 - p' maximizes [(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!