L02 Games: Definitions and Existence of Nash Equilibrium

CS 280 Algorithmic Game Theory Ioannis Panageas

Definition (Normal Form Games). A normal form game is specified by

- *set of n players* $[n] = \{1, ..., n\}$
- For each player i a set of strategies/actions S_i and a utility $u_i : \times_{j=1}^n S_j \to \mathbb{R}$ denoting the payoff of i.
- set of strategy profiles $S = S_1 \times ... \times S_n$.

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Example (Rock-Paper-Scissors). We have that:

- n = 2
- $S_1, S_2 = \{R, P, S\}.$
- $u_1(R,R) = 0$, $u_1(R,P) = -1$, $u_1(R,S) = 1$, $u_1(P,R) = 1$, $u_1(P,P) = -1$, $u_1(P,S) = -1$ $u_1(S,R) = -1$, $u_1(S,P) = 1$, $u_1(S,S) = 0$.
- $u_2 = -u_1$

Definition (Mixed strategies). The set of mixed strategies available to player i are all distributions over S_i

$$\Delta_i = \{x_i : \sum_{s_i \in S_i} x_i(s_i) = 1 \text{ and } x_i \ge \mathbf{0}\}$$

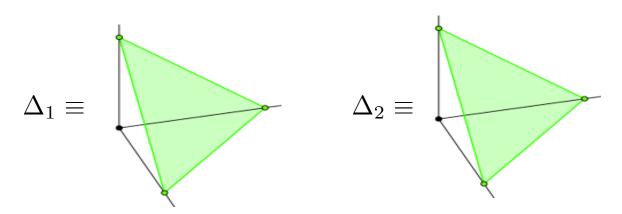
We also denote $\Delta = \Delta_1 \times ... \times \Delta_n$ the set of mixed strategies of all players and Δ_{-i} the set of mixed strategies of all players excluding i.

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Example (Rock-Paper-Scissors).

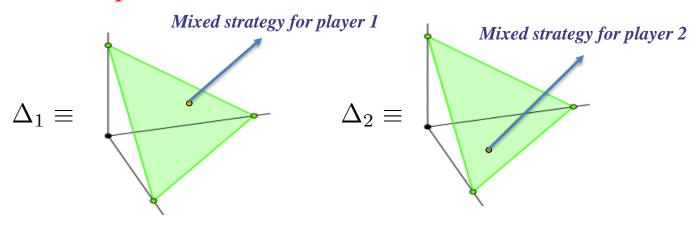


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Definition (Expected utility). Given a mixed strategy $x \in \Delta$, the expected utility of player i is

$$u_i(x) = \sum_{(s_1,...,s_n) \in S} u_i(s_1,...,s_n) \prod_{j=1} x_j(s_j)$$

or (in a more compact way)

$$u_i(x) = \mathbb{E}_{s \sim x} u_i(s).$$

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Example (Rock-Paper-Scissors). We have that:

- Say $x_1 = (\frac{3}{6}, \frac{2}{6}, \frac{1}{6})$ and $x_2 = (\frac{1}{2}, \frac{1}{2}, 0)$.
- Then $u_1(x_1, x_2) = \frac{3}{12}u_1(R, R) + \frac{3}{12}u_1(R, P) + 0u_1(R, S) + \frac{2}{12}u_1(P, R) + \frac{2}{12}u_1(P, P) + 0u_1(P, S) + \frac{1}{12}u_1(S, R) + \frac{1}{12}u_1(S, P) + 0u_1(S, S) = -\frac{3}{12} + \frac{2}{12} \frac{1}{12} + \frac{1}{12} = -\frac{1}{12}$
- $u_2(x_1, x_2) = \frac{1}{12}$

Existence of Nash Equilibrium

Definition (Nash equilibrium). A mixed strategy $x \equiv (x_1; ...; x_n) \in \Delta$ is a Nash equilbrium if and only if for all agents i and $x'_i \in \Delta_i$ we have

$$u_i(x_i; x_{-i}) \ge u_i(x_i'; x_{-i}).$$

Definition (ϵ -approximate Nash equilibrium). A mixed strategy $x \equiv (x_1; ...; x_n) \in \Delta$ is a ϵ -approximate Nash equilbrium if and only if for all agents i and $x'_i \in \Delta_i$ we have

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Theorem (Nash 51'). Every game with a finite number of players and actions has a Nash equilibrium.

Before we proceed with the proof, an important theorem from topology...

Theorem (Brouwer). Let D be a convex, compact subset of \mathbb{R}^d and $f: D \to D$ a continuous function. There always exists $x \in D$ such that

$$f(x) = x$$
.

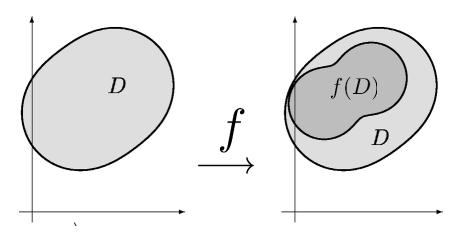
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Remark:

x is called a fixed point of f. Uniqueness is not true!



Intro to AGT

Proof. Consider any finite game. Define function $f: \Delta \to \Delta$ as follows

$$f_{is_i}(x) = \frac{x_i(s_i) + \max\{u_i(s_i; x_{-i}) - u_i(x), 0\}}{1 + \sum_{s' \in S_i} \max\{u_i(s'; x_{-i}) - u_i(x), 0\}}$$

Gain if i were to switch to pure strategy s_i instead of mixed x_i

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For RPS

$$f(x) = (f_{1R}(x), f_{1P}(x), f_{1S}(x), f_{2R}(x), f_{2P}(x), f_{2S}(x))$$

with
$$f_{1R}(x) = \left(\frac{x_{1R} + \max\{-x_{2P} + x_{2S} - u_1(x), 0\}}{1 + \max\{-x_{2P} + x_{2S} - u_1(x), 0\} + \max\{x_{2R} - x_{2S} - u_1(x), 0\} + \max\{-x_{2R} + x_{2P} - u_1(x), 0\}}\right)$$
 etc...

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Observations:

• For each player $i \Rightarrow \sum_{s' \in S_i} f_{is'}(x) = 1$. why?

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Gain if i were to switch to pure strategy s_i instead of mixed x_i

Observations:

- For each play Fixed point always exists!!
- f is continuous mapping from Δ to Δ (which is convex and compact).

Proof cont.

$$f_{is_i}(x) = \frac{x_i(s_i) + \max\{u_i(s_i; x_{-i}) - u_i(x), 0\}}{1 + \sum_{s' \in S_i} \max\{u_i(s'; x_{-i}) - u_i(x), 0\}}$$

Let x^* be a fixed point of f. We will show that x^* is a Nash Equilibrium!

Proof cont.

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Since $f(x^*) = x^*$ it holds that for all agents i and $s \in S_i$ that $f_{is}(x^*) = x_i^*(s)$

$$\Rightarrow x_i^*(s) \sum_{s' \in S_i} \max\{u_i(s'; x_{-i}^*) - u_i(x^*), 0\} = \max\{u_i(s; x_{-i}^*) - u_i(x^*), 0\}$$

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Cases:

• $x_i^*(s) = 0 \Rightarrow u_i(s; x_{-i}^*) \le u_i(x^*).$

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Cases:

- $x_i^*(s) = 0 \Rightarrow u_i(s; x_{-i}^*) \le u_i(x^*).$
- $x_i^*(s) > 0$ then if $u_i(s; x_{-i}^*) < u_i(x^*)$ we get $u_i(s'; x_{-i}^*) \le u_i(x^*)$ for all $s' \in S_1$. But then $u_i(x^*) = \sum_{s'} x_i^*(s') u_i(x^*) > \sum_{s'} u_i(s'; x_{-i}^*) x_i^*(s') = u_i(x^*)$.

Proof cont.

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 Contradiction!

Proof cont.

$$f_{is_i}(x) = \frac{x_i(s_i) + \max\{u_i(s_i; x_{-i}) - u_i(x), 0\}}{1 + \sum_{s' \in S_i} \max\{u_i(s'; x_{-i}) - u_i(x), 0\}}$$

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- $x_i^*(s) = 0 \Rightarrow u_i(s; x_{-i}^*) \le u_i(x^*).$
- $x_i^*(s) > 0$ then if $u_i(s; x_{-i}^*) \ge u_i(x^*)$.

In particular since $u_i(x^*) = \sum_{s'} u_i(s'; x_{-i}^*) x_i^*(s)$ we have $u_i(s; x_{-i}^*) = u_i(x^*)$ whenever $x_i^*(s) > 0$.

Proof cont.

$$f_{is_i}(x) = \frac{x_i(s_i) + \max\{u_i(s_i; x_{-i}) - u_i(x), 0\}}{1 + \sum_{s' \in S_i} \max\{u_i(s'; x_{-i}) - u_i(x), 0\}}$$

It holds for all agents i:

- $x_i^*(s) = 0 \Rightarrow u_i(s; x_{-i}^*) \le u_i(x^*).$
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Consider any \tilde{x}_i , we will show that

$$u_i(x_i^*; x_{-i}^*) \ge u_i(\tilde{x}_i; x_{-i}^*).$$

Proof cont.

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From two bullets we get $\tilde{x}_i(s)u_i(s, x_{-i}^*) \leq \tilde{x}_i(s)u_i(x^*)$ so taking the summation

$$u_i(\tilde{x}_i, x_{-i}^*) = \sum_{s'} \tilde{x}_i(s') u_i(s'; x_{-i}^*) \le \sum_{s'} \tilde{x}_i(s') u_i(x^*) = u_i(x^*)$$

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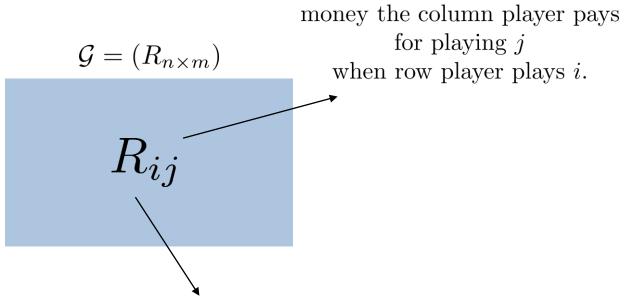
From two bullets

$$u_i(x_i^*; x_i^*) > u_i(\tilde{x}_i; x_i^*).$$

So finding Nash Equilibria is like computing fixed points! Can it be computationally hard? YES...

$$u_i(\tilde{x}_i, x_{-i}^*) = \sum_{s'} \tilde{x}_i(s') u_i(s'; x_{-i}^*) \le \sum_{s'} \tilde{x}_i(s') u_i(x^*) = u_i(x^*)$$

- 2 players: Row and Column
- *n*, *m* strategies available
- Payoff matrix R of size $n \times m$.



money the row player gets for playing i when column player plays j.

Column player chooses $y \in \Delta_m$

Row player chooses $x \in \Delta_n$

 R_{ij}

Row gets $x^{\top}Ry$. Column pays $x^{\top}Ry$.

Example: Two candidates are aiming for presidency.

	Tax-cuts	Society
Economy	3, -3	-1,1
Education	-2, 2	1,-1

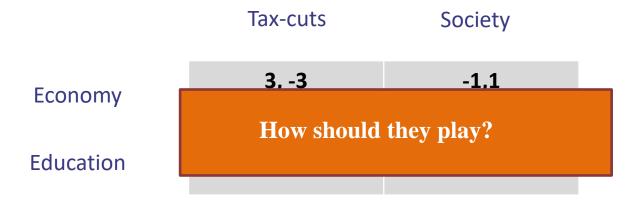
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Suppose row player plays (x_{11}, x_{12}) . How should column player respond?

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Suppose row player plays (x_{11}, x_{12}) . How should column player respond?

Answer: If she chooses Tax-cuts she gets in expectation $u_2(x_1, 'Tax - cuts') = -3x_{11} + 2x_{12}$ and if she chooses Society, she gets $u_2(x_1, 'Society') = x_{11} - x_{12}$.

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Column plays best response: Column should get

$$\max\{-3x_{11}+2x_{12},x_{11}-x_{12}\}.$$

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Intro to AGT

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If row wants to maximize her utility, she should play then

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Linear Program for Row player

$$\max z$$
s.t $3x_{11} - 2x_{12} \ge z$

$$-x_{11} + x_{12} \ge z$$

$$x_{11} + x_{12} = 1$$

$$x_{11}, x_{12} \ge 0$$

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Linear Program for Row player
$$\max z$$
 $s.t \ 3x_{11} - 2x_{12} \ge z$ $-x_{11} + x_{12} \ge z$ $x_{11} + x_{12} = 1$ $x_{11}, x_{12} \ge 0$ Sol $x_1 = (\frac{3}{7}, \frac{4}{7}), z = \frac{1}{7}$ Row gets at least 1/7!

Suppose now that column player plays (x_{21}, x_{22}) . How should row player respond?

Answer: If she chooses Economy she gets in expectation $u_1('Economy', x_2) = 3x_{21} - x_{22}$ and if she chooses Education, she gets $u_1('Education', x_2) = -2x_{21} + x_{22}$.

Row plays best response: Row should get

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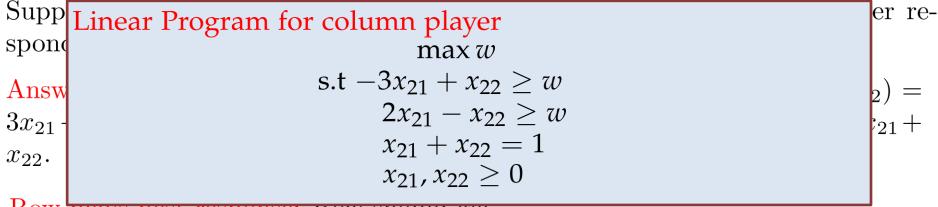
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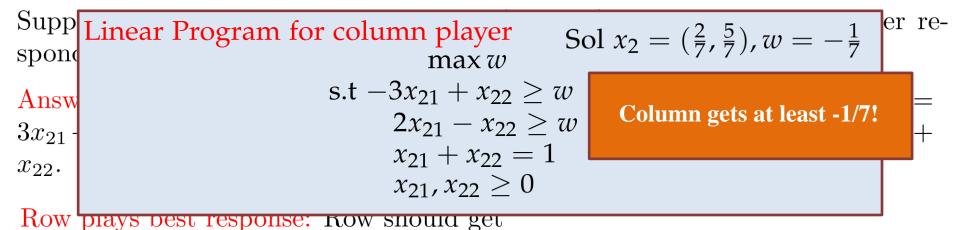
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Column gets (zero-sum):

$$\min\{-3x_{21} + x_{22}, 2x_{21} - x_{22}\}.$$

If Column wants to maximize her utility, she should play then

$$(x_{21}^*, x_{22}^*) = \arg\max_{x_{21}, x_{22}} \min\{-3x_{21} + x_{22}, 2x_{21} - x_{22}\}\$$

Supp Supp Linear Program for column player
$$\max w$$
 Sol $x_2 = (\frac{2}{7}, \frac{5}{7}), w = -\frac{1}{7}$ er respond

Since zero sum $(\frac{3}{7}, \frac{4}{7}), (\frac{2}{7}, \frac{5}{7})$ must be a Nash equilibrium!!

Next lecture LP duality to get NE and minimax theorem!

$$(x_{21}^*, x_{22}^*) = \arg\max_{x_{21}, x_{22}} \min\{-3x_{21} + x_{22}, 2x_{21} - x_{22}\}\$$