

Lecture 8

Games and Nash Equilibrium

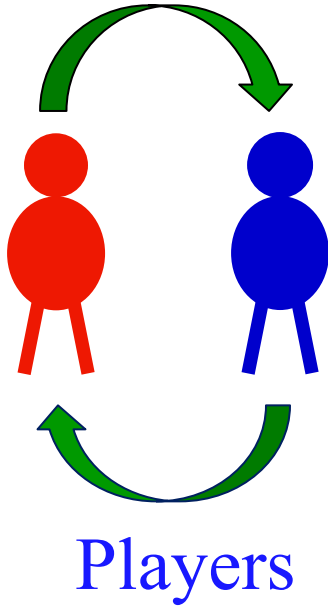
CS 598RM

22nd September 2020

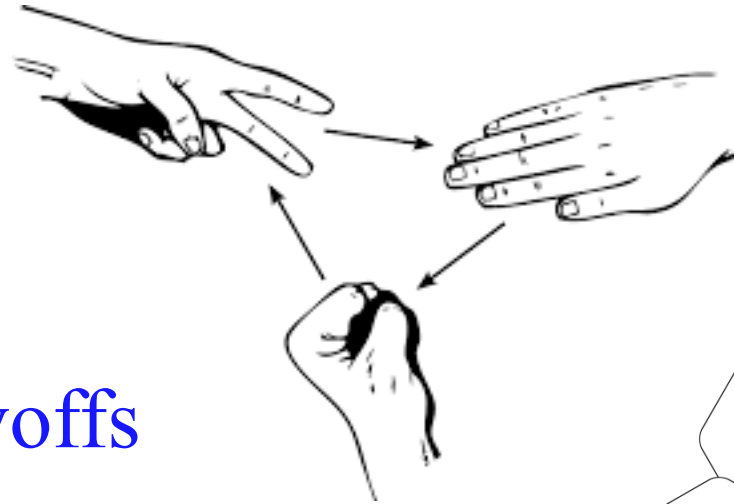
Instructor: [Ruta Mehta](#)



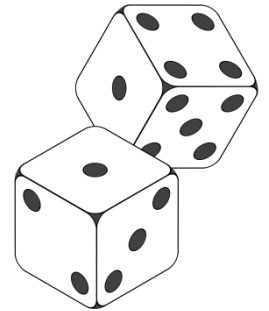
Games



Payoffs

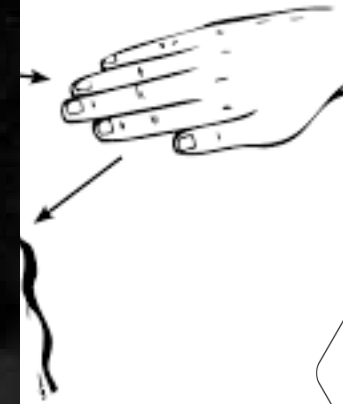
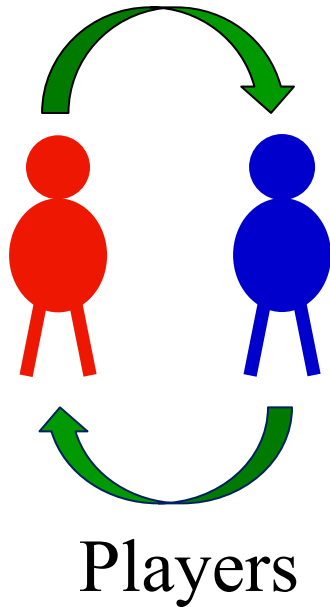


Strategies

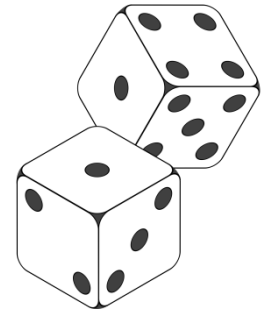


Randomize!

Games



Strategies



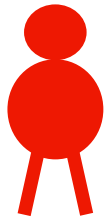
Randomize!

Nash (1950):

There exists a (stable) state where no player gains by unilateral deviation.

Nash equilibrium (NE)

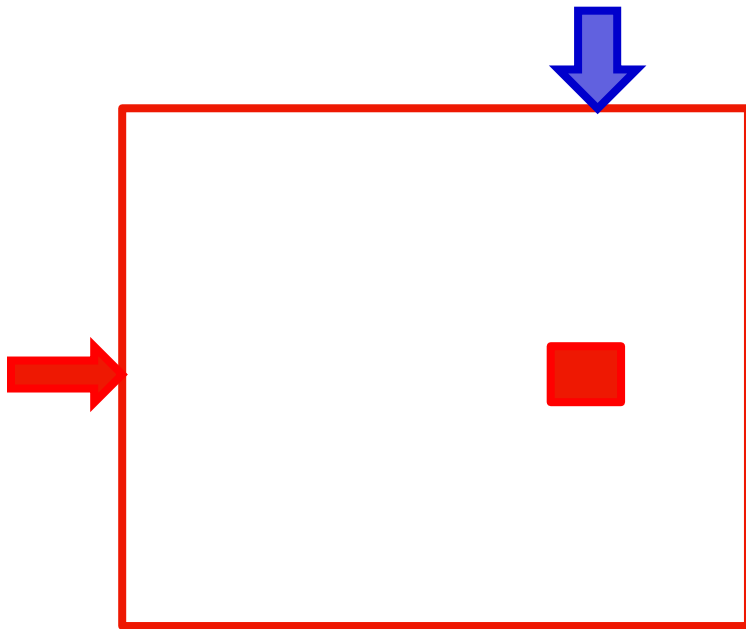
Our focus: Two-player games



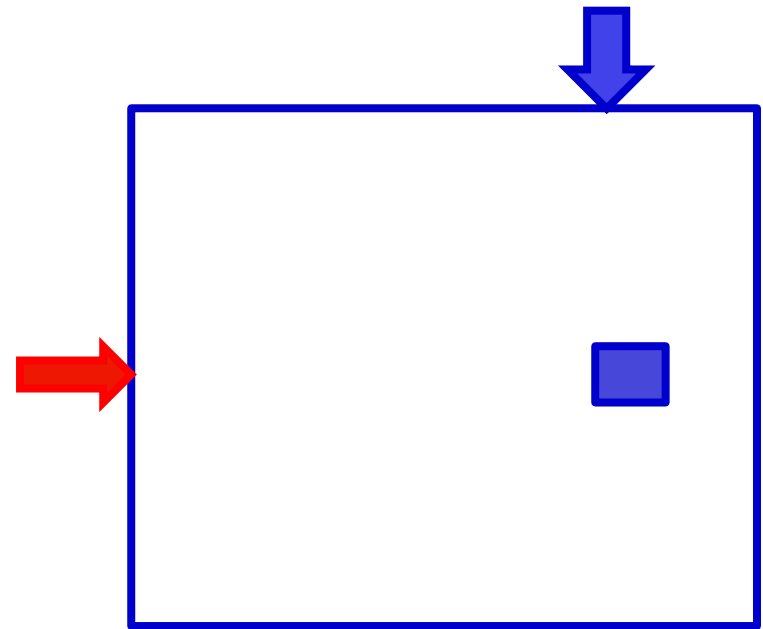
Alice
m strategies



Bob
n strategies



$A_{m \times n}$



$B_{m \times n}$

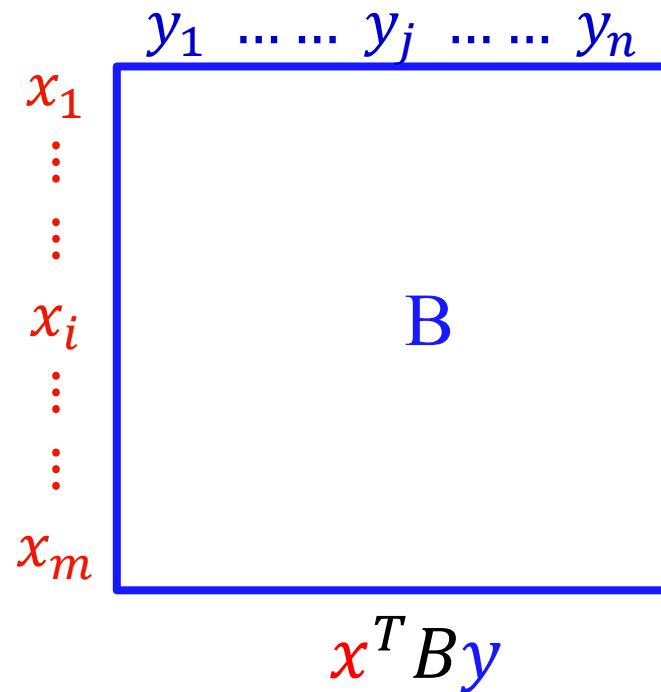
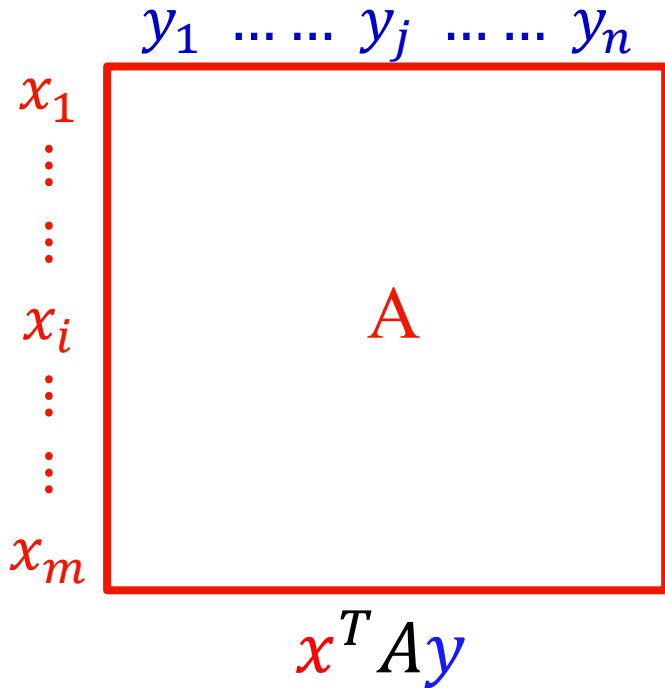


Alice



Bob

Randomize



NE: No unilateral deviation is beneficial

$$x^T A y \geq z^T A y, \quad \forall z \in \Delta_m$$

$$x^T B y \geq x^T B z, \quad \forall z \in \Delta_n$$

Example

	R	P	S
R	0 0	-1 1	1 -1
P	1 -1	0 0	-1 1
S	-1 1	1 -1	0 0

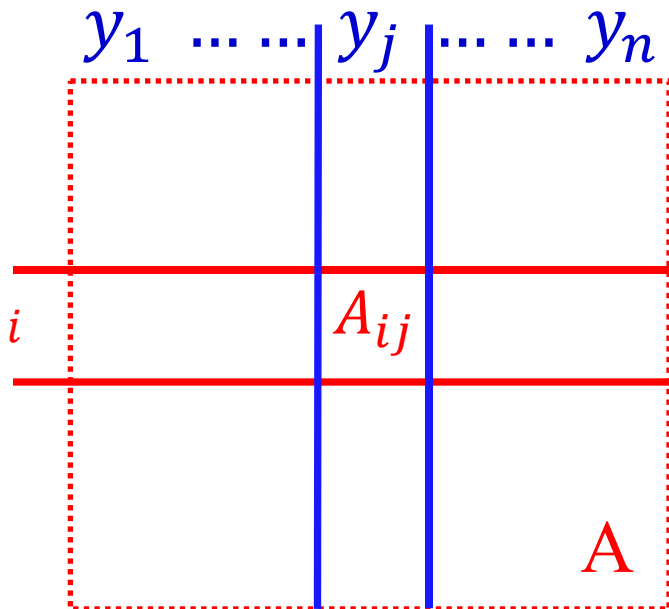


2-Nash Characterization

2-Nash Characterization



- For **Row**, i^{th} strategy gives

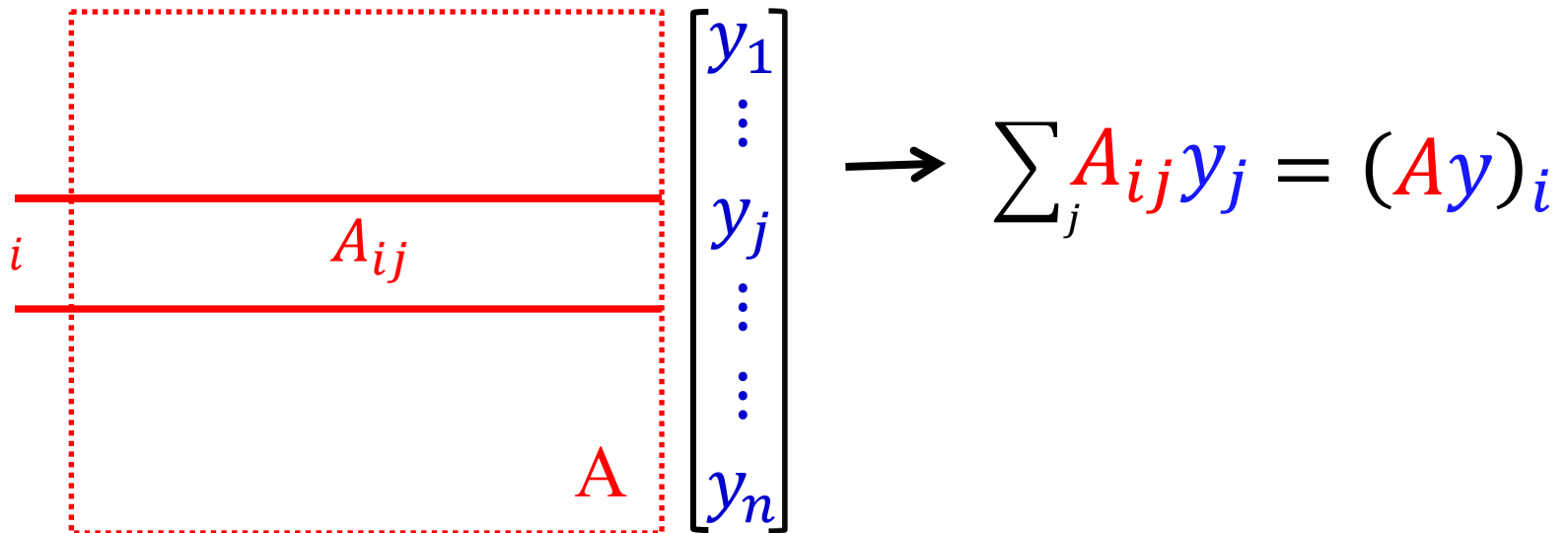


$$\longrightarrow \sum_j A_{ij} y_j$$

2-Nash Characterization



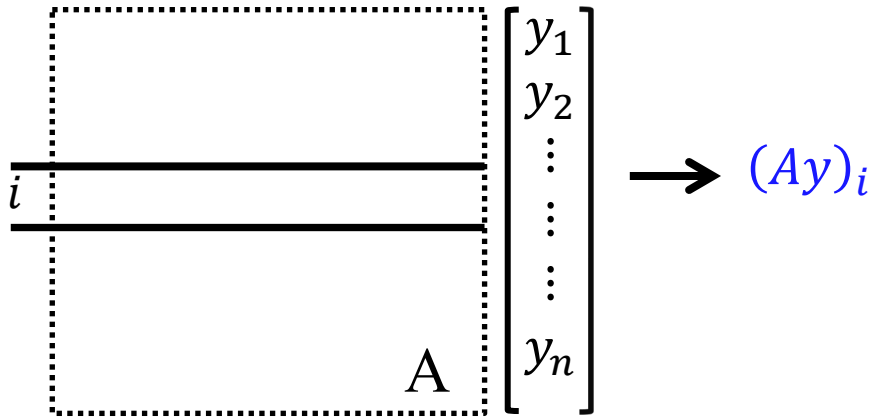
- For Row, i^{th} strategy gives



The diagram illustrates the calculation of the i^{th} element of the product Ay . A matrix A is shown with a red dotted box around its i^{th} row. The element A_{ij} is labeled in the i^{th} row and j^{th} column. To the right, a column vector y is shown with elements $y_1, \dots, y_j, \dots, y_n$. An arrow points to the equation:

$$\sum_j A_{ij} y_j = (Ay)_i$$

- i^{th} strategy gives Alice



- Max possible payoff: $\max_i e_i A y$

- x achieves max payoff iff

$$\forall i, \quad x^T A y \geq (A y)_i$$

\equiv

$$\forall k, \quad x_k > 0 \Rightarrow (A y)_k = \max_i (A y)_i$$

Complementarity

Polyhedra



max-payoff $\leq \pi_A$

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

max-payoff $\leq \pi_B$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$



P

$$\forall i, (Ay)_i \leq \pi_A$$
$$y \in \Delta_n$$



Q

$$\forall j, (x^T B)_j \leq \pi_B$$
$$x \in \Delta_m$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Sum of payoffs

At least the sum of
max payoffs

$$\underbrace{x^T (A + B) y}_{\text{Sum of payoffs}} - \underbrace{(\pi_A + \pi_B)}_{\text{At least the sum of max payoffs}} \leq 0$$

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Sum of payoffs

At least the sum of
max payoffs

$$x^T (A + B)y - (\pi_A + \pi_B) = 0$$

Complementarity

1. (x, y) is a NE
2. π_A and π_B are the max payoffs

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

Claim. For $(y, \pi_A, x, \pi_B) \in P \times Q$, (i) $x^T(A + B)y - (\pi_A + \pi_B) \leq 0$.
(ii) $x^T(A + B)y - (\pi_A + \pi_B) = 0$ if and only if (x, y) is a NE.

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Sum of payoffs
2-Nash

At least the sum of
max payoffs

$$\max: \underbrace{x^T (A + B)y}_{\text{Sum of payoffs}} - \underbrace{(\pi_A + \pi_B)}_{\text{At least the sum of max payoffs}} = 0$$

$$\text{s.t. } (y, \pi_A, x, \pi_B) \in P \boxtimes Q$$

Complementarity

1. (x, y) is a NE
2. π_A and π_B are the max payoffs

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Theorem. If (A, B) is zero-sum, i.e., $A + B = 0$, then
2-Nash \rightarrow linear programming

$$\max: x^T (A + B)y - (\pi_A + \pi_B)$$

$$\text{s.t. } (y, \pi_A, x, \pi_B) \in P \times Q$$

$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Theorem. If (A, B) is zero-sum, i.e., $A + B = 0$, then
2-Nash \rightarrow linear programming

$$\begin{array}{l} \text{max: } -(\pi_A + \pi_B) \\ \text{s.t. } (y, \pi_A, x, \pi_B) \in P \times Q \end{array}$$

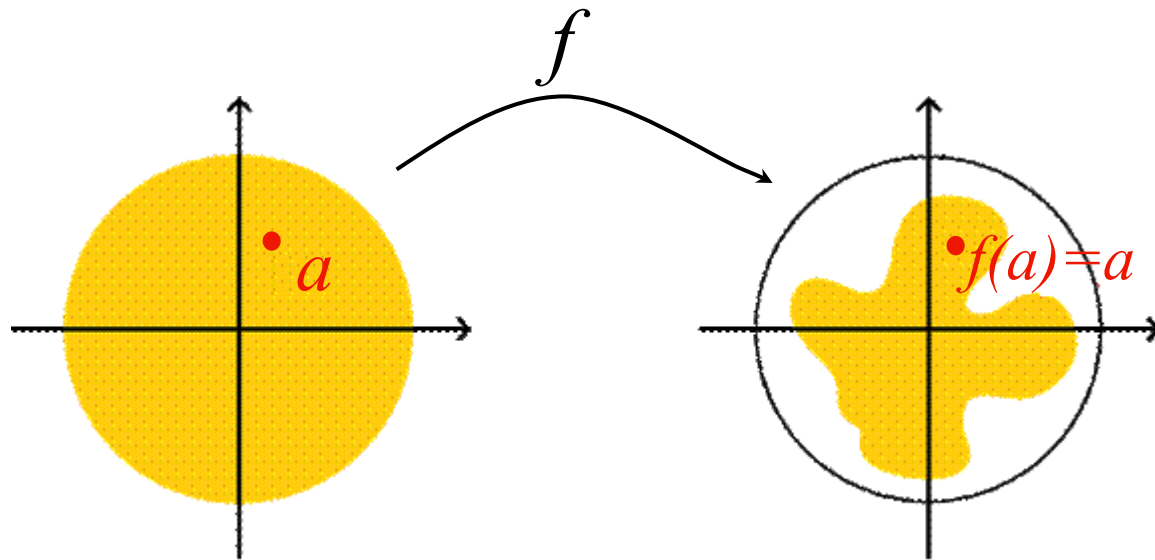
Theorem. [von Neumann'28] (max-min = min-max)

Game $(A, -A) \equiv$ wrt A , Alice is a maximizer and Bob minimizer

$$\max_x \min_y x^T A y = \min_y \max_x x^T A y \quad \& \text{ the max-min is NE.}$$

Computation in general?

NE existence via fixed-point theorem.



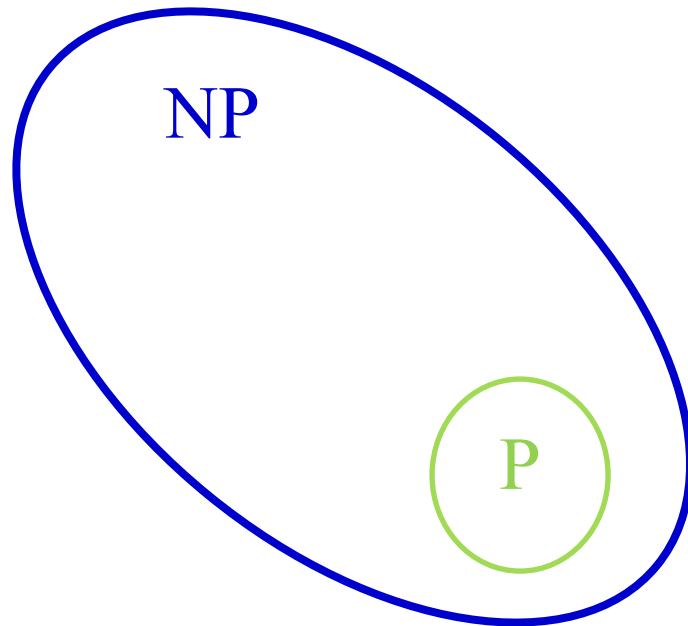
Computation? (in Econ)

- Special cases: Dantzig'51, Lemke-Howson'64, Elzen-Talman'88, Govindan-Wilson'03, ...
- Scarf'67: Approximate fixed-point.
 - Numerical instability
 - Not efficient!
- ...

Computation? (in CS)

Not easy!

\exists solution?



What if solution always exists, like Nash Eq.?

Computation? (in CS)

Megiddo and Papadimitriou'91 :

Nash is NP-hard \Rightarrow NP=Co-NP

NP-hardness is ruled out!

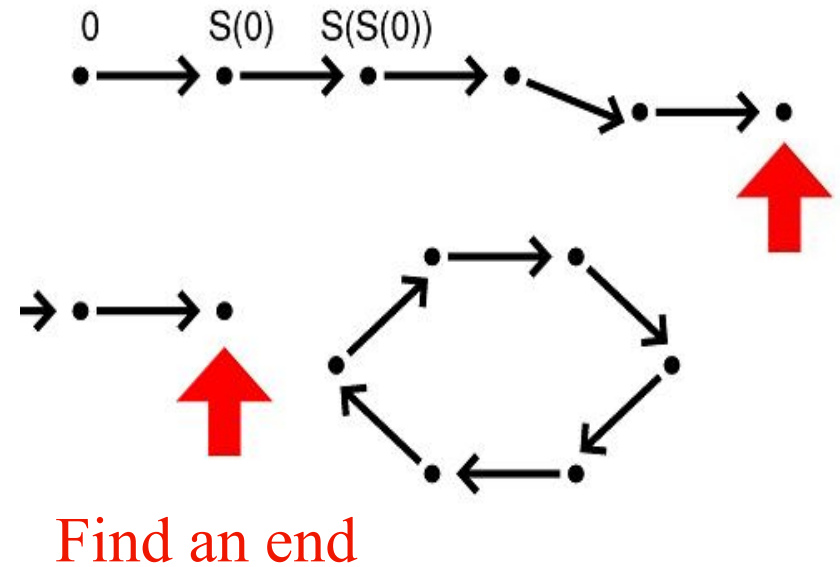
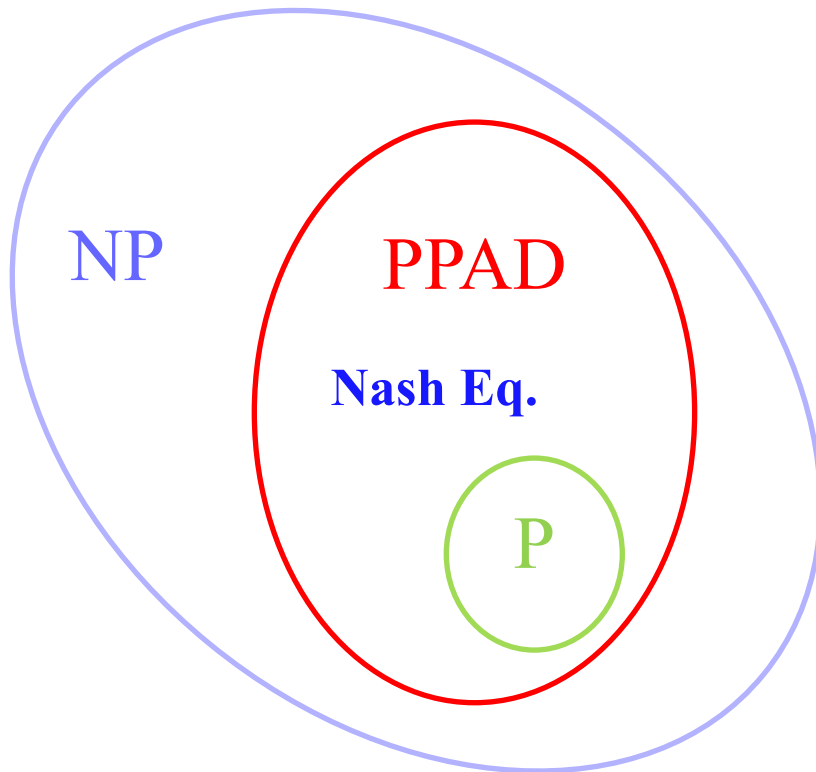
Complexity Classes

2-Nash is PPAD-complete!

[DGP'06, CDT'06]

Papadimitriou'94

PPAD Polynomial Parity Argument for Directed graph



$$P \quad \begin{array}{l} \forall i, (Ay)_i \leq \pi_A \\ y \in \Delta_n \end{array}$$

$$Q \quad \begin{array}{l} \forall j, (x^T B)_j \leq \pi_B \\ x \in \Delta_m \end{array}$$

$$(y, \pi_A, x, \pi_B) \in P \times Q$$

Theorem. If (A, B) is zero-sum, i.e., $A + B = 0$, then
2-Nash \rightarrow linear programming

$$\text{max: } -(\pi_A + \pi_B)$$

$$\text{s.t. } (y, \pi_A, x, \pi_B) \in P \times Q$$

Rank of a game: rank(A+B)

Zero-sum \equiv Rank-0 games

Rank 1 Game

$$A + B = u \cdot v^T$$

\downarrow
 $\in R^m$

\rightarrow
 $\in R^n$

Bilinear

2-Nash

$$\max: x^T (A + B) y - (\pi_A + \pi_B)$$

$$\text{s.t. } (y, \pi_A, x, \pi_B) \in P \times Q$$

Rank 1 Game [AGM.S'11]

$$A + B = u \cdot v^T$$

Product of two
linear terms

2-Nash

$$\max: \boxed{(x^T u)(v^T y)} - (\pi_A + \pi_B)$$

$$\text{s.t. } (y, \pi_A, x, \pi_B) \in P \times Q$$

Rank 1 QP is NP-hard in general

Rank 1 Game [AGM.S'11]

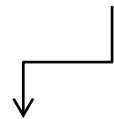
$$A + B = u \cdot v^T \longrightarrow (A, u, v)$$

2-Nash

$$\max: (x^T u)(v^T y) - (\pi_A + \pi_B)$$

s.t.

$$P \times Q$$



$$(x^T u)v^T - x^T A$$

Think Big!

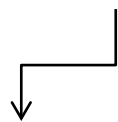
Consider game space $S = (A, *, v)$

2-Nash

$$\max: (x^T *)(v^T y) - (\pi_A + \pi_B)$$

s.t.

$$P \times Q$$



$$(x^T *)(v^T) - x^T A$$

Think Big!

Consider game space $S = (A, *, v)$ All NE of S

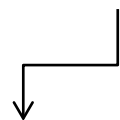
Complementarity

Captures

LP(λ)

$$\max: \lambda(v^T y) - (\pi_A + \pi_B)$$

$$\text{s.t.} \quad P \times Q$$



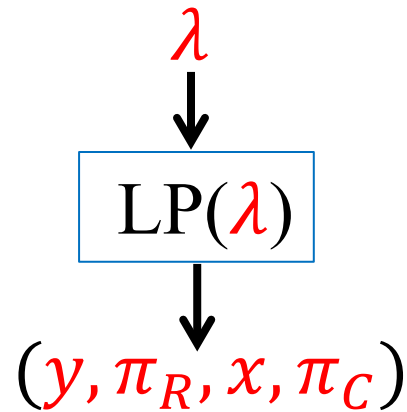
$$\lambda v^T - x^T A$$

Solutions of LP(λ)

$$\forall \lambda \in \mathbb{R}$$

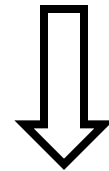
Claim. For any $\lambda \in \mathbb{R}$, optimal value of $\text{LP}(\lambda)$ is zero.

$$\begin{array}{ll} \max: & \lambda(v^T y) - (\pi_A + \pi_B) \\ \text{s.t.} & P \times Q \\ & \downarrow \\ \text{LP}(\lambda) & \lambda v^T - x^T A \end{array}$$



Goal: NE of $(R, \mathbf{u}, \mathbf{v})$

$(m-1)$ -dimensional
space in \mathcal{S}



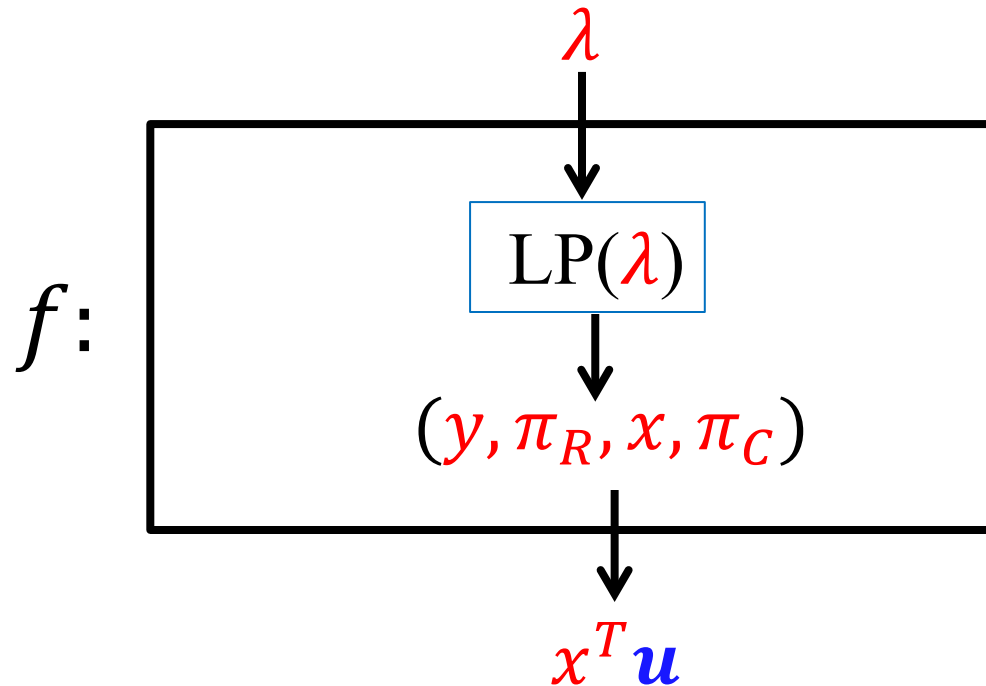
If \mathbf{u} one of them
then done!

Claim:

$$\forall \mathbf{c} \text{ s.t. } \mathbf{x}^T \mathbf{c} = \lambda,$$

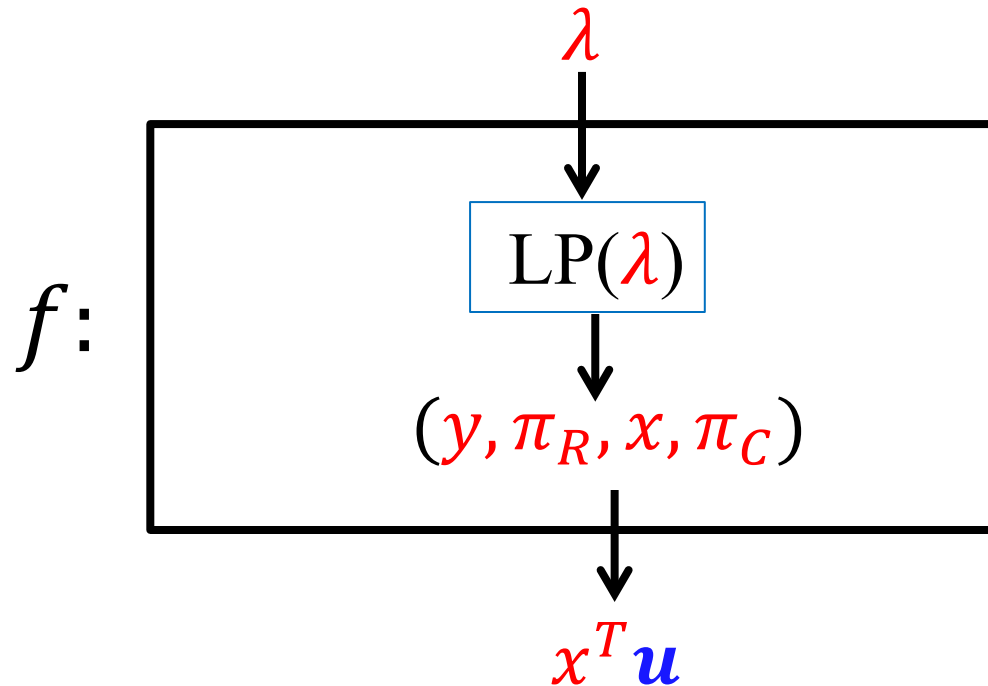
(\mathbf{x}, \mathbf{y}) is a NE of game $(R, \mathbf{c}, \mathbf{v})$

Goal: NE of game $(R, \mathbf{u}, \mathbf{v})$



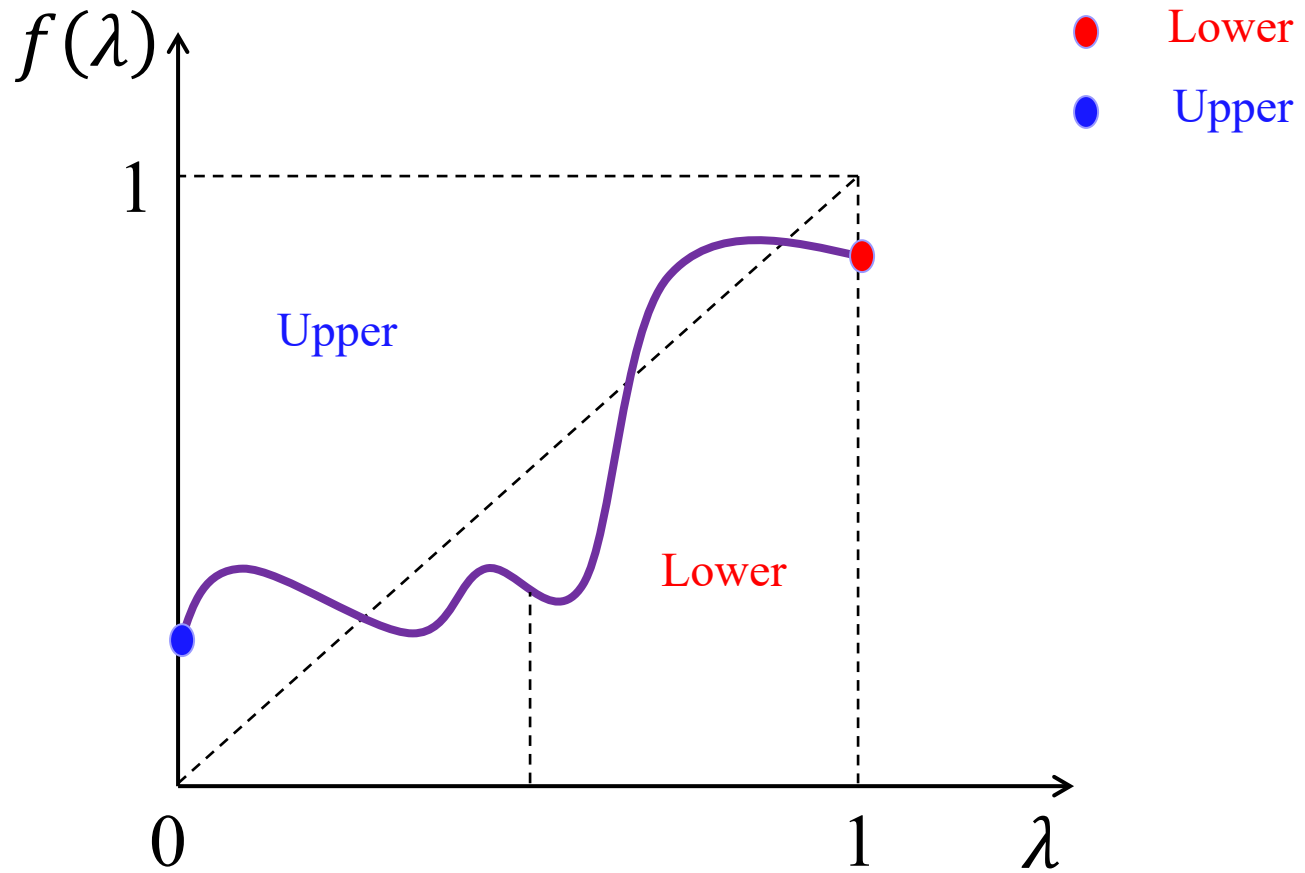
If $x^T \mathbf{u} = \lambda$ then done!

Goal: NE of game $(R, \mathbf{u}, \mathbf{v})$

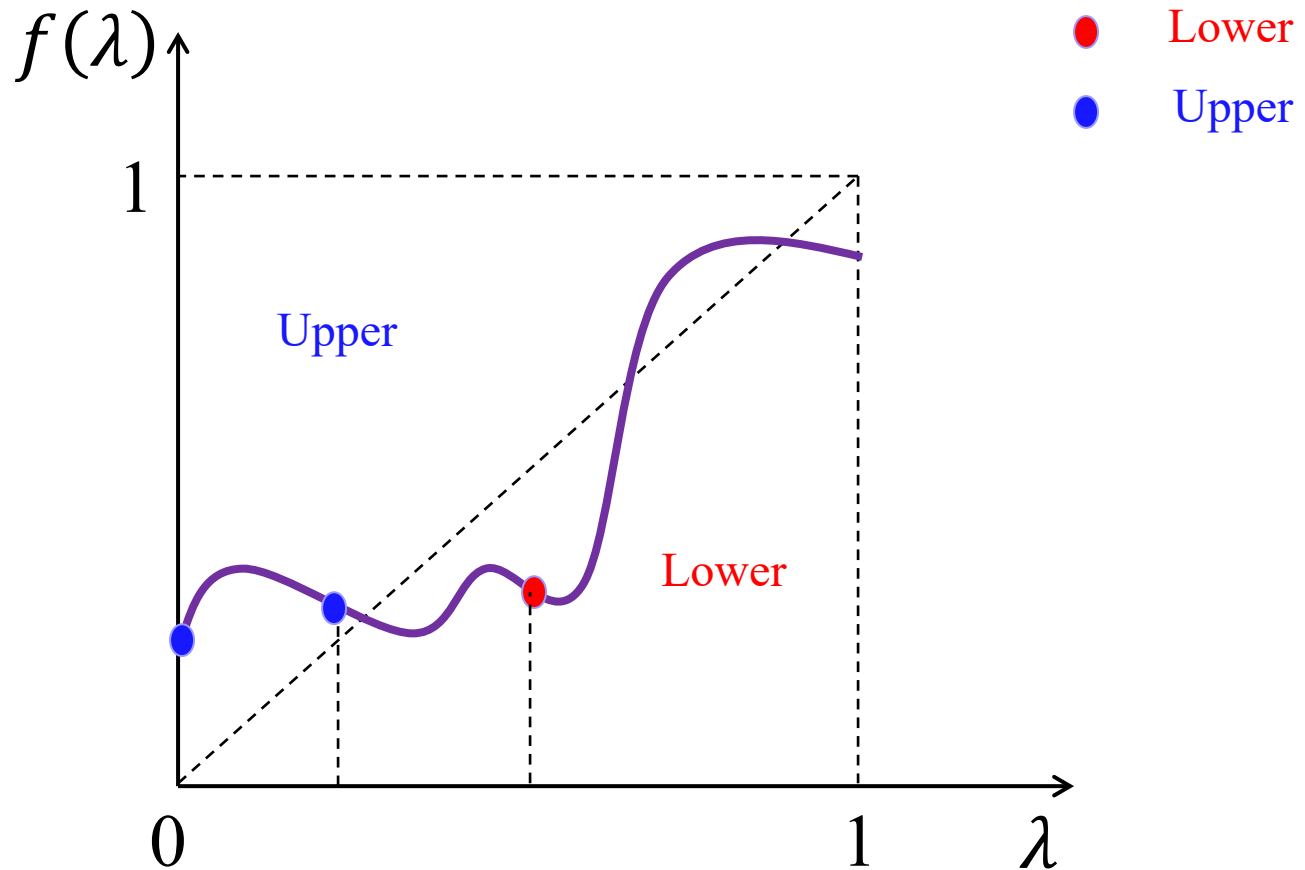


NE of game $(R, \mathbf{u}, \mathbf{v}) \Leftrightarrow \lambda \leftarrow$ Fixed points of f

1-D Fixed Point



1-D Fixed Point



And so on until the difference becomes small enough



What about rank-2 or more?

Rank-0 (zero-sum)
games $\text{rank}(A+B)=0$

→
Von Neumann
(1928)

Rank-1 games
 $\text{rank}(A+B)=1$

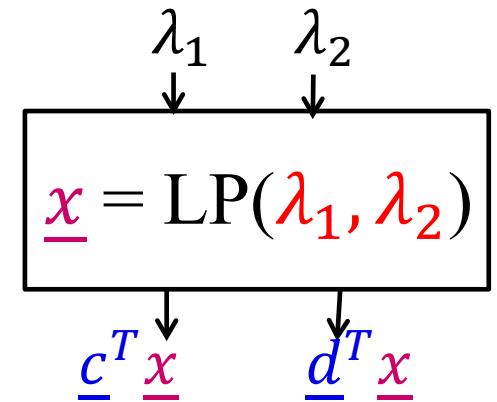
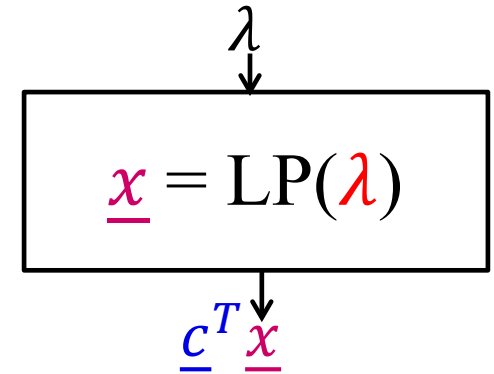
→

Rank-2 games
 $\text{rank}(A+B)=2$

→

⋮

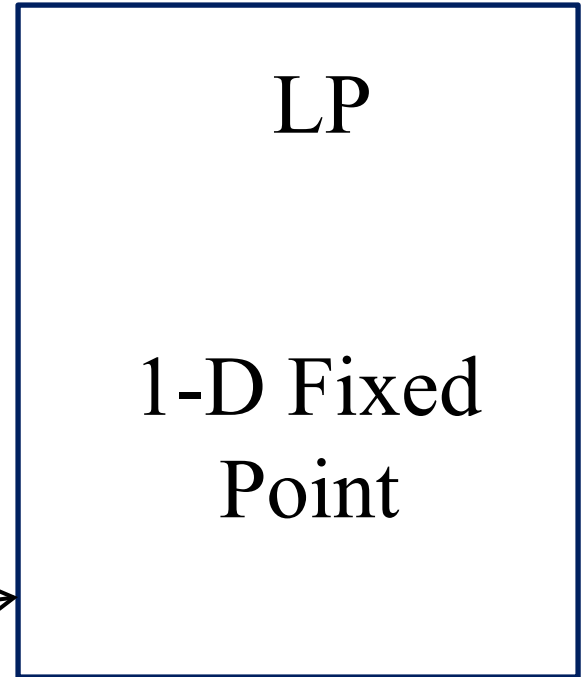
LP



⋮

Rank-0 (zero-sum)
games

→
Von Neumann
(1928)



Rank-1 games

→

In P →

Rank-2 games

→

⋮

2-D Fixed
Point

Rank-0 (zero-sum)
games

→
Von Neumann
(1928)

LP

Rank-1 games

→

1-D Fixed
Point

**PPAD-hard
in general**

→

Rank-2 games

→

2-D Fixed
Point

⋮

⋮

Rank-0 (zero-sum)
games

→
Von Neumann
(1928)

LP

Rank-1 games

→

1-D Fixed
Point

**PPAD-hard
in general**

→

Rank-2 games

←
?

2-D Fixed
Point

⋮

⋮

Rank-0 (zero-sum)
games

→
Von Neumann
(1928)

LP

Rank-1 games

→

1-D Fixed
Point

**PPAD-hard
in general**



Rank-2 games

←
[M.'14, COPY'16]

2-D Fixed
Point

⋮

⋮

⋮