

Today

- Zero-Sum Games

Strategic Games.

N players.

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Example:

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Example:

2 players

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Example:

2 players

Player 1: { **D**efect, **C**ooperate }.

Player 2: { **D**efect, **C**ooperate }.

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Example:

2 players

Player 1: $\{ \mathbf{D}$ efect, \mathbf{C} ooperate $\}$.

Player 2: $\{ \mathbf{D}$ efect, \mathbf{C} ooperate $\}$.

Payoff:

Strategic Games.

N players.

Each player has strategy set. $\{S_1, \dots, S_N\}$.

Vector valued payoff function: $u(s_1, \dots, s_n)$ (e.g., $\in \mathbb{R}^N$).

Example:

2 players

Player 1: { **D**efect, **C**ooperate }.

Player 2: { **D**efect, **C**ooperate }.

Payoff:

	C	D
C	(3,3)	(0,5)
D	(5,0)	(1,1)

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Defects! Payoff (5,0)

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Defects! Payoff (5,0)

What does player 2 do now?

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Defects! Payoff (5,0)

What does player 2 do now?

Defects! Payoff (.1,.1).

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Defects! Payoff (5,0)

What does player 2 do now?

Defects! Payoff (.1,.1).

Stable now!

Famous because?

	C	D
C	(3,3)	(0,5)
D	(5,0)	(.1,.1)

What is the best thing for the players to do?

Both cooperate. Payoff (3,3).

If player 1 wants to do better, what does he do?

Defects! Payoff (5,0)

What does player 2 do now?

Defects! Payoff (.1,.1).

Stable now!

Nash Equilibrium: neither player has incentive to change strategy.

Two Person Zero Sum Games

2 players.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ?

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no. (P, R) ?

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no. (P, R) ? no.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no. (P, R) ? no. (S, R) ?

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

"Player 1 pays a to player 2."

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no. (P, R) ? no. (S, R) ? no.

Two Person Zero Sum Games

2 players.

Each player has strategy set:

m strategies for player 1 n strategies for player 2

Payoff function: $u(i, j) = (-a, a)$ (or just a).

“Player 1 pays a to player 2.”

Zero Sum: Payoff for any pair of strategies sums to 0.

Payoffs by m by n matrix: G .

Row player maximizes, column player minimizes.

Roshambo: rock, paper, scissors.

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

Any Nash Equilibrium?

(R, R) ? no. (P, R) ? no. (S, R) ? no.

Mixed Strategies.

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

How do you play?

Mixed Strategies.

		R	P	S
R	$\frac{.33}{3}$	0	-1	1
P	$\frac{.33}{3}$	1	0	-1
S	$\frac{.33}{3}$	-1	1	0

How do you play?

Player 1: play each strategy with equal probability.

Mixed Strategies.

		R	P	S
		$\frac{.33}{.33}$	$\frac{.33}{.33}$	$\frac{.33}{.33}$
R	$\frac{.33}{.33}$	0	-1	1
P	$\frac{.33}{.33}$	1	0	-1
S	$\frac{.33}{.33}$	-1	1	0

How do you play?

Player 1: play each strategy with equal probability.

Player 2: play each strategy with equal probability.

Mixed Strategies.

		R	P	S
		$\frac{.33}{.33}$	$\frac{.33}{.33}$	$\frac{.33}{.33}$
R	$\frac{.33}{.33}$	0	-1	1
P	$\frac{.33}{.33}$	1	0	-1
S	$\frac{.33}{.33}$	-1	1	0

How do you play?

Player 1: play each strategy with equal probability.

Player 2: play each strategy with equal probability.

Mixed Strategies.

		R	P	S
		$\frac{.33}{3}$	$\frac{.33}{3}$	$\frac{.33}{3}$
R	$\frac{.33}{3}$	0	-1	1
P	$\frac{.33}{3}$	1	0	-1
S	$\frac{.33}{3}$	-1	1	0

How do you play?

Player 1: play each strategy with equal probability.

Player 2: play each strategy with equal probability.

Definitions.

Mixed strategies: Each players plays distribution over strategies.

Mixed Strategies.

		R	P	S
		$.3\bar{3}$	$.3\bar{3}$	$.3\bar{3}$
R	$.3\bar{3}$	0	-1	1
P	$.3\bar{3}$	1	0	-1
S	$.3\bar{3}$	-1	1	0

How do you play?

Player 1: play each strategy with equal probability.

Player 2: play each strategy with equal probability.

Definitions.

Mixed strategies: Each players plays distribution over strategies.

Pure strategies: Each player plays single strategy.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{3}$	$\frac{.33}{3}$	$\frac{.33}{3}$
R	$\frac{.33}{3}$	0	-1	1
P	$\frac{.33}{3}$	1	0	1
S	$\frac{.33}{3}$	-1	1	0

Payoffs?

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{}$	$\frac{.33}{}$	$\frac{.33}{}$
R	$\frac{.33}{}$	0	-1	1
P	$\frac{.33}{}$	1	0	1
S	$\frac{.33}{}$	-1	1	0

Payoffs? Can't just look it up in matrix!

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{3}$	$\frac{.33}{3}$	$\frac{.33}{3}$
R	$\frac{.33}{3}$	0	-1	1
P	$\frac{.33}{3}$	1	0	1
S	$\frac{.33}{3}$	-1	1	0

Payoffs? Can't just look it up in matrix!

Average Payoff.

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{3}$	$\frac{.33}{3}$	$\frac{.33}{3}$
R	$\frac{.33}{3}$	0	-1	1
P	$\frac{.33}{3}$	1	0	1
S	$\frac{.33}{3}$	-1	1	0

Payoffs? Can't just look it up in matrix!.

Average Payoff. Expected Payoff.

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{}$	$\frac{.33}{}$	$\frac{.33}{}$
R	$\frac{.33}{}$	0	-1	1
P	$\frac{.33}{}$	1	0	1
S	$\frac{.33}{}$	-1	1	0

Payoffs? Can't just look it up in matrix!.

Average Payoff. **Expected Payoff.**

Sample space: $\Omega = \{(i,j) : i,j \in [1, \dots, 3]\}$

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{}$	$\frac{.33}{}$	$\frac{.33}{}$
R	$\frac{.33}{}$	0	-1	1
P	$\frac{.33}{}$	1	0	1
S	$\frac{.33}{}$	-1	1	0

Payoffs? Can't just look it up in matrix!.

Average Payoff. **Expected Payoff.**

Sample space: $\Omega = \{(i, j) : i, j \in [1, \dots, 3]\}$

Random variable X (payoff).

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		.33	.33	.33
R	.33	0	-1	1
P	.33	1	0	1
S	.33	-1	1	0

Payoffs? Can't just look it up in matrix!

Average Payoff. Expected Payoff.

Sample space: $\Omega = \{(i, j) : i, j \in [1, \dots, 3]\}$

Random variable X (payoff).

$$E[X] = \sum_{(i,j)} X(i,j) Pr[(i,j)].$$

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{}$	$\frac{.33}{}$	$\frac{.33}{}$
R	$\frac{.33}{}$	0	-1	1
P	$\frac{.33}{}$	1	0	1
S	$\frac{.33}{}$	-1	1	0

Payoffs? Can't just look it up in matrix!

Average Payoff. **Expected Payoff.**

Sample space: $\Omega = \{(i,j) : i,j \in [1, \dots, 3]\}$

Random variable X (payoff).

$$E[X] = \sum_{(i,j)} X(i,j) Pr[(i,j)].$$

Each player chooses independently:

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		$\frac{.33}{}$	$\frac{.33}{}$	$\frac{.33}{}$
R	$\frac{.33}{}$	0	-1	1
P	$\frac{.33}{}$	1	0	1
S	$\frac{.33}{}$	-1	1	0

Payoffs? Can't just look it up in matrix!

Average Payoff. **Expected Payoff.**

Sample space: $\Omega = \{(i, j) : i, j \in [1, \dots, 3]\}$

Random variable X (payoff).

$$E[X] = \sum_{(i,j)} X(i,j) Pr[(i,j)].$$

Each player chooses independently:

$$Pr[(i,j)] = \frac{1}{3} \times \frac{1}{3} = \frac{1}{9}.$$

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		.33	.33	.33
R	.33	0	-1	1
P	.33	1	0	1
S	.33	-1	1	0

Payoffs? Can't just look it up in matrix!.

Average Payoff. Expected Payoff.

Sample space: $\Omega = \{(i, j) : i, j \in [1, \dots, 3]\}$

Random variable X (payoff).

$$E[X] = \sum_{(i,j)} X(i,j) Pr[(i,j)].$$

Each player chooses independently:

$$Pr[(i,j)] = \frac{1}{3} \times \frac{1}{3} = \frac{1}{9}.$$

$$E[X] = 0.$$

¹Remember zero sum games have one payoff.

Payoffs: Equilibrium.

		R	P	S
		.33	.33	.33
R	.33	0	-1	1
P	.33	1	0	1
S	.33	-1	1	0

Payoffs? Can't just look it up in matrix!

Average Payoff. Expected Payoff.

Sample space: $\Omega = \{(i, j) : i, j \in [1, \dots, 3]\}$

Random variable X (payoff).

$$E[X] = \sum_{(i,j)} X(i,j) Pr[(i,j)].$$

Each player chooses independently:

$$Pr[(i,j)] = \frac{1}{3} \times \frac{1}{3} = \frac{1}{9}.$$

$$E[X] = 0.^1$$

¹Remember zero sum games have one payoff.

Equilibrium

		R	P	S
		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
R	$\frac{1}{3}$	0	1	-1
P	$\frac{1}{3}$	-1	0	1
S	$\frac{1}{3}$	1	-1	0

Will Player 1 change strategy?

Equilibrium

		R	P	S
		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
R	$\frac{1}{3}$	0	1	-1
P	$\frac{1}{3}$	-1	0	1
S	$\frac{1}{3}$	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Equilibrium

		R	P	S
		$\frac{.33}{.33}$	$\frac{.33}{.33}$	$\frac{.33}{.33}$
R	$\frac{.33}{.33}$	0	1	-1
P	$\frac{.33}{.33}$	-1	0	1
S	$\frac{.33}{.33}$	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock?

Equilibrium

		R	P	S
		$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
R	$\frac{1}{3}$	0	1	-1
P	$\frac{1}{3}$	-1	0	1
S	$\frac{1}{3}$	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper?

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors?

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j)$$

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is **weighted av.** of **payoffs of pure strats.**

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Mixed strategy can't be better than the best pure strategy.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Mixed strategy can't be better than the best pure strategy.

Player 1 has no incentive to change!

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Mixed strategy can't be better than the best pure strategy.

Player 1 has no incentive to change! Same for player 2.

Equilibrium

		R	P	S
		.33	.33	.33
R	.33	0	1	-1
P	.33	-1	0	1
S	.33	1	-1	0

Will Player 1 change strategy? Mixed strategies uncountable!

Expected payoffs for pure strategies for player 1.

Expected payoff of Rock? $\frac{1}{3} \times 0 + \frac{1}{3} \times 1 + \frac{1}{3} \times -1 = 0$.

Expected payoff of Paper? $\frac{1}{3} \times -1 + \frac{1}{3} \times 0 + \frac{1}{3} \times 1 = 0$.

Expected payoff of Scissors? $\frac{1}{3} \times 1 + \frac{1}{3} \times 0 + \frac{1}{3} \times -1 = 0$.

No better pure strategy. \implies No better mixed strategy!

Mixed strat. payoff is weighted av. of payoffs of pure strats.

$$E[X] = \sum_{(i,j)} (Pr[i] \times Pr[j]) X(i,j) = \sum_i Pr[i] (\sum_j Pr[j] \times X(i,j))$$

Mixed strategy can't be better than the best pure strategy.

Player 1 has no incentive to change! Same for player 2.

Equilibrium!

Another example plus notation.

Rock, Paper, Scissors, prEempt.

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium?

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium? **(E,E)**.

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium? **(E,E)**. Pure strategy equilibrium.

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium? **(E,E)**. Pure strategy equilibrium.

Notation:

Another example plus notation.

Rock, Paper, Scissors, prEempt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium? **(E,E)**. Pure strategy equilibrium.

Notation: Rock is 1, Paper is 2, Scissors is 3, prEmpt is 4.

Another example plus notation.

Rock, Paper, Scissors, prEmpt.

PreEmpt ties preEmpt, beats everything else.

Payoffs.

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

Equilibrium? **(E,E)**. Pure strategy equilibrium.

Notation: Rock is 1, Paper is 2, Scissors is 3, prEmpt is 4.

Payoff Matrix.

$$G = \begin{bmatrix} 0 & -1 & 1 & -1 \\ 1 & 0 & -1 & -1 \\ -1 & 1 & 0 & -1 \\ 1 & 1 & 1 & 0 \end{bmatrix}$$

Playing the boss...

Row has extra strategy: Cheat.

Playing the boss...

Row has extra strategy: Cheat.

Ties with rock and scissors, beats paper. (Scissors, or no rock!)

Playing the boss...

Row has extra strategy: Cheat.

Ties with rock and scissors, beats paper. (Scissors, or no rock!)

Payoff matrix:

Rock is strategy 1, Paper is 2, Scissors is 3, and Cheat is 4 (for row.)

Playing the boss...

Row has extra strategy: Cheat.

Ties with rock and scissors, beats paper. (Scissors, or no rock!)

Payoff matrix:

Rock is strategy 1, Paper is 2, Scissors is 3, and Cheat is 4 (for row.)

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Note: column knows row cheats.

Playing the boss...

Row has extra strategy: Cheat.

Ties with rock and scissors, beats paper. (Scissors, or no rock!)

Payoff matrix:

Rock is strategy 1, Paper is 2, Scissors is 3, and Cheat is 4 (for row.)

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Note: column knows row cheats. Why play?

Row is column's boss.

Playing the boss...

Row has extra strategy: Cheat.

Ties with rock and scissors, beats paper. (Scissors, or no rock!)

Payoff matrix:

Rock is strategy 1, Paper is 2, Scissors is 3, and Cheat is 4 (for row.)

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Note: column knows row cheats. Why play?

Row is column's boss.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff?

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

Strategy 1: $\frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

Strategy 1: $\frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$

Strategy 2: $\frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

Strategy 1: $\frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$

Strategy 2: $\frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

$$\text{Payoff is } 0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6})$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

$$\text{Payoff is } 0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6}) = -\frac{1}{6}$$

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

$$\text{Payoff is } 0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6}) = -\frac{1}{6}$$

Column player: every column payoff is $-\frac{1}{6}$.

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

$$\text{Payoff is } 0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6}) = -\frac{1}{6}$$

Column player: every column payoff is $-\frac{1}{6}$.

Both only play optimal strategies!

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

$$\text{Strategy 1: } \frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$$

$$\text{Strategy 2: } \frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$$

$$\text{Strategy 3: } \frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$$

$$\text{Strategy 4: } \frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$$

$$\text{Payoff is } 0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6}) = -\frac{1}{6}$$

Column player: every column payoff is $-\frac{1}{6}$.

Both only play optimal strategies!

Why not play just one?

Equilibrium: play the boss...

$$G = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Equilibrium:

Row: $(0, \frac{1}{3}, \frac{1}{6}, \frac{1}{2})$. Column: $(\frac{1}{3}, \frac{1}{2}, \frac{1}{6})$.

Payoff? Remember: weighted average of pure strategies.

Row Player.

Strategy 1: $\frac{1}{3} \times 0 + \frac{1}{2} \times 1 + \frac{1}{6} \times -1 = \frac{1}{3}$

Strategy 2: $\frac{1}{3} \times -1 + \frac{1}{2} \times 0 + \frac{1}{6} \times 1 = -\frac{1}{6}$

Strategy 3: $\frac{1}{3} \times 1 + \frac{1}{2} \times -1 + \frac{1}{6} \times 0 = -\frac{1}{6}$

Strategy 4: $\frac{1}{3} \times 0 + \frac{1}{2} \times 0 + \frac{1}{6} \times -1 = -\frac{1}{6}$

Payoff is $0 \times \frac{1}{3} + \frac{1}{3} \times (-\frac{1}{6}) + \frac{1}{6} \times (-\frac{1}{6}) + \frac{1}{2} \times (-\frac{1}{6}) = -\frac{1}{6}$

Column player: every column payoff is $-\frac{1}{6}$.

Both only play optimal strategies!

Why not play just one? Changes payoff for other guy!

Two person zero sum games.

$m \times n$ payoff matrix G .

Two person zero sum games.

$m \times n$ payoff matrix G .

Row mixed strategy: $x = (x_1, \dots, x_m)$.

Two person zero sum games.

$m \times n$ payoff matrix G .

Row mixed strategy: $x = (x_1, \dots, x_m)$.

Column mixed strategy: $y = (y_1, \dots, y_n)$.

Two person zero sum games.

$m \times n$ payoff matrix G .

Row mixed strategy: $x = (x_1, \dots, x_m)$.

Column mixed strategy: $y = (y_1, \dots, y_n)$.

Payoff for strategy pair (x, y) :

Two person zero sum games.

$m \times n$ payoff matrix G .

Row mixed strategy: $x = (x_1, \dots, x_m)$.

Column mixed strategy: $y = (y_1, \dots, y_n)$.

Payoff for strategy pair (x, y) :

$$p(x, y) = x^t G y$$

That is,

$$\sum_i x_i \left(\sum_j G[i, j] y_j \right) = \sum_j \left(\sum_i x_i G[i, j] \right) y_j.$$

Two person zero sum games.

$m \times n$ payoff matrix G .

Row mixed strategy: $x = (x_1, \dots, x_m)$.

Column mixed strategy: $y = (y_1, \dots, y_n)$.

Payoff for strategy pair (x, y) :

$$p(x, y) = x^t G y$$

That is,

$$\sum_i x_i \left(\sum_j G[i, j] y_j \right) = \sum_j \left(\sum_i x_i G[i, j] \right) y_j.$$

Recall row maximizes, column minimizes.

Best Response/Defense.

Row goes first:

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Note: y can be $(0, 0, \dots, 1, \dots 0)$.

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Note: y can be $(0, 0, \dots, 1, \dots 0)$.

Example: Roshambo.

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Note: y can be $(0, 0, \dots, 1, \dots 0)$.

Example: Roshambo. Value of R ?

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Note: y can be $(0, 0, \dots, 1, \dots 0)$.

Example: Roshambo. Value of R ?

		R	P	S
x_1	R	0	-1	1
x_2	P	1	0	-1
x_3	S	-1	1	0

Pick x_1, x_2, x_3 to maximize

Best Response/Defense.

Row goes first:

Find x , where best column is not too low..

$$R = \min_y \max_x (x^t G y).$$

Note: y can be $(0, 0, \dots, 1, \dots 0)$.

Example: Roshambo. Value of R ?

		R	P	S
x_1	R	0	-1	1
x_2	P	1	0	-1
x_3	S	-1	1	0

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Best Defense: row

Pick x_1, x_2, x_3 to maximize

Best Defense: row

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Best Defense: row

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Linear program:

Best Defense: row

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Linear program: $z = \min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$

Best Defense: row

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Linear program: $z = \min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$

$$\max z$$

$$z \leq x_2 - x_3$$

$$z \leq -x_1 + x_3$$

$$z \leq x_1 - x_2$$

$$x_1 + x_2 = 1$$

or in standard form...

Best Defense: row

Pick x_1, x_2, x_3 to maximize

$$\min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$$

Linear program: $z = \min\{x_2 - x_3, -x_1 + x_3, x_1 - x_2\}$

$$\max z$$

$$z \leq x_2 - x_3$$

$$z \leq -x_1 + x_3$$

$$z \leq x_1 - x_2$$

$$x_1 + x_2 = 1$$

or in standard form...

$$\max z$$

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

Computing best defense: column.

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Example: Roshambo.

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Example: Roshambo. Value of C ?

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Example: Roshambo. Value of C ? 0.

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Example: Roshambo. Value of C ? 0.

	y_1	y_2	y_3
	R	P	S
R	0	-1	1
P	1	0	-1
S	-1	1	0

Computing best defense: column.

Column goes first:

Find y , where best row is not high.

$$C = \max_x \min_y (x^t G y).$$

Again: x of form $(0, 0, \dots, 1, \dots, 0)$.

Example: Roshambo. Value of C ? 0.

	y_1	y_2	y_3
	R	P	S
R	0	-1	1
P	1	0	-1
S	-1	1	0

Find y to minimize $\max\{y_2 - y_3, -y_1 + y_3, y_1 - y_2\}$.

Column Best Defense: LP.

Find y to minimize $\max\{y_2 - y_3, -y_1 + y_3, y_1 - y_2\}$.

Column Best Defense: LP.

Find y to minimize $\max\{y_2 - y_3, -y_1 + y_3, y_1 - y_2\}$.

$$\min w$$

$$y_2 - y_3 \geq w$$

$$-y_1 + y_3 \geq w$$

$$y_1 - y_2 \geq w$$

$$y_1 + y_2 = 1$$

Column Best Defense: LP.

Find y to minimize $\max\{y_2 - y_3, -y_1 + y_3, y_1 - y_2\}$.

$$\min w$$

$$y_2 - y_3 \geq w$$

$$-y_1 + y_3 \geq w$$

$$y_1 - y_2 \geq w$$

$$y_1 + y_2 = 1$$

..in standard form..

$$\min w$$

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual?

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

$$\implies z = w!$$

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

$$\implies z = w!$$

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

$$\implies z = w!$$

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Wow

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$y_2 - y_3 - w \geq 0$$

$$-y_1 + y_3 - w \geq 0$$

$$y_1 - y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

$$\implies z = w!$$

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Wow!

Both.

max z

$$x_2 - x_3 - z \geq 0$$

$$-x_1 + x_3 - z \geq 0$$

$$x_1 - x_2 - z \geq 0$$

$$x_1 + x_2 = 1$$

$$x_1, x_2 \geq 0$$

min w

$$-y_2 + y_3 - w \geq 0$$

$$y_1 - y_3 - w \geq 0$$

$$-y_1 + y_2 - w \geq 0$$

$$y_1 + y_2 = 1$$

$$y_1, y_2, y_3 \geq 0$$

Note: “column” defense lp is upper bound on “row” defense lp.

Linear programming dual? ...Yes!!!

$$\implies z = w!$$

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Wow! ! ! ! ! ! ! (von Neumann's minimax theorem.)

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Row goes first, column gets to respond.

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Row goes first, column gets to respond.

Test!

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Row goes first, column gets to respond.

Test!

Play rock first?

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Row goes first, column gets to respond.

Test!

Play rock first?

Mixed strategy.

Von Neumann's Minimax theorem.

$$\max_x \min_y x^t G y = \min_y \max_x x^t G y.$$

Column goes first, row gets to respond.

same as

Row goes first, column gets to respond.

Test!

Play rock first?

Mixed strategy. Play rock, paper, scissors uniformly.

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms:

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything:

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings!

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

We have time !

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

We have time !?

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

We have time !?

Sissa.

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

We have time !?

Sissa. Not so much!

Search Problems

Shortest paths, minimum msts, maximum matchings, maximum increasing subsequences, maximum flows.

Efficient algorithms: polynomial.

Try everything: $n!$ matchings! n^{n-2} spanning trees...

2^n or worse.

Moore: Computers doubling in speed every 18 month!

We have time !?

Sissa. Not so much!

Are there always *efficient* algorithms for optimization problems?

Perhaps not.

Satisfiability or **SAT**.

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem:

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ?

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Perhaps not.

Satisfiability or **SAT**.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT.

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Greedy

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Greedy and strongly connected components.

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Greedy and strongly connected components.

Efficient algorithm for 3-SAT?

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Greedy and strongly connected components.

Efficient algorithm for 3-SAT? Don't think so!

Perhaps not.

Satisfiability or SAT.

An Instance.

$$\phi = (x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x}) \vee (\bar{x} \vee \bar{y} \vee \bar{z})$$

Is there any way to satisfy the formula above?

Search problem: Given instance I find a solution S .

S is short and easy to check.

A *search problem* has efficient **checking** algorithm \mathcal{C} :

S is solution for I if and only if $\mathcal{C}(I, S) = \text{true}$.

For SAT, what is S ? assignment. What is \mathcal{C} ?

Checks if assignment satisfies formula.

Recall: Horn or 2-SAT. Efficient algorithms!

Greedy and strongly connected components.

Efficient algorithm for 3-SAT? Don't think so! More later.