

# Two-player zero sum game

$M(r,c)$  is the payout  
to each player

	c1	c2	c3
r1	(2,-2)	(1,-1)	(-1,1)
r2	(3,-3)	(0,0)	(-2,2)
r3	(-1,1)	(2,-2)	(1,-1)
r4	(0,0)	(0,0)	(1,-1)

Both players want to  
maximize *their* payout.

# Two-player zero sum game

$M(r,c)$  is the amount  
row player gets (from column player)

	c1	c2	c3
r1	2	1	-1
r2	3	0	-2
r3	-1	2	1
r4	0	0	1

Row player trying to maximize  
Column player trying to minimize

$M(r,c)$  is the payout  
to each player

	c1	c2	c3
r1	(2,-2)	(1,-1)	(-1,1)
r2	(3,-3)	(0,0)	(-2,2)
r3	(-1,1)	(2,-2)	(1,-1)
r4	(0,0)	(0,0)	(1,-1)

Both players want to  
maximize *their* payout.

# Who plays first? Two options

What if the row player first announces their “move”,  
*then* column player responds?

	c1	c2	c3
r1	2	1	-1
r2	3	0	-2
r3	-1	2	1
r4	0	0	1

Row player trying to maximize  
Column player trying to minimize

Results in (r4,c1) or (r4,c2)  
(Row) payoff equals 0.

$$\max_{r \in [R]} \min_{c \in C} M(r, c) = 0$$

Play order →

← “Solve order”

To play first, row player must calculate  
 $\min_c M(1,c)$ ,  $\min_c M(2,c)$ ,  $\min_c M(3,c)$   
and play the move that maximizes those.

# Let's reverse roles: now column player first

	c1	c2	c3
r1	2	1	-1
r2	3	0	-2
r3	-1	2	1
r4	0	0	1

Row player trying to maximize  
Column player trying to minimize

Results in (c3,r3) or (c3,r4)  
(Row) payoff equals 1.

$$\min_{c \in C} \max_{r \in [R]} M(r, c) = 1$$

Play order →

← "Solve order"

To play first, column player must calculate  
 $\max_r M(r, 1)$ ,  $\max_r M(r, 2)$ , etc  
and play the move that minimizes those.

$$\min_{c \in C} \max_{r \in [R]} M(r, c) = 1$$

$$\max_{r \in [R]} \min_{c \in C} M(r, c) = 0$$

**Theorem:**  $\min_{c \in [C]} \max_{r \in [R]} M(r, c) \geq \max_{r \in [R]} \min_{c \in [C]} M(r, c)$

**Proof:**  $\max_{r \in [R]} M(r, c) \geq M(r', c)$  for any  $r', c$

$$\min_{c \in [C]} \max_{r \in [R]} M(r, c) \geq \min_{c \in [C]} M(r', c) \text{ for any } r'$$

A “pure” strategy is a commitment to play a single action.

**For “pure strategies”, advantageous to go second.**

A “mixed” strategy is a distribution over actions.

Row player announces a distribution  $p$  over their actions.

Column player announces a distribution  $q$  over their actions.

Then,  $r \sim p, c \sim q$  and a payoff  $M(r, c)$  is realized.

Expected payoff is  $\mathbb{E}_{r \sim p, c \sim q}[M(r, c)] = p^T M q$

**Theorem:**  $\min_{q \in \Delta_C} \max_{p \in \Delta_R} p^T M q = \max_{p \in \Delta_R} \min_{q \in \Delta_C} p^T M q = v^*$

Value of the game

$$\exists p \in \Delta_R \forall q \in \Delta_C p^T M q \geq v^*$$

$$\exists q \in \Delta_C \forall p \in \Delta_R p^T M q \leq v^*$$

**For “mixed strategies”, order does not matter!**

## Zero-sum games [\[ edit \]](#)

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The minimax theorem was first proven and published in 1928 by [John von Neumann](#),<sup>[3]</sup> who is quoted as saying "*As far as I can see, there could be no theory of games ... without that theorem ... I thought there was nothing worth publishing until the Minimax Theorem was proved*".<sup>[4]</sup>

Formally, von Neumann's minimax theorem states:

Let  $X \subset \mathbb{R}^n$  and  $Y \subset \mathbb{R}^m$  be [compact convex](#) sets. If  $f : X \times Y \rightarrow \mathbb{R}$  is a continuous function that is concave-convex, i.e.

$f(\cdot, y) : X \rightarrow \mathbb{R}$  is [concave](#) for fixed  $y$ , and

$f(x, \cdot) : Y \rightarrow \mathbb{R}$  is [convex](#) for fixed  $x$ .

Then we have that

$$\max_{x \in X} \min_{y \in Y} f(x, y) = \min_{y \in Y} \max_{x \in X} f(x, y).$$

## Sion's minimax theorem

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From Wikipedia, the free encyclopedia

In [mathematics](#), and in particular [game theory](#), **Sion's minimax theorem** is a generalization of [John von Neumann's minimax theorem](#), named after [Maurice Sion](#).

It states:

Let  $X$  be a [compact convex](#) subset of a [linear topological space](#) and  $Y$  a convex subset of a linear topological space. If  $f$  is a real-valued [function](#) on  $X \times Y$  with

$f(x, \cdot)$  [upper semicontinuous](#) and [quasi-concave](#) on  $Y$ ,  $\forall x \in X$ , and

$f(\cdot, y)$  lower semicontinuous and quasi-convex on  $X$ ,  $\forall y \in Y$

then,

$$\min_{x \in X} \sup_{y \in Y} f(x, y) = \sup_{y \in Y} \min_{x \in X} f(x, y).$$