

# Algorithmic game theory

Martin Balko

2nd lecture

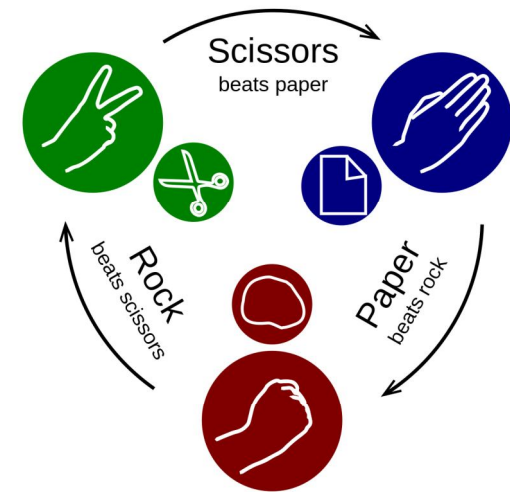
October 11th 2024



# Proof of Nash's Theorem

# Nash equilibria in normal-form games

	Rock	Paper	Scissors
Rock	(0,0)	(-1,1)	(1,-1)
Paper	(1,-1)	(0,0)	(-1,1)
Scissors	(-1,1)	(1,-1)	(0,0)



Sources: <https://en.wikipedia.org/>

- We introduced perhaps the most influential solution concept, which captures a notion of stability.
- The **best response** of player  $i$  to a strategy profile  $s_{-i}$  is a mixed strategy  $s_i^*$  such that  $u_i(s_i^*; s_{-i}) \geq u_i(s_i'; s_{-i})$  for each  $s_i' \in S_i$ .
- For a normal-form game  $G = (P, A, u)$  of  $n$  players, a **Nash equilibrium (NE)** in  $G$  is a strategy profile  $(s_1, \dots, s_n)$  such that  $s_i$  is a best response of player  $i$  to  $s_{-i}$  for every  $i \in P$ .
- Amazingly, **every normal-form game has a Nash equilibrium.**

# Nash's Theorem

## Nash's Theorem (Theorem 2.16)

Every normal-form game has a Nash equilibrium.

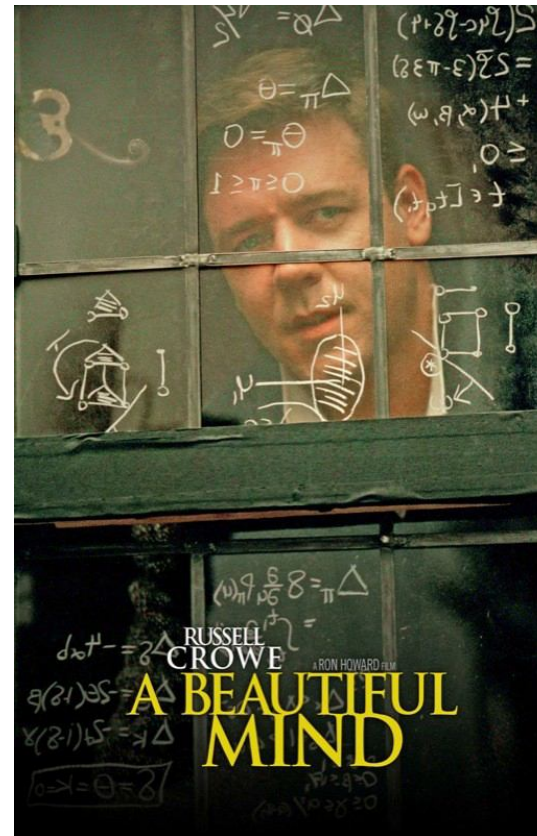


Figure: John Forbes Nash Jr. (1928–2015) and his depiction in the movie **A Beautiful mind**.

# Preparations for the proof of Nash's theorem

- The proof is essentially **topological**, as its main ingredient is a fixed-point theorem. We use a theorem due to **Brouwer**.
- For  $d \in \mathbb{N}$ , a subset  $X$  of  $\mathbb{R}^d$  is **compact** if  $X$  is closed and bounded.
- We say that a subset  $Y$  of  $\mathbb{R}^d$  is **convex** if every line segment containing two points from  $Y$  is fully contained in  $Y$ . Formally: for all  $x, y$  from  $Y$ ,  $tx + (1 - t)y \in Y$  for every  $t \in [0, 1]$ .
- For  $n$  affinely independent points  $x_1, \dots, x_n \in \mathbb{R}^d$ , an  **$(n - 1)$ -simplex**  $\Delta_n$  on  $x_1, \dots, x_n$  is the set of convex combinations of the points  $x_1, \dots, x_n$ . Each simplex is a compact convex set in  $\mathbb{R}^d$ .

## Lemma (Lemma 2.18)

For  $n, d_1, \dots, d_n \in \mathbb{N}$ , let  $K_1, \dots, K_n$  be compact sets, each  $K_i$  lying in  $\mathbb{R}^{d_i}$ . Then,  $K_1 \times \dots \times K_n$  is a compact set in  $\mathbb{R}^{d_1 + \dots + d_n}$ .

# Brouwer's Fixed Point Theorem

- For each  $d \in \mathbb{N}$ , let  $K$  be a non-empty compact convex set in  $\mathbb{R}^d$  and  $f: K \rightarrow K$  be a continuous mapping. Then, there exists a fixed point  $x_0 \in K$  for  $f$ , that is,  $f(x_0) = x_0$ .



Figure: L. E. J. Brouwer (1881–1966).

Source: <https://arxiv.org/pdf/1612.06820.pdf>

- [https://www.youtube.com/watch?v=csInNn6pfT4&t=268s&ab\\_](https://www.youtube.com/watch?v=csInNn6pfT4&t=268s&ab_)

# Proof of Nash's Theorem I

- Let  $G = (P, A, u)$  be a normal-form game of  $n$  players. Recall that  $S_i$  is the set of mixed strategies of player  $i$ .
- We want to apply **Brouwer's theorem**, thus we need to find a suitable compact convex body  $K$  and a continuous mapping  $f: K \rightarrow K$  whose fixed points are NE in  $G$ .
- We start with  $K$ . Let  $K = S_1 \times \cdots \times S_n$  be the set of all mixed strategies.
  - We verify that  $K$  is compact and convex.
  - By definition, each  $S_i$  is, a simplex which is compact and convex.
  - By **Lemma 2.18**, the set  $K = S_1 \times \cdots \times S_n$  is **compact**.
  - For any strategy profiles  $s = (s_1, \dots, s_n)$ ,  $s' = (s'_1, \dots, s'_n) \in K$  and a number  $t \in [0, 1]$ , the point

$$ts + (1 - t)s' = (ts_1 + (1 - t)s'_1, \dots, ts_n + (1 - t)s'_n)$$

is also a mixed-strategy profile in  $K$ . Thus,  $K$  is **convex**.



# Proof of Nash's Theorem II

- We now find the continuous mapping  $f: K \rightarrow K$ .
- For every player  $i \in P$  and action  $a_i \in A_i$ , we define a mapping  $\varphi_{i,a_i}: K \rightarrow \mathbb{R}$  by setting

$$\varphi_{i,a_i}(s) = \max\{0, u_i(a_i; s_{-i}) - u_i(s)\}.$$

- $\varphi_{i,a_i}(s) > 0$  iff  $i$  can improve his payoff by using  $a_i$  instead of  $s_i$ .
- By the definition of  $u_i$ , this mapping is **continuous**.
- Given  $s \in K$ , we define a new “improved” strategy profile  $s' \in K$  as

$$s'_i(a_i) = \frac{s_i(a_i) + \varphi_{i,a_i}(s)}{\sum_{b_i \in A_i} (s_i(b_i) + \varphi_{i,b_i}(s))} = \frac{s_i(a_i) + \varphi_{i,a_i}(s)}{1 + \sum_{b_i \in A_i} \varphi_{i,b_i}(s)}.$$

- “Increase probability at actions that are better responses to  $s_{-i}$ .”
- $s' \in K$  as each  $s'_i(a_i)$  lies in  $[0, 1]$  and  $\sum_{a_i \in A_i} s'_i(a_i) = 1$ .
- We then define  $f$  by setting  $f(s) = s'$ .



# Proof of Nash's Theorem III

- Then,  $f$  is **continuous**, since the mappings  $\varphi_{i,a_i}$  are.
- It remains to show that **fixed points of  $f$  are exactly NE** in  $G$ . Then, **Brouwer's theorem** gives us a fixed point of  $f$ , which is NE in  $G$ .
- **First**, if  $s$  is NE, then all functions  $\varphi_{i,a_i}$  are constant zero functions and thus  $f(s) = s$ . So  $s$  is a fixed point for  $f$ .
- **Second**, assume that  $s = (s_1, \dots, s_n) \in K$  is a fixed point for  $f$ .
  - For any player  $i$ , there is  $a'_i \in A_i$  with  $s_i(a_i) > 0$  such that  $u_i(a'_i; s_{-i}) \leq u_i(s)$ . Otherwise,  $u_i(s) < \sum_{a_i \in A_i} s_i(a_i) u_i(a_i; s_{-i})$ , which is impossible by the **linearity of the expected payoff**.
  - Then,  $\varphi_{i,a'_i}(s) = 0$  and we get  $s'_i(a'_i) = \frac{s_i(a'_i)}{1 + \sum_{b_i \in A_i} \varphi_{i,b_i}(s)}$ .
  - Since  $s$  is a fixed point, we get  $s'_i(a'_i) = s_i(a'_i)$  and, since  $s_i(a'_i) > 0$ , the denominator in the denominator is 1. This means that  $\varphi_{i,b_i}(s) = 0$  for every  $b_i \in A_i$ . It follows that  $s$  is NE as

$$u_i(s''_i; s_{-i}) = \sum_{b_i \in A_i} s''_i(b_i) u_i(b_i; s_{-i}) \leq \sum_{b_i \in A_i} s''_i(b_i) u_i(s) = u_i(s).$$



# Nash's Theorem: remarks

- Two pages worth of **Nobel prize!**

48 MATHEMATICS: J. F. NASH, JR. Proc. N. A. S.

This follows from the arguments used in a forthcoming paper.<sup>12</sup> It is proved by constructing an "abstract" mapping cylinder of  $\lambda$  and transcribing into algebraic terms the proof of the analogous theorem on CW-complexes.

<sup>12</sup> This note arose from consultations during the tenure of a John Simon Guggenheim Memorial Fellowship by MacLane.

<sup>13</sup> Whitehead, J. H. C., "Combinatorial Homotopy I and II," *Bull. A.M.S.*, 55, 214-246 and 453-480 (1949). We refer to these papers as CH I and CH II, respectively.

<sup>14</sup> By a complex we shall mean a connected CW complex, as defined in §5 of CH I. We do not restrict ourselves to finite complexes. A fixed 0-cell  $e^0$  of  $K$  will be the base point for all the homotopy groups in  $K$ .

<sup>15</sup> MacLane, S., "Cohomology Theory in Abstract Groups III," *Ann. Math.*, 50, 736-761 (1949), referred to as CT III.

<sup>16</sup> An (unpublished) result like Theorem 1 for the homotopy type was obtained prior to these results by J. A. Zilber.

<sup>17</sup> CT III uses in place of equation (2.4) the stronger hypothesis that  $\lambda B$  contains the center of  $A$ , but all the relevant developments there apply under the weaker assumption (2.4).

<sup>18</sup> Eilenberg, S., and MacLane, S., "Cohomology Theory in Abstract Groups II," *Ann. Math.*, 48, 829-941 (1947).

<sup>19</sup> Eilenberg, S., and MacLane, S., "Determination of the Second Homology ... by Means of Homotopy Invariants," *Proc. Festschrifts*, 32, 277-280 (1946).

<sup>20</sup> Birkhoff, G., "Some Relations Between Homology and Homotopy Groups," *Ann. Math.*, 49, 428-461 (1949), §12.

<sup>21</sup> The hypothesis of Theorem C, requiring that  $\pi^{-1}(1)$  not be cyclic, can be readily realized by suitable choice of the free group  $X$ , but this hypothesis is not needed here (cf. §7).

<sup>22</sup> Eilenberg, S., and MacLane, S., "Homology of Spaces with Operators II," *Trans. A.M.S.*, 65, 49-99 (1949), referred to as HSO II.

<sup>23</sup>  $C(K)$  here is the  $C(A)$  of CH II. Note that  $K$  exists and is a CW complex by (N) of p. 251 of CH I and that  $p^{-1}K^* = B^*$ , where  $p$  is the projection  $p: K \rightarrow K^*$ .

<sup>24</sup> Whitehead, J. H. C., "Simple Homotopy Types." If  $W = 1$ , Theorem 5 follows from (17-2) on p. 155 of S. Lefschetz, *Algebraic Topology*, (New York, 1942) and arguments in §6 of J. H. C. Whitehead, "On Simply Connected 4-Dimensional Polyhedra" (*Comm. Math. Helv.*, 22, 48-92 (1949)). However this proof cannot be generalized to the case  $W \neq 1$ .

## EQUILIBRIUM POINTS IN $n$ -PERSON GAMES

By JOHN F. NASH, JR.\*

PRINCETON UNIVERSITY

Communicated by S. Lefschetz, November 16, 1949

One may define a concept of an  $n$ -person game in which each player has a finite set of pure strategies and in which a definite set of payments to the  $n$  players corresponds to each  $n$ -tuple of pure strategies, one strategy being taken for each player. For mixed strategies, which are probability

VOL. 38, 1950 MATHEMATICS: G. POLYA 49

distributions over the pure strategies, the pay-off functions are the expectations of the players, thus becoming polynomial forms in the probabilities with which the various players play their various pure strategies.

Any  $n$ -tuple of strategies, one for each player, may be regarded as a point in the product space obtained by multiplying the  $n$  strategy spaces of the players. One such  $n$ -tuple counters another if the strategy of each player in the countering  $n$ -tuple yields the highest obtainable expectation for its player against the  $n-1$  strategies of the other players in the countered  $n$ -tuple. A self-countering  $n$ -tuple is called an equilibrium point.

The correspondence of each  $n$ -tuple with its set of countering  $n$ -tuples gives a one-to-many mapping of the product space into itself. From the definition of countering we see that the set of countering points of a point is convex. By using the continuity of the pay-off functions we see that the graph of the mapping is closed. The closedness is equivalent to saying: if  $P_1, P_2, \dots$  and  $Q_1, Q_2, \dots, Q_n, \dots$  are sequences of points in the product space where  $Q_n \rightarrow Q$ ,  $P_n \rightarrow P$  and  $Q_n$  counters  $P_n$  then  $Q$  counters  $P$ .

Since the graph is closed and since the image of each point under the mapping is convex, we infer from Kakutani's theorem<sup>1</sup> that there is an equilibrium point (i.e., point contained in its image). Hence there is an equilibrium point.

In the two-person zero-sum case the "main theorem"<sup>2</sup> and the existence of an equilibrium point are equivalent. In this case any two equilibrium points lead to the same expectations for the players, but this need not occur in general.

<sup>1</sup> The author is indebted to Dr. David Gale for suggesting the use of Kakutani's theorem to simplify the proof and to the A. E. C. for financial support.

<sup>2</sup> Kakutani, S., *Duke Math. J.*, 8, 467-489 (1941).

<sup>3</sup> Von Neumann, J., and Morgenstern, O., *The Theory of Games and Economic Behavior*, Chap. 3, Princeton University Press, Princeton, 1947.

## REMARK ON WEYL'S NOTE "INEQUALITIES BETWEEN THE TWO KINDS OF EIGENVALUES OF A LINEAR TRANSFORMATION"\*

By GEORGE POLYA

DEPARTMENT OF MATHEMATICS, STANFORD UNIVERSITY

Communicated by H. Weyl, November 25, 1949

In the note quoted above H. Weyl proved a Theorem involving a function  $\varphi(\lambda)$  and concerning the eigenvalues  $\alpha_i$  of a linear transformation  $A$  and those,  $\epsilon_i$ , of  $A^*A$ . If the  $\epsilon_i$  and  $\lambda_i = |\alpha_i|^2$  are arranged in descending order,

Sources: J. F. Nash: Equilibrium points in  $n$ -person games (1950).

- Requires **finite numbers of players and actions**, both assumptions are necessary. (Consider 2-player game "who guesses larger number wins".)
- The proof is **non-constructive**. How to find NE efficiently?

# Pareto optimality

# Pareto optimality

- A brief detour: another example of an **interesting solution concept**, other than NE.
- We want to capture “the best” state of a game. Might be difficult, consider the **Battle of sexes**.
- A strategy profile  $s$  in  $G$  **Pareto dominates**  $s'$ , written  $s' \prec s$ , if, for every player  $i$ ,  $u_i(s) \geq u_i(s')$ , and there exists a player  $j$  such that  $u_j(s) > u_j(s')$ .
  - The relation  $\prec$  is a **partial ordering** of the set  $S$  of all strategy profiles of  $G$ .
  - The outcomes of  $G$  that are considered best are the **maximal elements** of  $S$  in  $\prec$ .
- A strategy profile  $s \in S$  is **Pareto optimal** if there does not exist another strategy profile  $s' \in S$  that Pareto dominates  $s$ .
  - In zero-sum games, all strategy profiles are Pareto-optimal.
  - Not all NE are Pareto-optimal (the NE in Prisoner's dilemma)

# Vilfredo Pareto

- an Italian engineer, sociologist, economist, political scientist, and philosopher.

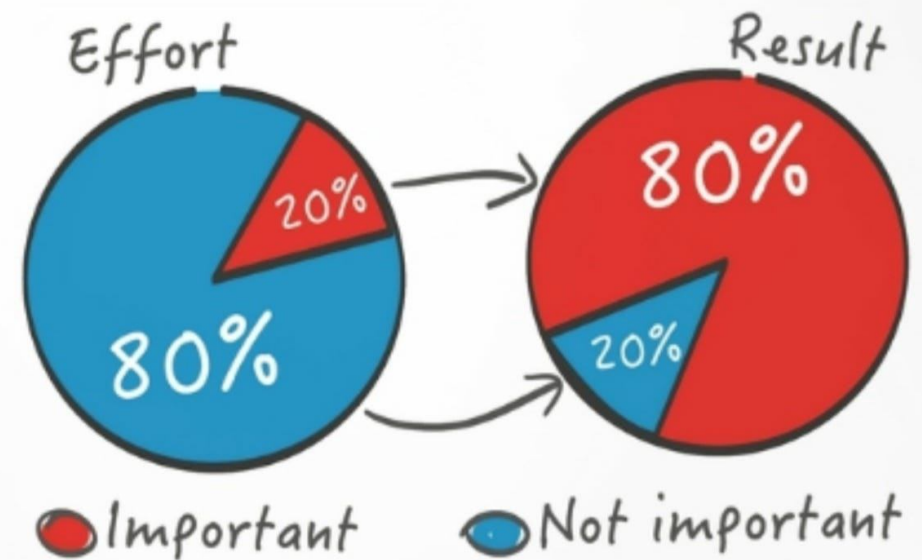
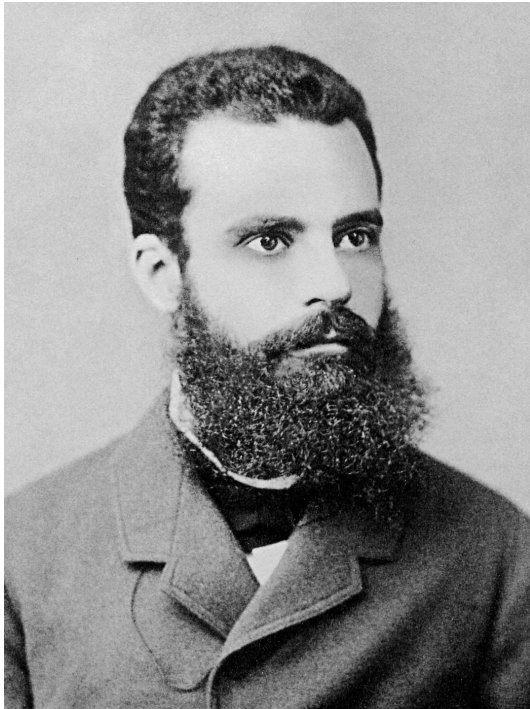


Figure: Vilfredo Pareto (1848–1923).

Sources: <https://en.wikipedia.org> and <https://medium.com/>

- **Pareto principle:** for many outcomes roughly 80% of consequences come from 20% of the causes.

# Finding Nash equilibria in zero-sum games

# Finding Nash equilibria

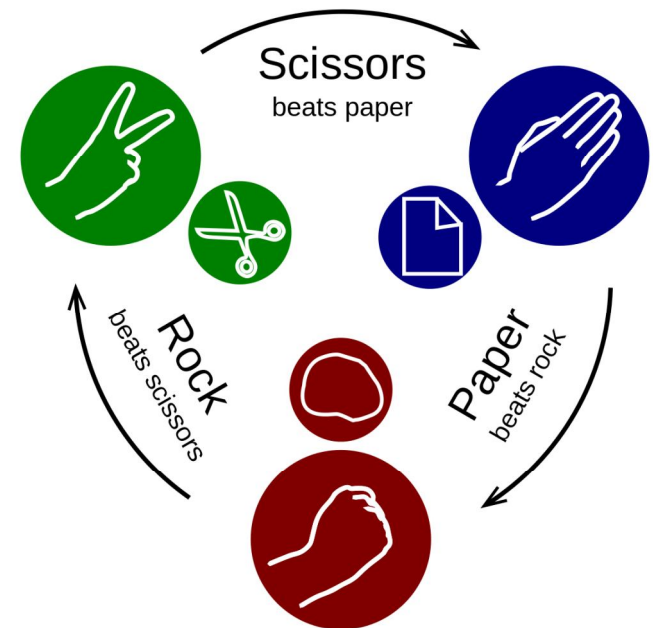
- We know that NE exist in every normal-form game (**Nash's theorem**).
- However, **we do not have any algorithm** for how to find them yet.
- We start with a simple class of 2-player games, so-called **zero-sum games**.
- We show that **we can find NE efficiently in this case**. In fact, we show that NE “solves” zero-sum games completely.
- Historically, zero-sum games were considered first in game theory (by **Morgenstern** and **Von Neumann** in the 1940s).



# Zero-sum games

- Two-player games  $(P, A, u)$  where  $u_1(a) = -u_2(a)$  for every  $a \in A$ .

	Rock	Paper	Scissors
Rock	(0,0)	(-1,1)	(1,-1)
Paper	(1,-1)	(0,0)	(-1,1)
Scissors	(-1,1)	(1,-1)	(0,0)



Sources: <https://en.wikipedia.org/>

# Zero-sum games examples: chess



Source: <https://edition.cnn.com/>

# Zero-sum games examples: table tennis



Source: <https://www.reddit.com/>

# Zero-sum games examples: derivative trading



Source: <https://www.linkedin.com/>



# Zero-sum games examples: elections



Source: <https://news.sky.com/>

# Zero-sum games examples: many more



Source: <https://lhongtortai.com/collection/what-is-a-non-zero-sum-game>

# Representing zero-sum games

- With zero-sum games, our notation simplifies.
- Let  $G = (P, A = A_1 \times A_2, u)$  be a zero-sum game. That is,  $u_1(a) + u_2(a) = 0$  for every  $a \in A$ .
- If  $A_1 = \{1, \dots, m\}$  and  $A_2 = \{1, \dots, n\}$ , then  $G$  can be represented with an  $m \times n$  payoff matrix  $M$  where  $M_{i,j} = u_1(i, j) = -u_2(i, j)$ .
- For a strategy profile  $(s_1, s_2)$ , we write  $x_i = s_1(i)$  and  $y_j = s_2(j)$ , representing  $(s_1, s_2)$  with **mixed strategy vectors**  $x = (x_1, \dots, x_m)$  and  $y = (y_1, \dots, y_n)$  that satisfy  $\sum_{i=1}^m x_i = 1$  and  $\sum_{j=1}^n y_j = 1$ .
- The **expected payoff** of player 1 then equals

$$u_1(s) = \sum_{a=(i,j) \in A} u_1(a) s_1(i) s_2(j) = \sum_{i=1}^m \sum_{j=1}^n M_{i,j} x_i y_j = x^\top M y = -u_2(s).$$



# Worst-case optimal strategies

- Thus, player's 2 best response to a strategy  $x$  of 1, is a vector  $y \in S_2$  that minimizes  $x^\top My$ . Player's 1 best response to a strategy  $y$  of 2 is  $x \in S_1$  that maximizes  $x^\top My$ .
- Let  $\beta(x) = \min_{y \in S_2} x^\top My$  be the best expected payoff of 2 against  $x$ . Let  $\alpha(y) = \max_{x \in S_1} x^\top My$  be the best expected payoff of 1 to  $y$ .
- A strategy profile  $(x, y)$  is then a **NE** if and only if it satisfies  $\beta(x) = x^\top My = \alpha(y)$ .
- Assume player 1 expects player 2 to select a best response to every strategy  $x$  he can come up with. Player 1 then chooses a mixed strategy  $\bar{x}$  from  $S_1$  that maximizes his expected payoff under this, rather pessimistic, assumption.
- This **worst-case optimal strategy** for 1 satisfies  $\beta(\bar{x}) = \max_{x \in S_1} \beta(x)$ . The worst-case optimal strategy for 2 is a mixed strategy  $\bar{y} \in S_2$  that satisfies  $\alpha(\bar{y}) = \min_{y \in S_2} \alpha(y)$ .

# Worst-case optimal strategies and NE

- To achieve NE in a zero-sum game, both players must select their worst-case optimal strategies.

## Lemma 2.20

- (a) For all  $x \in S_1$  and  $y \in S_2$ , we have  $\beta(x) \leq x^\top M y \leq \alpha(y)$ .
  - (b) If a strategy profile  $(x^*, y^*)$  is NE, then both strategies  $x^*$  and  $y^*$  are worst-case optimal.
  - (c) Any strategies  $x^* \in S_1$  and  $y^* \in S_2$  satisfying  $\beta(x^*) = \alpha(y^*)$  form NE  $(x^*, y^*)$ .
- 
- (a) This follows immediately from the definitions of  $\beta$  and  $\alpha$ .
  - (b) Part (a) implies that  $\beta(x) \leq \alpha(y^*)$  for every  $x \in S_1$ . Since  $(x^*, y^*)$  is NE, we have  $\beta(x^*) = \alpha(y^*)$  and thus  $\beta(x) \leq \beta(x^*)$  for every  $x \in S_1$ . Thus,  $x^*$  is a worst-case optimal for 1. Analogously for player 2.
  - (c) If  $\beta(x^*) = \alpha(y^*)$ , then (a) implies  $\beta(x^*) = (x^*)^\top M y^* = \alpha(y^*)$ . □

# The Minimax Theorem

## The Minimax Theorem (Theorem 2.21)

For every zero-sum game, worst-case optimal strategies for both players exist and can be efficiently computed. There is a number  $v$  such that, for any worst-case optimal strategies  $x^*$  and  $y^*$ , the strategy profile  $(x^*, y^*)$  is a Nash equilibrium and  $\beta(x^*) = (x^*)^\top M y^* = \alpha(y^*) = v$ .



Figure: John von Neumann (1903–1957) and Oskar Morgenstern (1902–1977).

# The Minimax Theorem: remarks

- It was a starting point of game theory.
- Proved by **Von Neumann** in 1928 (predates Nash's Theorem).
- *"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."* (Von Neumann).
- The Minimax theorem tells us everything about zero-sum games: **there is NE** and it can be found **efficiently**. Moreover, there is a unique **value of the game**  $v = (x^*)^\top M(y^*)$  of the payoff attained in any NE  $(x^*, y^*)$ .
- There are **no secrets in zero-sum games**: strategies known in advance change nothing, each player can choose a worst-case optimal strategy and get payoff  $\geq v$ . If the opponent chooses his worst-case optimal strategy, then his payoff is always  $\leq v$ .
- **The name**: the expanded equality  $\beta(x^*) = v = \alpha(y^*)$  becomes

$$\max_{x \in S_1} \min_{y \in S_2} x^\top M y = v = \min_{y \in S_2} \max_{x \in S_1} x^\top M y.$$

- Original proof uses **Brouwer's theorem**. We will use **linear programming**.



Source: <https://czthomas.files.wordpress.com>

Thank you for your attention.