

## Game-Theory Basics

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**Example 1.1** (Prisoner’s Dilemma). *Two criminals are interrogated separately. Each of them has two possible strategies: (C)onfess, remain (S)ilent. Confessing yields a smaller sentence if the other one is silent. If both confess, the sentence is larger for both (4 years) compared to when they both remain silent (2 years).*

		S	C
S		2	1
C		5	4
	1	2	5
	4	1	5

- If both players remain (S)ilent, the total cost is smallest.
- If both players (C)onfess, the cost is larger for both of them.
- Still, for each player confessing is always the preference!

**Definition 1.2.** A (normal form, cost minimization) game is a triple  $(\mathcal{N}, (S_i)_{i \in \mathcal{N}}, (c_i)_{i \in \mathcal{N}})$  where

- $\mathcal{N}$  is the set of players,  $n = |\mathcal{N}|$ ,
- $S_i$  is the set of (pure) strategies of player  $i$ ,
- $S = \prod_{i \in \mathcal{N}} S_i$  is the set of states,
- $c_i: S \rightarrow \mathbb{R}$  is the cost function of player  $i \in \mathcal{N}$ . In state  $s \in S$ , player  $i$  has a cost of  $c_i(s)$ .

We denote by  $s_{-i} = (s_1, \dots, s_{i-1}, s_{i+1}, \dots, s_n)$  a state  $s$  without the strategy  $s_i$ . This notation allows us to concisely define a unilateral deviation of a player. For  $i \in \mathcal{N}$ , let  $s \in S$  and  $s'_i \in S_i$ , then  $(s'_i, s_{-i}) = (s_1, \dots, s_{i-1}, s'_i, s_{i+1}, \dots, s_n)$ .

**Definition 1.3.** A pure strategy  $s_i$  is called a dominant strategy for player  $i \in \mathcal{N}$  if  $c_i(s_i, s_{-i}) \leq c_i(s'_i, s_{-i})$  for every  $s'_i \in S_i$  and every  $s_{-i}$ .

**Definition 1.4.** A state  $s \in S$  is called a dominant strategy equilibrium if for every player  $i \in \mathcal{N}$  strategy  $s_i \in S_i$  is a dominant strategy.

Not every game has a dominant strategy equilibrium.

**Example 1.5** (Battle of the Sexes). *Suppose Angelina and Brad go to the movies. Angelina prefers watching movie A, Brad prefers watching movie B. However, both prefer watching a movie together to watching movies separately.*

		A	B
A		1	6
B		5	2
	5	1	6
	2	5	1

There is no dominant strategy of either of the two player: In state (A,A) the preference for both is A. In state (B,B) the preference for both is B.

What is a likely outcome in this situation?

**Definition 1.6.** A strategy  $s_i$  is called a best response for player  $i \in \mathcal{N}$  against a collection of strategies  $s_{-i}$  if  $c_i(s_i, s_{-i}) \leq c_i(s'_i, s_{-i})$  for all  $s'_i \in S_i$ .

Note:  $s_i$  dominant strategy if and only if  $s_i$  best response for all  $s_{-i}$ .

**Definition 1.7.** A state  $s \in S$  is called a pure Nash equilibrium if  $s_i$  is a best response against the other strategies  $s_{-i}$  for every player  $i \in \mathcal{N}$ .

So, a pure Nash equilibrium is stable against unilateral deviation. No player can reduce his cost by only changing his only strategy.

Not every game has a pure Nash equilibrium.

**Example 1.8** (Rock-Paper-Scissors). The well-known game rock-paper-scissors can be represented by the following cost matrix.

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

There is no pure Nash equilibrium: In each of the nine states, at least one of the two players does not play a best response.

**Definition 1.9.** A mixed strategy  $\sigma_i$  for player  $i$  is a probability distribution over the set of pure strategies  $S_i$ .

We will only consider the case of finitely many pure strategies and finitely many players. In this case, we can write a mixed strategy  $\sigma_i$  as  $(\sigma_{i,s_i})_{s_i \in S_i}$  with  $\sum_{s_i \in S_i} \sigma_{i,s_i} = 1$ . The cost of a mixed state  $\sigma$  for player  $i$  is

$$c_i(\sigma) = \sum_{s \in S} p(s) \cdot c_i(s) ,$$

where  $p(s) = \prod_{i \in \mathcal{N}} \sigma_{i,s_i}$  is the probability that the outcome is pure state  $s$ .

**Definition 1.10.** A mixed strategy  $\sigma_i$  is a (mixed) best-response strategy against a collection of mixed strategies  $\sigma_{-i}$  if  $c(\sigma_i, \sigma_{-i}) \leq c(\sigma'_i, \sigma_{-i})$  for all other mixed strategies  $\sigma'_i$ .

**Definition 1.11.** A mixed state  $\sigma$  is called a mixed Nash equilibrium if  $\sigma_i$  is a best-response strategy against  $\sigma_{-i}$  for every player  $i \in \mathcal{N}$ .

Note that every pure strategy is also a mixed strategy and every pure Nash equilibrium is also a mixed Nash equilibrium.

It is enough to only consider deviations to pure strategies.

**Lemma 1.12.** A mixed strategy  $\sigma_i$  is a best-response strategy against  $\sigma_{-i}$  if and only if  $c_i(\sigma_i, \sigma_{-i}) \leq c_i(s'_i, \sigma_{-i})$  for all pure strategies  $s'_i \in S_i$ .

*Proof.* The “only if” part is trivial: Every pure strategy is also a mixed strategy.

For the “if” part, let  $\sigma_{-i}$  be an arbitrary mixed strategy profile for all players except for  $i$ . Furthermore, let  $\sigma_i$  be a mixed strategy for player  $i$  such that  $c_i(\sigma_i, \sigma_{-i}) \leq c_i(s'_i, \sigma_{-i})$  for all pure strategies  $s'_i \in S_i$ .

Observe that for any mixed strategy  $\sigma'_i$ , we have  $c_i(\sigma'_i, \sigma_{-i}) = \sum_{s'_i \in S_i} \sigma_{i,s'_i} c_i(s'_i, \sigma_{-i}) \geq \min_{s'_i \in S_i} c_i(s'_i, \sigma_{-i})$ . Using  $\min_{s'_i \in S_i} c_i(s'_i, \sigma_{-i}) \geq c_i(\sigma_i, \sigma_{-i})$ , we are done.  $\square$

While dominant-strategy and pure Nash equilibria do not necessarily exist, mixed Nash equilibria always exist if the number of players and the number of strategies is finite.

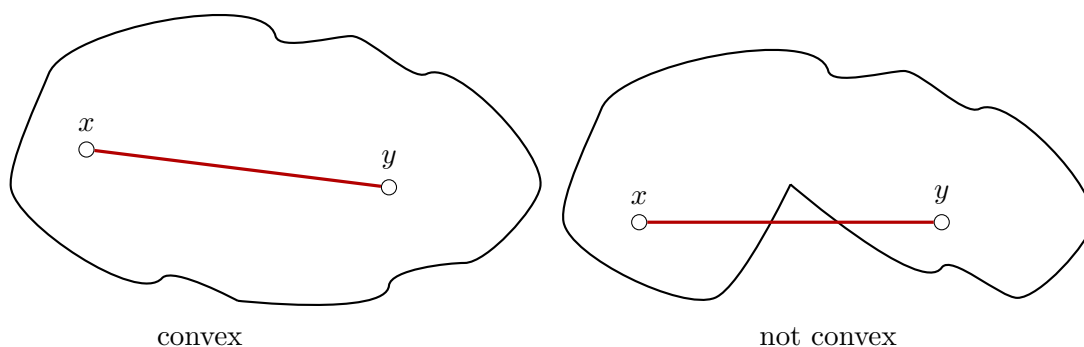
**Theorem 1.13** (Nash Theorem). Every finite normal form game has a mixed Nash equilibrium.

We will use Brouwer’s fixed point theorem to prove it.

**Theorem 1.14** (Brouwer Fixed Point Theorem). Every continuous function  $f: D \rightarrow D$  mapping a compact and convex nonempty subset  $D \subseteq \mathbb{R}^m$  to itself has a fixed point  $x^* \in D$  with  $f(x^*) = x^*$ .

As a reminder, these are the definitions of the terms used in Brouwer’s fixed point theorem. Here,  $\|\cdot\|$  denotes an arbitrary norm, for example,  $\|x\| = \max_i |x_i|$ .

- A set  $D \subseteq \mathbb{R}^m$  is *convex* if for any  $x, y \in D$  and any  $\lambda \in [0, 1]$  we have  $\lambda x + (1 - \lambda)y \in D$ .

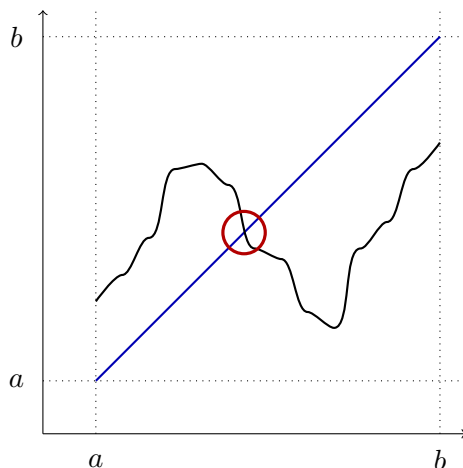


- A set  $D \subseteq \mathbb{R}^m$  is *compact* if and only if it is closed and bounded.
- A set  $D \subseteq \mathbb{R}^m$  is *bounded* if and only if there is some bound  $r \geq 0$  such that  $\|x\| \leq r$  for all  $x \in D$ .
- A set  $D \subseteq \mathbb{R}^m$  is *closed* if it contains all its limit points. That is, consider any convergent sequence  $(x_n)_{n \in \mathbb{N}}$  within  $D$ , i.e.,  $\lim_{n \rightarrow \infty} x_n$  exists and  $x_n \in D$  for all  $n \in \mathbb{N}$ . Then  $\lim_{n \rightarrow \infty} x_n \in D$ .

$[0, 1]$  is closed and bounded  
 $(0, 1]$  is not closed but bounded  
 $[0, \infty)$  is closed and unbounded

- A function  $f: D \rightarrow \mathbb{R}^m$  is *continuous at a point*  $x \in D$  if for all  $\epsilon > 0$ , there exists  $\delta > 0$ , such that for all  $y \in D$ : If  $\|x - y\| < \delta$  then  $\|f(x) - f(y)\| < \epsilon$ .  
 $f$  is called *continuous* if it is continuous at every point  $x \in D$ .

Equivalent formulation of Brouwer's fixed point theorem in one dimension:  
 For all  $a, b \in \mathbb{R}$ ,  $a < b$ , every continuous function  $f: [a, b] \rightarrow [a, b]$  has a fixed point.



*Proof of Theorem 1.13.* Consider a finite normal form game. Without loss of generality let  $\mathcal{N} = \{1, \dots, n\}$ ,  $S_i = \{1, \dots, m_i\}$ . So the set of mixed states  $X$  can be considered a subset of  $\mathbb{R}^m$  with  $m = \sum_{i=1}^n m_i$ .

Exercise: Show that  $X$  is convex and compact.

We will define a function  $f: X \rightarrow X$  that transforms a mixed strategy profile into another mixed strategy profile. The fixed points of  $f$  are shown to be the mixed Nash equilibria of the game.

For mixed state  $x$  and for  $i \in \mathcal{N}$  and  $j \in S_i$ , let

$$\phi_{i,j}(x) = \max\{0, c_i(x) - c_i(j, x_{-i})\} .$$

So,  $\phi_{i,j}(x)$  is the amount by which player  $i$ 's cost would reduce when unilaterally moving from  $x$  to  $j$  if this quantity is positive, otherwise it is 0.

Observe that by Lemma 1.12 a mixed state  $x$  is a Nash equilibrium if and only if  $\phi_{i,j}(x) = 0$  for all  $i = 1, \dots, n$ ,  $j = 1, \dots, m_i$ .

Define  $f: X \rightarrow X$  with  $f(x) = x' = (x'_{1,1}, \dots, x'_{n,m_n})$  by

$$x'_{i,j} = \frac{x_{i,j} + \phi_{i,j}(x)}{1 + \sum_{k=1}^{m_i} \phi_{i,k}(x)}$$

for all  $i = 1, \dots, n$  and  $j = 1, \dots, m_i$ .

Observe that  $x' \in X$ . That means,  $f: X \rightarrow X$  is well defined. Furthermore,  $f$  is continuous. Therefore, by Theorem 1.14,  $f$  has a fixed point, i.e., there is a point  $x^* \in X$  such that  $f(x^*) = x^*$ .

We only need to show that every fixed point  $x^*$  of  $f$  is a mixed Nash equilibrium. So, in other words, we need to show that  $f(x^*) = x^*$  implies that  $\phi_{i,j}(x^*) = 0$  for all  $i = 1, \dots, n$ ,  $j = 1, \dots, m_i$ .

Fix some  $i \in \mathcal{N}$ . Once we have shown that  $\phi_{i,j}(x^*) = 0$  for  $j = 1, \dots, m_i$ , we are done. We observe that there is  $j'$  with  $x^*_{i,j'} > 0$  and  $c_i(x^*) \leq c_i(j', x^*_{-i})$  because  $c_i(x^*)$  is defined to be  $\sum_{j=1}^{m_i} x^*_{i,j} \cdot c_i(j, x^*_{-i})$ . So, it is the weighted average of all costs and it is impossible that every pure strategy has strictly smaller cost than the weighted average. For this  $j'$ ,  $\phi_{i,j'}(x^*) = \max\{0, c_i(x^*) - c_i(j', x^*_{-i})\} = 0$ .

We now use the fact that  $x^*$  is a fixed point. Therefore, we have

$$x^*_{i,j'} = \frac{x^*_{i,j'} + \phi_{i,j'}(x^*)}{1 + \sum_{k=1}^{m_i} \phi_{i,k}(x^*)} = \frac{x^*_{i,j'}}{1 + \sum_{k=1}^{m_i} \phi_{i,k}(x^*)} .$$

As  $x_{i,j'}^* > 0$ , we also have

$$1 = \frac{1}{1 + \sum_{k=1}^{m_i} \phi_{i,k}(x^*)} ,$$

and so

$$\sum_{k=1}^{m_i} \phi_{i,k}(x^*) = 0 .$$

Since  $\phi_{i,k}(x^*) \geq 0$  for all  $k$ , we have to have  $\phi_{i,k}(x^*) = 0$  for all  $k$ . This completes the proof.  $\square$