

## Lecture 1.1: The Minimax Theorem

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In these notes, we show that a Nash equilibrium exists in every finite two-player zero-sum games. This theorem was established by John von Neumann in 1928. While von Neumann's proof was different, we present a proof via strong linear programming duality, provided by Dantzig in the 1940's. Importantly, his existence proof lends itself to efficient algorithms for computing Nash equilibrium in two-player zero-sum games using linear programming. Before showing our main result, we introduce some notation.

## 1 Two-player Game Notation

**Two-Player (Normal-Form) Games.** A *two-player normal-form game* can be specified via a pair  $(R, C)$  of  $m \times n$  *payoff matrices*, where  $m$  is the number of actions available to one player and  $n$  the number of actions available to the other player. We call the two players *row* and *column* player respectively.

As the players' names imply, the actions of the row player are in one-to-one correspondence with the rows of the payoff matrices, while the actions of the column player correspond to the columns. So if the row player plays the action corresponding to row  $i$  and the column player the action corresponding to row  $j$ , then their respective payoffs are given by  $R[i, j]$  and  $C[i, j]$ .

We represent the set of mixed strategies available to the row player by  $\Delta_m := \{\mathbf{x} \in \mathbb{R}_{\geq 0}^m \mid \sum_i x_i = 1\}$  and the set of mixed strategies of the column player by  $\Delta_n$ .

For a pair of mixed strategies  $\mathbf{x} \in \Delta_m$  and  $\mathbf{y} \in \Delta_n$ , the expected payoff of the row and column player are respectively  $\mathbf{x}^T R \mathbf{y}$  and  $\mathbf{x}^T C \mathbf{y}$ .

With this notation, we can re-write the Nash equilibrium conditions as follows.

**Nash Equilibrium.** A pair of strategies  $(\mathbf{x}, \mathbf{y})$  is said to be a *Nash equilibrium* iff

$$\begin{aligned} \mathbf{x}^T R \mathbf{y} &\geq \mathbf{x}'^T R \mathbf{y}, \quad \forall \mathbf{x}' \in \Delta_m; \\ \mathbf{x}^T C \mathbf{y} &\geq \mathbf{x}^T C \mathbf{y}', \quad \forall \mathbf{y}' \in \Delta_n. \end{aligned}$$

Notice that both players' expected payoffs are linear in the players' mixed strategies. Hence, the support of a player's strategy in a Nash equilibrium should include only those pure strategies that maximize her payoff given the other player's mixed strategy. Conversely, any strategy whose support satisfies the above condition must be a Nash equilibrium. This leads to a pair of alternative conditions defining a Nash equilibrium:

$$\begin{aligned} \forall i \quad \text{s.t. } x_i > 0, \quad i &\in \arg \max_k \{\mathbf{e}_k^T R \mathbf{y}\}; \\ \forall j \quad \text{s.t. } y_j > 0, \quad j &\in \arg \max_\ell \{\mathbf{x}^T C \mathbf{e}_\ell\}, \end{aligned}$$

where  $\mathbf{e}_k \in \Delta_m$ ,  $\mathbf{e}_\ell \in \Delta_n$  represent the vectors of appropriate dimensions with a single 1 at the coordinate  $k$  and  $\ell$  respectively; these correspond to the pure strategies  $k$  and  $\ell$  respectively for the row and column players.

Another pair of equivalent conditions asserts that a player cannot improve her payoff by unilaterally switching to a pure strategy, i.e.

$$\begin{aligned} \mathbf{x}^T R \mathbf{y} &\geq \mathbf{e}_i^T R \mathbf{y}, \quad \forall i \in [m] \\ \mathbf{x}^T C \mathbf{y} &\geq \mathbf{x}^T C \mathbf{e}_j, \quad \forall j \in [n], \end{aligned}$$

where  $\mathbf{e}_i$  and  $\mathbf{e}_j$  respectively represent the row player's pure strategy  $i$  and the column player's pure strategy  $j$ .

**Examples.** With this notation in mind, we can restate some classical game examples. Each is described by a table whose  $(i, j)$ -th entry has a pair of values, the first corresponding to  $R[i, j]$  and the second to  $C[i, j]$ .

### Rock Paper Scissors (RPS)

This is the familiar school-yard game with the same name.

	Rock	Paper	Scissors
Rock	0, 0	-1, 1	1, -1
Paper	1, -1	0, 0	-1, 1
Scissors	-1, 1	1, -1	0, 0

It can be checked that this game has a unique Nash equilibrium, namely when both players randomize uniformly over their strategies:  $\mathbf{x} = \mathbf{y} = \langle \frac{1}{3}, \frac{1}{3}, \frac{1}{3} \rangle$ .

### Prisoner's Dilemma

Two suspects are arrested by the police. The police have insufficient evidence for a conviction, and, having separated both prisoners, visit each of them to offer the same deal. If one testifies for the prosecution against the other and the other remains silent, the betrayer goes free and the silent accomplice receives the full 10-year sentence. If both remain silent, both prisoners are sentenced to only six months in jail for a minor charge. If each betrays the other, each receives a five-year sentence. Each prisoner must choose to betray the other or to remain silent. Each one is assured that the other would not know about the betrayal before the end of the investigation. How should the prisoners act?

	Silent	Betray
Silent	$-\frac{1}{2}, -\frac{1}{2}$	-10, 0
Betray	0, -10	-5, -5

This game has a pure Nash equilibrium where both players “Betray”. Observe that this is not the *social optimum*, that is the pair of strategies optimizing the sum of the players’ payoffs, since both players stand to gain if they simultaneously switched to “Silent”. This exposes a key feature of Nash equilibria—that they may fail to optimize social welfare and are stable only under unilateral deviations by the players.

### Battle of the Sexes

A pair of friends with different interests need to decide where to spend their afternoon. They make simultaneous suggestions to each other and will only go somewhere if their suggestions match. One prefers football and the other prefers theater. Their payoffs are as follows.

	Theater!	Football fine
Theater fine	1, 5	0, 0
Football!	0, 0	5, 1

This game has multiple Nash equilibria. In particular, both the squares with non-zero payoffs represent Nash equilibria. In addition,  $\mathbf{x} = \langle \frac{1}{6}, \frac{5}{6} \rangle$  and  $\mathbf{y} = \langle \frac{5}{6}, \frac{1}{6} \rangle$  is a Nash equilibrium. Thus, this game has 3 Nash equilibria. In fact, as we will establish later in this mini-course, *every “non-degenerate game” must have an odd number of Nash equilibria!*

## 2 Two-player Zero-sum Games

**Definition 1.** A two-player game is said to be zero-sum if  $R + C = 0$ , i.e.  $R[i, j] + C[i, j] = 0$  for all  $i \in [m]$  and  $j \in [n]$ .

For example, the RPS game described above is zero-sum. We give another less symmetric game below.

## Presidential Elections (adopted from [4])

Two candidates for presidency have two possible strategies on which to focus their campaigns. Depending on their decisions they may win the favor of a number of voters. The table below describes their gains in millions of voters.

	Morality	Tax-cuts
Economy	3, -3	-1, 1
Society	-2, 2	1, -1

Clearly, this is also a zero-sum game. Suppose now that the row player announces a strategy of  $(\frac{1}{2}, \frac{1}{2})$  in advance. This is presumably a suboptimal way to play the game, since it exposes the row player's strategy to the column player who can now react optimally. Given the row player's announced strategy, the column player will choose the strategy that maximizes her payoff given the row player's declared strategy; in this case, she will choose "Tax-cuts". More generally, if the row player announces strategy  $(x_1, x_2)$ , then the column player has the following expected payoffs:

$$\begin{aligned}\mathbb{E}[\text{"Morality"}] &= -3x_1 + 2x_2 \\ \mathbb{E}[\text{"Tax - cuts"}] &= x_1 - x_2.\end{aligned}$$

Hence, the payoff that the column player will get after reacting optimally to the row player's announced strategy is  $\max(-3x_1 + 2x_2, x_1 - x_2)$ . Since this is a zero-sum game, the resulting row player's payoff is  $-\max(-3x_1 + 2x_2, x_1 - x_2) \equiv \min(3x_1 - 2x_2, -x_1 + x_2)$ . So if the row player is forced to announce his strategy in advance he will choose

$$(x_1, x_2) \in \arg \max_{(x_1, x_2)} \min(3x_1 - 2x_2, -x_1 + x_2).$$

In other words, the row player's optimal announced strategy  $(x_1, x_2)$  can be computed via the following linear program, whose value will be equal to the row player's payoff after the column player's optimal response to  $(x_1, x_2)$ .

$$\begin{aligned}\max z \\ \text{s.t. } 3x_1 - 2x_2 &\geq z \\ -x_1 + x_2 &\geq z \\ x_1 + x_2 &= 1 \\ x_1, x_2 &\geq 0.\end{aligned}$$

Solving this LP yields  $x_1 = \frac{3}{7}, x_2 = \frac{4}{7}$  and  $z = \frac{1}{7}$ . Conversely, if the column player were forced to announce her strategy in advance, she would aim to solve the following LP.

$$\begin{aligned}\max w \\ \text{s.t. } -3y_1 + y_2 &\geq w \\ 2y_1 - y_2 &\geq w \\ y_1 + y_2 &= 1 \\ y_1, y_2 &\geq 0,\end{aligned}$$

which yields the following solution:  $y_1 = \frac{2}{7}, y_2 = \frac{5}{7}$  and  $w = -\frac{1}{7}$ .

Here is a remarkable observation. If the row player plays  $\mathbf{x} = \langle \frac{3}{7}, \frac{4}{7} \rangle$  and the column player plays  $\mathbf{y} = \langle \frac{2}{7}, \frac{5}{7} \rangle$  then

- The payoff of the row player is  $1/7$  and the payoff of the column payer is  $-1/7$ ; in particular, the sum of their payoffs is zero.

- Given strategy  $\mathbf{x} = \langle \frac{3}{7}, \frac{4}{7} \rangle$  for the row player, the column player cannot obtain a payoff greater than  $-\frac{1}{7}$  since the row player gets a payoff of at least  $\frac{1}{7}$  irrespective of the column player's strategy (by the definition of the first LP) and the game is zero-sum;
- Similarly, given  $\mathbf{y} = \langle \frac{2}{7}, \frac{5}{7} \rangle$  for the column player, the row player cannot obtain a payoff greater than  $\frac{1}{7}$  since the column player gets a payoff of at least  $-\frac{1}{7}$  irrespective of the row player's strategy.

Hence, the pair of strategies  $\mathbf{x} = \langle \frac{3}{7}, \frac{4}{7} \rangle$  and  $\mathbf{y} = \langle \frac{2}{7}, \frac{5}{7} \rangle$  is a Nash equilibrium! This was somewhat unexpected since the two LPs were formulated independently by the two players without any care/guessing about what their opponent might do. In particular, it seemed a priori suboptimal/too pessimistic for a player to announce her strategy and let the other player respond. Moreover, there was no reason to expect that the pessimistic strategies for the two players are in fact best responses to each other. As we show below, this coincidence happens for a deep reason, it is true in all two-player zero-sum games, and is a ramification of the strong LP duality.

### 3 The Min-Max Theorem

#### 3.1 LP Formulation

Let us now study general two-player zero-sum games. Recall that a two-player game is zero-sum iff the payoff matrices  $(R, C)$  satisfy  $R + C = 0$ . As in the previous section, let us write the LP for the row player if he is forced to announce his strategy in advance.

$$\begin{aligned} & \max z \\ \text{s.t. } & \mathbf{x}^T R \geq z \mathbf{1}^T \\ & \mathbf{x}^T \mathbf{1} = 1 \\ & \forall i, x_i \geq 0. \end{aligned} \quad LP(1)$$

Notice that the maximum of  $z$  is actually the same as

$$\max_x \min_y \mathbf{x}^T R \mathbf{y}.$$

Now let us consider the dual of  $LP(1)$ :

$$\begin{aligned} & \min z' \\ \text{s.t. } & -\mathbf{y}^T R^T + z' \mathbf{1}^T \geq \mathbf{0} \\ & \mathbf{y}^T \mathbf{1} = 1 \\ & \forall j, y_j \geq 0. \end{aligned} \quad LP(2)$$

Let us make the following change of variable  $z'' = -z'$  and use the fact that  $C = -R$  to change  $LP(2)$  into  $LP(3)$  below, whose value has flipped sign.

$$\begin{aligned} & \max z'' \\ \text{s.t. } & C \mathbf{y} \geq z'' \mathbf{1} \\ & \mathbf{y}^T \mathbf{1} = 1 \\ & \forall j, y_j \geq 0. \end{aligned} \quad LP(3)$$

It is not hard to see that the maximum of  $z''$  is equal to

$$\max_y \min_x \mathbf{x}^T C \mathbf{y} = - \min_y \max_x \mathbf{x}^T R \mathbf{y}.$$

Here is the interesting fact:  $LP(3)$  is nothing but the linear program that the column player would solve if she were forced to announce her strategy in advance! Using the strong duality of linear programming we are going to show next that the solutions to  $LP(1)$  and  $LP(3)$  constitute a Nash equilibrium of the game.

### 3.2 LP Strong Duality

Since  $LP(2)$  is the dual of  $LP(1)$ , LP strong duality implies the following: If  $(\mathbf{x}, z)$  is optimal for  $LP(1)$  and  $(\mathbf{y}, z')$  is optimal for  $LP(2)$ , then  $z = z'$ . On the other hand, given our construction of  $LP(3)$  it follows that, if  $(\mathbf{y}, z')$  is optimal for  $LP(2)$ , then  $(\mathbf{y}, -z')$  is optimal for  $LP(3)$ , and vice versa. Thus, if  $(\mathbf{x}, z)$  is optimal for  $LP(1)$  and  $(\mathbf{y}, z'')$  is optimal for  $LP(3)$ , then  $z = -z''$ .

We show next that the solutions to these LPs actually give us a Nash equilibrium.

**Theorem 1.** *If  $(\mathbf{x}, z)$  is optimal for  $LP(1)$ , and  $(\mathbf{y}, z'')$  is optimal for  $LP(3)$ , then  $(\mathbf{x}, \mathbf{y})$  is a Nash equilibrium of  $(R, C)$ . Moreover, the payoffs of the row/column player in this Nash equilibrium are  $z$  and  $z'' = -z$  respectively.*

**Proof:** Since  $(\mathbf{x}, z)$  is feasible for  $LP(1)$ , and  $(\mathbf{y}, z'')$  is feasible for  $LP(3)$ :

$$\mathbf{x}^T R \geq z \mathbf{1}^T \Rightarrow \mathbf{x}^T R \mathbf{y} \geq z \tag{1}$$

$$\begin{aligned} C \mathbf{y} \geq z'' \mathbf{1} &\Rightarrow \mathbf{x}'^T C \mathbf{y} \geq z'', \forall \mathbf{x}' \\ &\Rightarrow \mathbf{x}'^T R \mathbf{y} \leq -z'', \forall \mathbf{x}'. \end{aligned} \tag{2}$$

From equation (2) we know that, if the column player is playing  $\mathbf{y}$ , the row player's payoff is at most  $-z'' \equiv z$  (since we argued above that  $z'' = -z$  from strong LP duality). Moreover, equation (1) implies that if the row player plays  $\mathbf{x}$  against  $\mathbf{y}$ , her payoff will be at least  $z$ ; in fact, given (2) it is exactly  $z$ . Thus,  $\mathbf{x}$  is a best response for the row player against strategy  $\mathbf{y}$  of the column player, giving payoff of  $z$  to the row player.

Using a similar argument, we can show that  $\mathbf{y}$  is a best response for the column player against strategy  $\mathbf{x}$  of the row player, giving payoff  $z''$  to the column player. Hence,  $(\mathbf{x}, \mathbf{y})$  is a Nash equilibrium in which the players' payoffs are  $z$  and  $z'' = -z$  respectively.  $\square$

We give a couple of corollaries of the above theorem.

**Corollary 1.** *There exists a Nash equilibrium in every two-player zero-sum game.*

**Corollary 2** (The Minimax Theorem).

$$\max_{\mathbf{x}} \min_{\mathbf{y}} \mathbf{x}^T R \mathbf{y} = \min_{\mathbf{y}} \max_{\mathbf{x}} \mathbf{x}^T R \mathbf{y}.$$

Next, we argue that every Nash equilibrium of the two-player zero-sum game can be found by our approach.

**Theorem 2.** *If  $(\mathbf{x}, \mathbf{y})$  is a Nash equilibrium of  $(R, C)$ , then  $(\mathbf{x}, \mathbf{x}^T R \mathbf{y})$  is an optimal solution of  $LP(1)$ , and  $(\mathbf{y}, -\mathbf{x}^T C \mathbf{y})$  is an optimal solution of  $LP(2)$ .*

**Proof:** Let  $z = \mathbf{x}^T R \mathbf{y} = -\mathbf{x}^T C \mathbf{y}$ . Since  $(\mathbf{x}, \mathbf{y})$  is a Nash equilibrium.

$$\begin{aligned} \mathbf{x}^T C \mathbf{y} &\geq \mathbf{x}^T C \mathbf{e}_j \quad \forall j \\ \Rightarrow -\mathbf{x}^T R \mathbf{y} &\geq -\mathbf{x}^T R \mathbf{e}_j \quad \forall j \\ \Rightarrow \mathbf{x}^T R &\geq z \mathbf{1}. \end{aligned}$$

So we have showed that  $(\mathbf{x}, z)$  is a feasible solution for  $LP(1)$ ; similarly we can argue that  $(\mathbf{y}, z)$  is a feasible solution for  $LP(2)$ . Since  $LP(2)$  is the dual of  $LP(1)$ , by weak duality, we know that the value achieved by any feasible solution of  $LP(2)$  upper bounds the value achieved by any feasible solution of

$LP(1)$ . But, these values are equal here. So,  $(\mathbf{x}, z)$  and  $(\mathbf{y}, z)$  must be optimal solutions for  $LP(1)$  and  $LP(2)$  respectively.  $\square$

Since all Nash equilibria define a pair of optimal solutions to  $LP(1)$  and  $LP(2)$  and all pairs of optimal solutions to  $LP(1)$  and  $LP(2)$  define a Nash equilibrium, we obtain the following important corollaries of Theorems 1 and 2.

**Corollary 3** (Convexity of Optimal Strategies). *The set*

$$\{\mathbf{x} \mid \text{there exists } \mathbf{y} \text{ such that } (\mathbf{x}, \mathbf{y}) \text{ is a Nash equilibrium of } (R, C)\}$$

*of equilibrium strategies for the row player in zero-sum game is convex. Ditto for the column player.*

**Corollary 4** (Anything Goes). *If  $\mathbf{x}$  is any equilibrium strategy for the row player in a zero-sum game  $(R, C)$  and  $\mathbf{y}$  is any equilibrium strategy for the column player, then  $(\mathbf{x}, \mathbf{y})$  is a Nash equilibrium of the game.*

**Corollary 5** (Convexity of Equilibrium Set). *The equilibrium set*

$$\{(\mathbf{x}, \mathbf{y}) \mid (\mathbf{x}, \mathbf{y}) \text{ is a Nash equilibrium of } (R, C)\}$$

*of a zero-sum game is convex.*

**Corollary 6** (Uniqueness of Payoffs). *The payoff of the row player is equal in all Nash equilibria of a zero-sum game. Ditto for the column player.*

Inspired by this last corollary, we define the *value of a zero-sum game*, to be the payoff of the row player in any (all) Nash equilibria of the game.

**Definition 2** (Value of a Zero-Sum Game). *If  $(R, C)$  is zero-sum game, then the value of the game is the unique payoff of the row player in all Nash equilibria of the game.*

**Remark 1** (MinMax Equilibrium). *Given our proof of Theorems 1 and 2 as well as Corollary 2, a Nash equilibrium of a two-player zero-sum game is also called a MinMax equilibrium, or a MaxMin equilibrium, or a minimax equilibrium, or sometimes the minimax solution of the game.*

## 4 Exercises

1. **Adjacent Zero-Sum Games:** Players 1 and 2 are playing a zero-sum game with payoff matrices  $(A^{1,2}, A^{2,1})$ ; at the same time players 1 and 3 are playing another zero-sum game with payoff matrices  $(A^{1,3}, A^{3,1})$ . Instead of using different strategies in the two games, player 1 must use the same strategy in both games.

- a. Show that there exists a Nash equilibrium in this game using linear programming duality.
- b. Give an algorithm that can compute a Nash equilibrium efficiently.

2. **Triangle Zero-Sum Games:** Three players 1, 2, 3 are pairwise engaged in two-player games, which need not be zero-sum. Each player has the same strategy set in the two games s/he participates and needs to employ the same strategy in the two games. Moreover, assume that, in every pure strategy profile, the sum of all players' payoff is 0, i.e. that the overall game is zero-sum.

- a. Give an algorithm for computing a Nash equilibrium efficiently using linear programming.
- b. (Optional) Show that there exists a Nash equilibrium in this game using linear programming duality.

## 5 Historical Remarks

The study of games from a mathematical perspective goes back to the work of mathematician Émile Borel. The existence of a Nash equilibrium in two-player zero-sum games was established by John von Neumann in 1928, in a paper [6] that arguably gave birth to the field of Game Theory. Two decades after von Neumann's result it became understood in a discussion between von Neumann and Danzig—the father of linear programming—that the existence of an equilibrium in zero-sum games is implied by Strong Linear Programming duality, and that equilibria can be computed using linear programming (see, e.g., [2], [3]). Hence, as was established another three decades later by Khachiyan [5], finding an equilibrium in zero-sum games is computationally tractable. Danzig also proposed a (computational) reduction from a generic Linear Programming problem to the computation of equilibrium in a zero-sum game, noting that the reduction does always work. Despite this, it has since been quoted that the Min-Max theorem and the Strong Duality Theorem of linear programming are equivalent. In reality, it was only recently that Ilan Adler (himself a student of Danzig) provided a complete reduction from Linear Programming to Zero-sum Games [1], showing that the existence of equilibrium in two-player zero-sum games and strong linear programming duality are equivalent.

## References

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