# Game Theory Theory

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## Introduction

## 1.1 Games

Games provide valuable models for simulating a variety of real-world situations.

**Definition** (Game). A game is a structured process that includes the following components:

- A group of participants, referred to as players, with at least two members.
- An initial state or starting condition.
- A set of rules that define how players can act.
- A range of possible outcomes or end states.
- The preferences of each player concerning these potential outcomes.

# 1.2 Game Theory Assumptions

Game theory operates under the following key assumptions about the players involved:

- 1. Self-interested.
- 2. Rational.

## 1.2.1 Self-interest

Players are assumed to focus solely on their own preferences concerning the outcomes of the game. This is a mathematical assumption, not an ethical judgment. In fact, it is essential for defining what constitutes a rational choice within the framework of game theory.

## 1.2.2 Rationality

**Definition** (*Preference relation*). Let X be a set. A preference relation on X is a binary relation  $\leq$  that satisfies the following properties for all  $x, y, z \in X$ :

- Reflexive:  $x \leq x$  (every element is at least as preferred as itself).
- Complete:  $x \leq y$  or  $y \leq x$  (any two elements can be compared).
- Transitive: if  $x \leq y$  and  $y \leq z$ , then  $x \leq z$  (preferences are consistent across comparisons).

The transitive property ensures that preferences can be consistently ranked.

**Definition** (*Utility function*). Given a preference relation  $\leq$  over a set X, a utility function representing  $\leq$  is a function  $u: X \to \mathbb{R}$  such that:

$$u(x) \ge u(y) \Leftrightarrow x \le y$$

While a utility function may not always exist in specific cases, it does exist in general settings, particularly when X is finite. If a utility function does exist, there are infinitely many such functions, differing by any strictly increasing transformation of the original function.

Each player i is assigned a set  $X_i$ , representing all the choices available to them. Therefore, the set  $X = xX_i$  over which the utility function u is defined represents the combined choices of all players.

Rationality assumptions The following assumptions define the rational behavior of players:

- 1. Consistent preferences: players can establish a preference relation over the game's outcomes, and this ordering is consistent.
- 2. *Utility representation*: players can define a utility function that represents their preference relations when needed.
- 3. Consistent use of probability: players apply the laws of probability consistently, including computing expected utilities and updating probabilities according to Bayes' rule.
- 4. *Understanding consequences*: players comprehend the outcomes of their actions, the impact on other players, and the resulting chain of consequences.
- 5. Application of decision theory: players use decision theory to maximize their utility. Given a set of alternatives X and a utility function u, each player seeks  $\bar{x} \in X$  such that:

$$u(\bar{x}) > u(x) \qquad \forall x \in X$$

One significant consequence of these axioms is the principle of eliminating strictly dominated strategies: a player will not choose an action a if there exists another action b that yields a strictly better outcome, regardless of the actions of other players.

#### Example:

Consider the following games:

| Gain | Probability |
|------|-------------|
| 2500 | 33%         |
| 2400 | 66%         |
| 0    | 1%          |

Table 1.1: Game A

| Gain | Probability |
|------|-------------|
| 2500 | 0%          |
| 2400 | 100%        |
| 0    | 0%          |

Table 1.2: Game B

In a sample of 72 participants, 82% chose to play Game B, indicating a preference for certainty—characteristic of risk-averse individuals. According to expected utility theory, this decision is rational if:

$$u(2400) > \frac{33}{100}u(2500) + \frac{66}{100}u(2400)$$

This simplifies to:

$$\frac{34}{100}u(2400) > \frac{33}{100}u(2500)$$

Now consider the following alternatives:

| Gain | Probability |
|------|-------------|
| 2500 | 33%         |
| 0    | 67%         |

Table 1.3: Game C

| Gain | Probability |
|------|-------------|
| 2400 | 34%         |
| 0    | 66%         |

Table 1.4: Game D

In this new setup, 83% of participants preferred Game C, reflecting a preference for a larger gain even with a lower probability of success. Rationality in this scenario requires:

$$\frac{34}{100}u(2400) < \frac{33}{100}u(2500)$$

This contradicts the earlier experiment, where the opposite preference was observed. Such behavior violates the independence axiom in expected utility theory, which states that consistent preferences should hold under similar probabilistic transformations.

This contradiction is known as the Allais Paradox, demonstrating that individuals do not always act as fully rational decision-makers.

#### Example:

A group of players is asked to choose an integer between 1 and 100. The mean of all chosen numbers, M is then calculated. The objective of the game is to select the number closest to qM, where 0 < q < 1.

A purely rational player would conclude that the optimal number to choose is 1, regardless of the value of q. However, this player is likely to lose.

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For example, let  $q = \frac{1}{2}$ . Since  $M \le 100$ , in the first step, it seems irrational to choose a number greater than  $\frac{1}{2} \cdot 100$ , as this is the initial target value based on the game's rules.

However, in the second step, assuming all players are rational and recognize that others are also rational, each player would realize that others will also choose numbers below 50. Therefore, the new logical step would be to pick a number less than  $\left(\frac{1}{2}\right)^2 \cdot 100$ .

This reasoning continues iteratively: at step n, it becomes irrational to choose a number greater than  $\left(\frac{1}{2}\right)^n \cdot 100$ . Ultimately, after enough steps, the only rational choice would appear to be selecting the smallest possible number (1).

Despite this reasoning, experiments show that the actual winning number is far higher than 1. In fact, the winning number tends to increase as the value of q increases, revealing that real-life behavior often deviates from purely rational game theory predictions.

## 1.3 Bimatrices

Conventionally, Player 1 selects a row, while Player 2 selects a column. This results in a pair of values that represent the utilities for Player 1 and Player 2, respectively. These options can be conveniently displayed in a bimatrix.

#### Example:

Consider the following bimatrix:

$$\begin{pmatrix}
(8 & 8) & (2 & 7) \\
(7 & 2) & (0 & 0)
\end{pmatrix}$$

In this example, Player 1's utilities are given by:

$$\begin{pmatrix} 8 & 2 \\ 7 & 0 \end{pmatrix}$$

Since the second row is strictly dominated by the first (i.e., Player 1's utility in the first row is higher for any choice by Player 2), Player 1 will rationally choose the first row. Similarly, Player 2 will select the first column, as it strictly dominates the second column.

While the principle of eliminating strictly dominated strategies may seem simplistic, it can lead to surprisingly powerful insights and outcomes.

#### Example:

Consider the following two games:

$$\begin{pmatrix}
(10 & 10) & (3 & 15) \\
(15 & 3) & (5 & 5)
\end{pmatrix}$$

$$\begin{pmatrix}
(8 & 8) & (2 & 7) \\
(7 & 2) & (0 & 0)
\end{pmatrix}$$

Note that in the first game, players have outcomes like (10 10) and (15 3), which individually seem to offer higher utilities than most outcomes in the second game. However, applying rational decision-making principles leads to a surprising result.

According to the principle of elimination of dominated strategies, players will end up choosing the outcome pair (8 8) in the second game because it dominates other available

1.3. Bimatrices 5

outcomes. This leads them to prefer the second game over the first game, despite the fact that the first game contains outcomes with higher individual utilities, like (10 10) and (15 3).

Now, consider the expanded form of the first game, which contains even more outcomes:

$$\begin{pmatrix}
(1 & 1) & (11 & 0) & (4 & 0) \\
(0 & 11) & (8 & 8) & (2 & 7) \\
(0 & 4) & (7 & 2) & (0 & 0)
\end{pmatrix}$$

This expanded version of the first game includes all the outcomes from the second game, plus some additional options. However, rationality axioms suggest that in the first game, players should choose the outcome (10 10), which dominates the other possibilities.

Interestingly, in the second game, where fewer options are available, the players end up selecting (8–8). This leads to a paradoxical outcome: having fewer available actions can actually make players better off by simplifying the decision-making process and avoiding suboptimal choices.

#### Example:

Consider the rational outcomes of the following game.

$$\begin{pmatrix} \begin{pmatrix} 0 & 0 \end{pmatrix} & \begin{pmatrix} 1 & 1 \end{pmatrix} \\ \begin{pmatrix} 1 & 1 \end{pmatrix} & \begin{pmatrix} 0 & 0 \end{pmatrix} \end{pmatrix}$$

While we may not know the rational outcomes formally, it is clear that the preferred outcome for both players is (1 1). However, this leads to a coordination problem.

Both pairs of actions result in the same outcome (1 1), but there is no clear way for the players to distinguish between these two strategies. As a result, while the rational outcome is obvious, the players face difficulty coordinating on which specific actions to take to achieve it.

#### Example:

Consider a voting game with three players, each having the following preferences:

- 1. Player 1:  $A \not\supseteq B \not\supseteq C$
- 2. Player 2:  $B \not\subseteq C \not\subseteq A$
- 3. Player 3:  $C \not\subseteq A \not\subseteq B$

Here, the notation  $A \not\subseteq B$  indicates that Player 1 prefers  $B \subseteq A$ , but not vice versa. The winner is determined by the alternative that receives the most votes. However, if there is a tie among three different votes, the alternative chosen by Player 1 will win.

Let's now analyse the rational outcome of the game through the elimination of dominated actions:

- Alternative A is a weakly dominant strategy for Player 1.
- Players 2 and 3 have their least preferred choice as a weakly dominated strategy.

To avoid their worst outcome, Player 2 retains options B and C (ordered in rows), while Player 3 keeps C and A (ordered in columns). Player 1 will consistently choose A. This simplifies the game to a  $2 \times 2$  table with the following outcomes:

1.3. Bimatrices 6

|              | $\mathbf{C}$ | A |
|--------------|--------------|---|
| В            | Α            | Α |
| $\mathbf{C}$ | C            | A |

Since  $C \not\subseteq A$  for both Players 2 and 3, they will choose the outcome in the second row and first column, leading to the final result being C, which is the worst outcome for Player 1.

#### Example:

Consider the classic Chicken Game. Two cars are driving toward each other on a narrow road, and there isn't enough room for both cars to pass. If both cars keep going straight, they will crash, but if at least one deviates, the crash is avoided. The payoff bimatrix for the game is:

$$\begin{pmatrix} (-1 & -1) & (1 & 10) \\ (10 & 1) & (-10 & -10) \end{pmatrix}$$

The payoffs are as follows:

- ullet Both players receive a utility of -1 if they both deviate.
- A player earns 10 for going straight when the other deviates, while the deviating player receives 1.
- If both go straight and crash, they both receive -10.

The best outcomes are either (10 1) and (1 10). However, there is no way to decisively determine which of these two outcomes will occur, as both are equally viable equilibria.

# Extensive form games

## 2.1 Games tree

#### Example:

Three politicians are tasked with deciding whether to raise their salaries. The voting is public and happens sequentially. Each politician prefers a salary increase but also wants to vote against it to maintain public support.

The optimal outcome for each politician is to receive a salary increase while voting against it. The game has the following characteristics:

- 1. The voting process is sequential, with politicians voting one after another.
- 2. Every possible situation is fully known to all players: they are aware of the entire history and all possible future actions.
- 3. The final outcome is determined by the majority vote.

To represent such a game we can use a tree, where each branch represents a player's vote: a YES vote goes left, and a NO vote goes right. The utilities depend on each player's vote and the final outcome:

- 1. YES, but no raise.
- 2. NO, and no raise.
- 3. YES, and a raise.
- 4. NO, and a raise.

The corresponding decision tree looks like this:

2.1. Games tree

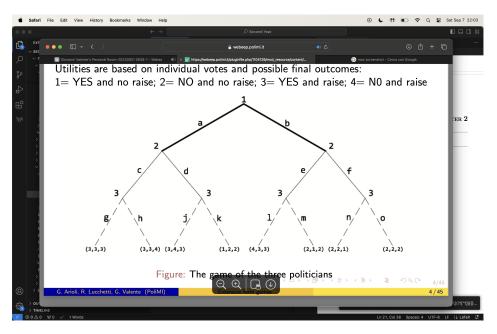


Figure 2.1: Voting game tree

This is an example of a game with perfect information, where each player is fully aware of all prior events.

#### Example:

Two players, Player 1 and Player 2, must decide sequentially whether to participate in a game. If both choose to play, a coin is flipped (random component R): Player 1 wins if it lands heads, and Player 2 wins if it lands tails. The corresponding decision tree is as follows:

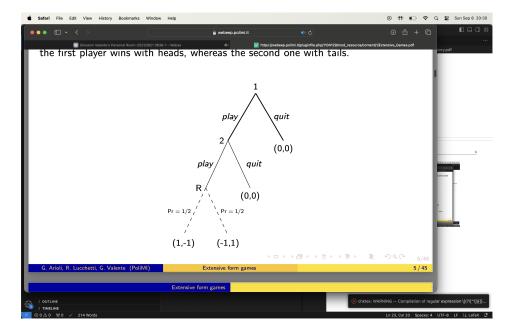


Figure 2.2: Chance game tree

**Definition** (Finite directed graph). A finite directed graph is a pair (V, E) where:

- V is a finite set, called the set of vertices.
- $E \subset V \times V$  is a set of ordered pairs of vertices, called the directed edges.

**Definition** (Path). A path from a vertex  $v_1$  to a vertex  $v_{k+1}$  is a finite sequence of vertices and edges  $v_1, e_1, v_2, \ldots, v_k, e_k, v_{k+1}$  such that  $e_i \neq e_j$  if  $i \neq j$  and  $e_j = (v_j, v_{j+1})$ .

The number k is called the length of the path.

**Definition** (Oriented graph). An oriented graph is a finite directed graph with no bidirectional edges. That is, for all vertices  $v_j$  and  $v_k$ , at most one of  $(v_j, v_k)$  and  $(v_k, v_j)$  can be an edge in the graph.

**Definition** (*Tree*). A tree is a triple  $(V, E, x_0)$  where (V, E) is an oriented graph, and  $x_0$  is a vertex in V such that there is a unique path from  $x_0$  to any other vertex  $x \in V$ .

**Definition** (Child). A child of a vertex v is any vertex x such that  $(v, x) \in E$ .

**Definition** (*Leaf*). A vertex is called a leaf if it has no children.

We say that the vertex x follows the vertex v if there is a path from v to x.

## 2.2 Extensive games

**Definition** (Extensive form game with perfect information). An extensive form game with perfect information consists of:

- 1. A finite set  $N = \{1, ..., n\}$  of players.
- 2. A game tree  $(V, E, x_0)$ .
- 3. A partition of the non-leaf vertices into sets  $P1, P2, \ldots, Pn + 1$ .
- 4. A probability distribution for each vertex in  $P_{n+1}$ , defined on the edges from that vertex to its children.

The game includes the following:

- 1. The set  $P_i$ , for  $i \leq n$ , consists of nodes where Player i must choose a child of v, representing a possible move by that player.
- 2.  $P_{n+1}$  is the set of nodes where chance plays a role (i.e., random events). Here, n+1 refers to the players plus the random component.  $P_{n+1}$  can be empty, indicating no random events in the game.
- 3. When  $P_{n+1}$  is empty, the *n* players only have preferences over the leaves, meaning no random component influences the outcome, so a utility function isn't required.

## 2.2.1 Extensive game solution

To determine the optimal outcome, we apply the axioms of rationality.

#### Example:

For the voting game described earlier:

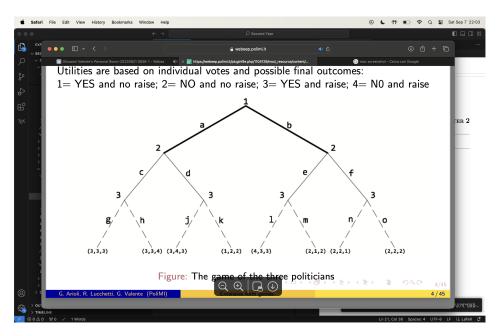


Figure 2.3: Voting game tree

We can identify the preferences as follows:

• For Player 3:

$$-h = (3 \ 3 \ 4) \text{ over } g = (3 \ 3 \ 3).$$

$$-j = (3 \ 4 \ 3) \text{ over } k = (1 \ 2 \ 2).$$

$$-l = (4 \ 3 \ 3) \text{ over } m = (2 \ 1 \ 2).$$

$$-o = (2 \ 2 \ 2) \text{ over } n = (2 \ 2 \ 1).$$

• For Player 2:

$$-d = (3 \ 4 \ 3) \text{ over } c = (3 \ 3 \ 4).$$

$$-e = (4 \ 3 \ 3) \text{ over } f = (2 \ 2 \ 2).$$

• For Player 1:

$$-b = (4 \ 3 \ 3) \text{ over } a = (3 \ 4 \ 3).$$

Thus, the optimal outcome is found.

**Definition** (*Length*). The length of a game is defined as the length of the longest path in the game tree.

Using decision theory and the assumption of rationality:

- Rationality assumption 5 allows us to solve games of length 1.
- Rationality assumption 4 allows us to solve games of length i + 1 if all games of length at most i have already been solved.

This iterative process is called backward induction, where we work backwards from the leaves of the tree to the root to determine the optimal sequence of actions.

2.3. Chess theorem

**Theorem 2.2.1** (First rationality theorem). The rational outcomes of a finite game with perfect information are those determined by the backward induction procedure.

Backward induction can be applied because every vertex v in the game is the root of a subgame consisting of all the vertices that follow v. These subgames are derived from the original game.

#### Example:

For the chance game:

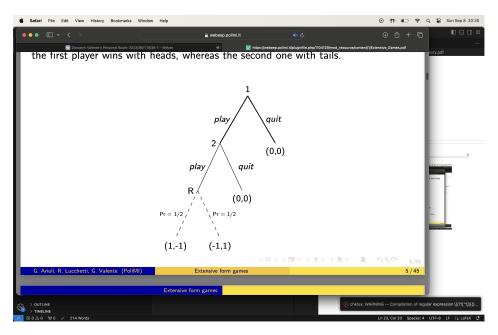


Figure 2.4: Chance game tree

The outcomes obtained through backward induction are  $(4\ 3)$  and  $(3\ 4)$ . Player 2 has no strict preference between  $(4\ 3)$  and  $(0\ 3)$ , indicating that in general, the solutions may not be unique.

## 2.3 Chess theorem

**Theorem 2.3.1** (Von Neumann). In the game of chess, one and only one of the following alternatives holds:

- 1. The White has a way to win, no matter what the Black does.
- 2. White has a strategy to guarantee a win, regardless of what Black does.
- 3. Both White and Black can force at least a draw, regardless of the opponent's actions.

*Proof.* Assume the game has a finite length of 2K moves, where each player makes K moves. Let  $a_i$  represent White's move at their i-th stage, and  $b_i$  represent Black's corresponding move. The first possibility in the theorem can be formulated as follows:

$$\exists a_1 : \forall b_1 \exists a_2 : \forall b_2 \dots \exists a_K : \forall b_K \implies \text{white wins}$$

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Now, suppose this is false. Then the negation is:

$$\forall a_1 \exists b_1 : \forall a_2 : \exists b_2 : \dots \forall a_K : \exists b_K \implies \text{white does not win}$$

This means Black has the possibility to prevent White from winning, ensuring at least a draw. If White does not have a winning strategy, then Black can secure at least a draw. Similarly, if Black does not have a winning strategy, then White can secure at least a draw. Therefore, if neither of the first two possibilities holds, the third one must be true.

#### 2.3.1 Extension

Von Neumann's theorem can be extended to any finite game of perfect information where the possible outcomes are either a win for one player or a tie.

Corollary 2.3.1.1. Consider a finite, perfect information game with two players, where the only possible outcomes are a win for one of the players or a tie. Then, exactly one of the following holds:

- 1. Player 1 has a winning strategy, no matter what the second player does.
- 2. Player 2 has a winning strategy, no matter what the second player does.

The possible solutions for a game are classified as follows:

- Very weak solution: the game has a rational outcome, but it is not accessible in practice, as with chess.
- Weak solution: the outcome of the game is known, but the method to achieve it is not generally understood.
- Solution: there exists an algorithm that can determine the outcome.

#### Example:

Consider the game of chomp. In this game we have a grid in which each player can remove a tile with all the others on the right and above it. In this game we have a solution if the grid is a square, and a weak solution if the grid is rectangular. In the latter case we may have the following situation:

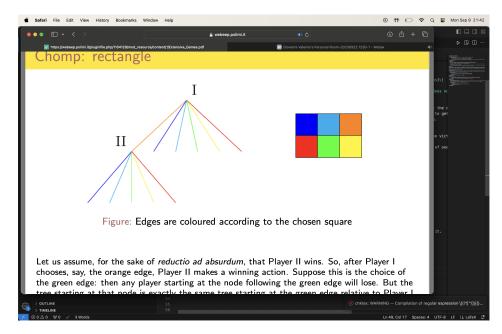


Figure 2.5: Rectangular chomp

Let us assume, for the sake of reductio ad absurdum, that Player II wins. So, after Player I chooses, say, the orange edge, Player II makes a winning action. Suppose this is the choice of the green edge: then any player starting at the node following the green edge will lose. But the tree starting at that node is exactly the same tree starting at the green edge relative to Player I. Thus Player I has a move available that guarantees victory, whereby Player II would have to lose. Since we derived a contradiction, the original assumption must be false: it is Player I that wins!

# 2.4 Impartial combinatorial games

**Definition** (*Impartial combinatorial game*). An impartial combinatorial game is a game such that:

- 1. There are two players moving in alternate order.
- 2. There is a finite number of positions in the game.
- 3. The players follow the same rules.
- 4. The game ends when no further moves are possible.
- 5. The game does not involve chance.
- 6. In the classical version, the winner is the player leaving the other player with no available moves, in the mis'ere version the opposite.

#### Example:

Examples of impartial combinatorial games are:

- k piles of cards. At her turn each player takes as many cards as she wants (at least one!) from one and only one pile.
- k piles of cards. At her turn each player takes as many cards as she wants (at least one) from no more than j < k piles.

• k cards in a row. At her turn each player takes either  $j_1$  or i or  $j_l$  cards.

In all these variants of the game, the player remaining without cards loses. In the first two cases the positions are  $(n_1, \dots, n_k)$  where  $n_i$  is a non negative integer for all i. In the last example positions can be seen as all non negative integers smaller or equal to k.

To solve this type of games we start by partitioning the set of all possible (finitely many) positions into two sets:

- 1. P-positions (i.e. previous player has a winning strategy): loosing.
- 2. N-positions (i.e. next player has a winning strategy): winning.

Note: it is the state of the game that matters, and not who is called to move. Rules for the partition (for the classical version):

- Terminal position  $(0,0,\cdots,0)$  is a P-position (it is a losing position, as the player does not have any card left)
- From a P-position only N-positions are available
- From a N-position it is possible, yet not necessary, to go to a P-position.

Therefore, the player starting from a N-position wins.

## 2.4.1 Nim game

The Nim game is defined as  $(n_1, \dots, n_k)$  where  $n_i$  is a positive integer for all i. At her turn any player is supposed to take one (and only one)  $n_i$  and substitute it with  $\hat{n}_i < n_i$ . The winner is the player who arrives at the position  $(0, \dots, 0)$ .

- Actions: taking away cards from one pile.
- Goal: to clear the whole table.

**Theorem 2.4.1** (Bouton). In the Nim game the position  $(n_1, n_2, ..., n_k)$  is a P-position if and only if:

$$n_1 \oplus n_2 \oplus \cdots \oplus n_k = 0$$

*Proof.* Terminal position  $(0,0,\ldots,0)$  is a P-position, with zero Nim-sum.

Positions with  $n_1 \oplus n_2 \oplus \cdots \oplus n_N = 0$  go only to positions with non-zero Nim-sum. For, suppose that the next position  $(\hat{n}_1, n_2, \dots, n_N)$  is such that  $\hat{n}_1 \oplus n_2 \oplus \cdots \oplus n_N = 0 = n_1 \oplus n_2 \oplus \cdots \oplus n_N$ : then, by the cancelation law one has  $\hat{n}_1 = n_1$ , which is impossible since the game requires  $\hat{n}_1 < n_1$ .

Positions with  $n_1 \oplus n_2 \oplus \cdots \oplus n_N \neq 0$  can go to positions with zero Nim-sum Let  $z := n_1 \oplus n_2 \oplus \cdots \oplus n_N \neq 0$ . Take a pile having 1 in the first column on the left of the expansion of z and put 0 there; then go right, leaving unchanged digits corresponding to 0 and changing them otherwise. Provably, the result is smaller than the original number.

#### Example:

From the following arrangement yielding non-zero Nim-sum: 4, 6, 5. by taking a card out of the first row we go to the next one with zero Nim-sum 3, 6, 5. So, there are three initial good moves, one for each row.

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#### 2.4.2 Conclusions

Games with perfect information can be solved by using backward induction. However backward induction is a concrete solution method only for very simple games, because of limited rationality. Depending on the game, we can reach different levels of solutions.

# 2.5 Strategies

In Backward induction a move must be specified at any node. Let  $P_i$  be the set of all the nodes where player i is called upon to make a move.

**Definition** (Pure strategy). A pure strategy for player i is a function defined on the set  $P_i$ , associating to each node v in  $P_i$  a child x, or equivalently an edge (v, x).

**Definition** (*Mixed strategy*). A mixed strategy is a probability distribution on the set of the pure strategies.

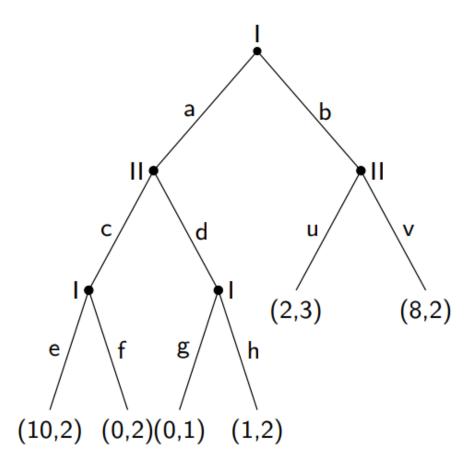
When a player has n pure strategies, the set of her mixed strategies is:

$$\sum_{n} := \left\{ p = (p_1, \dots, p_n) | p_i \ge 0 \text{ and } \sum_{i=1}^{n} p_i = 1 \right\}$$

 $\sum_{n}$  is the fundamental simplex in *n*-dimensional space.

#### Example:

Consider the following tree:



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| The strategies in | the | tree | are: |
|-------------------|-----|------|------|
|-------------------|-----|------|------|

|     | cu     | cv     | du    | dv    |
|-----|--------|--------|-------|-------|
| aeg | (10,2) | (10,2) | (0,1) | (0,1) |
| aeh | (10,2) | (10,2) | (1,2) | (1,2) |
| afg | (0,2)  | (0,2)  | (0,1) | (0,1) |
| afh | (0,2)  | (0,2)  | (1,2) | (1,2) |
| beg | (2,3)  | (8,2)  | (2,3) | (8,2) |
| beh | (2,3)  | (8,2)  | (2,3) | (8,2) |
| bfg | (2,3)  | (8,2)  | (2,3) | (8,2) |
| bfh | (2,3)  | (8,2)  | (2,3) | (8,2) |

Remember that the first player is in the rows, while the second is in the columns. All combinations are listed even if equivalent (e.g. strategies b.. for Player II) The table has repeated pairs: different strategies can lead to the same outcomes

- Extensive form: the different moves of the players are presented in sequence
- Strategic form: all strategies of the players are presented at the same time

**Theorem 2.5.1** (Von Neumann for strategies). In the chess game one of the following alternatives holds:

- 1. the white has a winning strategy.
- 2. the black has a winning strategy.
- 3. both players have a strategy leading them at least to a tie.

We have the first outcome if there is a raw with all winning elements. We have the second outcome if there is a column with all winning elements. The third outcome has a mixed outcomes, comprising also ties (but not all three outcomes in the same row or column).

If  $P_i = \{v_1, \ldots, v_k\}$  and  $v_j$  has  $n_j$  children, then the number of strategies of Player i is  $n_1 \cdot n_2 \cdot \cdots \cdot n_k$ . This shows that the number of strategies even in short games is usually very high.

#### Example:

If Tic-Tac-Toe is stopped after three moves, the first player has (without exploiting symmetries)  $9 \cdot 7^{(8 \times 9)}$  strategies

# 2.6 Games with imperfect information

Sometimes players must make moves at the same time, and so they cannot have full knowledge of each other's moves. This can be still represented with a tree.

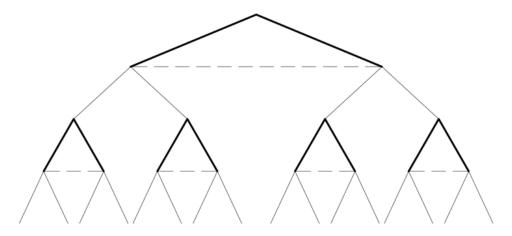


Figure 2.6: Tree with imperfect information

Dashed line: the player does not know exactly which vertex she finds herself in.

**Definition** (Information set). An information set for a player i is a pair  $(U_i, A(U_i))$  with the following properties:

- 1.  $U_i \subset P_i$  is a nonempty set of vertices  $v_1, \dots, v_k$
- 2. each  $v_i \in U_i$  has the same number of childre.
- 3.  $A_i(U_i)$  is a partition of the children of  $v_1 \cup \cdots \cup v_k$  with the property that each element of the partition contains exactly one child of each vertex  $v_i$

Accordingly, player i knows to be in  $U_i$ , but not in which vertex she is. The partition yields the choice function, meaning that each set in  $A_i(U_i)$  represents an available move for the player (graphically, it is the same choice, i.e. an edge, coming out of the different vertices).

**Definition** (Extensive form game with imperfect information). An Extensive form Game with imperfect information is constituted by:

- 1. A finite set  $N = \{1, \ldots, n\}$  of players.
- 2. A game tree  $(V, E, x_0)$ .
- 3. A partition made by sets  $P_1, P_2, \ldots, P_{n+1}$  of the vertices which are not leaves.
- 4. A partition  $(U_i^j)$ ,  $j = 1, \ldots, ki$  of the set  $P_i$ , for all i, with  $(U_i^j, A_i^j)$  information set for all players i for all vertices j (with the same number of children).
- 5. A probability distribution, for each vertex in  $P_{n+1}$ , defined on the edges going from the vertex to its children.
- 6. An *n*-dimensional vector attached to each leaf

Note that if the partition comprises just a single vertex, then a game with imperfect information becomes the same as a game with perfect information.

**Definition** (Pure strategy). A pure strategy for player i in an imperfect information game is a function defined on the collection  $\mathcal{U}$  of his information sets and assigning to each  $U_i$  in  $\mathcal{U}$  an element of the partition  $A(U_i)$ . A mixed strategy is a probability distribution over the pure strategies.

A game of perfect information is a particular game of imperfect information where all information sets of all players are singletons.

# Zero sum games

## 3.1 Introduction

**Definition** (*Zero sum game*). A two player zero sum game in strategic form is the triplet  $(X, Y, f: X \times Y \to \mathbb{R})$ . Here:

- X is the strategy space of Player 1.
- Y the strategy space of Player 2.
- f(x,y) is what Player 1 gets from Player 2, when they play x,y respectively.

Given that f is the utility function of Player 1, by definition of zero sum games the utility function g of Player 2 must be:

$$q = -f$$

In the finite case  $X = \{1, 2, ..., n\}$ ,  $Y = \{1, 2, ..., m\}$  the game is described by a payoff matrix P, wherein Player 1 selects row i while Player 2 selects column j:

$$\begin{pmatrix} p_{11} & \cdots & p_{1m} \\ \cdots & p_{ij} & \cdots \\ p_{n1} & \cdots & p_{nm} \end{pmatrix}$$

Here,  $p_{ij}$  is the payment of Player 2 to Player 1 when they play i, j respectively. In order to choose the optimal strategy, each player can reason as follows:

- Player 1 can guarantee herself to get at least  $v_1 = \max_i \min_i p_{ii}$ .
- Player 2 can guarantee himself to pay at most  $v_2 = \min_i max_i p_{ij}$ .

 $v_1$  and  $v_2$  are said to be the conservative values of Player 1 and Player 2, respectively.

#### Example:

Consider the game:

$$\begin{pmatrix} 4 & 3 & 1 \\ 7 & 5 & 8 \\ 8 & 2 & 0 \end{pmatrix}$$

For the player 1 we pick the minimum for each row, that is: (1 5 0) and then we choose

3.1. Introduction

the maximum between them. Thus, the conservative value for the player 1 is  $v_1 = 5$ .

For the player 2 we pick the maximum for each column, that is:  $(8 \ 5 \ 8)$  and then we choose the minimum between them. Thus, the conservative value for the player 2 is  $v_2 = 5$ . Accordingly, the rational outcome is 5 and the rational behavior is  $(\bar{i} = 2, \bar{j} = 2)$ .

## 3.1.1 Rationality in zero sum games

Let us suppose the following:

- $v_1 = v_2 := v$ ,
- $\bar{i}$  the row such that  $p_{\bar{i}\bar{j}} \geq v_1 = v$  for all j
- $\bar{j}$  the column such that  $p_{\bar{i}\bar{j}} \leq v_2 = v$  for all i

Then  $p_{i\bar{j}} = v$  and  $p_{i\bar{j}} = v$  is the rational outcome of the game

 $\bar{i}$  is an optimal strategy for Pl1, because she cannot get more than  $v_2$ , since  $v_2$  is the conservative value of the second player  $\bar{j}$  is an optimal strategy for Pl2, because he cannot pay less than  $v_1$ , since  $v_1$  is the conservative value of the first player  $\bar{i}$  maximizes the function  $\alpha(i) = \min_j p_{ij}$ ;  $\bar{j}$  minimizes the function  $\beta(j) = \max_i p_{ij}$ .

## 3.1.2 Extension to arbitrary games

Let the triplet  $(X, Y, f : X \times Y \to \mathbb{R})$  represent a game between two players, wherein their respective strategy spaces X and Y may not be finite sets. For the sake of rational behaviour, the players can guarantee to themselves the following outcomes:

- Player 1:  $v_1 = \sup_x \inf_y f(x, y)$
- Player 2:  $v_2 = \inf_y \sup_x f(x, y)$

The outcomes  $v_1$ ,  $v_2$  are the conservative values of the players. If  $v_1 = v_2$ , we set  $v = v_1 = v_2$  and we say that the game has value v.

**Optimality** Let X and Y be arbitrary sets. Suppose:

- 1.  $v_1 = v_2 := v$
- 2. there exists strategy  $\bar{x}$  such that  $f(\bar{x},y) > v$  for all  $y \in Y$
- 3. there exists strategy  $\bar{y}$  such that  $f(x,\bar{y}) \leq v$  for all  $x \in X$

(the last two conditions are needed if the sets are infinite and not compact). Then:

- v is the rational outcome of the game.
- $\bar{x}$  is an optimal strategy for Pl1.
- $\bar{y}$  is an optimal strategy for Pl2.

Observe  $\bar{x}$  is optimal for Pl1 since it maximizes the function  $\alpha(x) = \inf_y f(x, y) \bar{y}$  is optimal for Pl2 since it minimizes the function  $\beta(y) = \sup_x f(x, y)$  where  $\alpha(x)$  is the value of the optimal choice of Pl2 if he knows that Pl1 plays x, and  $\beta(y)$  is the value of the optimal choice of Pl1 if she knows that Pl2 plays y.

3.1. Introduction

**Proposition.** Let X,Y be nonempty sets and let  $f:X\times Y\to\mathbb{R}$  be an arbitrary real valued function. Then:

$$v_1 = \sup_{x} \inf_{y} f(x, y) \le \inf_{y} \sup_{x} f(x, y) = v_2$$

*Proof.* By definition, for all x, y:

$$\inf_{y} f(x,y) \le f(x,y) \le \sup_{x} f(x,y)$$

Thus:

$$\alpha(x) = \inf_{y} f(x, y) \le \sup_{x} f(x, y) = \beta(y)$$

Since for all  $x \in X$  and  $y \in Y$  it holds that

$$\alpha(x) \le \beta(y)$$

it follows that

$$\sup_{x} \alpha(x) \le \inf_{y} \beta(y)$$

As a consequence, in every game  $v_1 \leq v_2$ .

#### Example:

Consider the rock, scissors, and paper game:

$$\begin{pmatrix}
0 & 1 & -1 \\
-1 & 0 & 1 \\
1 & -1 & 0
\end{pmatrix}$$

The conservative values are not the same: in fact,  $v_1 = -1$  and  $v_2 = 1$ .

There is no winning strategy since each player always plays randomly. But if the game is repeated many times, the rational solution for both players is to play each option one-third of the times, so that in the long run their expected utility is zero. By extending the game with mixed strategies, both conservative values become 0.

Case  $v_1 < v_2$  Consider the case of mixed strategies with a gmae with an  $n \times m$  utility matrix P. In this case the strategy spaces are defined as:

$$\sum_{k} = \left\{ x = (x_1, \dots, x_k) | x_i \ge 0 \text{ and } \sum_{i=1}^{k} x_i = 1 \right\}$$

with k = n for Pl1 and k = m for Pl2. The utility function is:

$$f(x,y) = \sum_{i=1,...,n,j=i,...,m} x_i y_j p_{ij} = (x, Py)$$

Here  $p_{ij}$  is an element of P corresponding to the utility of Pl1 when she plays row i and Pl2 plays column j (of course the utility of Pl2 is just the opposite). Thus, the mixed extension of the initial game is

$$\left(\sum_{n}, \sum_{m}, f(x, y) = (x, Py)\right)$$

# Mathematical concepts

# 4.1 Binary sum

Define an operation  $\oplus$  on  $\mathbb{N} = \{0, 1, 2, \dots, n, \dots\}$  in the following way: for  $n_1, n_2 \in \mathbb{N}$ :

- 1. Write  $n_1$ ,  $n_2$  in binary form, denoted by  $[n_1]_2$ ,  $[n_2]_2$ .
- 2. Write the sum  $[n_1]_2 \oplus [n_2]_2$  in binary form where  $\oplus$  is the (usual) sum, but without carry.
- 3. What you get is the result in binary form.

#### Example:

The  $\oplus$  operation applied to 1,2,4, and 1. In binary they are 001 010, 100, and 001. The sum is 110, thus 6.

# 4.2 Group

**Definition** (Group). A nonempty set A with a binary operation  $\cdot$  defined on it is called a group provided that:

- 1. for  $a, b \in A$  the element  $a \cdot b \in A$
- 2.  $\cdot$  is associative:  $(a \cdot b) \cdot c = a \cdot (b \cdot c)$
- 3. there is a (unique) element e, called identity, such that  $a \cdot e = e \cdot a = a$  for all  $a \in A$
- 4. for every  $a \in A$  there is  $b \in A$  such that  $a \cdot b = b \cdot a = e$ : such an element is unique and called inverse of a

If  $a \cdot b = b \cdot a$  for all  $a, b \in A$  the group is called abelian.

#### Example:

Examples of abelian groups: The integers  $\mathbb{Z}$ , equipped with the usual sum The real numbers excluded 0, equipped with the usual product

Examples of non-abelian groups: The  $n \times n$  matrices with non-zero determinant, equipped with the usual product

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**Proposition.** Let  $(A, \cdot)$  be a group. Then the cancellation law holds:

$$a \cdot b = a \cdot c \implies b = c$$

*Proof.* By multiplying by  $a^{-1}$  both sides of the equation  $a \cdot b = a \cdot c$ , one obtains  $a^{-1}a \cdot b = a^{-1}a \cdot c$ . Then, insofar  $a^{-1}$  is the inverse of a, this expression reduces to  $e \cdot b = e \cdot c$ , which by the property of the identity e is exactly equal to b = c.

**Proposition.** The set of the natural numbers with operation  $\oplus$  is an abelian group.

*Proof.* The identity element is of course 0. The inverse of n is n itself. Associativity and commutativity of  $\oplus$  are easy to show.

Therefore, the cancellation law holds:  $n_1 \oplus n_2 = n_1 \oplus n_3 \implies n_2 = n_3$ .

## 4.3 Convexity

**Definition** (*Convex set*). A set  $C \subset \mathbb{R}^n$  is convex just in case for any  $x, y \in C$ , provided  $\lambda \in [0, 1]$ , one has:

$$\lambda x + (1 - \lambda)y \in C$$

The intersection of an arbitrary family of convex sets is convex. A closed convex set with nonempty interior coincides with the closure of its internal points.

**Definition** (Convex combination). We shall call a convex combination of elements  $x_1, \ldots, x_n$  any vector x of the form

$$x = \lambda_1 x_1 + \dots + \lambda_n x_n$$

with 
$$\lambda_1 \geq 0, \dots, \lambda_n \geq 0$$
 and  $\sum_{i=1}^n \lambda_i = 1$ 

**Proposition.** A set C is convex if and only if for every  $\lambda_1 \geq 0, \ldots, \lambda_n \geq 0$  such that  $\sum_{i=1}^n \lambda_i = 1$ , for every  $c_1, \ldots, c_n \in C$ , for all n, then  $\sum_{i=1}^n \lambda_i c_i \in C$ .

If C is not convex, then there is a smallest convex set containing C: it is the intersection of all convex sets containing C.

**Definition** (Convex hull). The convex hull of a set C, denoted by co C, is:

$$\operatorname{co}\, C = \bigcap_{A \in \mathcal{C}} A$$

where  $\mathcal{C} = \{A | C \subset A \text{ and } A \text{ is convex}\}.$ 

**Proposition.** Given a set C, then

co 
$$C = \left\{ \sum_{i=1}^{n} \lambda_i c_i | \lambda_i \ge 0, \sum_{i=1}^{n} \lambda_i = 1, c_i \in C \quad \forall i, n \in N \right\}$$

The convex hull of any set C is the set of all convex combinations of points in C. When C is a finite collection of points, its convex hull co C is called a polytope. E.g. if C contains three points, co C is the triangle (i.e. a polygone) with such points at its angles, which includes also all the points inside

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**Theorem 4.3.1.** Given a closed convex set C and a point x outside C, there is a unique element  $p \in C$  such that, for all  $c \in C$ :

$$||p - x|| \le ||c - x||$$

The projection p is characterized by the following properties:

1.  $p \in C$ 

2. 
$$(x-p,c-p) \leq 0$$
 for all  $c \in C$ 

That is, p is the closest point to x belonging to the set C, and by the last property it forms an obtuse angle between x and any  $c \in C$ .

**Theorem 4.3.2.** Let C be a convex proper subset of the Euclidean space  $\mathbb{R}^l$  and assume  $\bar{x} \in cl\ C^c$ . Then there is an element  $0 \neq x^* \in \mathbb{R}^l$  such that,  $\forall c \in C$ :

$$(x^*, c) \ge (x^*, \bar{x})$$

This theorem means that a criterion to tell apart an external point from any internal point.

*Proof.* Suppose  $\bar{x} \notin cl\ C$  and call p its projection on cl C. Based on the previous theorem, in follows that  $(\bar{x}-p,c-p) \leq 0$  for all  $c \in C$ . By setting  $x^* := p - \bar{x} \neq 0$ , the inequality becomes  $(-x^*,c-\bar{x}-x^*) = (-x^*,-x^*) + (-x^*,c-\bar{x}) \leq 0$ , that is  $(x^*,c-\bar{x}) \geq ||x^*||^2$ . Then, since  $||x^*||^2 > 0$ , by linearity one obtains

$$(x^*, c) \ge (x^*, \bar{x}) \quad \forall c \in C$$

As  $x^*$  appears in both sides, by renormalization we can choose  $||x^*|| = 1$ . If  $\bar{x} \in cl\ C \setminus C$ , take a sequence  $\{x_n\} \subset C^c$  such that  $x_n \to \bar{x}$ . From the first step of the proof, we can find some norm one  $x_n^*$  such that

$$(x_n^*, c) \ge (x_n^*, x_n) \quad \forall c \in C$$

So, given that for some sub-sequence one has  $x_n^* \to x^*$ , taking the limit of the above inequality yields

$$(x^*, c) \ge (x^*, \bar{x}) \quad \forall c \in C$$

**Corollary 4.3.2.1.** Let C be a closed convex set in a Euclidean space, let x be on the boundary of C. Then there is a hyperplane containing x and leaving all of C in one of the halfspaces determined by the hyperplane.

Such an hyperplane is said to be an hyperplane supporting C at x

Corollary 4.3.2.2. Let C be a closed convex set in a Euclidean space. Then C is the intersection of all half-spaces containing it.

**Theorem 4.3.3.** Let A, C be closed convex subsets of  $\mathbb{R}^l$  such that intA is nonempty and  $intA \cap C = \emptyset$ . Then there are  $0 \neq x^*$  and  $b \in \mathbb{R}$  such that,  $\forall a \in A, \forall c \in C$ :

$$(x^*, a) \ge b \ge (x^*, c)$$

This means a criterion to determine whether a point is in A or in C.

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*Proof.* Since  $\bar{x} = 0 \in (\text{int}A - C)^c$ , from the previous separation theorem with  $\bar{x} = 0$  there is  $x^* \neq 0$  such that

$$(x^*, x) \ge 0 \quad \forall x \in \text{int} A - C$$

Thus, for x = a - c by linearity we obtain

$$(x^*, a) \ge (x^*, c) \quad \forall a \in \text{int} A, \forall c \in C$$

By extension this implies

$$(x^*, a) \ge (x^*, c) \quad \forall a \in \text{cl int} A = A, \forall c \in C$$

 $H = \{x : (x^*, x) = b\}$  is called the separating hyperplane: A and C are contained in the two different half-spaces generated by H.