# Algorithmic Game Theory

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### 25. Juli 2023

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# 1 Random Definitions

- weakly solved: outcome if both players play perfect
- strongly solved: entire game tree/graph is known

## 2 Non-Cooperative-Games

Normalform G = (P, S, u)

- Players:  $P = \{1, 2, ..., n\}$
- Strategies:  $S = (S_1, S_2, S_n)$
- Utility functions:  $u = (u_1, u_2, ..., u_n)$  with  $u_i : S \to \mathbb{R}$

Note: Only **one decision** is made (choice of strategy) and moves are **simultaneous** Strategy Profile  $s = (s_1, s_2, ...s_n) \in S_1 \times S_2 \times ... \times S_n = S$ 

Weakly Dominant Strategies ( $\leq$ ): The strategy has the best payoff regardless of the choices of the other players.

**Strictly Dominant Strategies** (<): The strategy has the greatest payoff among all other strategies. (there can only be at most one such strategy)

### 2.1 pure and mixed strategies

#### pure strategies:

every player chooses the same action in a given situation (like a spreadsheet on what to do). Each move is deterministic. Example: Play Paper every time in Rock-Paper-Scissors

#### mixed strategies:

use probabilities to determine a pure strategy for each situation. Each move is non-deterministic. Example: choose Rock, Paper, Scissors with p=1/3 each.

- Models that players confuse each other by switching
- adds uncertainty about actions of others
- models repeated play and population dynamic

#### 2.2 Pareto

- s weakly Parteo-dominates t:  $\forall i \in I : u_i(s) \ge u_i(t)$ . payoff of s is at least as good as for t for each utility function
- s Pareto-dominates t:  $\exists j \in I : u_j(s) > u_j(t)$ . payoff of s is better than for t for a given utility function

- s strongly Pareto-dominates t:  $\forall i \in I : u_i(s) > u_i(t)$ . payof of s is better than for t for all utility functions
- **t is Pareto-optimal**:  $\neg \exists s \in S$  : s Pareto-dominates t
- **t is weakly Pareto-optimal**:  $\neg \exists s \in S$ : s stongly Pareto-dominates t

In a Pareto optimum no Player can gain switch strategies without another player being worse off

**Example**: (S, S), (S, C), (C, S) are Pareto-optimas for the prisoners dilemma. (S, S) strongly dominates (C, C).

All strategies in rock paper scissors are pareto-optimal

**Best Response**: A Strategy is a best response to another strategy iff it produces the best payoff for the player given the strategies of the other players Note: dominant strategy  $\rightarrow$  best response but not  $\leftarrow$ 

#### **2.3** Nash

#### pure:

- Nash equilibrium in pure strategies:  $\forall i \in I : s_i \text{ is a best response to } s_{-1}$
- strict Nash equilibrium in pure strategies:  $\exists ! s : s$  is a best response to  $s_{-1}$   $s_{-i}$  is the strategy profile s without the strategy of player i

The prisoners dilemma has (C,C) as the single pure Nash equilibrium where every player plays his dominant strategies.

There can be none, one or multiple such equilibria

Note: The pure Nash equilibria can be computed in **PTIME** by exhaustive search for best response

#### 2.3.1 Nashs Theorem

For a non-cooperative game G with finite Players, finite Strategies there exists a Nash equilibrium in mixed strategies

#### **Proof sketch:**

- Model strategies as unit vectors in  $\mathbb{R}^{|S|}$
- Mixed strategies are points on a simplex in this vector space
- define some function  $f:\Pi\to\Pi$  ( $\Pi$  is the set of all mixed strategy profiles)
- use Brouwers fixpoint theorem to show that f has at least one fixpoint

#### 2.3.2 Computation

- translate problem intro mixed integer programming
- solve MIP (NP-complete)
- Given a Nash equilibrium for a game, finding the next equilibrium is FNP-compete
  - Functional Complexity: Functions that given input x can be compute some y in some Time/Space
  - FP and FNP are functional P and NP

#### 2.4 Prisoners Dilemma

- $P = \{1, 2\}$
- $S_1 = S_2 = \{S, C\}$
- $S = \{(S, S), (S, C), (C, S), (C, C)\}$
- $u_1 = \{(S, S) \mapsto 4, (S, C) \mapsto 0, (C, S) \mapsto 5, (C, C) \mapsto 3\}$
- $u_2 = \{(S, S) \mapsto 4, (S, C) \mapsto 5, (C, S) \mapsto 0, (C, C) \mapsto 3\}$

# **3** Sequential Games

- Several decisions
- Strategies can be seen as advice
- Players move sequentially
- can be represented using a Graph (maybe even Tree) (e.g. in Chess board state diagram)

Complete Information: All players know possible moves and utilities of all players

#### 3.1 Zermelos Theorem

Every finite sequential game tree has a solution (strategy profile and utility) that can be obtained by backward induction

#### **Backwards Induction**

• start at the last round of the game (only one decision is left)

- determine for each of those nodes (in this depth) the best choice for the current player (and delete the rest)
- move one layer up and repeat till you reach root (only one path is left)

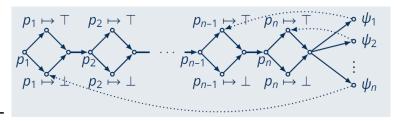
### 4 Combinatorial Games

Can be solved using backwards induction

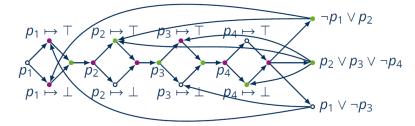
- Two Players
- Either One Player wins (1) and the other loses (-1) or its a draw (0)
- zero-sum-game

#### Geography is PSPace-complete

- $\in PSapce$ : use DFS on game tree and only keep current path in memory (depth is poly)
- *hardness*: by reduction to TrueQBF:
  - $-\varphi = \exists p_1 \forall p_2 \cdots \psi \text{ with } \psi = \psi_1 \wedge \cdots \wedge \psi_m$



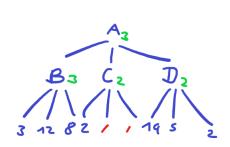
- 'diamonds' select for each prop-var if its true/false and in the end formula gets checked by edges to diamond nodes
- Example:



### 5 Minimax Tree Search

used to solve zero-sum games

- there is a min and a max player
- if min players turn select child node with min value (analogous for max)
- min and max switch in each layer
- Complexity:  $O(b^d)$  (b: branching factor, d: depth), hence its impractical for larger games
- $\rightarrow$  solution: alpha beta pruning:
  - stop search when when we know that a nodes does not gets visited anyway
  - Example:



- because B is already 3 we can stop at C=2 because Max does not care anyway
- we have to search all of D because D=2 is the right most element (unlucky)

- better heuristic helps
- Worst-case complexity:  $O(b^d)$  (all nodes have to be searched (because they are on the right))
- Best-case complexity:  $O(\sqrt{b}^d)$  (left nodes terminates search already)

### 6 Monte Carlo

Algorithms that use randomness and whose output may be incorrect with a certain probability

#### 6.1 Tree Search

- Simulate random moves from current to terminal states
- do statistics on moves that lead to wins
- Playout: complete move sequence from a state to a terminal state
- playout policy: bias simulation towards optimal play
- selection policy: determins where the simulation starts (exporations vs. exploitation)
- Steps:
  - Selection
  - Expansion
  - Simulation
  - Backpropagation

#### **6.2** Selection Policies

**UTC** (Upper confidence bounds applied to trees)

- ranks moves in terms of their upper confidence bound values
- $UCB1(n) = \frac{U(n)}{N(n)} + c \cdot \sqrt{\frac{\ln N(n')}{N(n)}}$
- n': unique parten of n in the tree
- U(n): total utility of node n
- N(m): total number of playout through node m
- c: some constant chosen empirically (for balance exploitation vs. exploration)

# 7 Games with missing information

#### **Extensive Form Game**

- $P = \{1, 2, ..., n\}$  Set of players (min. 2)
- n+1 tuple  $M = (M_1, ..., M_n, M_{nature})$  of Moves
- H Set of histories (seq. of moves)
- $T \subseteq H$  terminal histories
- $J_1,...,J_k=H\setminus T$  Partition into information sets s.t. all  $h_1,h_2\in J_i$  have the same legal moves
- $\{1,...,k\} \rightarrow P \cup \{Nature\}$  player function
- $u=(u_1,...,u_n)$  utility functions s.t.  $u_i:T\to\mathbb{R}$

#### **Information Sets**

• the turn player does not have the information to distinguish between states in the set (but from other sets)