

# Artificial Intelligence

## 3. Adversarial Search

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# Szenario

- Multitagent environments
  - Cooperative
  - Competitive
- Contingencies (unpredictability of other agents)
- Agents goals are in conflict:  
adversarial search (game)

# Game Theory

- Branch of mathematic/economics
- Game
  - Multiagent environment, significant impact on each other
- AI: special kind of games
  - Deterministic
  - Turn-taking (alternating agents)
  - Two player
  - Zero sum (utility values are equal but opposite)
  - Perfect information (full observable)

# Game as search problem

- Initial state
  - Successor function (legal moves)
  - Terminal test (terminal states  $\approx$  goal states)
  - Utility function
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- Two players: MAX, MIN
  - Game tree: half-move = ply

# MINIMAX-ALGORITHM

- $\text{MINIMAX-VALUE}(n) =$ 
  - =  $\text{UTILITY}(n)$ , if  $n$  is a terminal state
  - =  $\max(\text{MINIMAX-VALUE of the successors of } n)$ , if  $n$  is a MAX node
  - =  $\min(\text{MINIMAX-VALUE of the successors of } n)$ , if  $n$  is a MIN node
- Recursive counting of the minimax-value
- Gives optimal decision in games
- Space Complexity of the algorithm:  $O(m)$
- Time Complexity of the algorithm:  $O(b^m)$

# Optimal decisions in multiplayer games

- Modified MINIMAX-algorithm
  - Vector of utilities
  
- Collaboration, Alliances

# Alpha-Beta Pruning

- $\alpha$  = best choice we have found so far for MAX
- $\beta$  = best choice we have found so far for MIN
- Subtrees not improving the utility are not visited
- Reduces complexity
  - „ideal“ ordering of child-nodes:  $O(b^{m/2})$
  - random ordering:  $O(b^{3m/4})$
  - Chess: the „ideal“ complexity almost reachable

# Real-time Decisions

- Evaluation functions
  - Expected value of utility
  - Features
  - $\text{EVAL}(s) = w_1f_1(s) + w_2f_2(s) + \dots + w_nf_n(s)$
- Cutting off search
  - At a given depth  $d$
  - Quiescene search
  - Horizon effect
  - Singular extensions
  - Forward pruning



# Games with element of chance

- $\text{EXPECT-MINIMAX-VALUE}(n) =$ 
  - =  $\text{UTILITY}(n)$ , if  $n$  is a terminal state
  - =  $\max(\text{MINIMAX-VALUE of the successors of } n)$ , if  $n$  is a MAX node
  - =  $\min(\text{MINIMAX-VALUE of the successors of } n)$ , if  $n$  is a MIN node
  - =  $\sum P(s) \text{ EXPECT-MINIMAX-VALUE}(s)$ , for all successors  $s$  of  $n$ ,  
if  $n$  is a chance node