Deterministic Games

- States: S (starts at s_0)
- Players: P = 1...N (usually take turns)
- Actions: A (may depend on player / state)
- Transition function: $SxA \longrightarrow S$
- Terminal Test: $S \longrightarrow \{\text{true}, \text{false}\}$
- Terminal Utilities: $SxP \longrightarrow R$
- Solution for a player is a policy: $S \longrightarrow A$

Zero-Sum Games

- Agents have opposite utilities (values on outcomes)
- Think of a single value that one maximizes and the other minimizes
- Adversarial, pure competition

Solving Zero-Sum Games

- Have to think about how the other agent will respond to a move
- Each state has a value which is the best possible outcome at that state
 - In a tree, it is the highest value of all child states
- terminality $\Longrightarrow V(s) \stackrel{.}{=} known$
- nonterminal $\Longrightarrow V(s) = \max(V(s') \, \forall \, s' \in \operatorname{successors}(s))$
 - The opponent tries to minimize V(s) instead of maximize it

Adversarial Search (Minimax)

• Minimax value := the best achievable utility against a rational (optimal) adversary

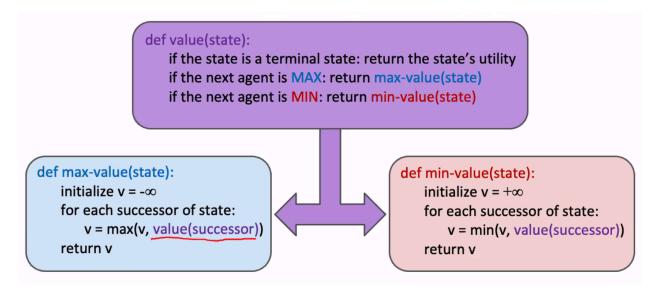


Figure 1: Screenshot_2023-09-12_at_5.42.27_PM.png

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- Optimal against a perfect player but not imperfect players
- b := branching factor
- m := maximum depth of game

- Time $\in O(b^m)$
- Space $\in O(bm)$

Alpha-Beta Pruning

- Used to minimize excess calculation
- Once you have a possible value for a parent node, only evaluate children if their value could be propagated with optimality
- Keep track of α in maximizer nodes; if the value $\leq \alpha$ for a node, ignore node and children
- Keep track of β in minimizer nodes; if the value $\geq \beta$ for a node, ignore the node and children

```
α: MAX's best option on path to root
                                β: MIN's best option on path to root
def max-value(state, \alpha, \beta):
                                                            def min-value(state, \alpha, \beta):
                                                                 initialize v = +\infty
    initialize v = -\infty
    for each successor of state:
                                                                 for each successor of state:
         v = max(v, value(successor, \alpha, \beta))
                                                                      v = min(v, value(successor, \alpha, \beta))
         if v \ge \beta return v
                                                                      if v \le \alpha return v
         \alpha = \max(\alpha, v)
                                                                      \beta = \min(\beta, v)
    return v
                                                                 return v
```

Figure 2: Screenshot 2023-09-12 at 6.02.01 PM.png

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- Has no effect on the minmax value computed for the root
- Values of intermediate nodes might be wrong
- Good child ordering improved effectiveness of pruning
- Time with perfect ordering $\in O(b^{m/2})$
 - Doubles solvable depth for a given computational power
- Simple example of metareasoning (computing about what to compute)

Depth-limited Search

- Search only to a limited depth in the tree
- Replace terminal utilities with an evaluation function for non-terminal positions
- Can get in an infinite loop if evaluation function is not specific enough
 - Usually a weighted linear sum of features

General Games

- Agents have independent utilities (values on outcomes)
- Cooperation, indifference, competition, and more are all possible