

Game playing

- We will focus on two player, perfect information zero-sum games:
 - perfect information — no hidden state
 - zero-sum — what is good for one player is bad for the other
- We will use search to analyze these games...
- But this search is different because the two players alternate and have different objectives.
- Assume both players will always make the best move
- Key question: What is the best move?

MINIMAX search

MINIMAX(n)

1. return MAX-PLAYER(n)

MAX-PLAYER(n)

1. if game corresponding to node n is over,
return value
2. find the children C of node n
3. for each child $c_i \in C$, let $v_i = \text{MIN-PLAYER}(c_i)$
4. return maximum v_i

MIN-PLAYER(n)

1. if game corresponding to node n is over,
return value
2. find the children C of node n
3. for each child $c_i \in C$, let $v_i = \text{MAX-PLAYER}(c_i)$
4. return minimum v_i

MINIMAX search with alpha-beta pruning

AB/MINIMAX(n)

1. return AB/MAX-PLAYER($n, -\infty, \infty$)

AB/MAX-PLAYER(n, α, β)

1. if game is over or depth cutoff reached, evaluate game state and return value.
2. find the children C of node n
3. for each child $c_i \in C$
 - $\alpha \leftarrow \max(\alpha, \text{AB/MIN-PLAYER}(c_i, \alpha, \beta))$
 - if $\alpha \geq \beta$, return β
4. return α

AB/MIN-PLAYER(n, α, β)

1. if game is over or depth cutoff reached, evaluate game state and return value.
2. find the children C of node n
3. for each child $c_i \in C$
 - $\beta \leftarrow \min(\beta, \text{AB/MIN-PLAYER}(c_i, \alpha, \beta))$
 - if $\alpha \geq \beta$, return α
4. return β