Game Playing Part 1 Minimax Search

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Sadly, not these games (not in this course) ...













Overview

- two-player zero-sum discrete finite deterministic game of perfect information
- Minimax search
- Alpha-beta pruning
- Large games
- two-player zero-sum discrete finite NON-deterministic game of perfect information

Two-player zero-sum discrete finite deterministic games of perfect information

Definitions:

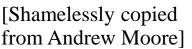
- Zero-sum: one player's gain is the other player's loss.
 Does not mean fair.
- Discrete: states and decisions have discrete values
- Finite: finite number of states and decisions
- Deterministic: no coin flips, die rolls no chance
- Perfect information: each player can see the complete game state. No simultaneous decisions.



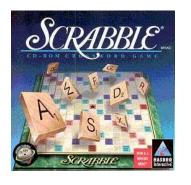












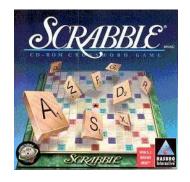




















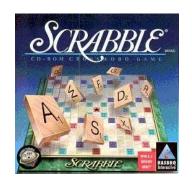
[Shamelessly copied from Andrew Moore]



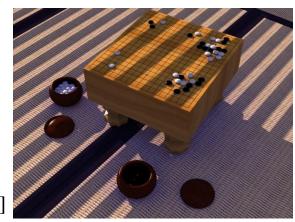


Zero-sum: one player's gain is the other player's loss. Does not mean *fair*.

Discrete: states and decisions have discrete values









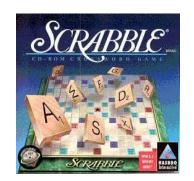


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Discrete: states and decisions have discrete values

Finite: finite number of states

and decisions







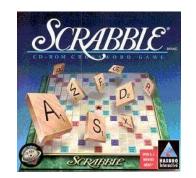


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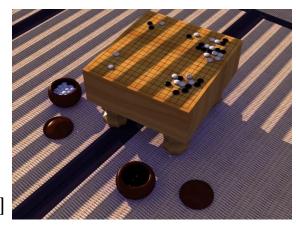
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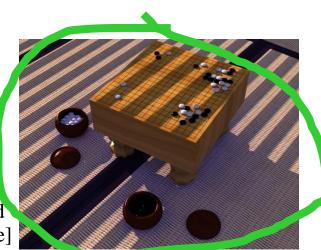
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II-Nim: Max simple game

- There are 2 piles of sticks. Each pile has 2 sticks.
- Each player takes one or more sticks from one pile.
- The player who takes the last stick loses.

(ii, ii)

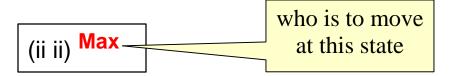
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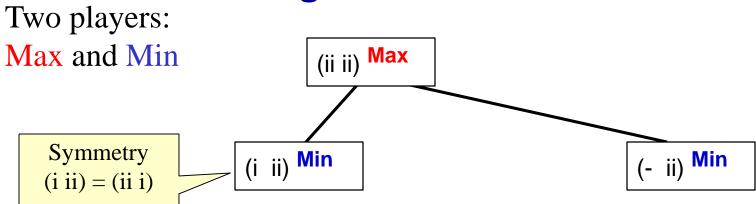
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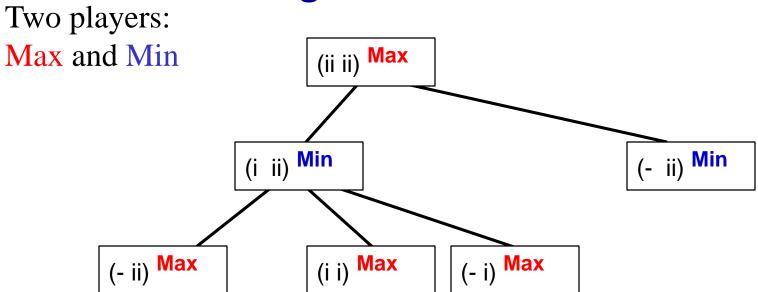
- Two players: Max and Min
- If Max wins, the score is +1; otherwise -1
- Min's score is –Max's
- Use Max's as the score of the game

Two players: Max and Min



Convention: score is w.r.t. the first player Max. Min's score = - Max





Two players:

Max and Min

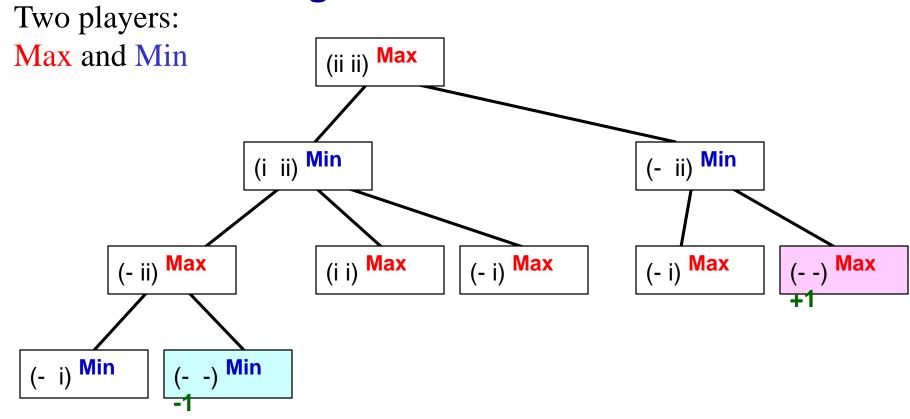
(i ii) Max

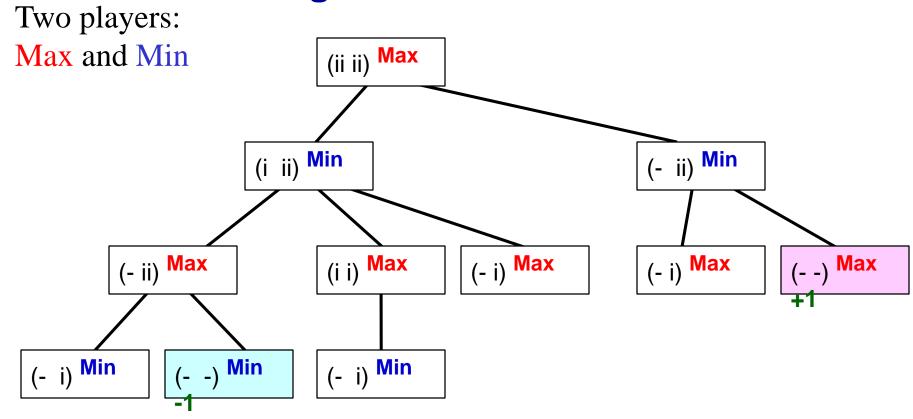
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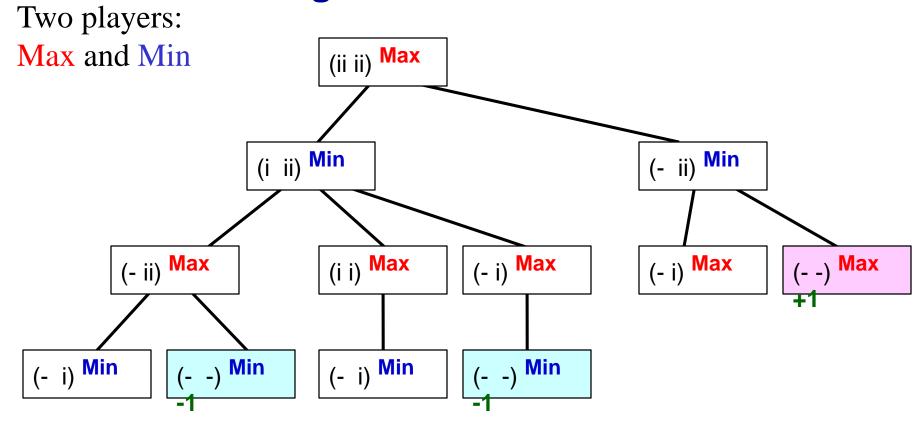
(- i) Max

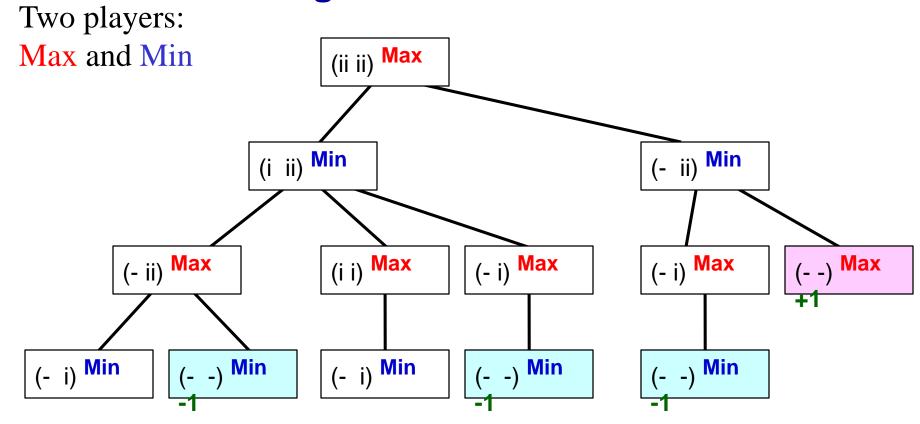
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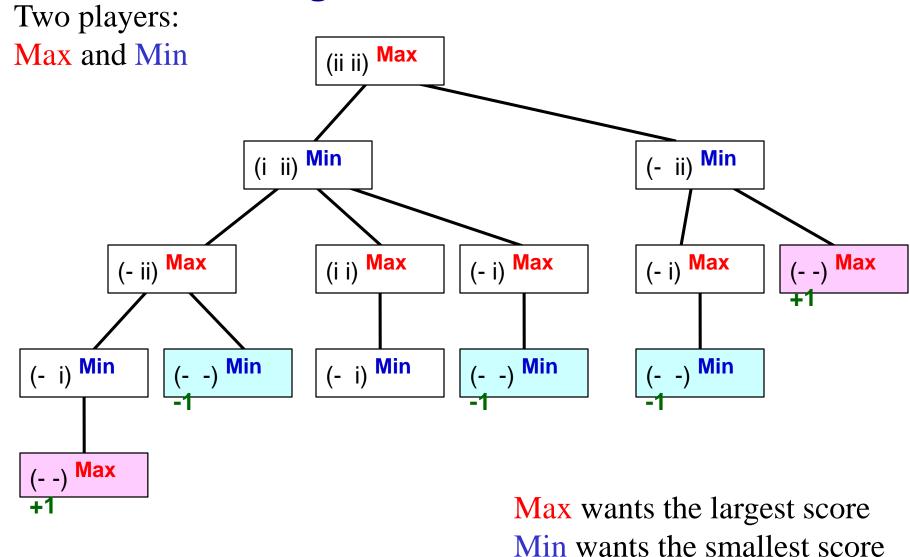
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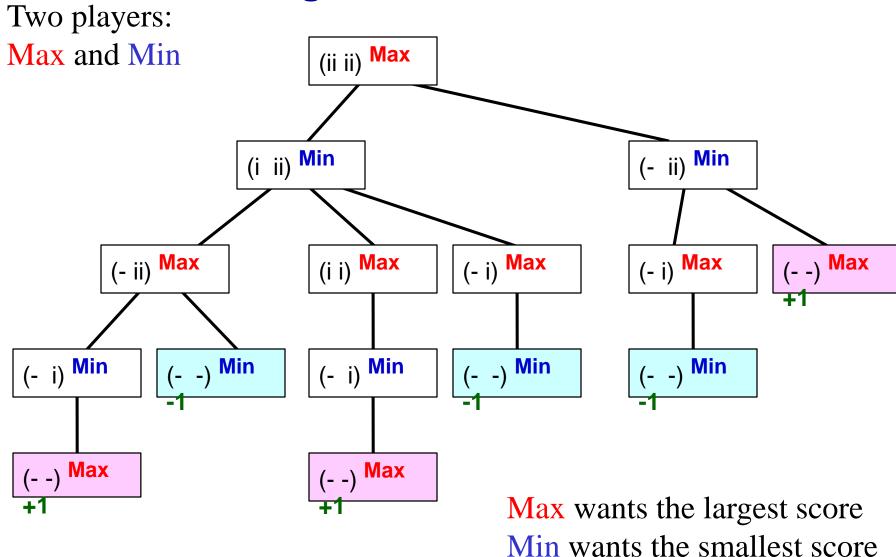






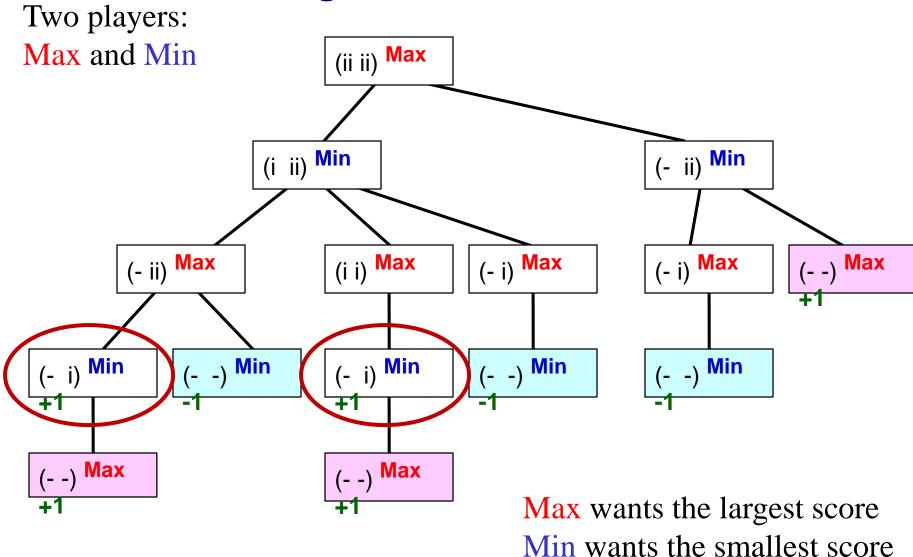


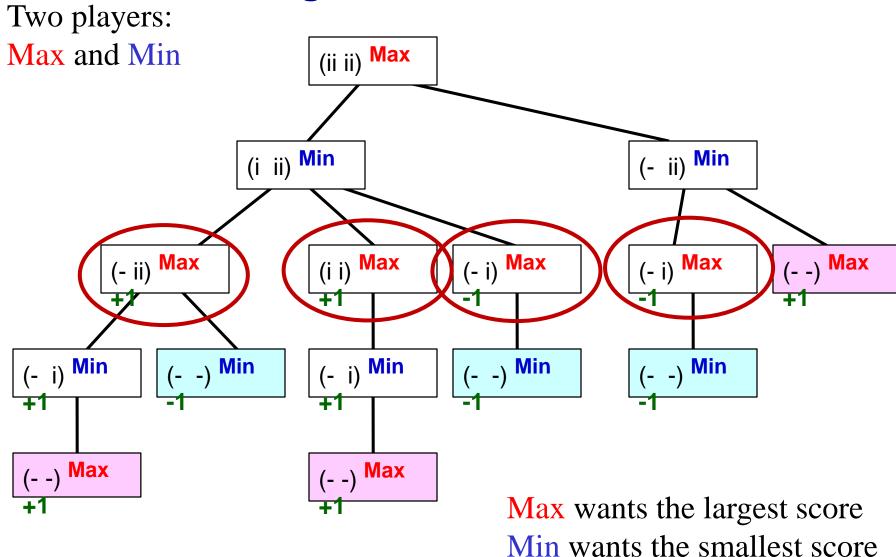


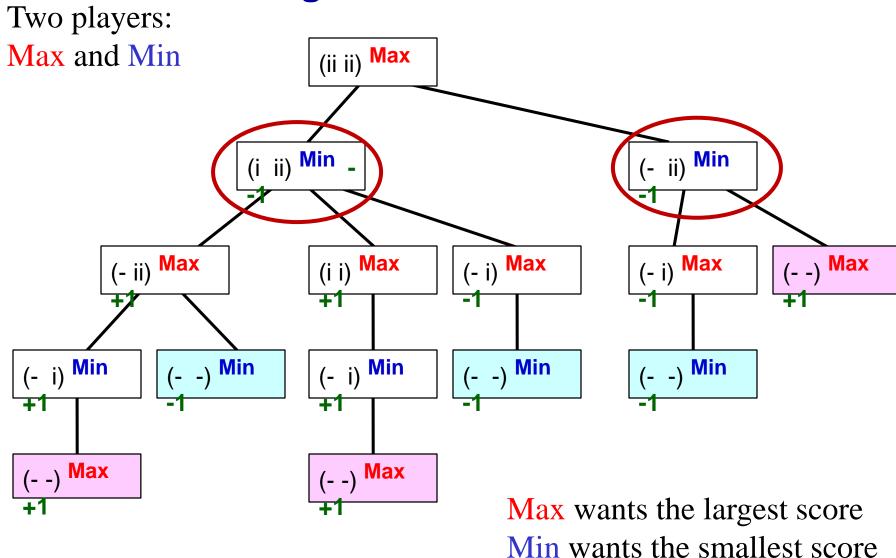


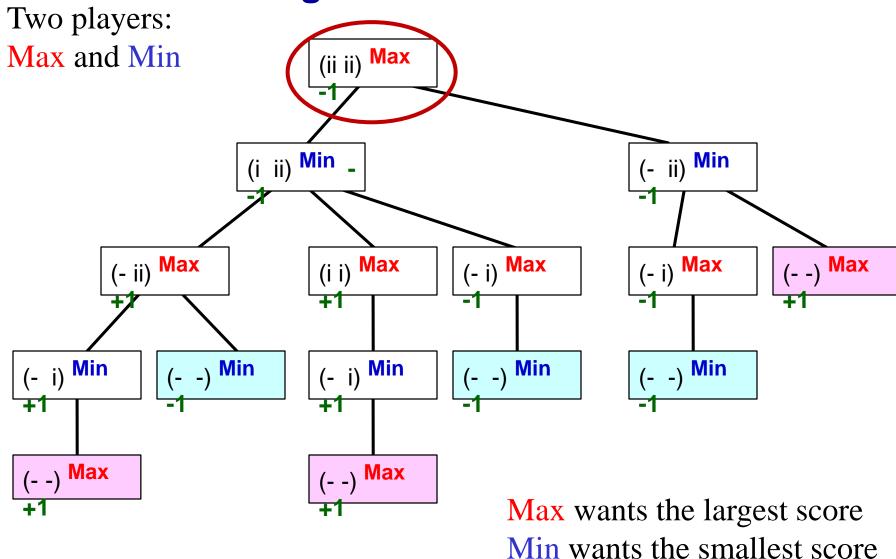
Game theoretic value

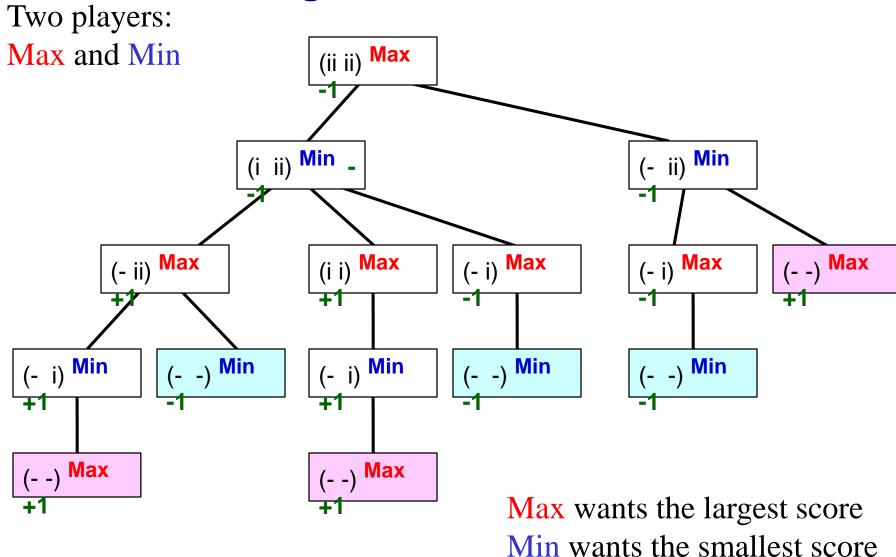
 Game theoretic value (a.k.a. minimax value) of a node = the score of the terminal node that will be reached if both players play optimally.

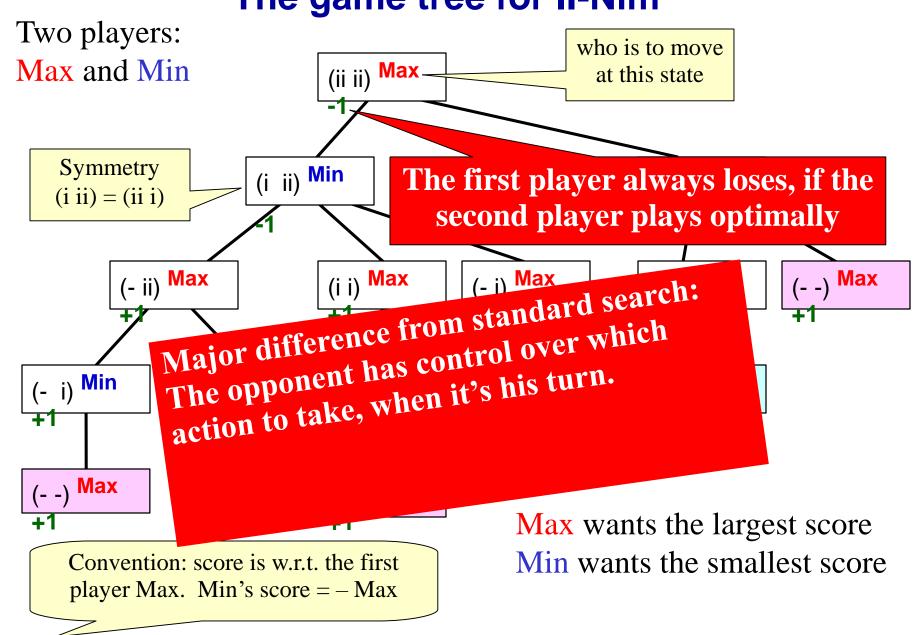












Game theoretic value

- Game theoretic value (a.k.a. minimax value) of a node = the score of the terminal node that will be reached if both players play optimally.
- = The numbers we filled in.
- Computed bottom up
 - In Max's turn, take the max of the children (Max will pick that maximizing action)
 - In Min's turn, take the min of the children (Min will pick that minimizing action)
- Implemented as a modified version of DFS: minimax algorithm

Minimax algorithm

```
function Max-Value(s)
inputs:
    s: current state in game, Max about to play
output: best-score (for Max) available from s
    if (s is a terminal state)
    then return (terminal value of s)
    else
              \alpha := -\infty
              for each s' in Succ(s)
                  \alpha := \max(\alpha, Min-value(s'))
    return a
function Min-Value(s)
output: best-score (for Min) available from s
   if (s is a terminal state)
    then return (terminal value of s)
    else
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              for each s' in Succs(s)
                  \beta := \min(\beta, \frac{Max-value(s')}{s'})
    return β
```

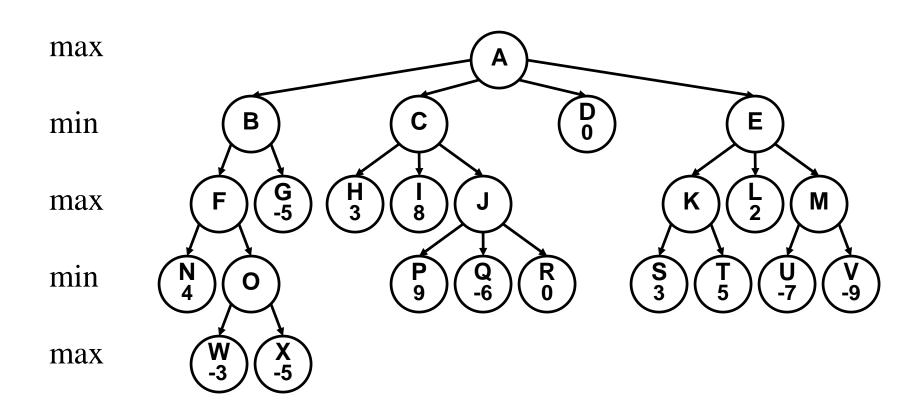
- Time complexity?
- Space complexity?

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```

- Time complexity?O(b^m) ← bad
- Space complexity?O(bm)

Minimax example



What are the game theoretic values? In particular, A's



Against a dumber opponent?

