

# Game Playing Summary

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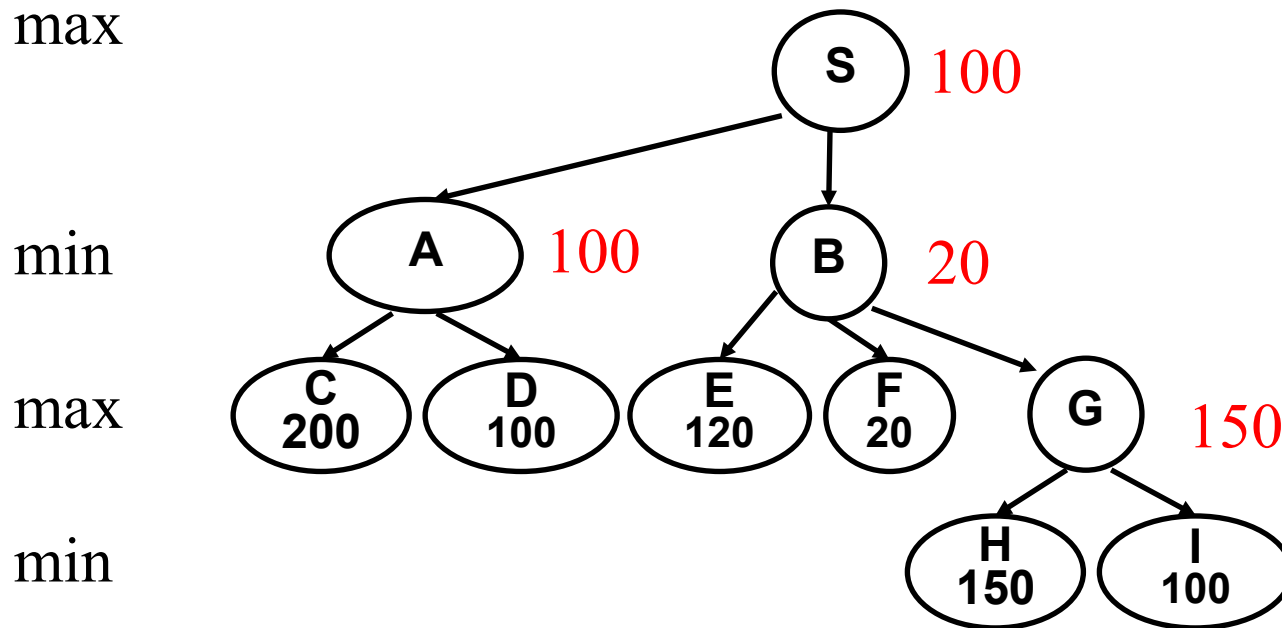
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[based on slides from A. Moore, C. Dyer, J. Skrentny, Jerry Zhu]

# The Kind of Games Focused

- Two-player
- Zero-sum
- Discrete finite
- Deterministic
- Perfect information

# Key Concept: Game Theoretic Value



The bottom-up approach: needs the whole game tree.  
The minimax algorithm: a variant of DFS to save space.

# Key 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)

$\alpha := \max(\alpha, \text{Min-value}(s'))$

return  $\alpha$

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

$\beta := \infty$

for each s' in Succs(s)

$\beta := \min(\beta, \text{Max-value}(s'))$

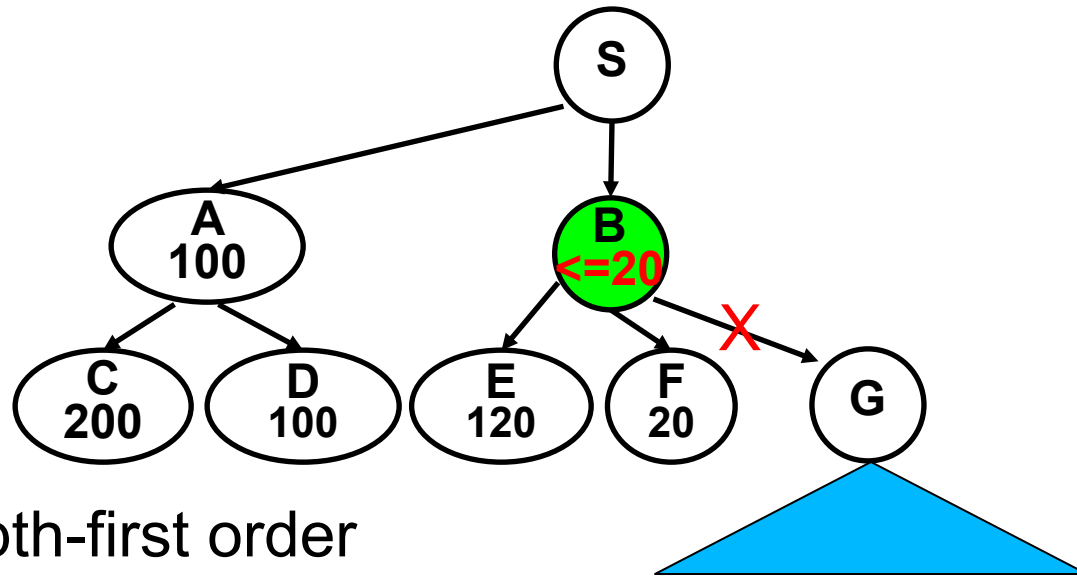
return  $\beta$

- Time complexity  $O(b^m)$
- Space complexity  $O(bm)$

# Alpha-Beta Pruning Intuition

max

min



- Depth-first order
- After returning from A, Max can get at least 100 at S
- After returning from F, Max can get at most 20 at B
- At this point, Max loses interest in B
- There is no need to explore G. The subtree at G is pruned. Saves time.

# Key Algorithm: Alpha-beta pruning

function **Max-Value** (s,  $\alpha$ ,  $\beta$ )

**inputs:**

s: current state in game, Max about to play

$\alpha$ : best score (highest) for Max along path to s

$\beta$ : best score (lowest) for Min along path to s

**output:**  $\min(\beta, \text{best-score (for Max) available from s})$

if ( s is a terminal state )

then return ( terminal value of s )

else for each s' in Succ(s)

$\alpha := \max(\alpha, \text{Min-value}(s', \alpha, \beta))$

if (  $\alpha \geq \beta$  ) then return  $\beta$  /\* alpha pruning \*/

return  $\alpha$

function **Min-Value**(s,  $\alpha$ ,  $\beta$ )

**output:**  $\max(\alpha, \text{best-score (for Min) available from s})$

if ( s is a terminal state )

then return ( terminal value of s )

else for each s' in Succs(s)

$\beta := \min(\beta, \text{Max-value}(s', \alpha, \beta))$

if (  $\alpha \geq \beta$  ) then return  $\alpha$  /\* beta pruning \*/

return  $\beta$

Starting from the root:

Max-Value(root,  $-\infty, +\infty$ )