# Artificial Intelligence

Lec 10: Adversarial Search

Pratik Mazumder

- Deterministic, zero-sum games:
  - Tic-tac-toe, chess, checkers
  - One player maximizes result
  - The other minimizes result
- Minimax search:
  - A state-space search tree
  - Players alternate turns
  - Compute each node's minimax value:
    - the best achievable utility against a rational (optimal) adversary

# Minimax values: computed recursively max max node min

Terminal values: part of the game

```
def max-value(state):
    if terminal-test(state):
        return utility(state)
    initialize v = -∞
    for each successor of state:
        v = max(v, min-value(successor))
    return v
```

$$V(s) = \max_{s' \in \text{successors}(s)} V(s')$$

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def min-value(state):
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 $V(s') = \min_{s \in \text{successors}(s')} V(s)$ 

#### Minimax Terminology

- move: a move by both players
- ply: a half move, i.e., action by one player
- backed-up value
  - Of a max position: the value of its largest successor
  - Of a min position: the value of its smallest successor
- Minimax procedure:
  - Search down several levels.
  - At the bottom level, apply the utility function.
  - o Back-up values all the way up to the root node
  - Select a move starting from the root node [if you perform the first move of the game].

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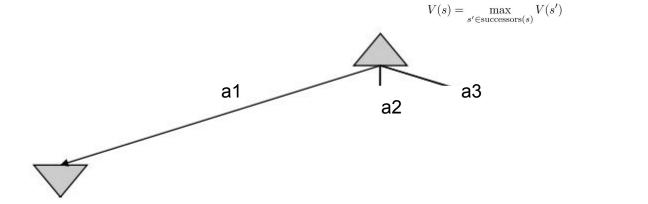
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b1

Is Terminal?

b2

b3

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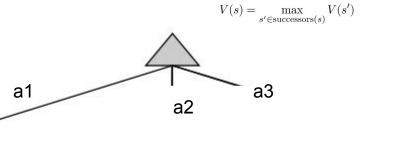
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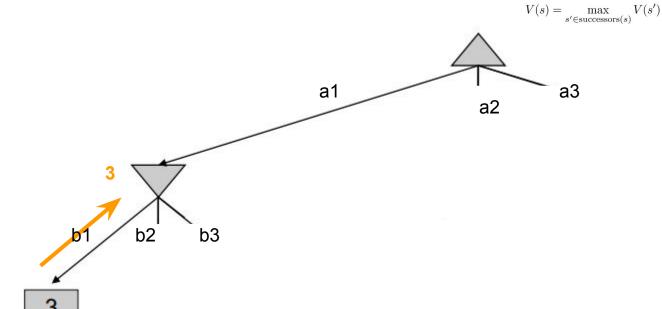
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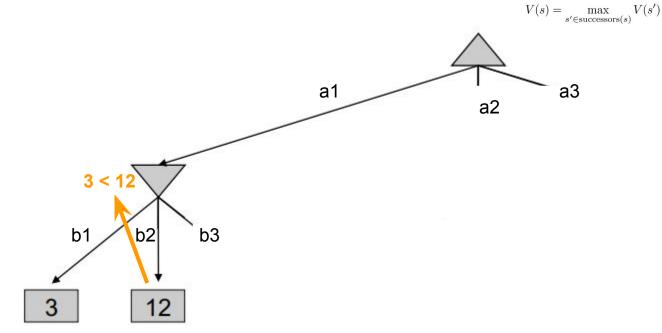
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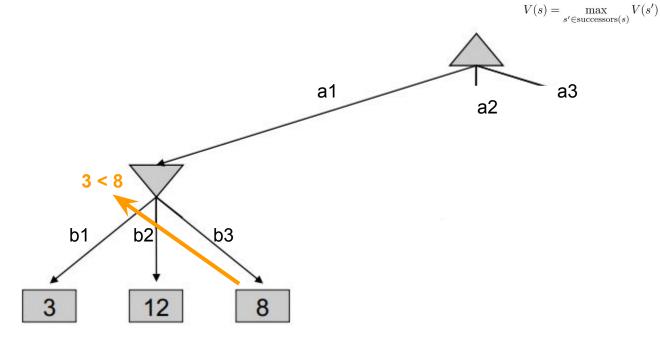
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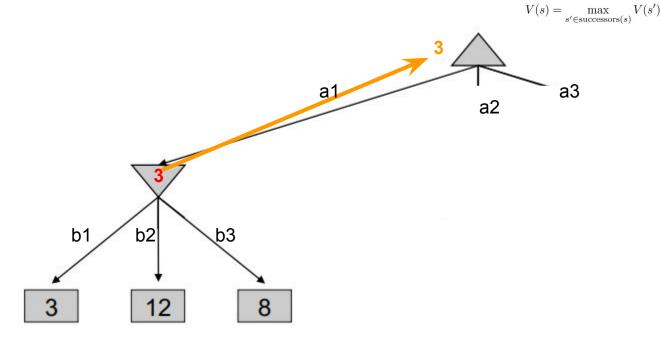
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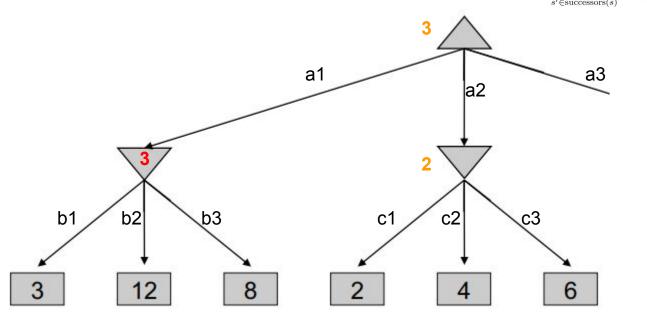
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8

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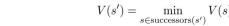
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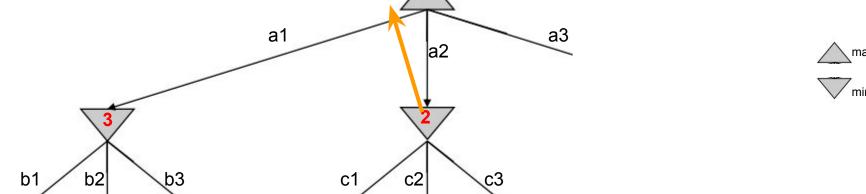
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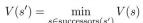


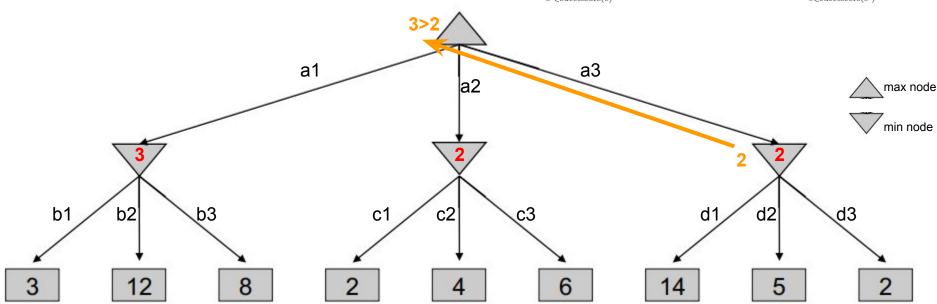


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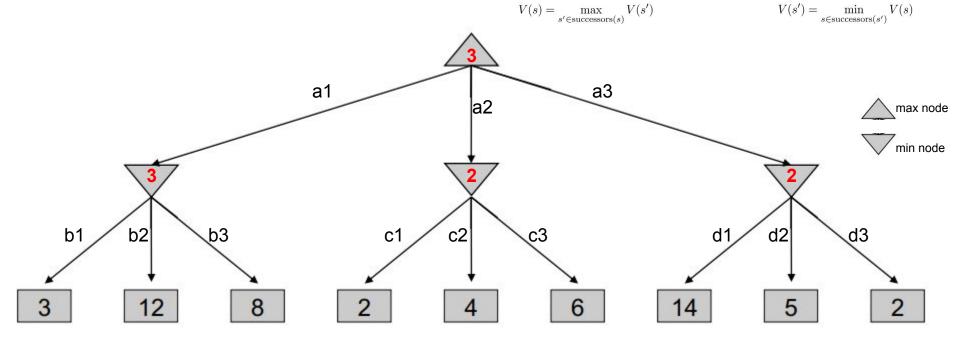


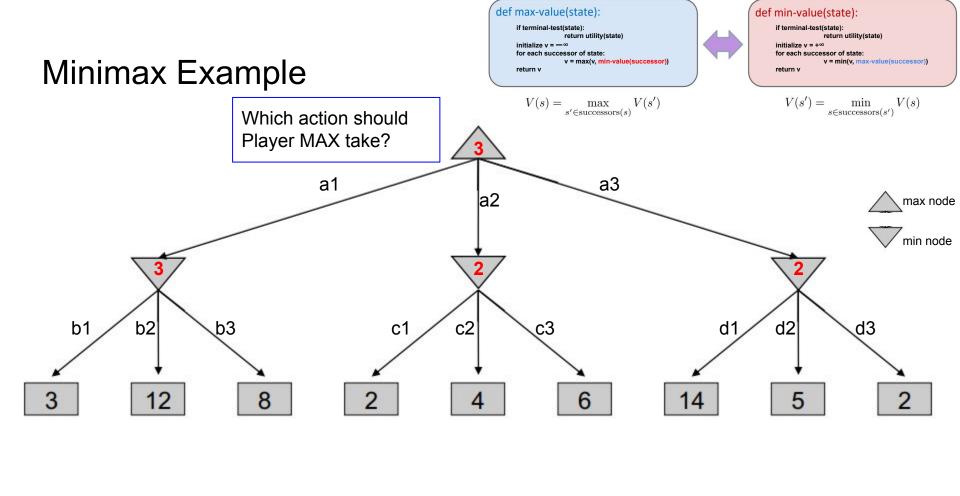
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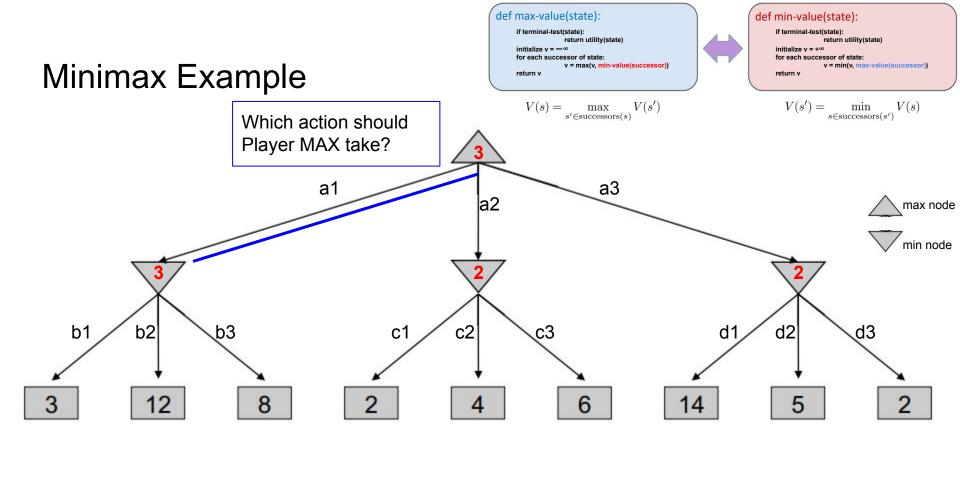
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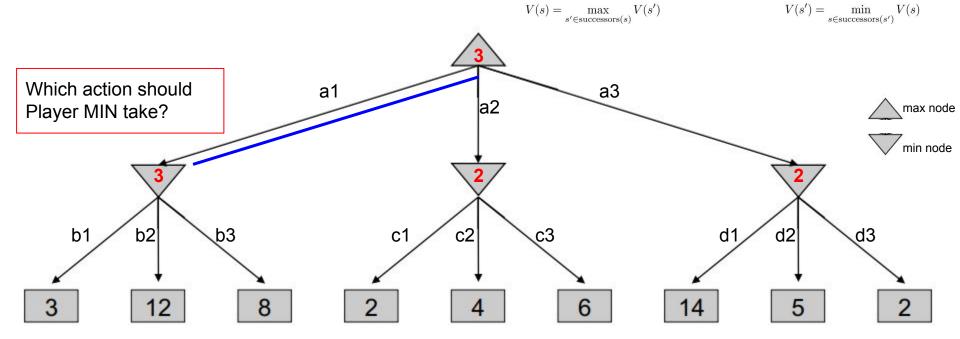
return v

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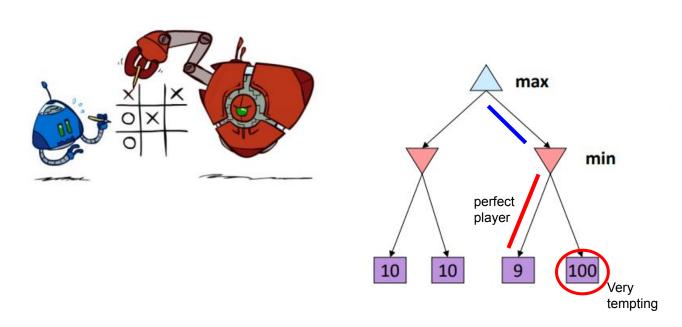
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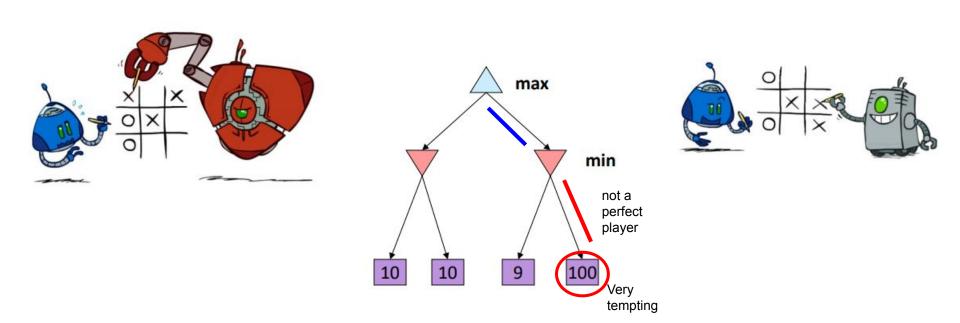


### Minimax Properties



Optimal against a perfect player.

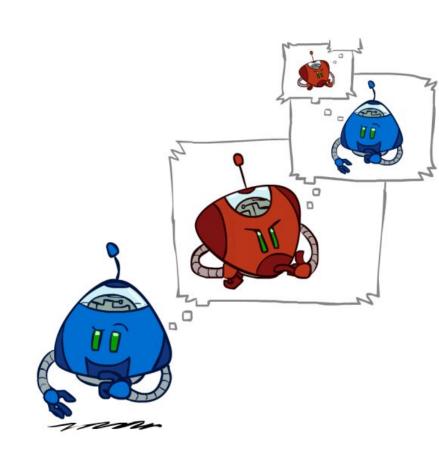
#### Minimax Properties



Optimal against a perfect player. Otherwise?

#### Minimax Efficiency

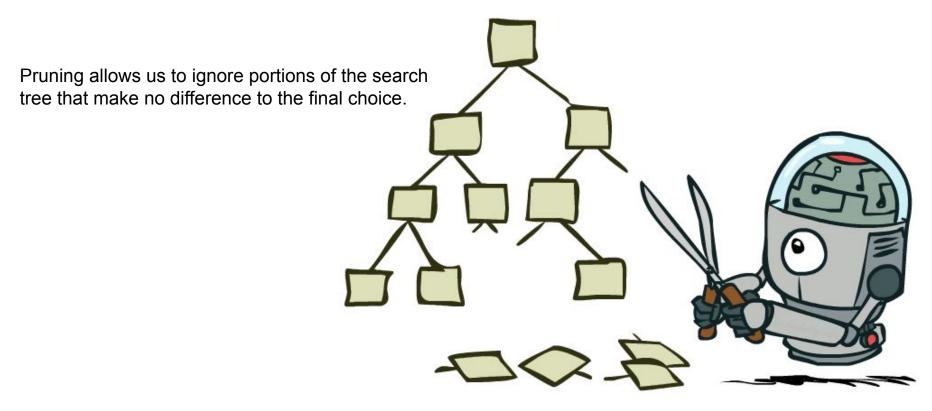
- How efficient is minimax?
  - Just like (exhaustive) DFS
- Example: For chess, branching factor b ≈ 35, solution depth/how many turns the game lasts typically m ≈ 100
  - Search space  $b^m = 35^{100} \approx 10^{154}$
- Interesting Analogy: Universe
  - Number of atoms  $\approx 10^{78}$
  - Age  $\approx 10^{18}$  seconds
  - Avg no of chemical reactions: 108/sec
  - $\circ$  10<sup>8</sup> moves/sec x 10<sup>78</sup> x 10<sup>18</sup> = 10<sup>104</sup>
- For Go, b ≈ 250-300, m ≈ 150
- Exact solution is not very feasible.
- But, do we need to explore the whole tree?

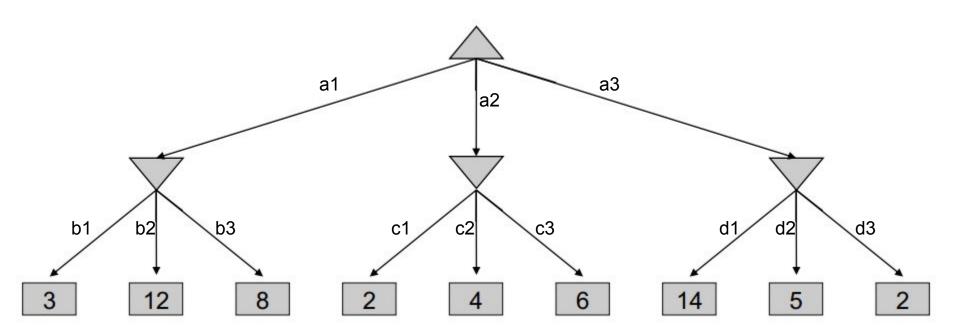


### Overcoming Resource Limits

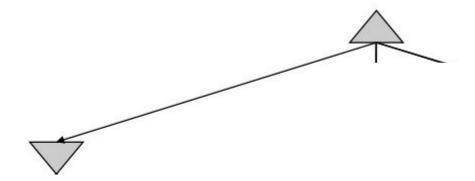


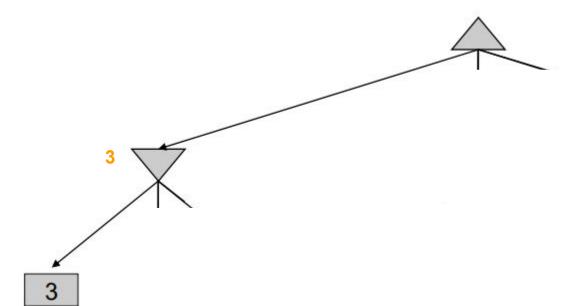
### Overcoming Resource Limits: Game Tree Pruning

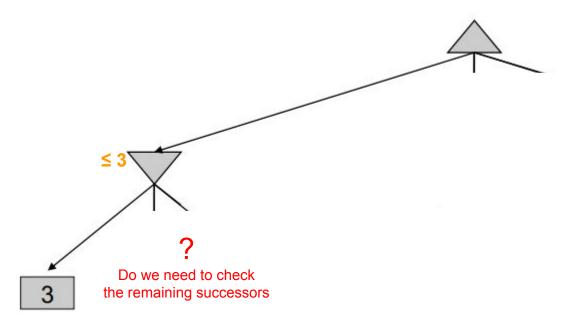


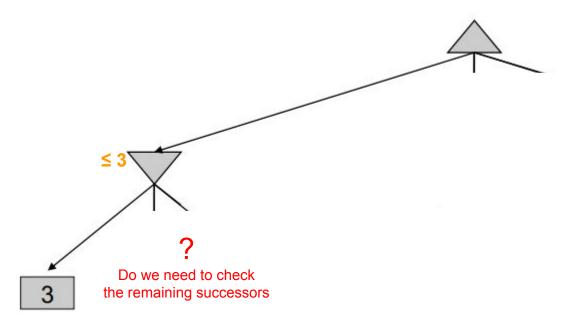


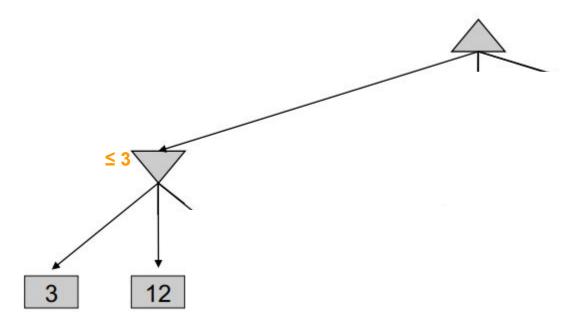


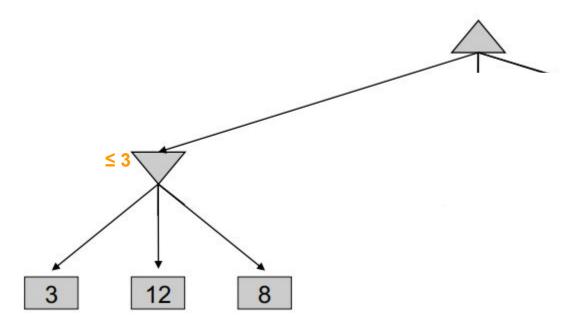


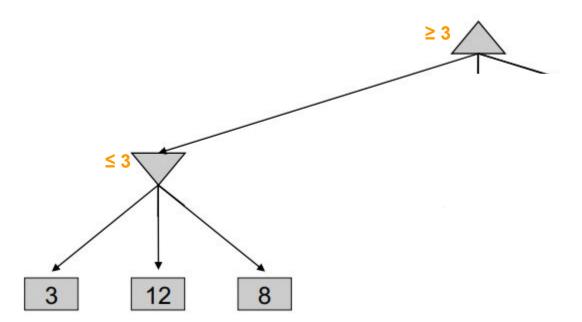


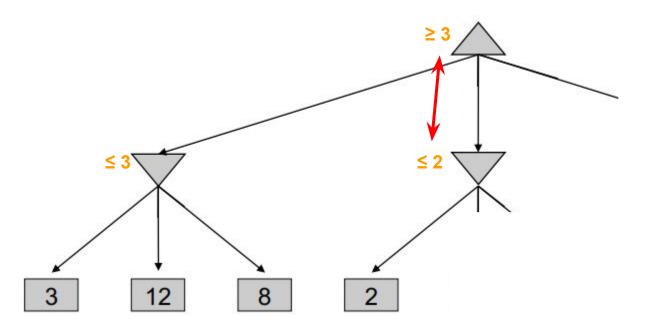


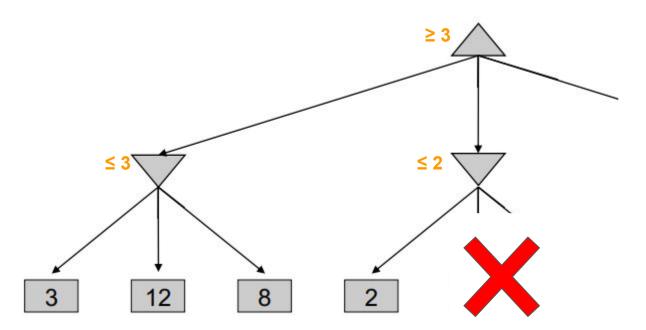




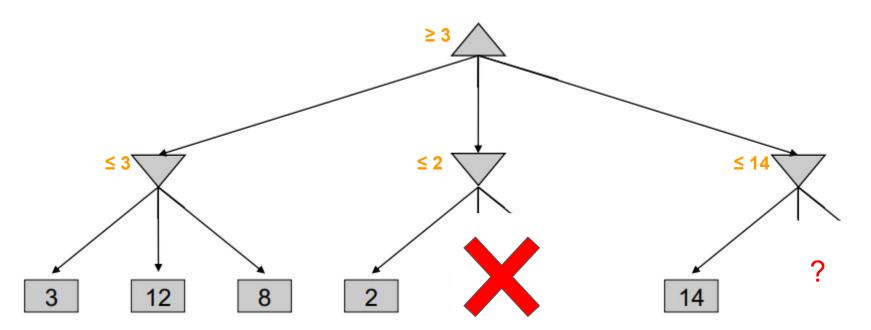




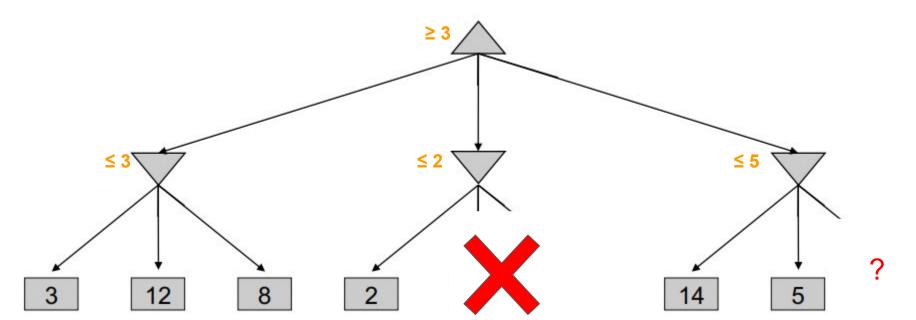




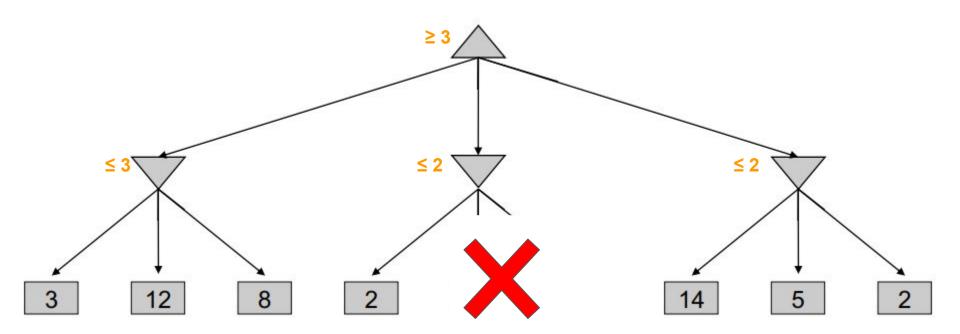
# Pruning



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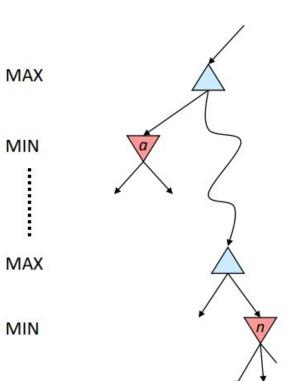


# Pruning



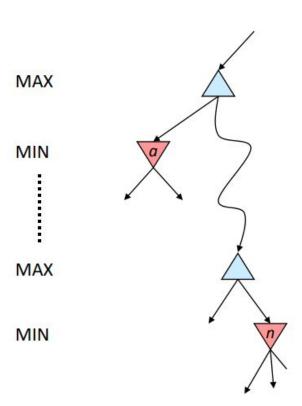
## Alpha-Beta Pruning

- The problem with minimax search is that the number of game states it has to examine is exponential in the depth of the tree.
- We can effectively **cut** the **search** and compute the correct minimax decision **without looking at every node** in the game tree.
- Alpha-beta pruning can be applied to trees of any depth, and it is often possible to prune entire subtrees rather than just leaves.
- Alpha—beta pruning gets its name from the following two parameters that describe bounds on the backed-up values that appear anywhere along the path:
  - $\alpha$  = the value of the best (i.e., highest-value) choice we have found so far at any choice point along the path for MAX.
  - $\circ$   $\beta$  = the value of the best (i.e., lowest-value) choice we have found so far at any choice point along the path for MIN.



## Alpha-Beta Pruning

- Alpha-beta search **updates the values of**  $\alpha$  **and**  $\beta$  as it goes along.
- It prunes the remaining branches at a node (i.e., terminates the recursive call) as soon as the value of the current node is known to be worse than the current  $\alpha$  or  $\beta$  value depending on whether the current node is MAX or MIN.



## Alpha-Beta Implementation

```
def max-value(state):

if terminal-test(state):

return utility(state)

initialize v = -∞

for each successor of state:

v = max(v, min-value(successor))

return v

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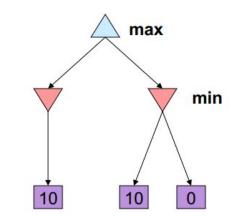
return v
```

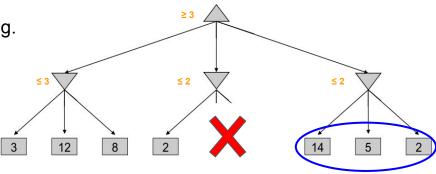
α: MAX's best option on path to root β: MIN's best option on path to root

```
\begin{aligned} &\text{def max-value(state, } \alpha, \, \beta): \\ &\text{if terminal-test(state):} \\ &\text{return utility(state)} \\ &\text{initialize } v = -\infty \\ &\text{for each successor of state:} \\ &\text{$v = \max(v, \min\text{-value(successor, } \alpha, \, \square))$} \\ &\text{if $v \geq \square$ return $v$} \\ &\alpha = \max(\alpha, v) \\ &\text{return $v$} \end{aligned}
```

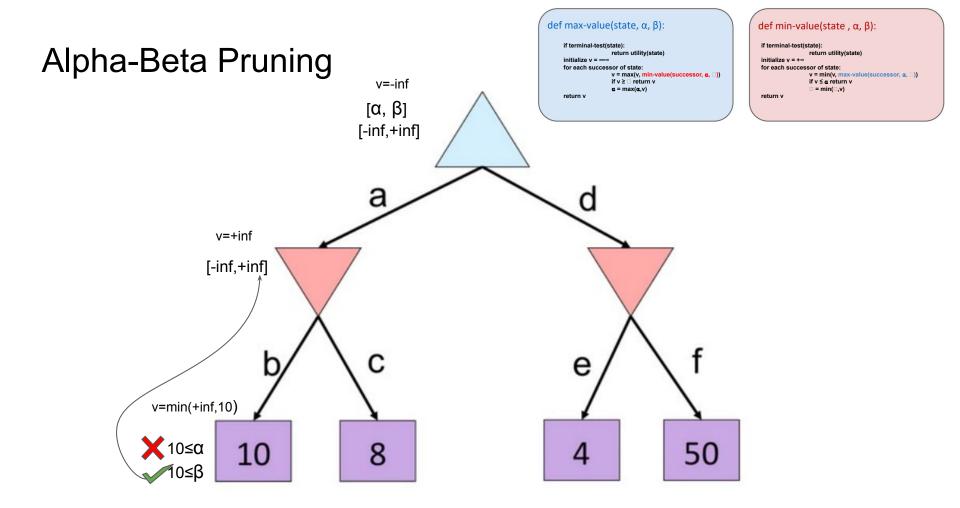
## Alpha-Beta Pruning Properties

- This pruning has no effect on the minimax value computed for the root!
- Values of intermediate nodes might be wrong.
  - Important: children of the root may have the wrong value.
  - So the most naive version won't let you do action selection.
- Good child ordering improves the effectiveness of pruning.
- With "perfect ordering":
  - o Time for search goes down.
  - Doubles solvable depth!
  - Full search of, e.g., chess, is still hopeless...

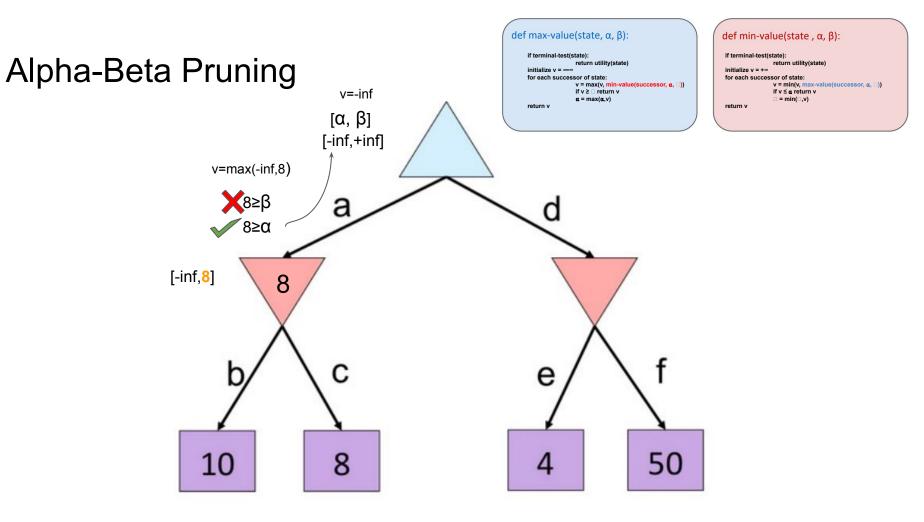


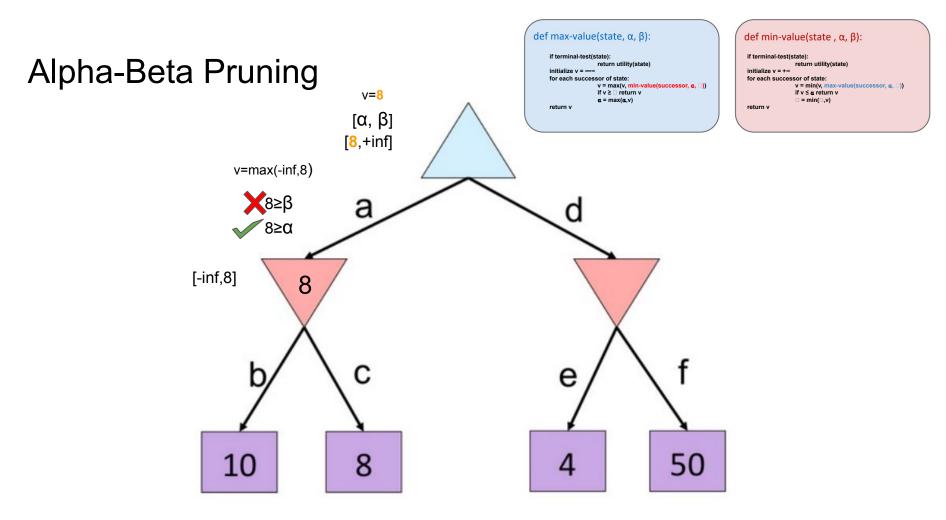


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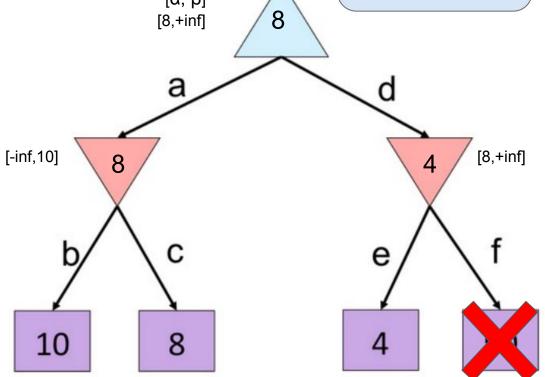
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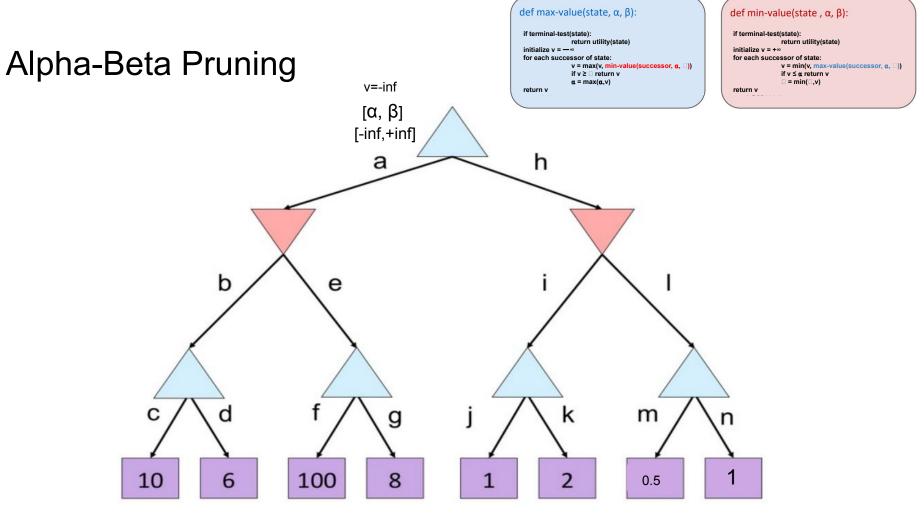
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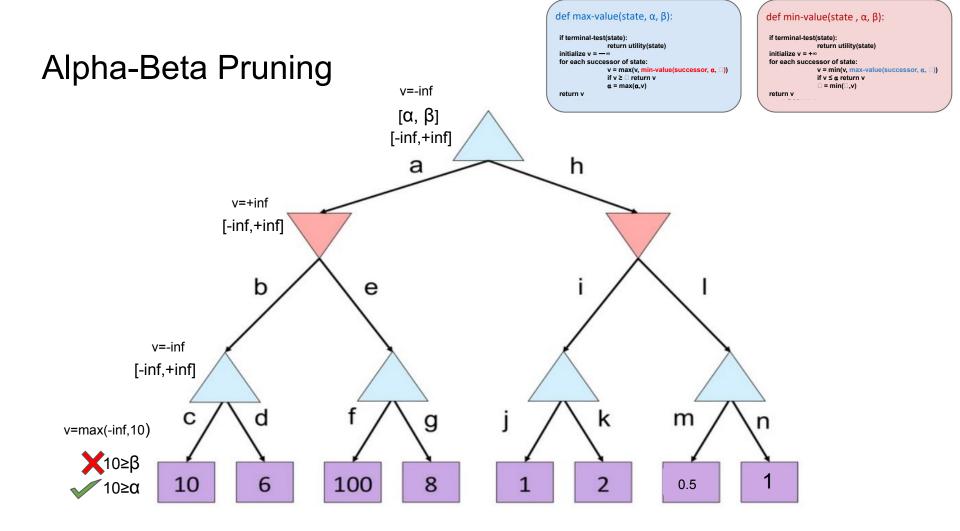
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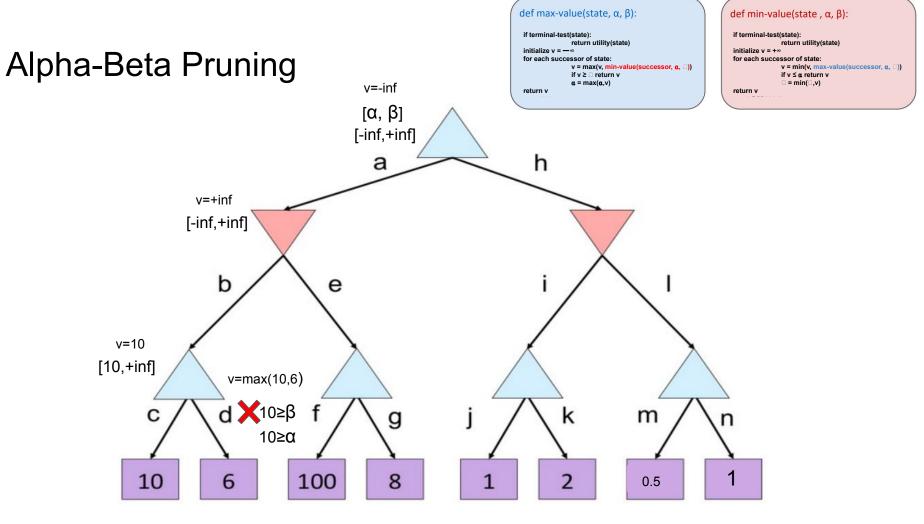
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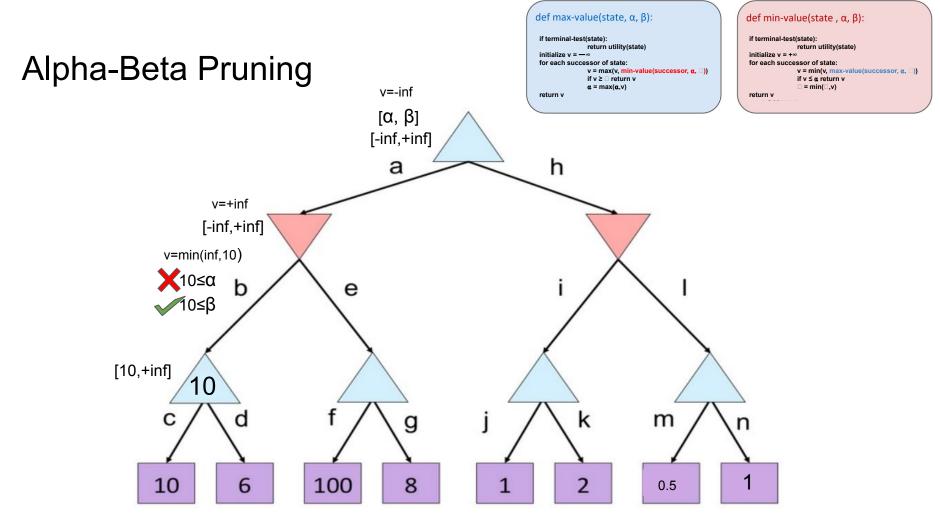
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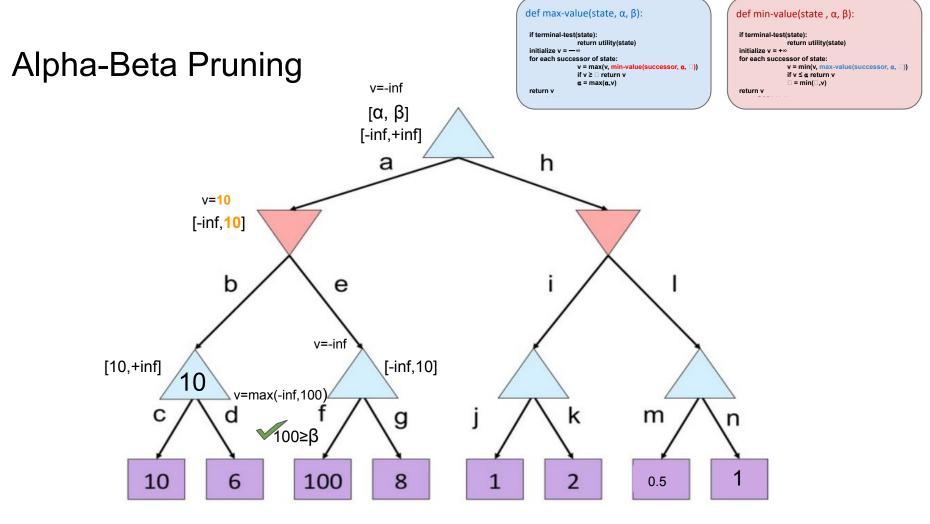


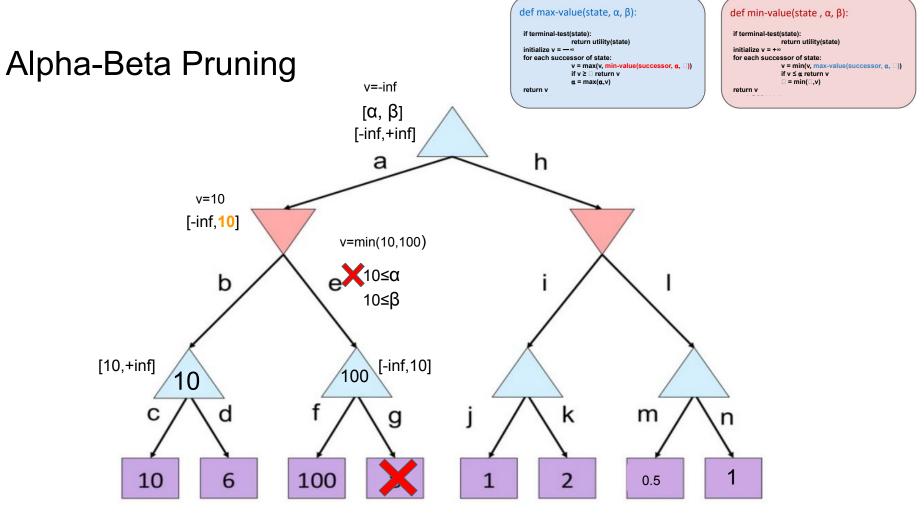




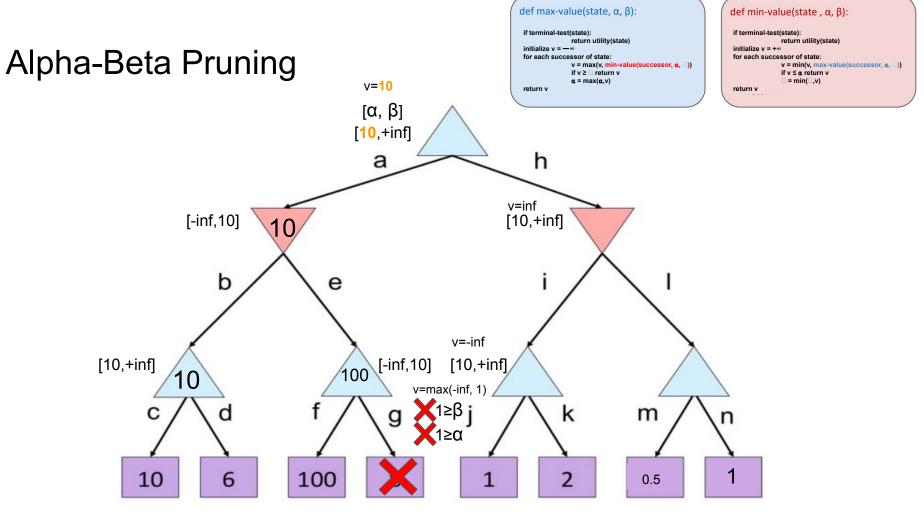


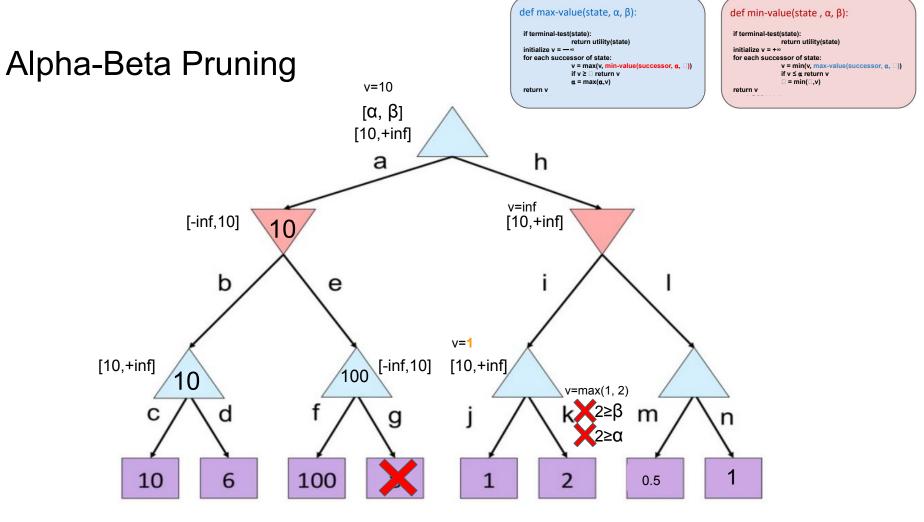


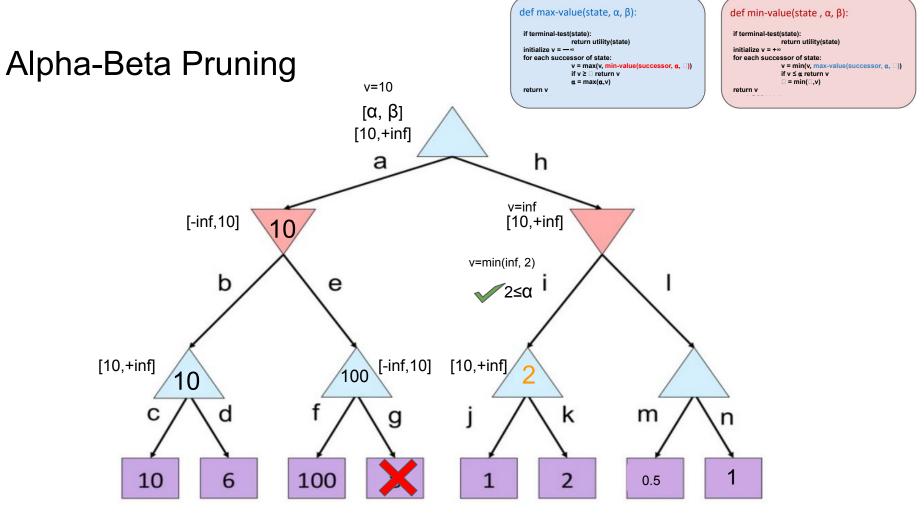


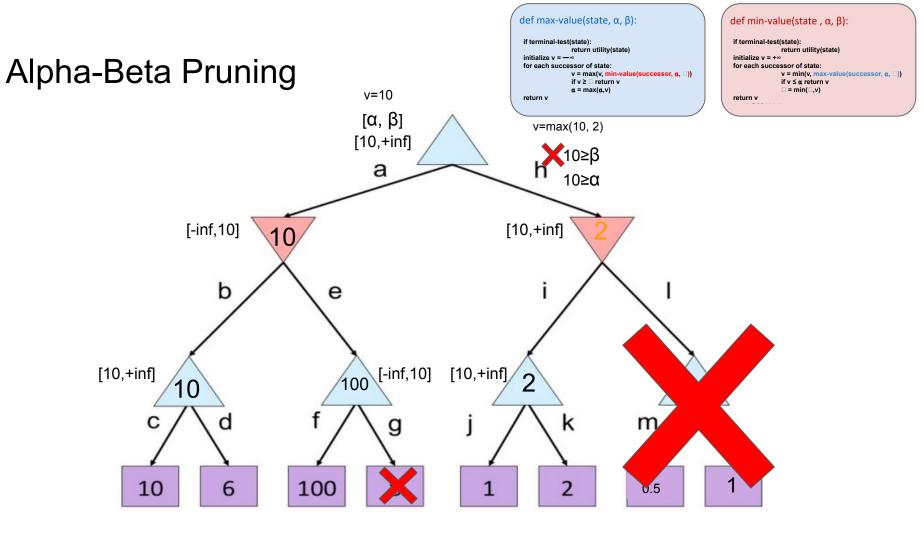


#### def max-value(state, $\alpha$ , $\beta$ ): def min-value(state, $\alpha$ , $\beta$ ): if terminal-test(state): if terminal-test(state): return utility(state) return utility(state) initialize v = -∞ initialize v = +∞ Alpha-Beta Pruning for each successor of state: for each successor of state: $v = max(v, min-value(successor, \alpha, \Box))$ $v = min(v, max-value(successor, \alpha, \square))$ if v ≥ □ return v if v ≤ α return v $\alpha = \max(\alpha, v)$ □ = min(□,v) v=-inf return v return v v=max(-inf,10) [α, β] **X**10≥β [-inf,+inf] h a [-inf,10] b e [10,+inf] [-inf,10] m 6 100 10 0.5



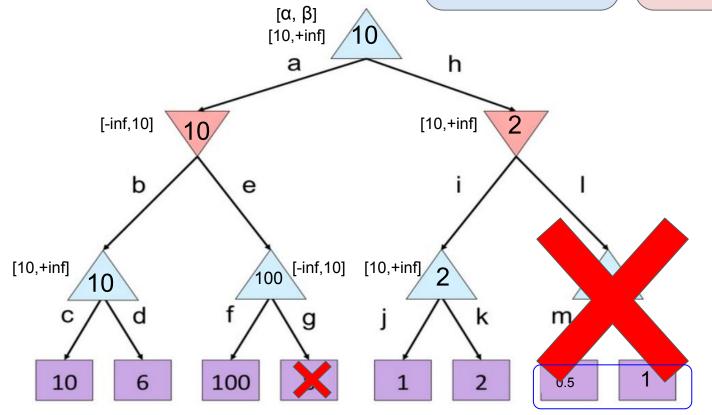






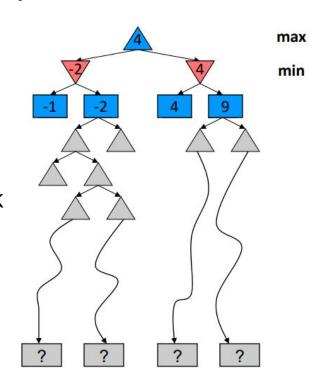
## Alpha-Beta Pruning

$$\label{eq:continuous} \begin{split} & \text{def max-value(state, } \alpha, \ \beta): \\ & \text{if terminal-test(state):} \\ & \text{return utility(state)} \\ & \text{initialize } v = --\infty \\ & \text{for each successor of state:} \\ & v = \max(v, \min\text{-value(successor, } \alpha, \dots)) \\ & \text{if } v \geq \square \text{ return } v \\ & \alpha = \max(\alpha, v) \\ & \text{return } v \end{split}$$

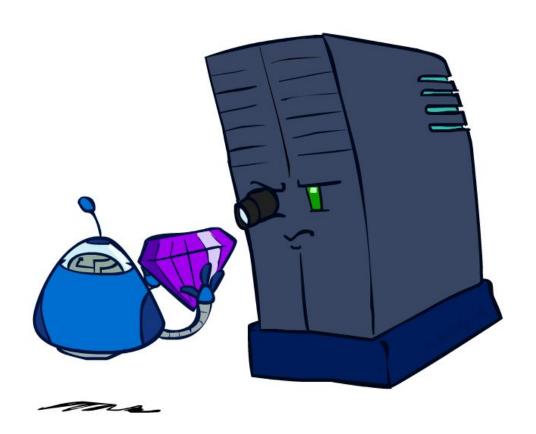


## Overcoming Resource Limits: Limiting Depth

- Problem: In realistic games, you cannot search upto leaves!
- Solution: Depth-limited search
  - o Instead, search only to a limited depth in the tree
  - Use an evaluation function for non-terminal positions
- Example:
  - Suppose we have 100 seconds for a move, and can explore 10K nodes per second.
  - So can check 1M nodes per move
- Guarantee of optimal play is gone.
- More plies/moves makes a BIG difference.
- Use iterative deepening.

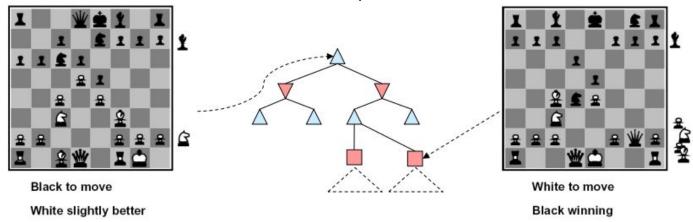


## **Evaluation Function**



### **Evaluation Function**

Evaluation functions score non-terminals in depth-limited search



- Ideal function: returns the actual minimax value of the position
- In practice: typically weighted linear sum of features:

$$Eval(s) = w_1 f_1(s) + w_2 f_2(s) + \dots + w_n f_n(s)$$

e.g. f1(s) = (num white queens – num black queens), etc.

## **Evaluation Function**

Evaluation functions are always imperfect.

• The deeper in the tree the evaluation function is buried, the less the quality of the evaluation function matters.



Tradeoff between complexity of features and complexity of computation

