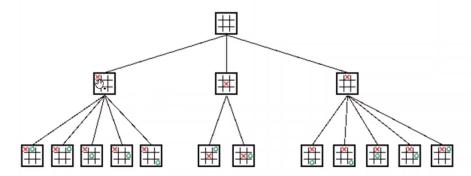
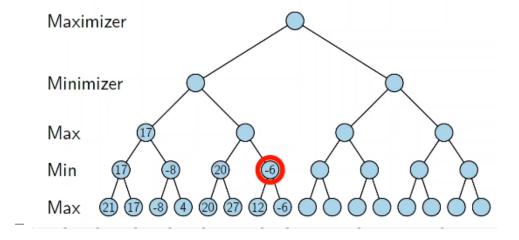
Classical Approach

- starting position as root
 - during game the current position
- states as nodes
- possible moves as edges to child nodes
 - states may be reached by multiple move sequences
 - performance improvement possible
 - * no longer a tree
- tree depth is bounded
 - exponential overhead
 - combinatorial explosion
- half move
 - own move + opponents move = full move
 - must also consider the opponents answer move
 - k half moves
 - * $\frac{k}{2}$ (+1) moves per player

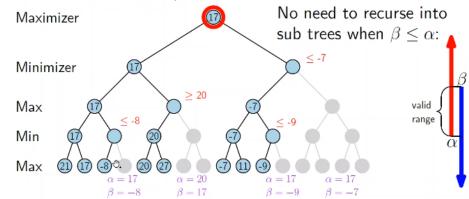


- not shown possible moves are symmetrical
- leaves nodes are evaluated using heuristics
 - heuristic scores the game state
 - terminal states (win, lose) have extreme values
 - non-leave nodes evaluated by traversing the tree downwards
 - * chooses maximum/minimum score of available moves
- maximizer and minimizer
 - one player wants to maximize the value of all child nodes
 - other player wants to minimize the value of all child nodes



- Pruning

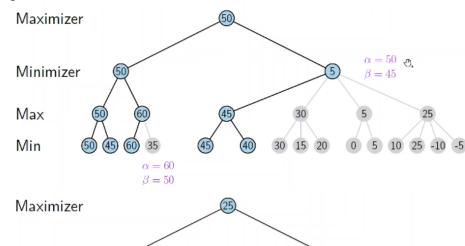
- prevents combinatorial explosion
- reduces states to consider
 - combines states with identical outcome
- more efficient and equal results as the classical approach
- stops traversing tree downwards upon reaching
 - a minimum value smaller than current maximum
 - $-\,$ a maximum value larger than current minimum
 - $\alpha :$ Lower bound for maximizer: no need to consider states with scores below $\alpha .$
 - β : Upper bound for minimizer: no need to consider states with scores above β .



• pseudo code

```
evaluate (node, alpha, beta)
    if node is a leaf
       return (heuristic value of node)
    if node is a minimizing node
       for each child of node
           beta = min (beta, evaluate (child, alpha, beta))
           if beta <= alpha
              return (alpha)
       return (beta)
    if node is a maximizing node
       for each child of node
           alpha = max (alpha, evaluate (child, alpha, beta))
           if beta <= alpha
              return (beta)
       return (alpha)
evaluate(root,-\infty,\infty);
```

• examples



Minimizer 20 30 55 30 45 25 Min 20 10 5 30 65 40 30 15 20 5 45 10 25 10 5

How much can we save with α - β pruning?

- Sometimes nothing!
- pre-sorted pruning
 - efficiency depends on order of states

Consider potentially best states first (large values for maximizer, small (large negativ) for minimizer)