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Connect 4 Game

Pseudocodes for the algorithms Used :

```
function alphabeta(node, depth,  $\alpha$ ,  $\beta$ , maximizingPlayer) is
  if depth == 0 or node is terminal then
    return the heuristic value of node
  if maximizingPlayer then
    value :=  $-\infty$ 
    for each child of node do
      value := max(value, alphabeta(child, depth - 1,  $\alpha$ ,  $\beta$ , FALSE))
      if value >  $\beta$  then
        break (*  $\beta$  cutoff *)
     $\alpha$  := max( $\alpha$ , value)
    return value
  else
    value :=  $+\infty$ 
    for each child of node do
      value := min(value, alphabeta(child, depth - 1,  $\alpha$ ,  $\beta$ , TRUE))
      if value <  $\alpha$  then
        break (*  $\alpha$  cutoff *)
     $\beta$  := min( $\beta$ , value)
    return value
```

```
function minimax(node, depth, maximizingPlayer) is
  if depth = 0 or node is a terminal node then
    return the heuristic value of node
  if maximizingPlayer then
    value :=  $-\infty$ 
    for each child of node do
      value := max(value, minimax(child, depth - 1, FALSE))
    return value
  else (* minimizing player *)
    value :=  $+\infty$ 
    for each child of node do
      value := min(value, minimax(child, depth - 1, TRUE))
    return value
```

```

function expectiminimax(node, depth)
  if node is a terminal node or depth = 0
    return the heuristic value of node
  if the adversary is to play at node
    // Return value of minimum-valued child node
    let  $\alpha := +\infty$ 
    foreach child of node
       $\alpha := \min(\alpha, \text{expectiminimax}(\text{child}, \text{depth}-1))$ 
  else if we are to play at node
    // Return value of maximum-valued child node
    let  $\alpha := -\infty$ 
    foreach child of node
       $\alpha := \max(\alpha, \text{expectiminimax}(\text{child}, \text{depth}-1))$ 
  else if random event at node
    // Return weighted average of all child nodes' values
    let  $\alpha := 0$ 
    foreach child of node
       $\alpha := \alpha + (\text{Probability}[\text{child}] \times \text{expectiminimax}(\text{child}, \text{depth}-1))$ 
  return  $\alpha$ 

```

Data Structures Used :

Numpy array to represent board

Lists used to get the windows in the array to check the heuristic value

Tuples used in checking winning move

Set used to store winning moves to prevent checking duplicates

Deque used to traverse nodes

Properties of minimax

- Optimal (opponent plays optimally) and complete (finite tree)
- DFS time: $O(b^m)$
- DFS space: $O(bm)$

Alpha Beta Pruning Strategy :

$\alpha - \beta$ pruning

- **Strategy:** Just like minimax, it performs a DFS.
 - **Parameters:** Keep track of two bounds
 - α : largest value for Max across seen children (current lower bound on MAX's outcome).
 - β : lowest value for MIN across seen children (current upper bound on MIN's outcome).
 - **Initialization:** $\alpha = -\infty$, $\beta = \infty$
 - **Propagation:** Send α , β values *down* during the search to be used for pruning.
 - Update α , β values by *propagating upwards* values of terminal nodes.
 - Update α only at Max nodes and update β only at Min nodes.
 - **Pruning:** Prune any remaining branches whenever $\alpha \geq \beta$.
-
- **Worst ordering:** no pruning happens (best moves are on the right of the game tree). Complexity $O(b^m)$.
 - **Ideal ordering:** lots of pruning happens (best moves are on the left of the game tree). This solves tree twice as deep as minimax in the same amount of time. Complexity $O(b^{m/2})$ (in practice). The search can go deeper in the game tree.

Heuristic Function Used :

Positive Scoring:

- Achieving 4 consecutive AI colors earns 4 points.
- Achieving 3 consecutive candidate AI colors earns 3 points.
- Achieving 2 consecutive candidate AI colors earns 2 points.
- Preventing the opponent from achieving a point earns 1 point.

Negative Scoring:

- Allowing the opponent to achieve 4 consecutive Human colors results in a deduction of 4 points.
- Allowing the opponent to achieve 3 consecutive candidate Human colors results in a deduction of 3 points.
- Allowing the opponent to achieve 2 consecutive candidate Human colors results in a deduction of 2 points.
- Failing to prevent opponent from achieving a point results in a deduction of 1 point.

Number Of Nodes expanded and time for all types and diff k

K = 3

Expanded Nodes: 399
Elapsed Time: 0.079993

Expect MiniMax

Expanded Nodes: 140
Elapsed Time: 0.024989

MiniMax with AlphaBeta pruning

Expanded Nodes: 399
Elapsed Time: 0.108992

MiniMax

K = 2

Expanded Nodes: 56
Elapsed Time: 0.010989

MiniMax

Expanded Nodes: 32
Elapsed Time: 0.006004

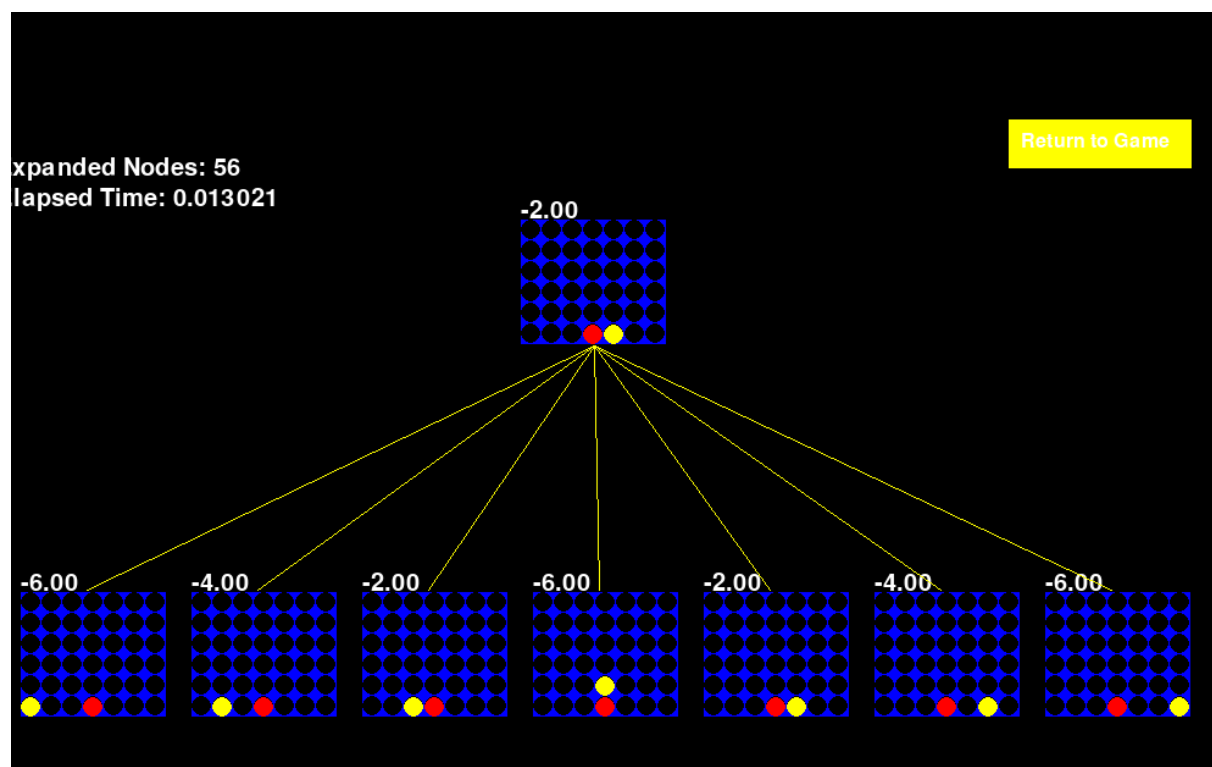
MiniMax with AlphaBeta pruning

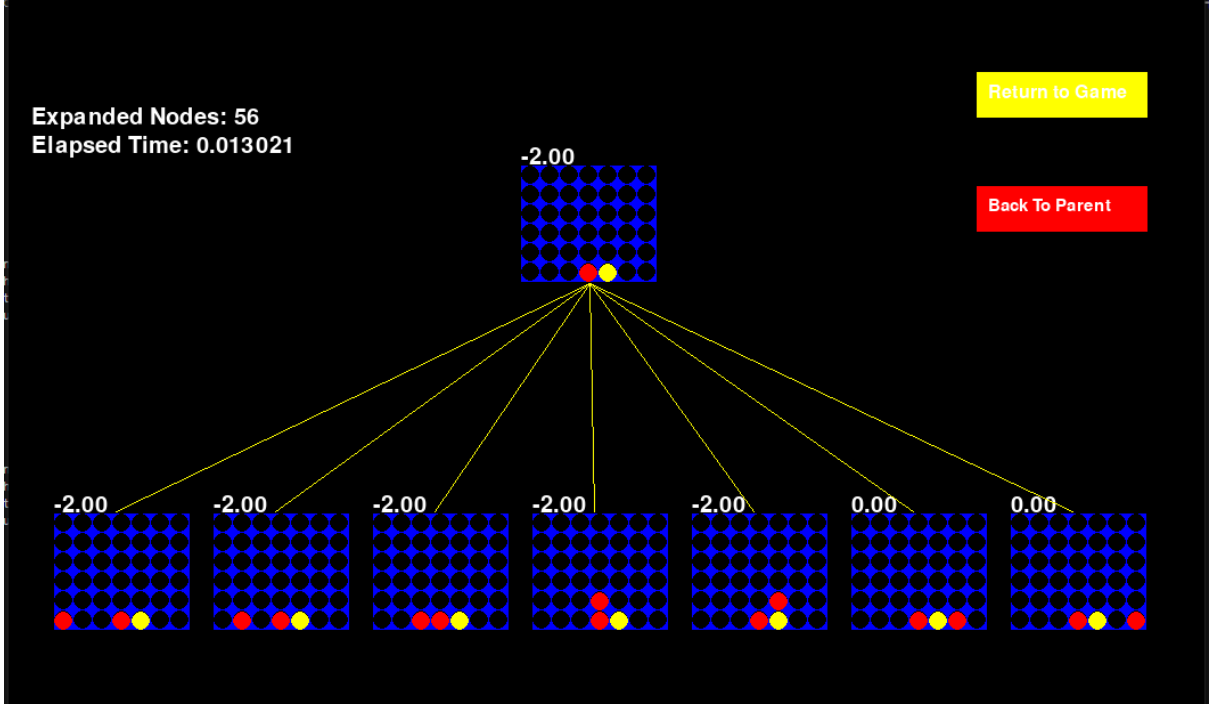
Expanded Nodes: 56
Elapsed Time: 0.012981

Expect MiniMax

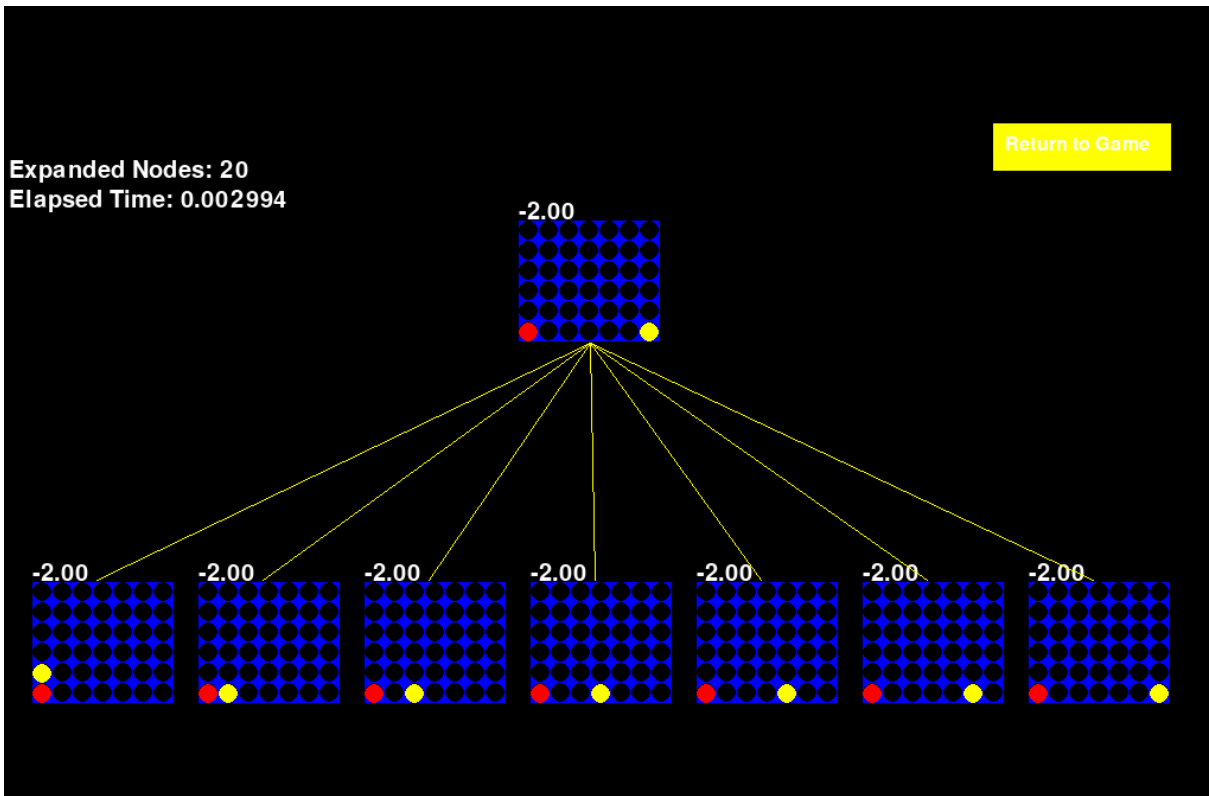
Full Sample Run

No Pruning





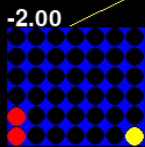
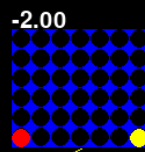
Full Sample Run With Alpha-Beta Pruning



Expanded Nodes: 20
Elapsed Time: 0.002994

Return to Game

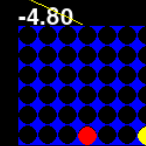
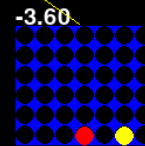
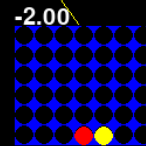
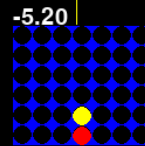
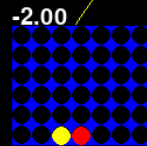
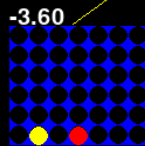
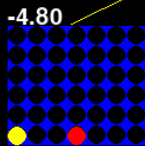
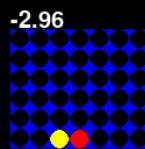
Back To Parent



Full Sample Run With Expect-MiniMax

Expanded Nodes: 56
Elapsed Time: 0.010996

Return to Game



Expanded Nodes: 56
Elapsed Time: 0.010998

Return to Game

Back To Parent

