

Introduction to Artificial Intelligence with Python

Adversarial Search

	O	X
O	X	X
O	X	

	O	X
O	X	X
O	X	

Minimax

O	X	X
O	O	
O	X	X

-1

X	O	X
O	O	X
X	X	O

0

O		X
	X	O
X	O	X

1

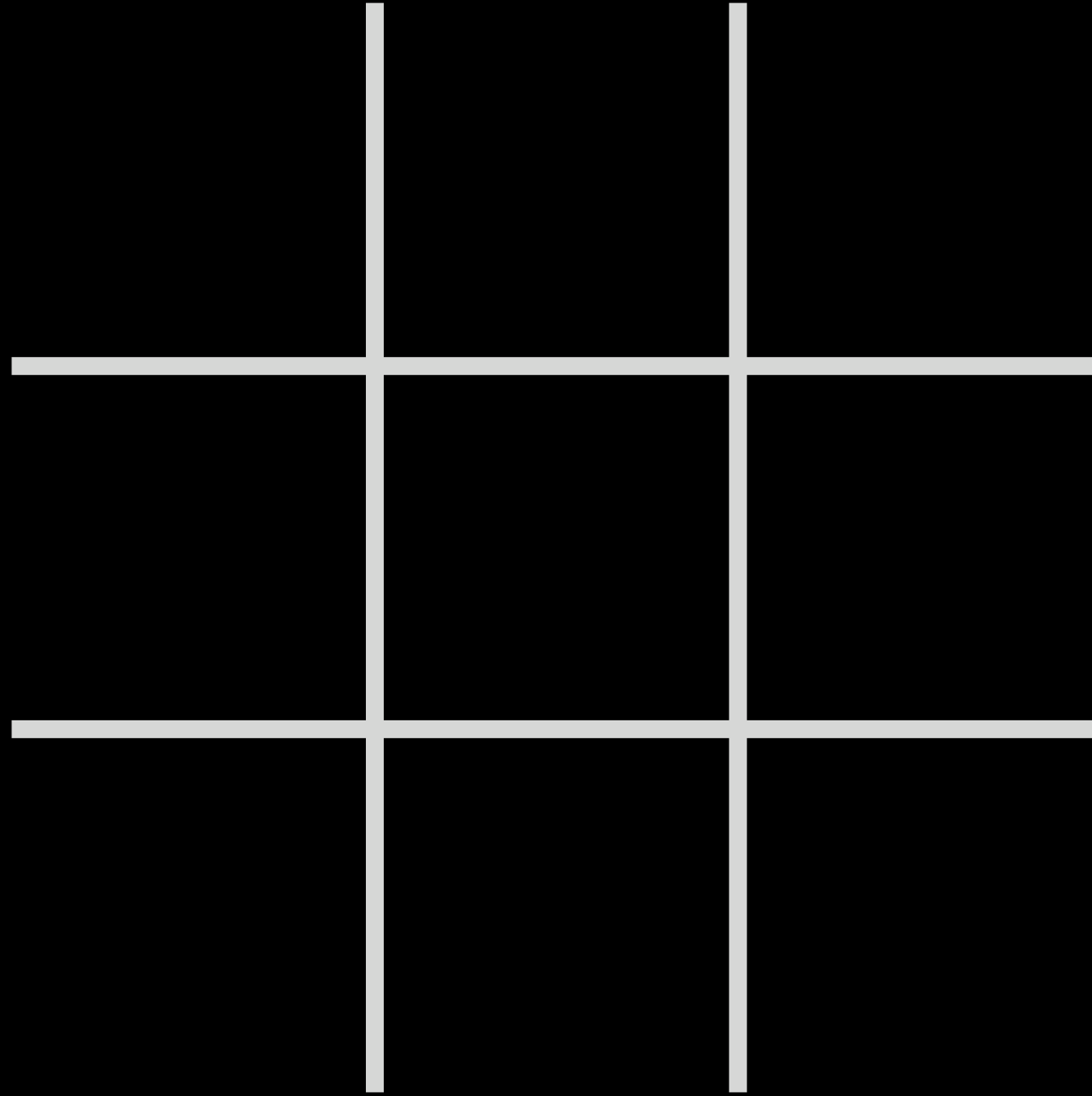
Minimax

- MAX (X) aims to maximize score.
- MIN (O) aims to minimize score.

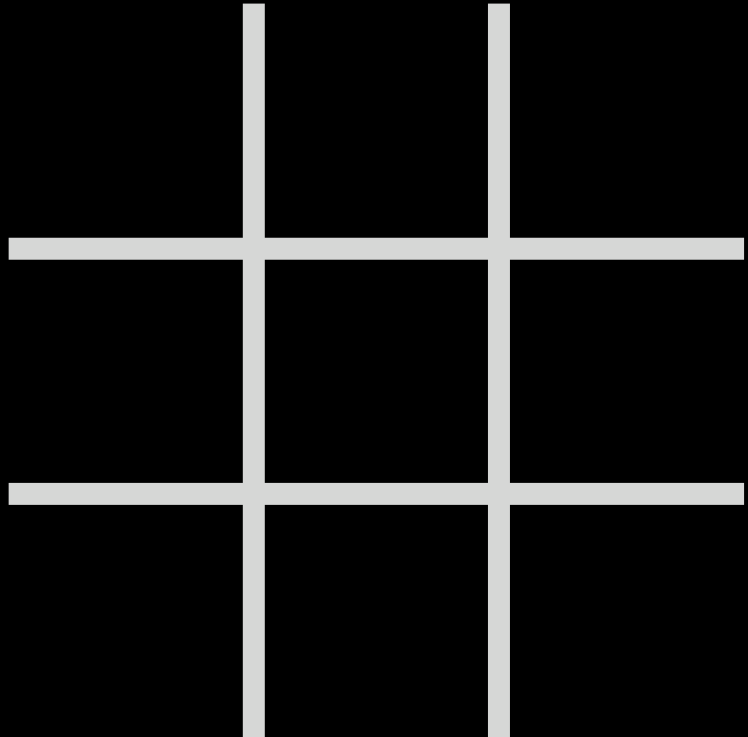
Game

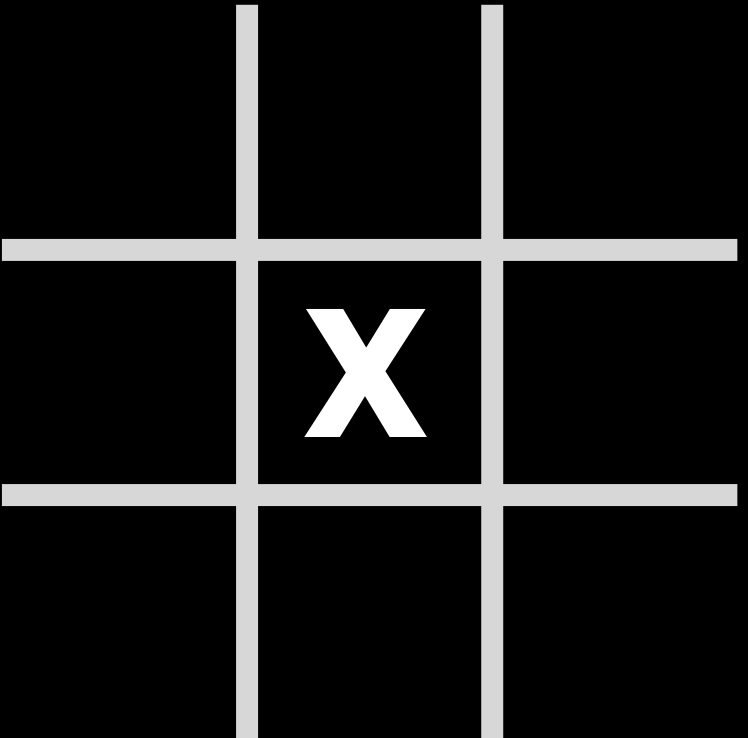
- S_0 : initial state
- $\text{PLAYER}(s)$: returns which player to move in state s
- $\text{ACTIONS}(s)$: returns legal moves in state s
- $\text{RESULT}(s, a)$: returns state after action a taken in state s
- $\text{TERMINAL}(s)$: checks if state s is a terminal state
- $\text{UTILITY}(s)$: final numerical value for terminal state s

Initial State



PLAYER(*s*)

PLAYER() = **X**

PLAYER() = **O**

ACTIONS(*s*)

ACTIONS(

	X	O
O	X	X
X		O

) = {

O		

 ,

	O	

 }

RESULT(*s*, *a*)

RESULT(

	X	O
O	X	X
X		O

 ,

O		

) =

O	X	O
O	X	X
X		O

TERMINAL(*s*)

TERMINAL(

o		
o	x	
x	o	x

) = false

TERMINAL(

o		x
o	x	
x	o	x

) = true

UTILITY(*s*)

$$\text{UTILITY}\left(\begin{array}{c|c|c} \mathbf{o} & & \mathbf{x} \\ \hline \mathbf{o} & \mathbf{x} & \\ \hline \mathbf{x} & \mathbf{o} & \mathbf{x} \end{array}\right) = 1$$

$$\text{UTILITY}\left(\begin{array}{c|c|c} \mathbf{o} & \mathbf{x} & \mathbf{x} \\ \hline \mathbf{x} & \mathbf{o} & \\ \hline \mathbf{o} & \mathbf{x} & \mathbf{o} \end{array}\right) = -1$$

O	X	O
O	X	X
X	X	O

VALUE: 1

PLAYER(s) = O

MIN-VALUE:
0

	X	O
O	X	X
X		O

MAX-VALUE:
1

O	X	O
O	X	X
X		O

MAX-VALUE:
0

	X	O
O	X	X
X	O	O

VALUE:
1

O	X	O
O	X	X
X	X	O

VALUE:
0

X	X	O
O	X	X
X	O	O

PLAYER(s) = O

MIN-VALUE:
0

	X	O
O	X	X
X		O

MAX-VALUE:
1

O	X	O
O	X	X
X		O

MAX-VALUE:
0

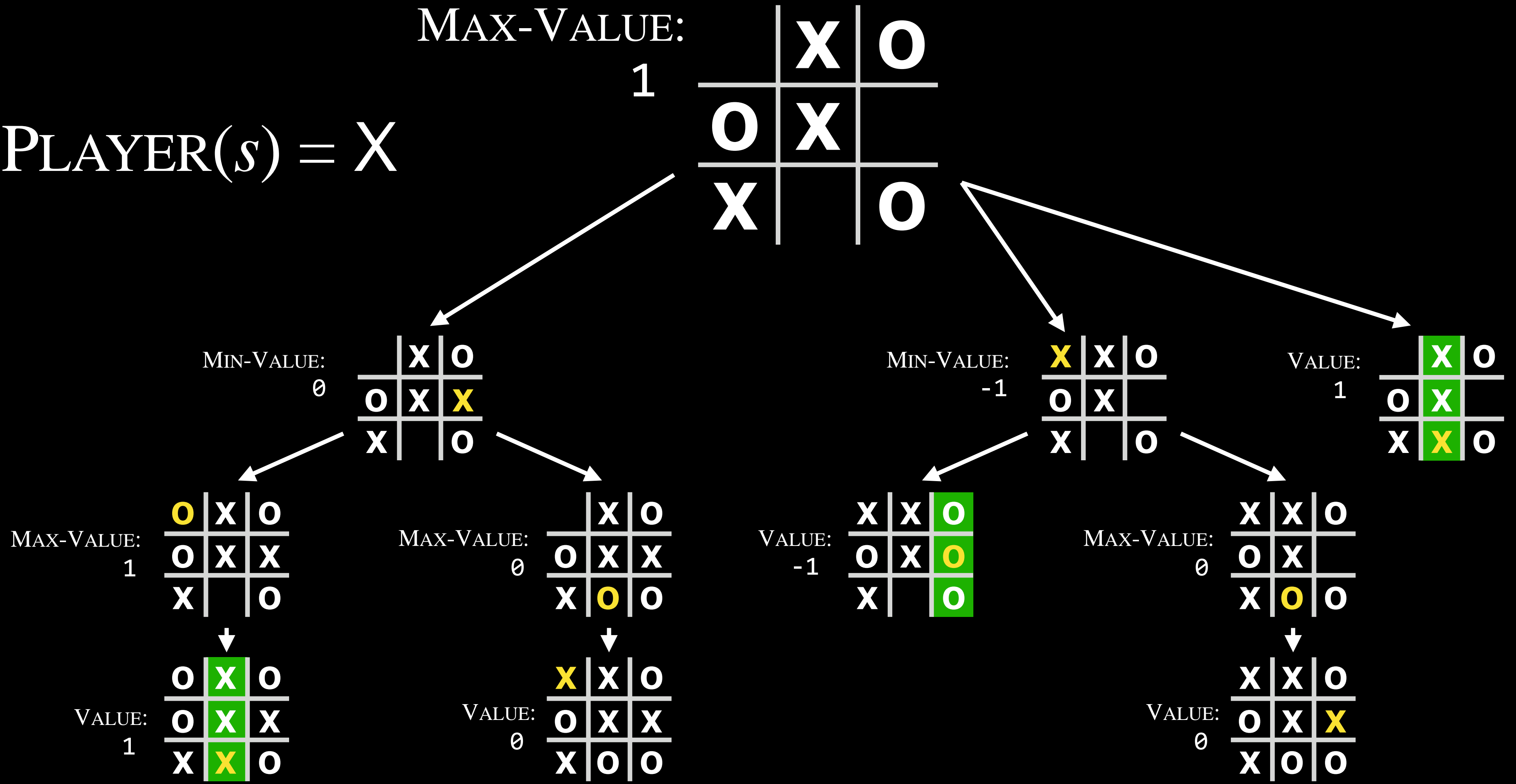
	X	O
O	X	X
X	O	O

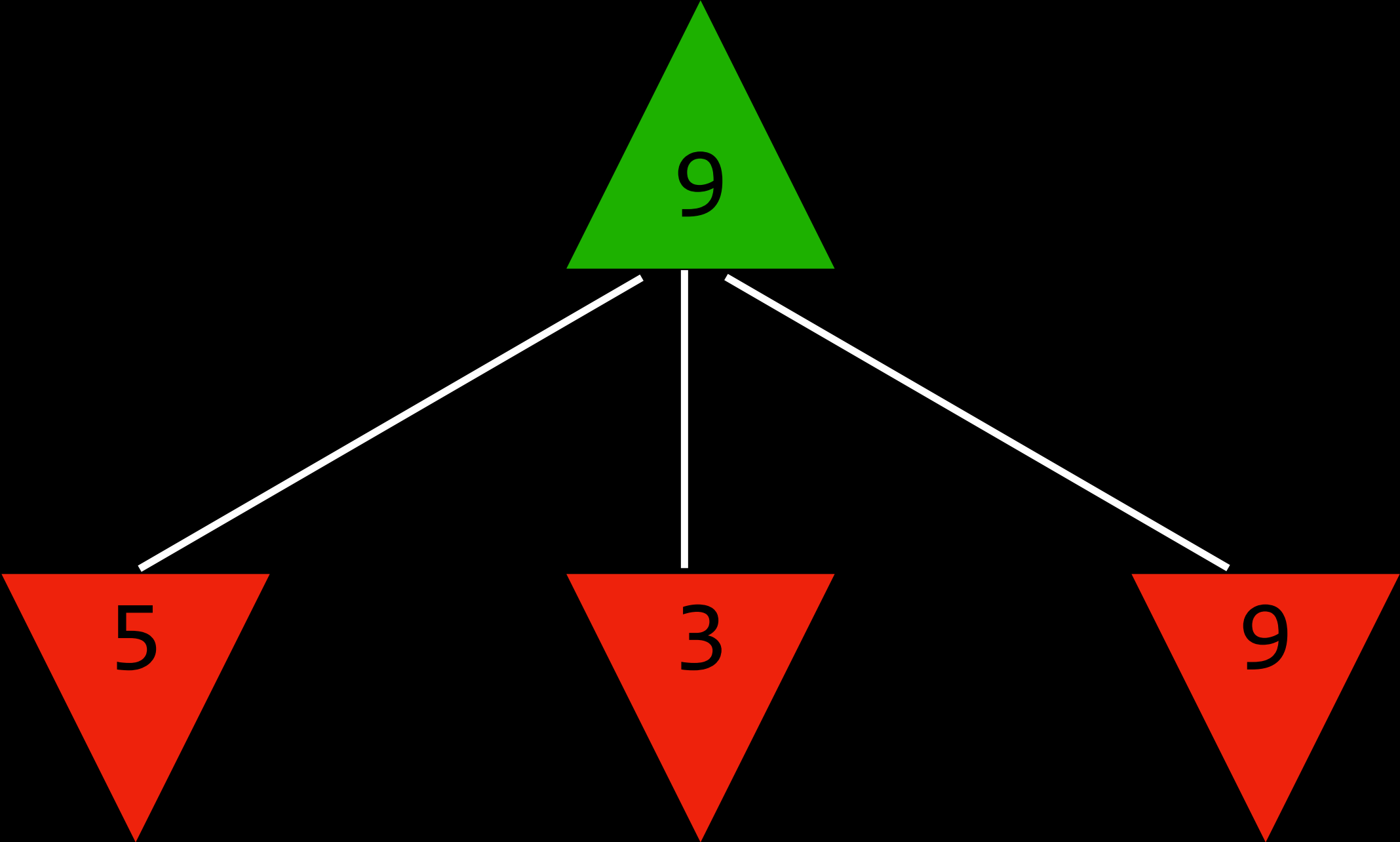
VALUE:
1

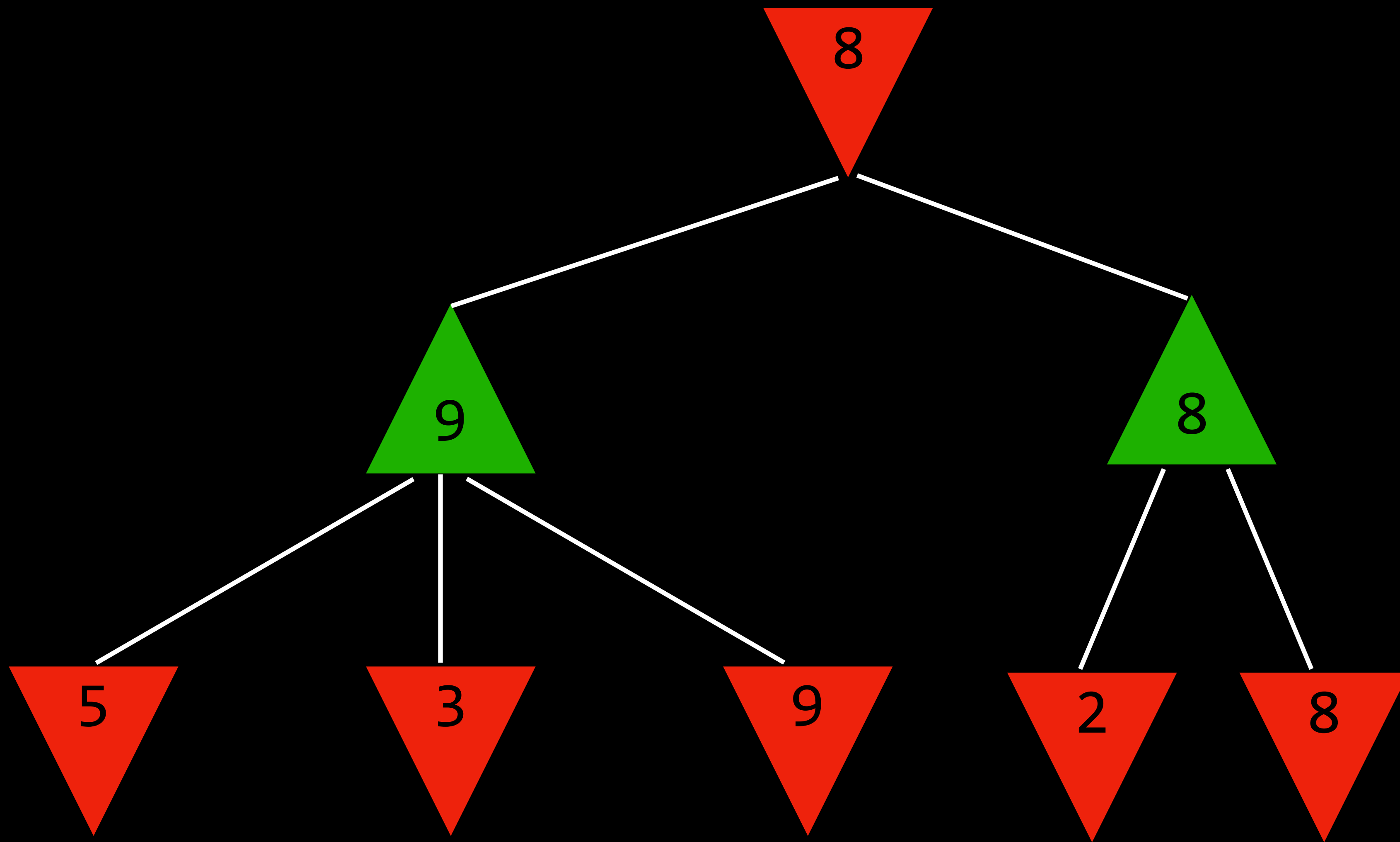
O	X	O
O	X	X
X	X	O

VALUE:
0

X	X	O
O	X	X
X	O	O







Minimax

- Given a state s :
 - MAX picks action a in $ACTIONS(s)$ that produces highest value of $MIN-VALUE(RESULT(s, a))$
 - MIN picks action a in $ACTIONS(s)$ that produces smallest value of $MAX-VALUE(RESULT(s, a))$

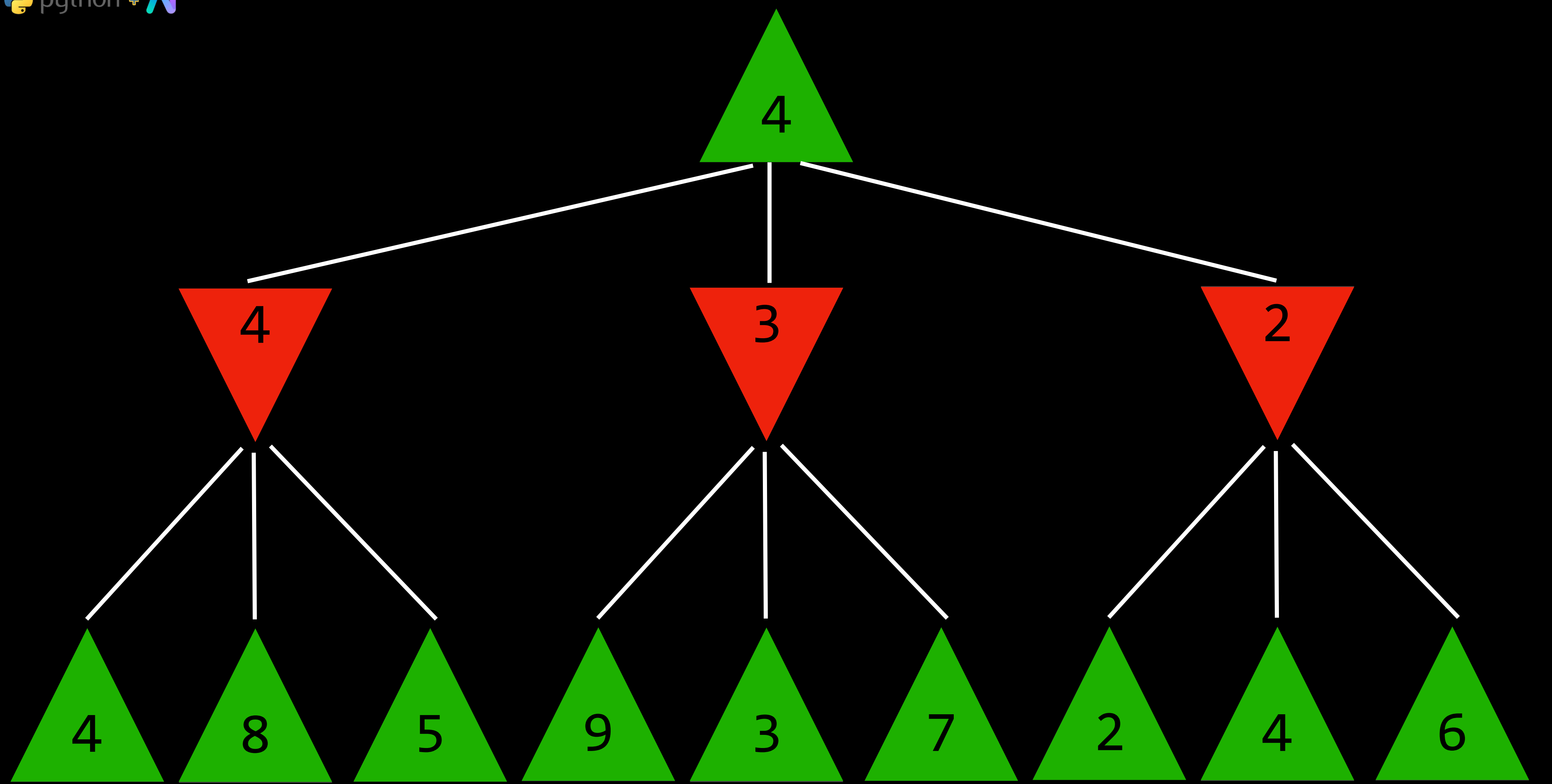
Minimax

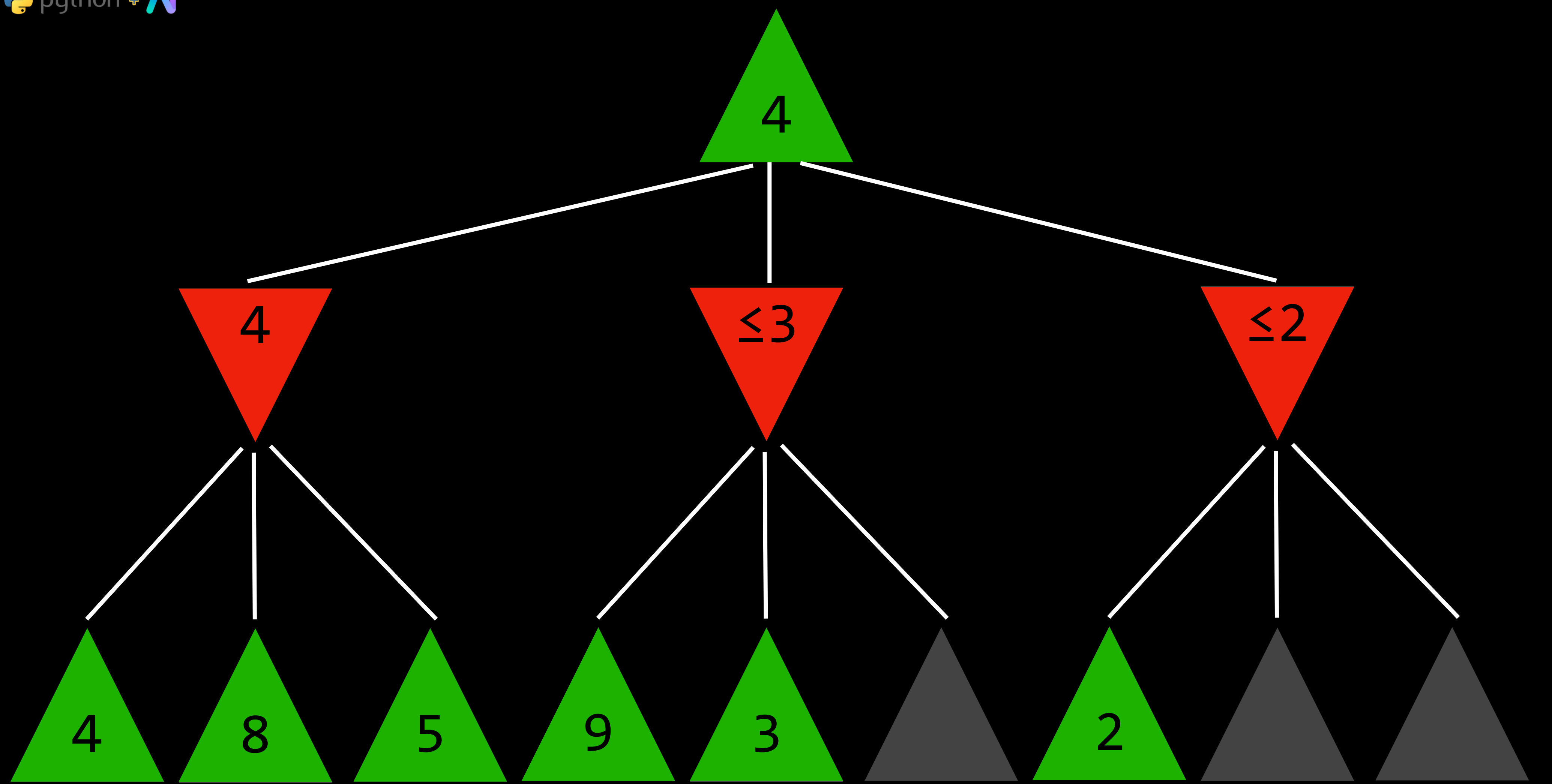
```
function MAX-VALUE(state):  
    if TERMINAL(state):  
        return UTILITY(state)  
     $v = -\infty$   
    for action in ACTIONS(state):  
         $v = \text{MAX}(v, \text{MIN-VALUE}(\text{RESULT}(\textit{state}, \textit{action})))$   
    return  $v$ 
```

Minimax

```
function MIN-VALUE(state):  
    if TERMINAL(state):  
        return UTILITY(state)  
     $v = \infty$   
    for action in ACTIONS(state):  
         $v = \text{MIN}(v, \text{MAX-VALUE}(\text{RESULT}(\textit{state}, \textit{action})))$   
    return  $v$ 
```

Optimizations





Alpha-Beta Pruning

255,168

total possible Tic-Tac-Toe games

288,000,000,000

total possible chess games
after four moves each

10^{29000}

total possible chess games
(lower bound)

 10^{17}

Age of universe
in seconds

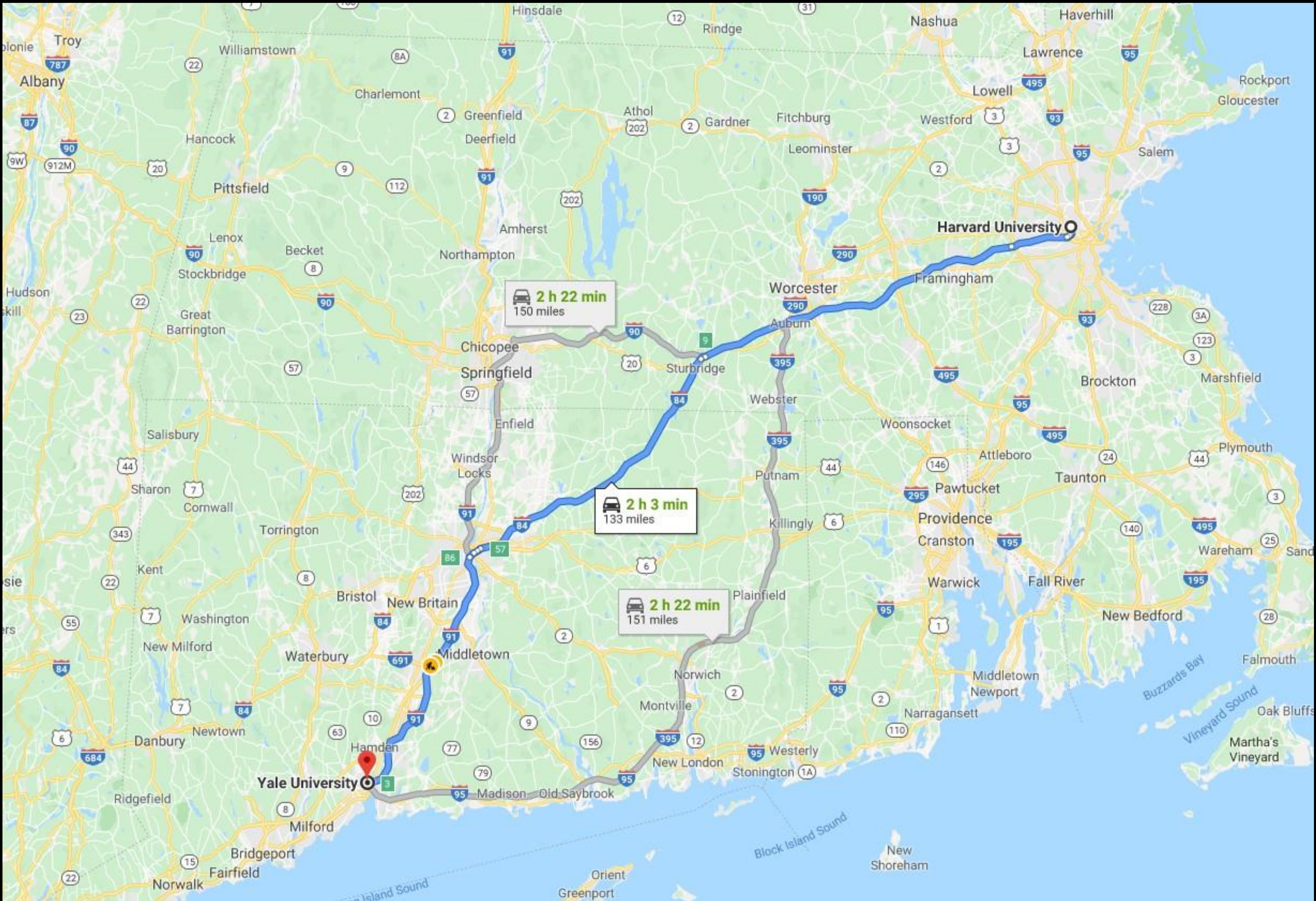
$10^{10^{48}}$

total possible go games
(lower bound)

Depth-Limited Minimax

evaluation function

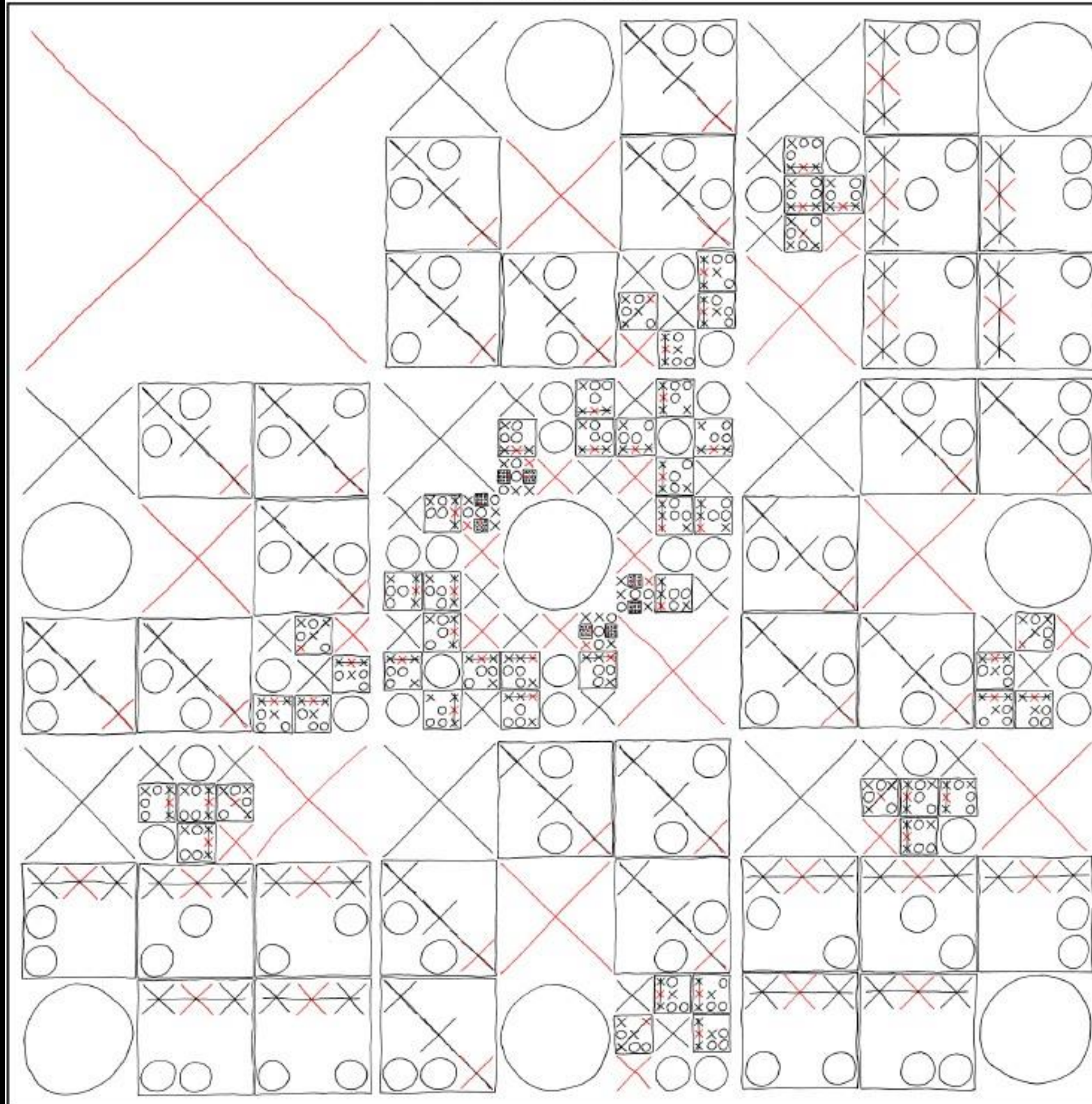
function that estimates the expected utility
of the game from a given state



COMPLETE MAP OF OPTIMAL TIC-TAC-TOE MOVES

YOUR MOVE IS GIVEN BY THE POSITION OF THE LARGEST RED SYMBOL ON THE GRID. WHEN YOUR OPPONENT PICKS A MOVE, ZOOM IN ON THE REGION OF THE GRID WHERE THEY WENT. REPEAT.

MAP FOR X:



Adversarial Search

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