

# COMP250: Artificial Intelligence

## 2: Designing AI behaviours

# Noughts and Crosses



Clone the following repository:

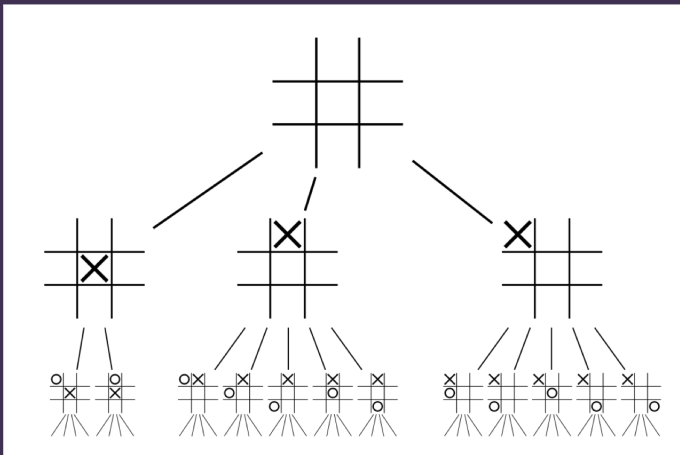
```
https://github.com/Falmouth-Games-Academy/  
bsc-live-coding
```

Open `COMP250/02_oxo` in PyCharm and run `oxo.py`

# Minimax search



# Game trees



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- ▶ Therefore I want to **maximise** the **minimum** value my opponent can achieve

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- ▶ I want to **maximise** the value
- ▶ My opponent wants to **minimise** the value
- ▶ Therefore I want to **maximise** the **minimum** value my opponent can achieve
- ▶ This is generally only true for **two-player zero-sum** games

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- ▶ Consider each possible “next state”, i.e. each possible move
- ▶ If it's my turn, the value is the **maximum** value over next states
- ▶ If it's my opponent's turn, the value is the **minimum** value over next states

# Minimax search pseudocode

**procedure** MINIMAX(state, currentPlayer)

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        bestValue =  $-\infty$

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**for each** possible nextState **do**

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```
procedure MINIMAX(state, currentPlayer)
  if state is terminal then
    return value of state
  else if currentPlayer = 1 then
    bestValue =  $-\infty$ 
    for each possible nextState do
      v = MINIMAX(nextState, 3 - currentPlayer)
```



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```
procedure MINIMAX(state, currentPlayer)
  if state is terminal then
    return value of state
  else if currentPlayer = 1 then
    bestValue =  $-\infty$ 
    for each possible nextState do
       $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$ 
      bestValue = MAX(bestValue, v)
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      bestValue = MAX(bestValue, v)
    end for
    return bestValue
  else if currentPlayer = 2 then
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procedure MINIMAX(state, currentPlayer)
  if state is terminal then
    return value of state
  else if currentPlayer = 1 then
    bestValue =  $-\infty$ 
    for each possible nextState do
       $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$ 
      bestValue = MAX(bestValue,  $v$ )
    end for
    return bestValue
  else if currentPlayer = 2 then
    bestValue =  $+\infty$ 
    for each possible nextState do
       $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$ 
      bestValue = MIN(bestValue,  $v$ )
    end for
    return bestValue
```

# Minimax search pseudocode

```
procedure MINIMAX(state, currentPlayer)
  if state is terminal then
    return value of state
  else if currentPlayer = 1 then
    bestValue =  $-\infty$ 
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       $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$ 
      bestValue = MAX(bestValue,  $v$ )
    end for
    return bestValue
  else if currentPlayer = 2 then
    bestValue =  $+\infty$ 
    for each possible nextState do
       $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$ 
      bestValue = MIN(bestValue,  $v$ )
    end for
    return bestValue
  end if
end procedure
```

# Stopping early

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for each possible nextState do  
     $v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$   
     $\text{bestValue} = \text{MAX}(\text{bestValue}, v)$   
end for
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- State values are always between  $-1$  and  $+1$

# Stopping early

**for each** possible nextState **do**

$v = \text{MINIMAX}(\text{nextState}, 3 - \text{currentPlayer})$

$\text{bestValue} = \text{MAX}(\text{bestValue}, v)$

**end for**

- ▶ State values are always between  $-1$  and  $+1$
- ▶ So if we ever have  $\text{bestValue} = 1$ , we can stop early



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for each possible nextState do  
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- ▶ State values are always between  $-1$  and  $+1$
- ▶ So if we ever have  $\text{bestValue} = 1$ , we can stop early
- ▶ Similarly when minimising if  $\text{bestValue} = -1$

# Using minimax search

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- ▶ To decide what move to play next...
- ▶ Calculate the minimax value for each move
- ▶ Choose the move with the maximum score
- ▶ If there are several with the same score, choose one at random

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- ▶ The game tree for noughts and crosses has only a few thousand states
- ▶ Most games are too large to search fully, e.g. chess has  $\approx 10^{47}$  states
- ▶ Later we will look at **heuristics** and **pruning** to cut down the size of the tree