



**Universitat**  
de les Illes Balears

# Artificial Intelligence

## Lesson 3.3

Games & Adversarial Search



## Chess



### AlphaZero vs. Stockfish



## Shogi



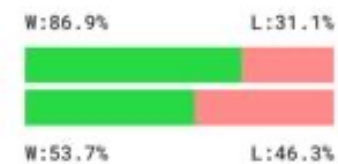
### AlphaZero vs. Elmo



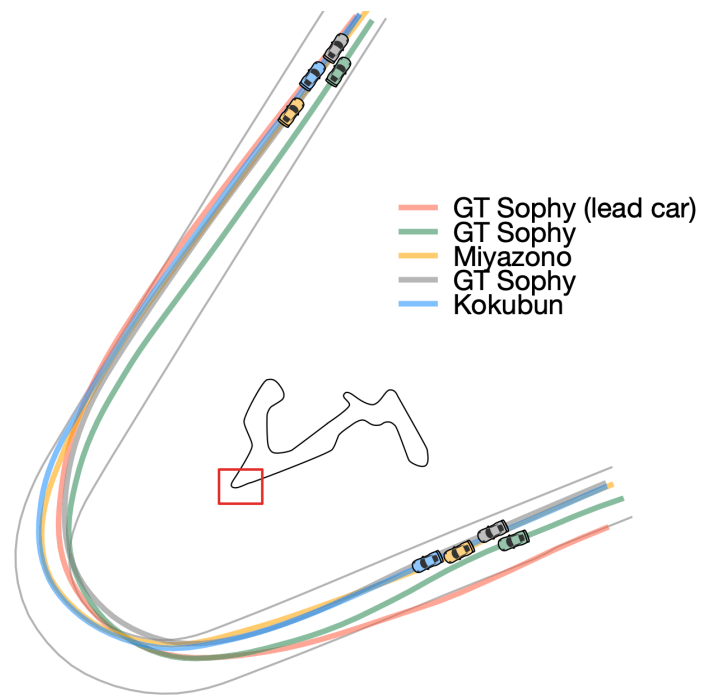
## Go



### AlphaZero vs. AGO



AZ wins AZ draws AZ loses AZ white ○ AZ black ●

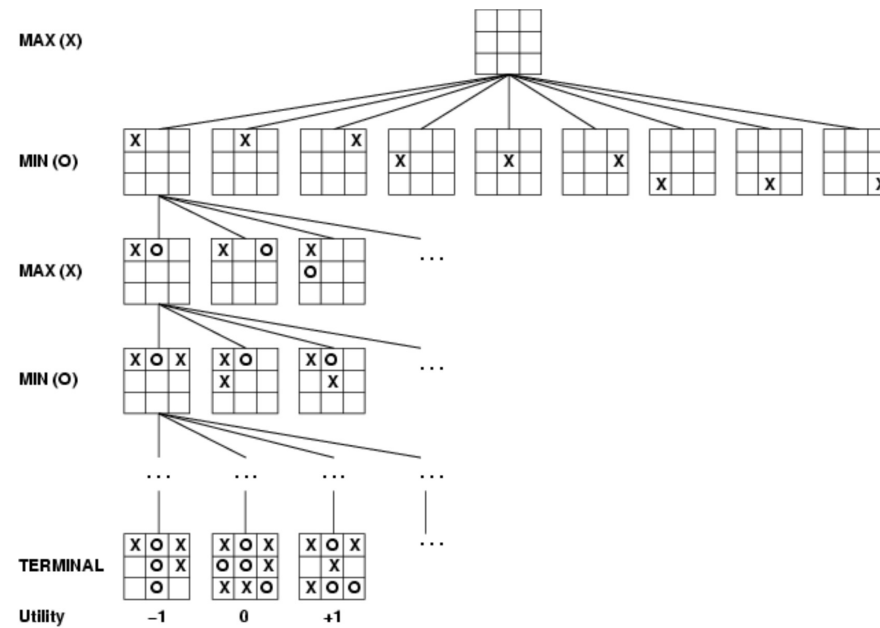


# Games

|                              | <b>Deterministic</b> | <b>Stochastic</b>           |
|------------------------------|----------------------|-----------------------------|
| <b>Perfect information</b>   | Go, Chess, Shogi,..  | Backgammon,<br>Monopoly,... |
| <b>Imperfect information</b> | Battleship,..        | Poker,...                   |

- Assumptions
  - Fully observable environments
  - Deterministic, turn-taking, zero-sum, perfect information
  - Non zero-sum: Prisoner's Dilemma (Nash equilibrium)

# Idea: Game tree



How do we search this tree to find the optimal move?

# Informed search vs adversarial search

- Informed search
  - Solution is a method for finding goal
  - GBFS & A\* algorithms can find a optimal solution
  - Heuristic: estimate cost from start to goal through a given node
- Adversarial search
  - Solution is a strategy (specifies move for every possible opponent reply)
  - Time limits force an approximate solution
  - Heuristic: evaluate “goodness” of game position

# Adversarial search

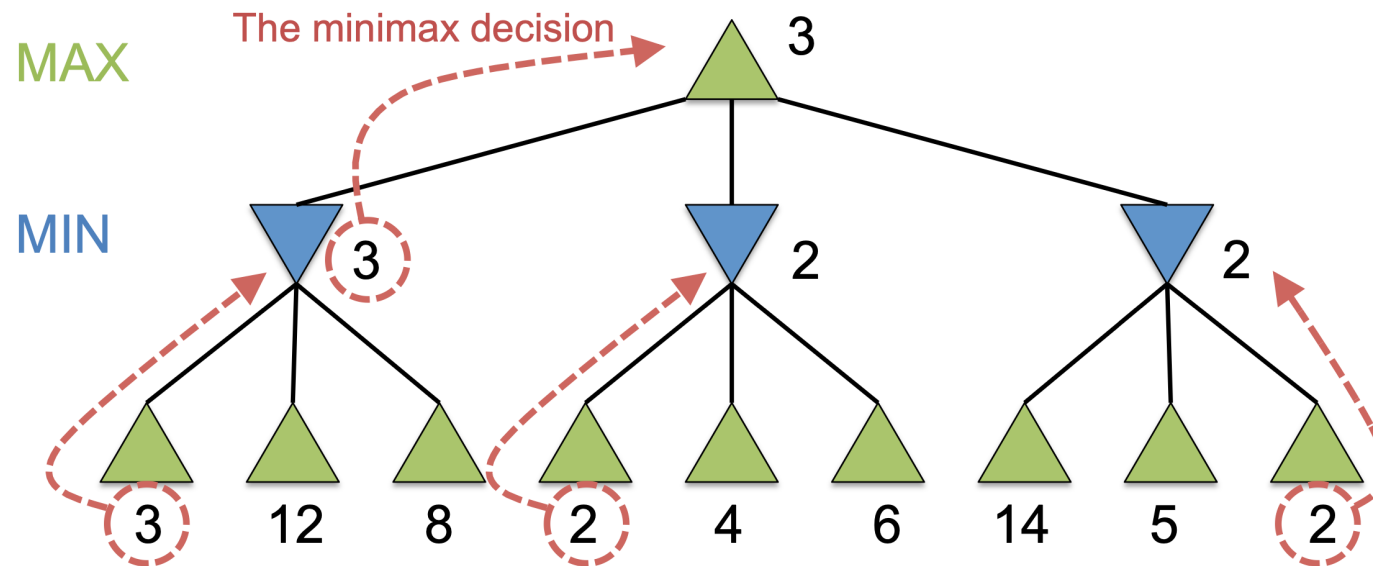
- Formal definition of a game as adversarial search
  - Two players: MAX, MIN
  - Initial state: set-up defined by rules
  - Actions(s): set of legal moves in a state
  - Terminal-Test(s): true if the game is finished; false otherwise
  - Utility(s, p): the numerical value of terminal states for player p
  - MAX uses search tree to determine “best” next move

# Minimax algorithm

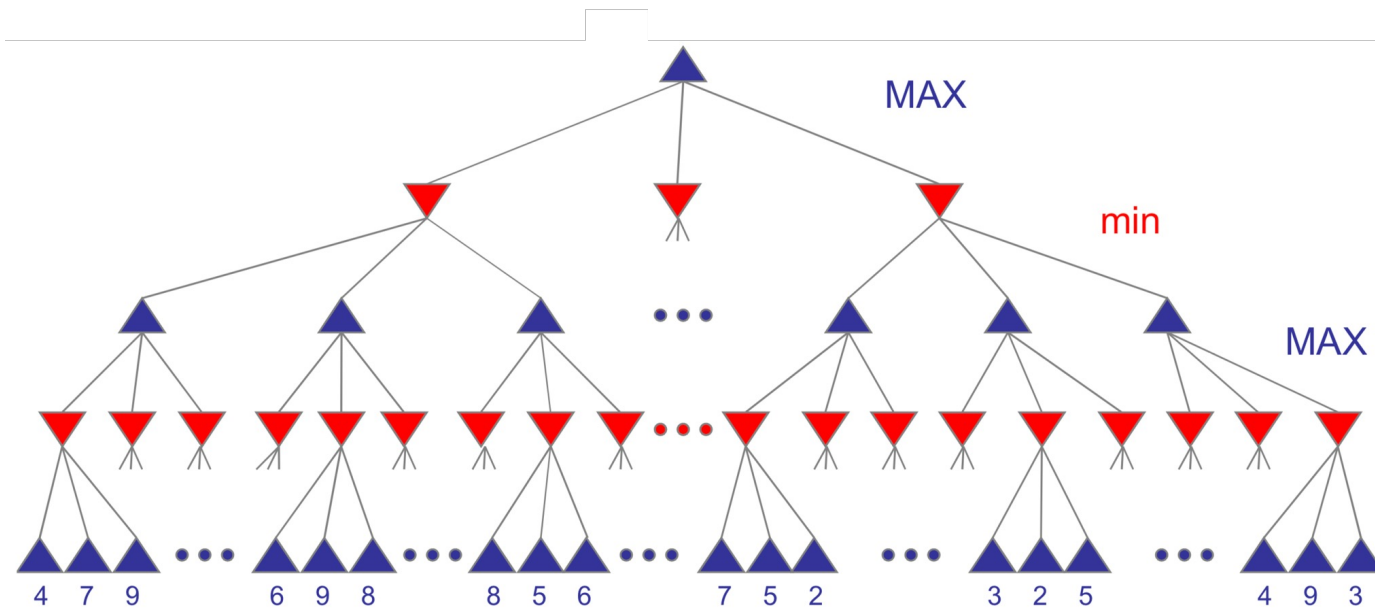
- Optimum algorithm for minimizing the possible loss for a worst case (maximum loss) scenario.
  1. Generate the whole game tree to leaves
  2. Apply heuristic function to leaves
  3. Back-up values from leaves toward the root:
    1. a Max node computes the max of its child values
    2. a Min node computes the min of its child values
  4. At root: choose move leading to the child of highest value



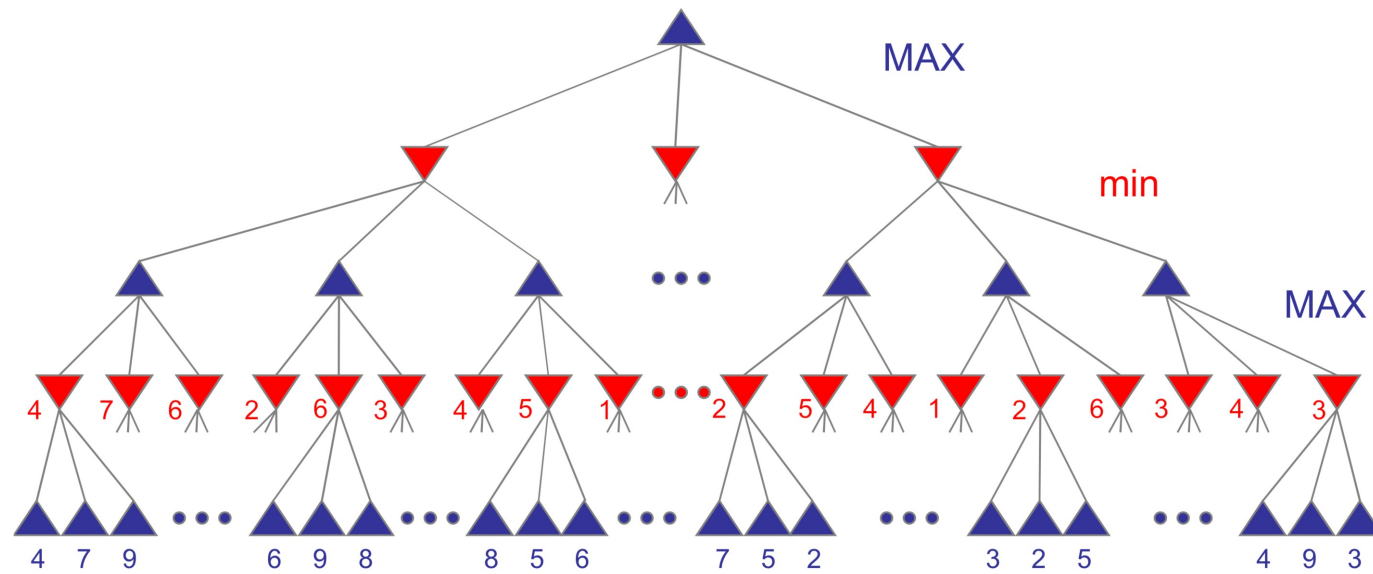
# Example: Two player game tree



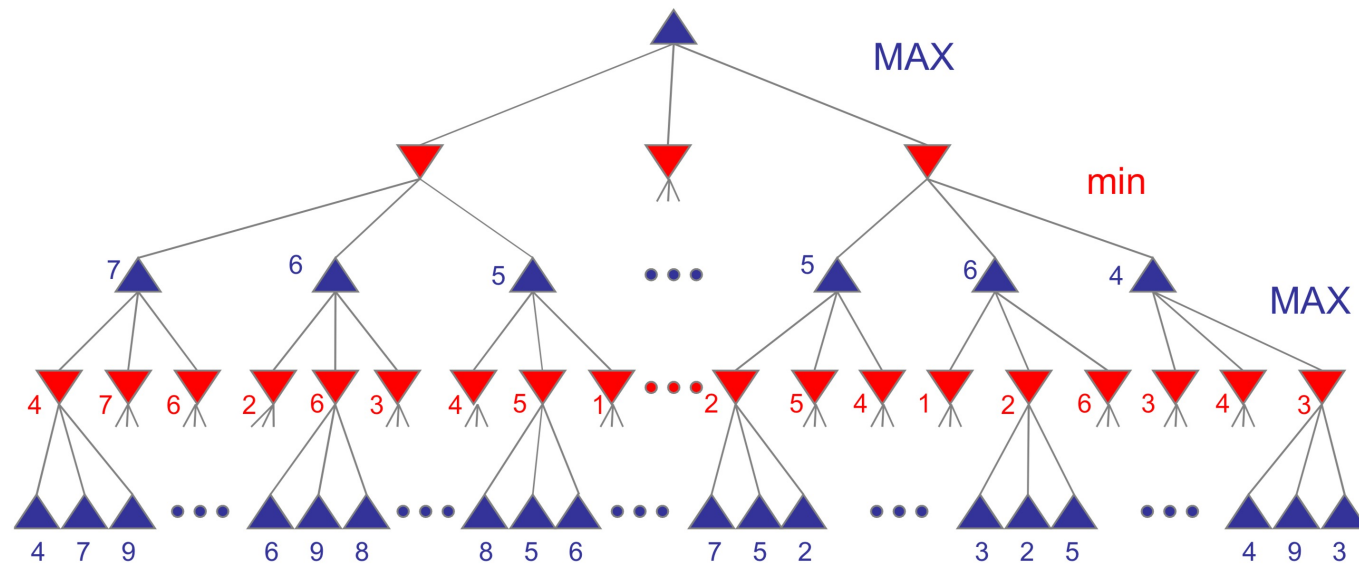
# Example



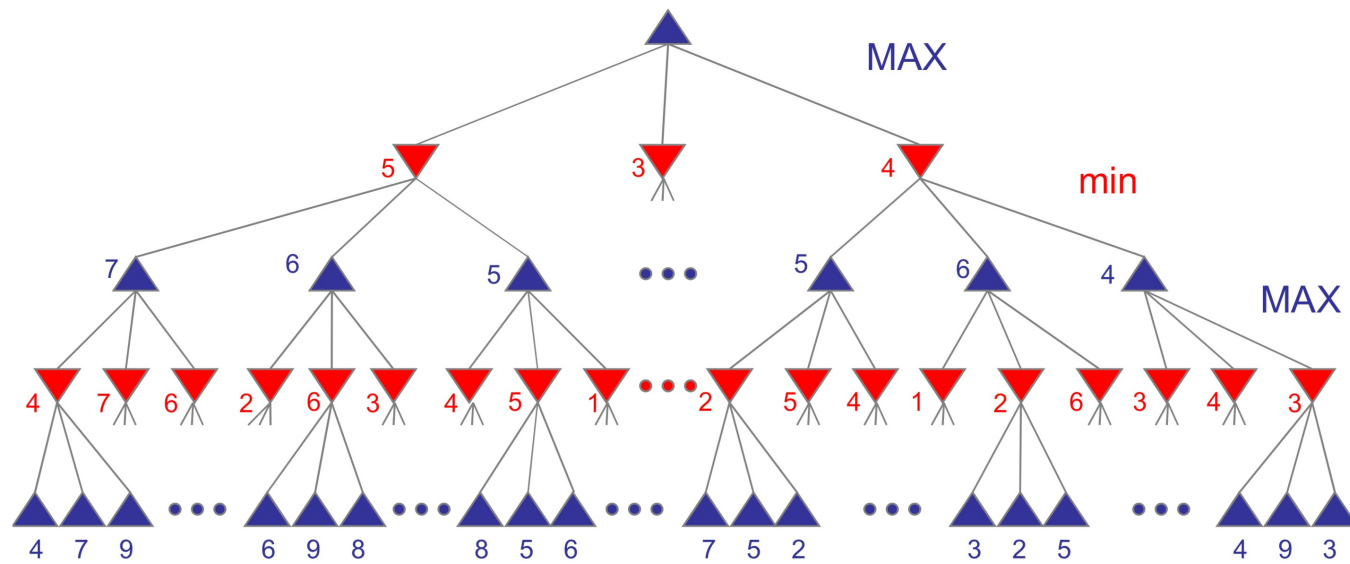
# Example



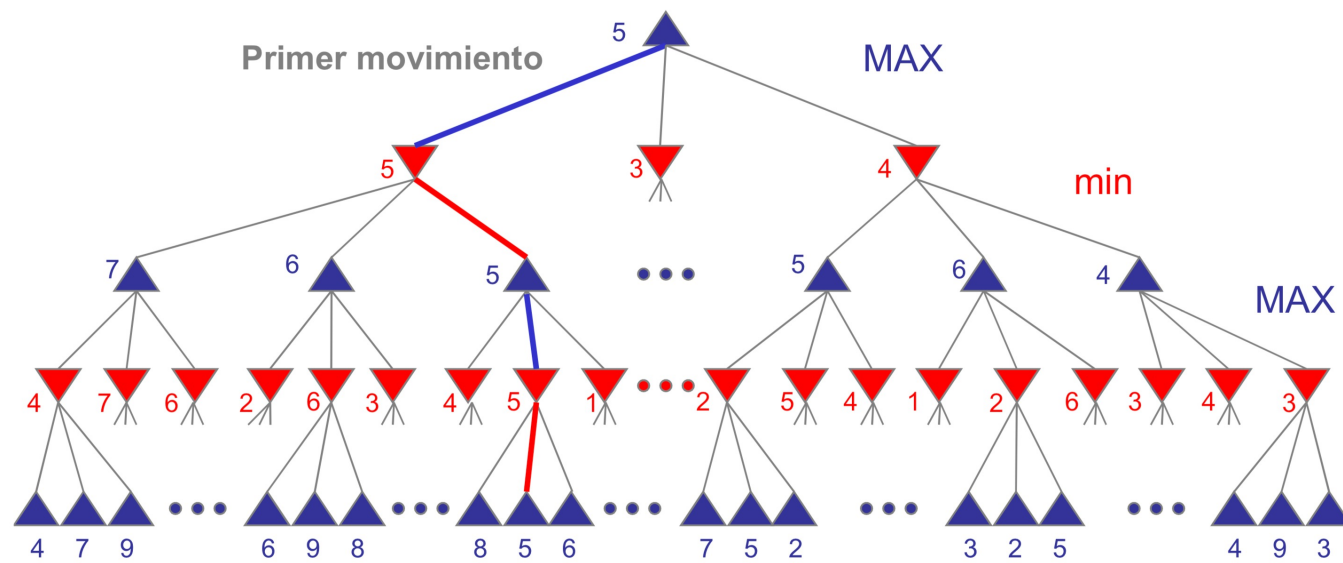
# Example



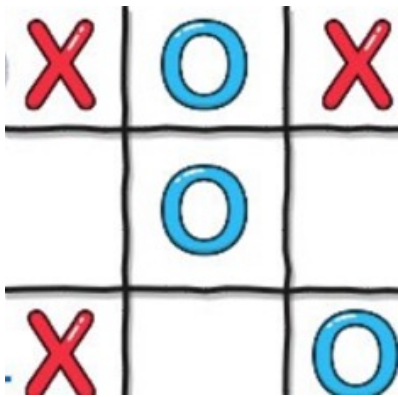
# Example



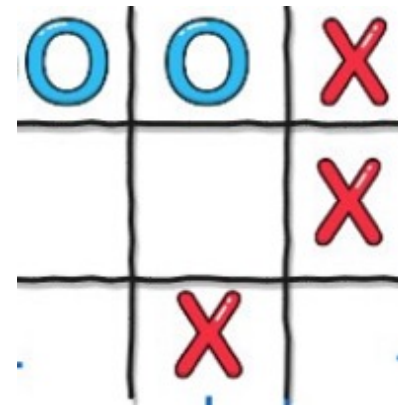
# Example



# Exercise



Best movement for X Player... ?



Who wins?

# Game tree size

- Tic-tac-toe
  - Approx branching factor: 5
  - Approx depth: 9
  - $b^d = 5^9$ , exact solution is quite reasonable

- Chess
  - Approx branching factor: 35
  - Approx depth: 100
  - $b^d = 35^{100}$ , exact solution is infeasible!

1) Limited depth

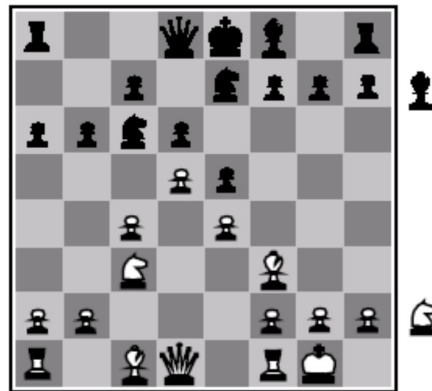
2) Alpha-beta pruning!



# Static Heuristics

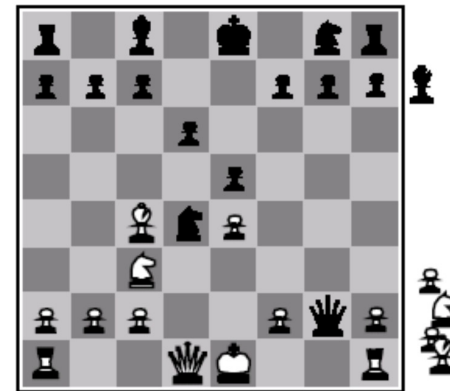
- A static heuristic for adversarial search
  - Estimate how good the current board configuration is for a player.
  - Typically, evaluate how good it is for the player, and how good it is for the opponent, and subtract the opponent's score from the player's.
  - Often called “static” because it is called on a static board position
  - Ex: Othello: Number of white pieces - Number of black pieces
  - Ex: Chess: Value of all white pieces - Value of all black pieces
  - Board evaluation:  $X$  for one player  $-X$  for opponent

# Chess...



Black to move

White slightly better



White to move

Black winning

For chess, typically *linear* weighted sum of *features*

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

e.g.,  $w_1 = 9$  with

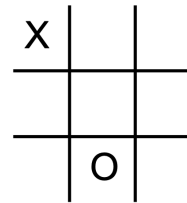
$f_1(s) = (\text{number of white queens}) - (\text{number of black queens}), \text{ etc.}$

# Tic-tac-toe...

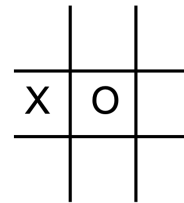
- Heuristic?
- Best first move?

# Tic-tac-toe...

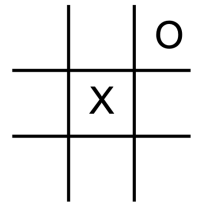
- Heuristic?
  - Number of possible win lines



H= 6-5



H= 4-6



H= 5-4

- Best first move?

- Heuristic?

- Heuristic?



# Alpha-beta pruning

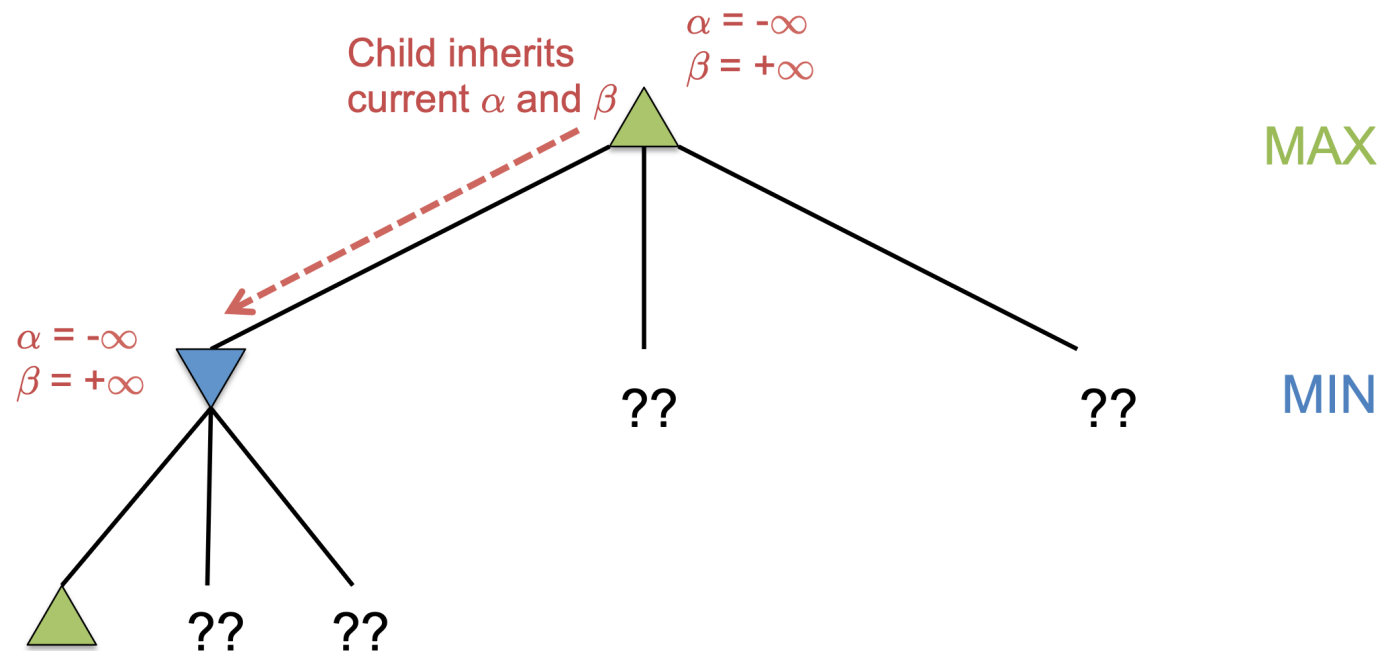
- Minimax problem: exponential states
- **Idea**: it is possible to calculate the correct minimax decision without looking at all the nodes in the tree
  - If a position is provably bad
    - It's no use searching to find out just how bad
  - If the adversary can force a bad position
    - It's no use searching to find the good positions the adversary won't let you achieve
- **Alpha-beta pruning** allows you to remove large parts of the tree, without influencing the final decision
  - An example

# Alpha-beta pruning

- Each node is analysed taking into account its value and the value that its parent currently has.
- This determines at each moment one interval  $(\alpha, \beta)$  of possible values that the node could take
- Intuitive meaning of  $\alpha$  and  $\beta$  at each moment:
  - MAX nodes:  $\alpha$  is the current value of the node (which will have that or more) and  $\beta$  is the current value of the parent (which will have that or less)
  - MIN nodes:  $\beta$  is the current value of the node (which will have that or less) and  $\alpha$  is the current value of the parent (which will have that or more)
- Pruning occurs if at some node  $\alpha \geq \beta$ :
  - There is no need to analyse the remaining successors of the node
  - At MIN nodes, it is called  $\beta$  pruning and at MAX nodes, pruning  $\alpha$

# Alpha-beta pruning

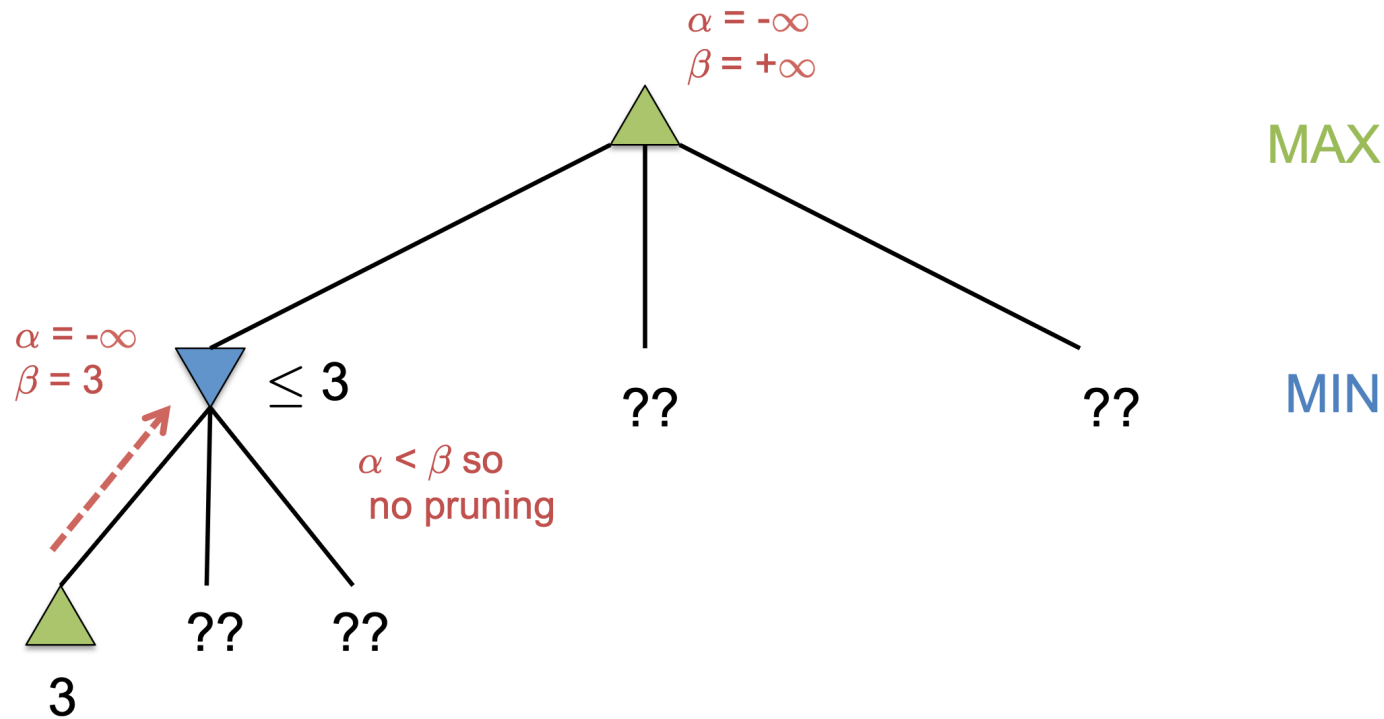
- Initially, possibilities are unknown: range ( $\alpha = -\infty, \beta = \infty$ )
- Do a depth-first search to the first leaf





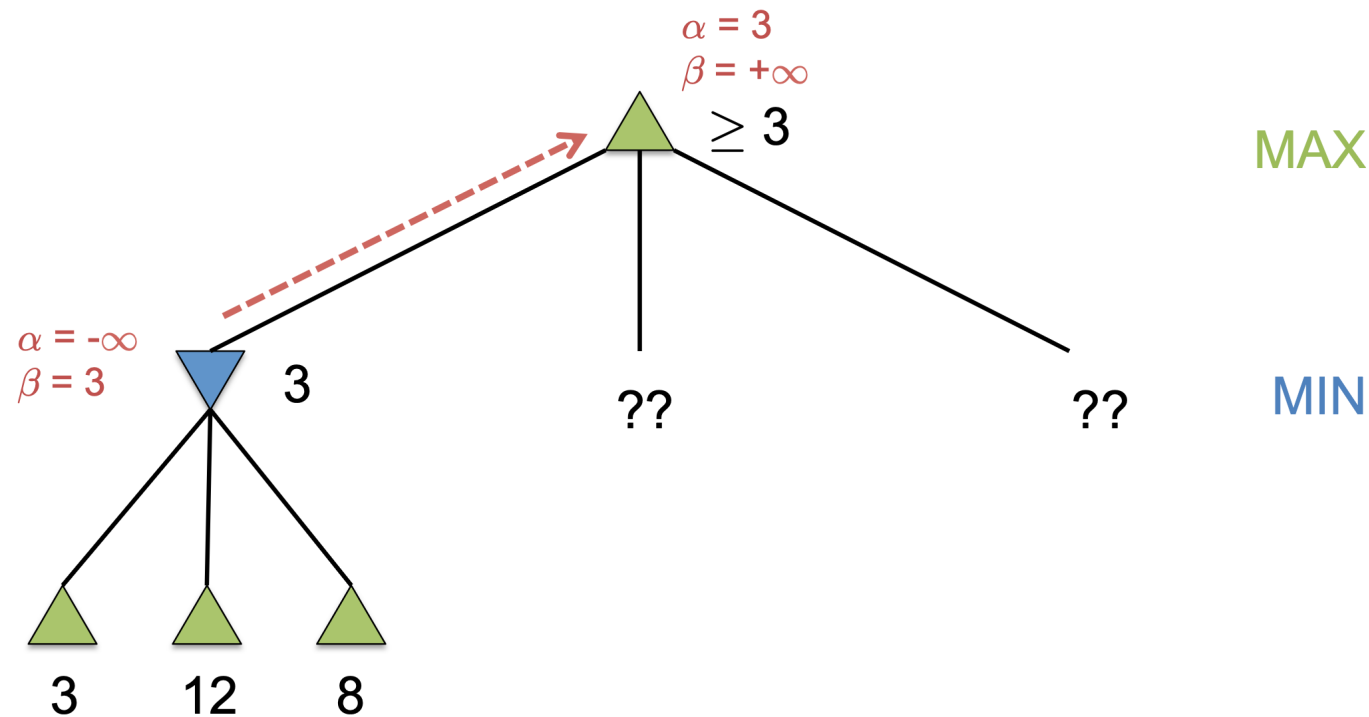
# Alpha-beta pruning

- See the first leaf, after MIN's move: MIN updates  $\beta$



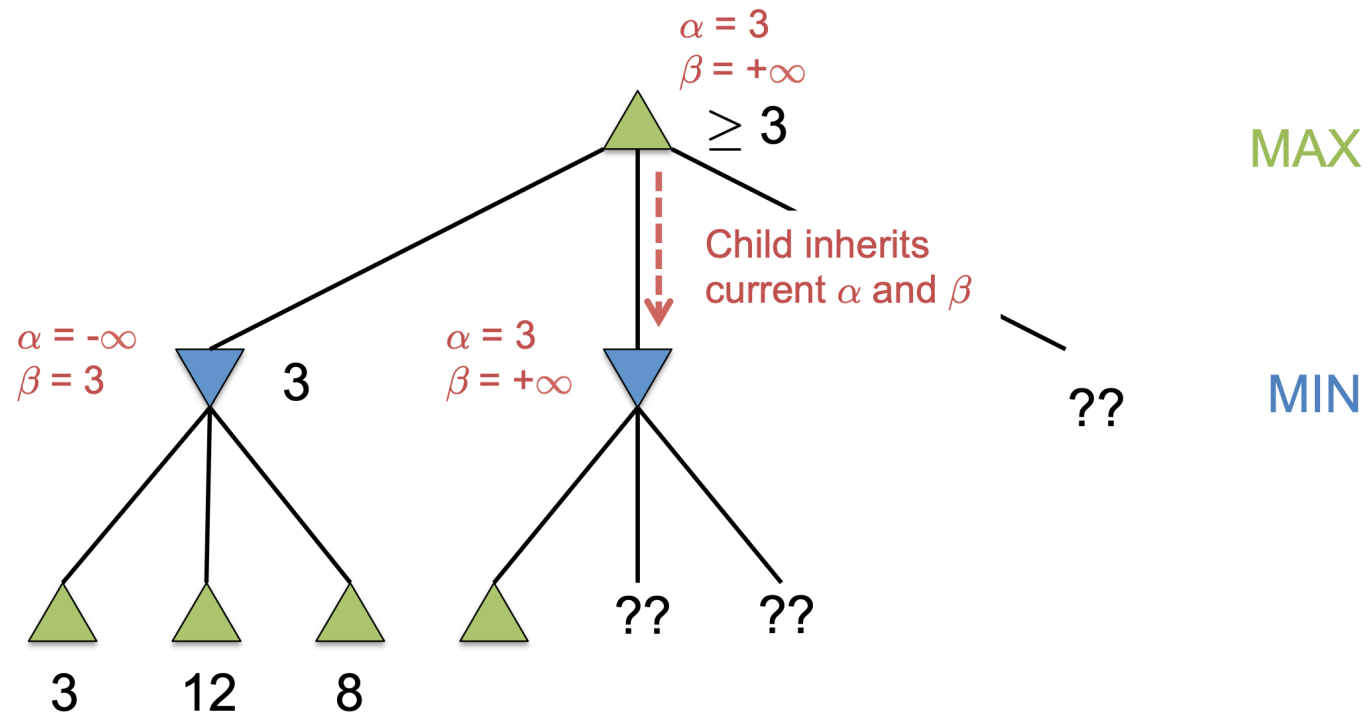
# Alpha-beta pruning

- See remaining leaves; value is known
- Pass outcome to caller; MAX updates  $\alpha$



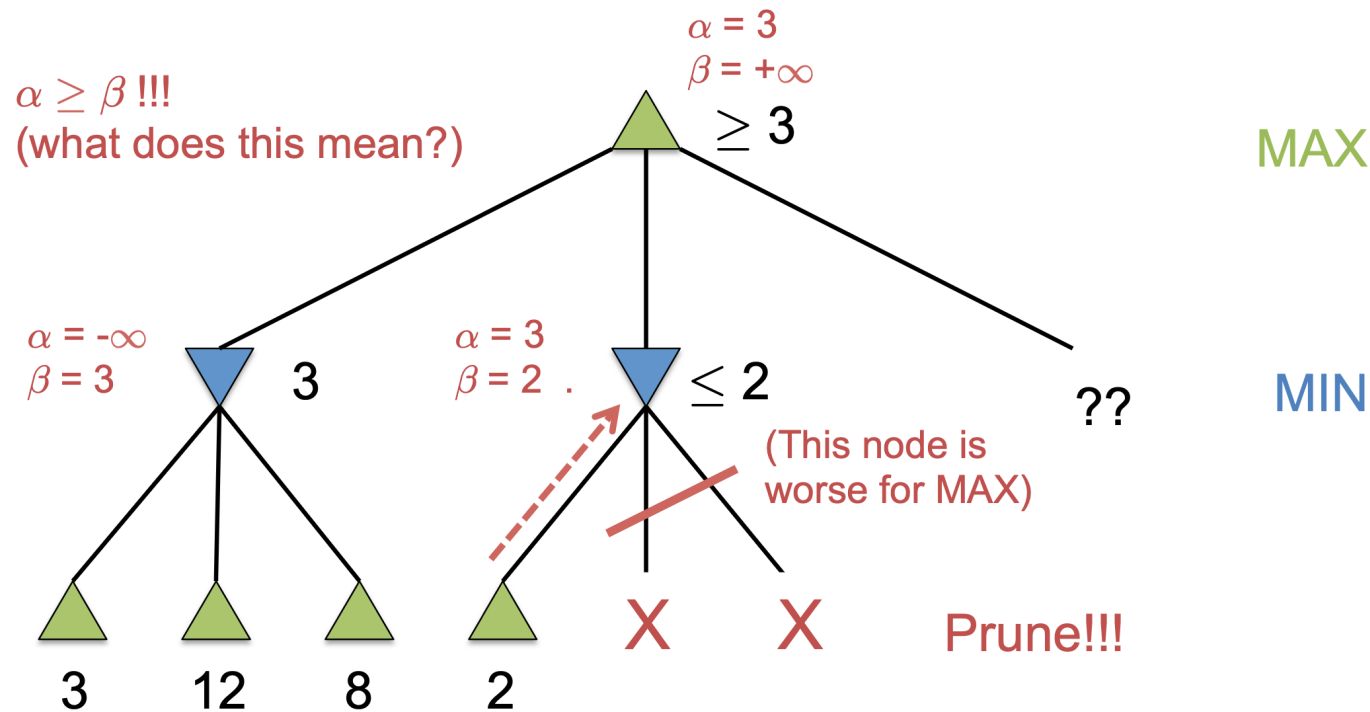
# Alpha-beta pruning

- Continue depth-first search to next leaf.
- Pass  $\alpha, \beta$  to descendants

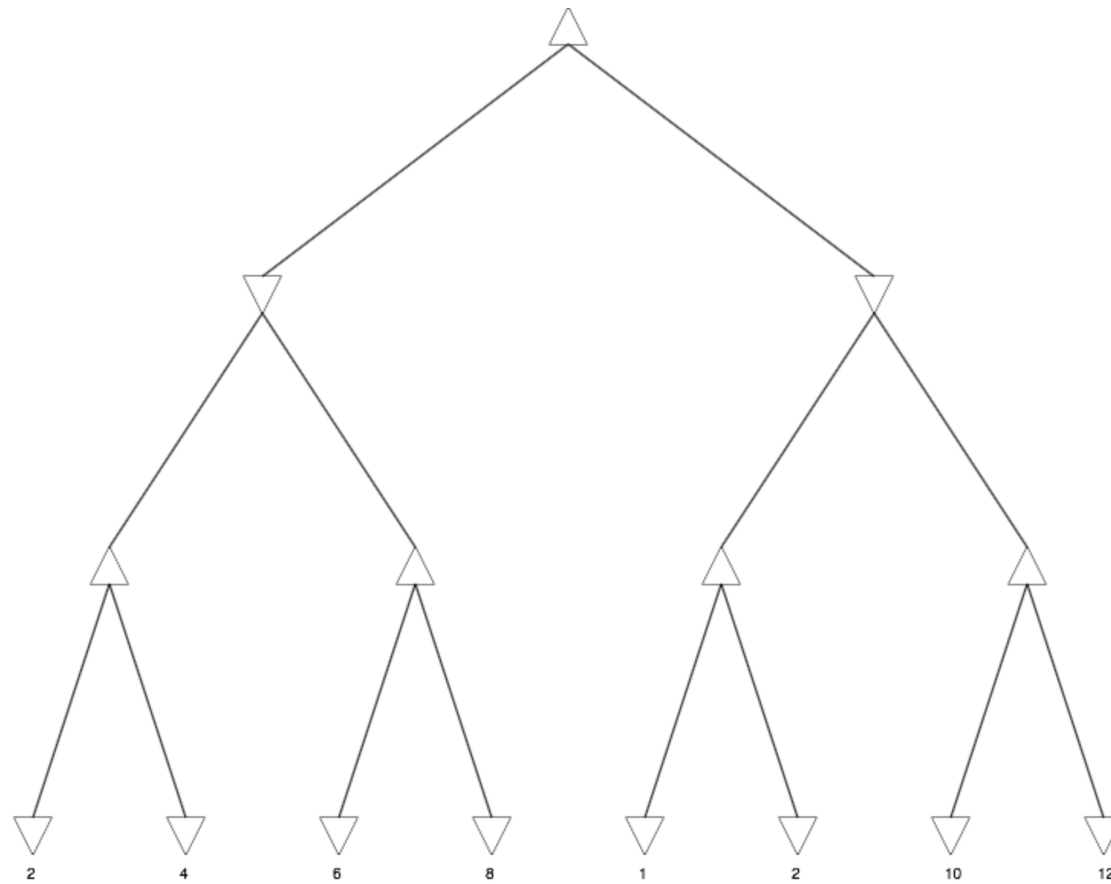


# Alpha-beta pruning

- Observe leaf value; MIN's level; MIN updates beta Prune
- play will never reach the other nodes!



# Exercise...



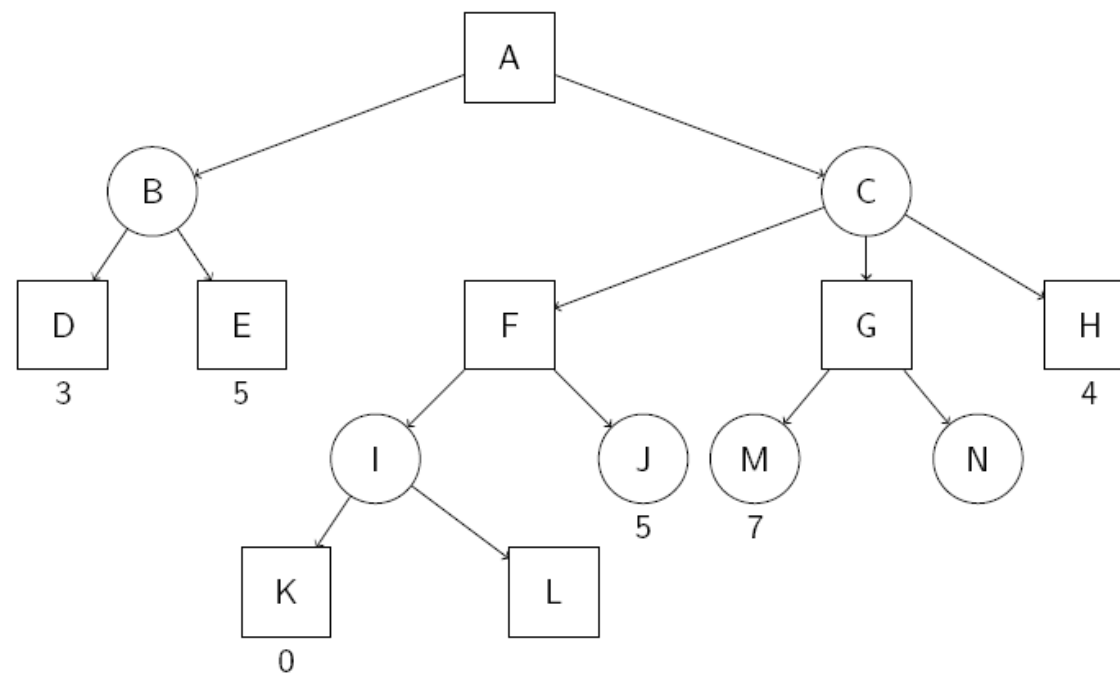
# Effectiveness of alpha-beta search

- Worst case: Minimax
- In practice often get  $O(b^{d/2})$  rather than  $O(b^d)$
- This is the same as having a branching factor of  $\sqrt{b}$ ,
- In chess go from  $b \sim 35$  to  $b \sim 6$
- Permitting much deeper search in the same amount of time

# Summary

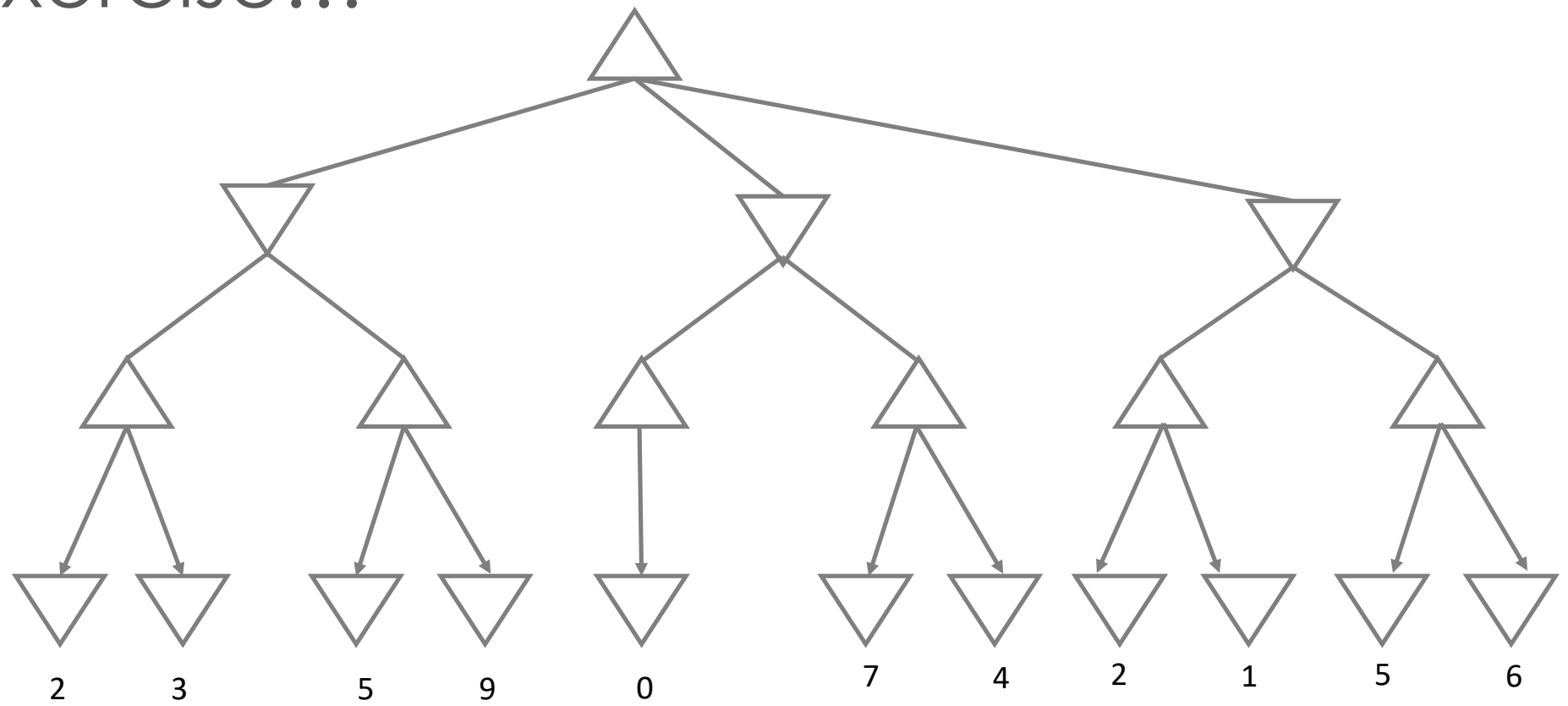
- Game playing could be modelled as a search problem
- Game trees represent alternate computer/opponent moves
- Heuristic functions estimate the quality of a given board configuration for the MAX player.
- Minimax is a procedure which chooses moves by assuming that the opponent will always choose the move which is best for them
- Alpha-Beta is a procedure which can prune large parts of the search tree and allow search to go deeper
- For many well-known games, computer algorithms based on heuristic search match or out-perform human world experts.

# Exercise...





# Exercise...



# Exercise: Chomp game

Consider the following game:

- Two players have a bar of chocolate with  $m \times n$  squares.
- The square in the top left corner is known to be poisonous.
- The players play in turn, where the rule is: a player chooses one entire row or column (of 1 or more squares)
- Goal: the player who has to eat the poisonous piece loses.
- Solve for a bar of chocolate with  $3 \times 2$  squares, that is

Hint: possible first actions

