

# 21CS2213RA

## AI for Data Science

### Session -8

**Contents:** *Adversarial Search*

# Adversarial Search

- Examine the problems that arise when we try to plan ahead in a world where other agents are planning against us.
- A good example is in board games.
- Adversarial games, while much studied in AI, are a small part of game theory in economics.

# Typical AI assumptions

- Two agents whose actions alternate
- Utility values for each agent are the opposite of the other
  - creates the adversarial situation
- Fully observable environments
- In game theory terms: Zero-sum games of perfect information.
- We'll relax these assumptions later.

# Search versus Games

- Search – no adversary
  - Solution is (heuristic) method for finding goal
  - Heuristic techniques can find *optimal* solution
  - Evaluation function: estimate of cost from start to goal through given node
  - Examples: path planning, scheduling activities
- Games – adversary
  - Solution is **strategy** (strategy specifies move for every possible opponent reply).
  - **Optimality depends on opponent.**
  - Time limits force an *approximate* solution
  - Evaluation function: evaluate “goodness” of game position
  - Examples: chess, checkers, Othello, backgammon

# Types of Games

	deterministic	Chance moves
Perfect information	Chess, checkers, go, othello	Backgammon, monopoly
Imperfect information (Initial Chance Moves)	Bridge, Skat	Poker, scrabble, blackjack

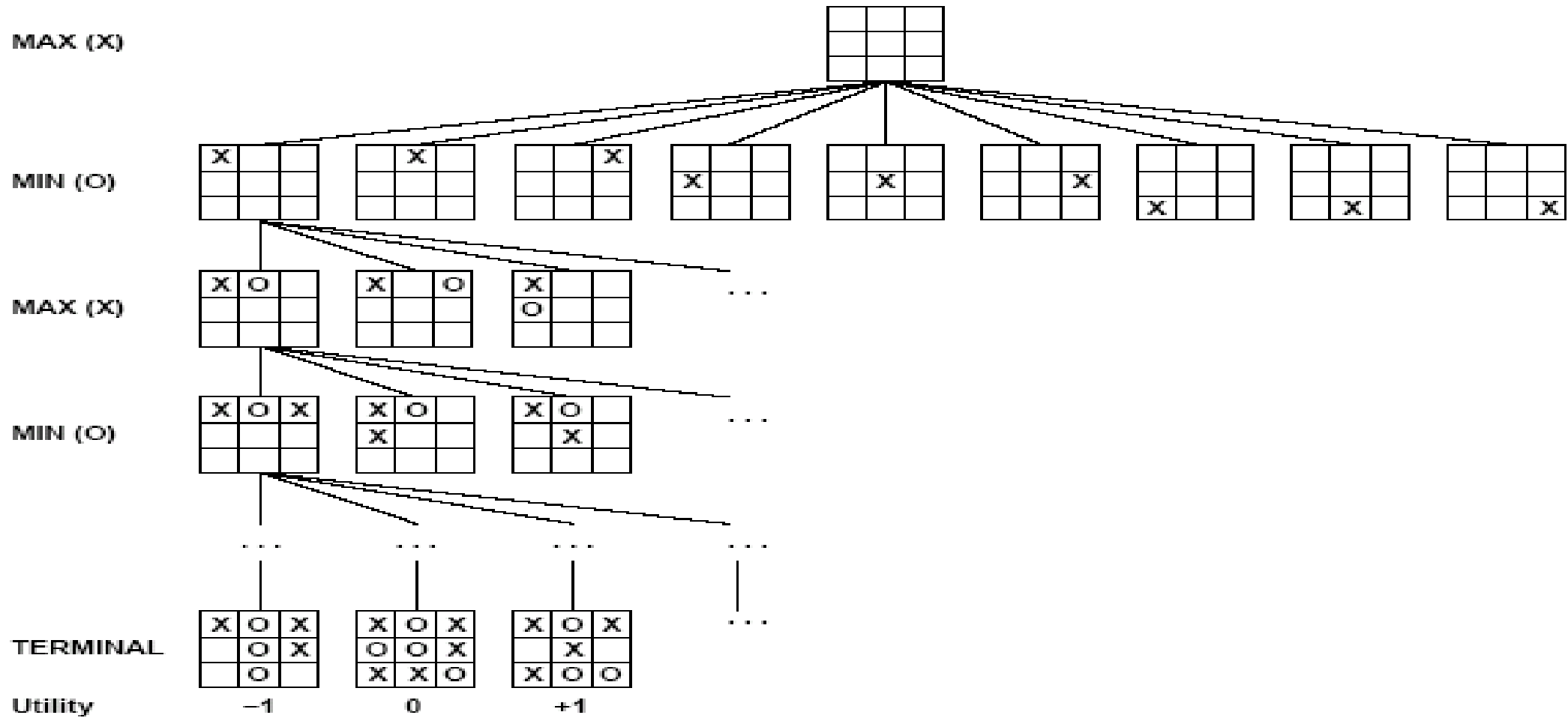
# Game Setup

- Two players: MAX and MIN
- MAX moves first and they take turns until the game is over
  - Winner gets award, loser gets penalty.
- Games as search:
  - Initial state: e.g. board configuration of chess
  - Successor function: list of (move,state) pairs specifying legal moves.
  - Terminal test: Is the game finished?
  - Utility function: Gives numerical value of terminal states. E.g. win (+1), lose (-1) and draw (0) in tic-tac-toe or chess

# Size of search trees

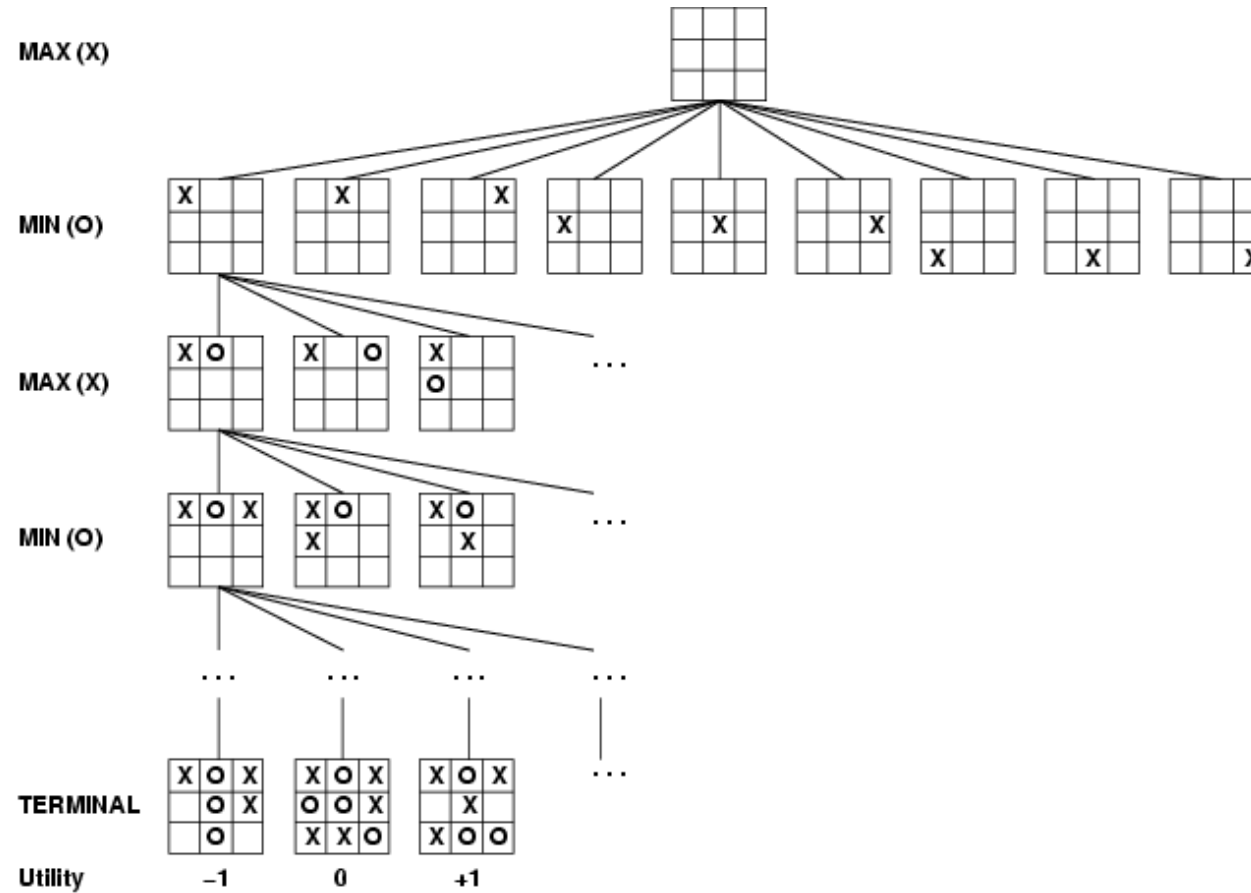
- $b$  = branching factor
- $m$  = number of moves by both players
- Search tree is  $O(b^m)$
- Chess
  - $b \sim 35$
  - $m \sim 100$ 
    - search tree is  $\sim 10^{154}$  (!!)
    - completely impractical to search this
- Game-playing emphasizes being able to make optimal decisions in a finite amount of time
  - Somewhat realistic as a model of a real-world agent
  - Even if games themselves are artificial

## Partial Game Tree for Tic-Tac-Toe





# Game tree (2-player, deterministic, turns)

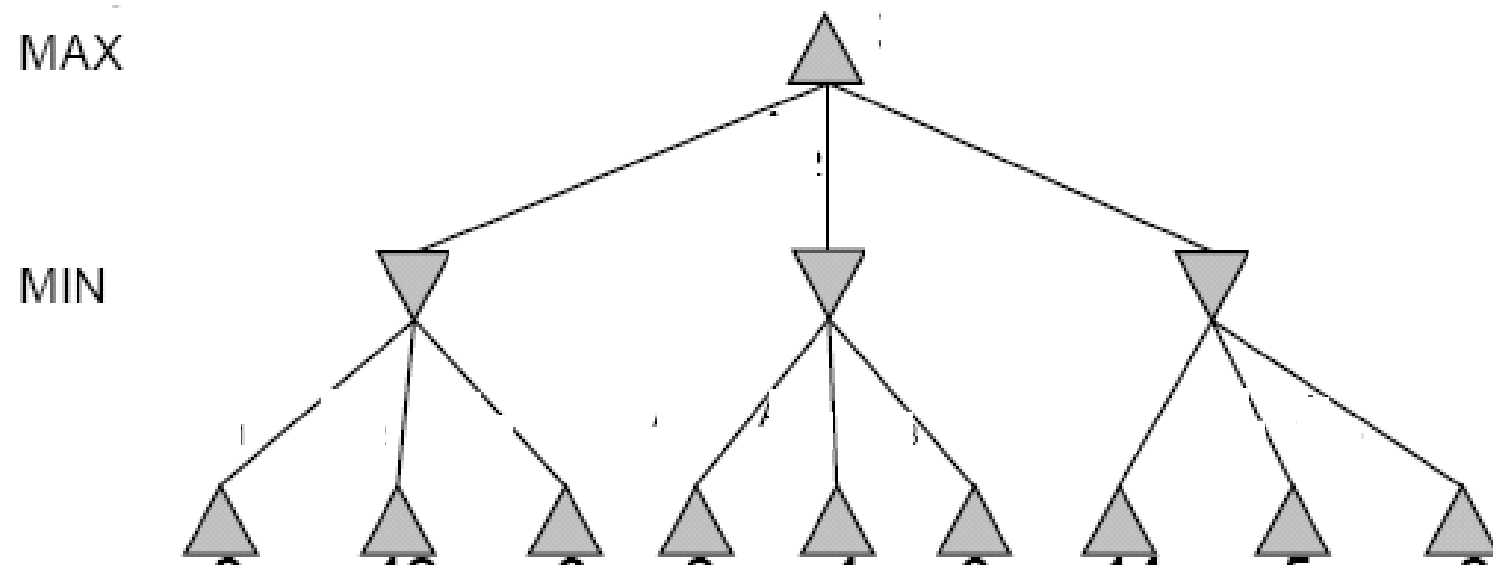


How do we search this tree to find the optimal move?

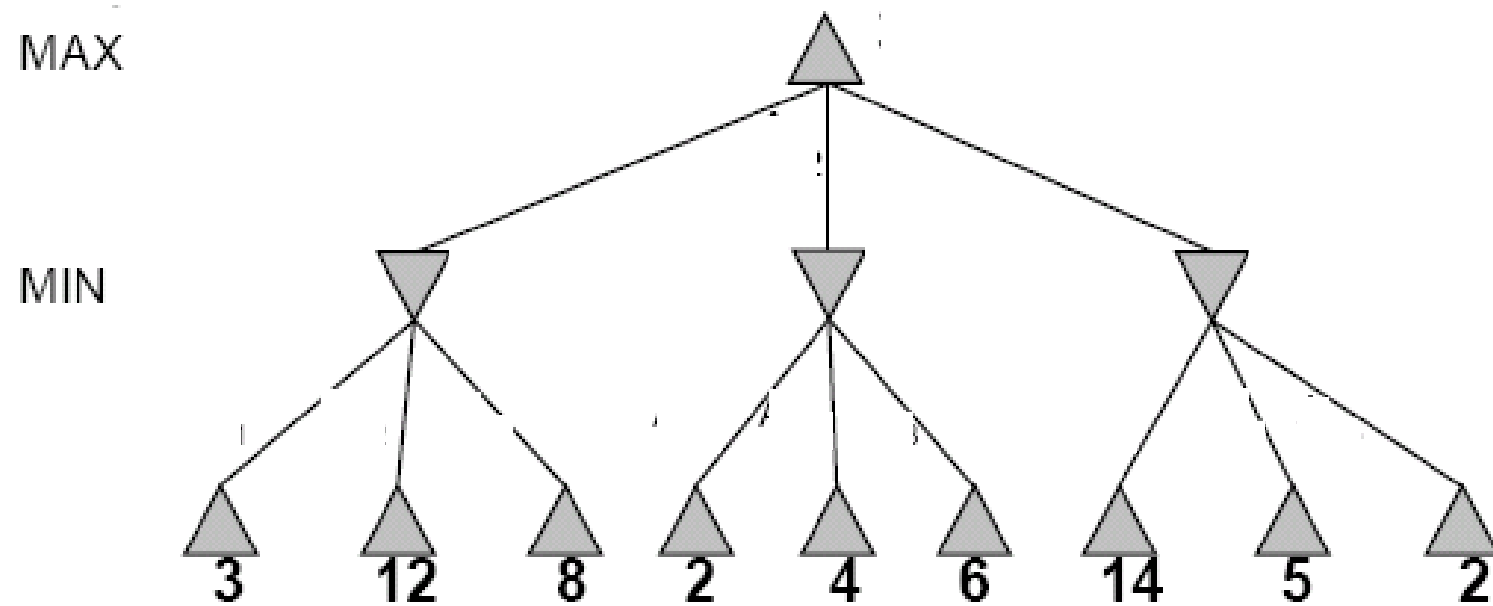
## Minimax strategy:

- Find the optimal *strategy* for MAX assuming an infallible MIN opponent
  - Need to compute this all the down the tree
  - [Game Tree Search Demo](#)
- Assumption: Both players play optimally!
- Given a game tree, the optimal strategy can be determined by using the **minimax value of each node**.
- Zermelo 1912.

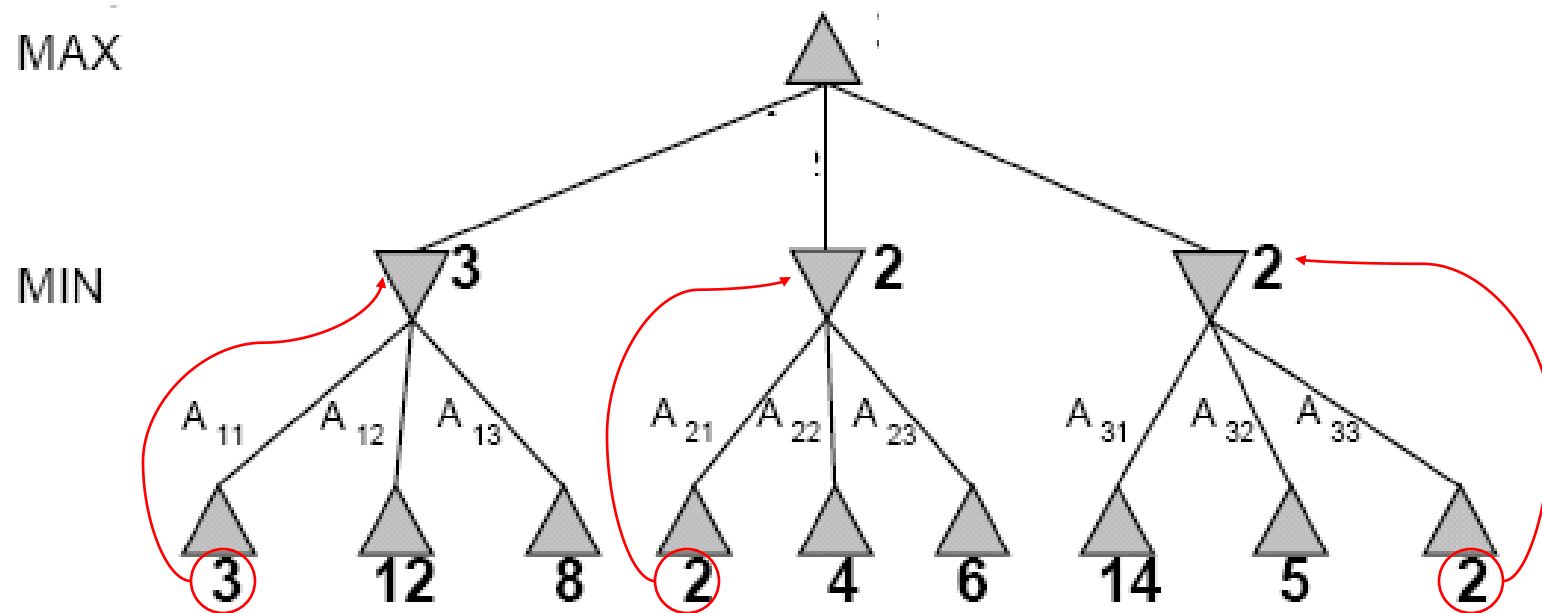
# Two-Ply Game Tree



# Two-Ply Game Tree

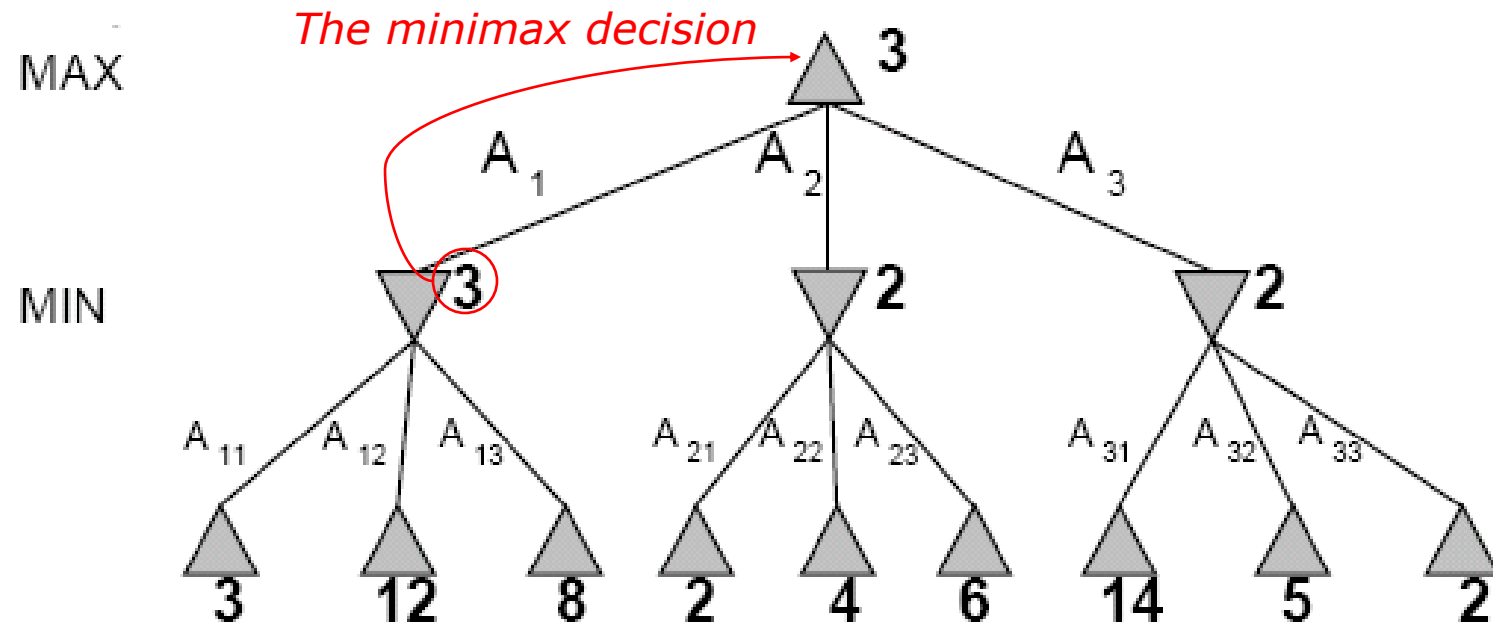


# Two-Ply Game Tree



# Two-Ply Game Tree

**Minimax maximizes the utility for the worst-case outcome for max**



# Pseudocode for Minimax Algorithm

**function** MINIMAX-DECISION(*state*) **returns** *an action*

**inputs:** *state*, current state in game

$v \leftarrow \text{MAX-VALUE}(\textit{state})$

**return** the *action* in SUCCESSORS(*state*) with value  $v$

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**function** MAX-VALUE(*state*) **returns** *a utility value*

**if** TERMINAL-TEST(*state*) **then return** UTILITY(*state*)

$v \leftarrow -\infty$

**for**  $a, s$  in SUCCESSORS(*state*) **do**

$v \leftarrow \text{MAX}(v, \text{MIN-VALUE}(s))$

**return**  $v$

---

**function** MIN-VALUE(*state*) **returns** *a utility value*

**if** TERMINAL-TEST(*state*) **then return** UTILITY(*state*)

$v \leftarrow \infty$

**for**  $a, s$  in SUCCESSORS(*state*) **do**

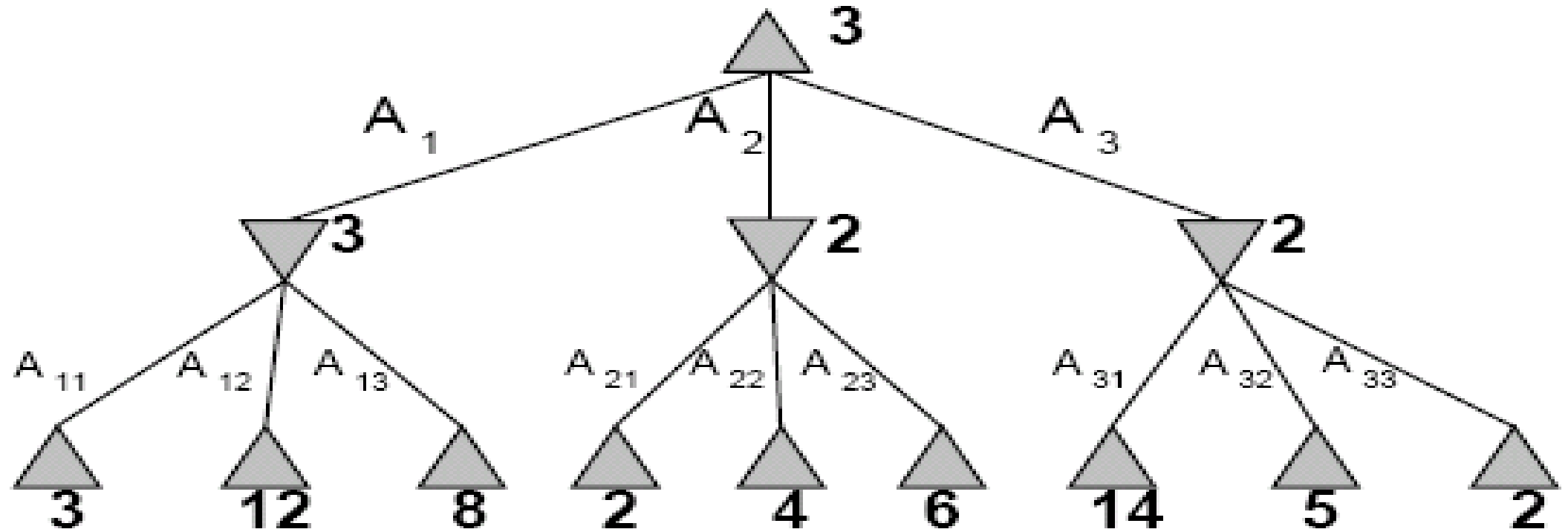
$v \leftarrow \text{MIN}(v, \text{MAX-VALUE}(s))$

**return**  $v$

## Example-1

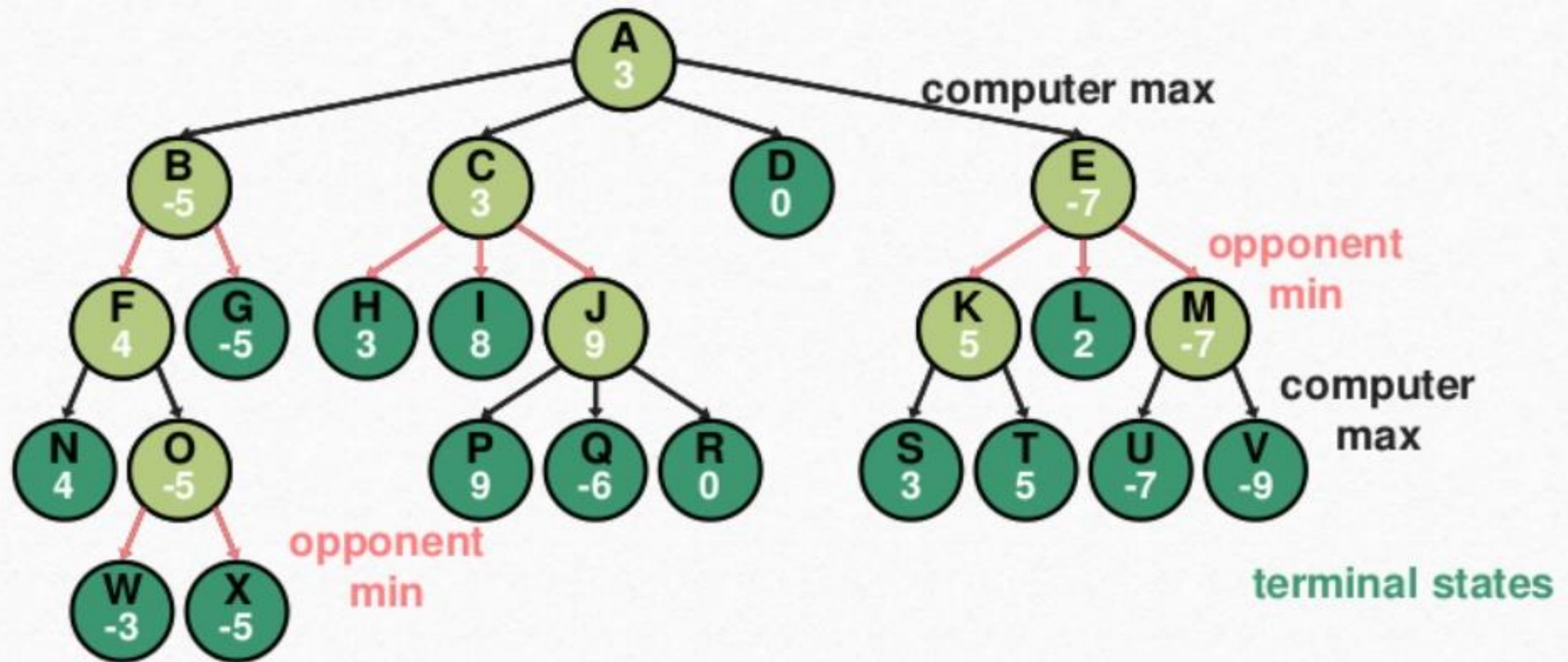
MAX

MIN

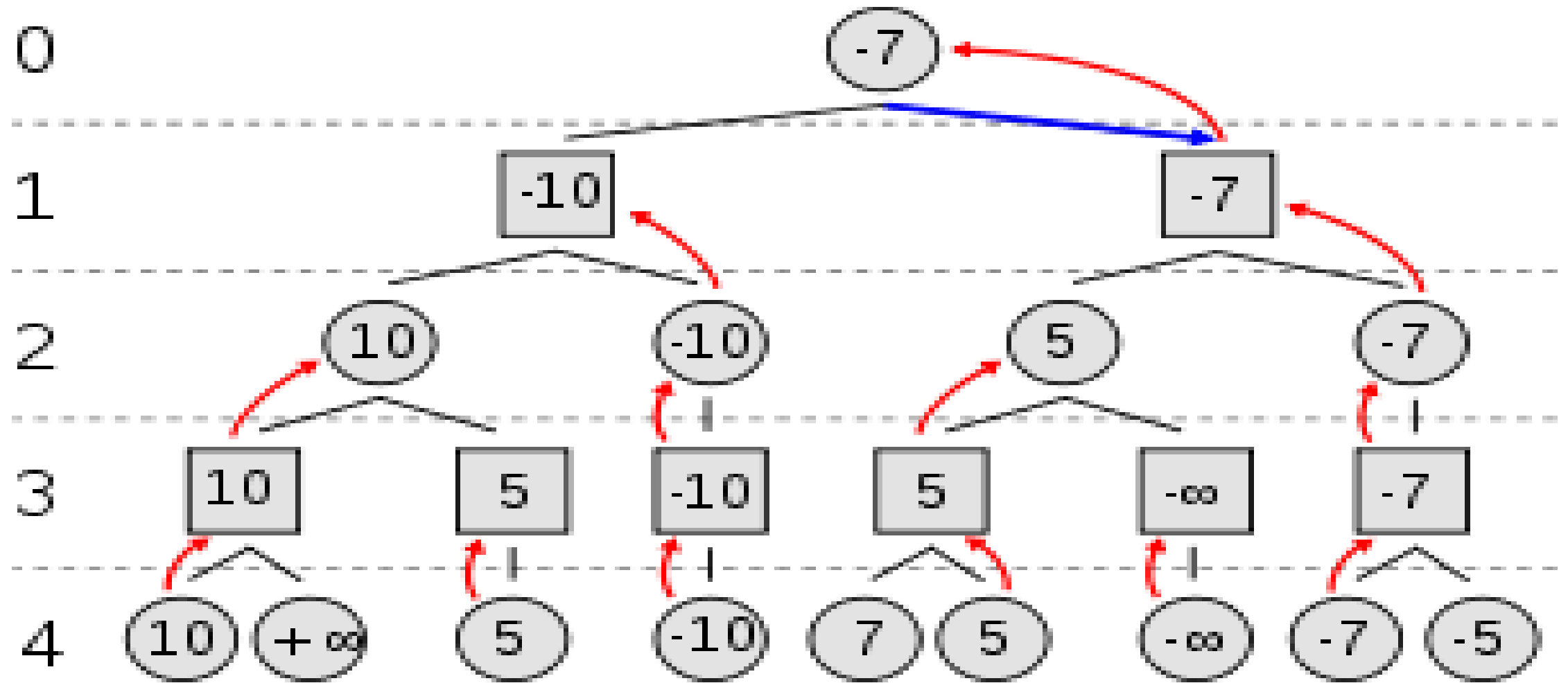




## Example-2



### Example-3



# Properties of Minimax

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- **Complete? Yes (if tree is finite)**
- **Optimal? Yes (against an optimal opponent)**
- **Time complexity?  $O(b^m)$**
- **Space complexity?  $O(bm)$  (depth-first exploration)**

**Thank you**