

# Artificial Intelligence

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Games



# What is game?

## **Definition:**

A Multi agent environment

## **Multi agent environments :**

- 1- Cooperative
- 2- Competitive (Adversarial Search)



# Search vs Games

## **Search — no adversary**

- Solution is (heuristic) method for finding goal
- Heuristics and CSP 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).
- Time limits force an approximate solution
- Evaluation function: evaluate “goodness” of game position

Examples: Chess – Checkers – Othello - Backgammon



# Types of Games

	Deterministic	Stochastic
Observable	Chess, Checkers, Go, Othello	Backgammon, Monopoly
Partially Observable		Bridge, Poker, Scrabble nuclear war



# Game properties

**Players:** 2 Players named MIN and MAX

**Game Progress:** MAX starts the game. The Game Continues till the end.

**Decision Making:** MAX uses Tree Search to select the next move

## Games as Search:

**Initial State:** The very first state of the game

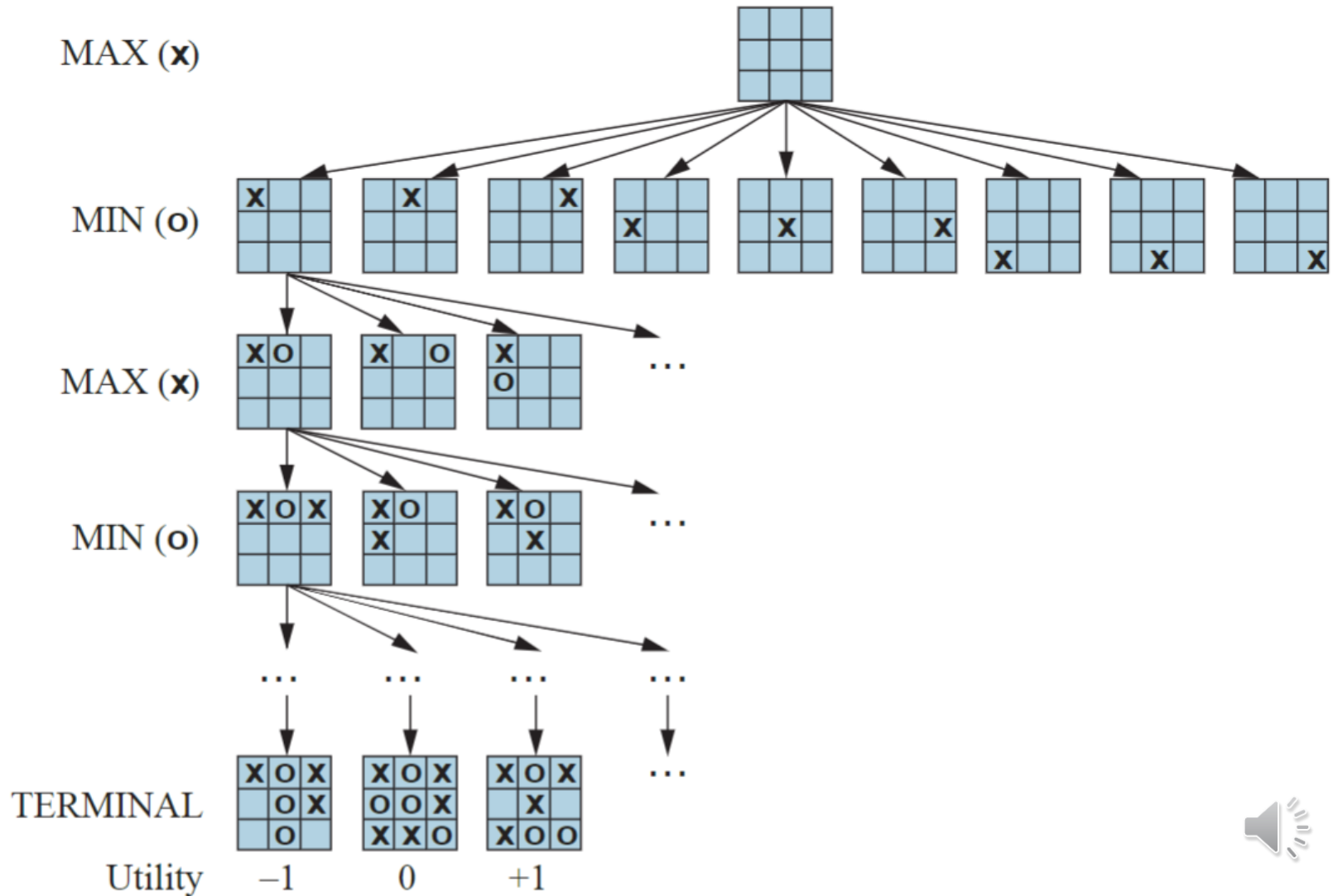
**Successor Function:** list of (move,state) pairs specifying legal moves.

**Terminal Test:** is the game finished?

**Utility Function:** Gives numerical value to the terminal states. e.g win(+1), loose(-1) and draw(0)



# TIC-TAC-TOE Example



# MINIMAX Algorithm

**Purpose:** Finding the best strategy for MAX (player)

**Default:** Both players play as (well/smart/efficient) as possible.

$\text{MINIMAX}(s) =$

$$\begin{cases} \text{UTILITY}(s, \text{MAX}) & \text{if IS-TERMINAL}(s) \\ \max_{a \in \text{Actions}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if TO-MOVE}(s) = \text{MAX} \\ \min_{a \in \text{Actions}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if TO-MOVE}(s) = \text{MIN} \end{cases}$$



**function** MINIMAX-SEARCH(*game, state*) **returns** *an action*

$\text{player} \leftarrow \text{game.TO-MOVE}(\text{state})$

$\text{value}, \text{move} \leftarrow \text{MAX-VALUE}(\text{game}, \text{state})$

**return** *move*

**function** MAX-VALUE(*game, state*) **returns** *a (utility, move) pair*

**if** *game.IS-TERMINAL(state)* **then return** *game.UTILITY(state, player), null*

$v \leftarrow -\infty$

**for each** *a* **in** *game.ACTIONS(state)* **do**

$v_2, a_2 \leftarrow \text{MIN-VALUE}(\text{game}, \text{game.RESULT}(\text{state}, a))$

**if**  $v_2 > v$  **then**

$v, \text{move} \leftarrow v_2, a$

**return** *v, move*

**function** MIN-VALUE(*game, state*) **returns** *a (utility, move) pair*

**if** *game.IS-TERMINAL(state)* **then return** *game.UTILITY(state, player), null*

$v \leftarrow +\infty$

**for each** *a* **in** *game.ACTIONS(state)* **do**

$v_2, a_2 \leftarrow \text{MAX-VALUE}(\text{game}, \text{game.RESULT}(\text{state}, a))$

**if**  $v_2 < v$  **then**

$v, \text{move} \leftarrow v_2, a$

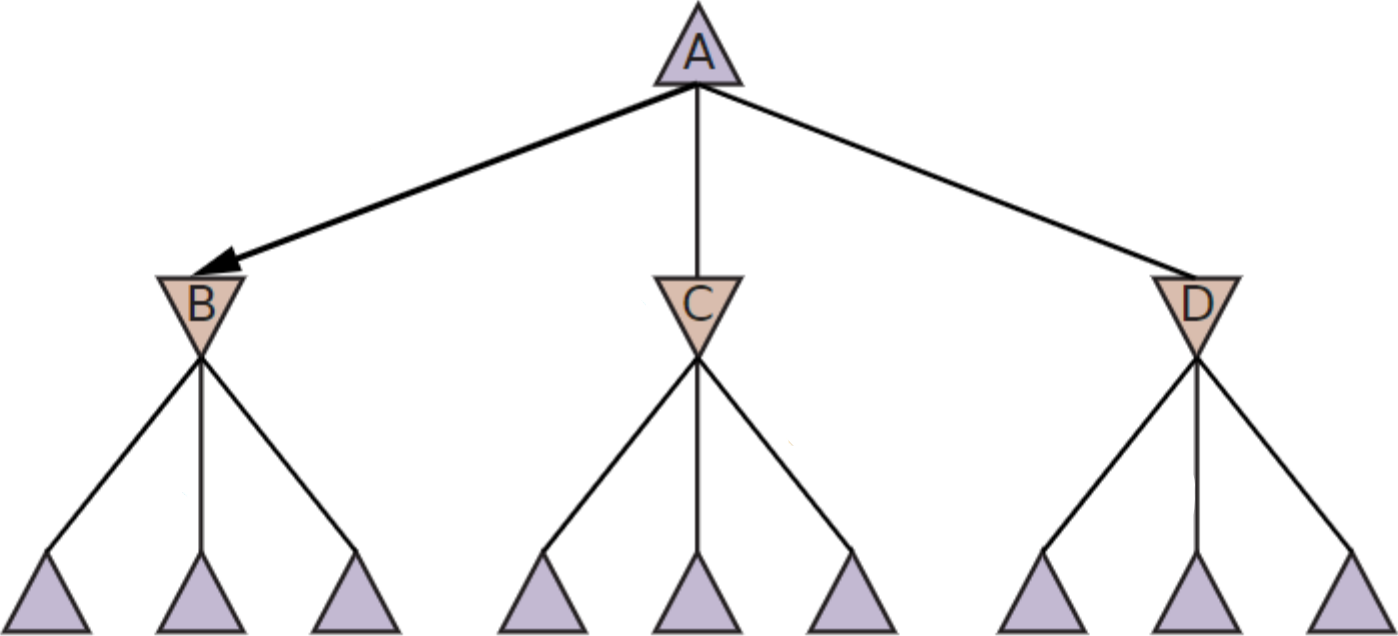
**return** *v, move*





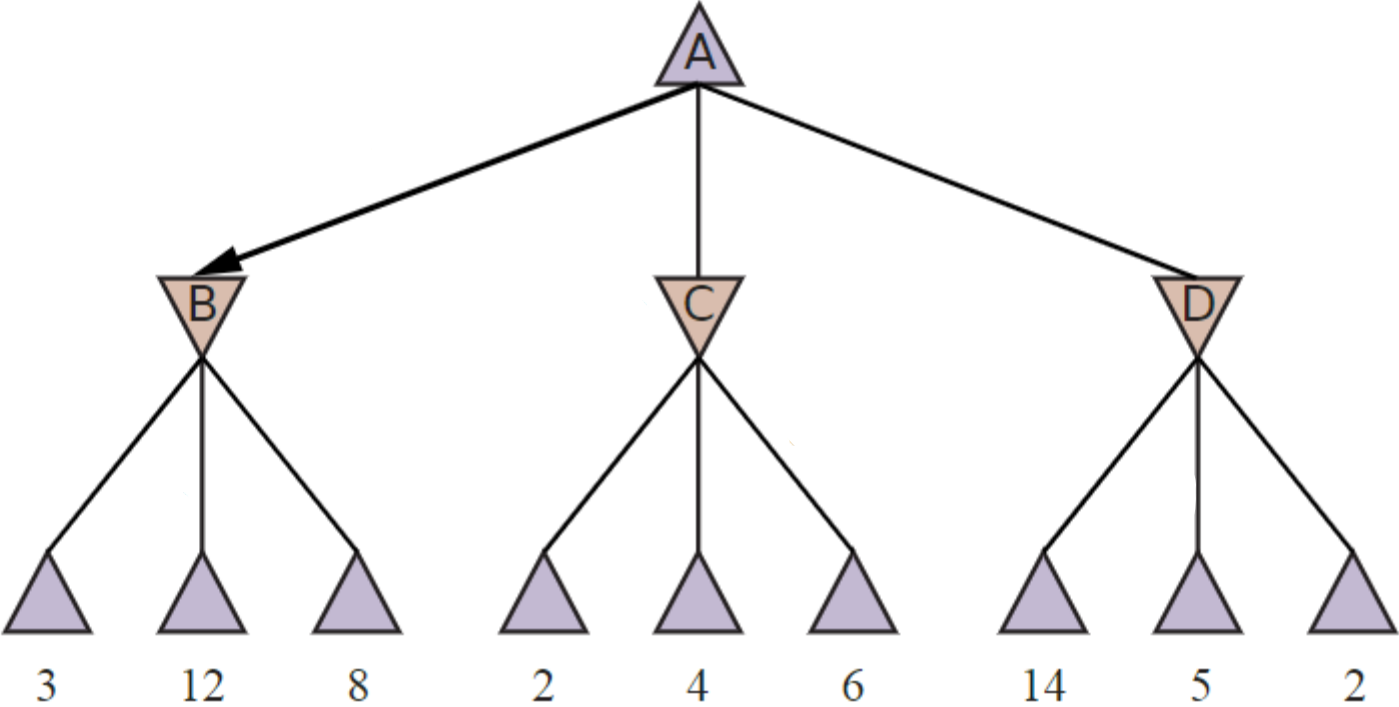
MAX

MIN



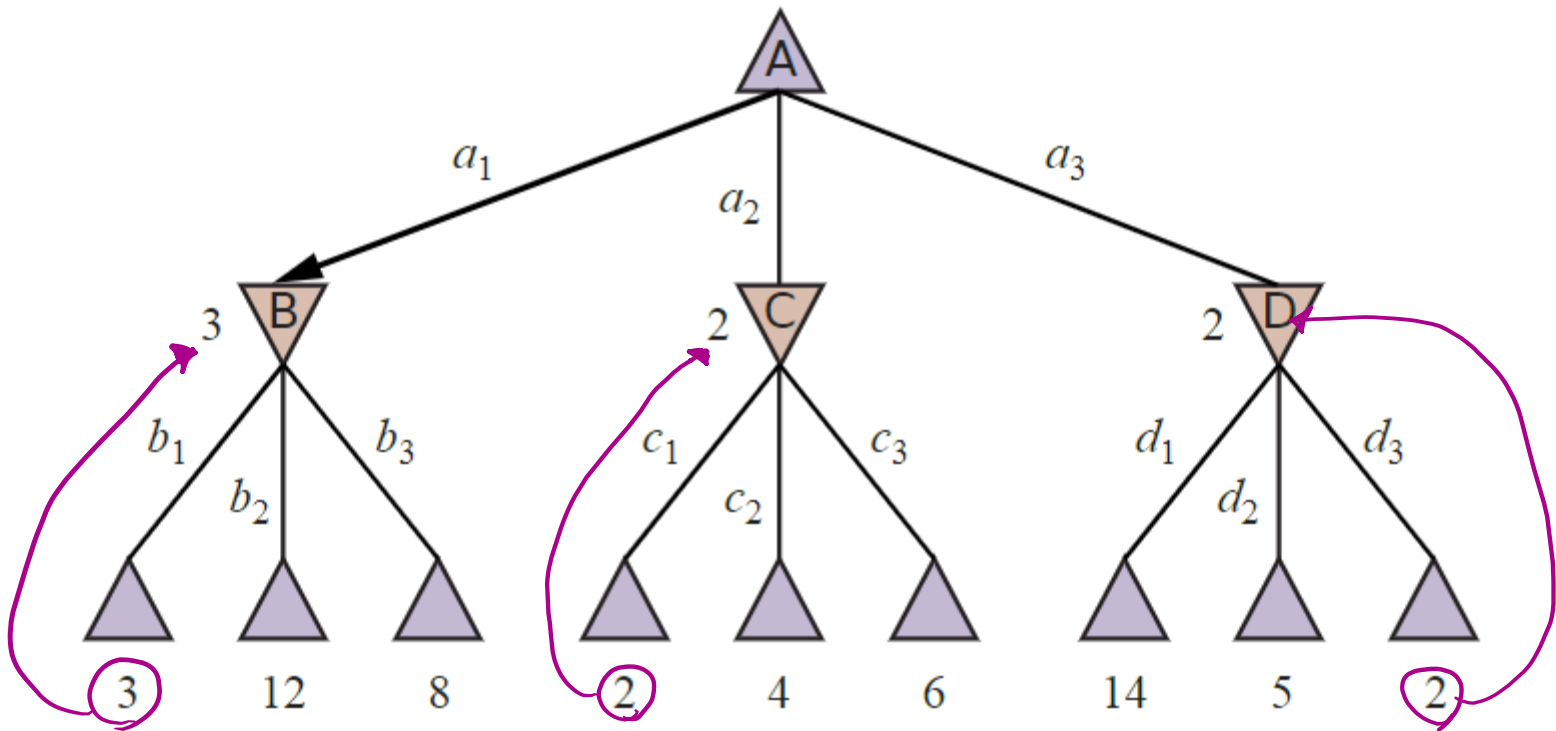
MAX

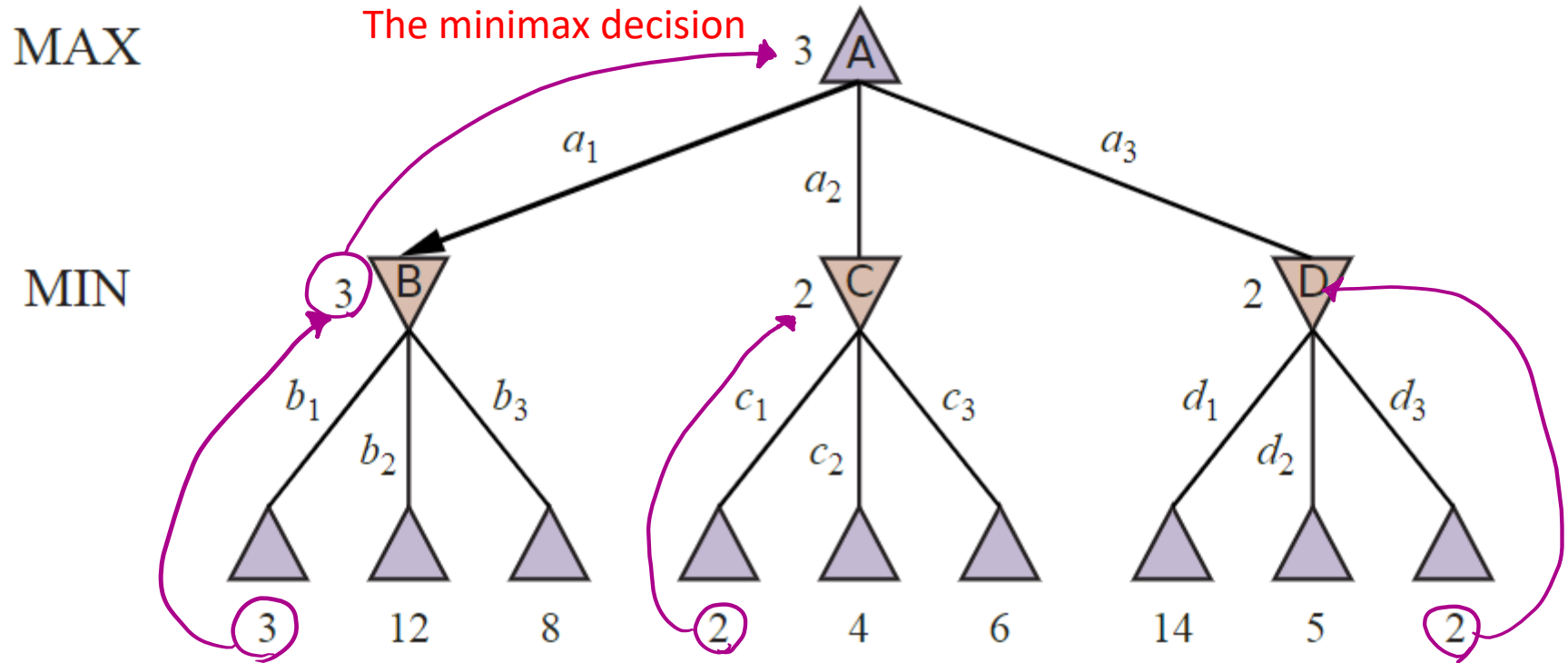
MIN



MAX

MIN





Minimax Algorithm maximizes the worst-case outcome for MAX.



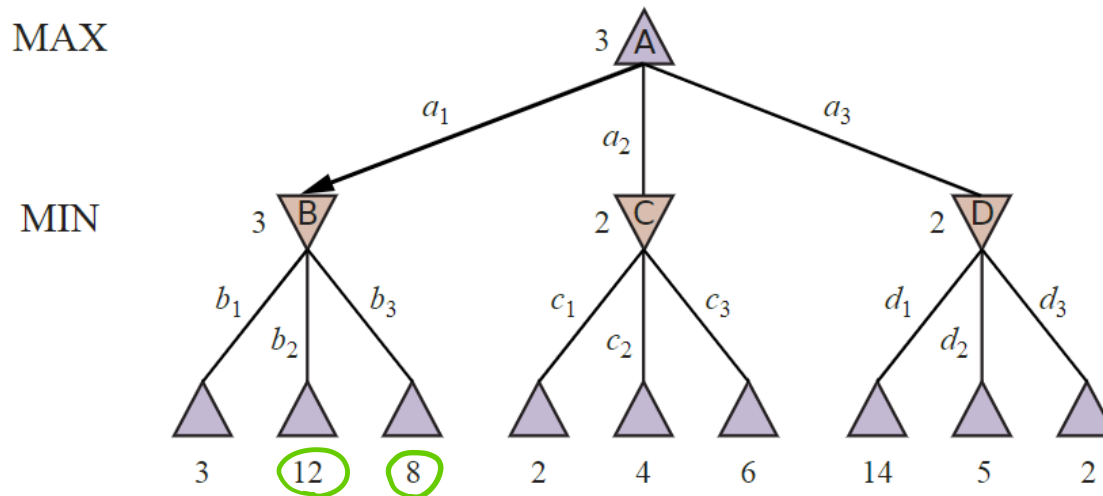
# Question

**Q:**

What if the MIN player doesn't play as (well/smart/efficient) as possible?

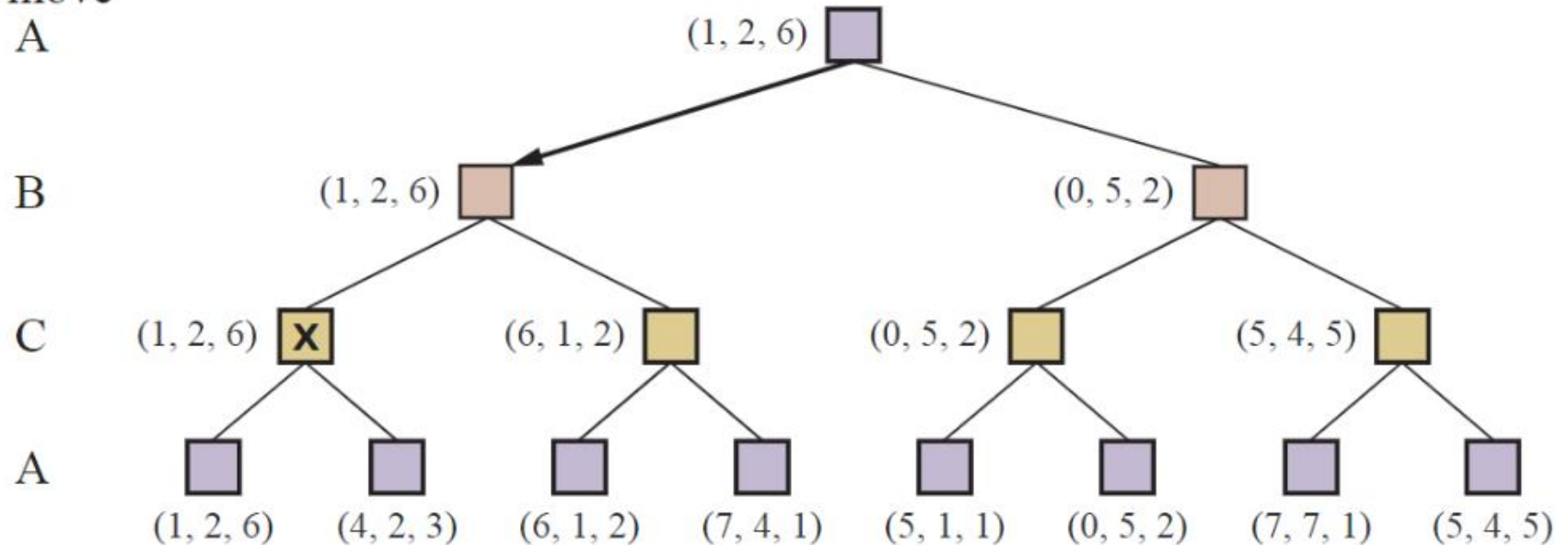
**A:**

The result would be better then!



# Games with more than two players

to move  
A



**Figure 5.4** The first three ply of a game tree with three players ( $A$ ,  $B$ ,  $C$ ). Each node is labeled with values from the viewpoint of each player. The best move is marked at the root.



# MINIMAX

Algorithm	MINIMAX
Complete	Yes
Time	$O(b^m)$
Space	$O(bm)$
Optimal	Yes



# Alpha-Beta Pruning

**Problem:** The number of states is exponential in compared to the number of moves.

**Solution:** Don't check all nodes.

**Alpha** = Maximum value in the node MAX

**Beta** = Minimum Value in the node MIN

## Rules:

- **First Rule:**

In the MIN node P, prune if  $\text{Alpha}(\text{Parent}(P)) > \text{Beta}(P)$

- **Second Rule:**

In the MAX node Q, prune if  $\text{Alpha}(Q) > \text{Beta}(\text{Parent}(Q))$





**function** ALPHA-BETA-SEARCH(*game, state*) **returns** an action  
     $\text{player} \leftarrow \text{game.TO-MOVE}(\text{state})$   
     $\text{value}, \text{move} \leftarrow \text{MAX-VALUE}(\text{game}, \text{state}, -\infty, +\infty)$   
    **return** *move*

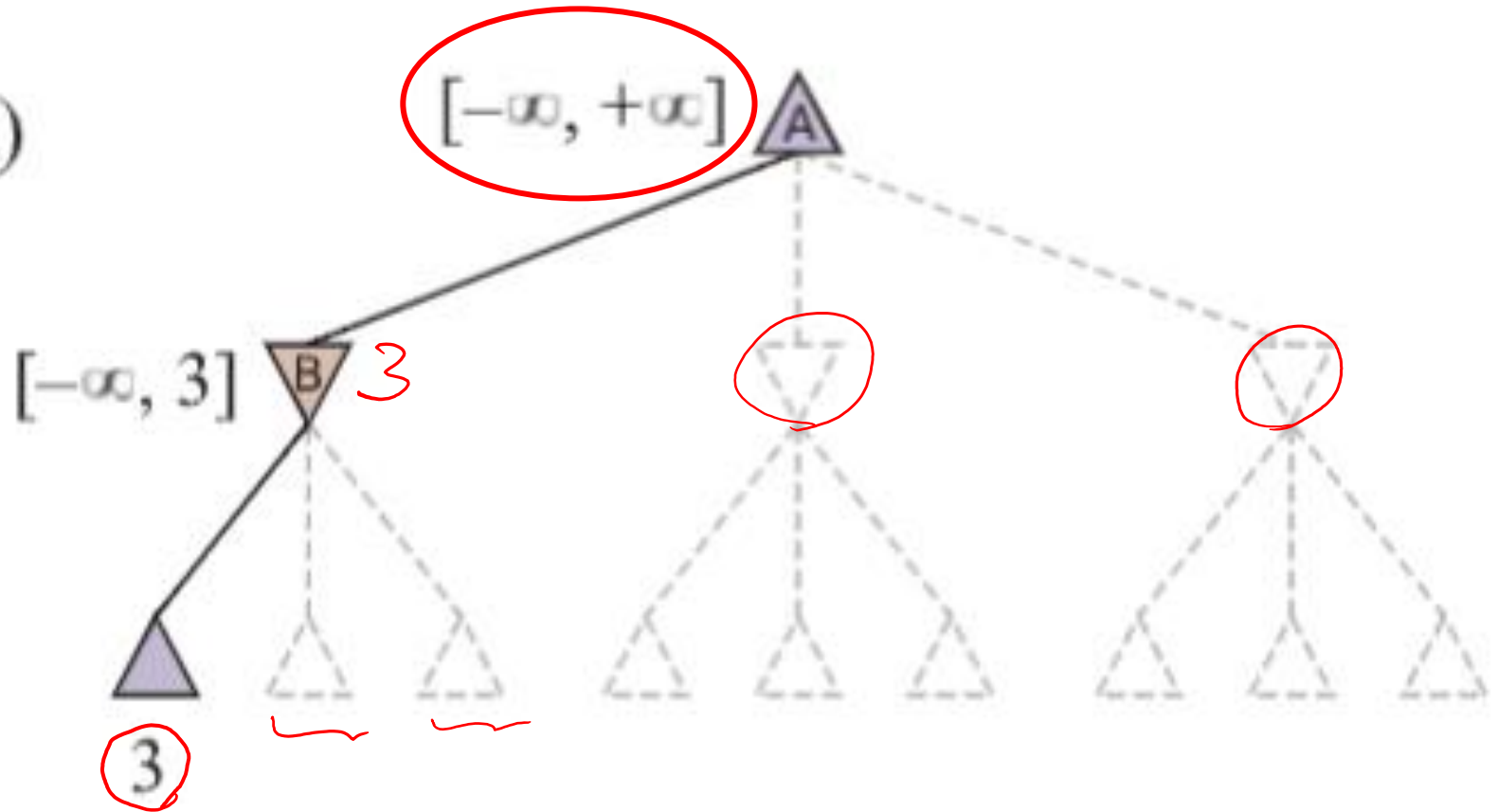
**function** MAX-VALUE(*game, state,  $\alpha, \beta$* ) **returns** a (*utility, move*) pair  
    **if** *game.IS-TERMINAL(state)* **then return** *game.UTILITY(state, player), null*  
     $v \leftarrow -\infty$   
    **for each** *a* **in** *game.ACTIONS(state)* **do**  
         $v2, a2 \leftarrow \text{MIN-VALUE}(\text{game}, \text{game.RESULT}(\text{state}, a), \alpha, \beta)$   
        **if**  $v2 > v$  **then**  
             $v, \text{move} \leftarrow v2, a$   
             $\alpha \leftarrow \text{MAX}(\alpha, v)$   
        **if**  $v \geq \beta$  **then return**  $v, \text{move}$   
    **return**  $v, \text{move}$

**function** MIN-VALUE(*game, state,  $\alpha, \beta$* ) **returns** a (*utility, move*) pair  
    **if** *game.IS-TERMINAL(state)* **then return** *game.UTILITY(state, player), null*  
     $v \leftarrow +\infty$   
    **for each** *a* **in** *game.ACTIONS(state)* **do**  
         $v2, a2 \leftarrow \text{MAX-VALUE}(\text{game}, \text{game.RESULT}(\text{state}, a), \alpha, \beta)$   
        **if**  $v2 < v$  **then**  
             $v, \text{move} \leftarrow v2, a$   
             $\beta \leftarrow \text{MIN}(\beta, v)$   
        **if**  $v \leq \alpha$  **then return**  $v, \text{move}$   
    **return**  $v, \text{move}$

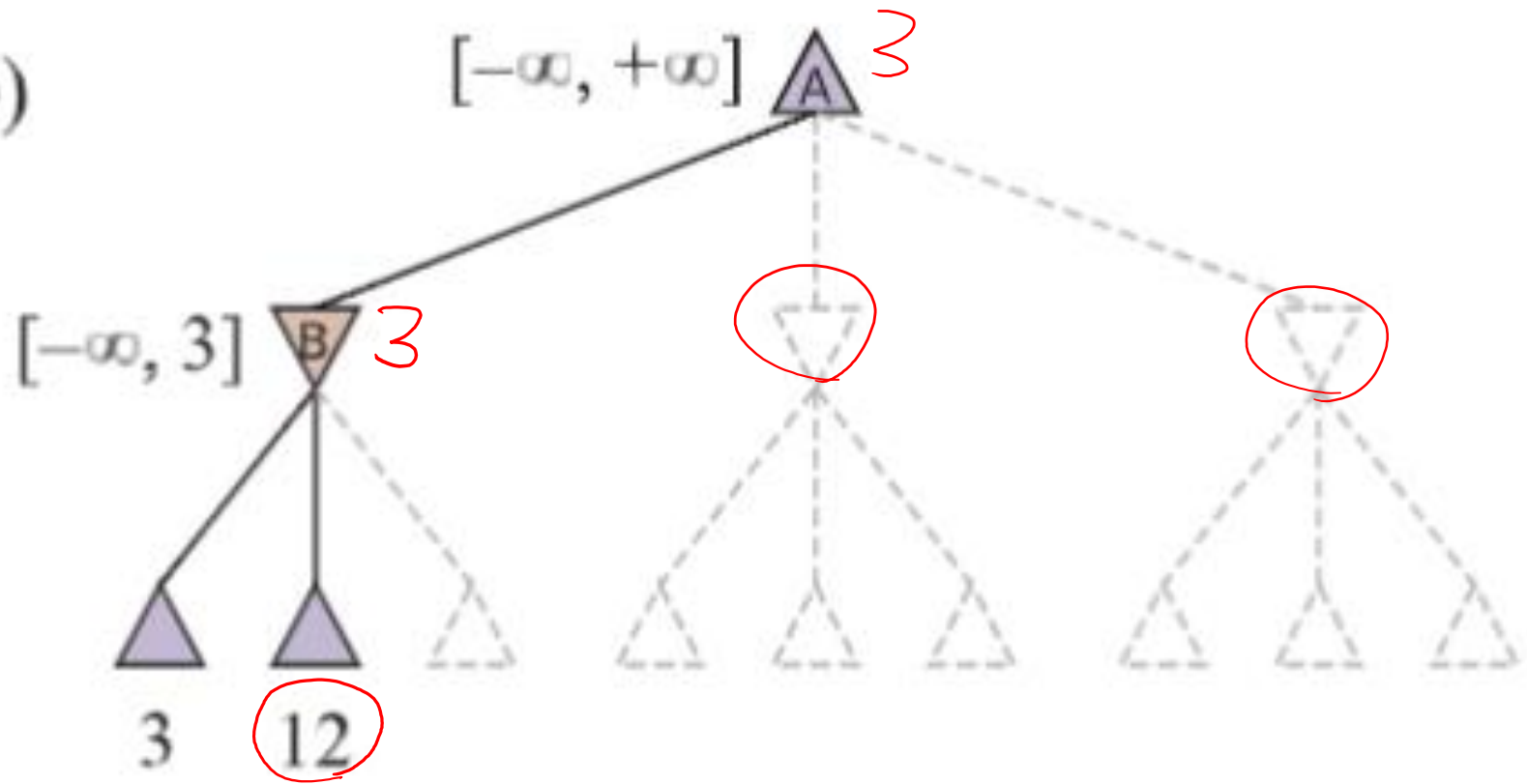


Range of possible values

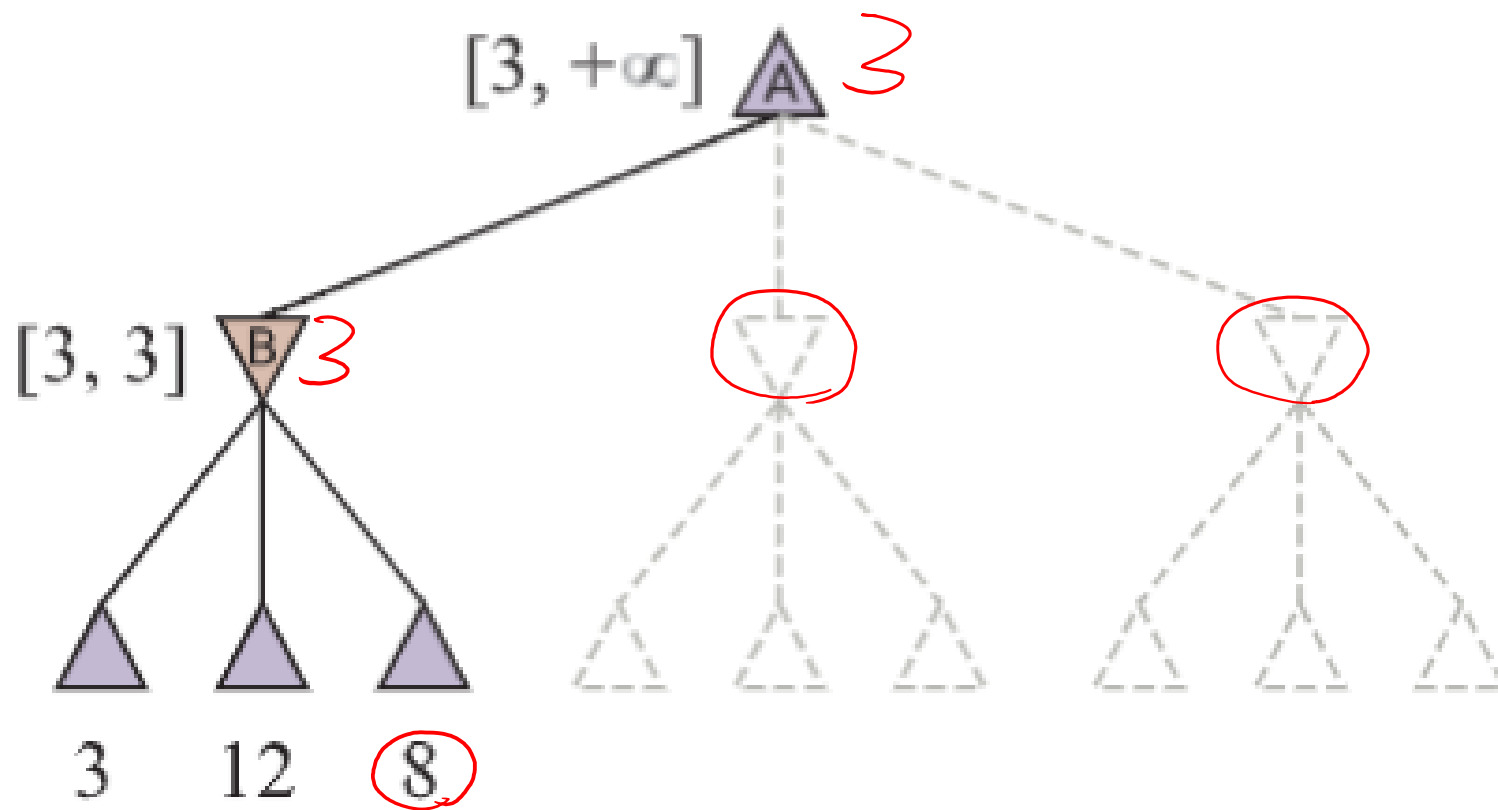
(a)



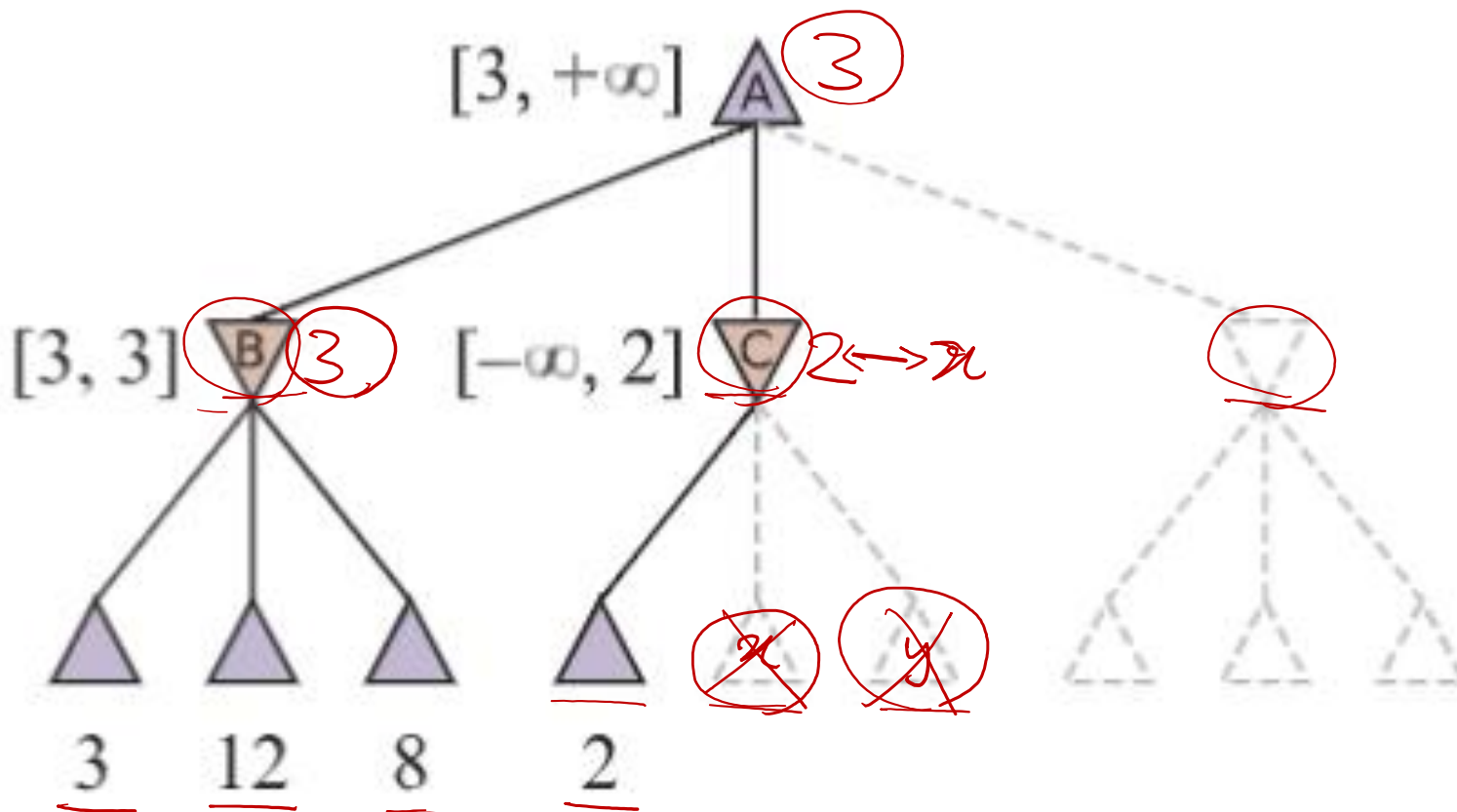
(b)



(c)



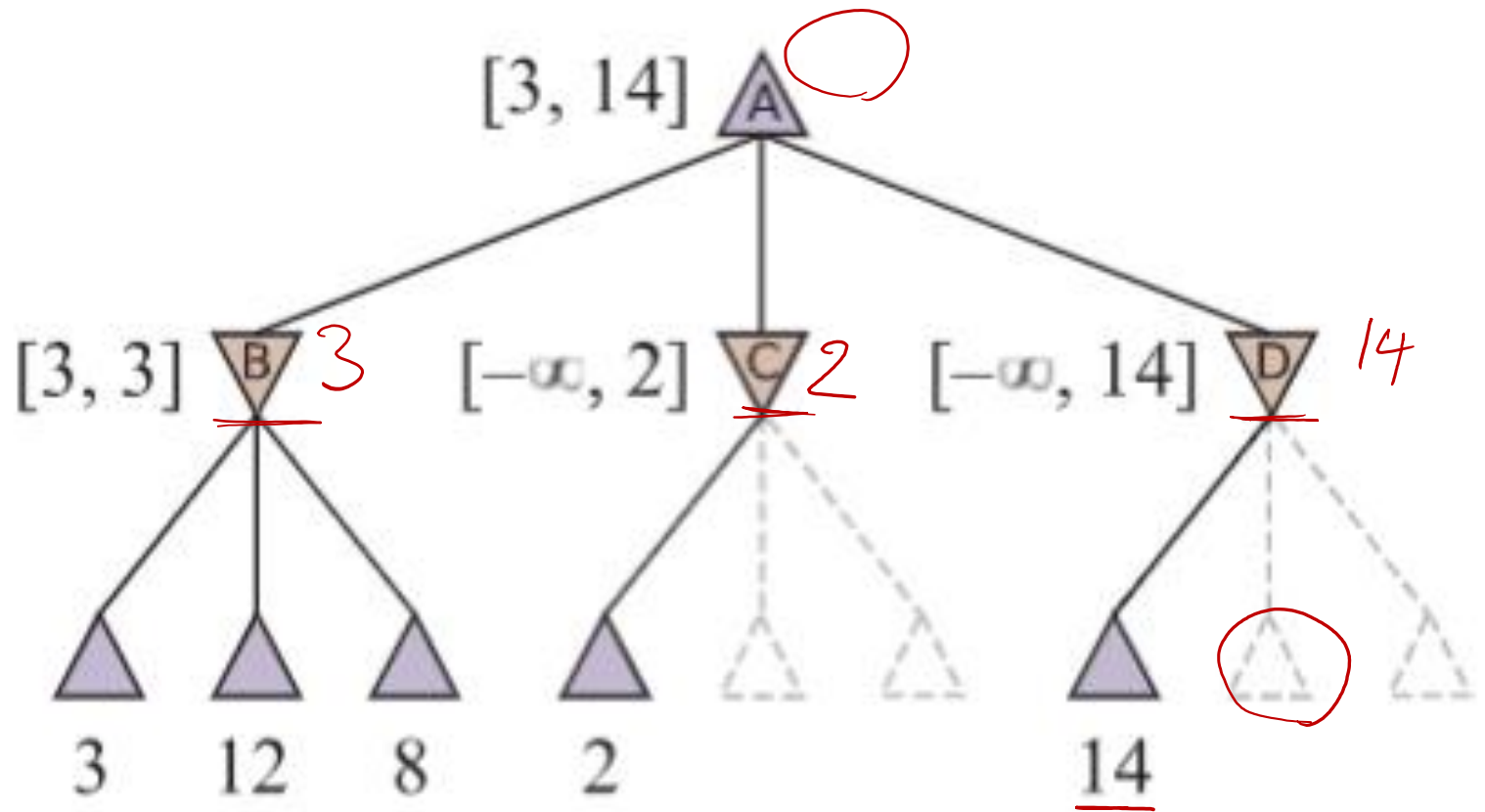
(d)



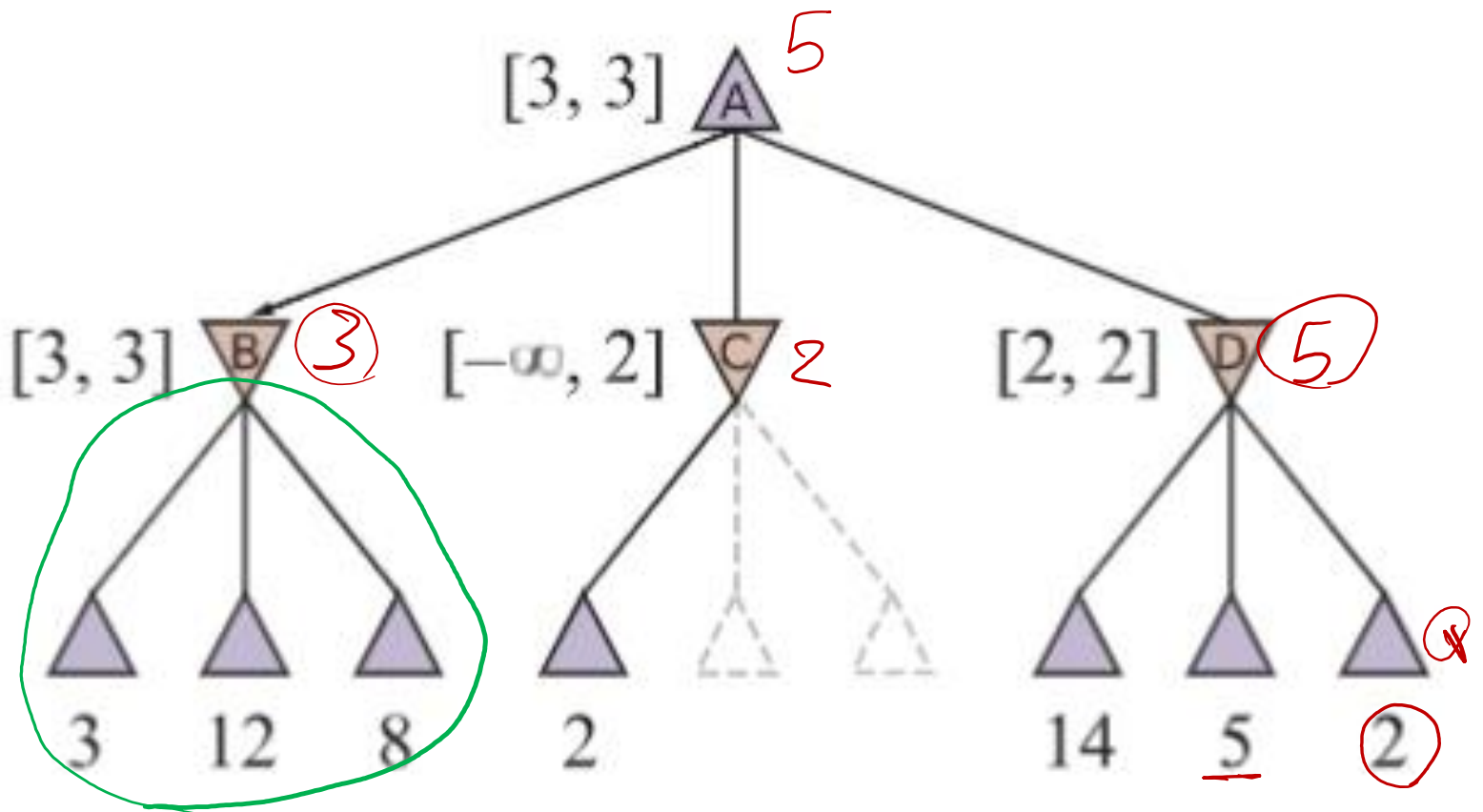
$x, y \geq 2$  ①

$x, y \leq 2$  ②

(e)



(f)



④, 3, 2, ...

