

# CSC242: Intro to AI

Lecture 5

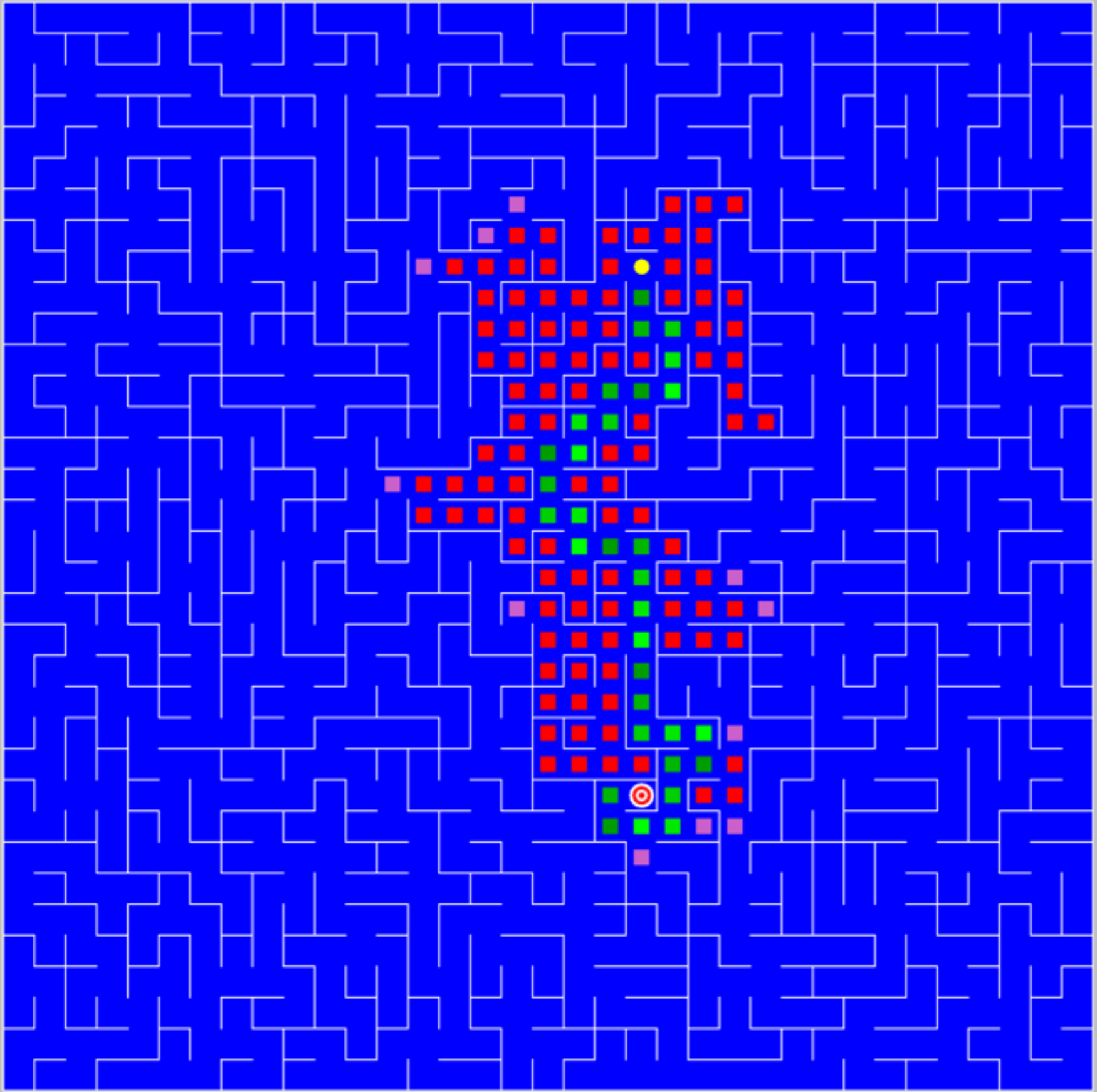
Adversarial Search

# Help Sessions

- Grad TA Xiaowan Dong
  - 5:00–6:00 pm Tuesday & Wednesday, CSB 724
- Undergrad TAs Sean Esterkin and Dan Scarafoni
  - 2:30–3:30 pm Thursdays, CS Majors Lab CSB 633
  - 5:00–6:00 p.m. Mondays, CS Majors Lab CSB 633
- Instructor Henry Kautz
  - 12:00–1:00 pm Friday, CSB 709



**BACK**   
**TO**  
**THE PAST**



Create

Width: 35

Height: 35

Cycles: 0

Clear

Search:

- DFS
- BFS
- IDS
- Best
- A\*

Solution

Length: 33

Nodes: 133

**and now it's time for something  
completely different**



# Adversarial Search





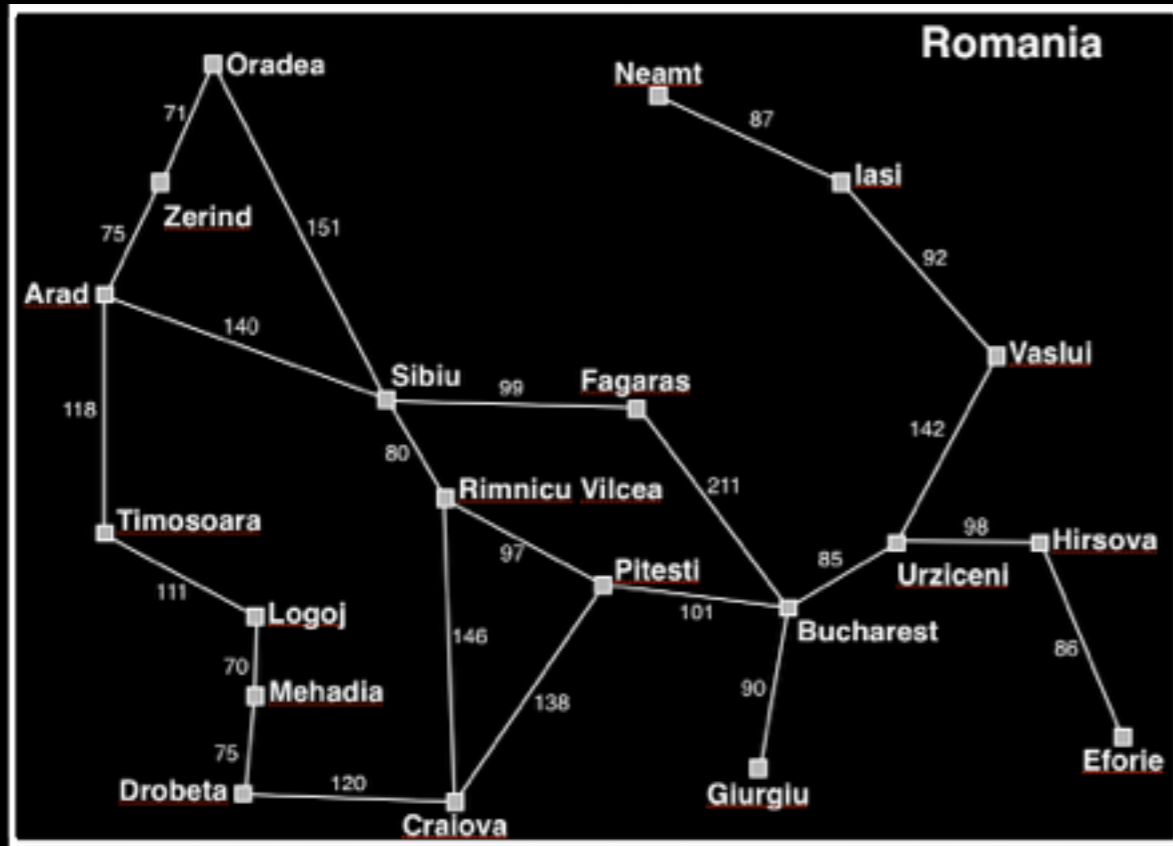


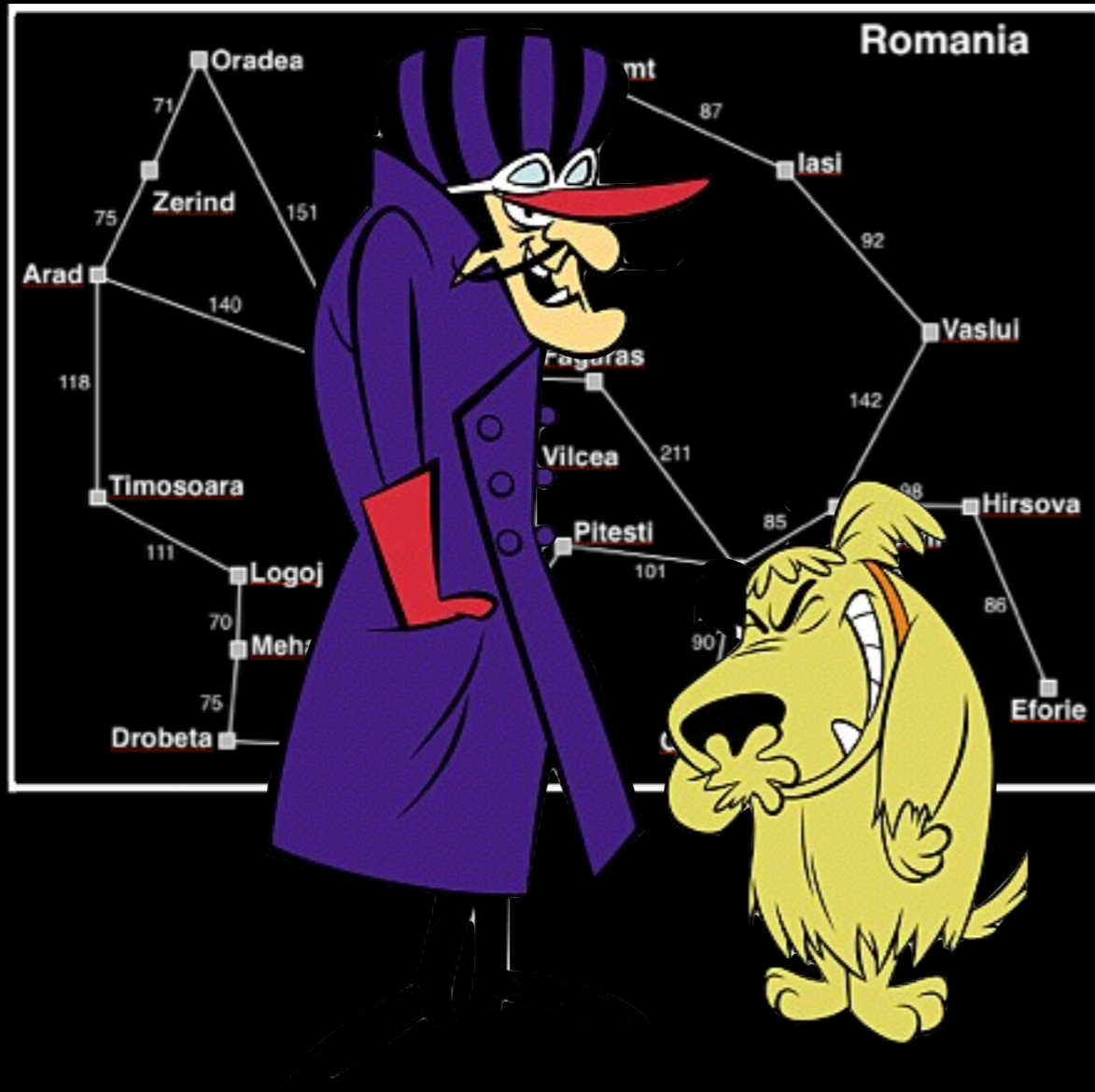


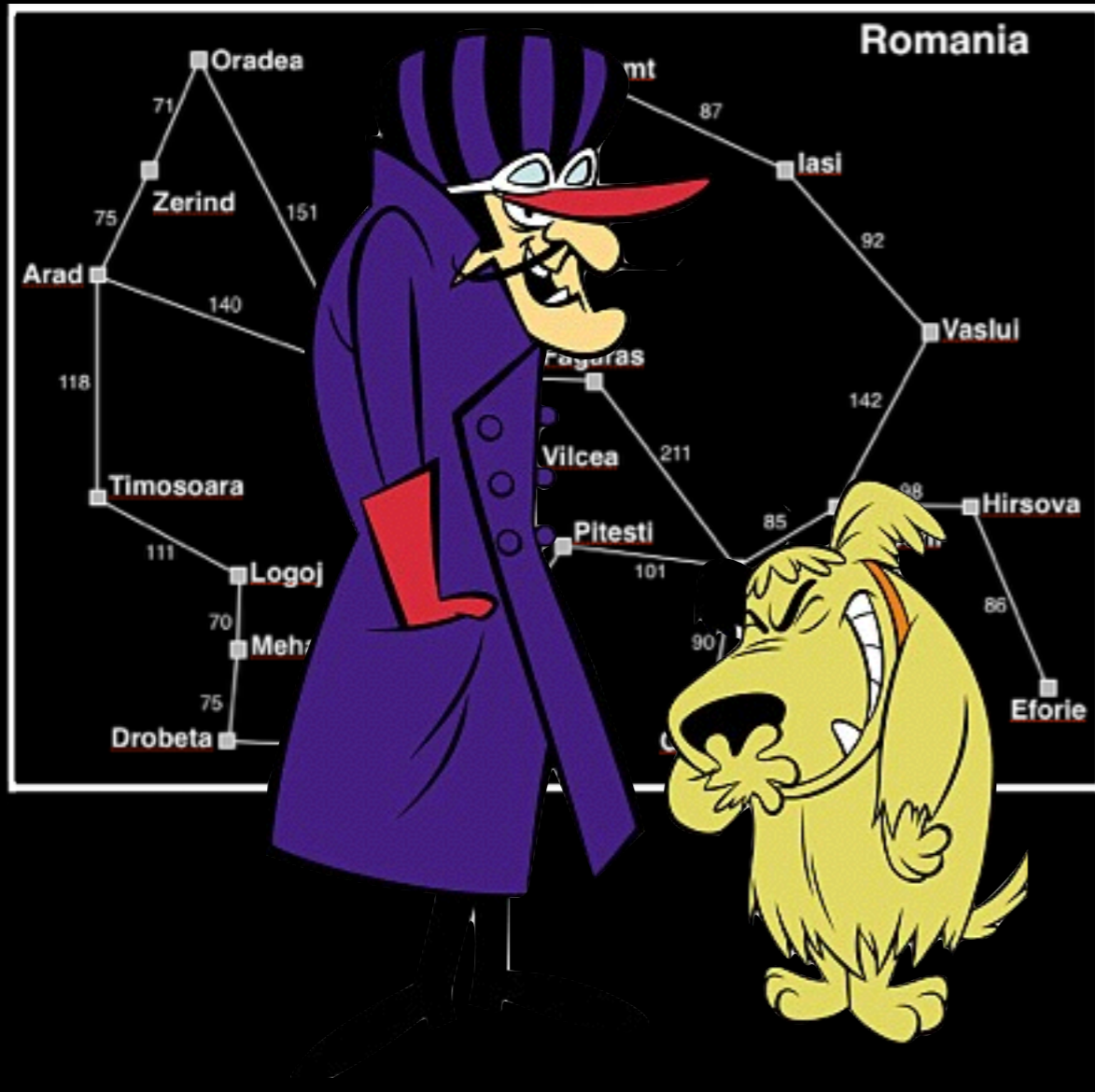


How are the games the same?  
How are they different?

# Adversarial Search

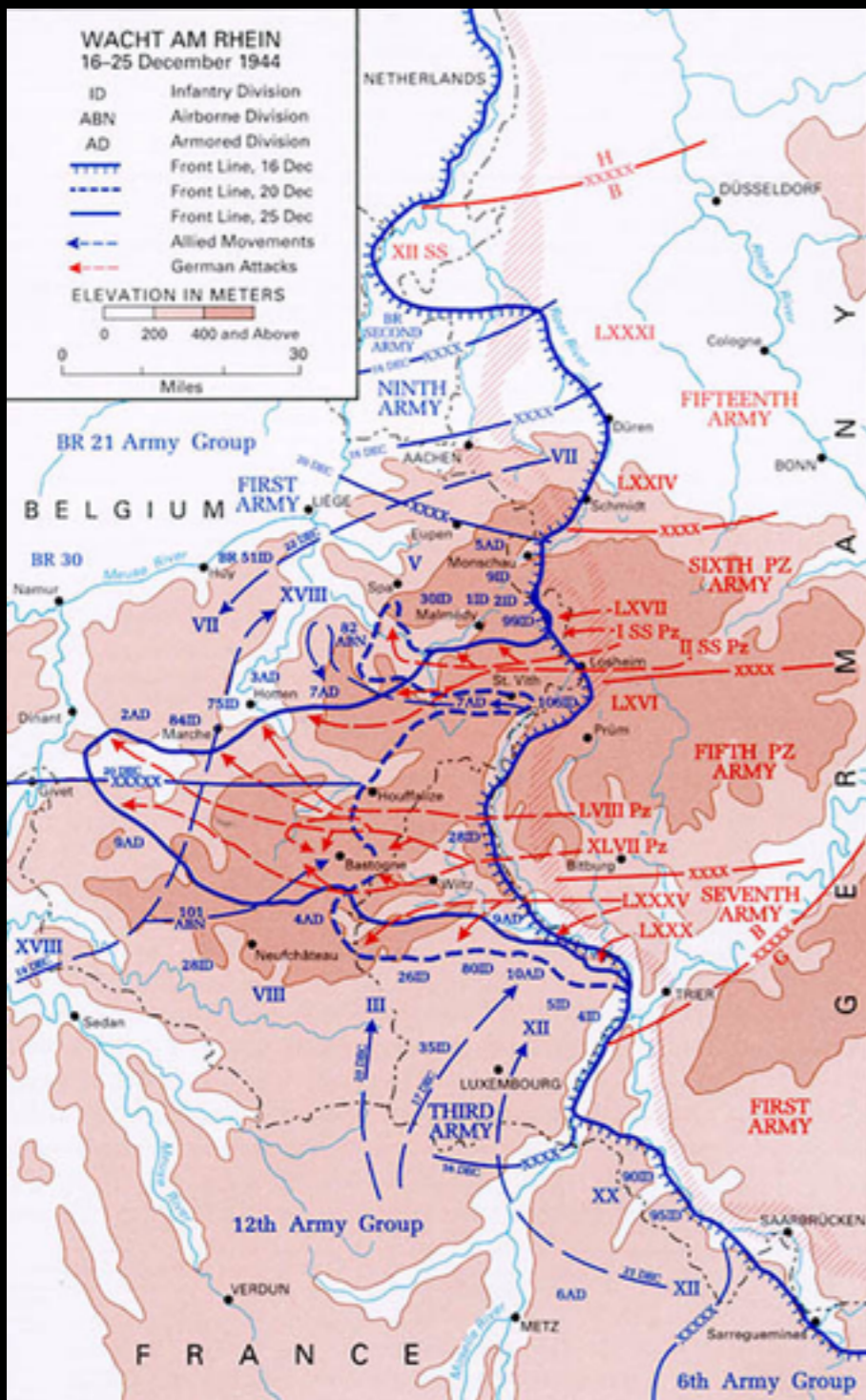




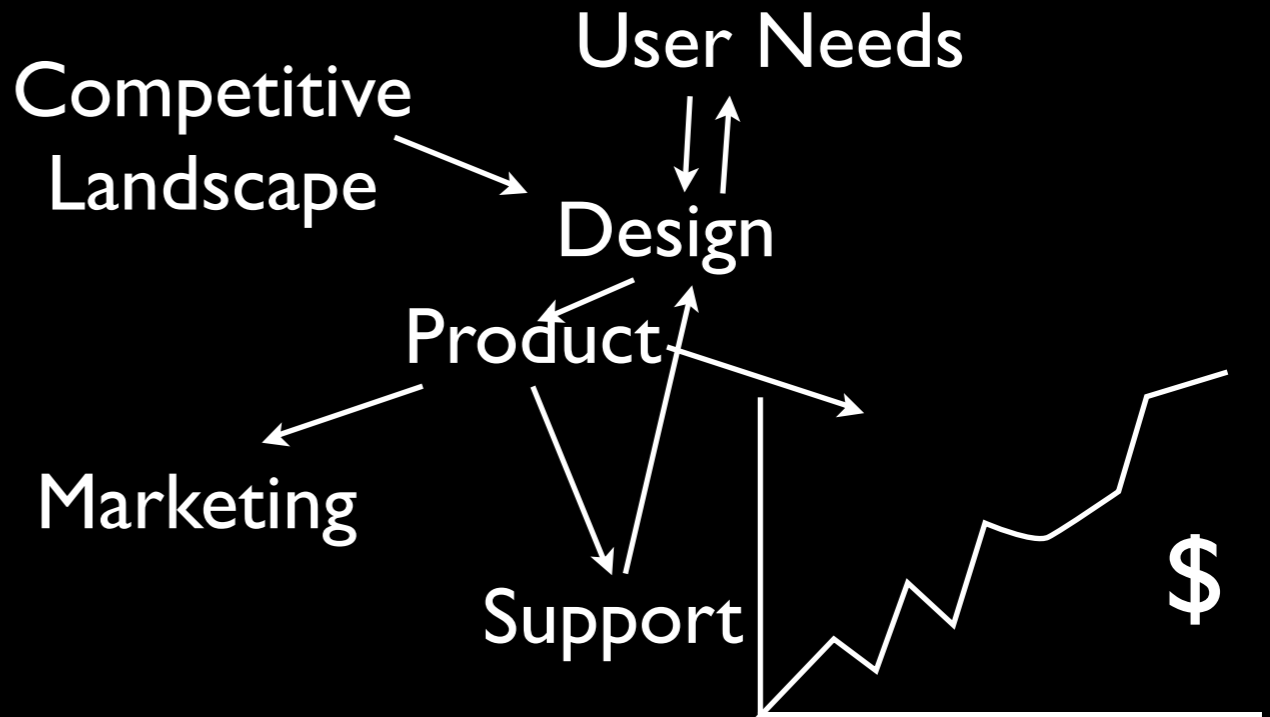
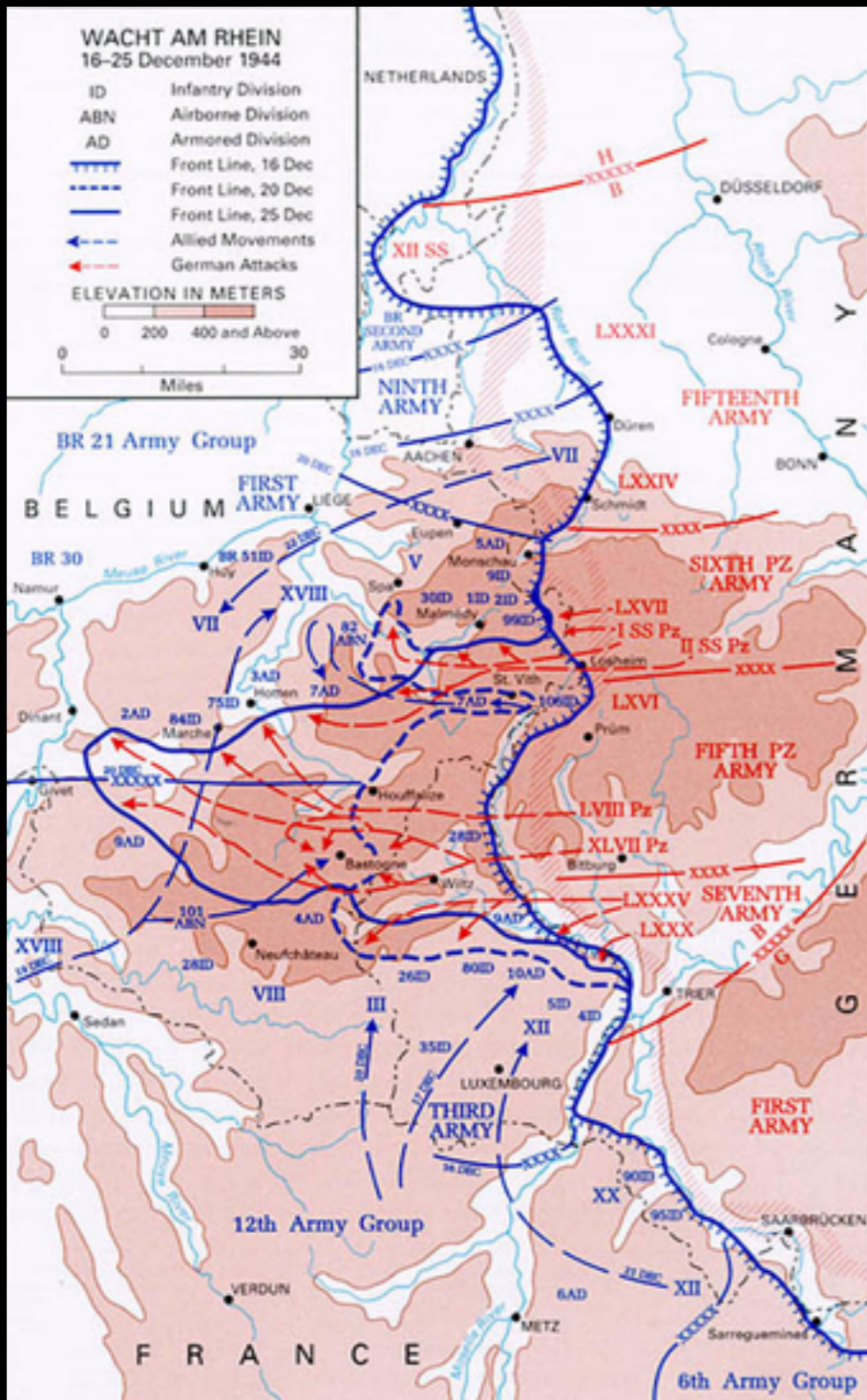




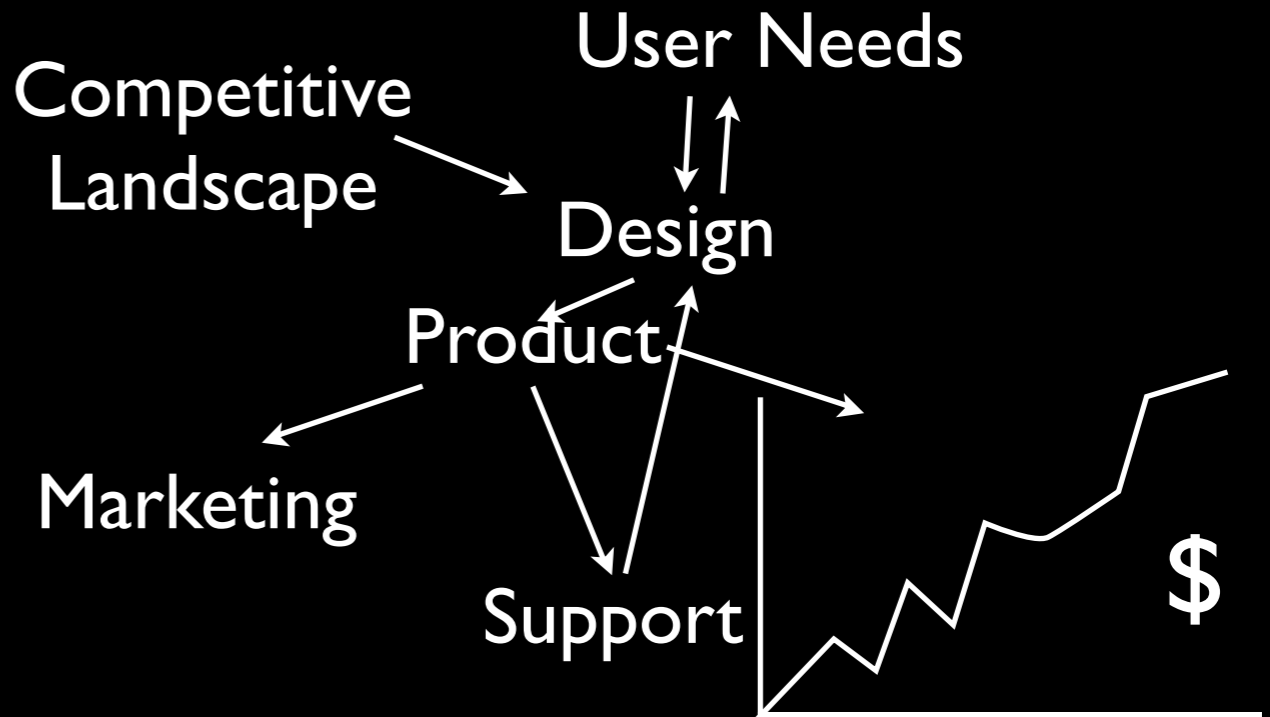
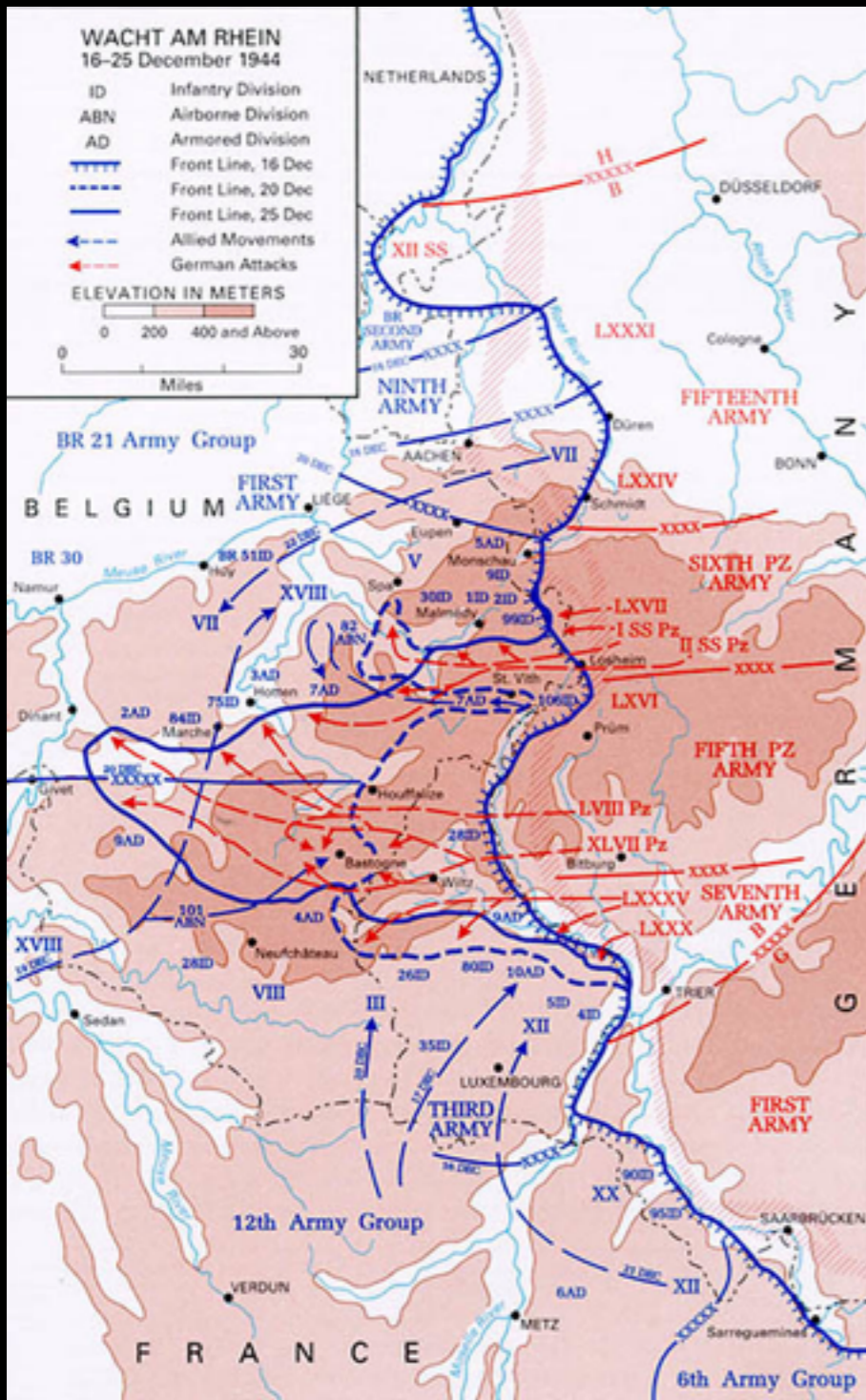




**Battle of The Bulge**  
**16-25 Dec 1944**



**Battle of The Bulge**  
**16-25 Dec 1944**



**Battle of The Bulge**  
**16-25 Dec 1944**



# Multi-Agent Environments

- Unpredictability of other agents → contingencies (strategies)
- Agents goals in conflict → competition

# Games

# Games



# Games





# Games

- “Require the ability to make *some* decision even when calculating the *optimal* decision is infeasible”
- “Penalize inefficiency severely”

# Games $\neq$ Toy Problems



$$9! = 362880$$



$$35^{100} = 10^{154}$$



$$2 \times 10^{170}$$

# Abstract Games

# Abstract Games

- Deterministic (no chance)
- Nondeterministic (dice, cards, etc.)

# Abstract Games

- Perfect information (fully observable)
- Imperfect information (partially observable)

# Abstract Games

- Zero-sum (total payoff the same in any game)
  - What's good for me is bad for you and vice-versa
- Arbitrary utility function

# Types of Games

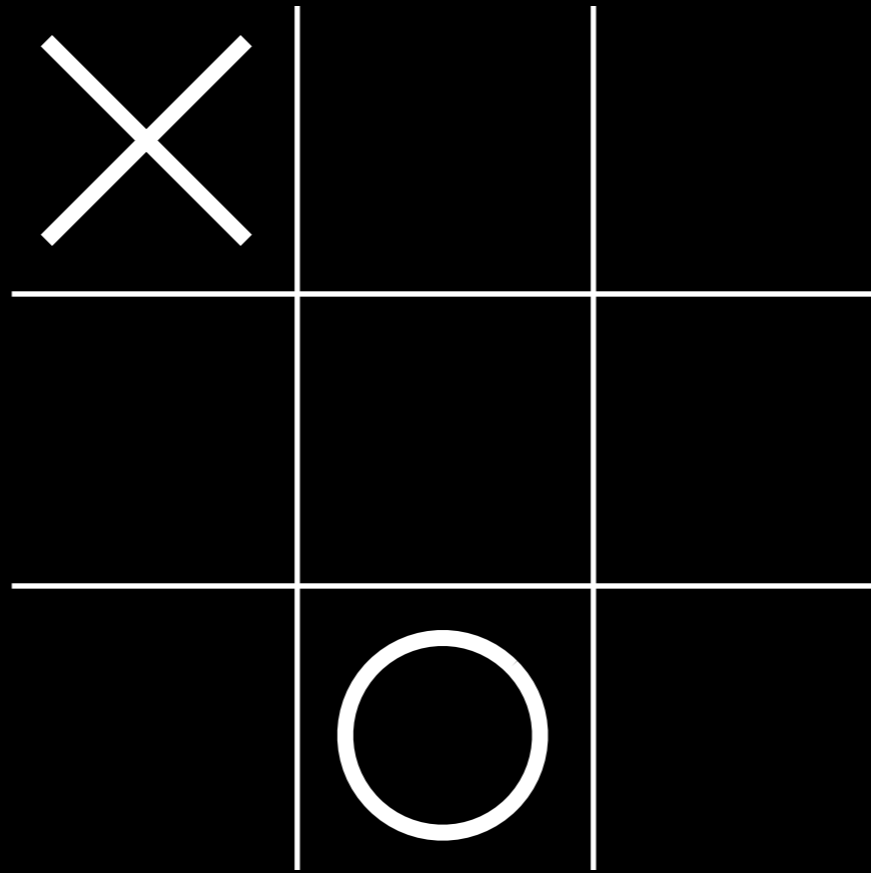
Deterministic (no chance)	Nondeterministic (dice, cards, etc.)
Perfect information (fully observable)	Imperfect information (partially observable)
Zero-sum (total payoff the same in any game)	Arbitrary utility functions

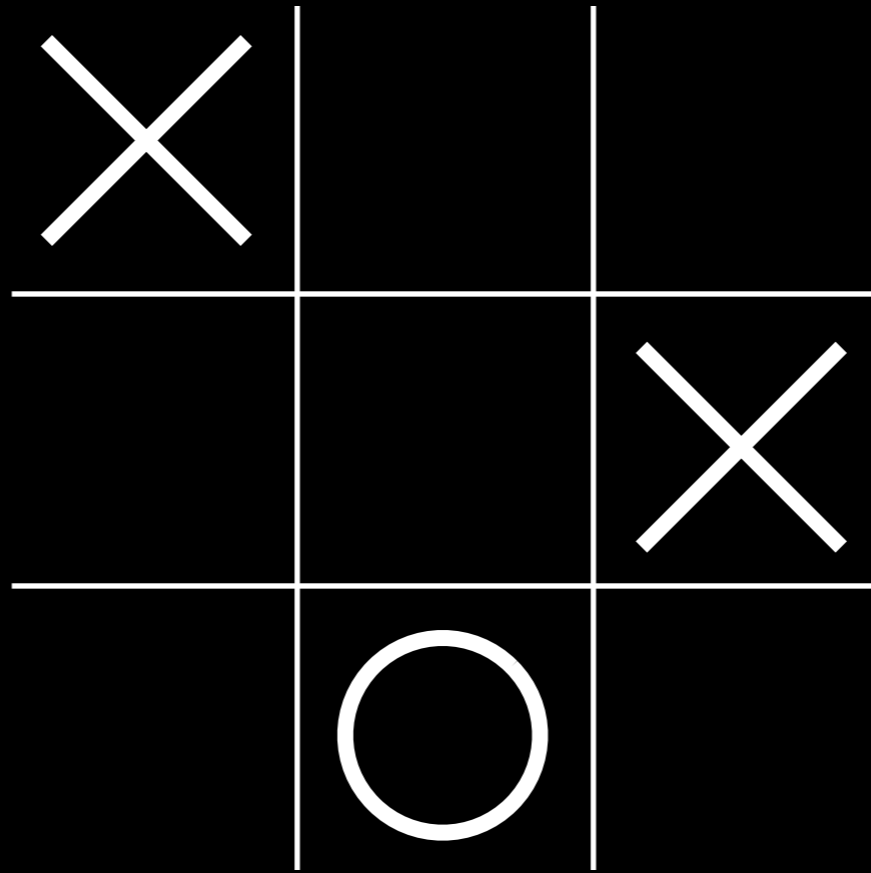
# Outline

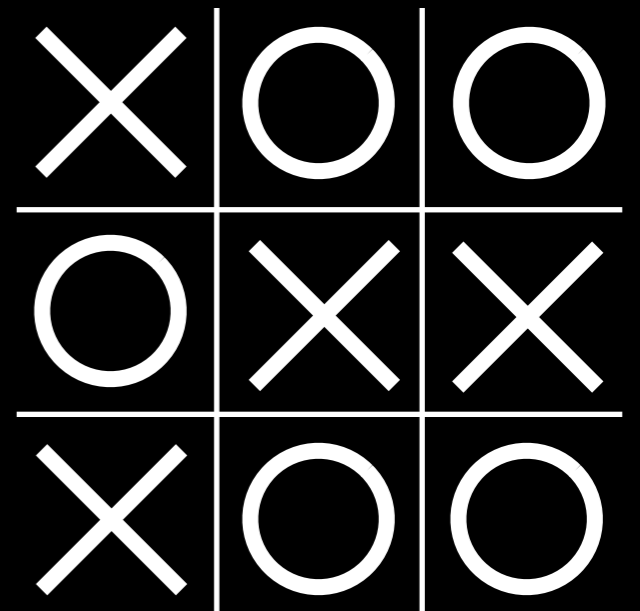
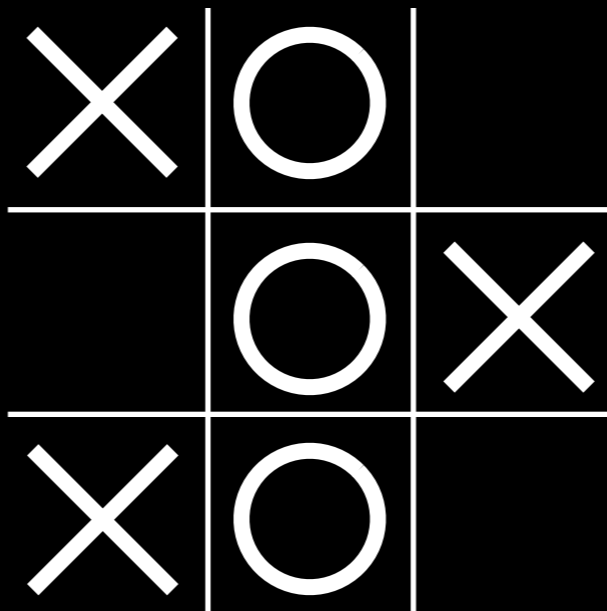
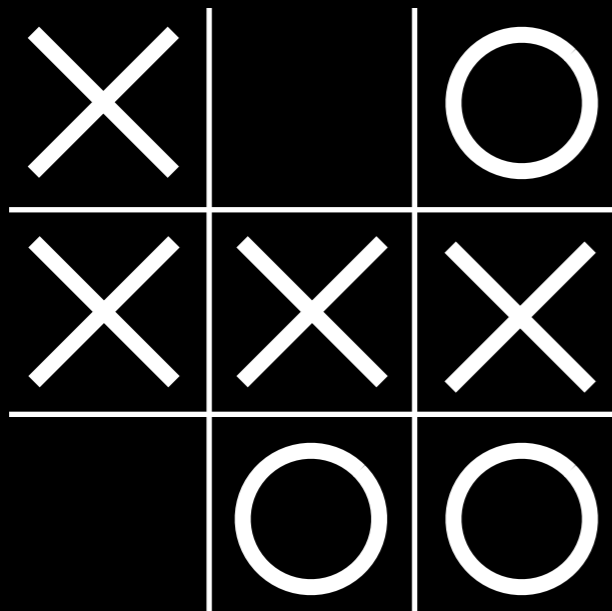
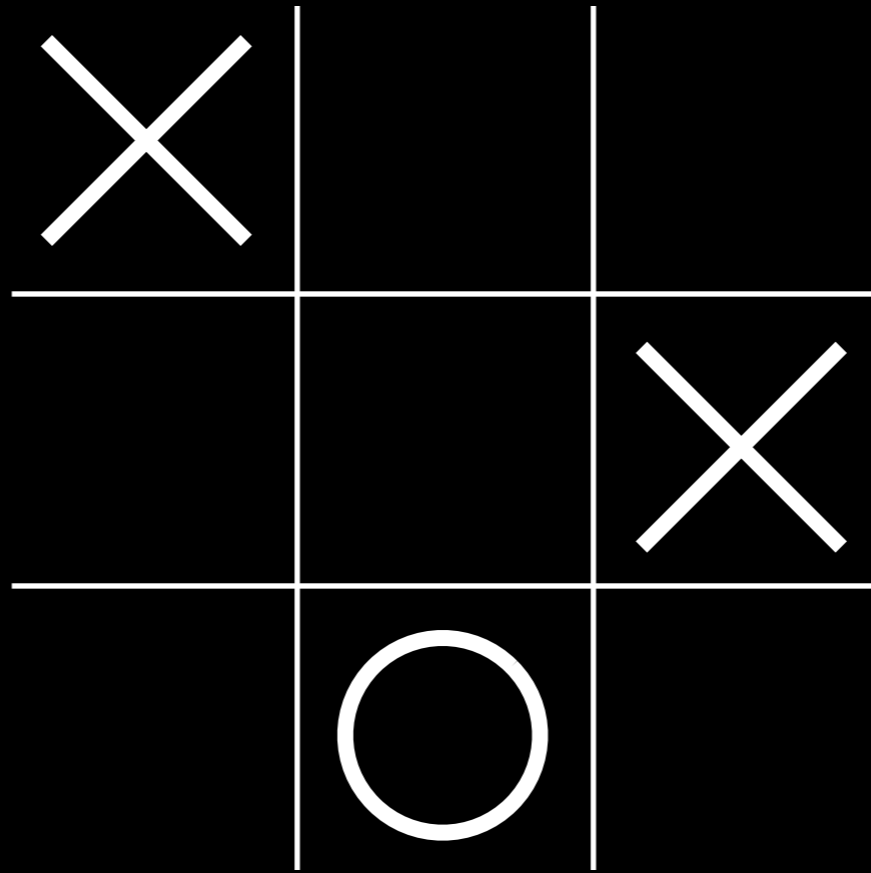
- In deterministic, perfect information, zero-sum games:
  - How to find optimal moves in deterministic games
  - How to find good moves when time is limited

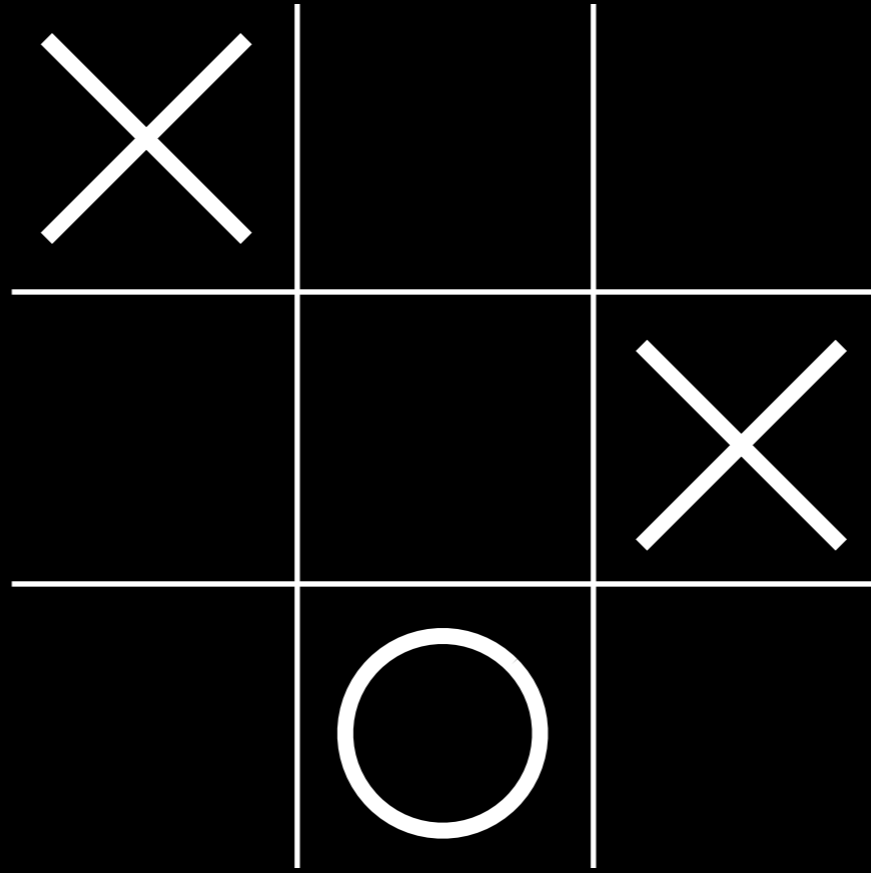


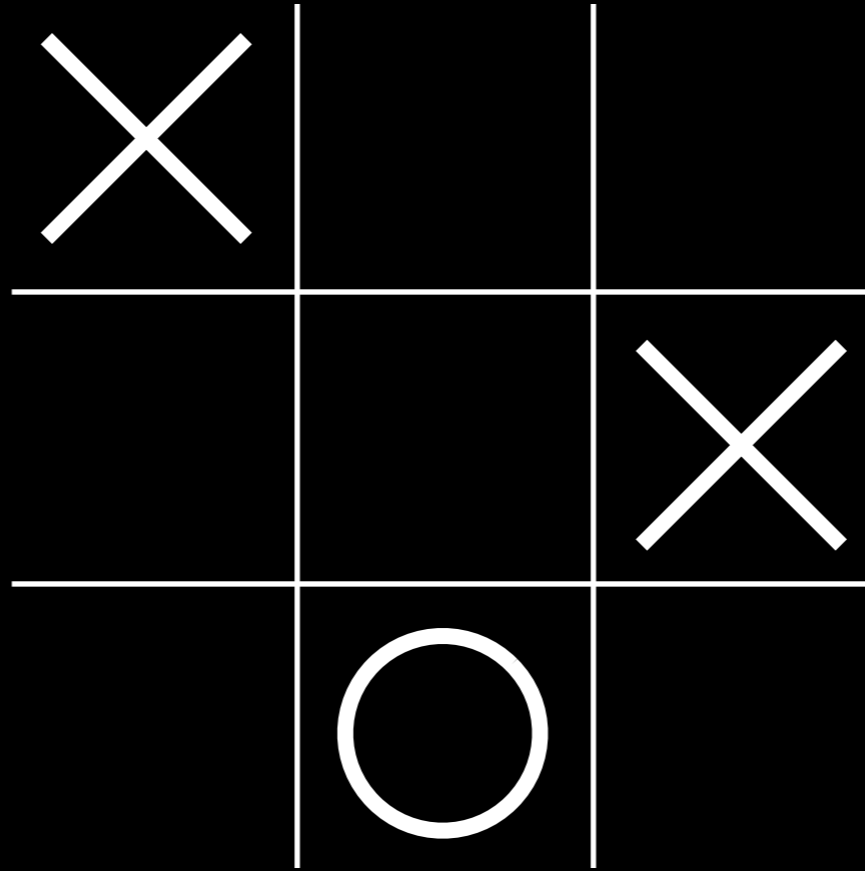




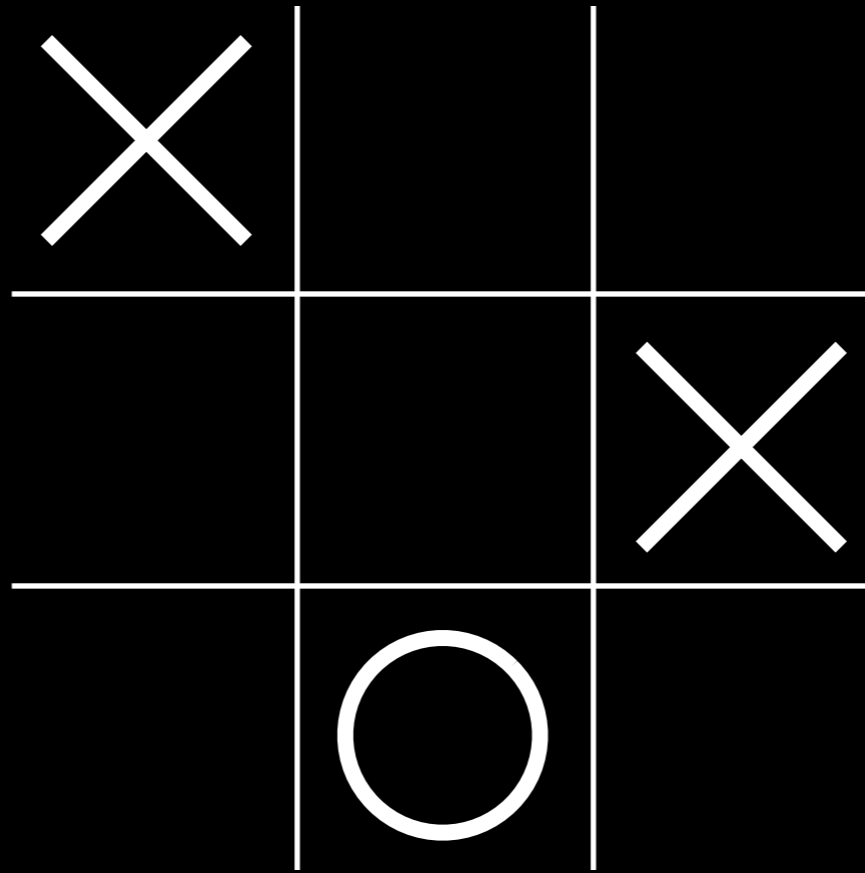




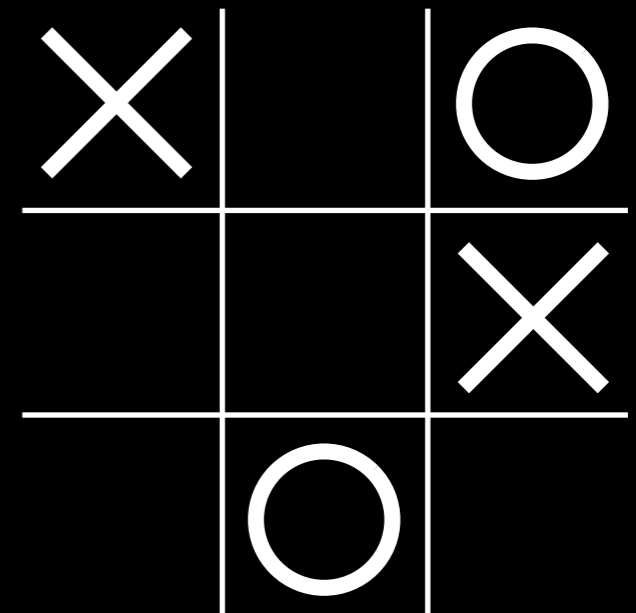
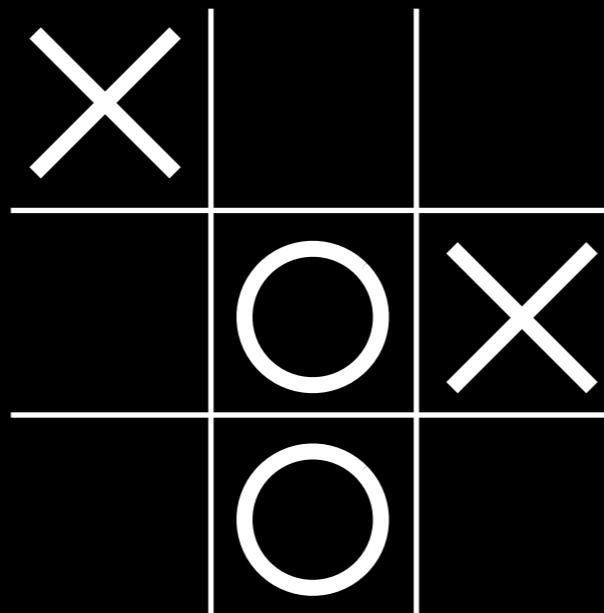
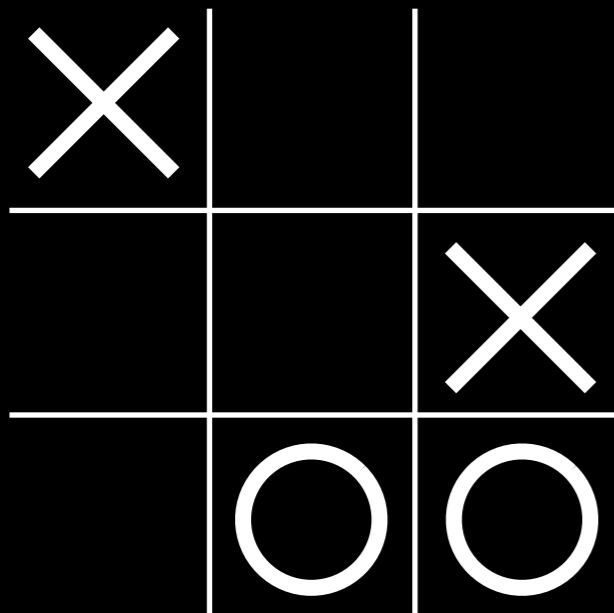




Next Player: O

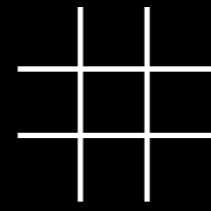


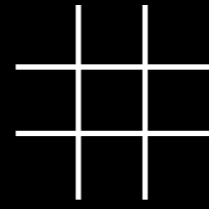
Next Player: O



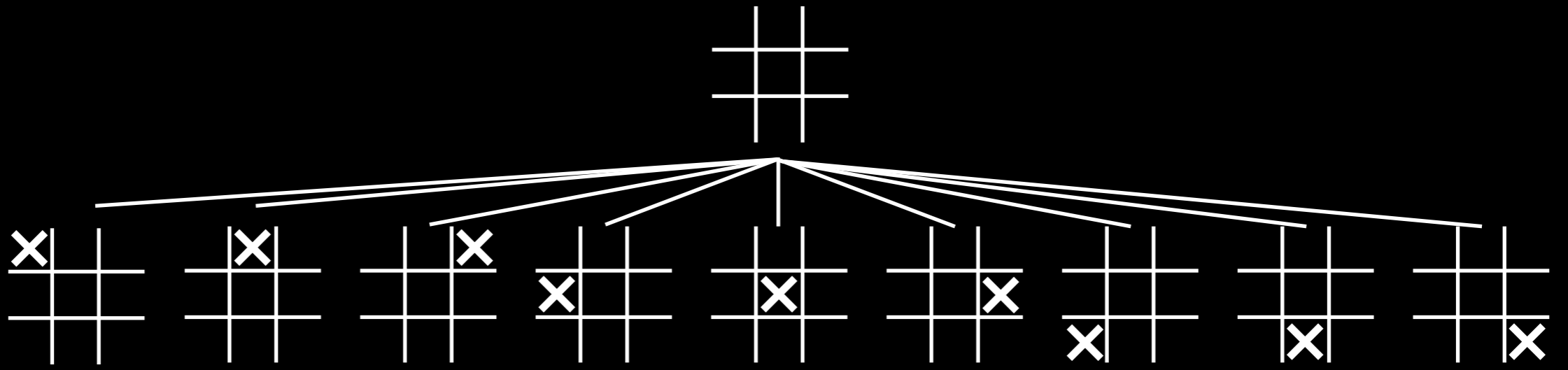






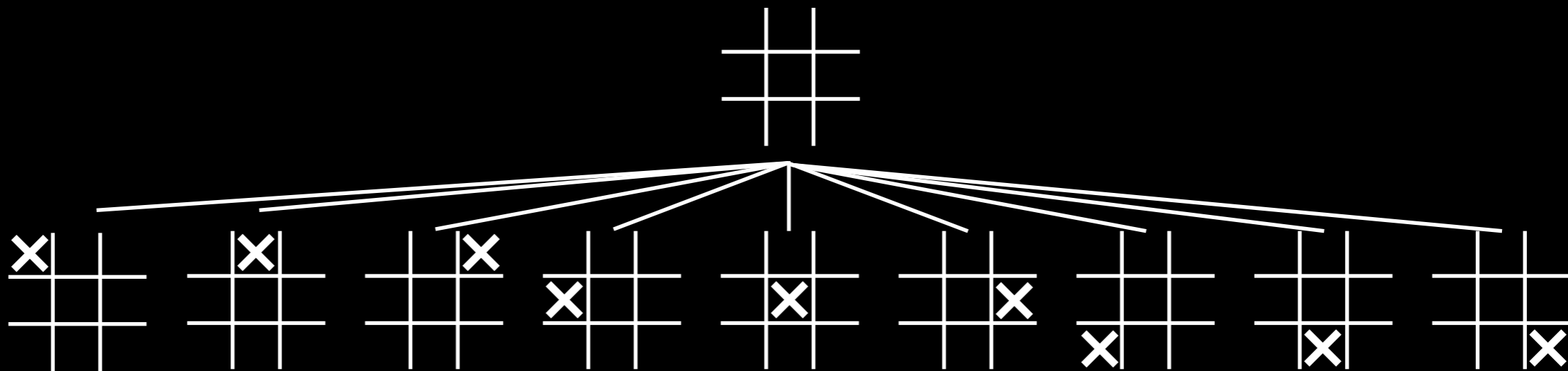


X



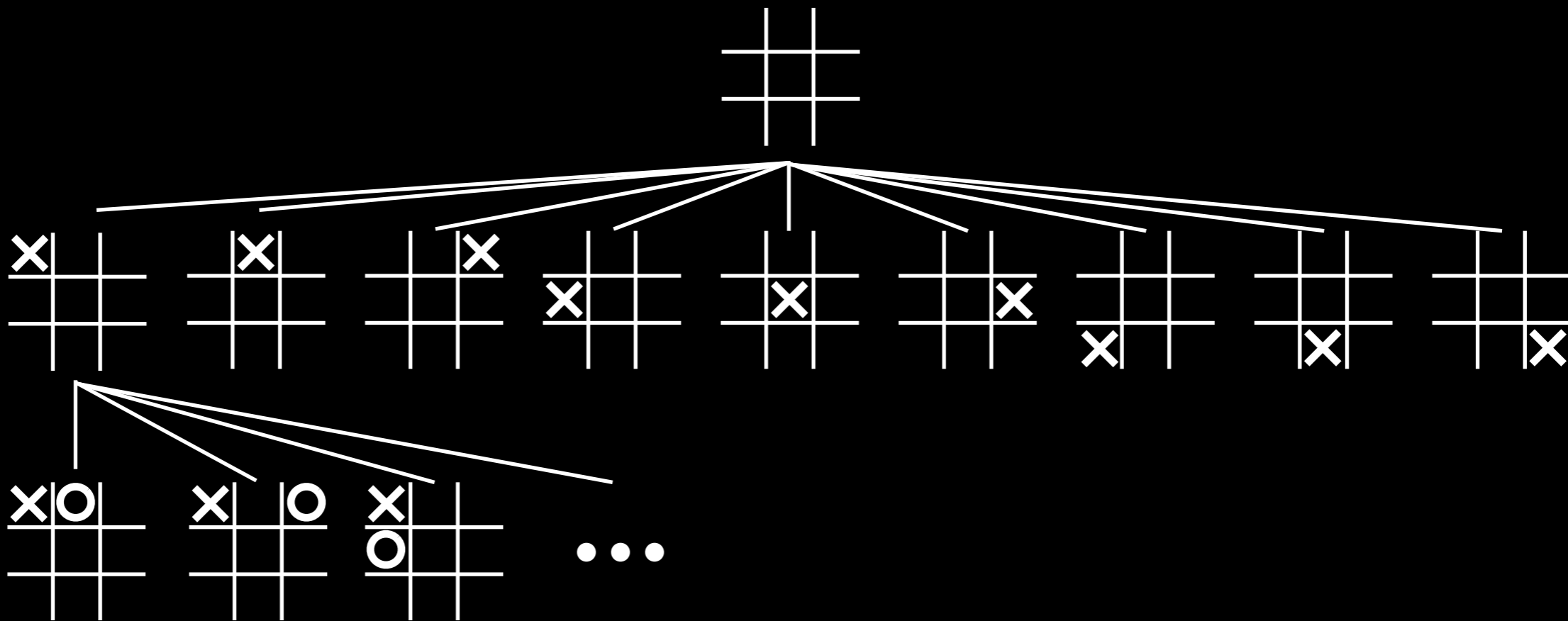
X

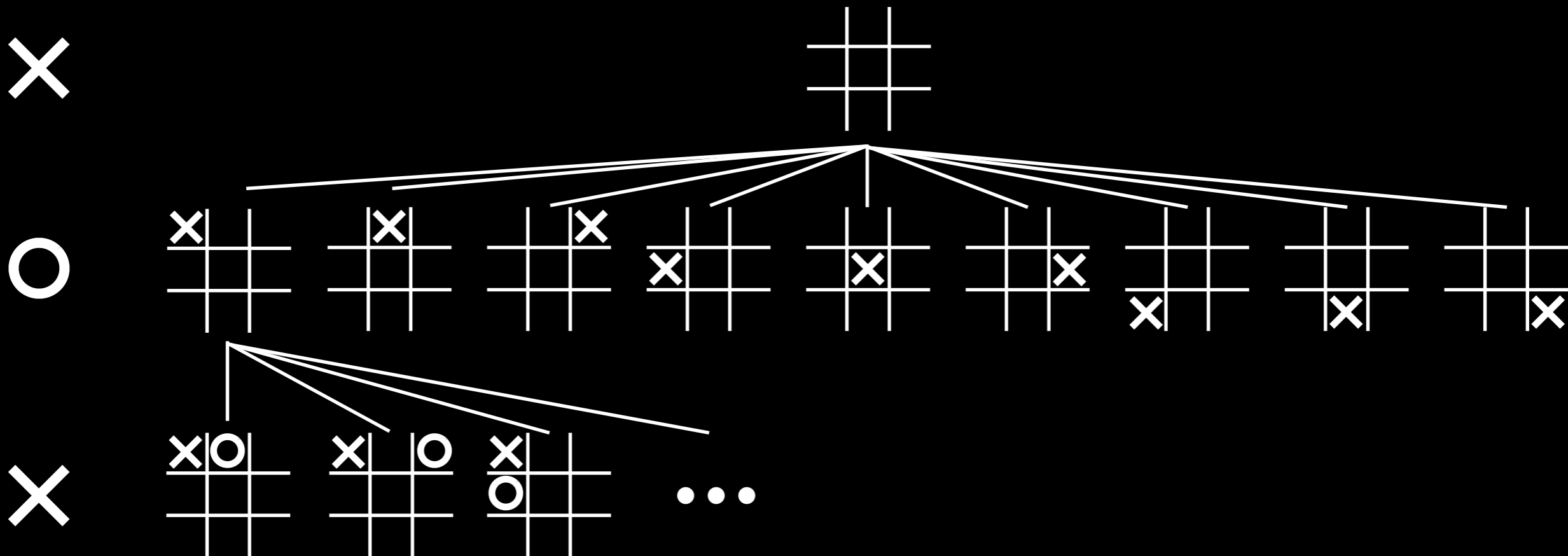
O



X

O





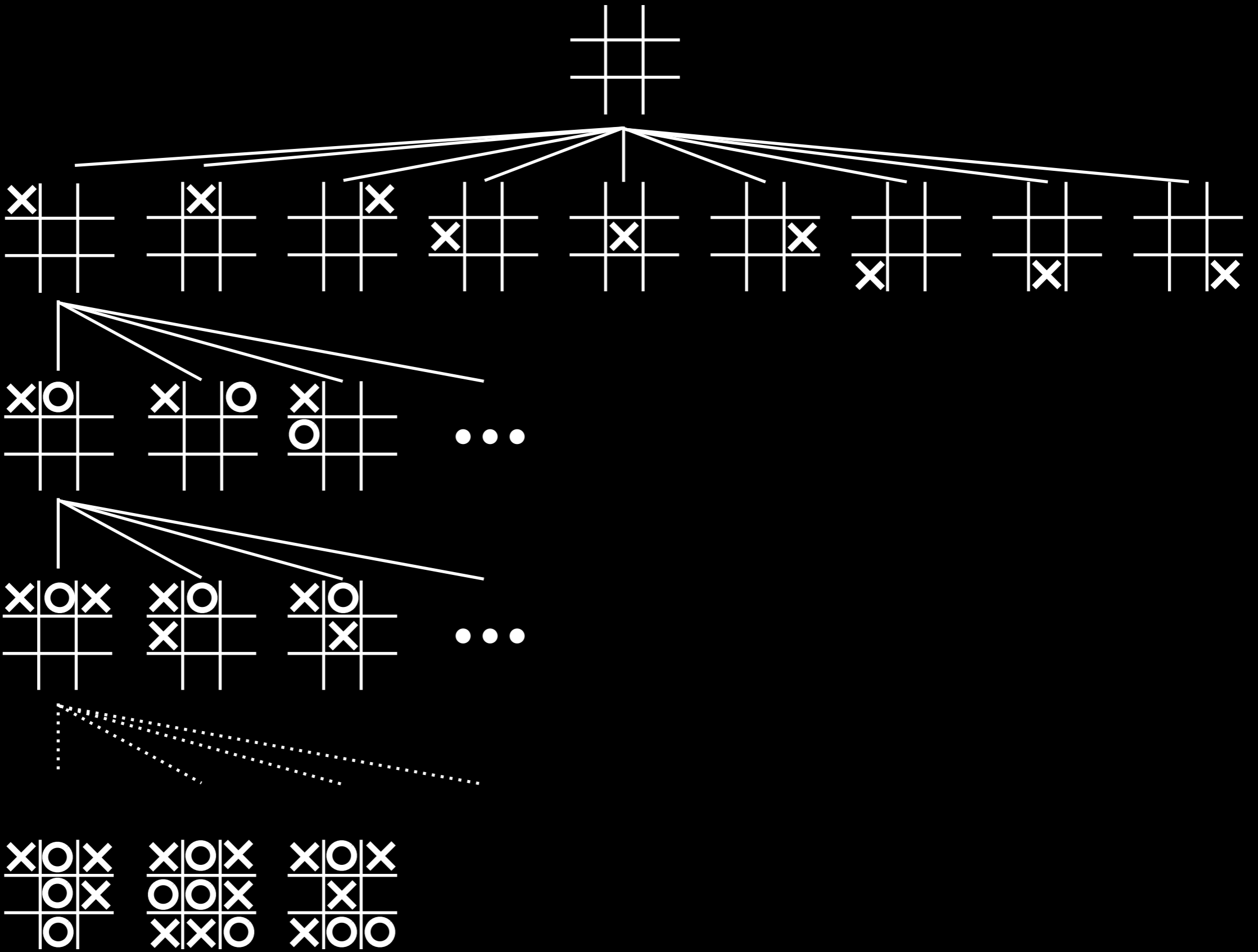
X

O

X

O

Term.



X	O	X
	O	X
	O	

X	O	X
O	O	X
X	X	O

X	O	X
	X	
X	O	O

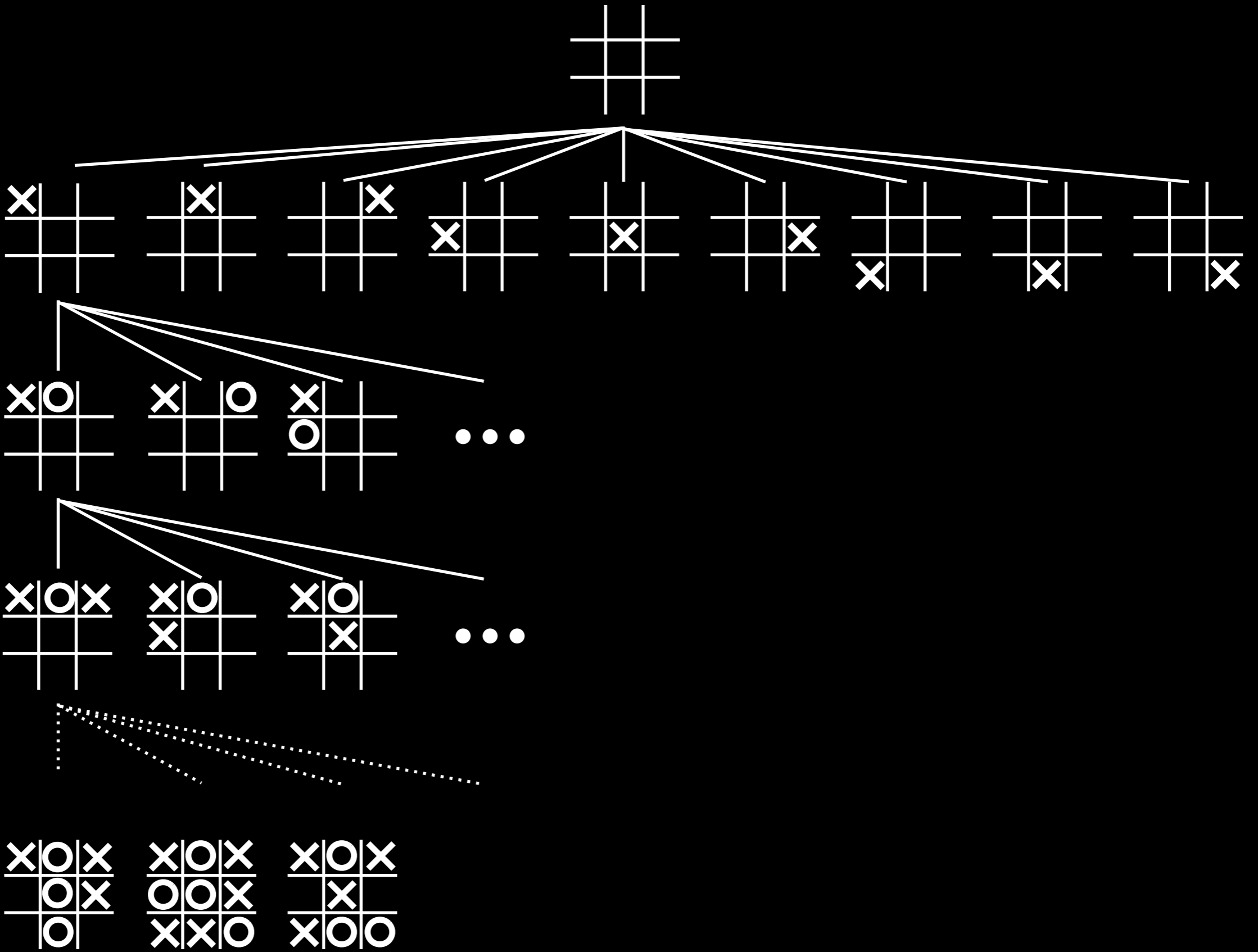
X

O

X

O

Term.





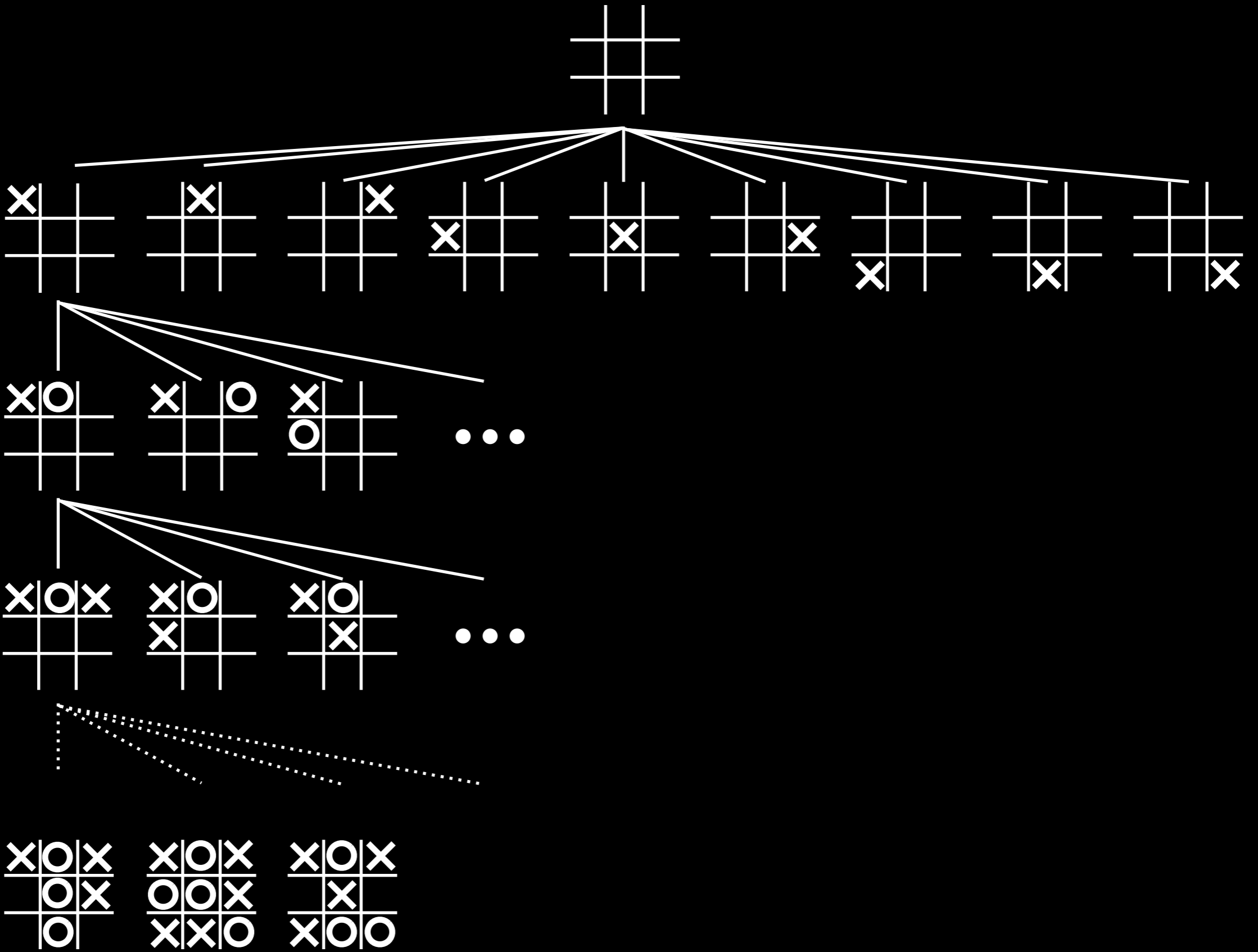
X

O

X

O

Term.



Deciding what to do in a game requires thinking about what the opponent will do, and having a strategy that takes that into account

Deciding what to do in a game requires thinking about what the opponent will do, and having a strategy that takes that into account



Not simply a sequence of actions, but a contingency plan that specifies what to do depending on what state one finds oneself in (AIMA 4.3)

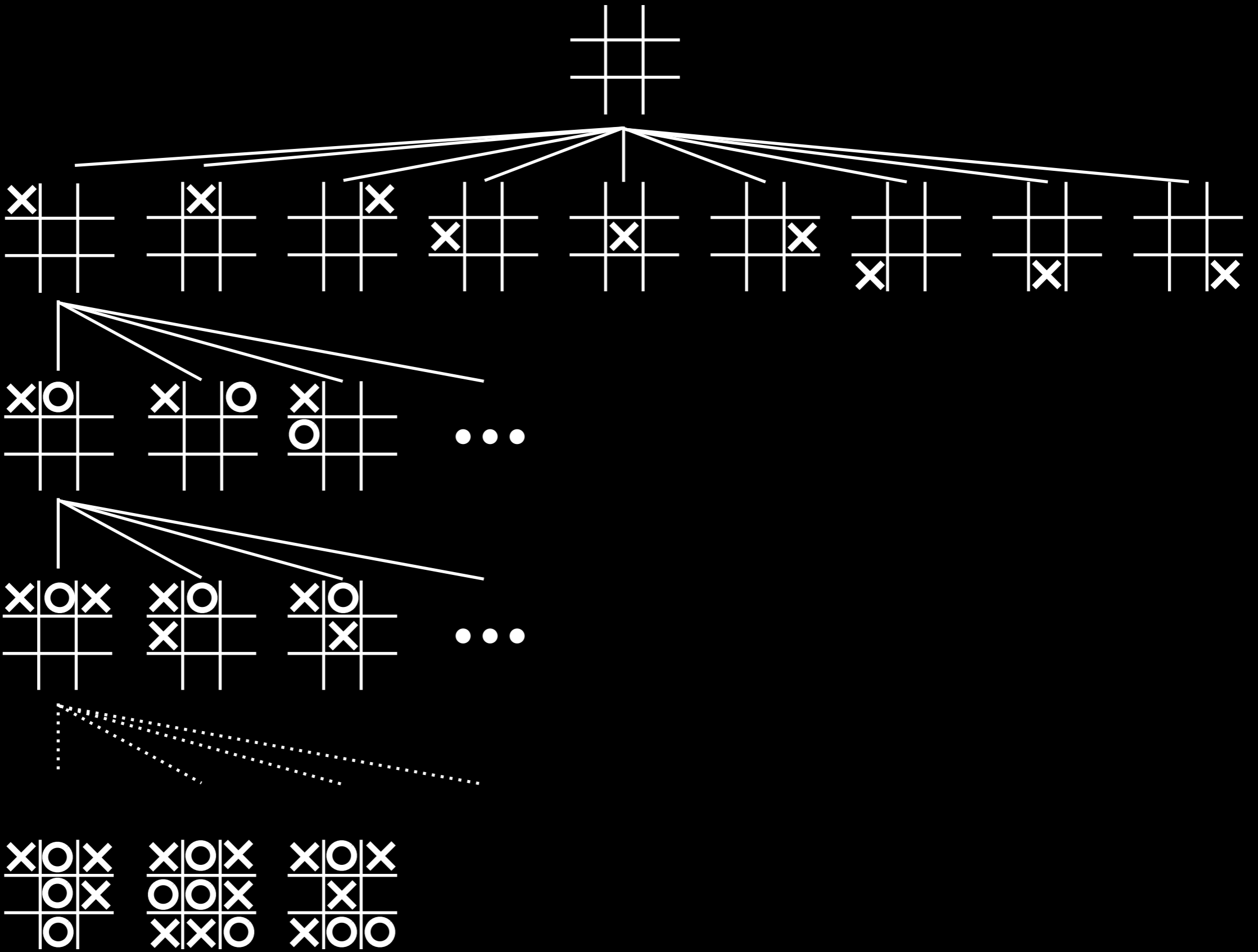
X

O

X

O

Term.

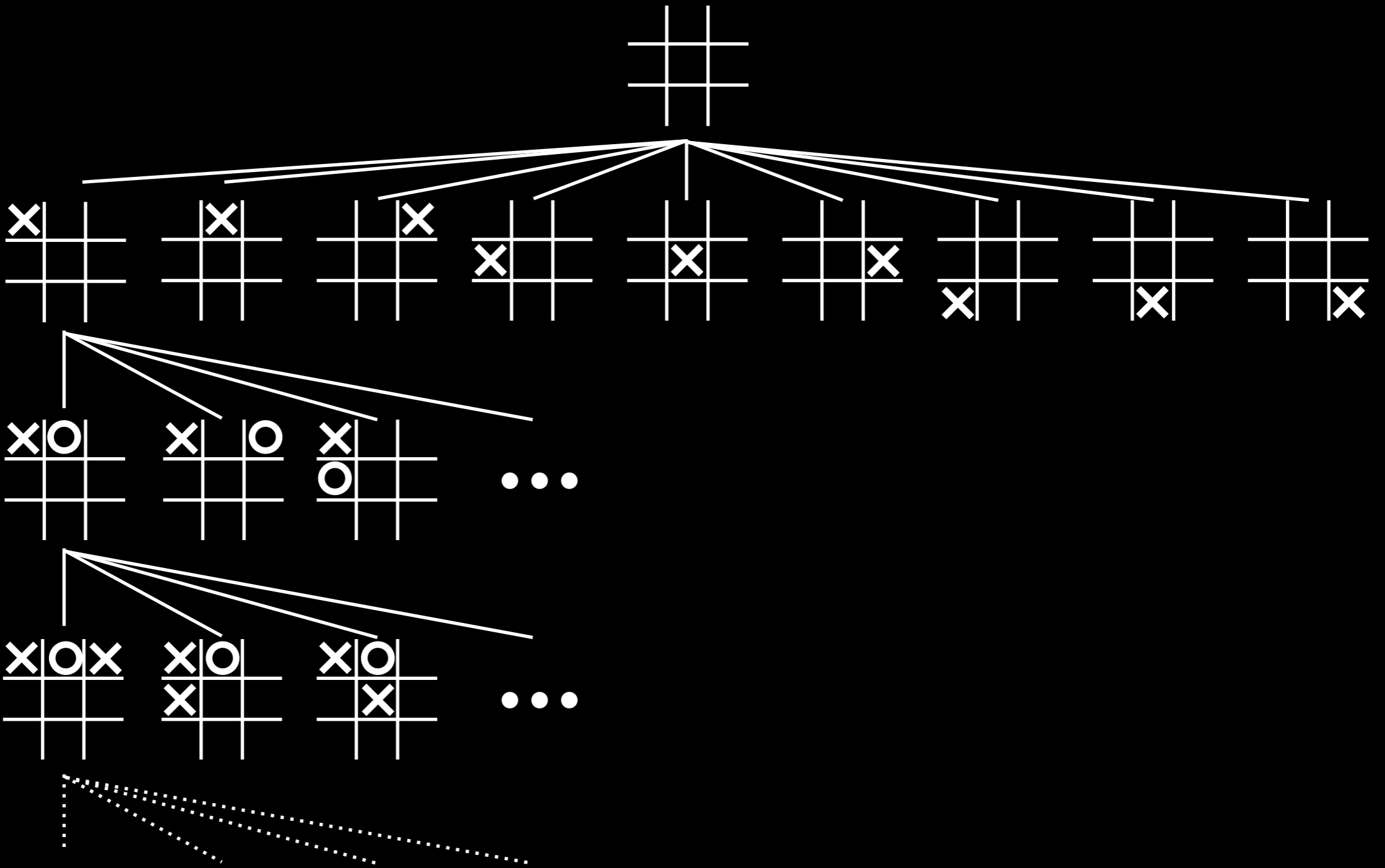


X

O

X

O



Term.



Utility

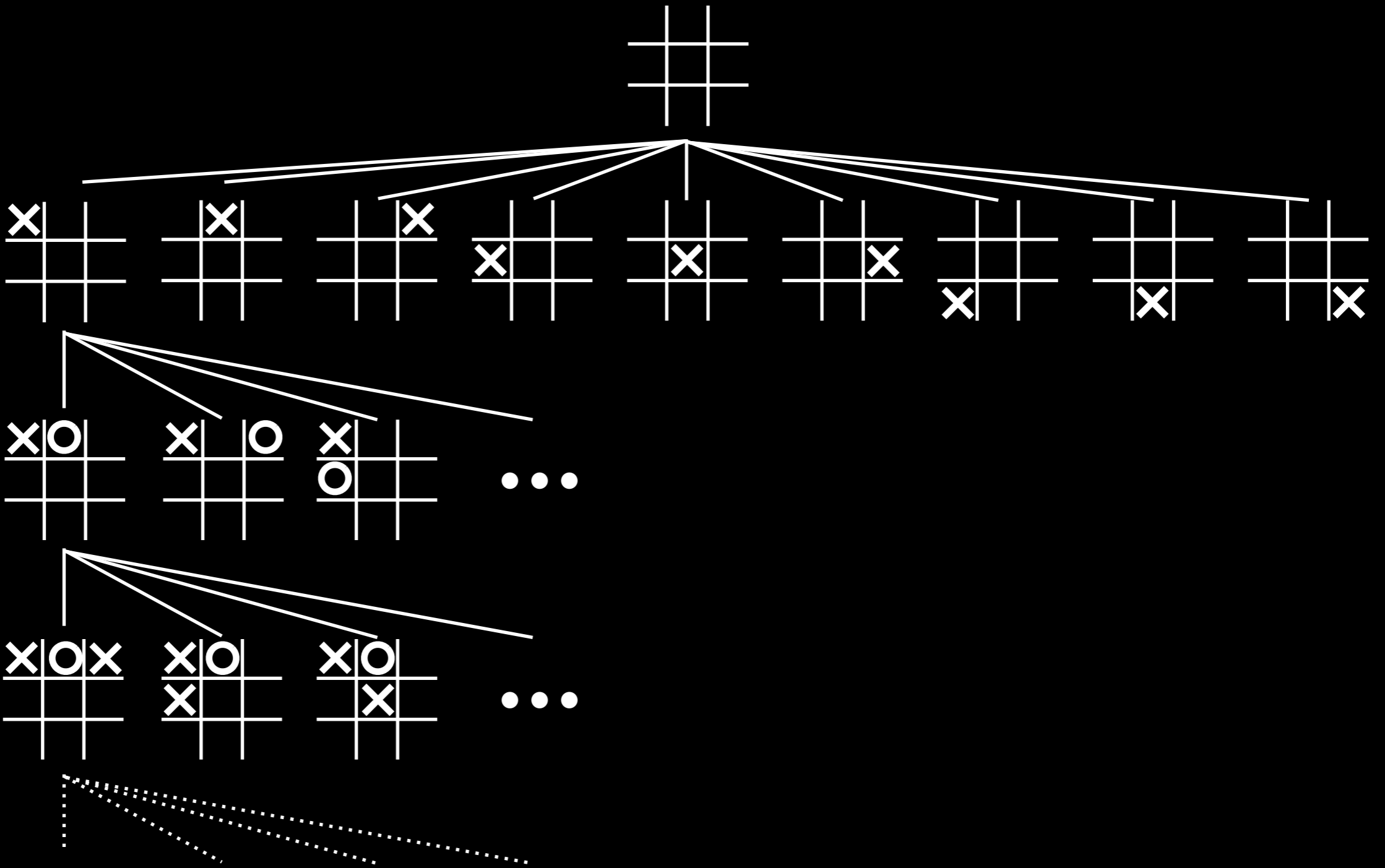
-1    0    +1

X

O

X

O

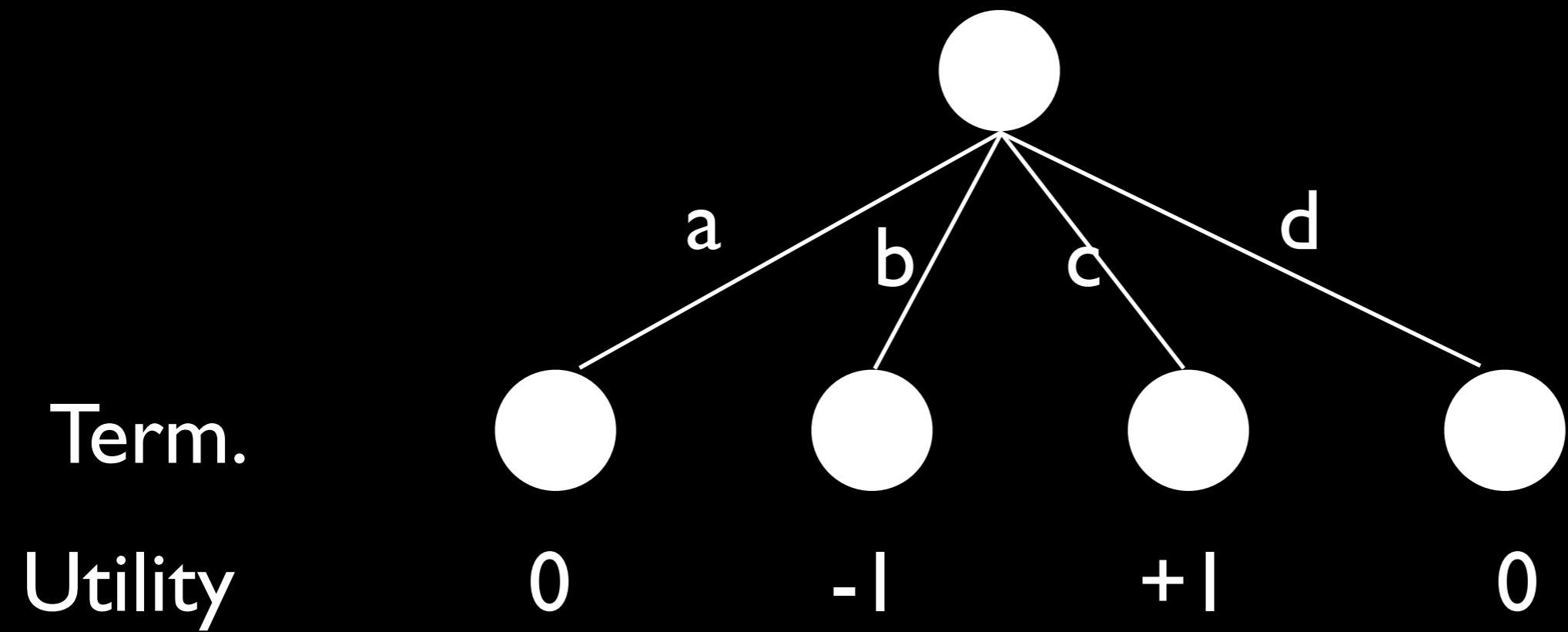


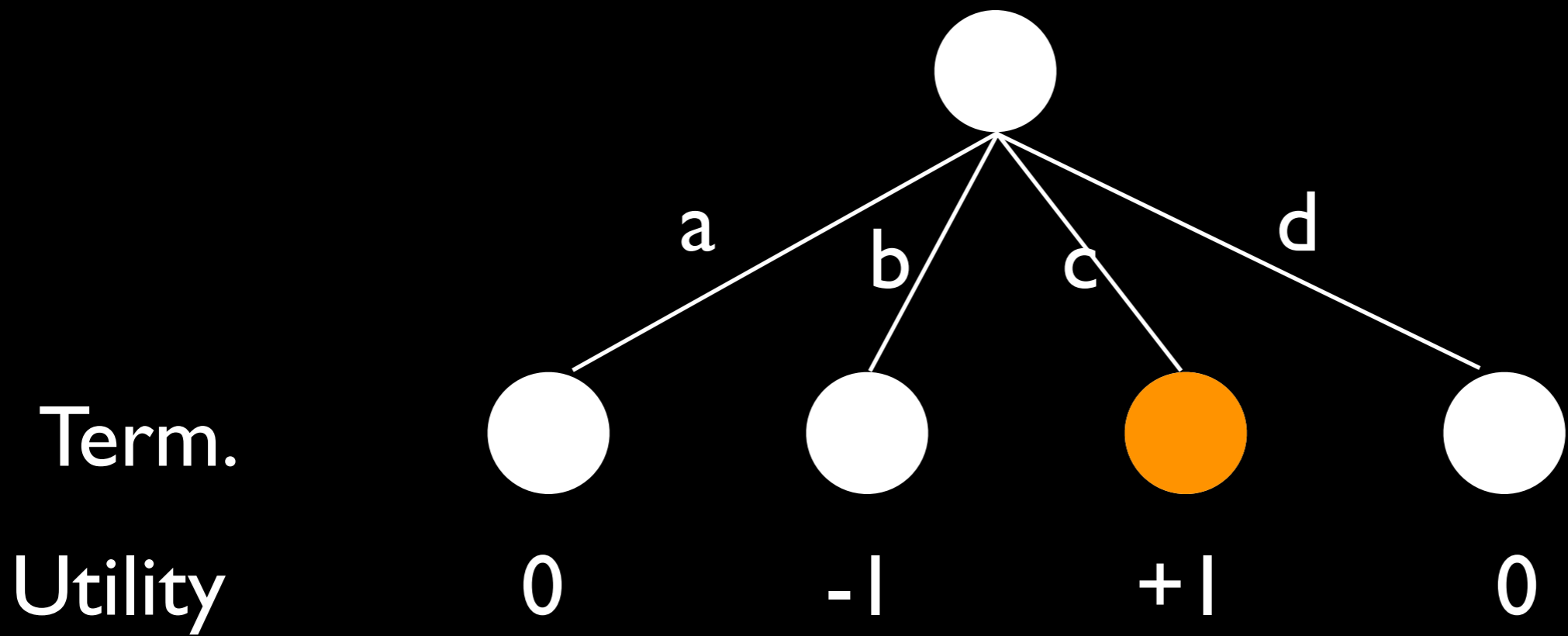
Term.



Utility

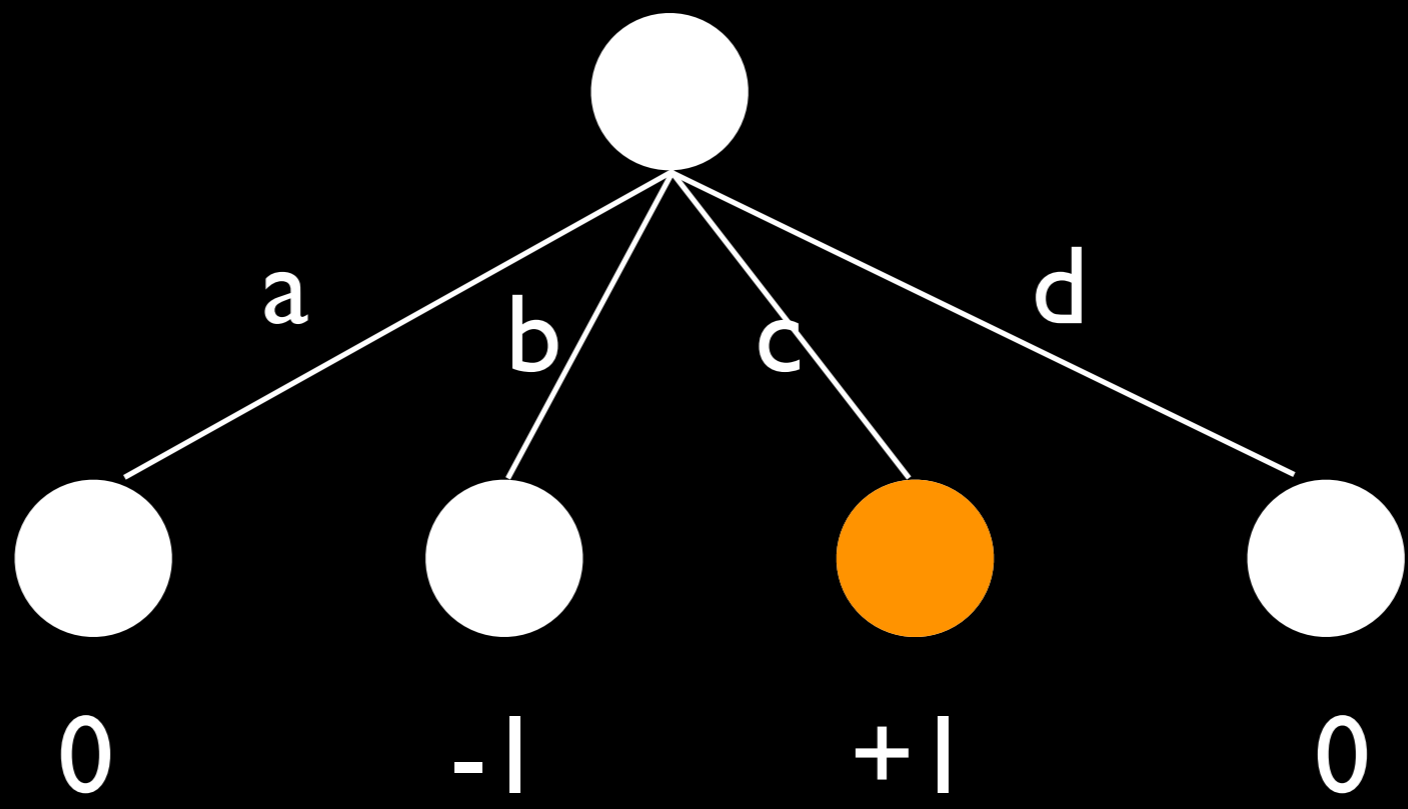
-1    0    +1







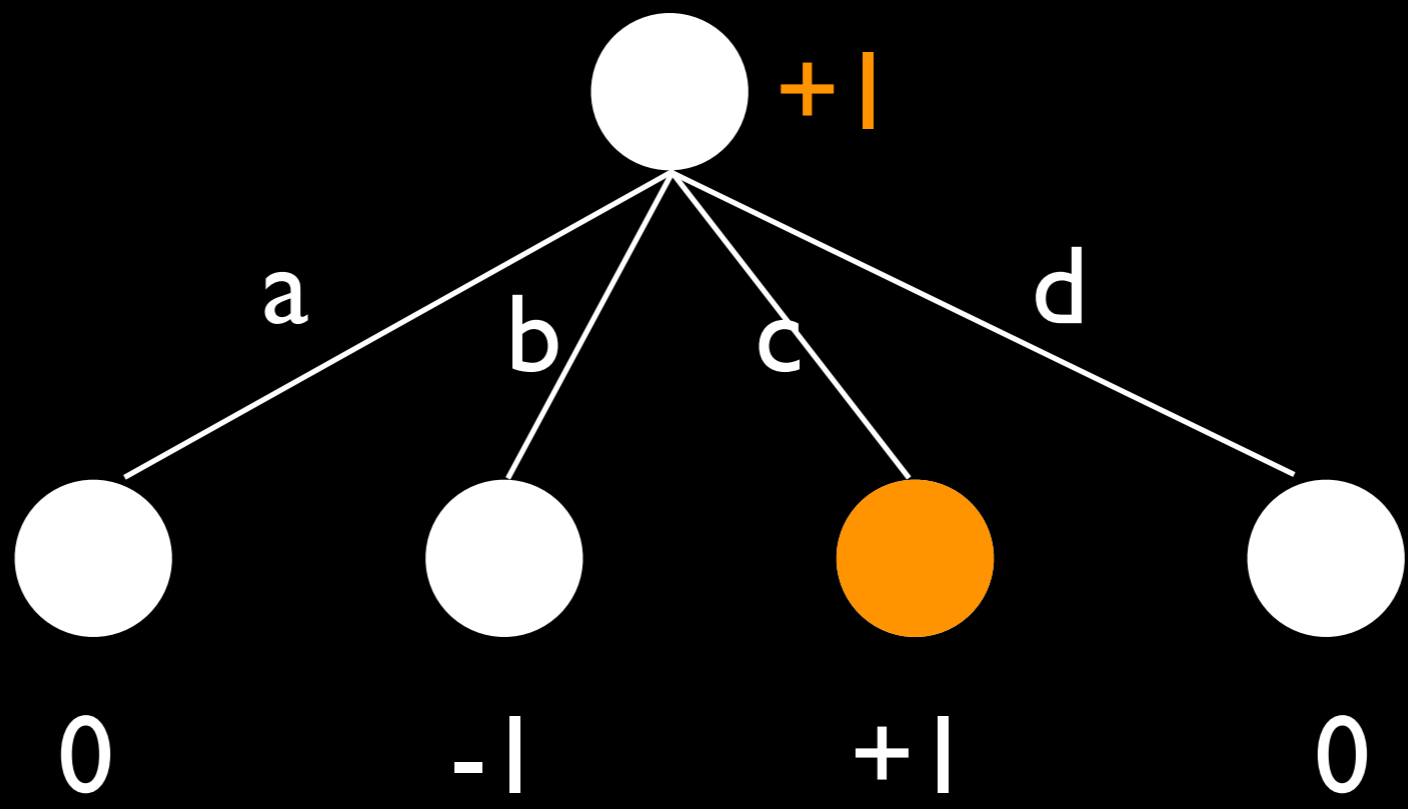
MAX



Term.

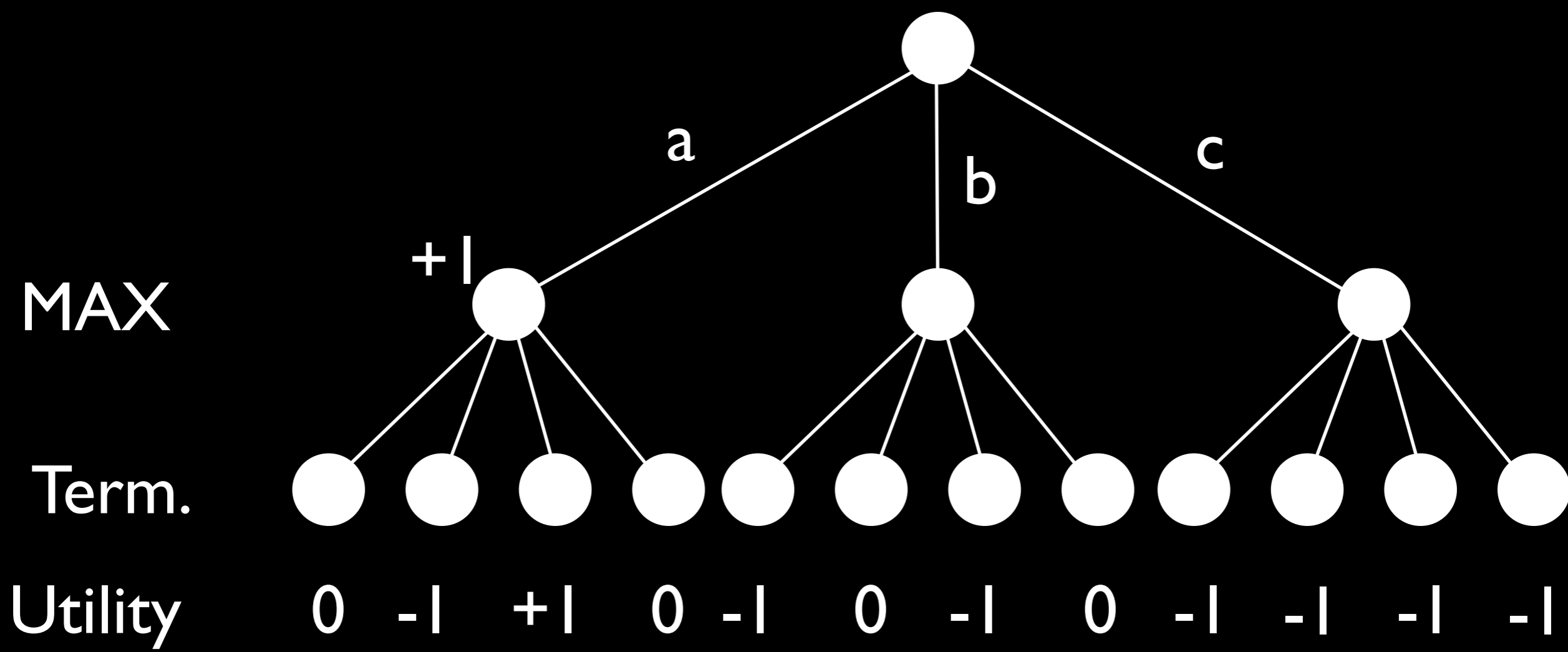
Utility

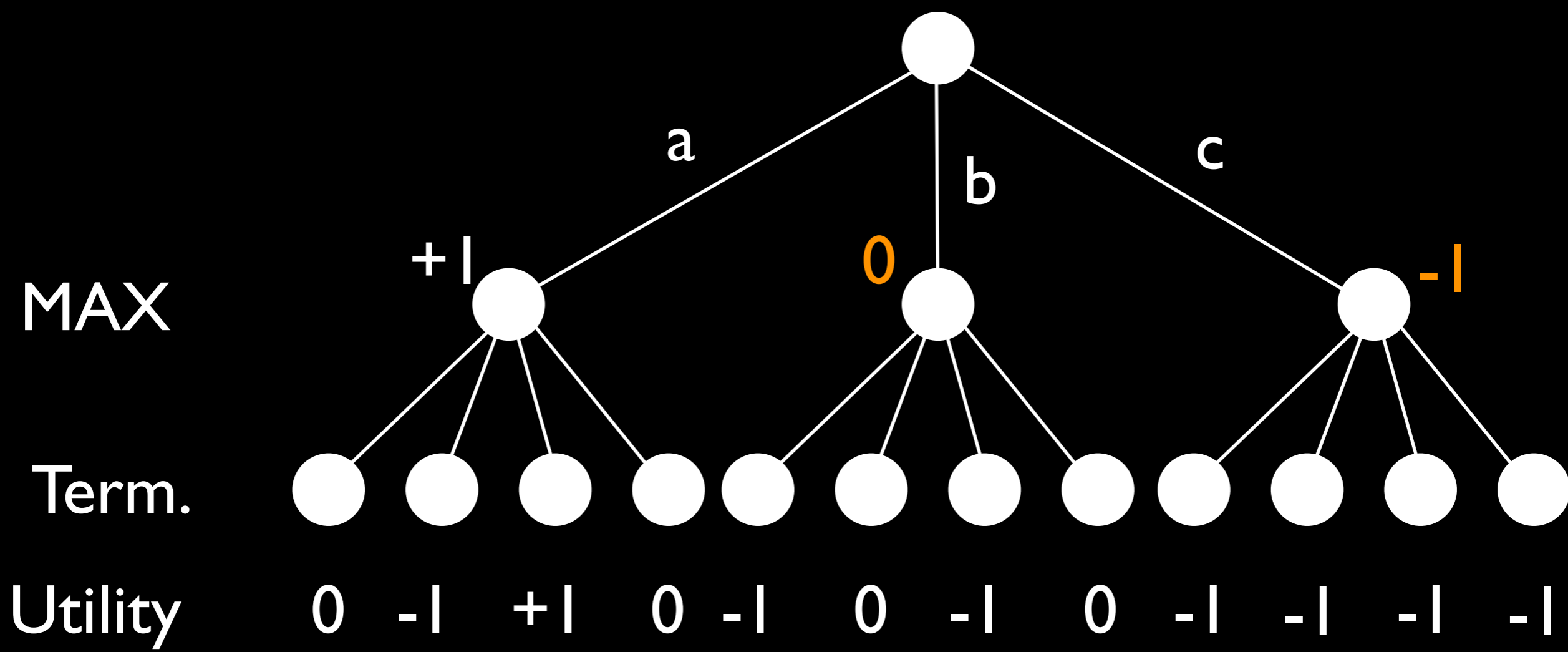
MAX

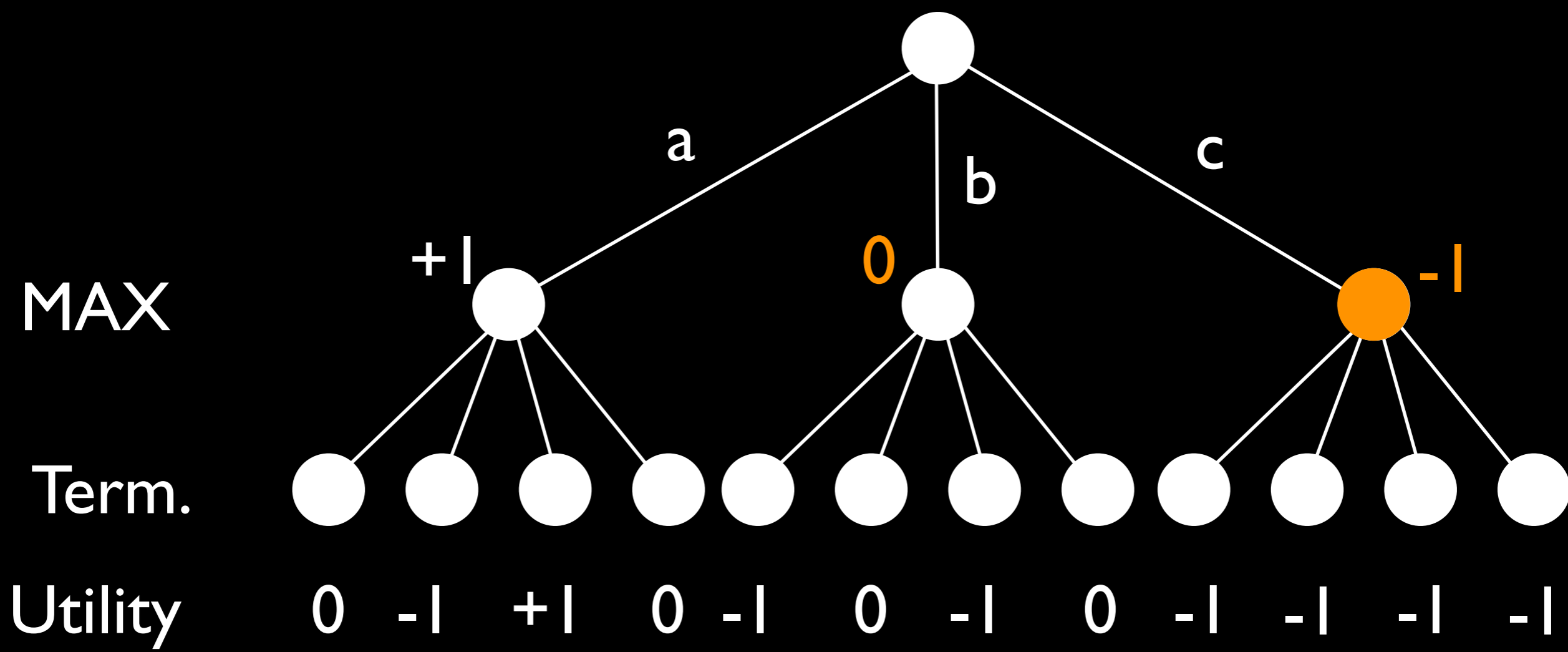


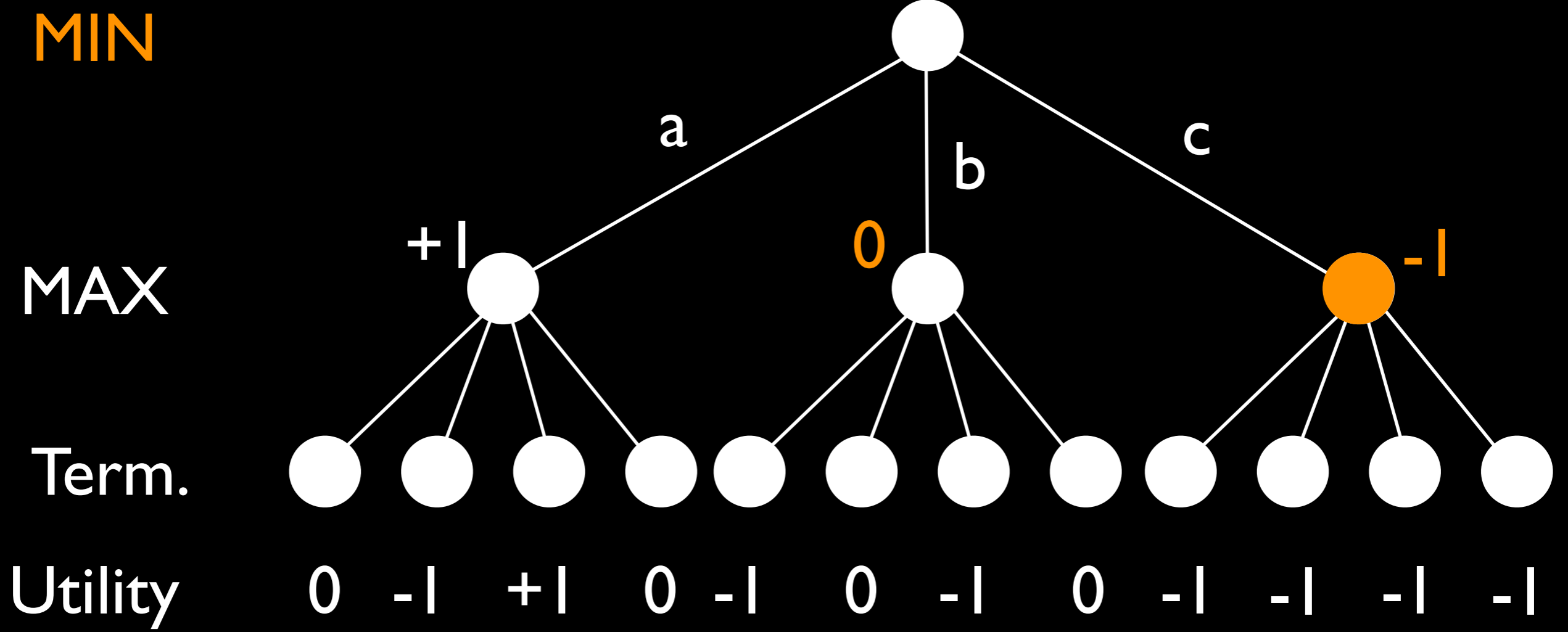
Term.

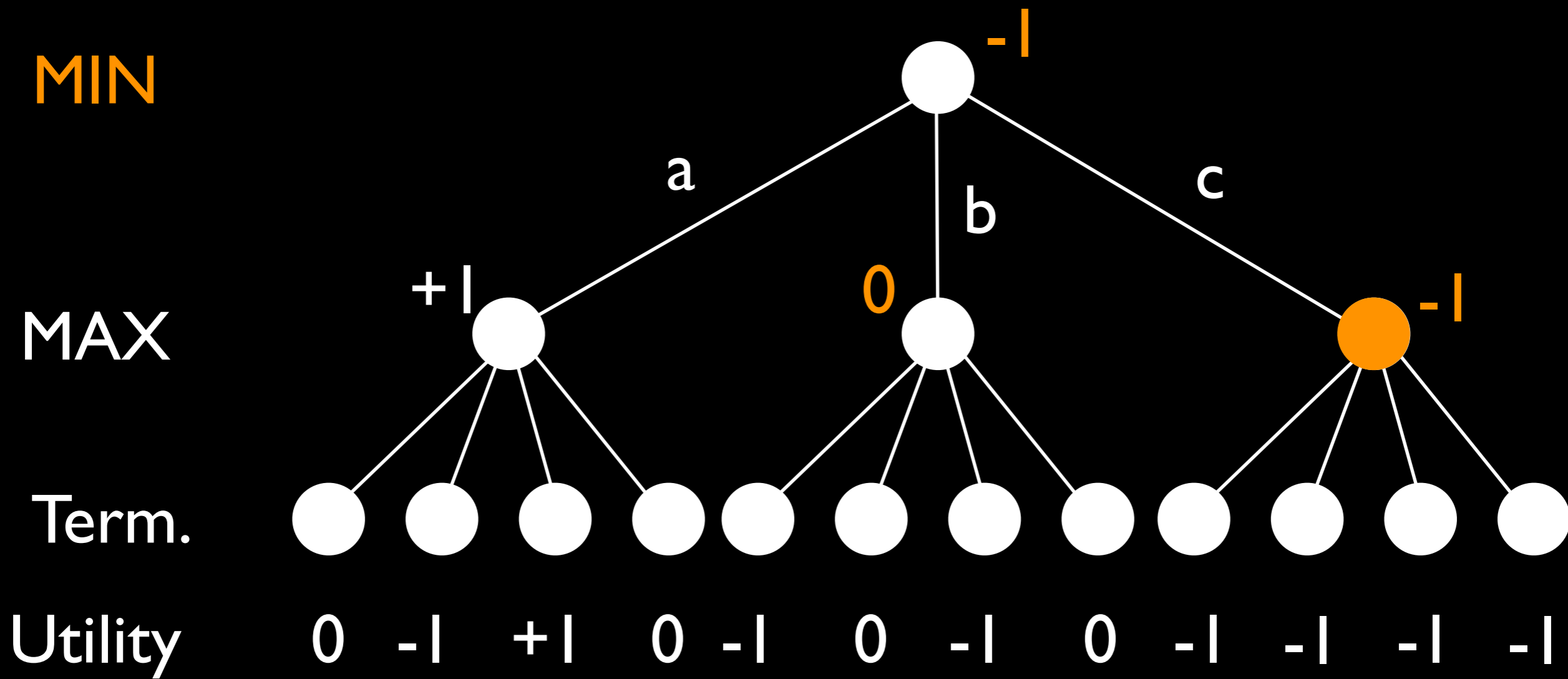
Utility

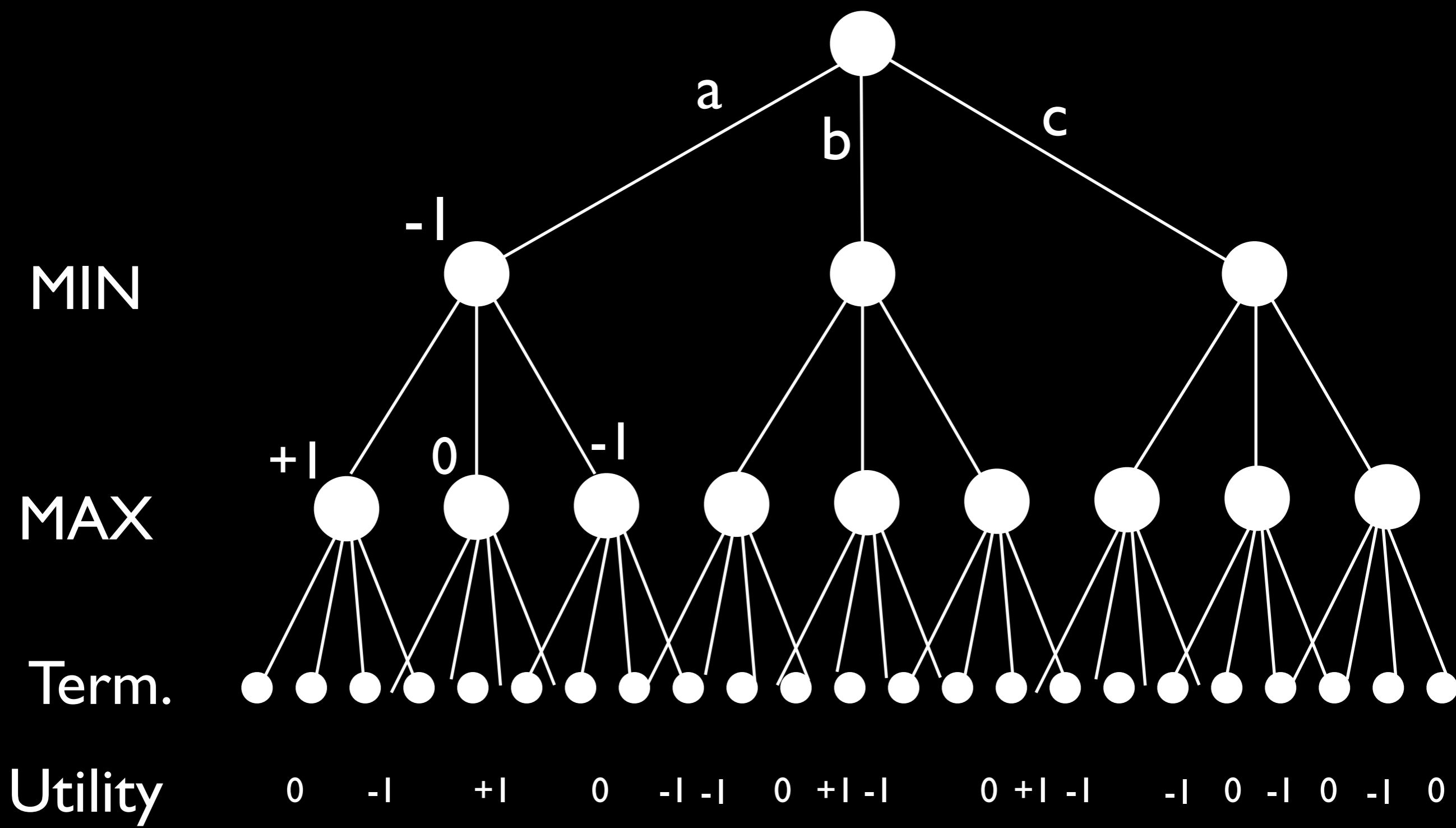




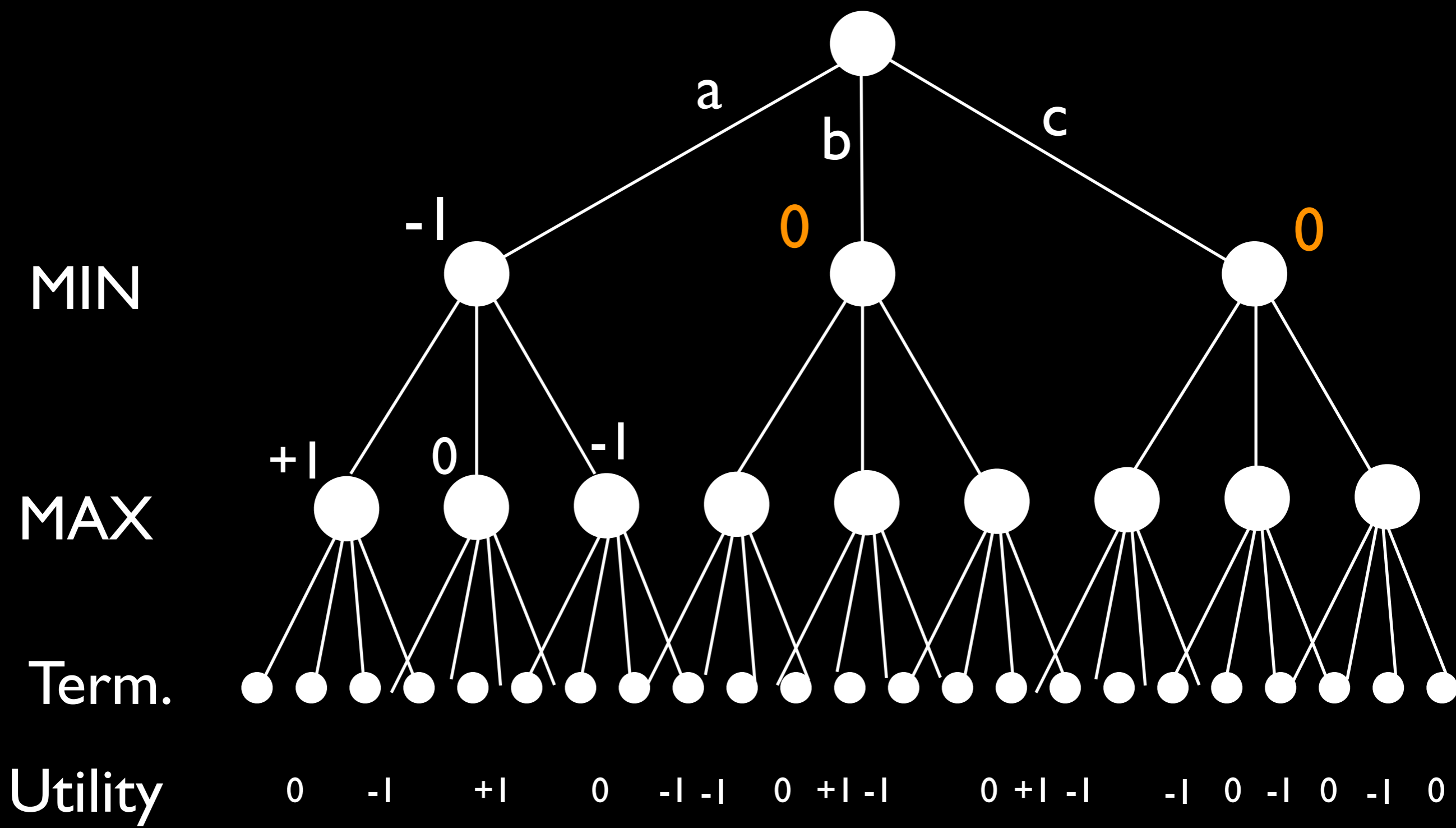












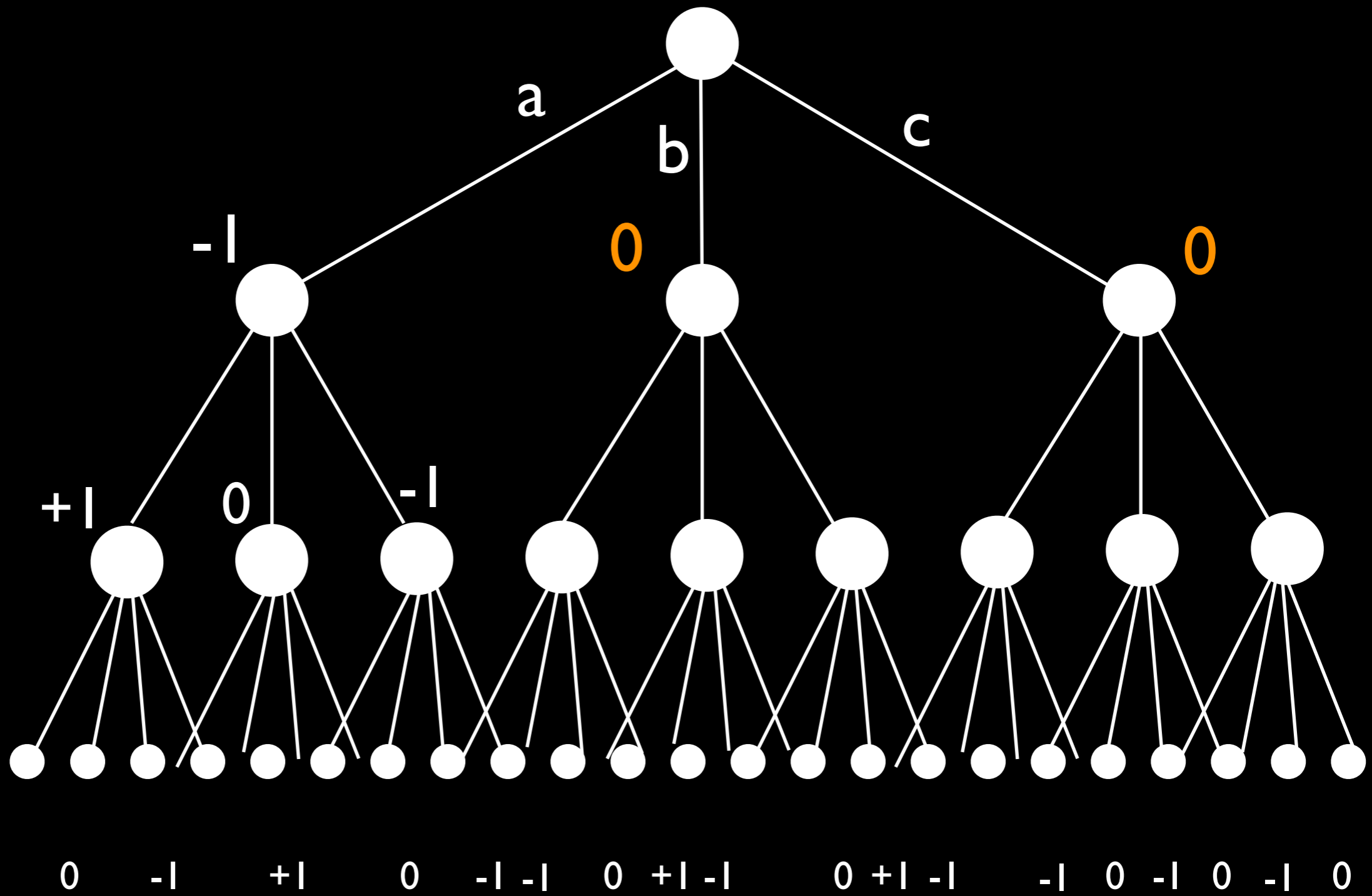
MAX

MIN

MAX

Term.

Utility



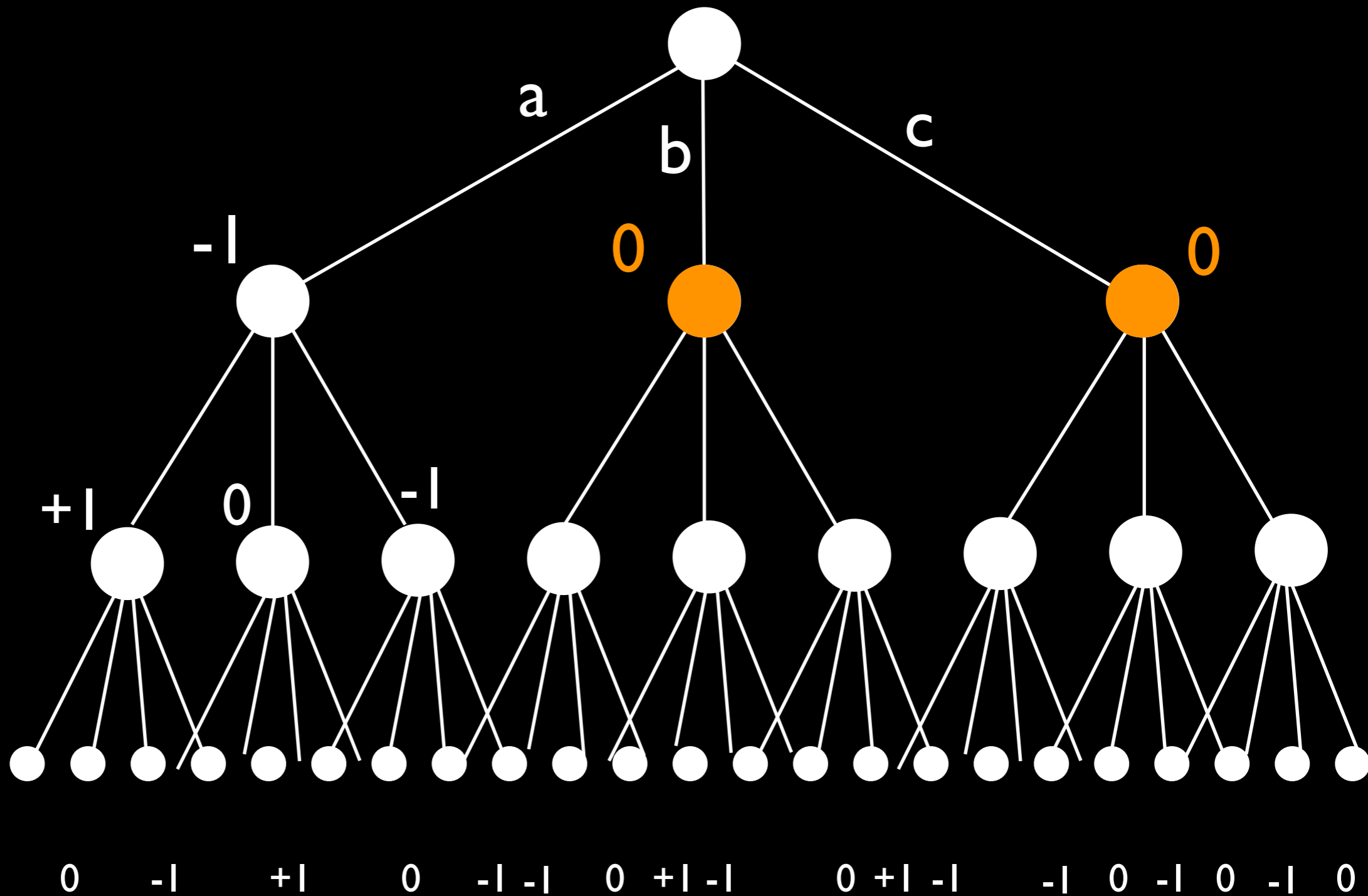
MAX

MIN

MAX

Term.

Utility



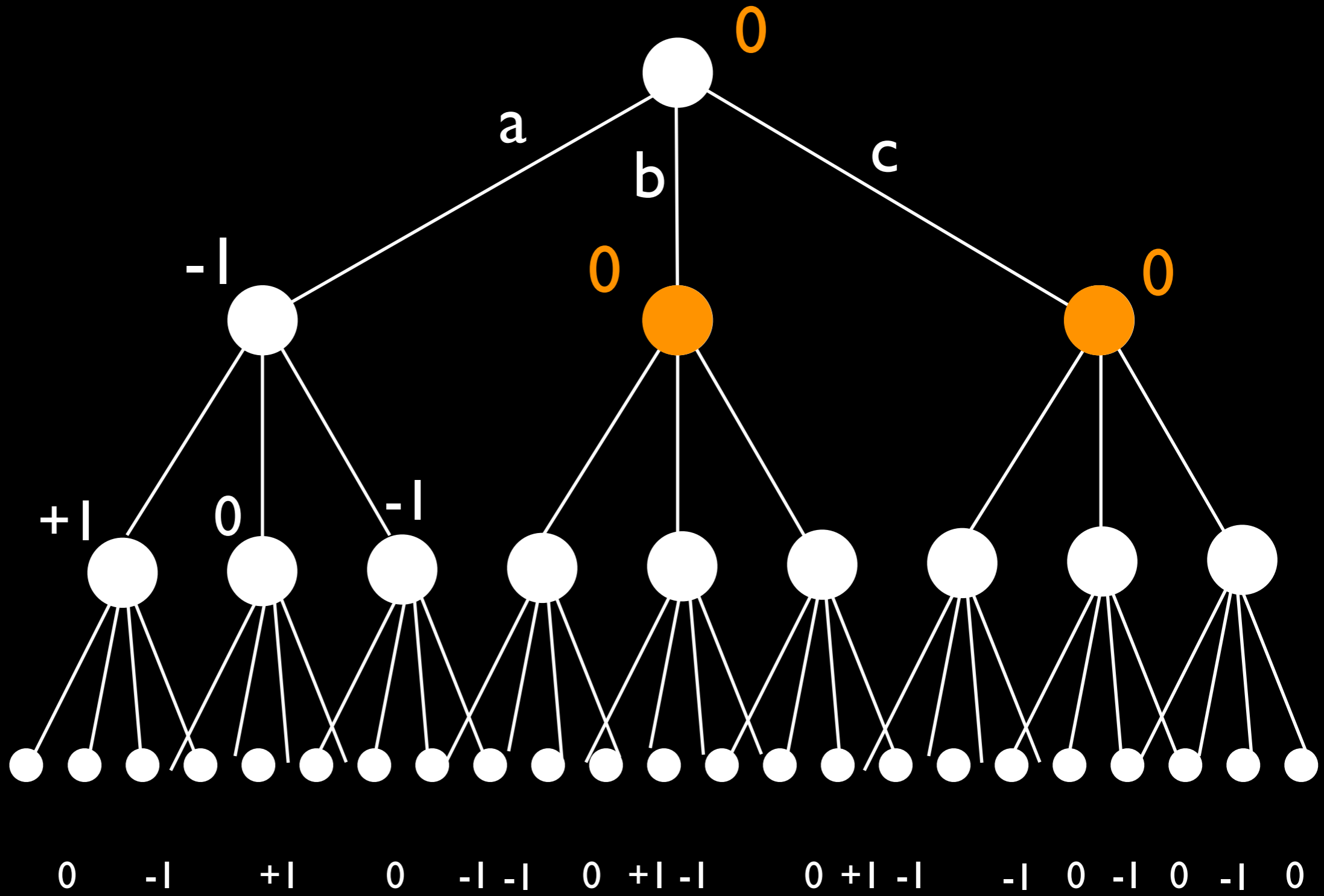
MAX

MIN

MAX

Term.

Utility

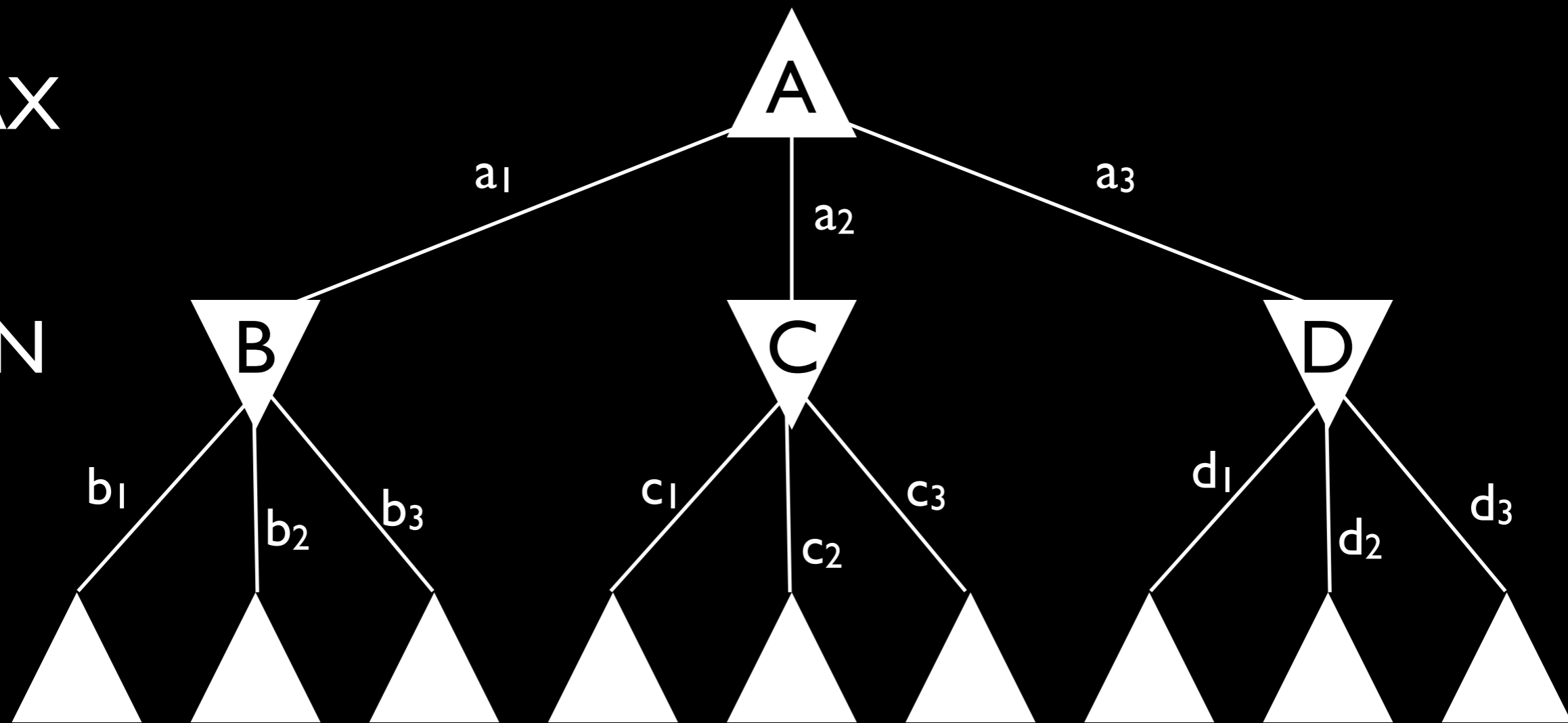


# Minimax Algorithm

$$\text{MINIMAX}(s) = \begin{cases} \text{UTILITY}(s) & \text{if } \text{TERMINAL-TEST}(s) \\ \max_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MAX} \\ \min_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MIN} \end{cases}$$

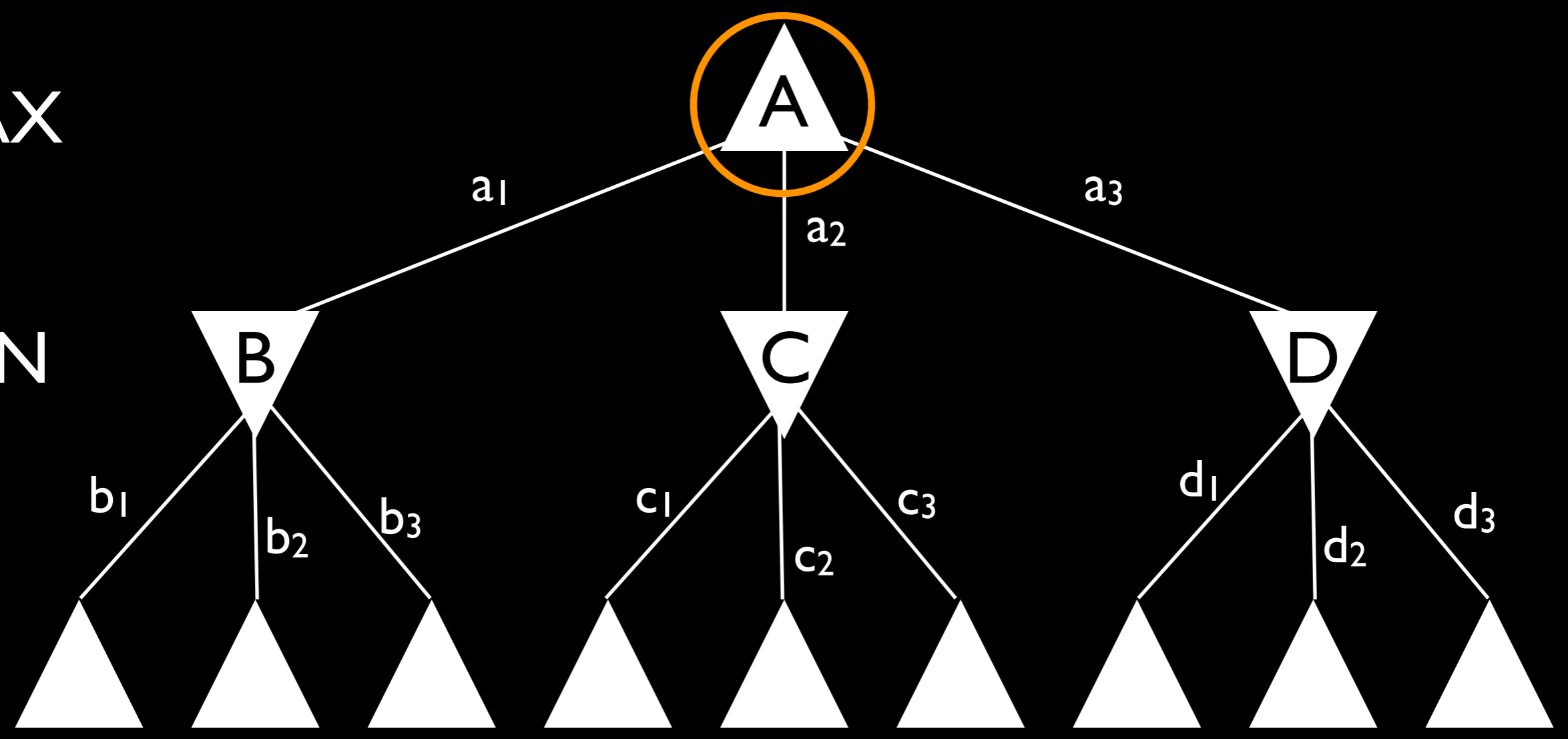
MAX

MIN



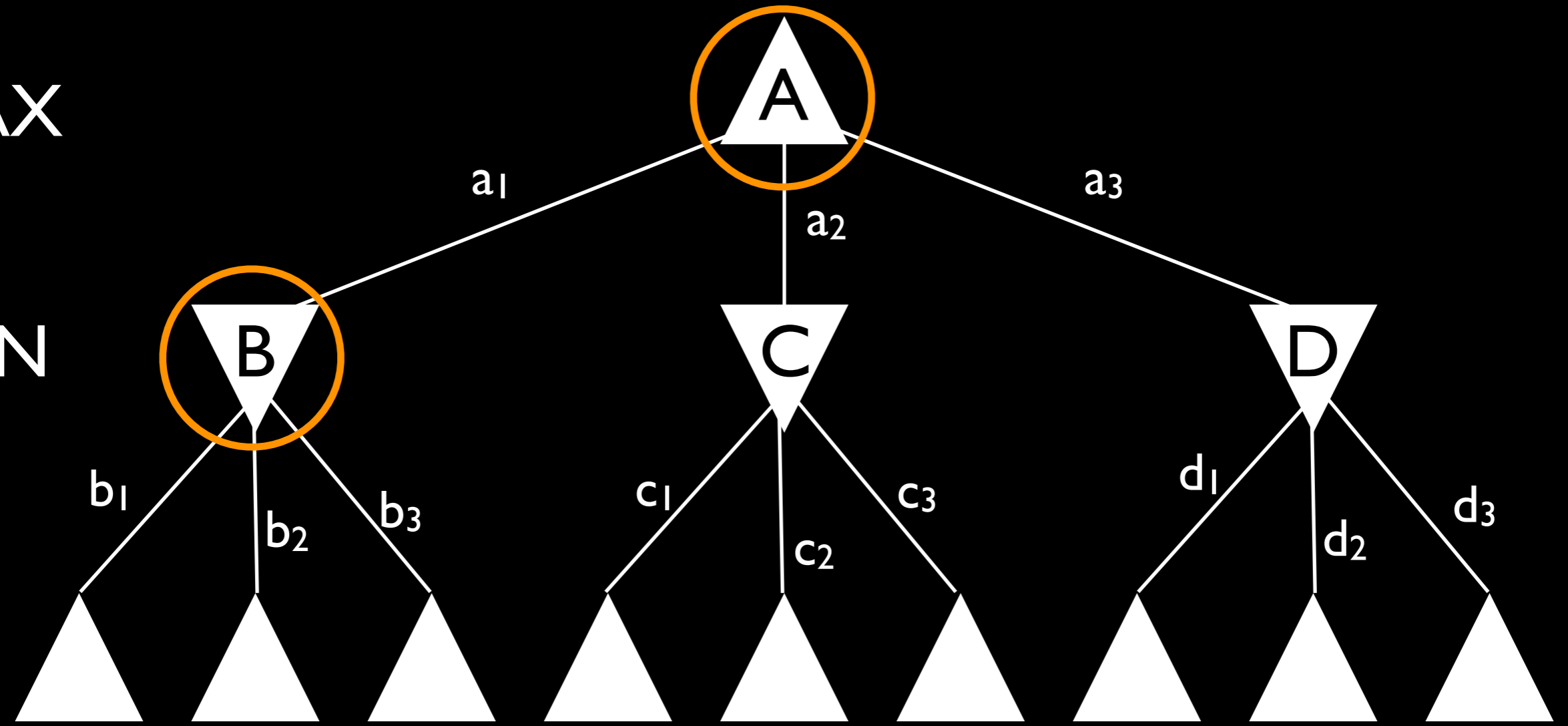
MAX

MIN



MAX

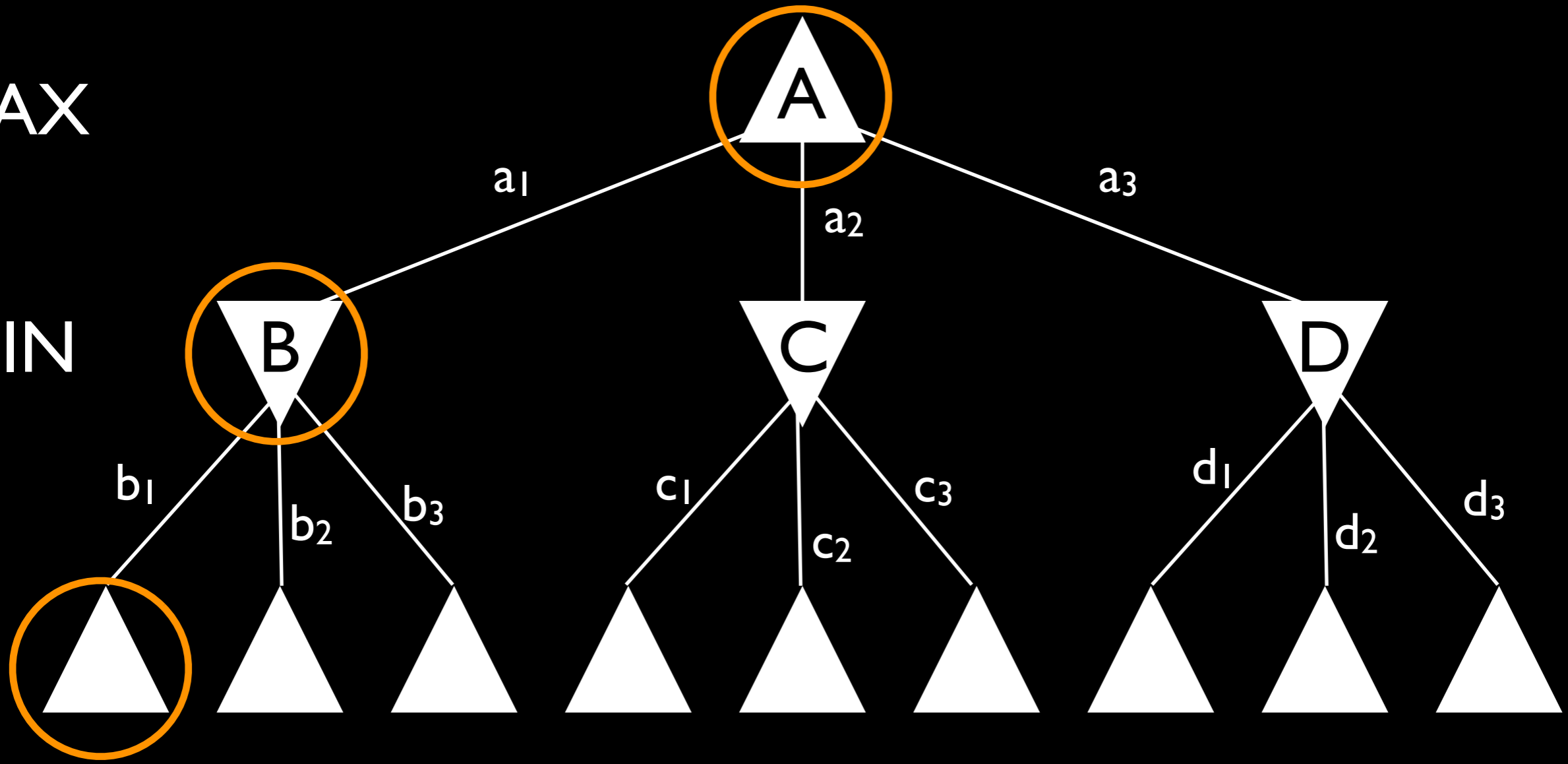
MIN





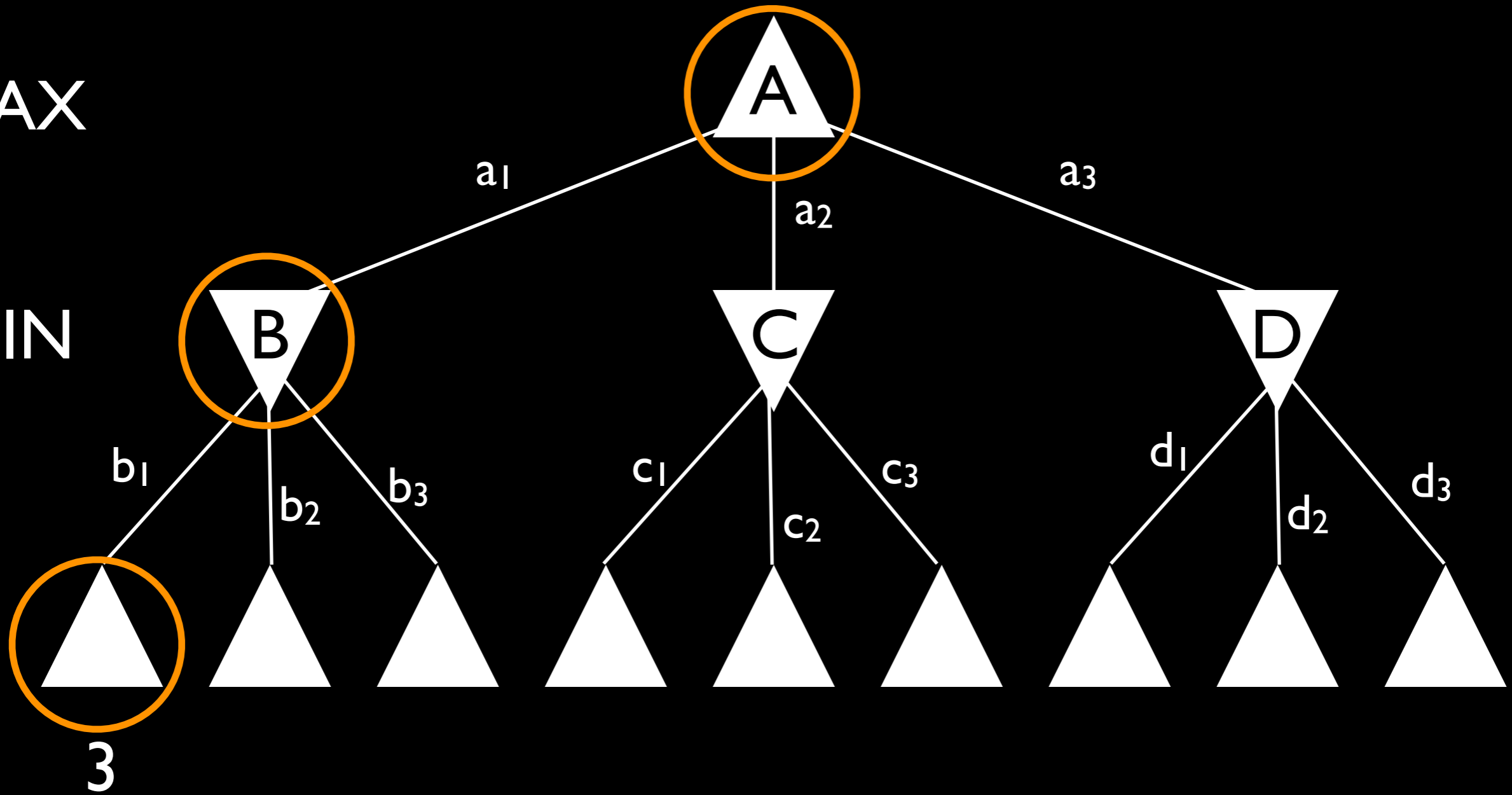
MAX

MIN



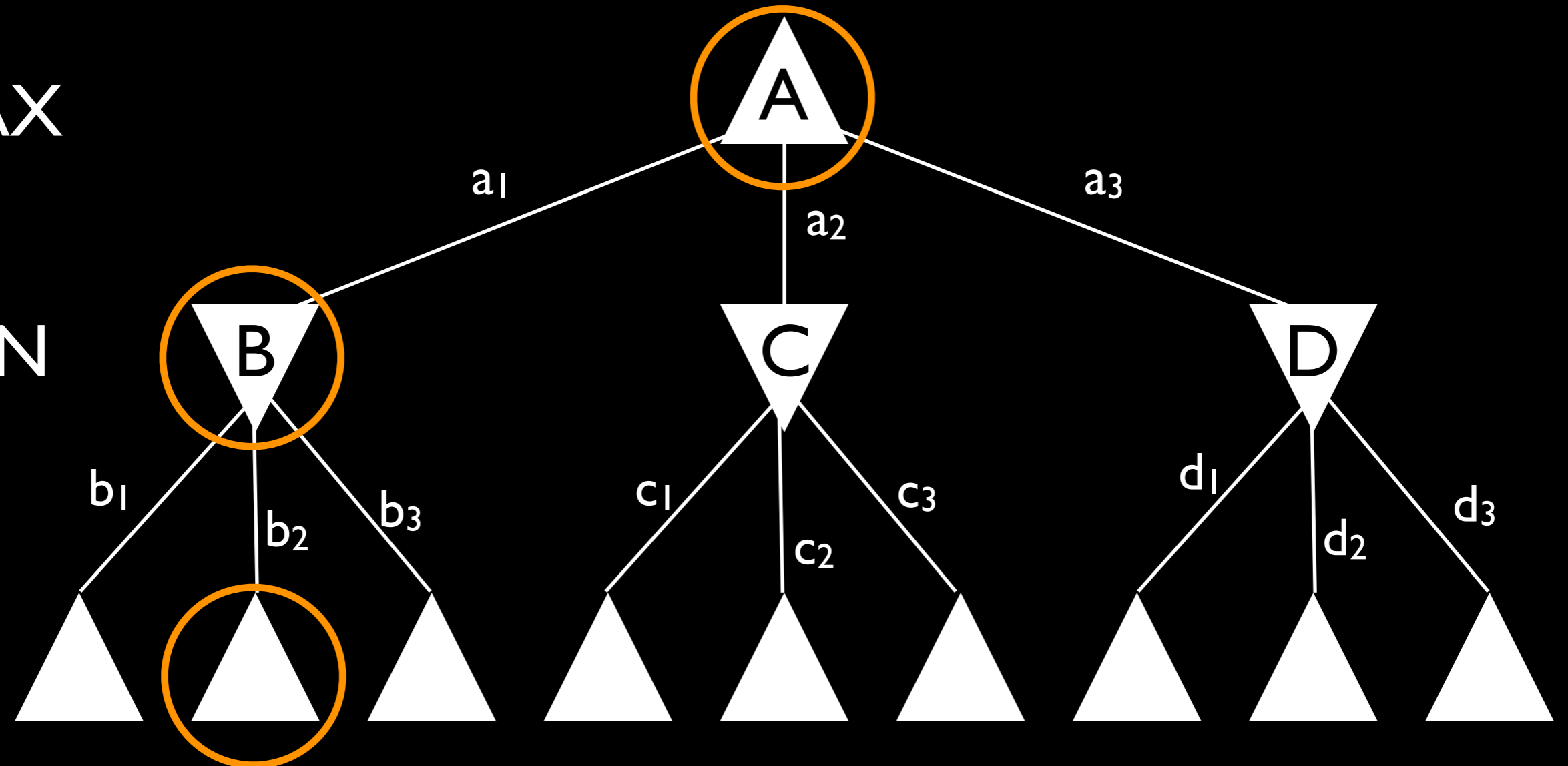
MAX

MIN



MAX

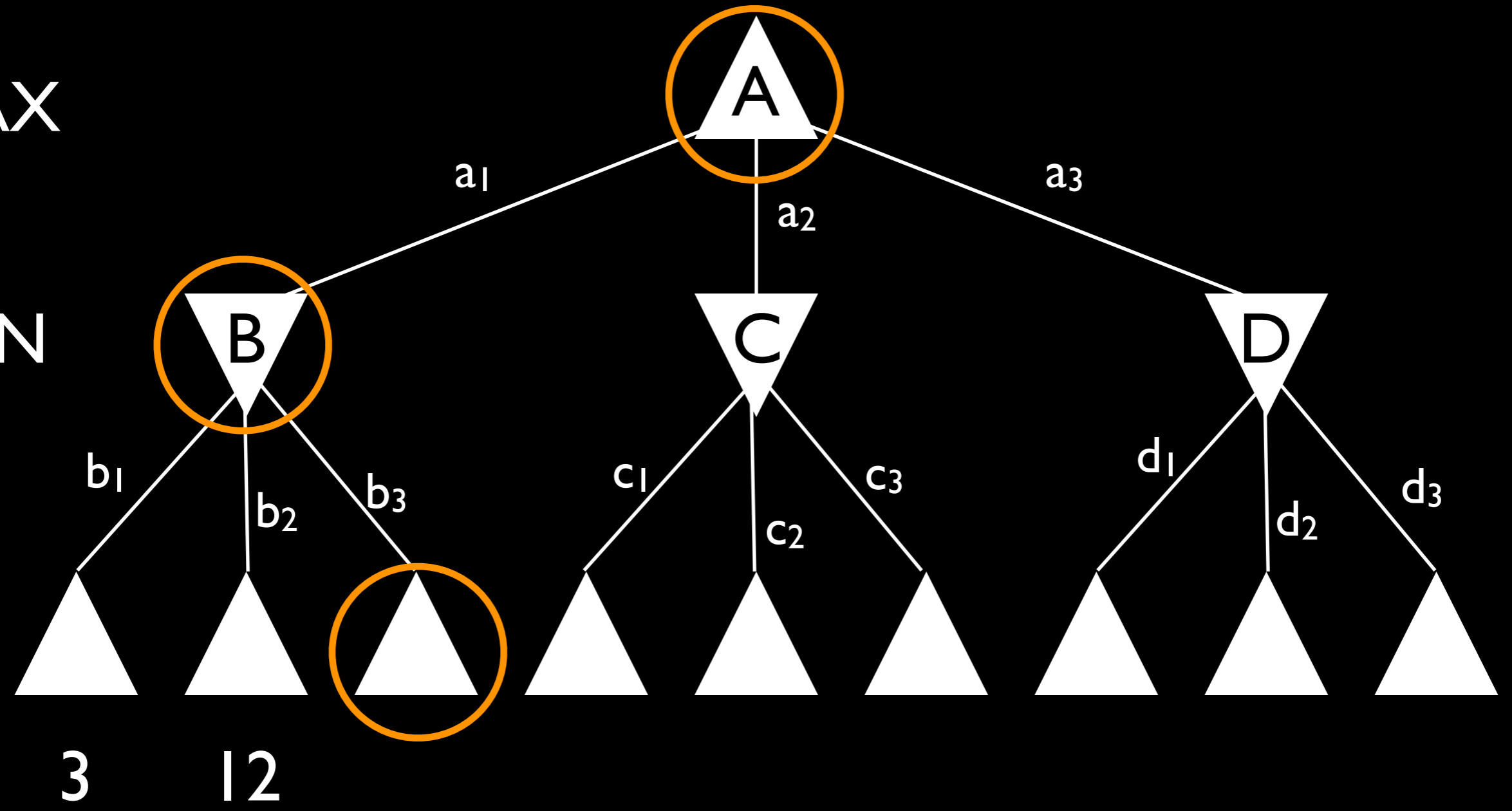
MIN



3

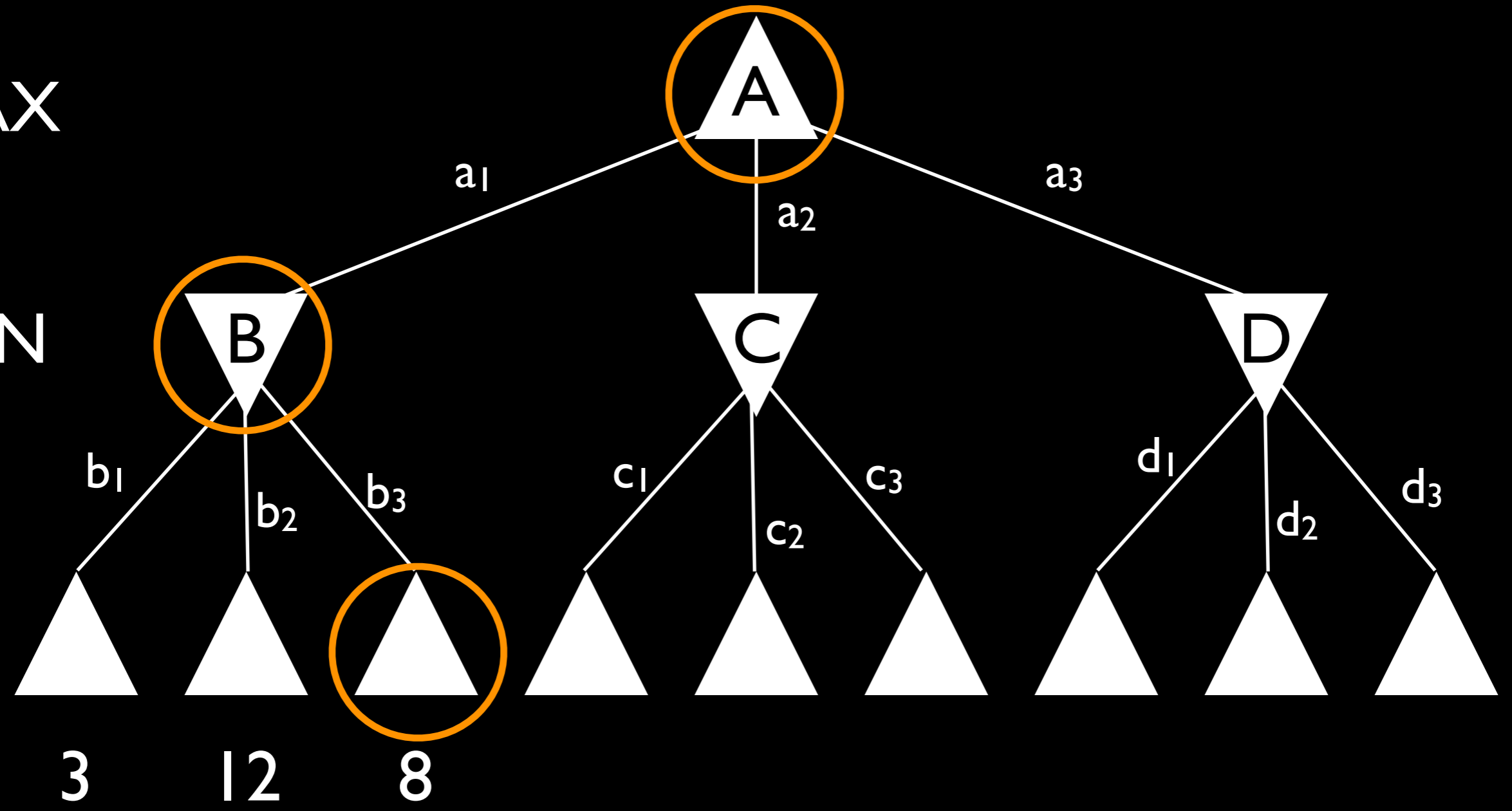
MAX

MIN



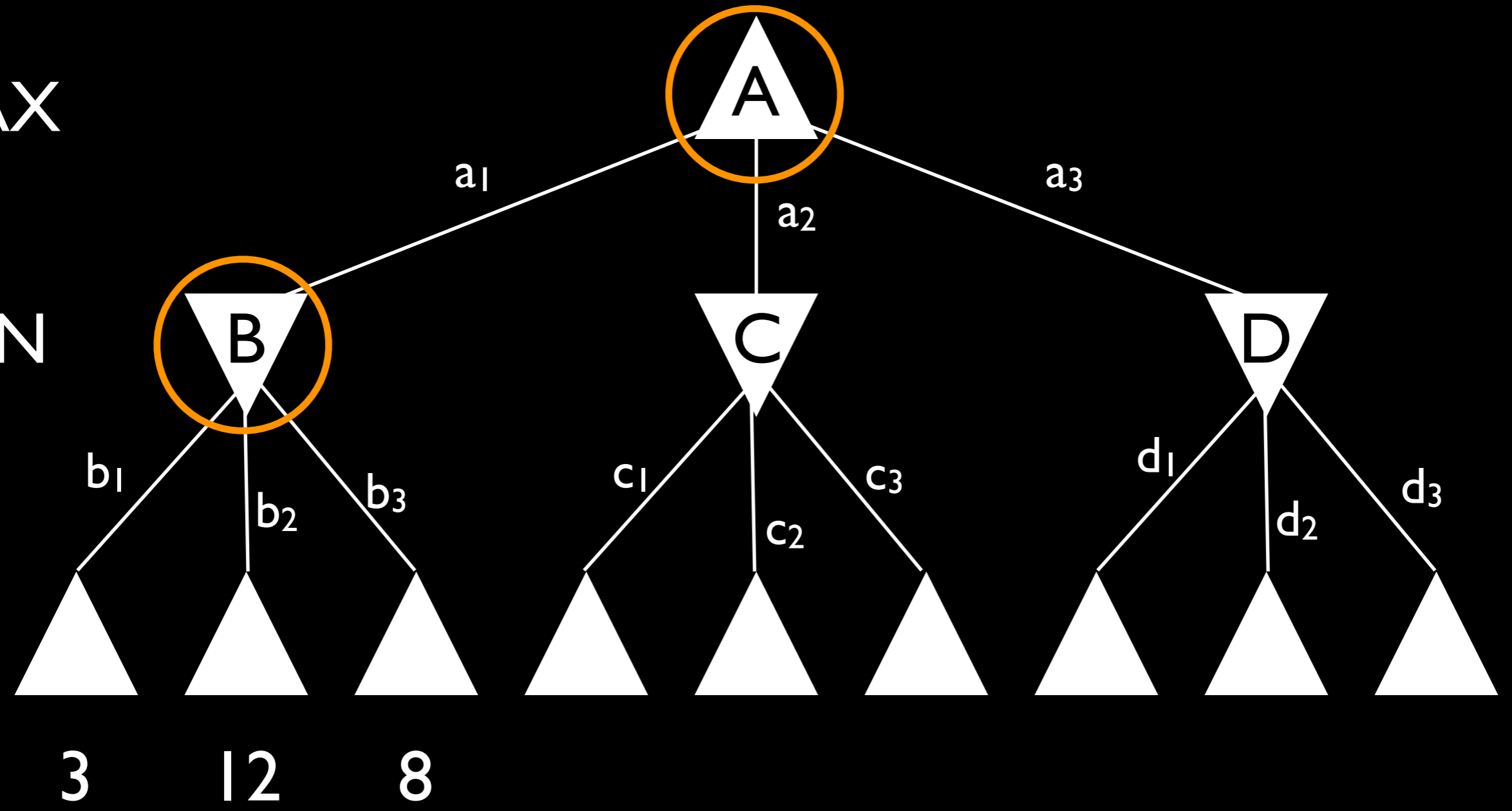
MAX

MIN



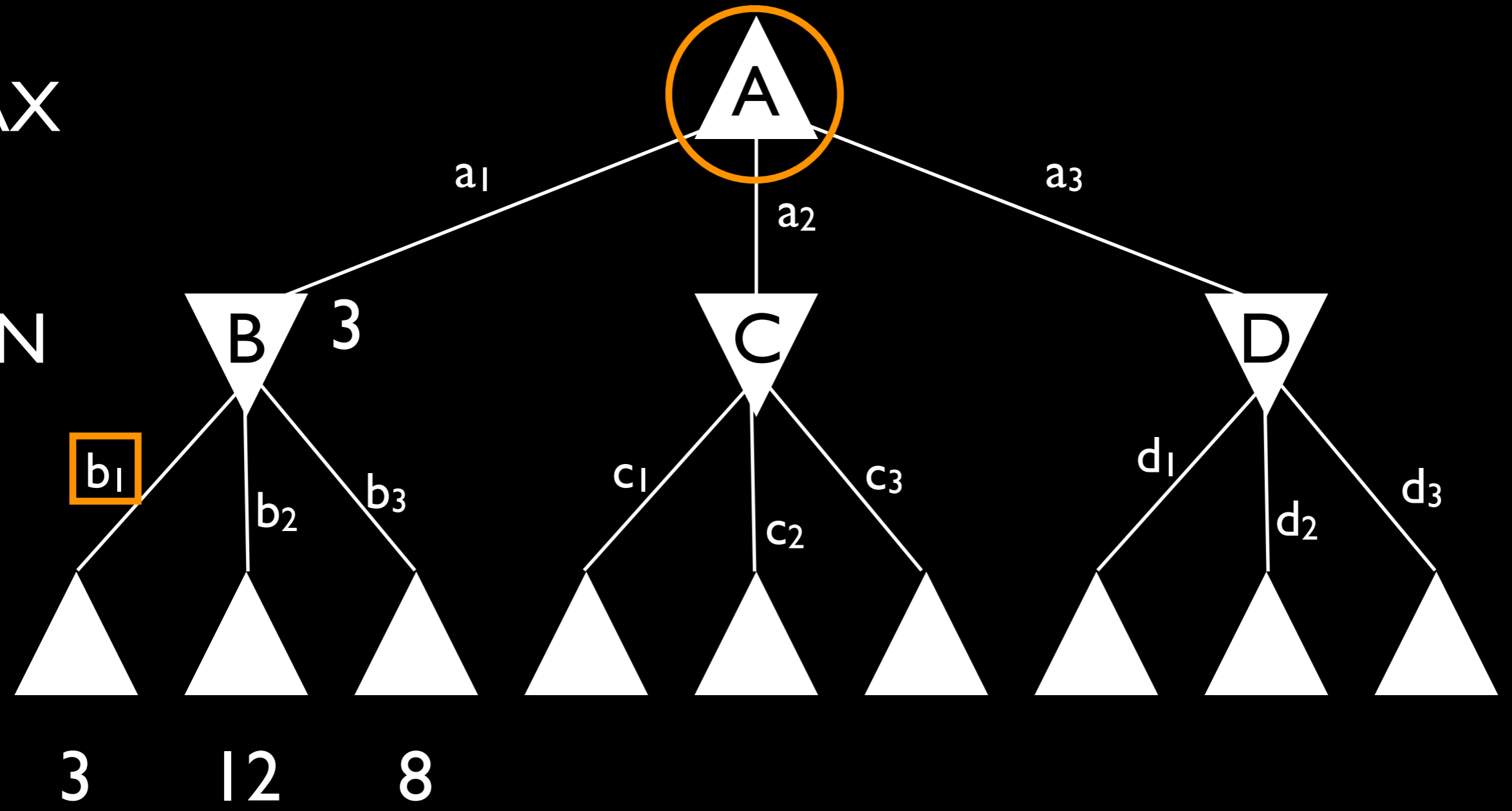
MAX

MIN



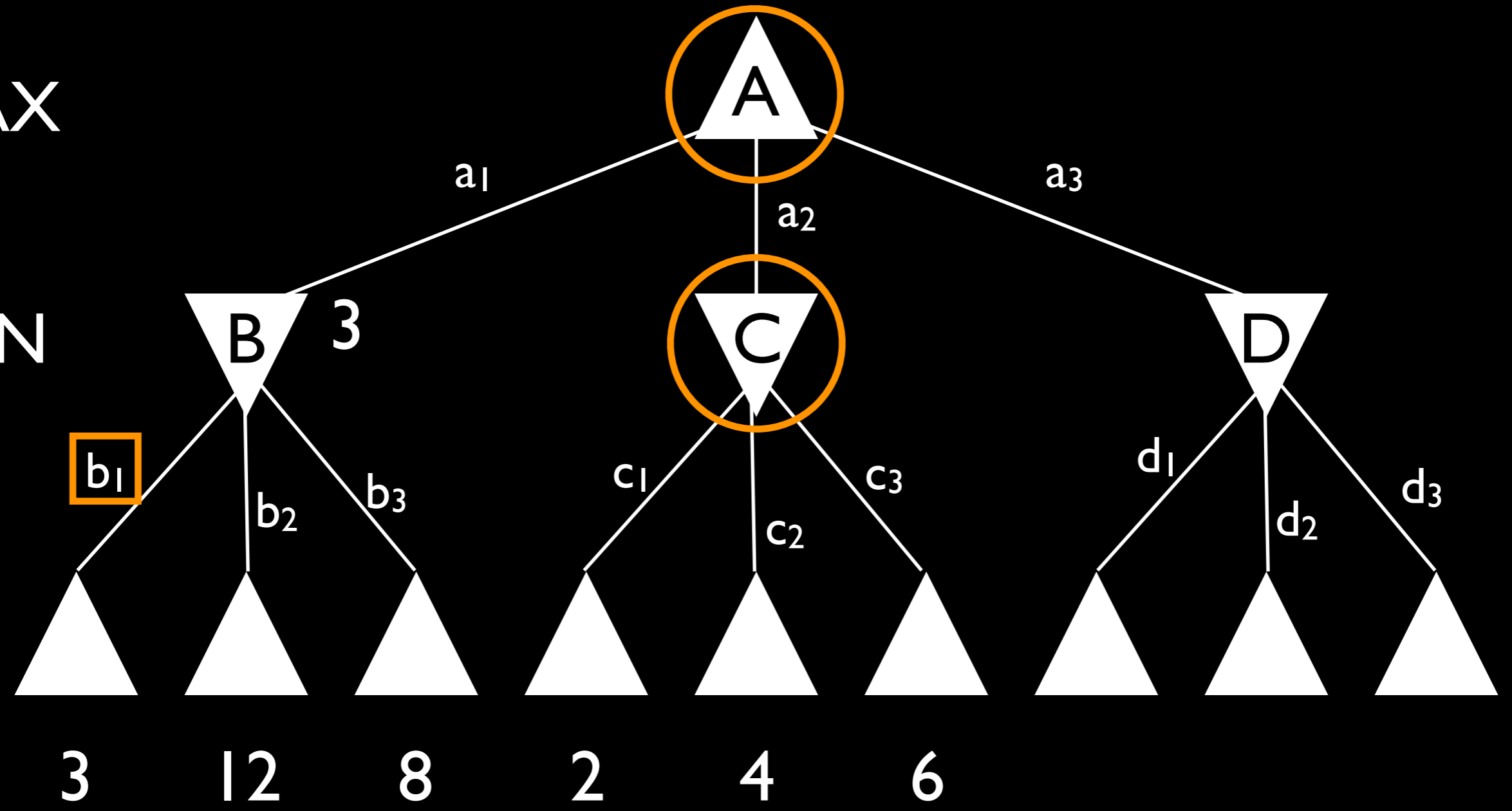
MAX

MIN



MAX

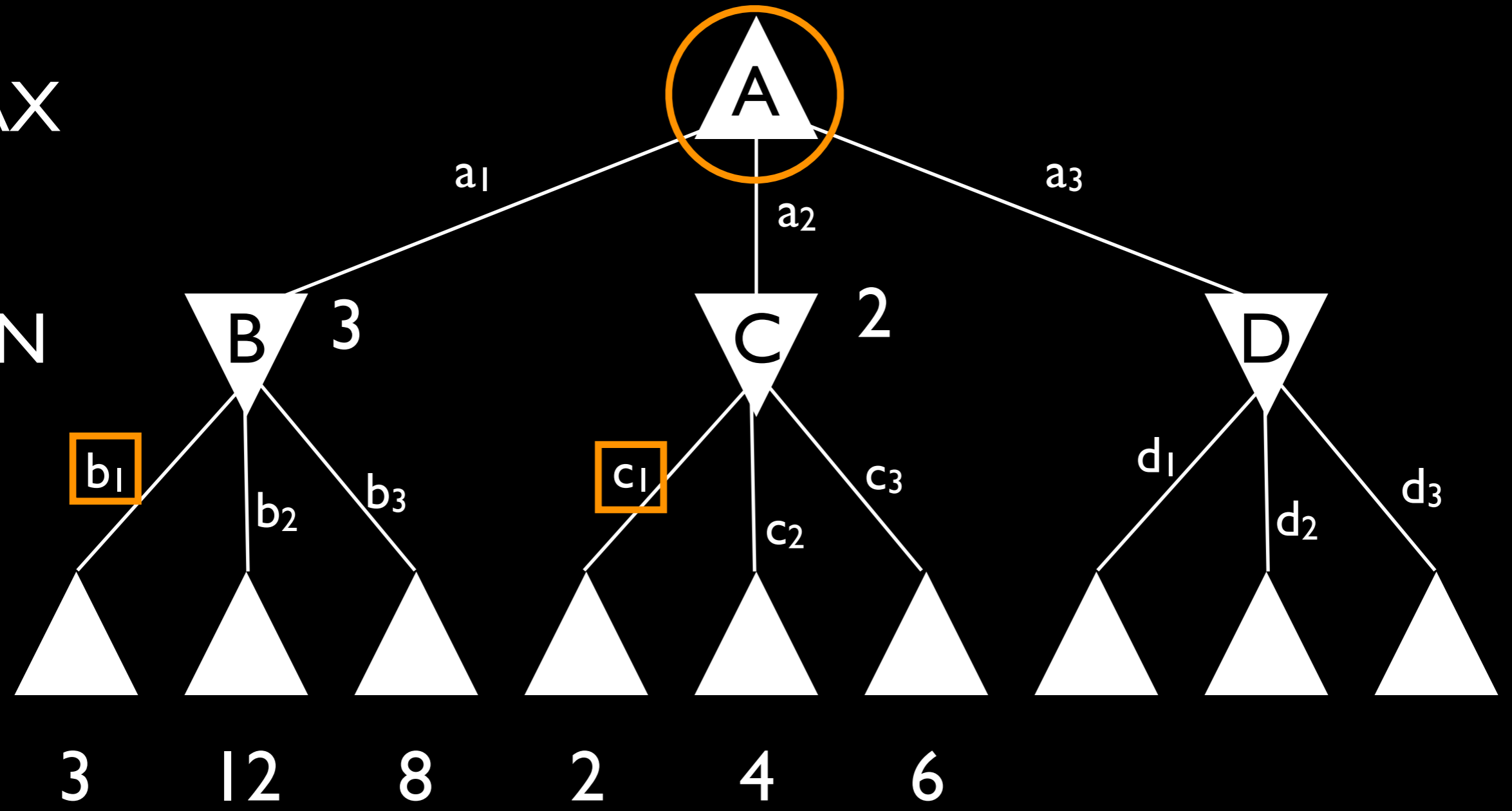
MIN





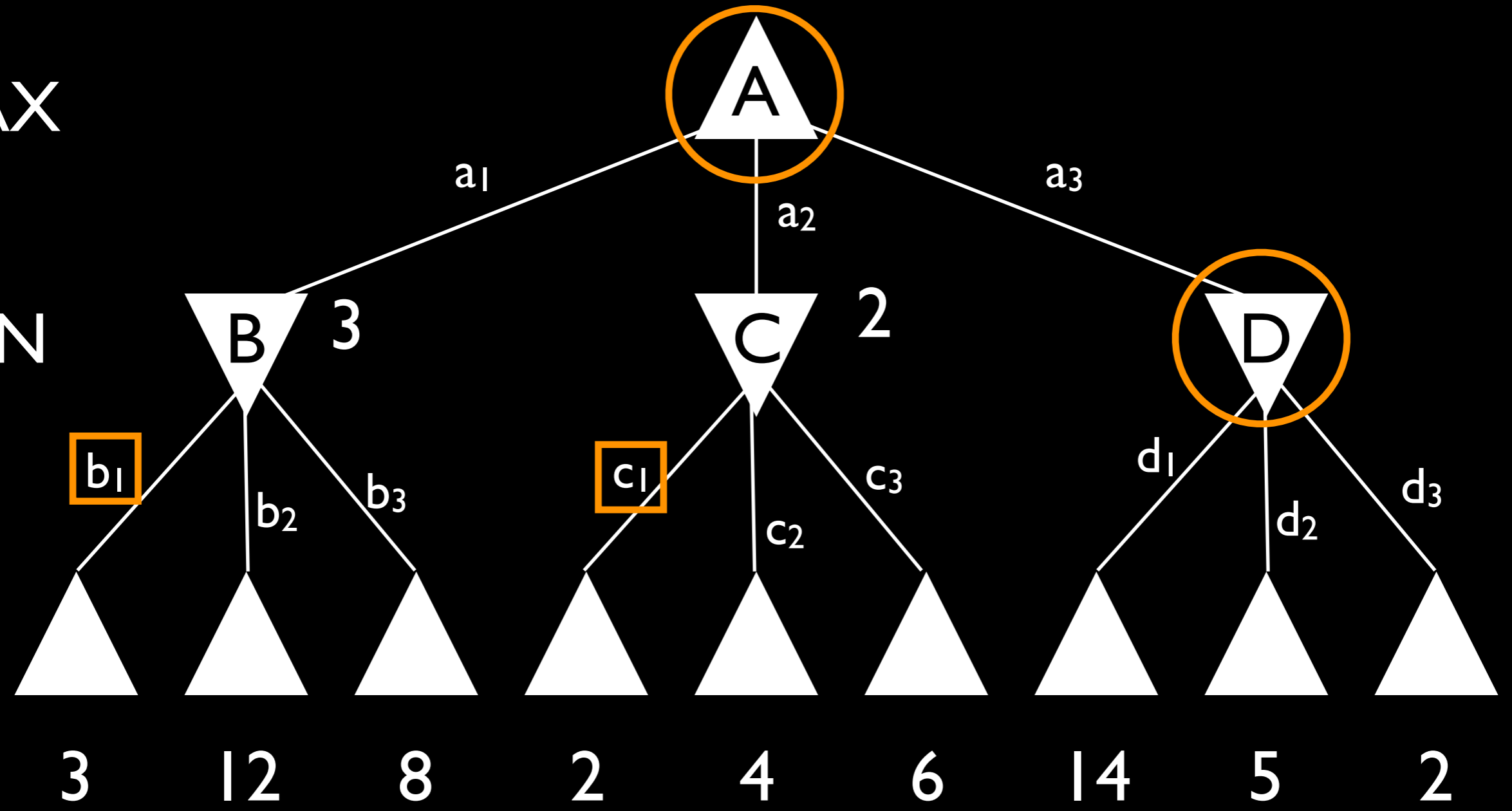
MAX

MIN



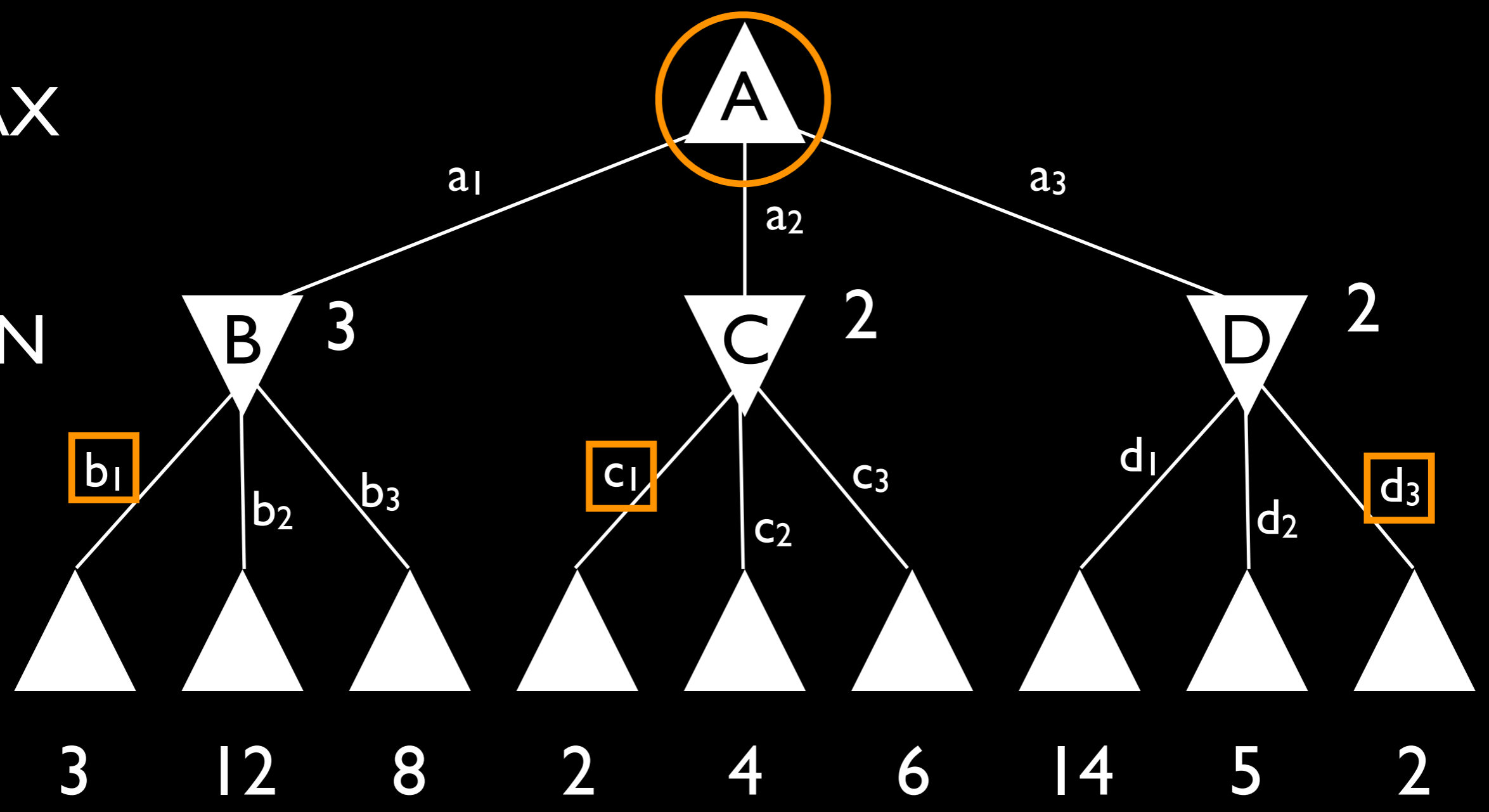
MAX

MIN



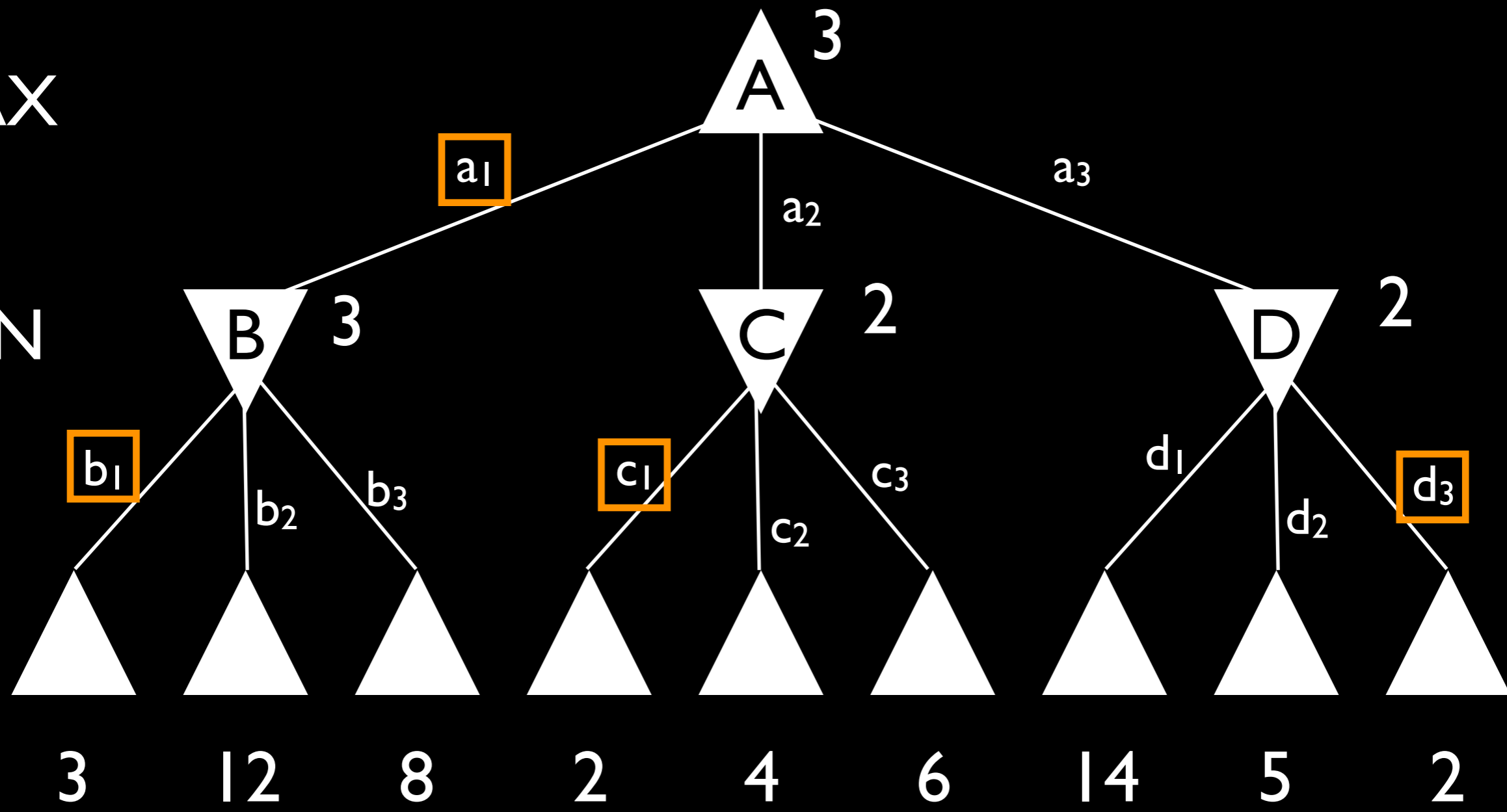
MAX

MIN



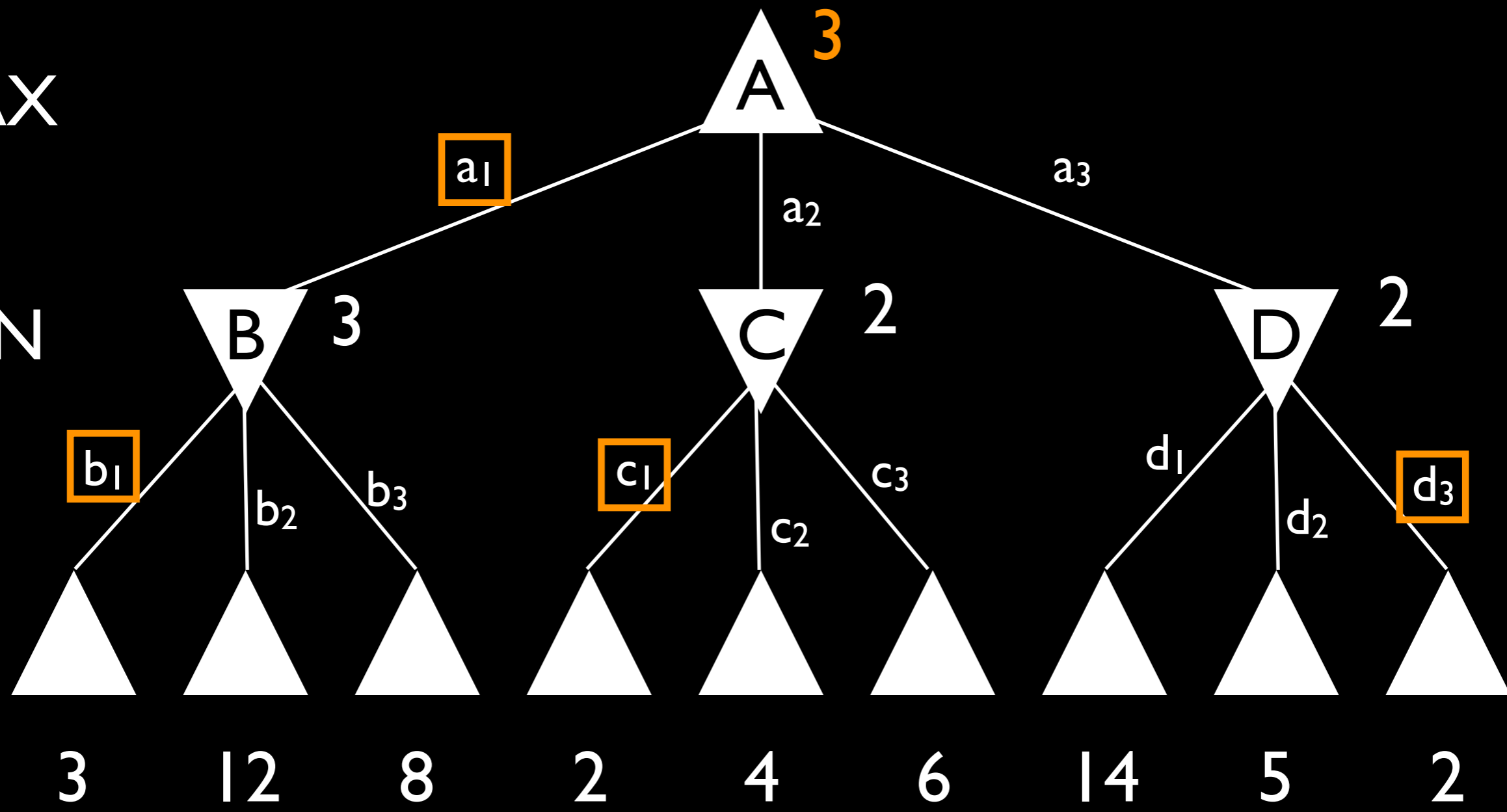
MAX

MIN



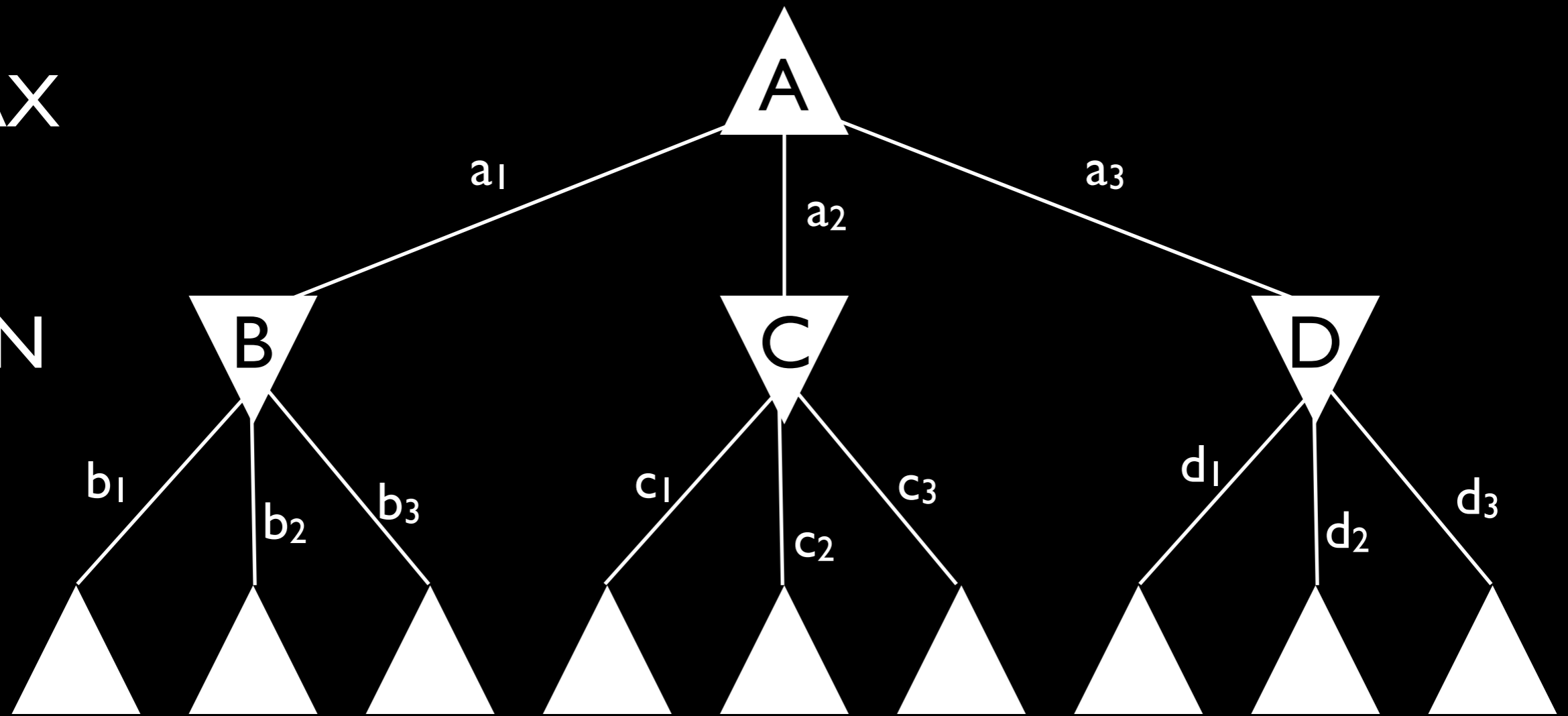
MAX

MIN



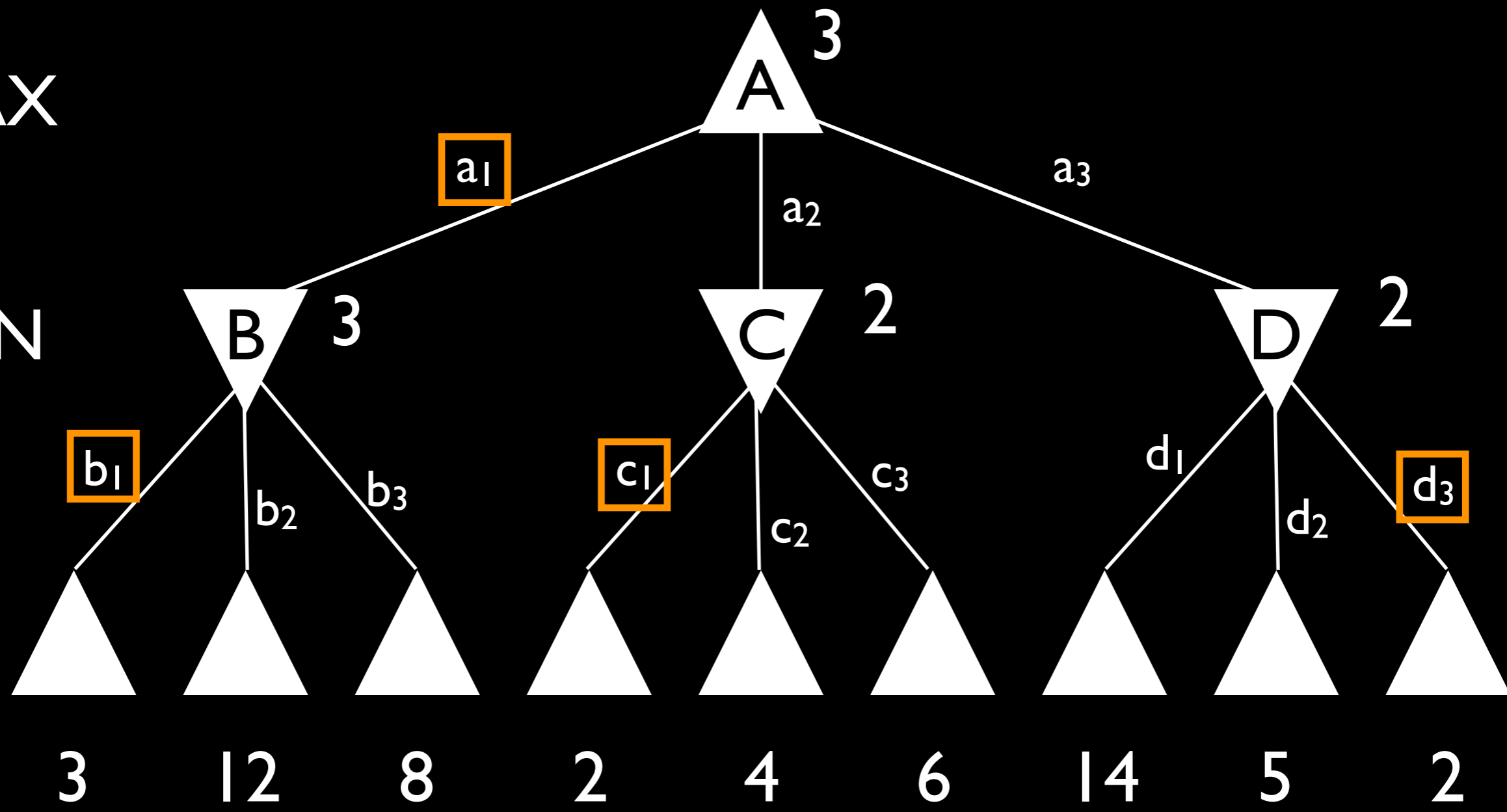
MAX

MIN




MAX

MIN



# Minimax Analysis


$$O(b^m)$$

Time Complexity


$$O(bm)$$

Space Complexity



# Minimax Summary

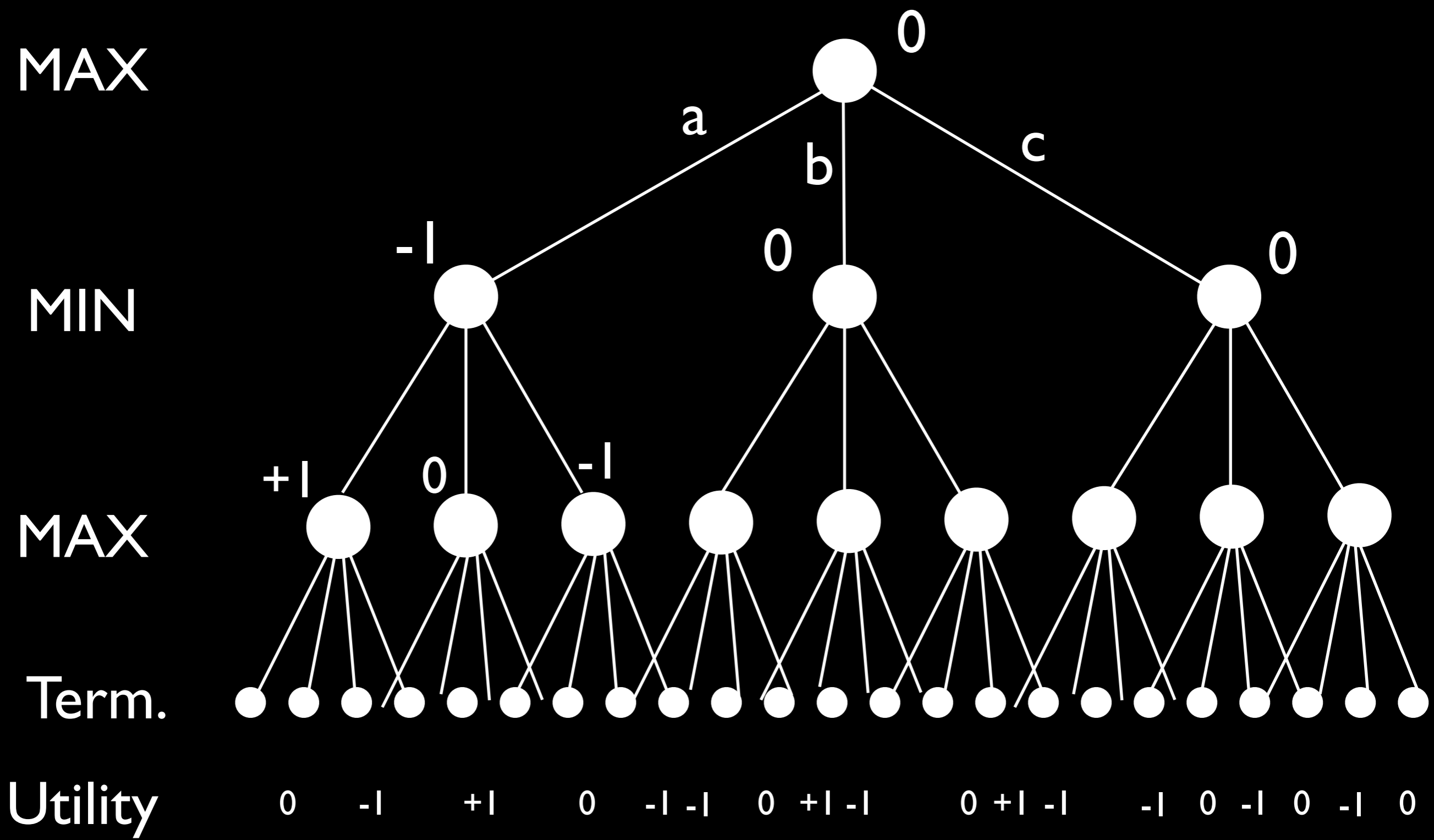
- Computes the optimal move assuming opponent also plays optimally (i.e., worst-case outcome)
- Explores game tree depth-first all the way to terminal states (end of game)
- Backs up utility values through alternating MIN and MAX (what's best for me is worst for you, and vice-versa)

# Minimax Code

AIMA 5.2.1 and Figure 5.3 (page 166)

Code template will appear on course page





# A Problem for Minimax

- You can't search all the way to the terminal nodes
- You can't evaluate a node (state) unless you're at a terminal node
- Utility function is defined on terminal states

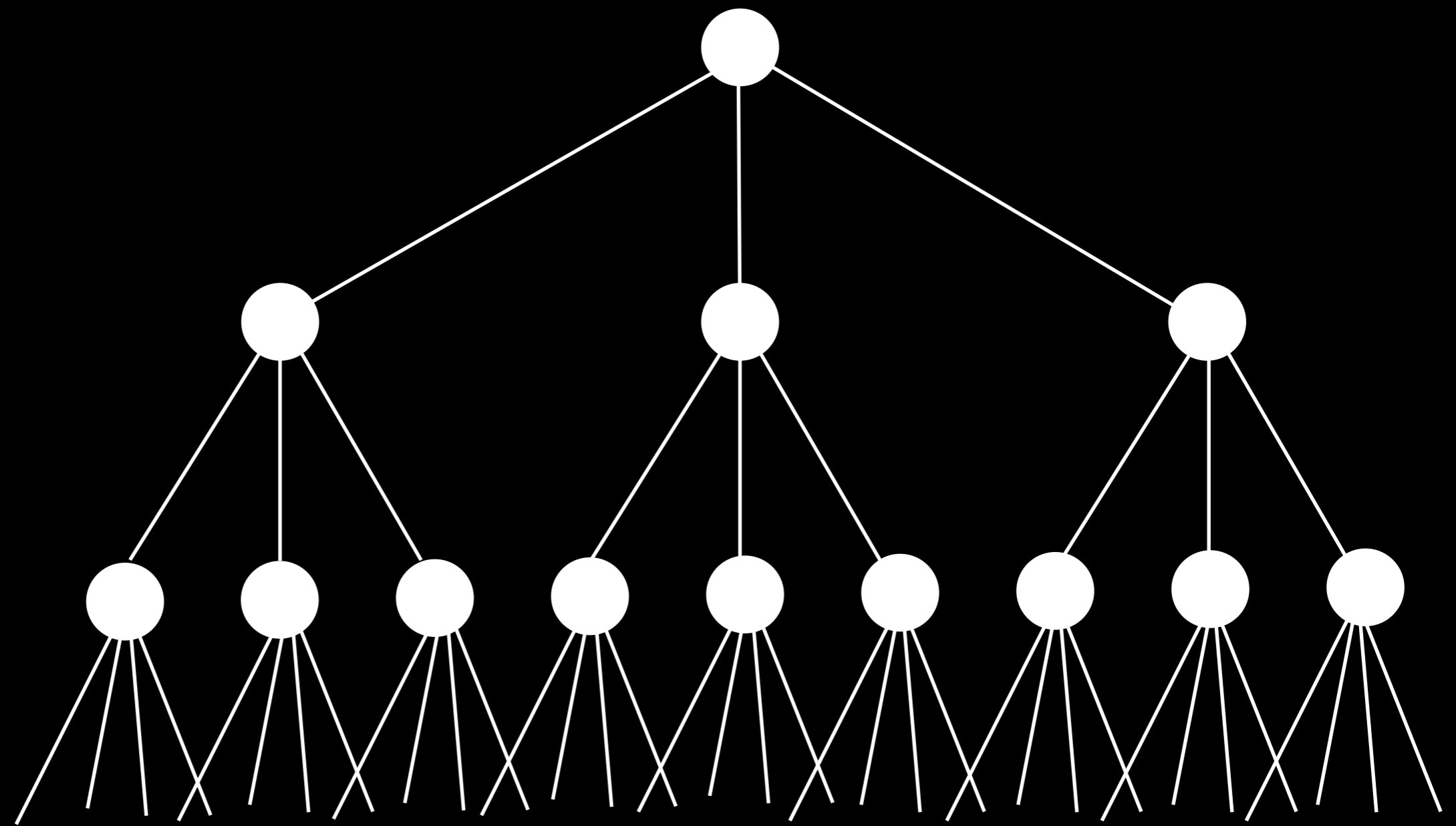
# Imperfect Real-Time Decisions

MAX

MIN

MAX

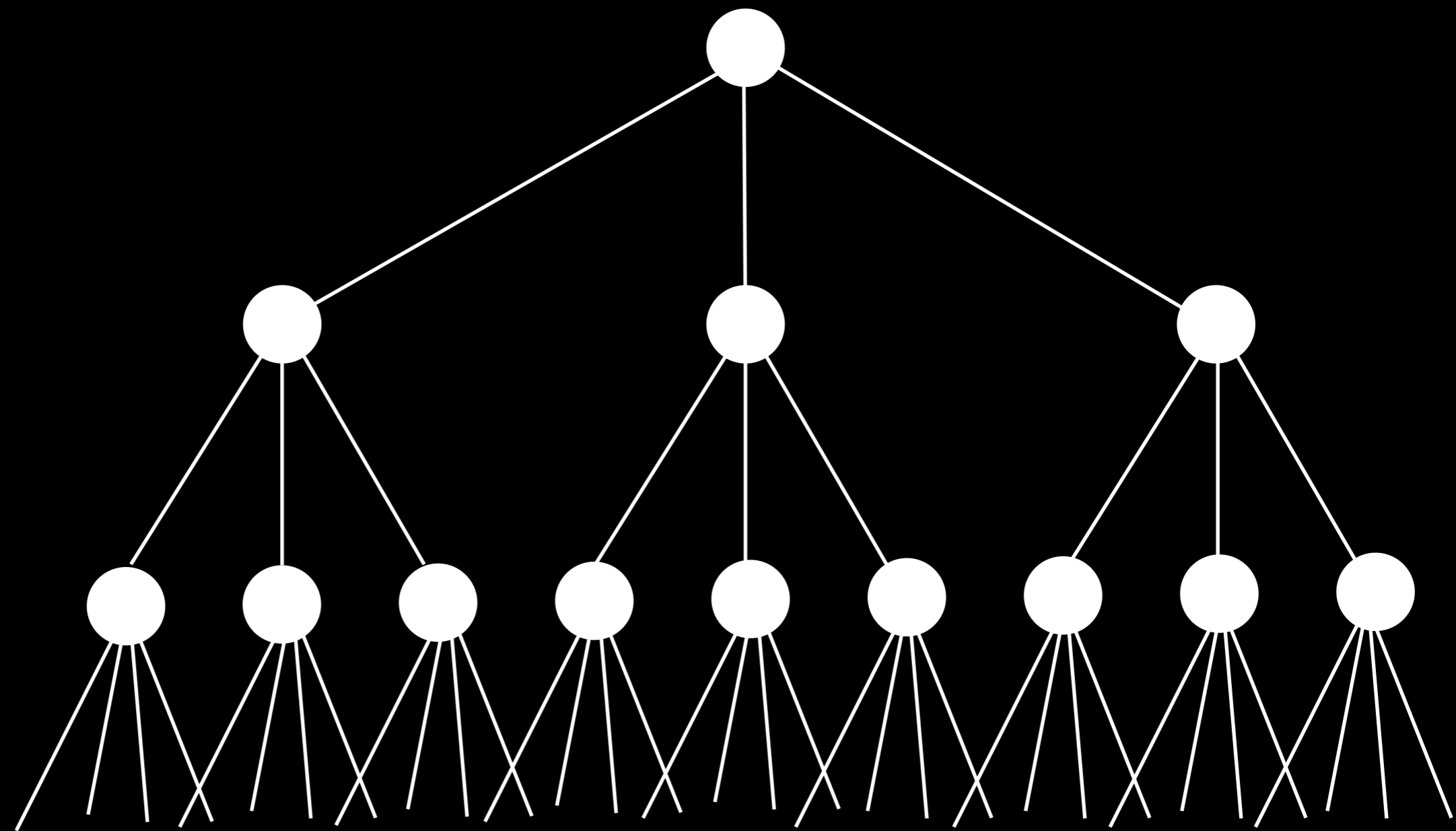
Term.



MAX

MIN

MAX



Term.

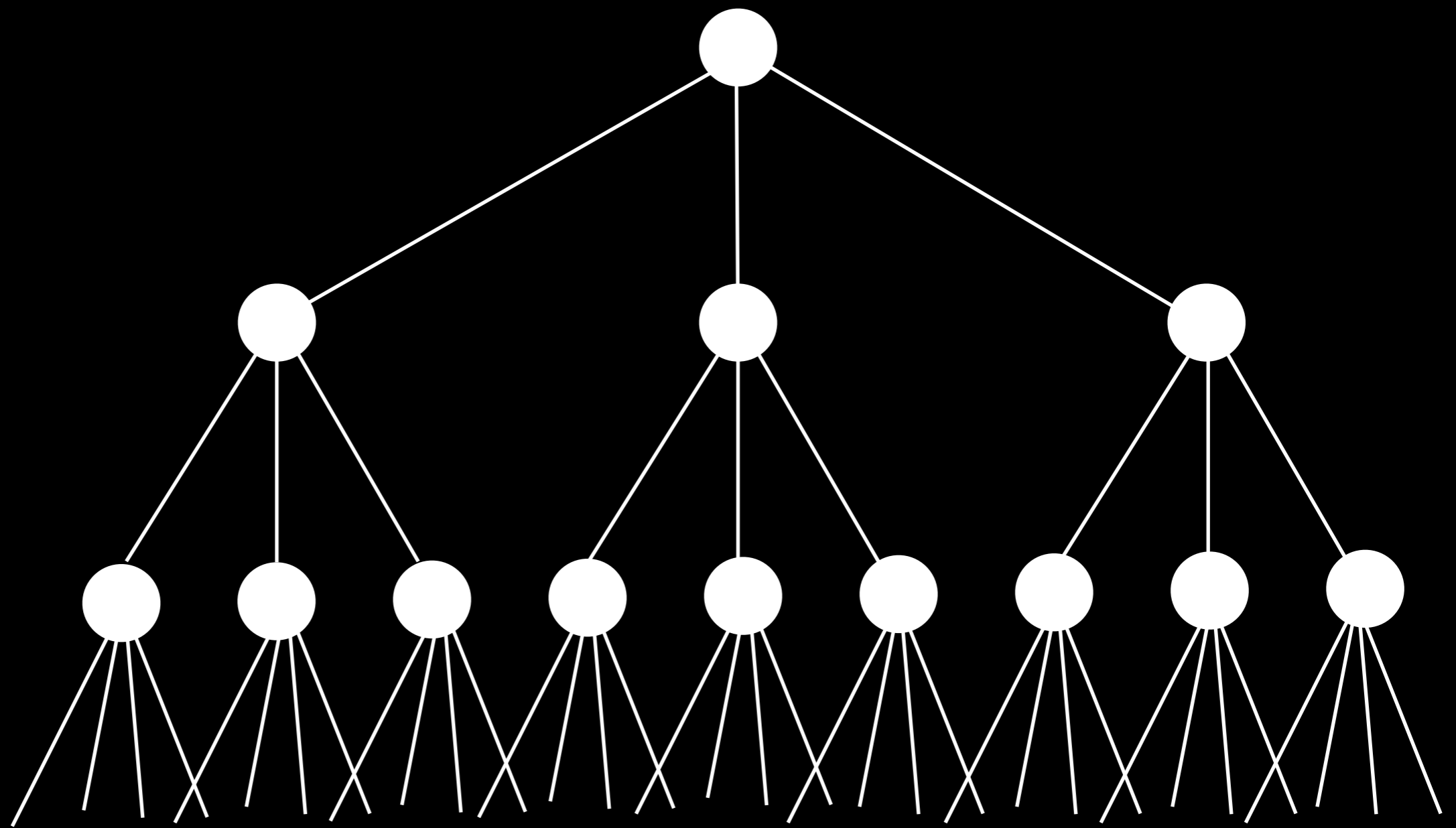




MAX

MIN

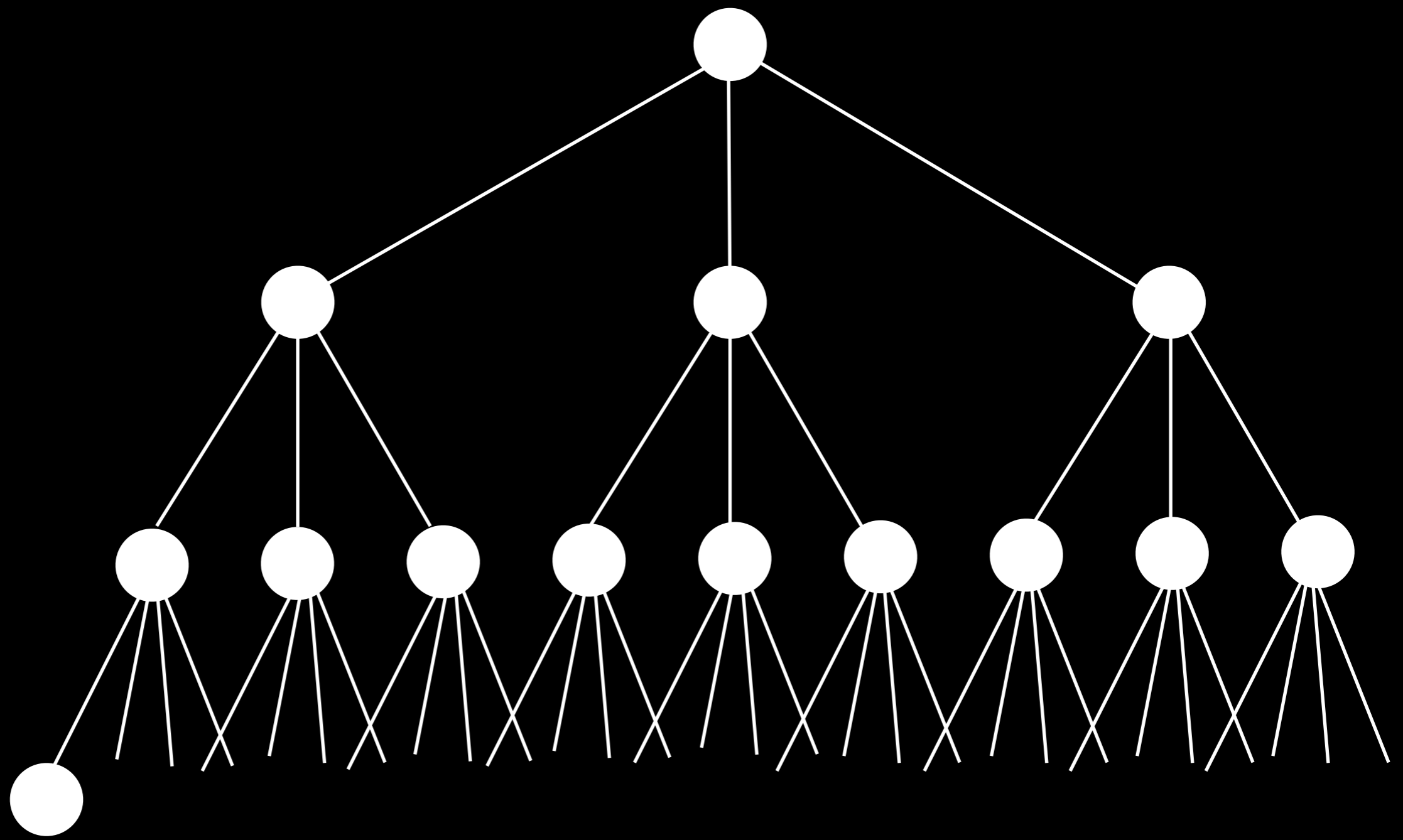
MAX



MAX

MIN

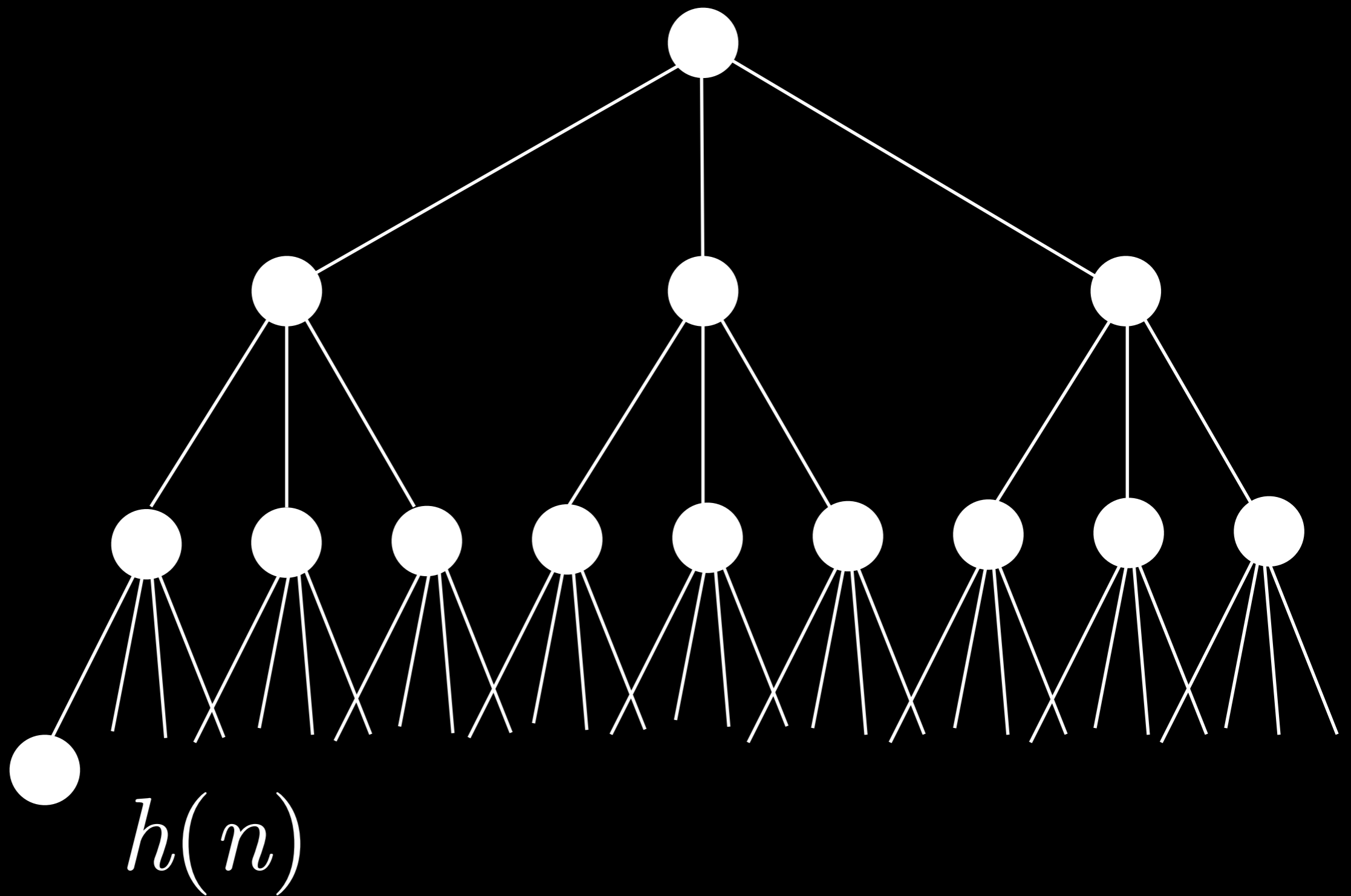
MAX



MAX

MIN

MAX



# Minimax Algorithm

$$\text{MINIMAX}(s) = \begin{cases} \text{UTILITY}(s) & \text{if } \text{TERMINAL-TEST}(s) \\ \max_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MAX} \\ \min_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if } \text{PLAYER}(s) = \text{MIN} \end{cases}$$

# Heuristic Minimax

H-MINIMAX( $s$ ) =

$$\begin{cases} h(s) & \text{if CUTOFF-TEST}(s) \\ \max_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if PLAYER}(s) = \text{MAX} \\ \min_{a \in \text{ACTIONS}(s)} \text{MINIMAX}(\text{RESULT}(s, a)) & \text{if PLAYER}(s) = \text{MIN} \end{cases}$$

# H-Minimax: Cutoff

- When to cutoff search?
  - Time
  - Depth
  - “Quiescence”
- Can adjust dynamically
- AIMA 5.4.2

# H-Minimax: Heuristic

- How to evaluate non-terminal state?
  - AIMA 5.4.1
  - Chess example
    - Material value
    - Weighted sum
    - Strategic considerations

# Heuristic Minimax

- Cutoff search before reaching terminal nodes (time, depth, “quiescence”)
- Use heuristic evaluation function to estimate state utility
- Backs up utility values through alternating MIN and MAX (what’s best for me is worst for you, and vice-versa)



# Adversarial Search: MINIMAX

## MINIMAX

- Searches to terminal nodes
- Uses utility function

## H-MINIMAX

- Cuts off search before terminals
- Uses heuristic function

Backs up utility values through alternating  
MIN and MAX (zero-sum game)

# For Next Class

- Read rest of AIMA Chapter 5
- Complete Homework 01 – solutions will be given out
- Project 1 launch: Othello Tournament
- Lecture 06: Games of chance and partial information