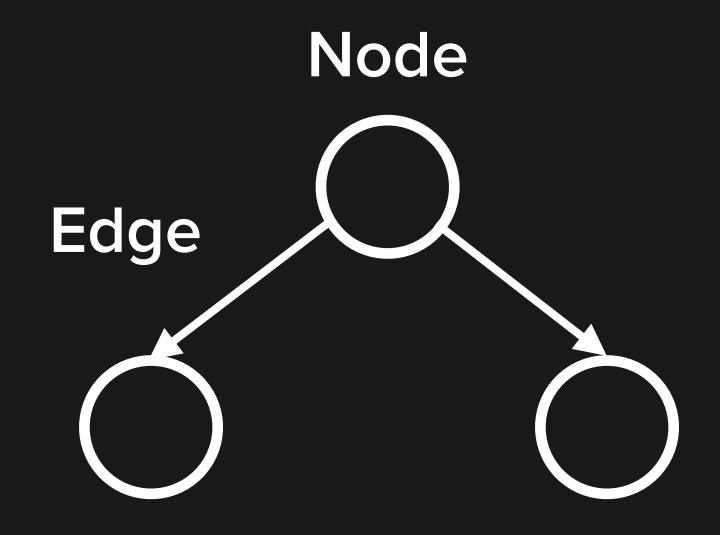
## TRES

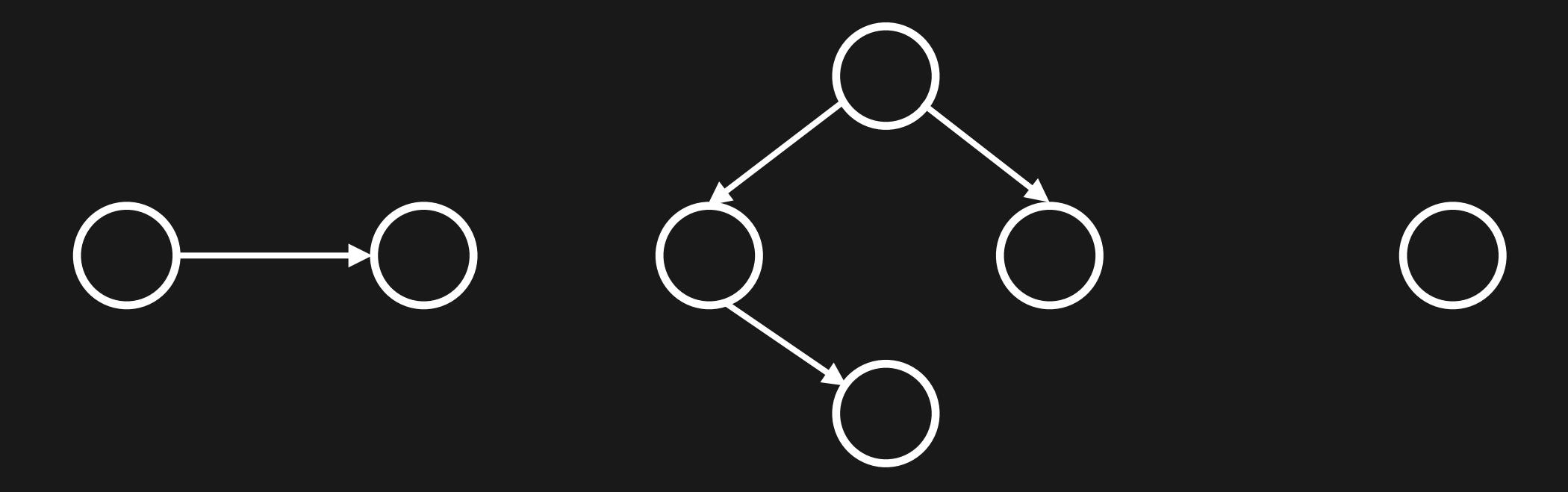
So log-arithmic. (Get it!?)

#### TREE

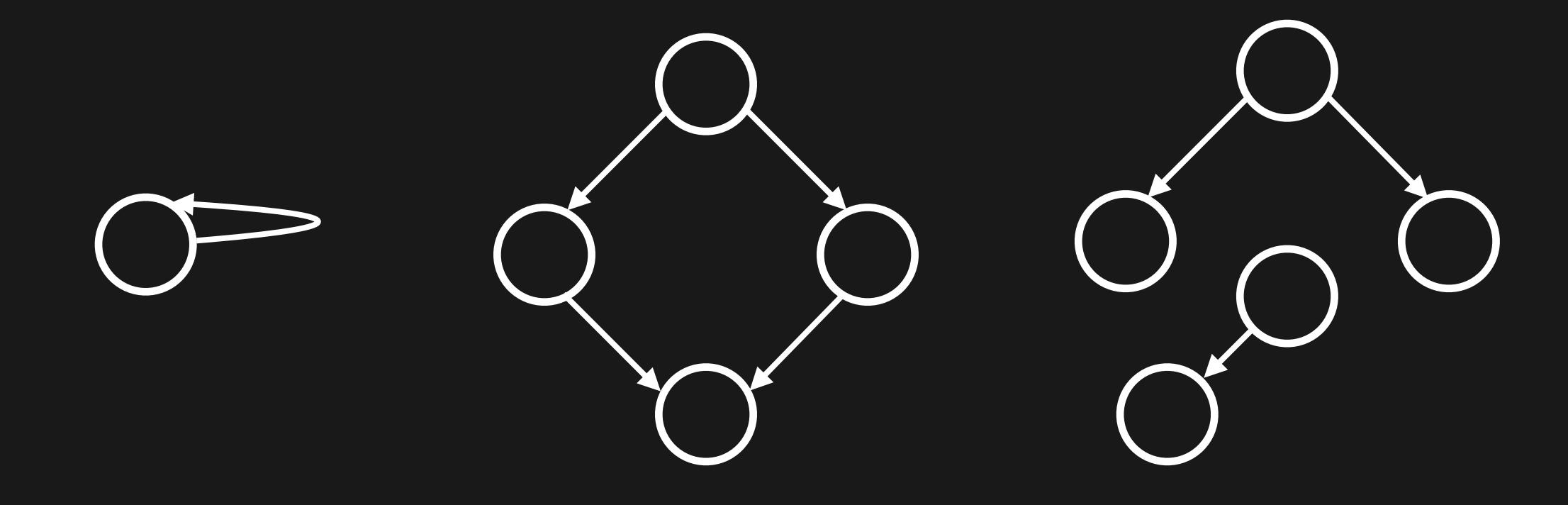
Set of nodes and edges (references to child nodes) without any cycles



# TREES

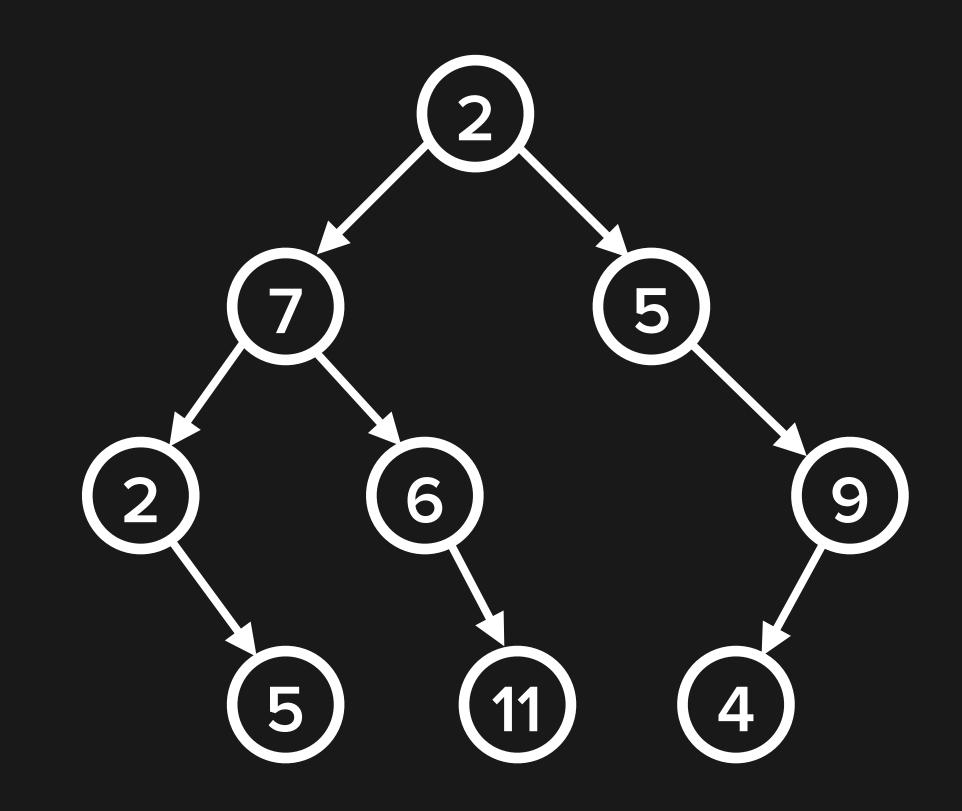


# NOT TREES



### BINARY TREE

Tree in which each node has at most two children



Size 9 Height 3

root unique topmost node of tree that can reach all other nodes

ancestor node reachable from child to parent, grandparent, etc.

parent → child node above → node below

leaf / external node node with no children

descendant node reachable from parent to child, grandchild, etc.

branch / internal node node with at least one child size (tree) number of nodes in the tree

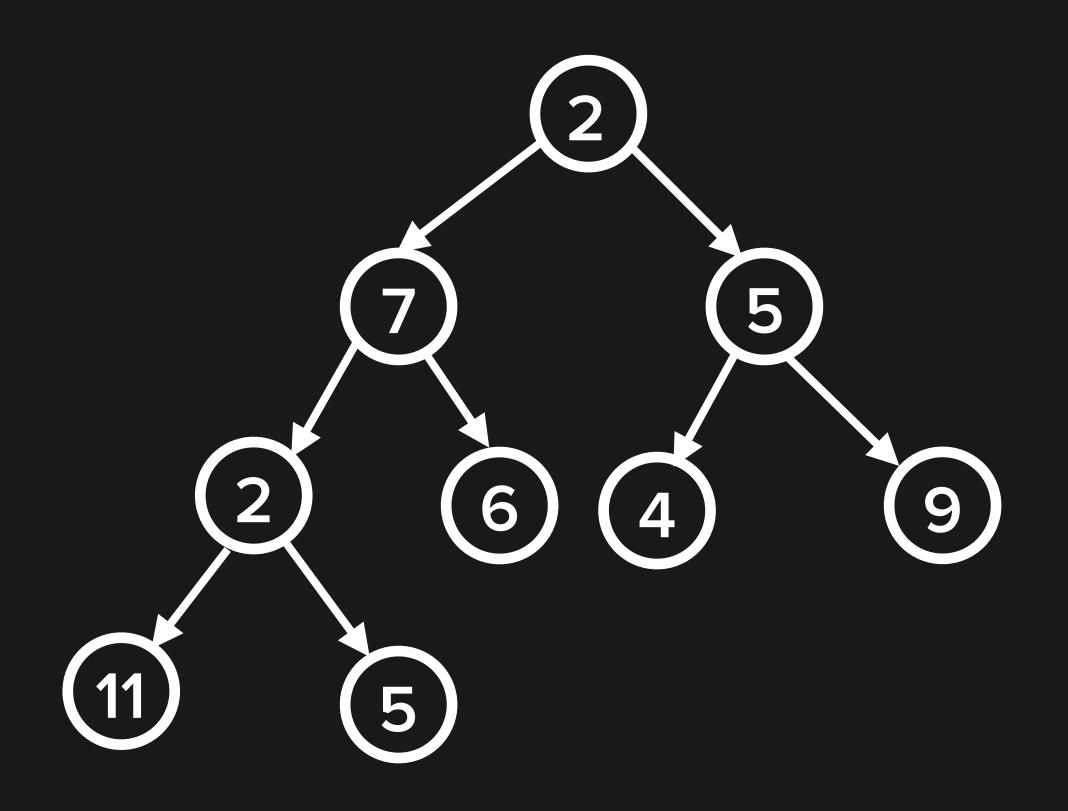
height (tree) number of edges on *longest* downward path from *root* to leaf

height (node) number of edges on *longest* downward path from *node* to leaf depth (node)
number of edges between the
node and the root

level (node)
1 + number of edges between
the node and the root

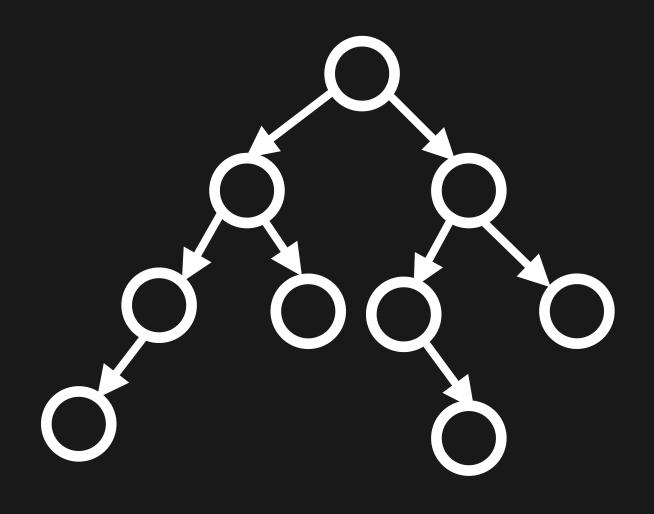
#### COMPLETE TREE

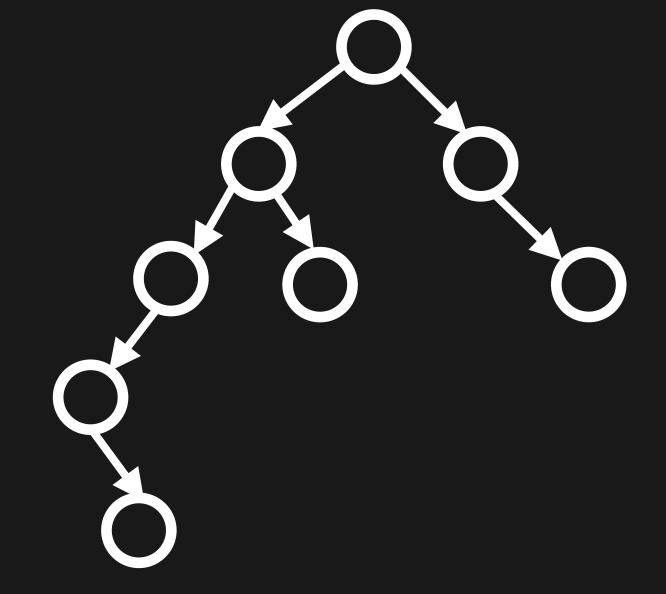
Every level except possibly last is completely filled and nodes are as far left as possible



## BALANCED TREE

All leaves are at minimum possible depth





Balanced

Unbalanced

### BINARY SEARCH TREE

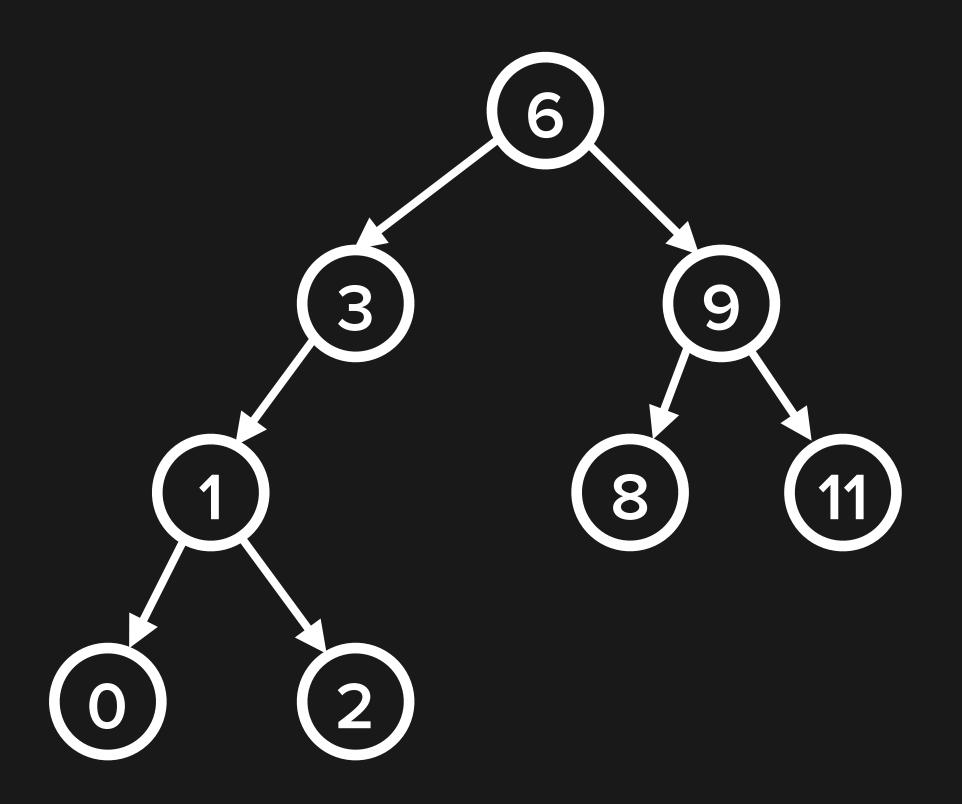
Always sorted

For each node

Left children are smaller

Right children are larger

No duplicate keys (usually)



#### WHY USE A BST?

Fast search, insertion, deletion - especially when balanced

Sort as you go instead of all at once

Fairly simple implementation for good performance

#### WHY USE A BST?

Only allocates memory as it's needed

Doesn't have to reallocate memory to grow (like a hash table)

#### ANOTE ON LOG N

When discussing time and space complexity of algorithms, log n usually means log<sub>2</sub>n

#### BINARY LOGARITHM

$$log_2n = x \leftrightarrow 2^x = n$$

the power by which 2 must be raised by to obtain n

$$log_2 16 = 4$$
  $log_2 32 = 5$   $log_2 143 = 7.15987$ 

#### WHY IS LOG N GOOD?

Imagine a binary search tree with 232 nodes

2<sup>32</sup> is 4,294,967,296 (that's over 4 billion!)

When searching, we only have to visit a maximum of 32 nodes to find the node containing the data we're looking for (assuming the tree is perfectly balanced)

#### SEARCH

```
# Call initially with node == root node
def find_recursive(key, node):
    if node is None or node.key == key:
        return node
    elif key < node.key:
        return find_recursive(key, node.left)
    else:
        return find_recursive(key, node.right)</pre>
```

#### INSERTION

Same as search except once you find a node without a child on the next side you're traversing, add it there.

### DELETION

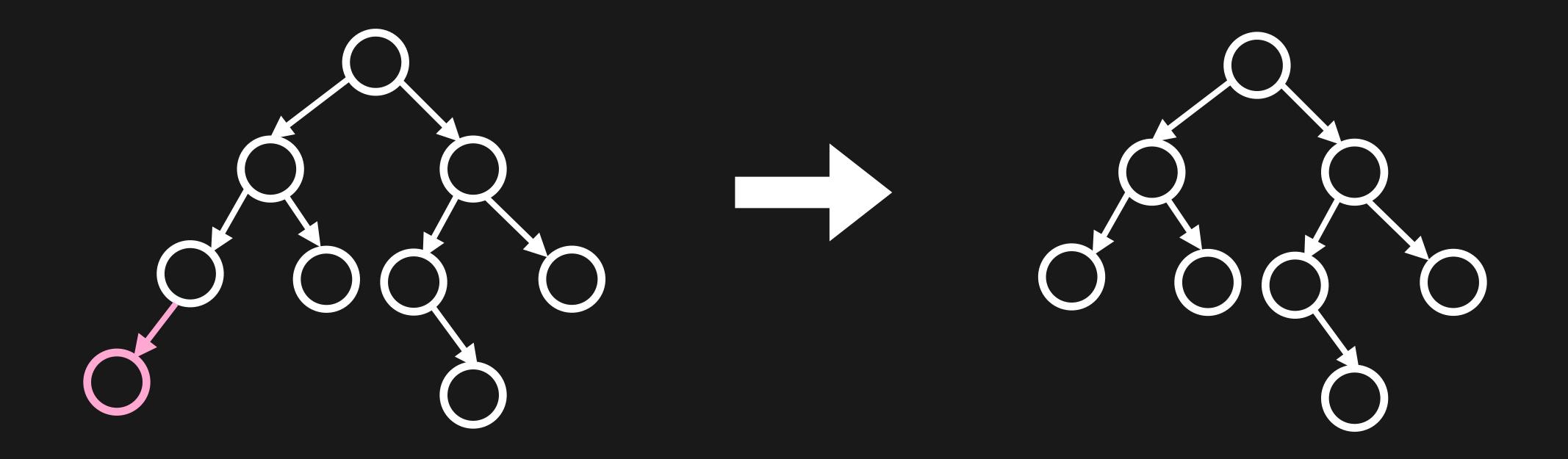
Three cases

No children

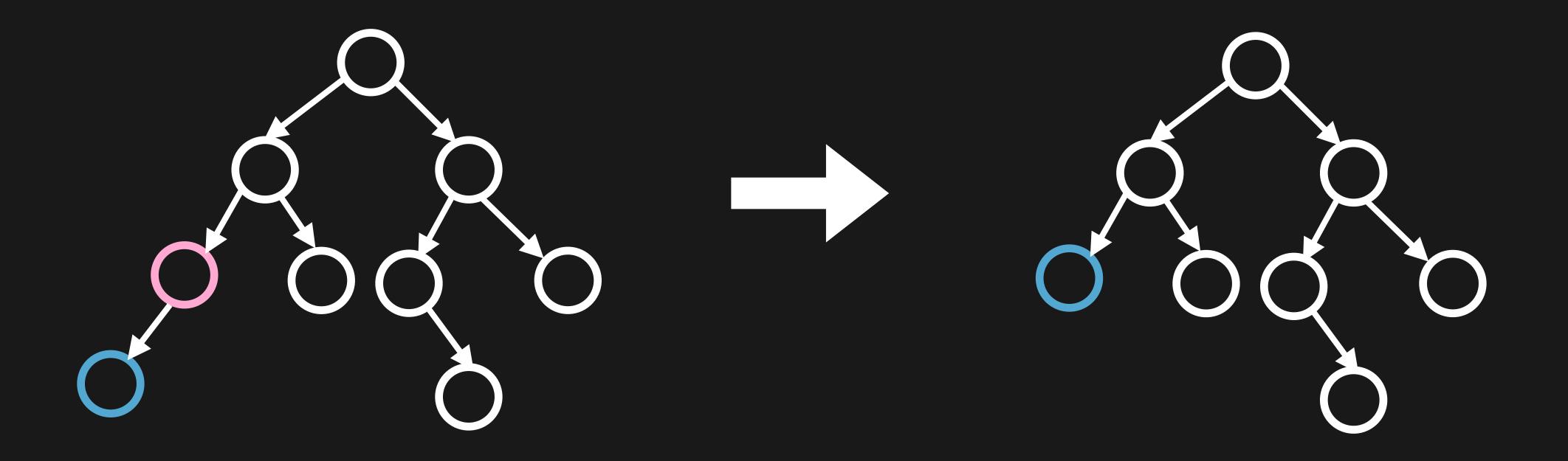
One child

Two children

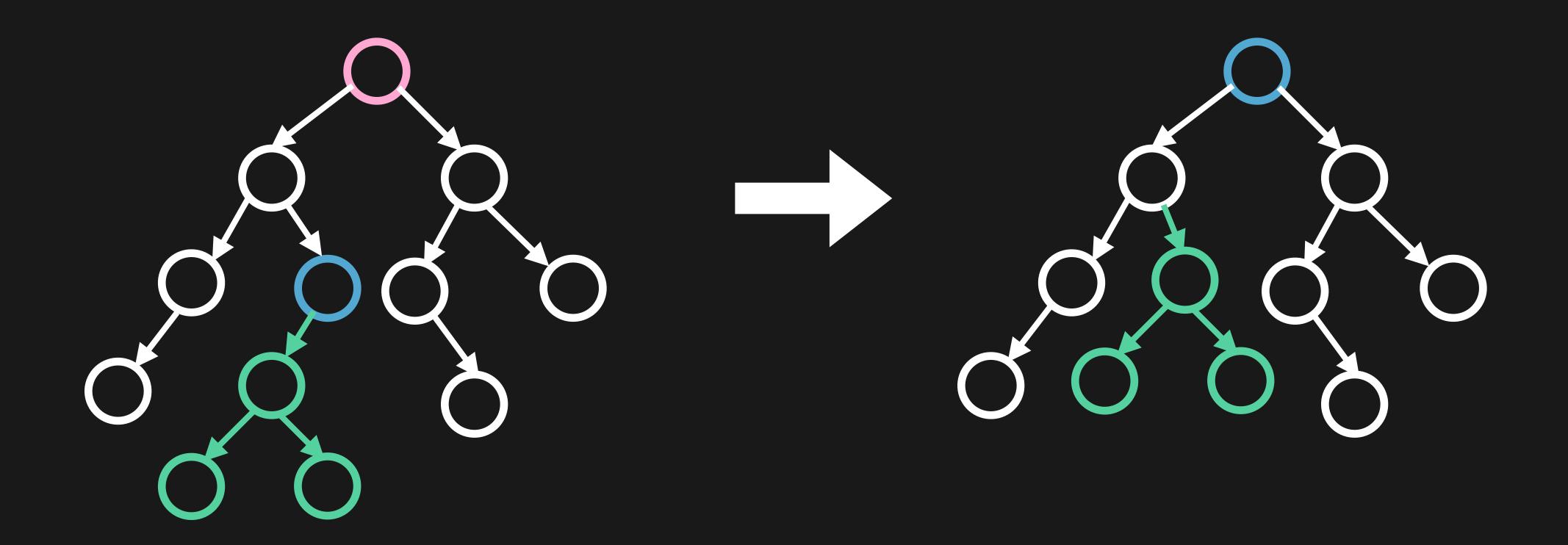
### NO CHILDREN



## ONE CHILD



## TWO CHILDREN



### COMPLEXITY

	Average Case	Worst Case
Space	O(n)	O(n)
Search	O(log n)	O(n)
Insert	O(log n)	<b>O</b> (n)
Delete	O(log n)	O(n)