

# Algorithms and Data Structures for Data Science

## Binary Search Tree

CS 277

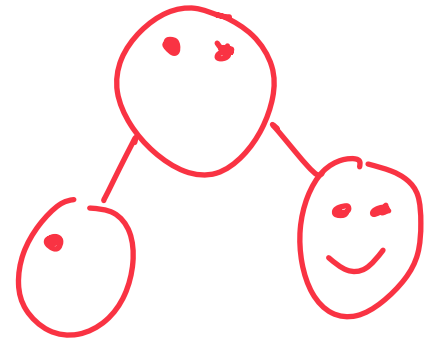
Brad Solomon

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UNIVERSITY OF  
**ILLINOIS**  
URBANA - CHAMPAIGN

Department of Computer Science



# Learning Objectives

Review understanding of Binary Trees

Practice tree traversals

Introduce the dictionary ADT

Extend ADT to Binary Search Trees

Practice recursion in the context of trees

*"Binary tree"*

???

*"Binary Search tree"*

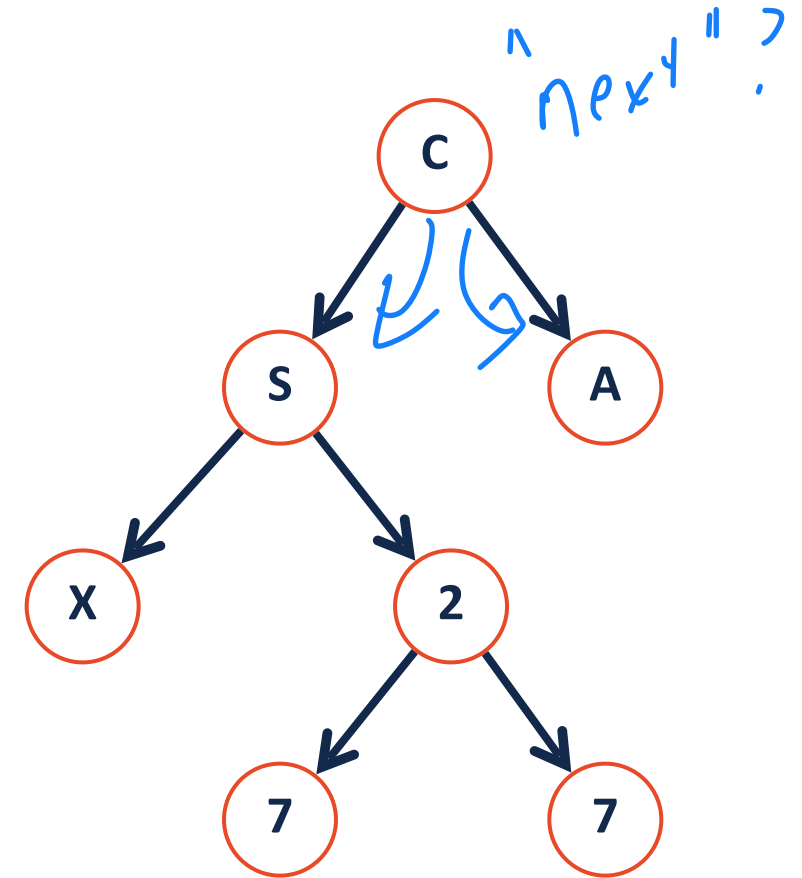
# (Binary) Tree Recursion

A **binary tree** is a tree  $T$  such that:

$T = None$

or

$T = treeNode(val, T_L, T_R)$



```
1 class treeNode:
2     def __init__(self, val, left=None, right=None):
3         self.val = val
4         self.left = left
5         self.right = right
```

```
1 class binaryTree:
2     def __init__(self):
3         self.root = None
4
5
```

# Tree ADT

**Constructor**: Build a new (empty) tree

**Insert**: Add an object into tree

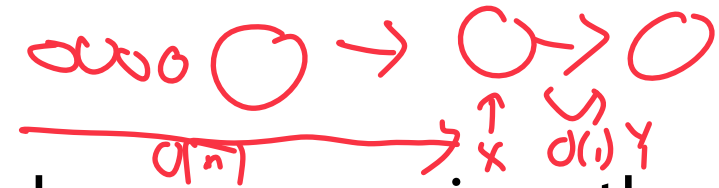
**Remove**: Remove a specific object from tree

**Traverse**: Visit every node in tree (all objects)

**Search**: Find a specific object in the tree

} find or get specific  
values

# Binary Tree Insert and Remove



Last class we implemented insert and remove where we are given the parent node and direction (allowing us to reach the node of interest)

Insert was a lot like what previous data structure:

1) the necessary info is very similar

2) the runtime was similar  $O(1)$

↳ If not need to find parent  $O(n)$

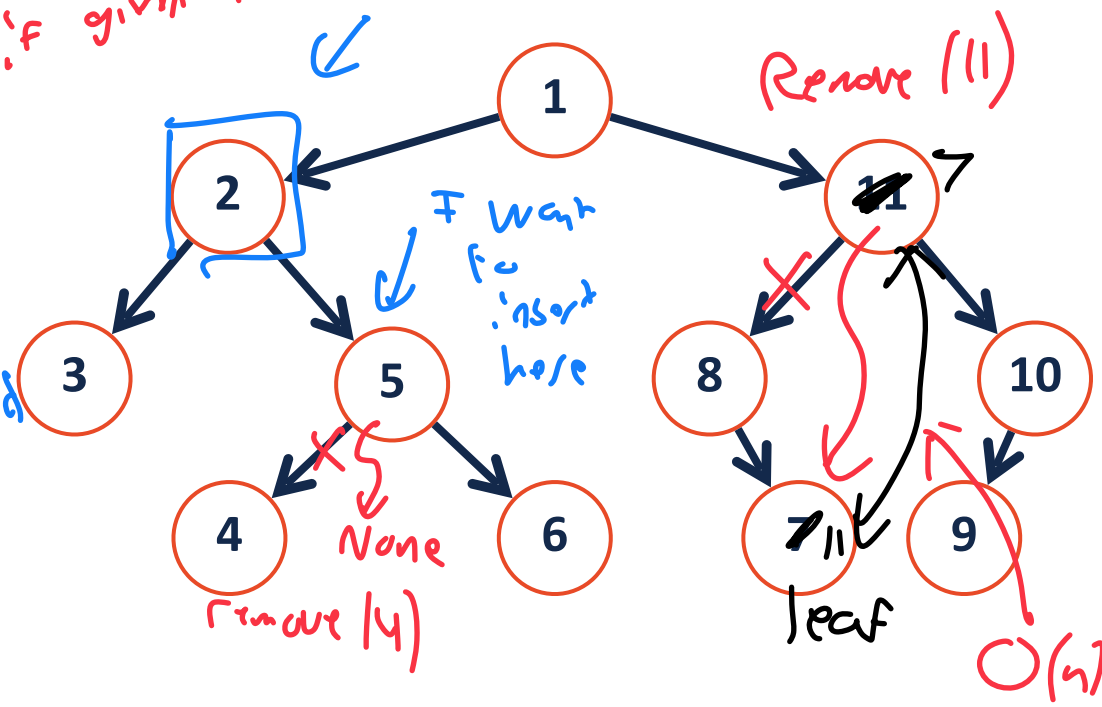
if given node so not need this node

Remove has one bad case, which was:

0 or 1 child at node to be removed

↳  $O(1)$

2 child removal  $\rightarrow O(n)$



# Tree Traversal

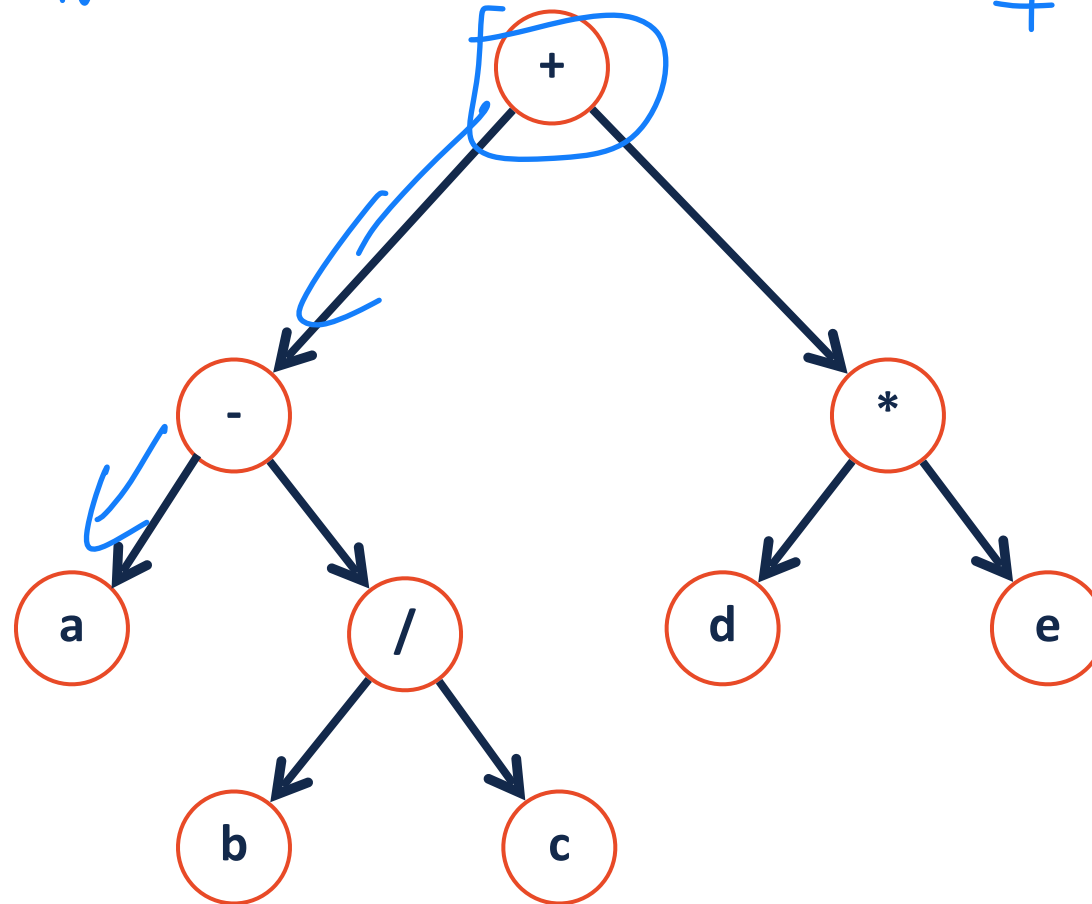
$\rightarrow O(n)$

A **traversal** of a tree T is an ordered way of visiting every node once.

1) Look at / process Node  
↳ print

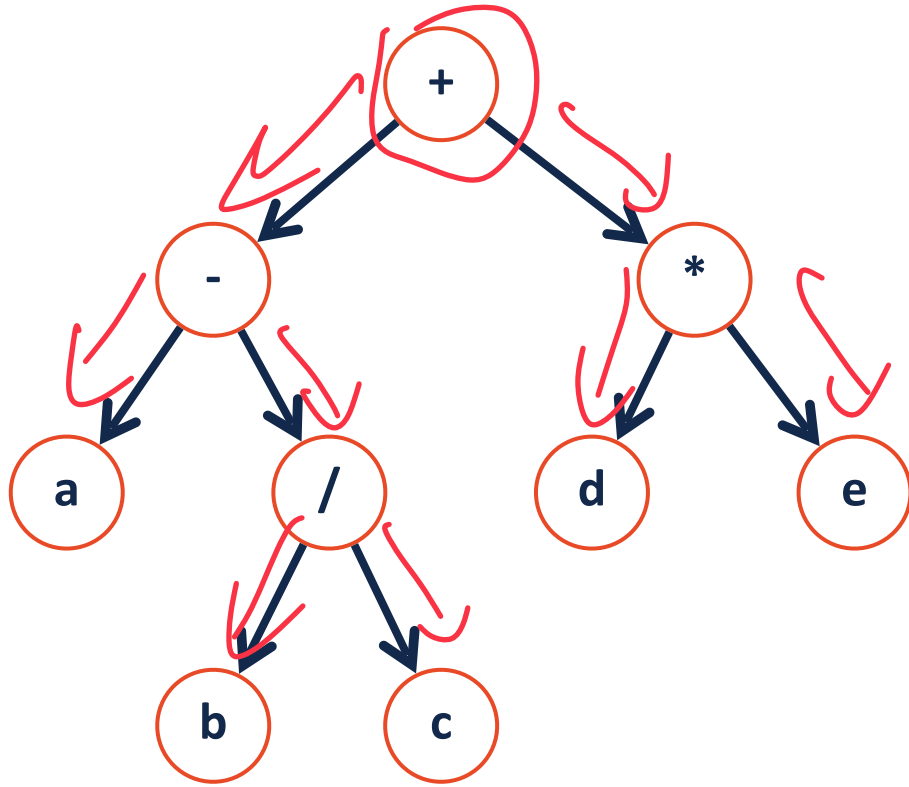
2) Recurse left

3) Recurse right



+ - 9

# Pre-order Traversal



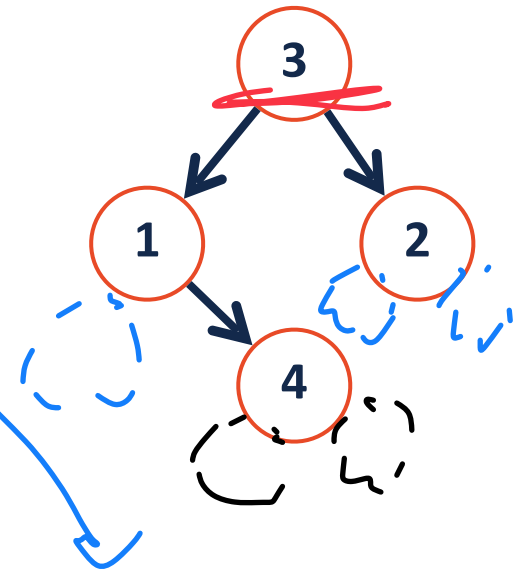
```
1 def preorderTraversal (node) :
2   if node:
3
4     print (node.val) process
5
6     preorderTraversal (node.left)
7
8     preorderTraversal (node.right)
9
10
11
```

**Pre-order:**

*+ - a / b c \* d e*

# Pre-order Traversal Visualized

3 1 4 2



```

1 def preorderTraversal (<3>):
2   if node:
3     print (<3>.val)
4     preorderTraversal (<3>.left)
5     preorderTraversal (<3>.right)
6
  
```

```

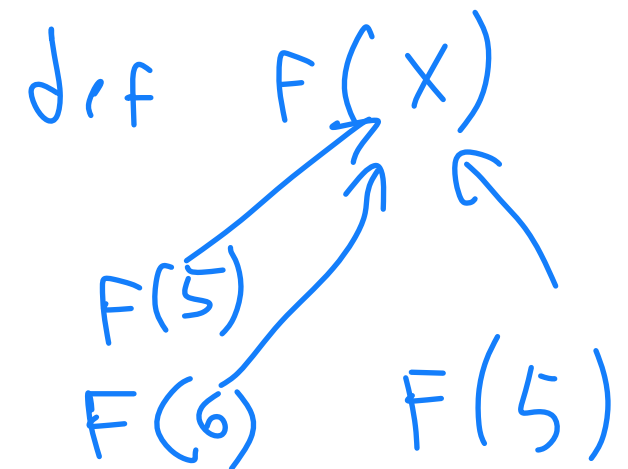
1 def preorderTraversal (<1>):
2   if node:
3     print (<1>.val)
4     preorderTraversal (<1>.left)
5     preorderTraversal (<1>.right)
6
  
```

```

1 def preorderTraversal (<4>):
2   if node:
3     print (<4>.val)
4     preorderTraversal (<4>.left)
5     preorderTraversal (<4>.right)
6
  
```

```

1 def preorderTraversal (<2>):
2   if node:
3     print (<2>.val)
4     preorderTraversal (<2>.left)
5     preorderTraversal (<2>.right)
6
  
```



preorder Traversal (None)  
Return None

None

return None

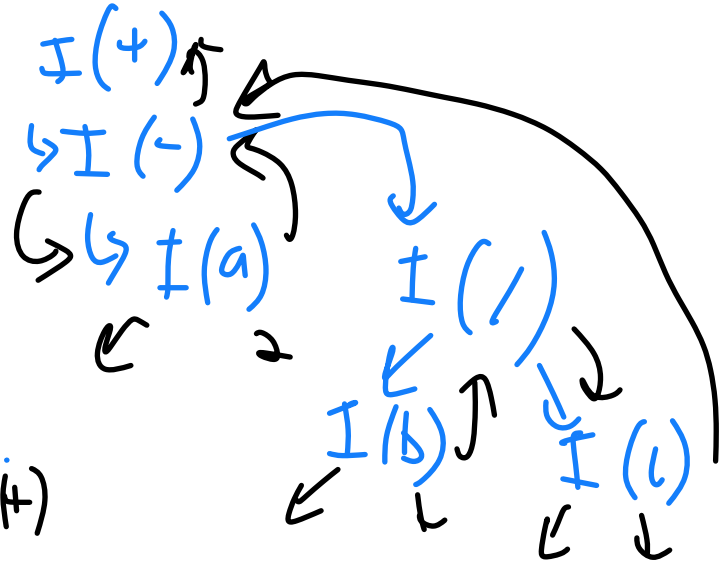
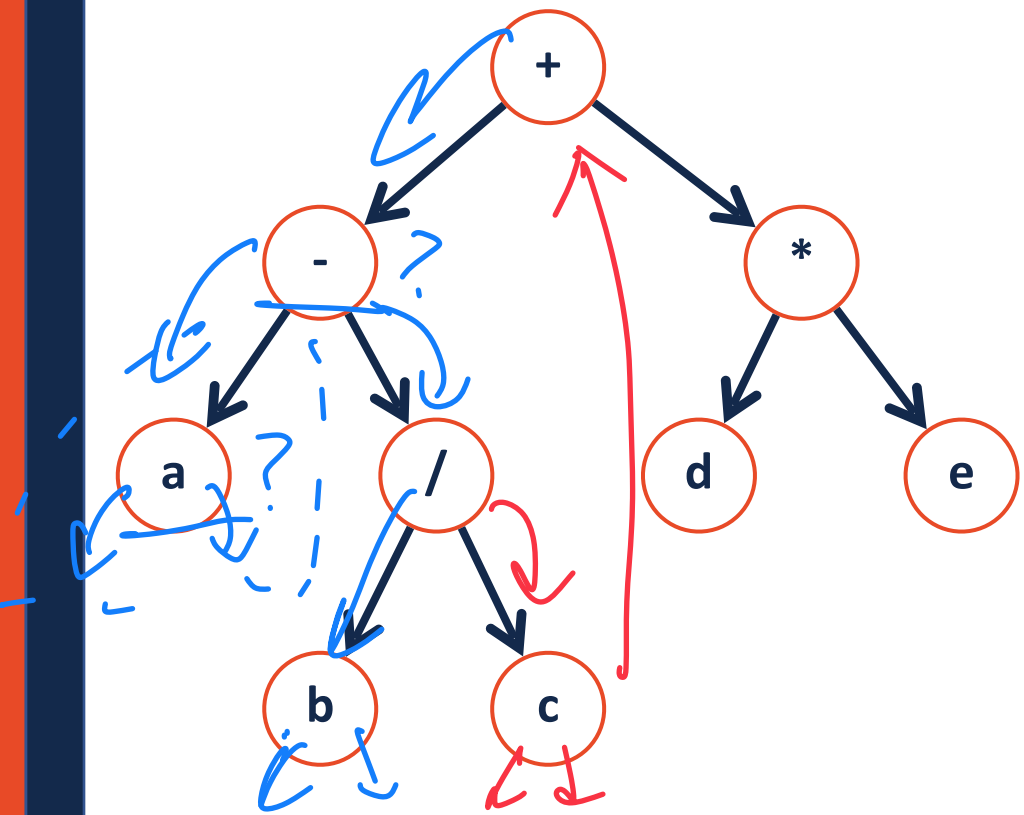
1  
2  
3



# In-order Traversal

- 1) Recurse Left
- 2) Process Node
- 3) Recurse Right

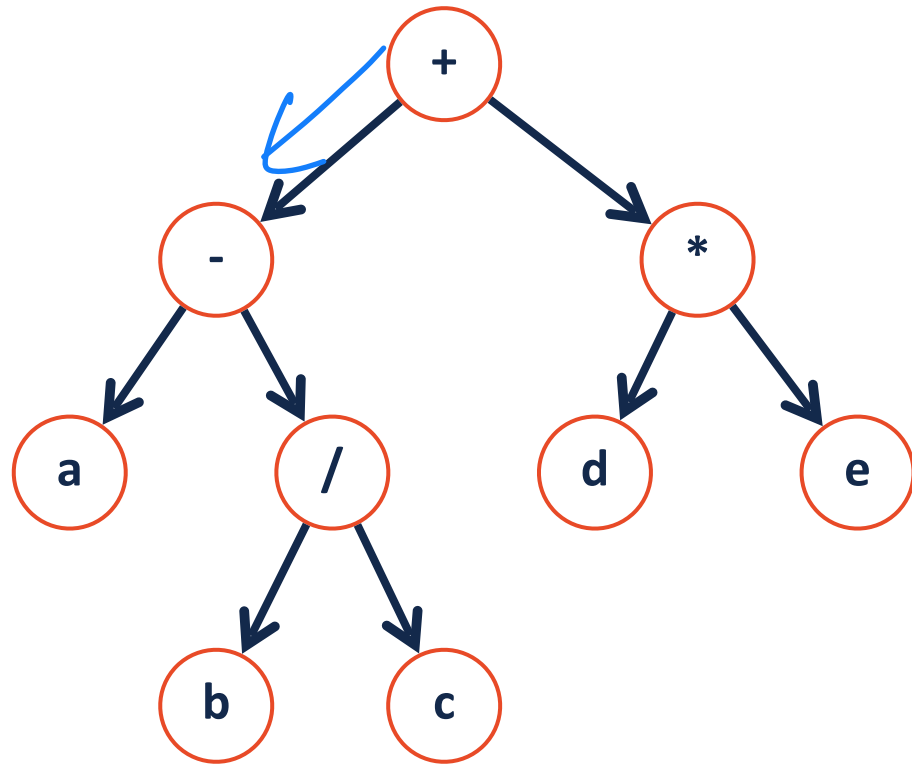
mit / current Node



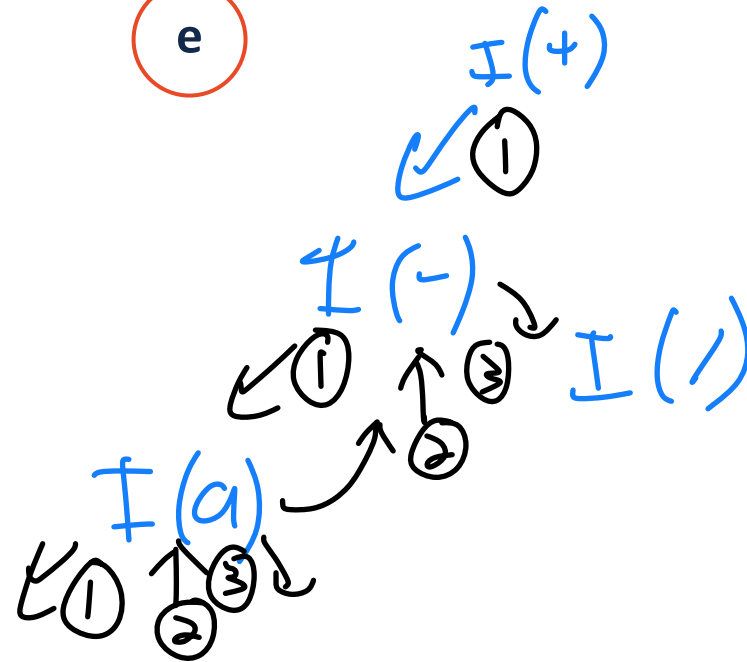
printed all left of (+)

**In-order:** a — b / c ; + ; d \* e

# In-order Traversal

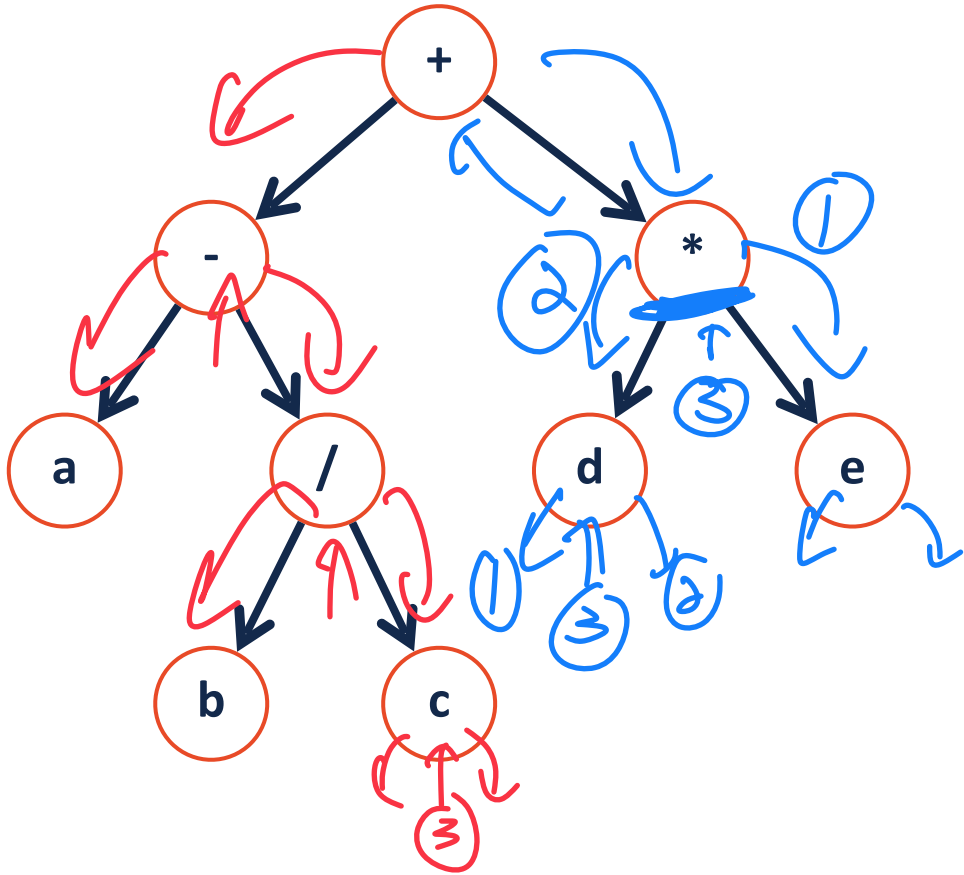


- 1) Recurse Left
- 2) Process Node mit / current Node
- 3) Recurse Right

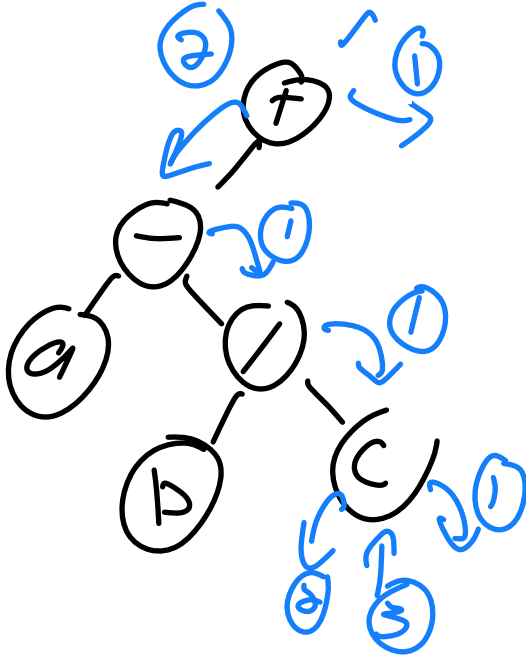


In-order: a —

# Post-order Traversal



- 1) recurse (right)
- 2) recurse (left)
- 3) process (current)



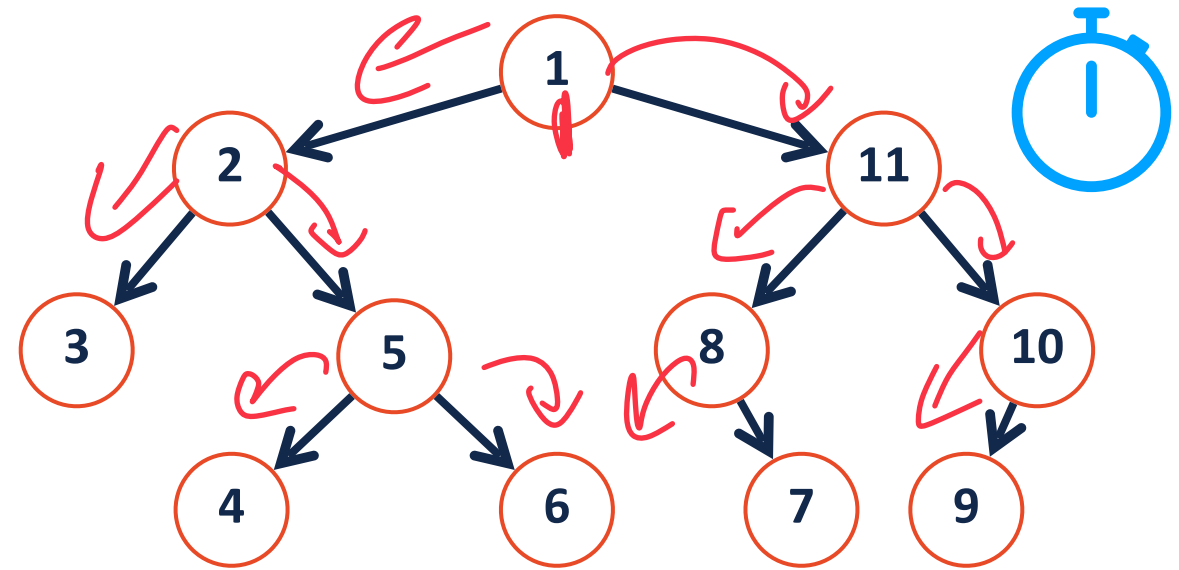
Post-order:

e d \* c b / a - f

# Tree Traversals

Lets practice our traversals!

*Left  
myself  
Right*



**Pre-order:**

**In-order:**

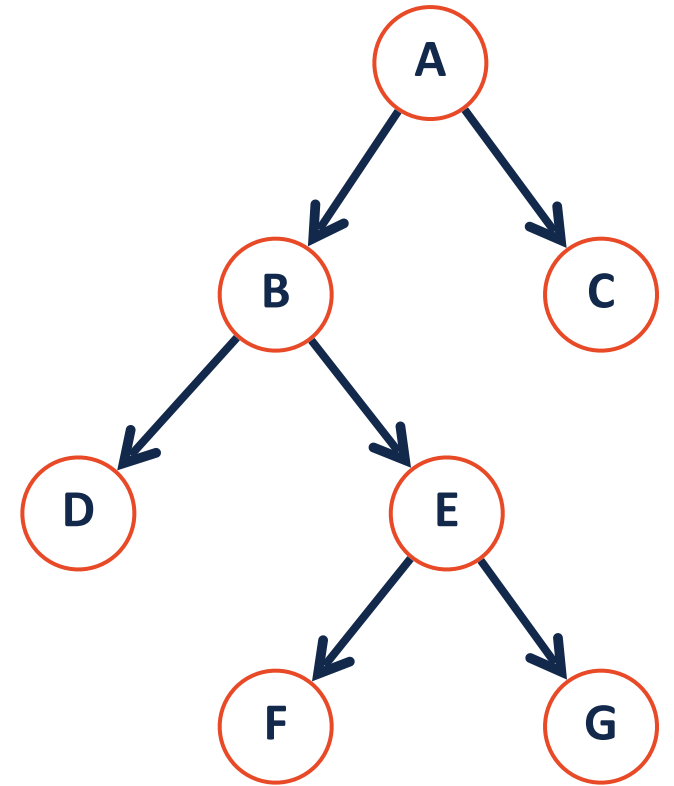
*3 2 4 5 6 1 8 7 11 9 10*

**Post-order:**

# Traversal vs Search

**Traversal** - Always looks at every Node  
↳  $O(n)$

**Search** - Might look at every Node  
↳  $O(n)$  <sup>worst case</sup>



D → D ↔ D → D

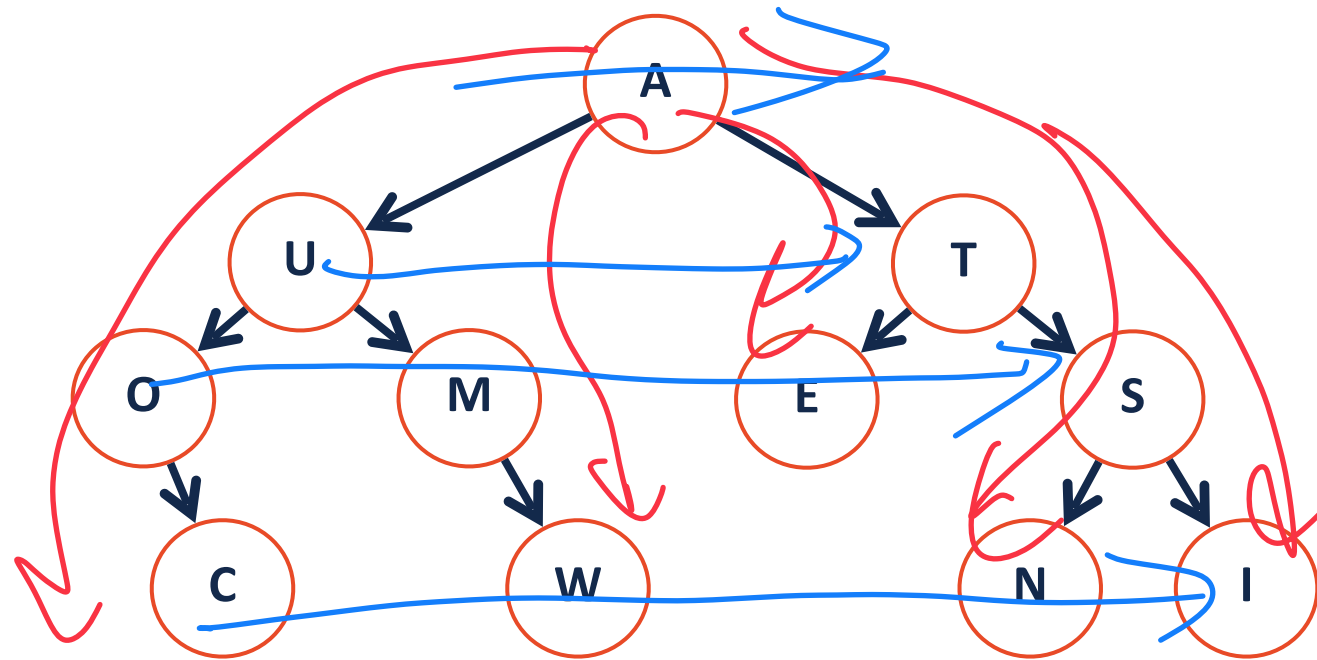
X

# Searching a Binary Tree

There are two main approaches to searching a binary tree:

Depth first

Breadth first



# Depth First Search

1, 11, 10, 9, 8, 7

Explore as far along one path as possible before backtracking

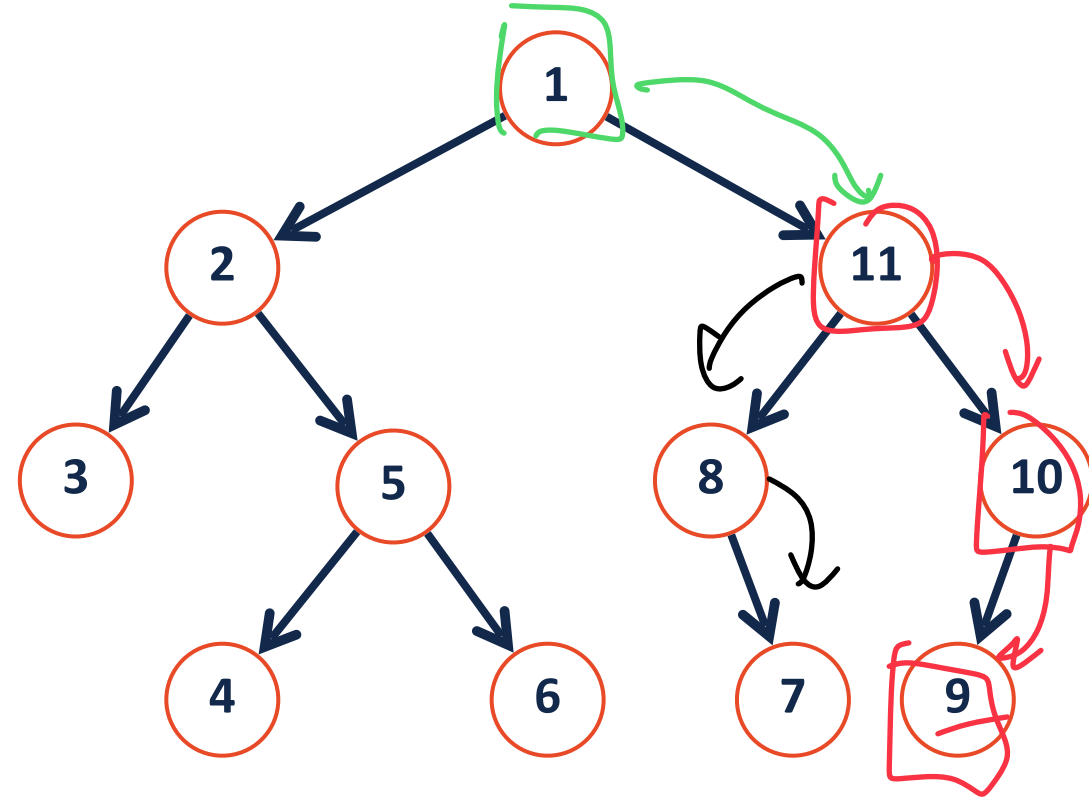
1) Make stack. Stack.push(1)

2) while stack is not empty:

n = Stack.pop()

2.5) Add children of n to stack

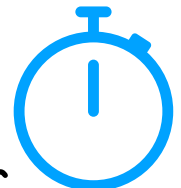
2.6) Process n ("Print")



~~1~~ 2 ~~11~~ 8 ~~10~~ 9 ~~7~~

# Breadth First Search

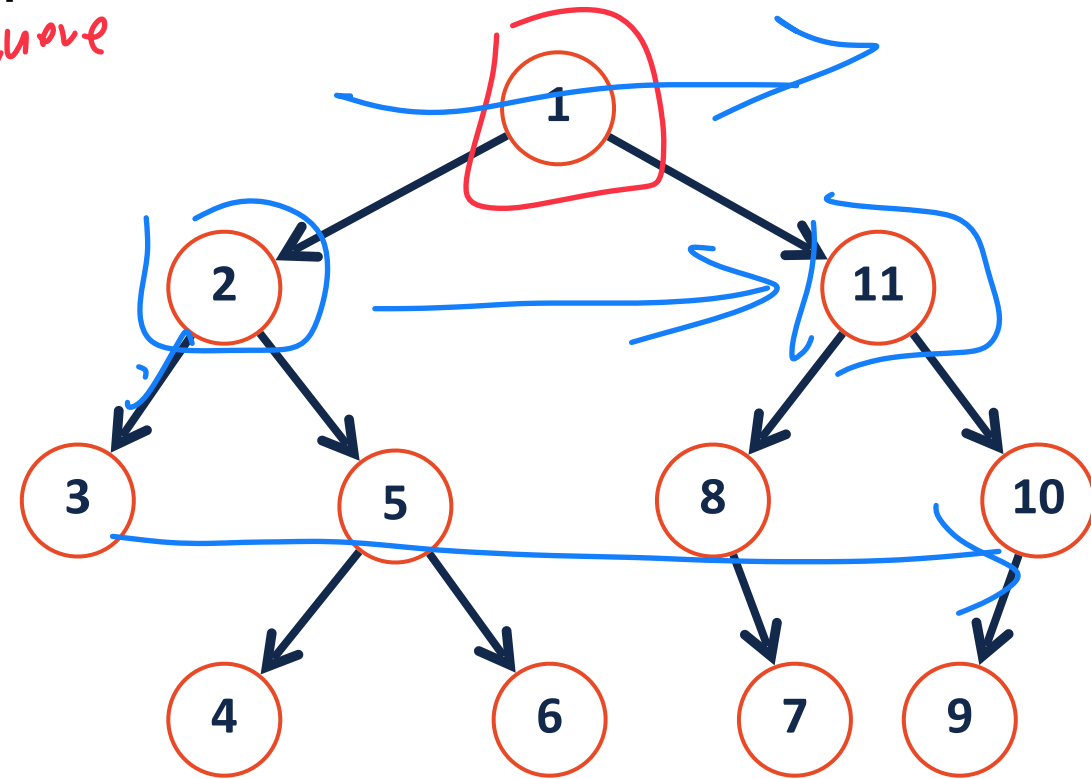
1, 2, 11, 3, 5, 8, 10  
 4, 6, 7, 9



Fully explore depth  $i$  before exploring depth  $i+1$

↳ take stack code, replace w/ queue

- 1) Make queue, enqueue (root)
- 2) while queue not empty
  - a.5)  $n = q.dequeue()$
  - process Node / add children to  $q$





# Traversal vs Search II

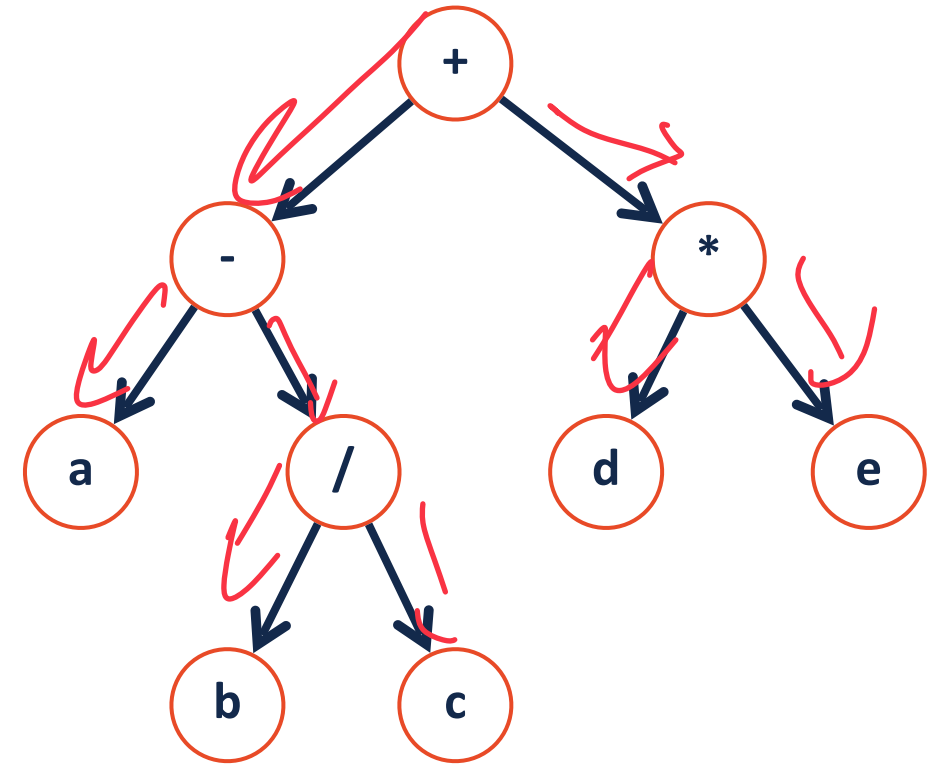
Pre-order, in-order, and post-order are three ways of doing which search?

*Depth-first search strategies*

**Pre-order:** + - a / b c \* d e

**In-order:** a - b / c + d \* e

**Post-order:** a b c / - d e \* +



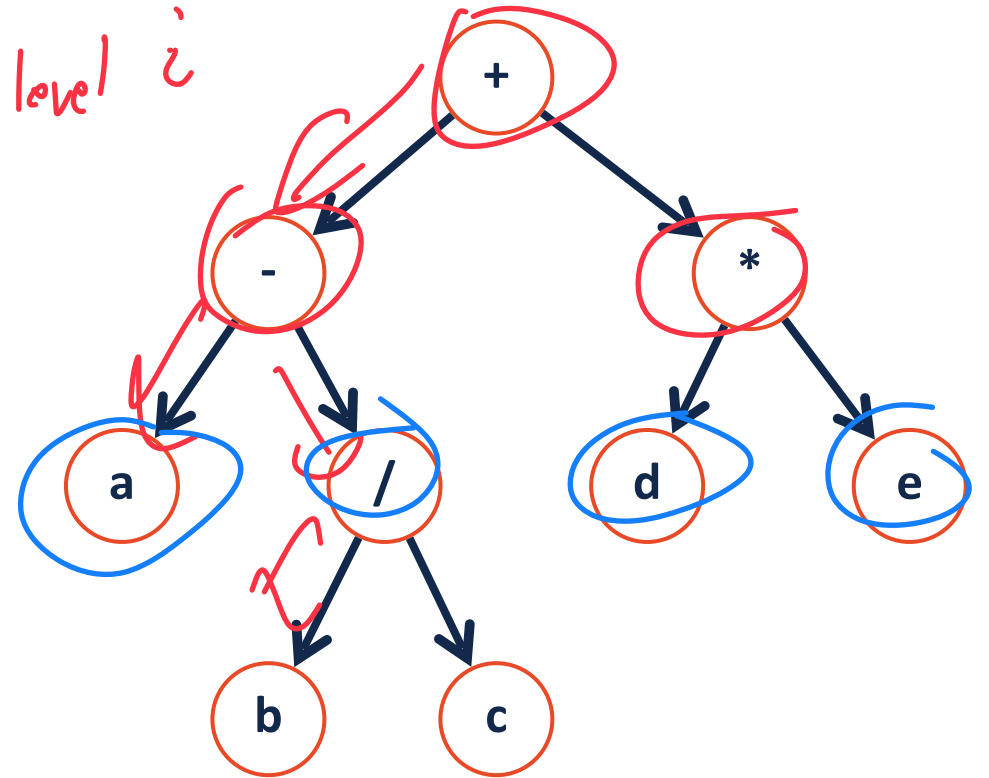
# Level-Order Traversal

A tricky recursive implementation but an easier queue implementation!

It's possible

1) Recurse through tree but only print level  $i$   
 $i \pm 1$

2) adjust code to return recursive list at depth

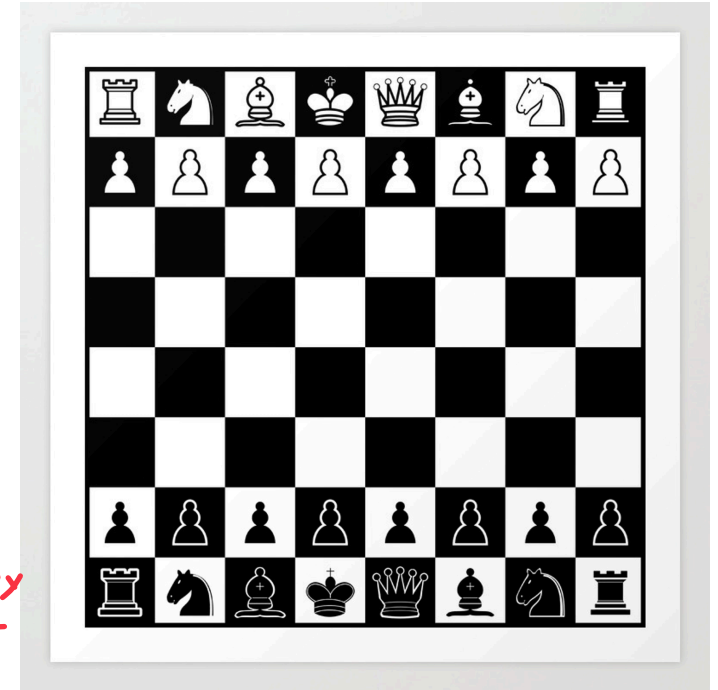


**Level-order:**



# What search algorithm is best?

The average 'branch factor' for a game of chess is ~31. If you were searching a decision tree for chess, which search algorithm would you use?



~50

31

Iterative deepening  
DFS  
↳ DFS but limit height. Expand if necessary

Neither really work!

DFS

∞ branch

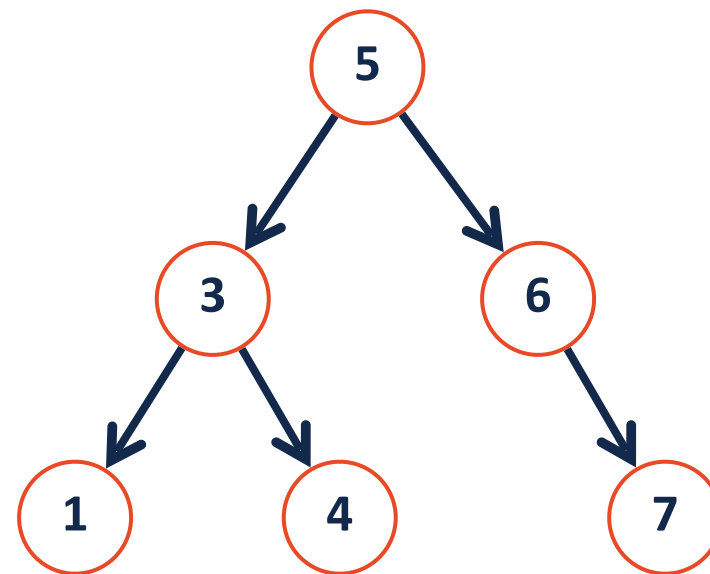
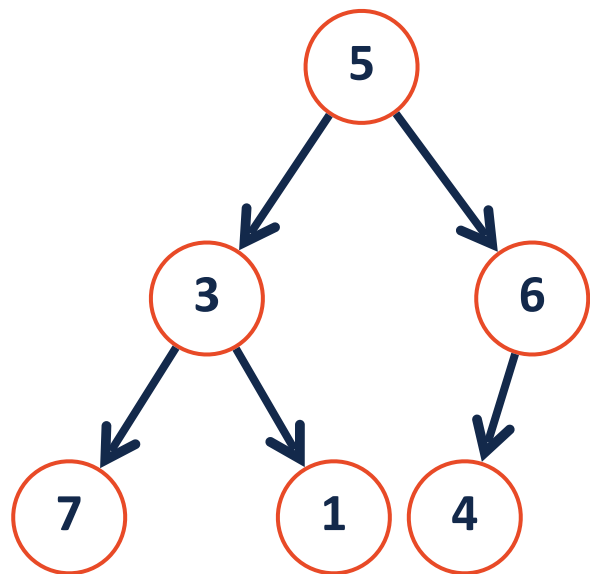
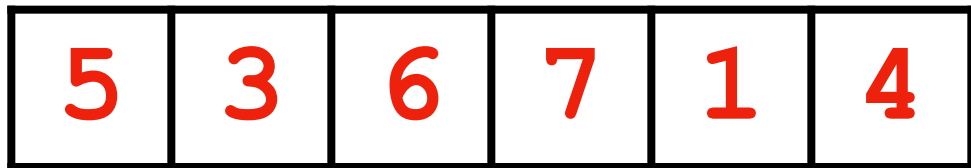
BFS level

1  
2  
3

31  
31  
31

Memory problems

# Improved search on a binary tree

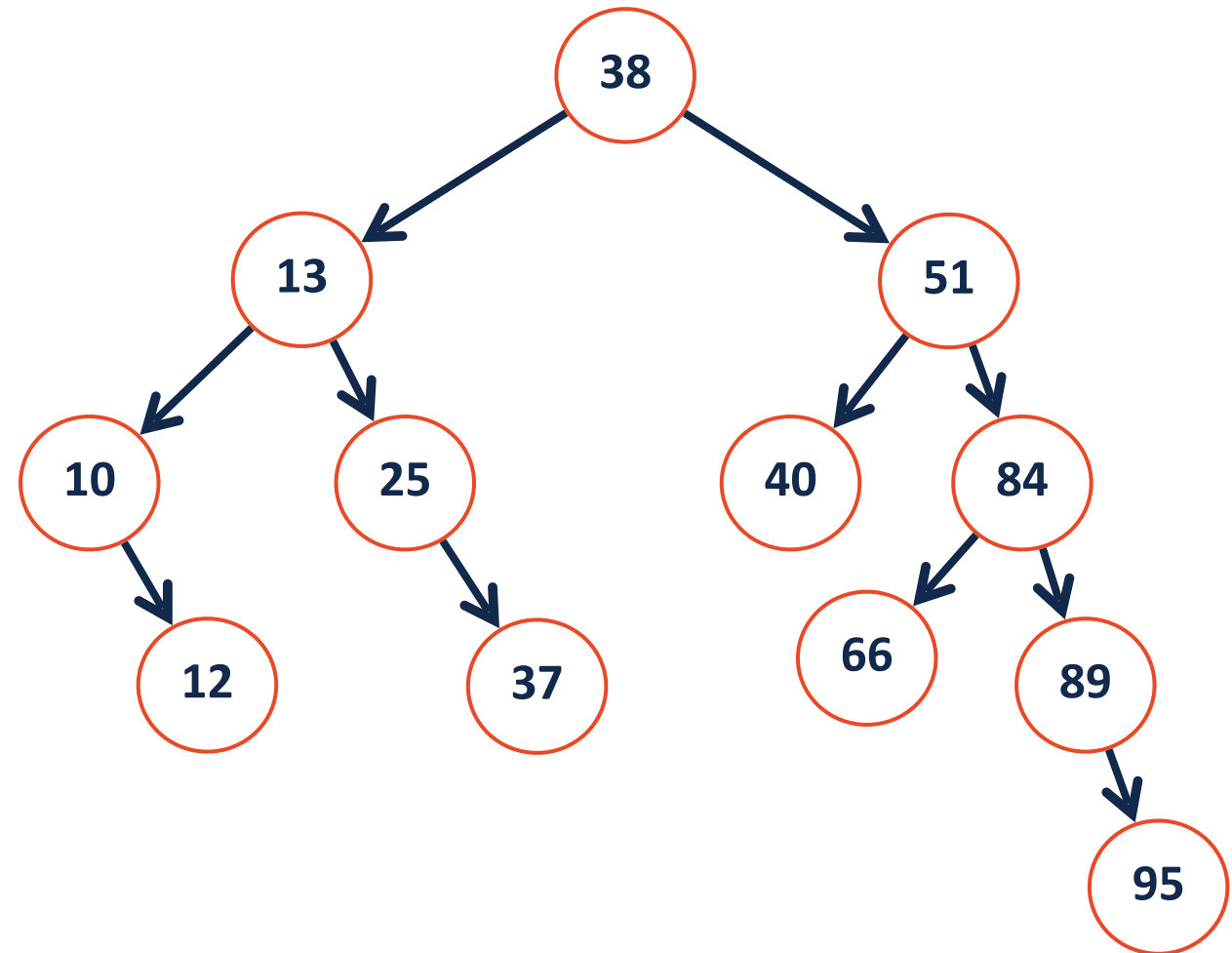


# Binary Search Tree (BST)

A **BST** is a binary tree  $T = treeNode(val, T_L, T_r)$  such that:

$\forall n \in T_L, n.val < T.val$

$\forall n \in T_R, n.val > T.val$



# Dictionary ADT

**Data is often organized into key/value pairs:**

Word → Definition

Course Number → Lecture/Lab Schedule

Node → Edges

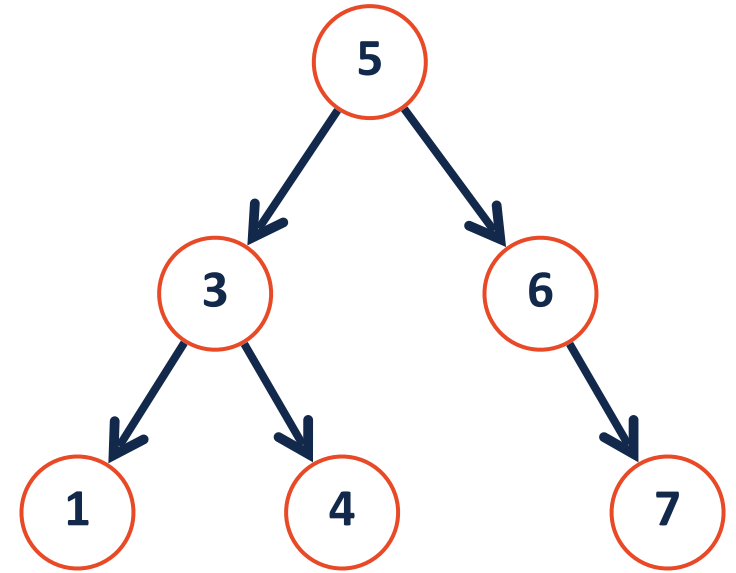
Flight Number → Arrival Information

URL → HTML Page

Average Image Color → File Location of Image

# Dictionary as a Binary Search Tree

```
1 class bstNode:  
2     def __init__(self, key, val, left=None, right=None):  
3         self.key = key  
4         self.val = val  
5         self.left = left  
6         self.right = right
```



Key	5	3	6	7	1	4
Value	A	B	C	D	E	F

# Binary Search Tree ADT



**Constructor:** Build a new (empty) tree

**Insert:** Add an object into tree

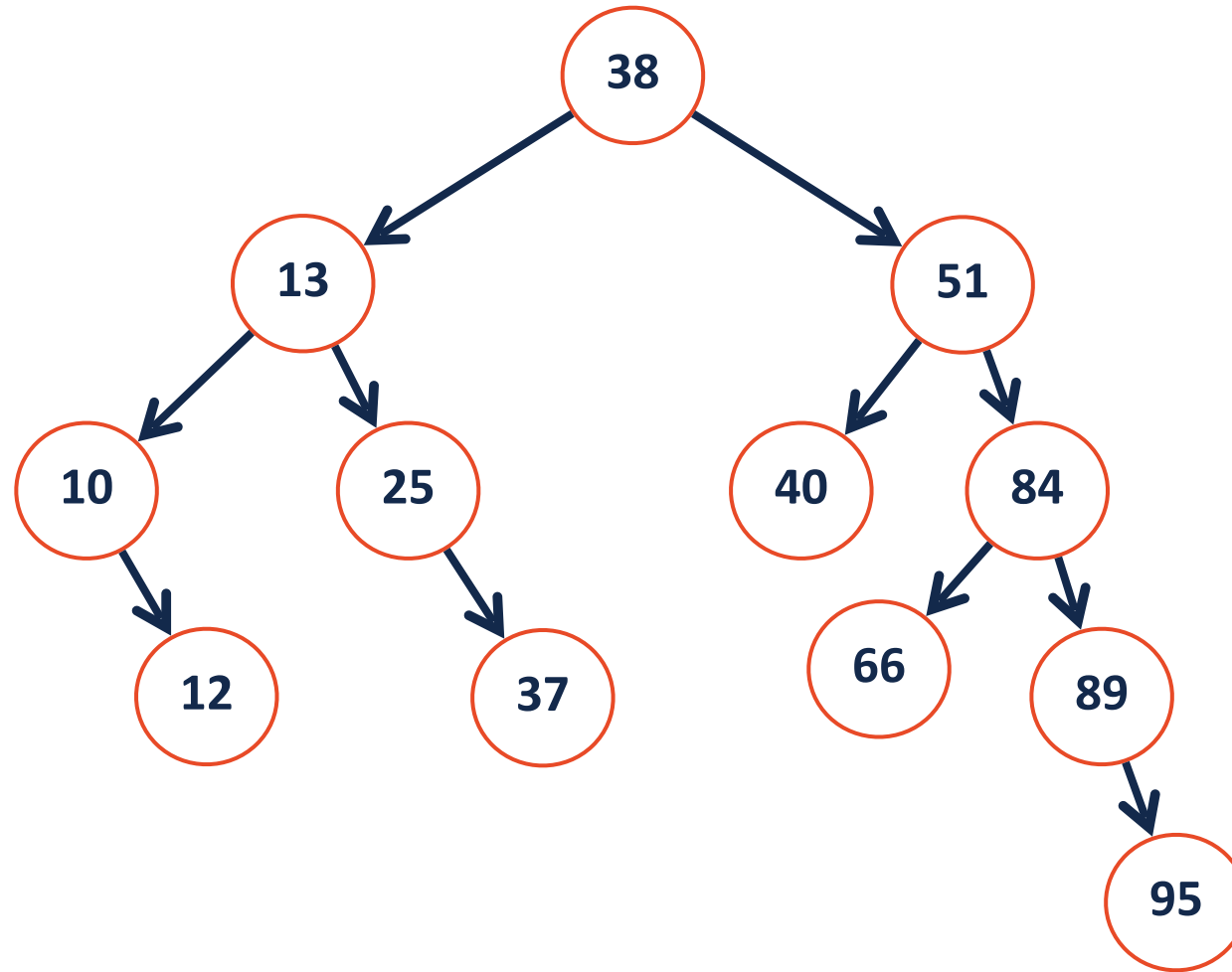
**Remove:** Remove a specific object from tree

**Traverse:** Visit every node in tree (all objects)

**Search:** Find a specific object in the tree

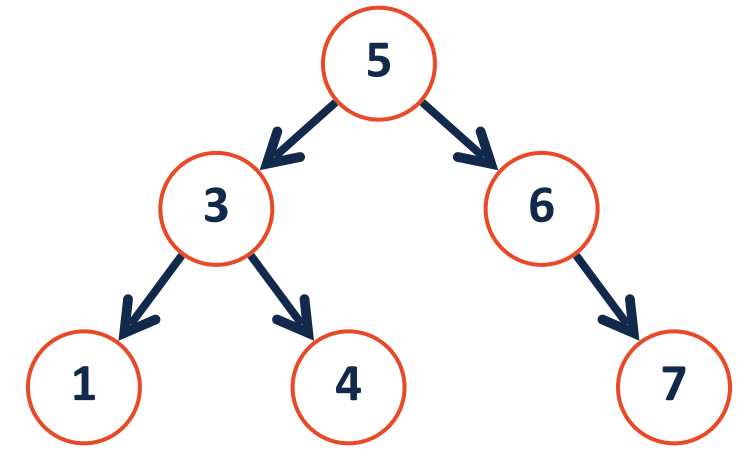


# BST In-Order Traversal



# BST Insert

**Base Case:**

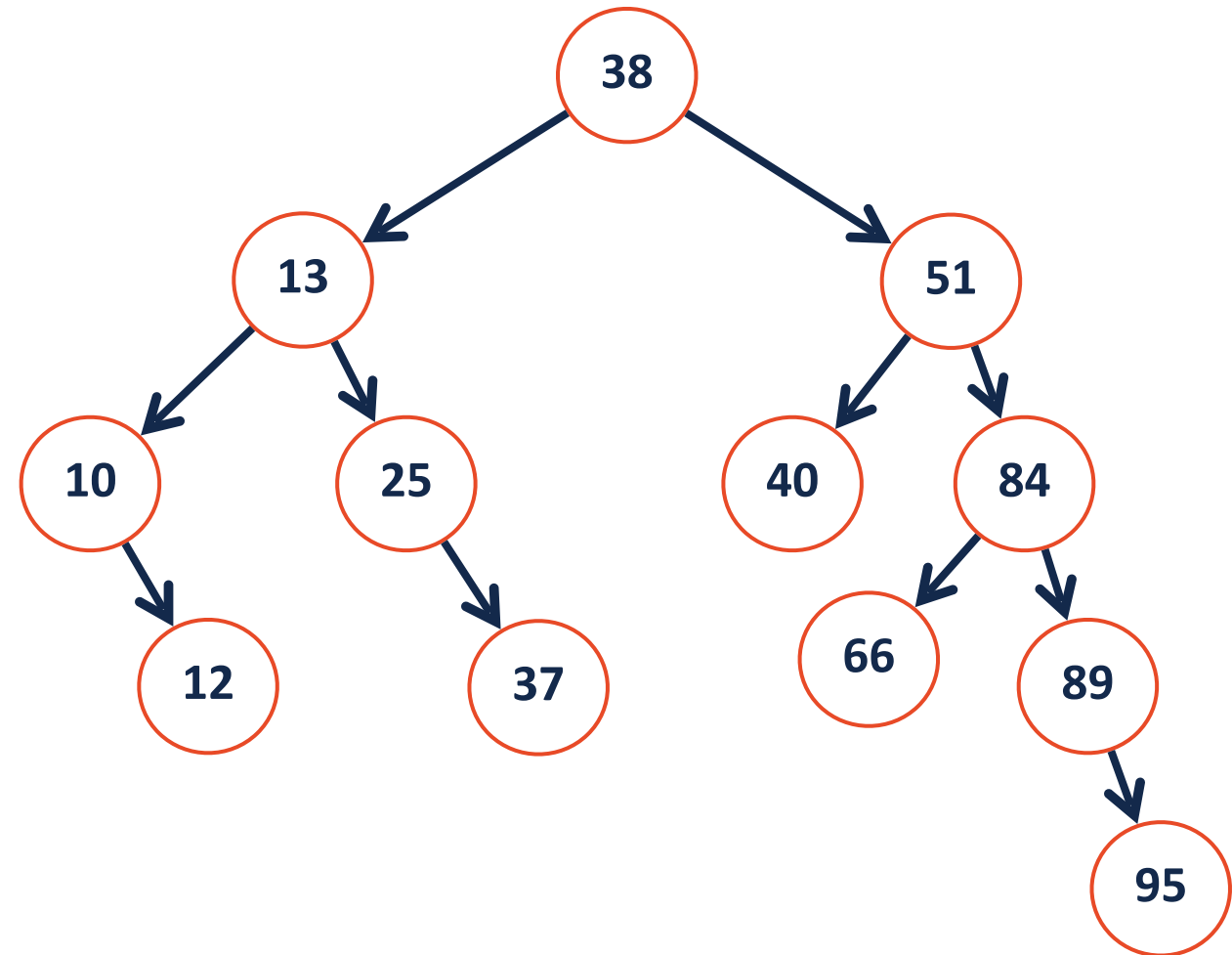


**Recursive Step:**

**Combining:**

# BST Insert

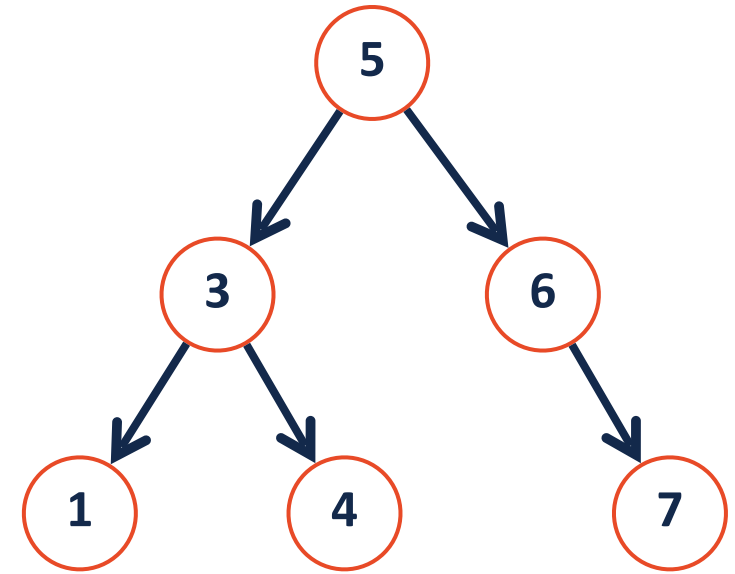
**insert(33)**



# BST Insert



```
1 # inside class bst
2 def insert(self, key, val):
3     self.root = self.insert_helper(self.root, key, val)
4
5 def insert_helper(self, node, key, val):
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
```

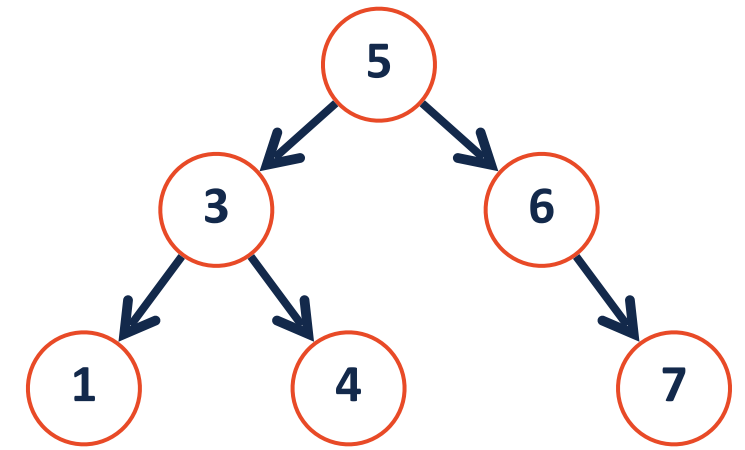


# BST Insert

What binary would be formed by inserting the following sequence of integers: [3, 7, 2, 1, 4, 8, 0]

# BST Find

**Base Case:**

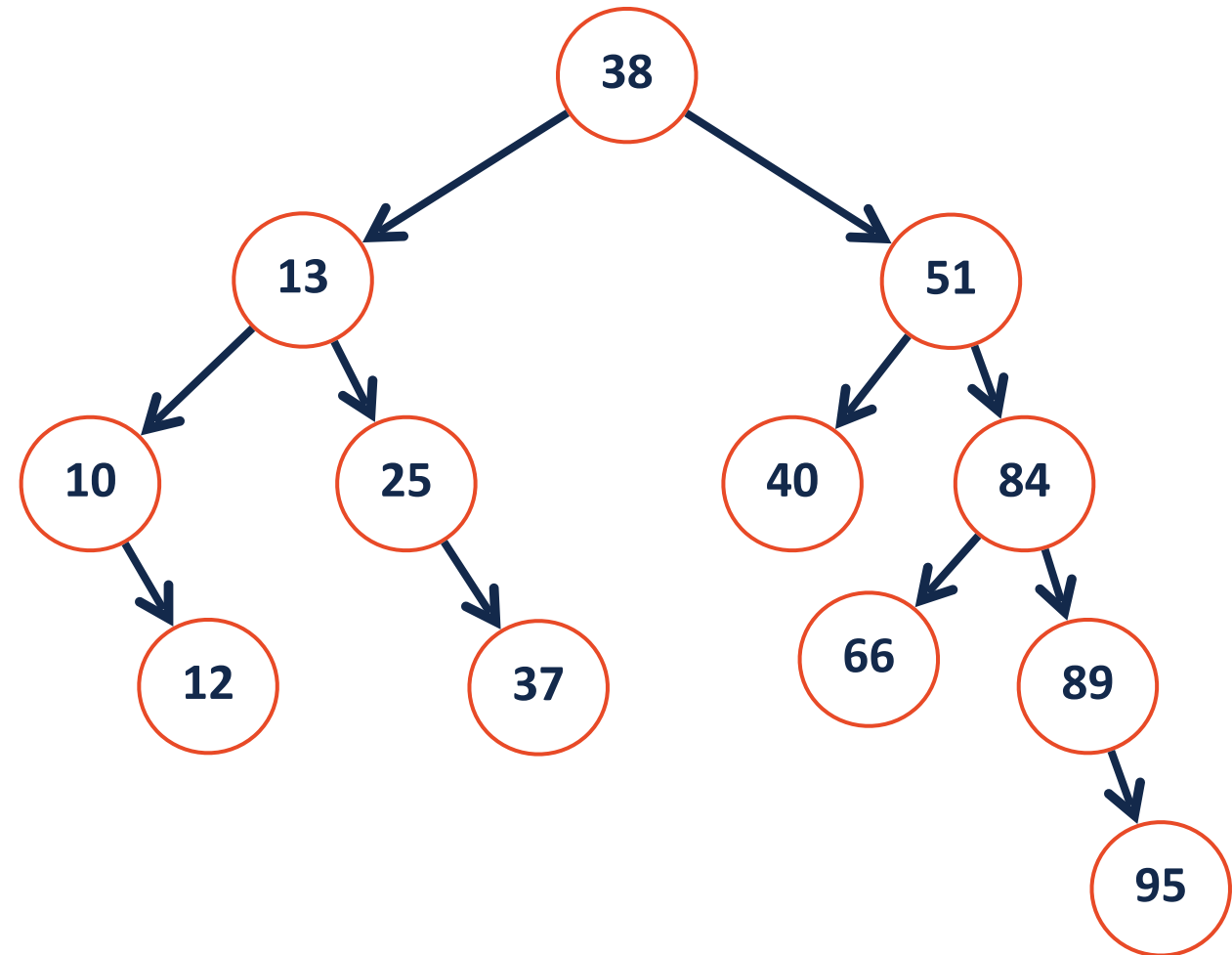


**Recursive Step:**

**Combining:**

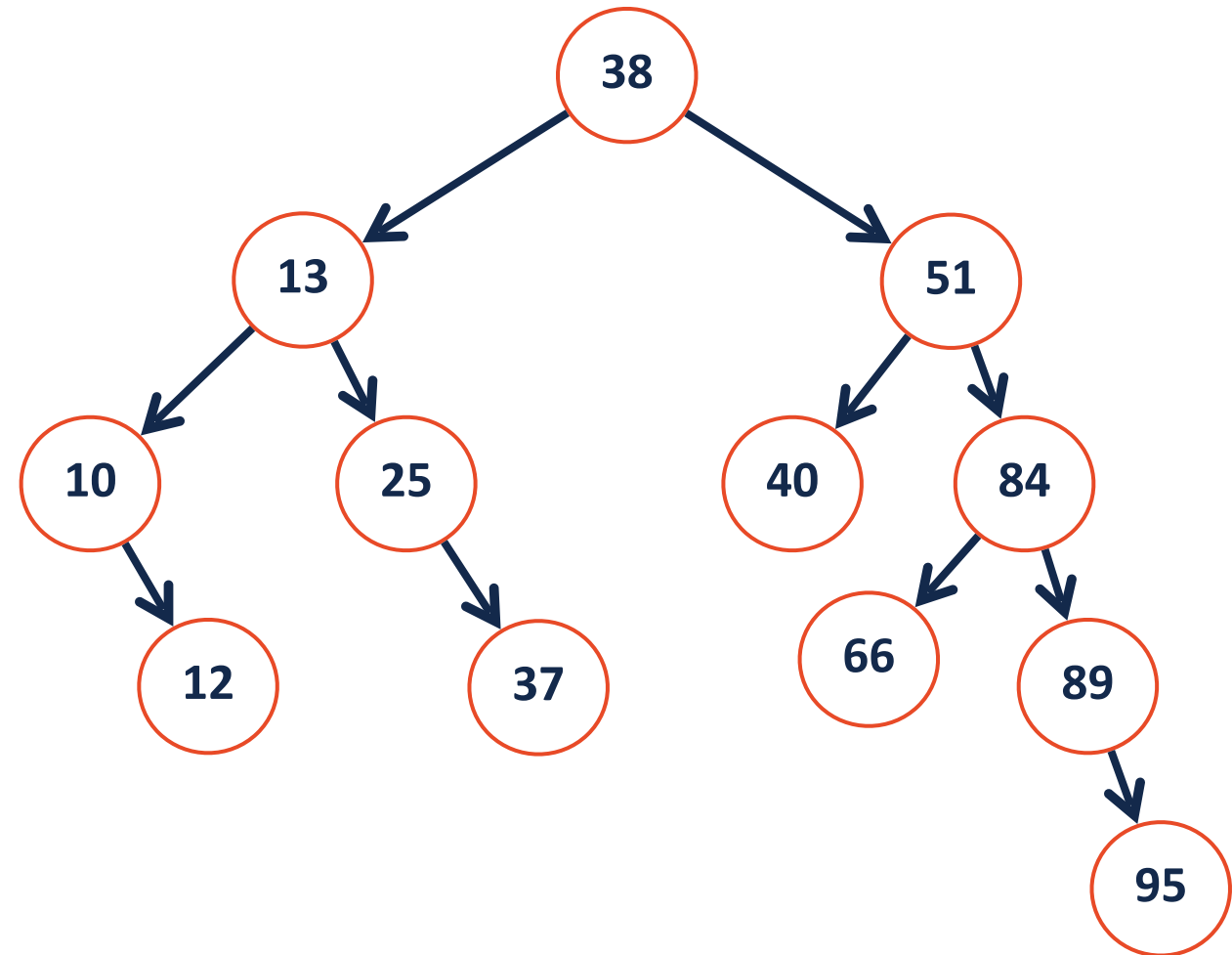
# BST Find

**find(66)**



# BST Find

**find(9)**

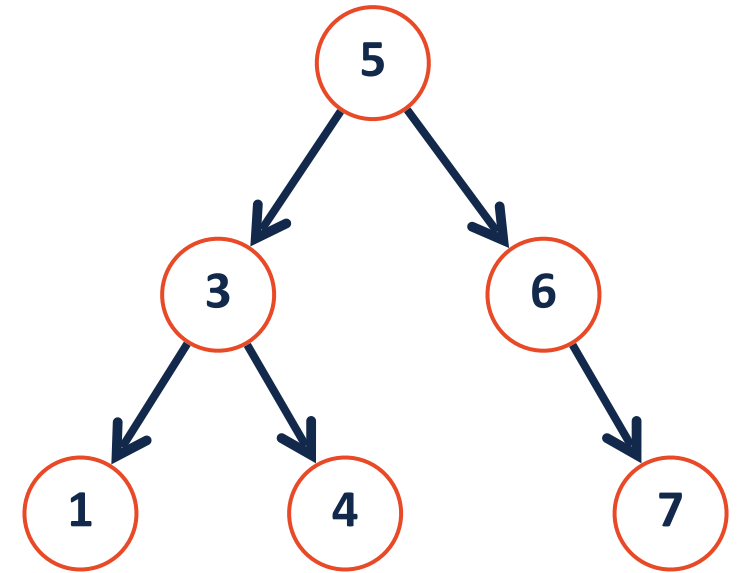




# BST Find

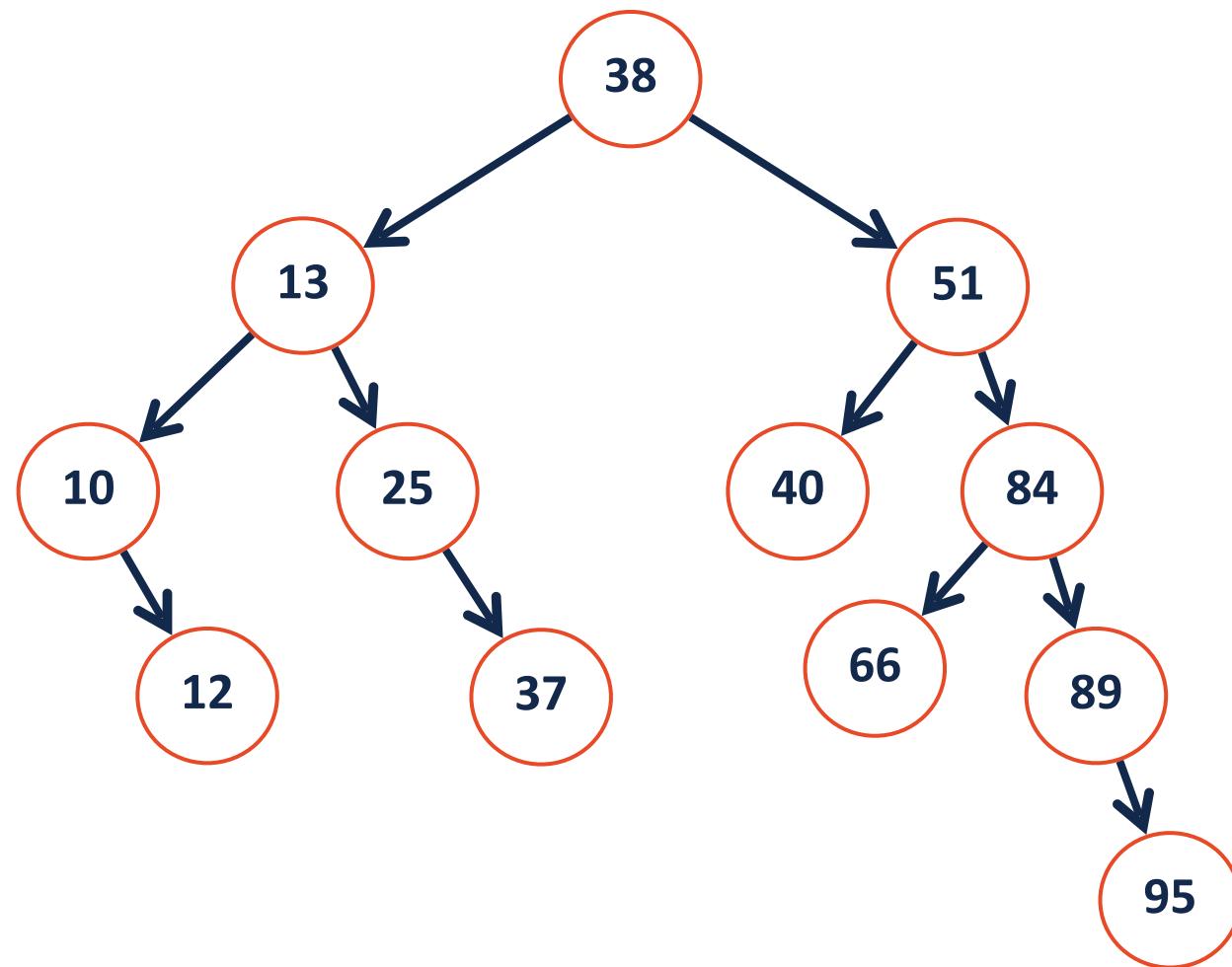


```
1 #inside class bst
2 def find(self, key):
3
4
5
6
7
8
9 def find_helper(self, node, key):
10
11
12
13
14
15
16
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19
20
21
22
23
```



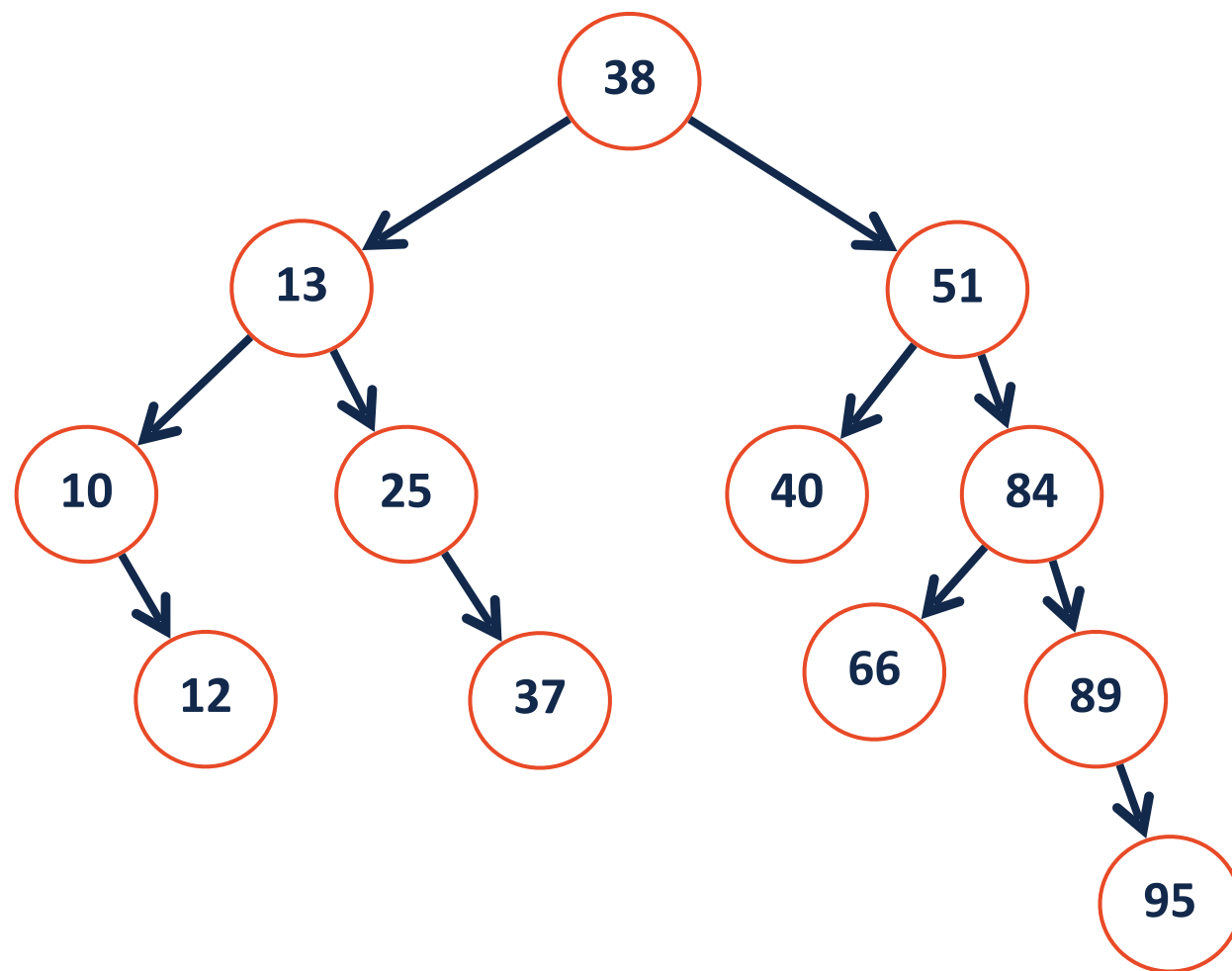
# BST Remove

**remove (40)**



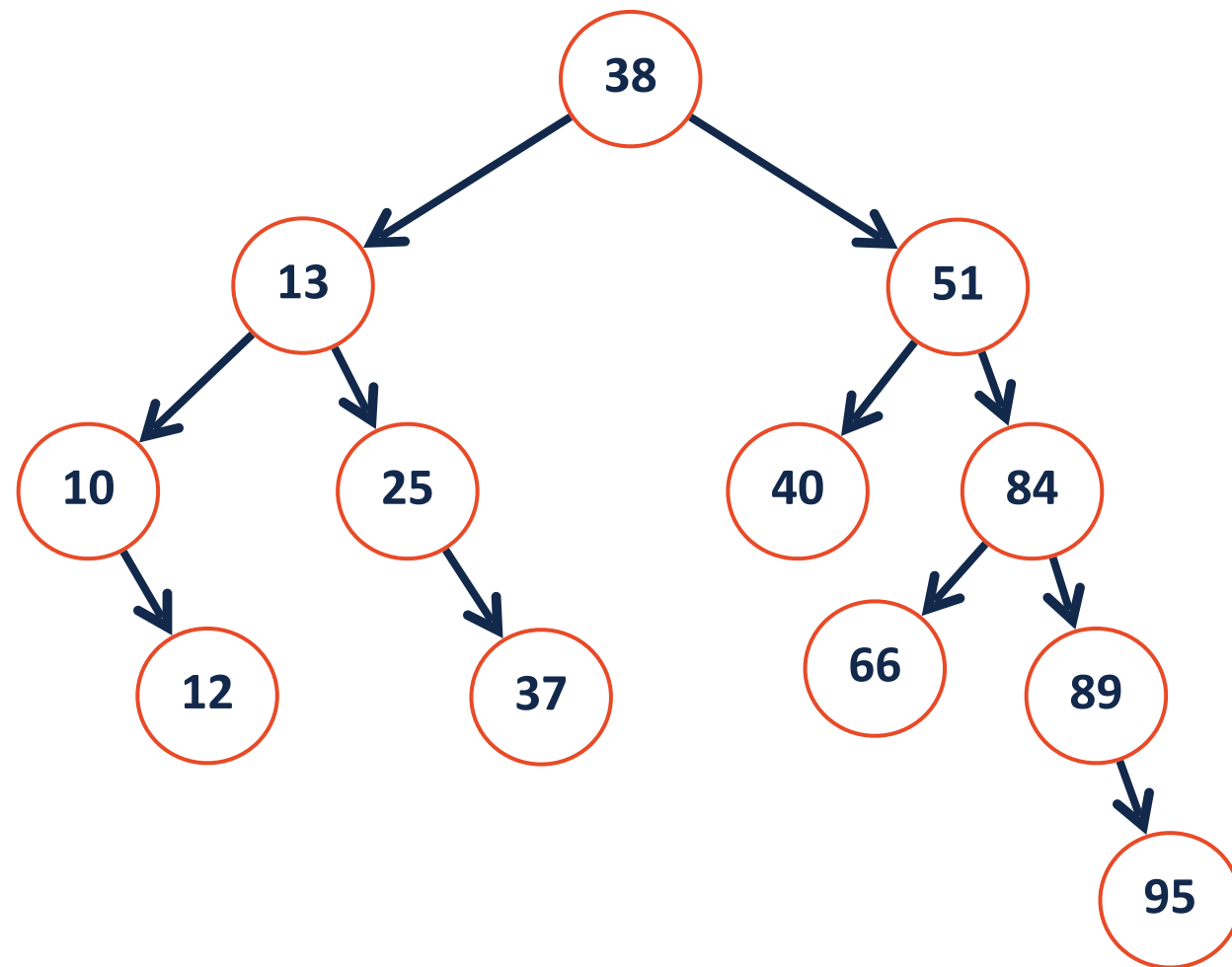
# BST Remove

**remove (25)**



# BST Remove

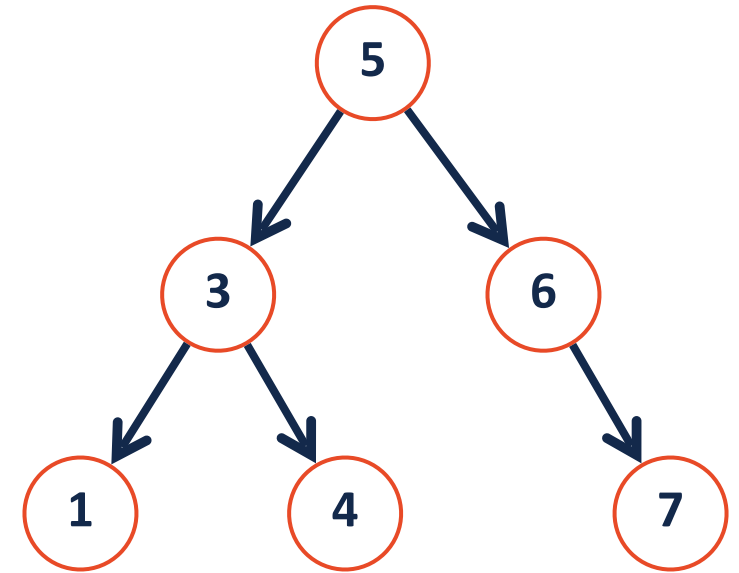
**remove (13)**



# BST Remove

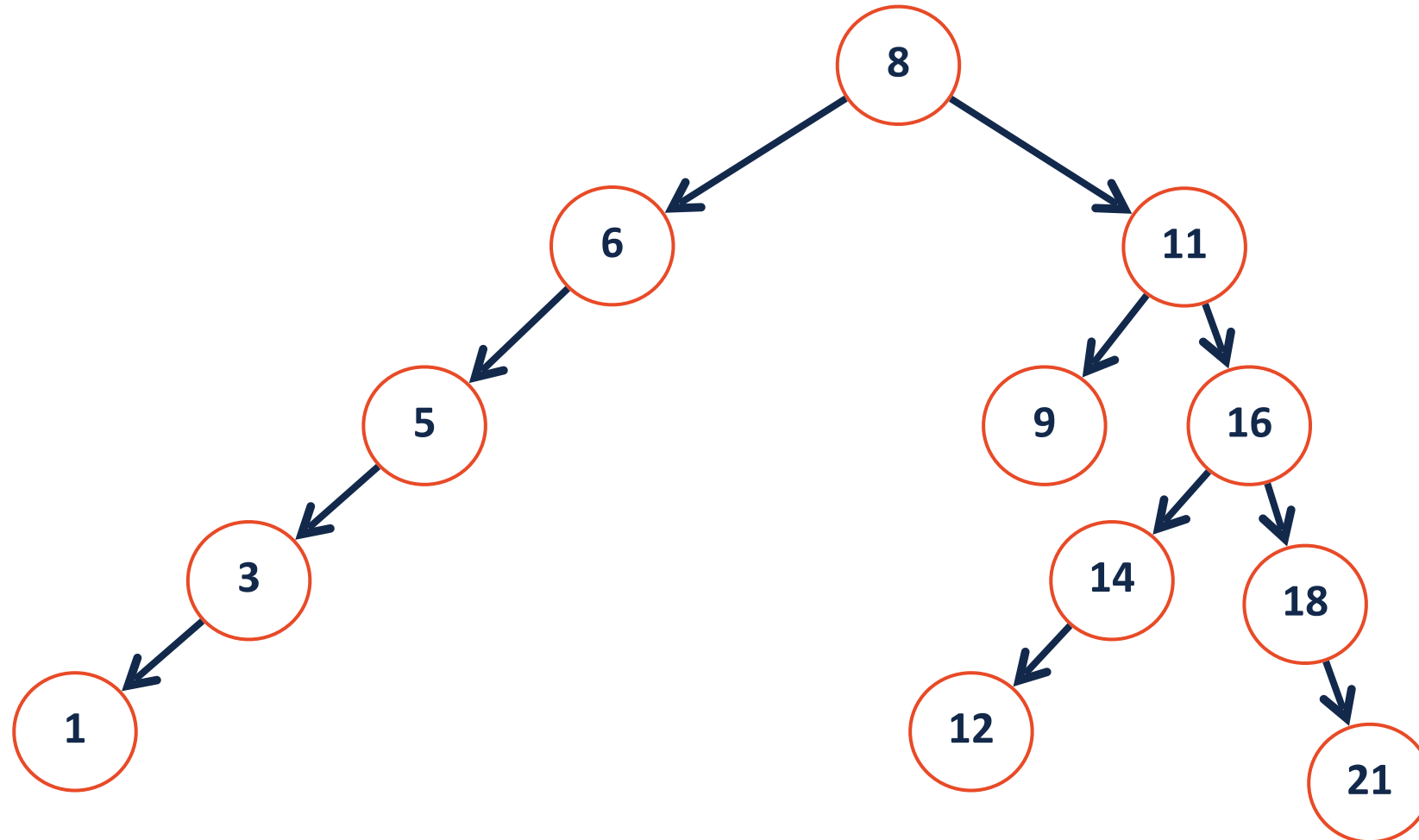


```
1 def remove(self, key):
2     self.root = self.remove_helper(self.root, key)
3
4 def remove_helper(self, node, key):
5
6
7
8
9
10
11
12
13
14
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16
17
18
19
20
21
22
23
24
25
```



# BST Remove

What will the tree structure look like if we remove node 16 using IOS?

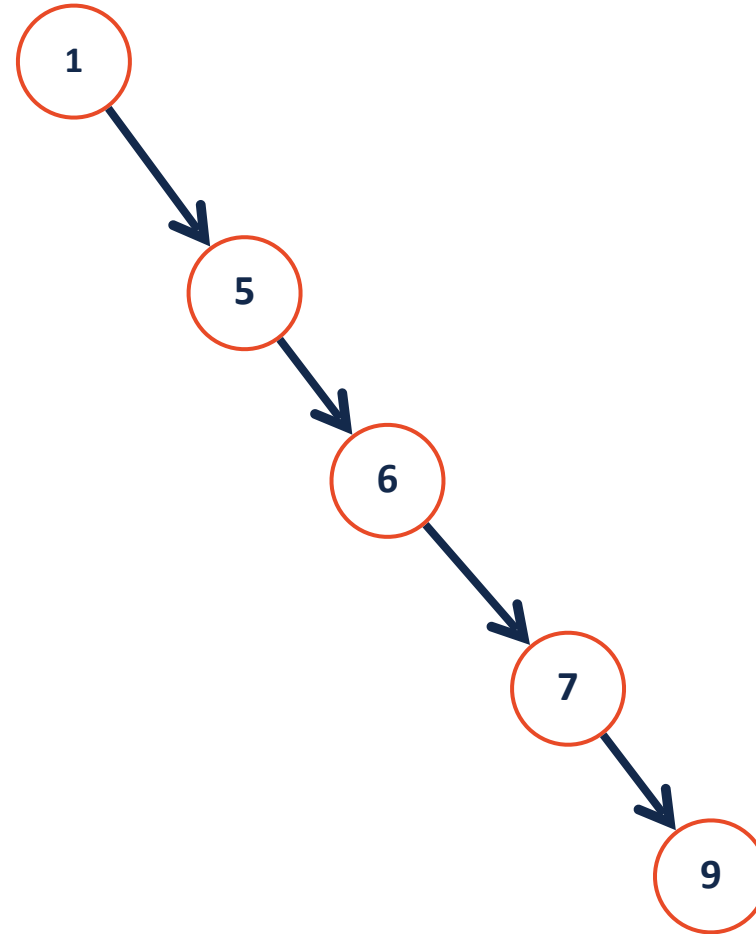
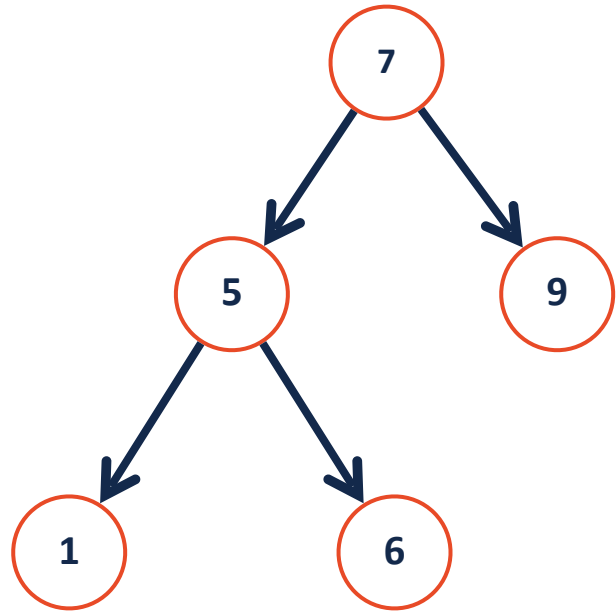


# BST Analysis – Running Time



Operation	BST Worst Case
find	
insert	
delete	
traverse	

# Limiting the height of a tree





# Option A: Correcting bad insert order

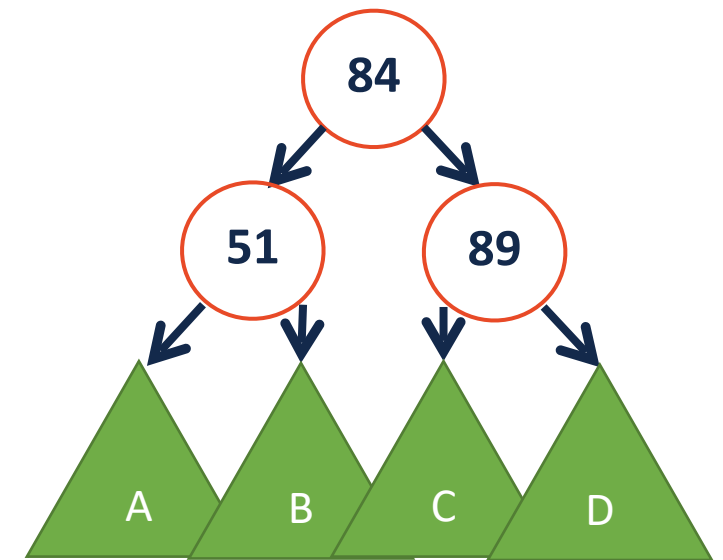
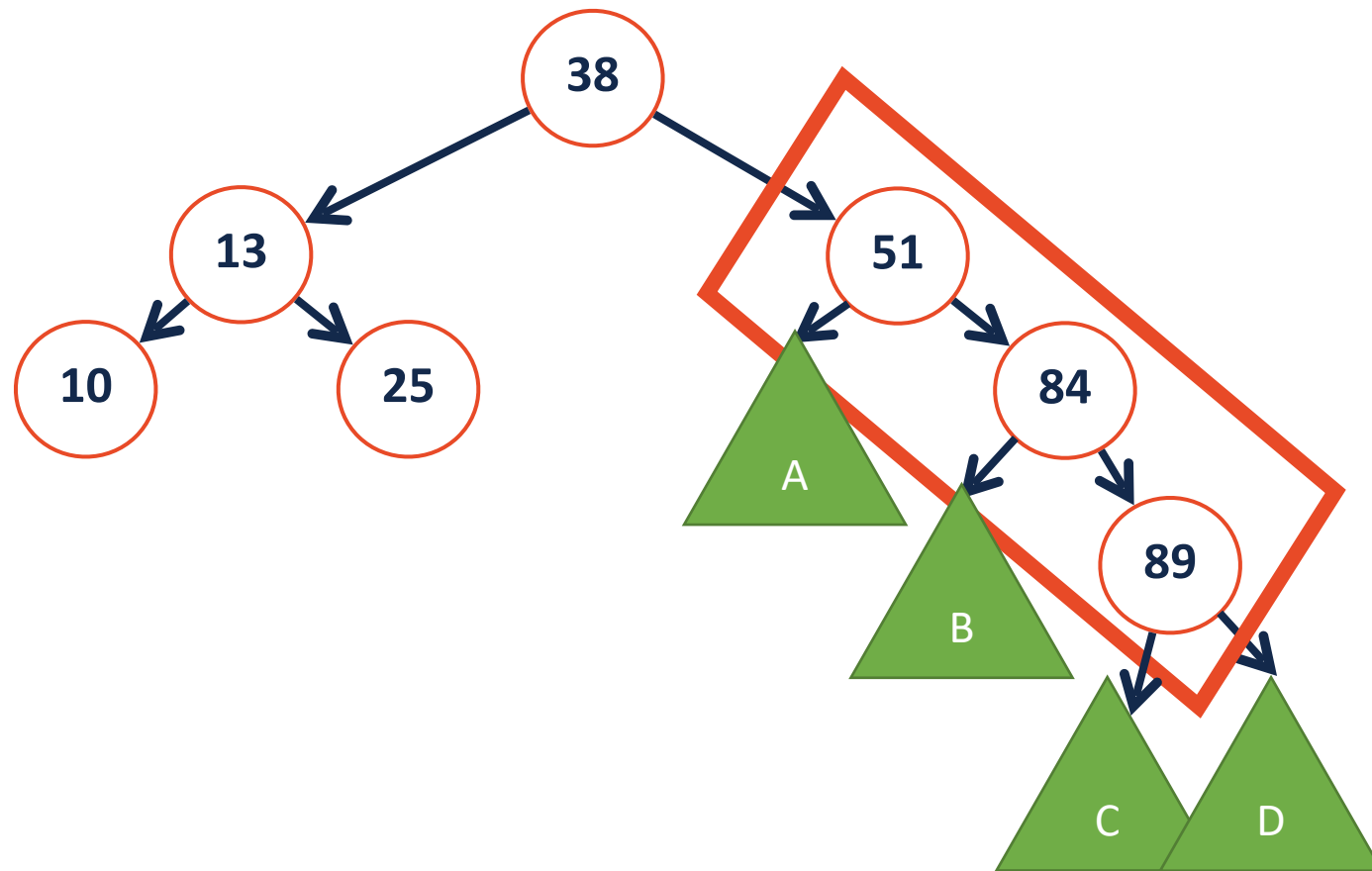
The height of a BST depends on the order in which the data was inserted

**Insert Order:** [1, 3, 2, 4, 5, 6, 7]

**Insert Order:** [4, 2, 3, 6, 7, 1, 5]

# AVL-Tree: A self-balancing binary search tree

Rather than fixing an insertion order, just correct the tree as needed!



# When would we use a tree?

Pretend for a moment that we always have an optimal BST.

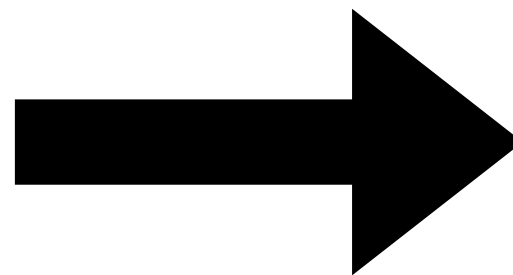
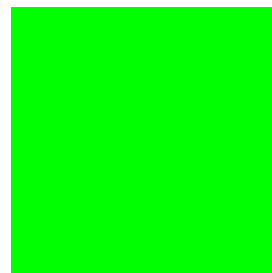
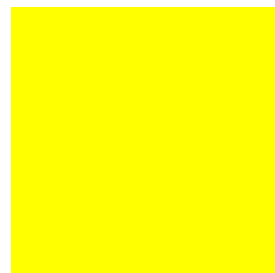
What is the running time of **find**?

What is the running time of **insert**?

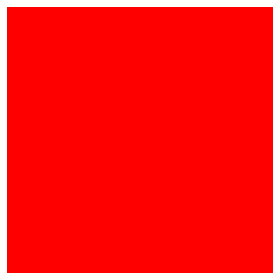
What is the running time of **remove**?

Is there a data structure with a *better* running time for all of these?

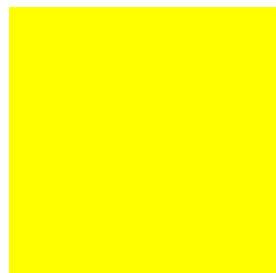
# Real World Use Case: Nearest neighbor search



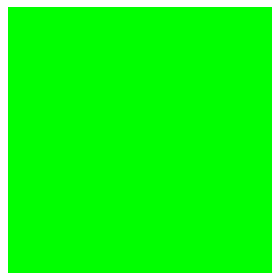
# Real World Use Case: Nearest neighbor search



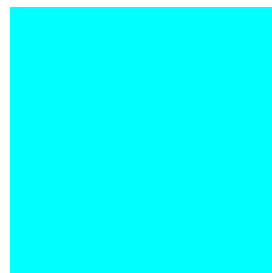
(255, 0, 0)



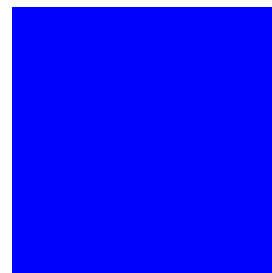
(255, 255, 0)



(0, 255, 0)



(0, 255, 255)



(0, 0, 255)



(255, 0, 255)



[[ 45 218 0], [223 147 243], [116 57 223], [187 9 9], [238 208 236]]

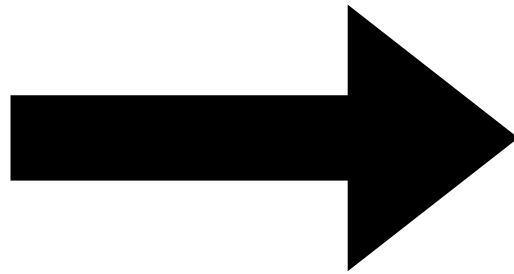
[[216 190 15], [193 64 80], [184 35 215], [ 95 152 180], [128 36 41]]

[[101 128 53], [224 122 191], [237 212 74], [ 35 98 227], [214 66 167]]

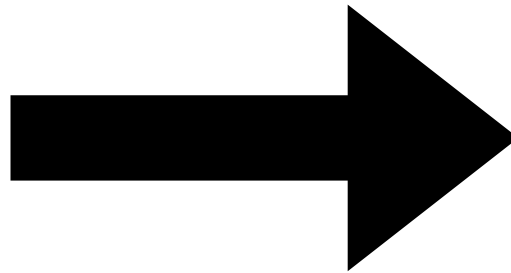
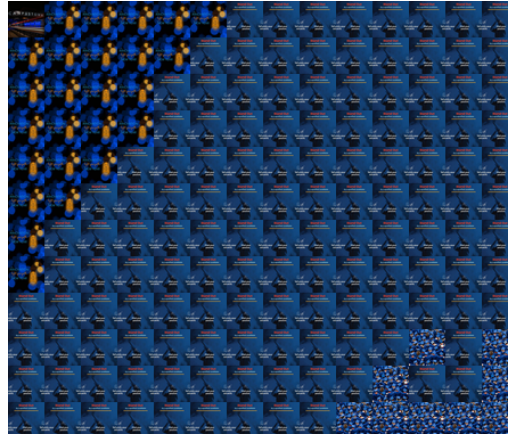
[[188 3 211], [217 142 33], [210 229 167], [208 57 22], [ 3 213 235]]

[[ 11 172 37], [225 191 57], [184 130 34], [136 33 51], [ 26 220 67]]

# Real World Use Case: Nearest neighbor search



# Real World Use Case: Nearest neighbor search



# Real World Use Case: Nearest neighbor search

Given an input image, how can we find the closest match from a collection collection of other images?