

DSC40B:  
Theoretical Foundations of Data  
Science II

Lecture 8: *Binary search tree*

Instructor: Yusu Wang

# (Dynamic) Set operations

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- ▶ Imagine you are maintaining a database indexed by some keys (real values), and you hope to support the following operations:
  - ▶ Search
  - ▶ Maximum
  - ▶ Minimum
  - ▶ Successor
  - ▶ Predecessor
  
  - ▶ Insert
  - ▶ Delete
  - ▶ Extract-Max
  - ▶ Increase-key



# (Dynamic) Set operations

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- ▶ Imagine you are maintaining a database indexed by some keys (real values), and you hope to support the following operations:

- ▶ Search
- ▶ Maximum
- ▶ Minimum
- ▶ Successor
- ▶ Predecessor

Static operations!

- ▶ Insert
- ▶ Delete
- ▶ Extract-Max
- ▶ Increase-key

Dynamic operations!

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# (Dynamic) Set operations

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- ▶ Imagine you are maintaining a database indexed by some keys (real values), and you hope to support the following operations:

First approach: sort the array of keys

- |                |                   |
|----------------|-------------------|
| ▶ Search       | ▶ $\Theta(\lg n)$ |
| ▶ Maximum      | ▶ $\Theta(1)$     |
| ▶ Minimum      | ▶ $\Theta(1)$     |
| ▶ Successor    | ▶ $\Theta(1)$     |
| ▶ Predecessor  | ▶ $\Theta(1)$     |
| ▶ Insert       | ▶                 |
| ▶ Delete       | ▶ $\Theta(n)$     |
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# (Dynamic) Set operations

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| ▶ Increase-key | ▶                 |

Using a sorted array can handle all static operations efficiently

How to have a good data structure so we can support all these operations efficiently?



# Today

---

- ▶ **Binary search tree**
  - ▶ support all the operations from previous slide
    - ▶ in time proportional to height of tree
- ▶ **(Review): how to implement key operations, and time complexity**
  - ▶ search, insert (and delete)
- ▶ **Extension to **balanced** binary search tree**
- ▶ **Select query: **augmenting** data structure**
  - ▶ median, order statistics



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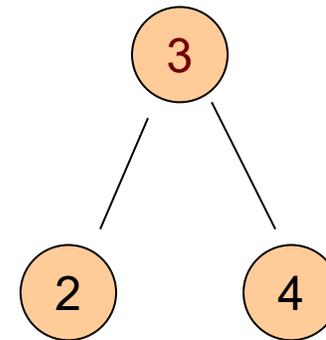
Part A:  
What is binary search tree?



# First: Binary tree

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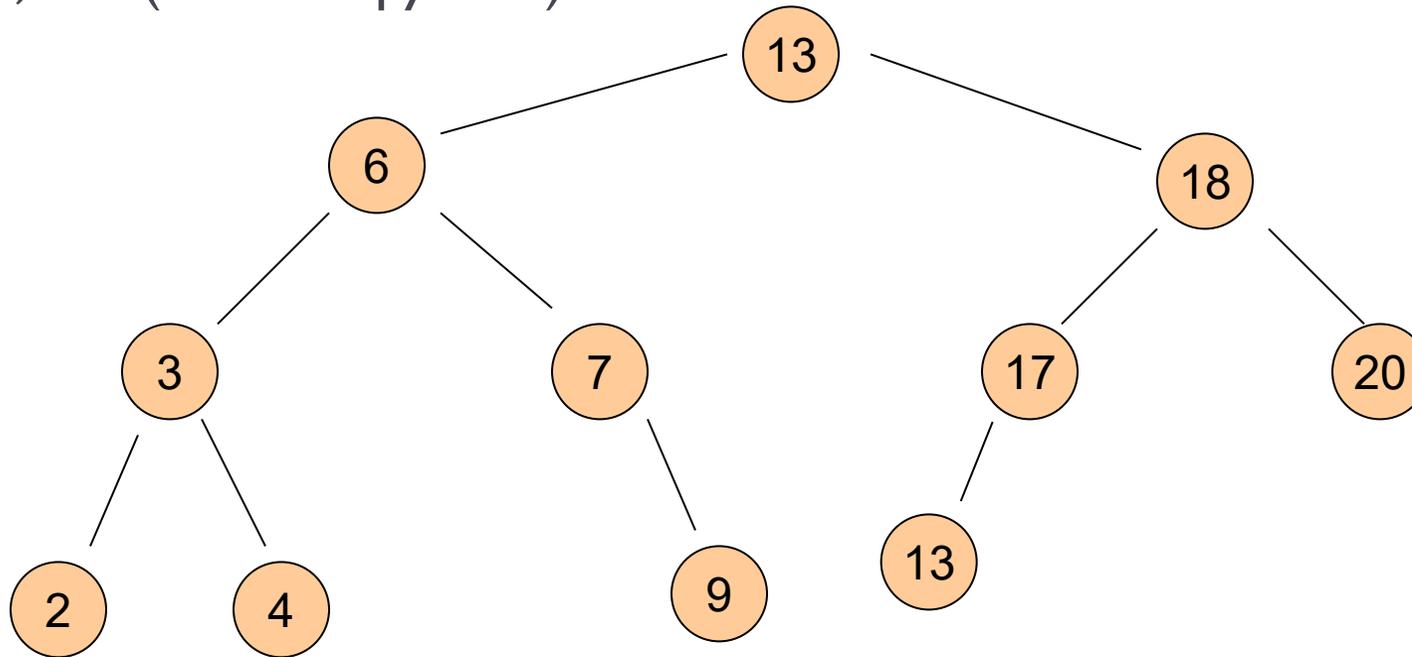
- ▶ A binary tree is a rooted tree
  - ▶ where each node has at most 2 children
- ▶ Represented by a linked data structure
- ▶ Each node contains at least fields:
  - ▶ *Key*
  - ▶ *Left*
  - ▶ *Right*
  - ▶ *Parent*



# Example

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- ▶ From root, following left pointers, we will visit
  - ▶ 13, 6, 3, 2, *Nil* (**None** in python)



# Create a single node tree in Python

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```
class Node:
    def __init__(self, key, parent=None):
        self.key = key
        self.parent = parent
        self.left = None
        self.right = None
```

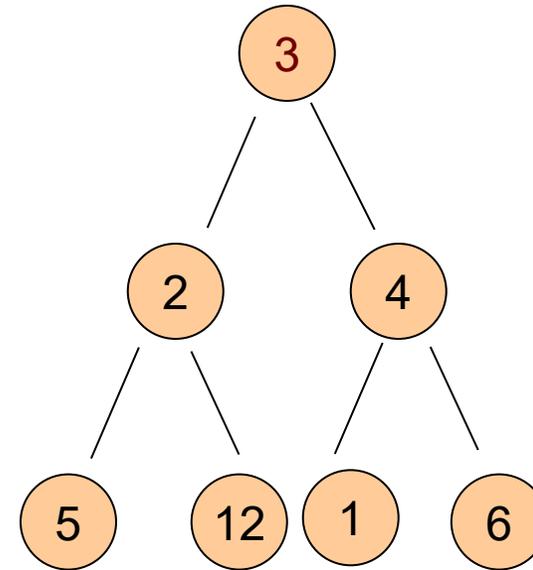
```
class BinarySearchTree:
    def __init__(self, root: Node):
        self.root = root
```



# Binary tree

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- ▶ A binary tree is a rooted tree where
  - ▶ each node has at most 2 children
- ▶ A node is the root of the tree
  - ▶ if its parent is Nil
- ▶ A node is a leaf
  - ▶ if both children are Nil
- ▶ Left sub-tree, right sub-tree



- ▶ A **complete binary tree** is a binary tree
    - ▶ where each node has two children other than leaves
    - ▶ and each level (except possibly last level) is filled, and all nodes are as left as possible.
- 



# Binary search tree (BST)

---

- ▶ **Binary-search-tree property**

- ▶ For any node  $x \in T$ ,

- ▶ if  $y$  is in the left subtree of  $x$ , then  $y.Key \leq x.Key$  and

- ▶ if  $y$  is in the right subtree of  $x$ , then  $y.Key \geq x.Key$

- ▶ A binary tree  $T$  is a **binary search tree (BST)** if

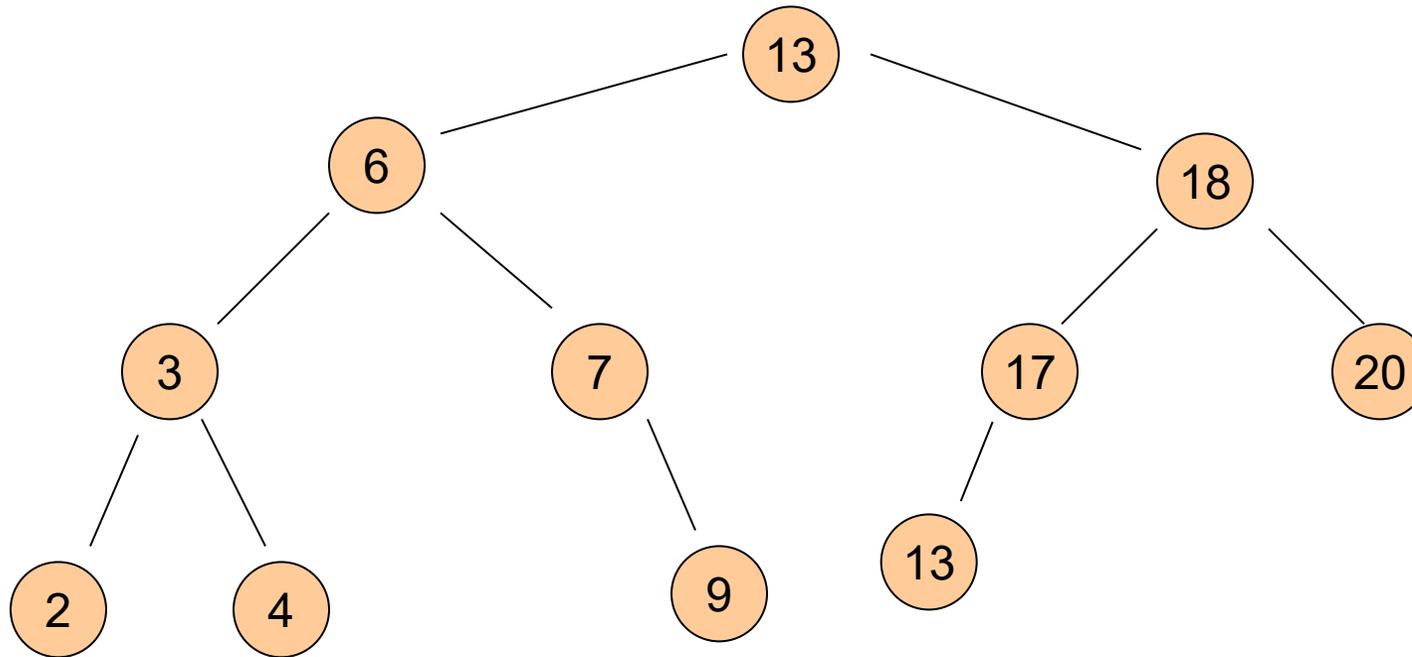
- ▶ it satisfies the binary search tree property



# Example

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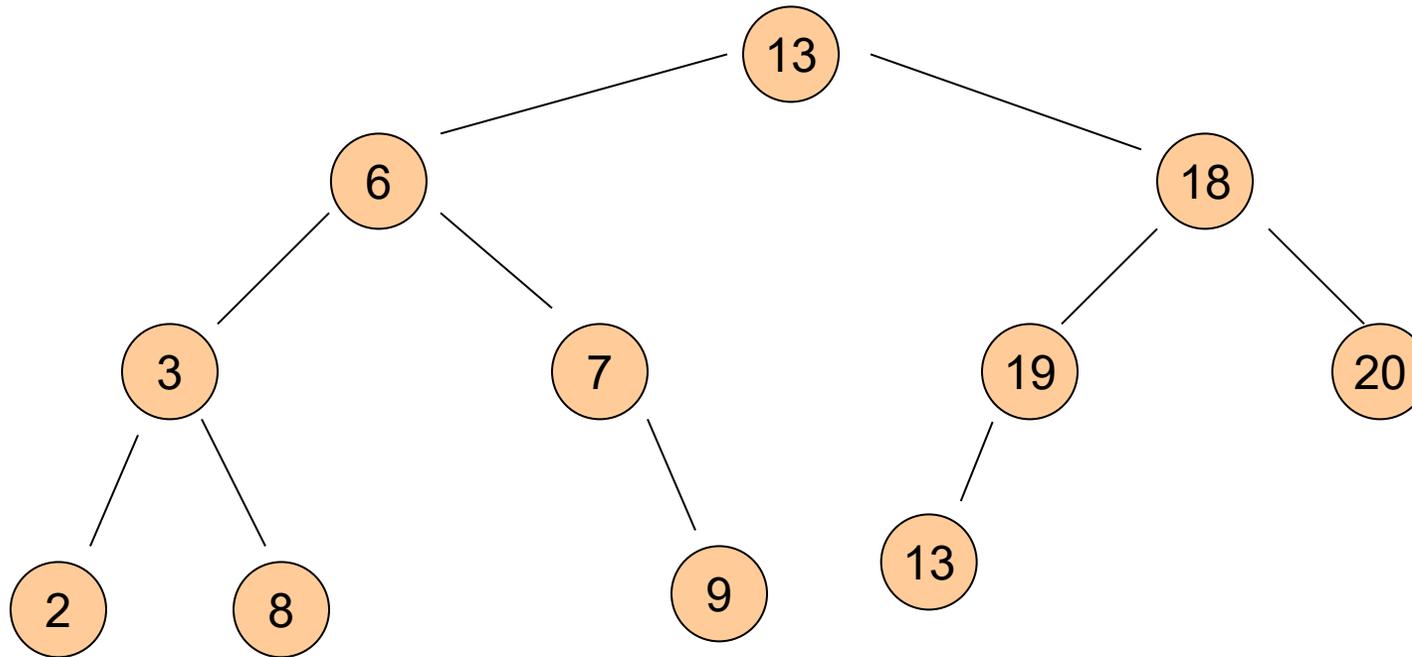
- ▶ A valid BST



# Example

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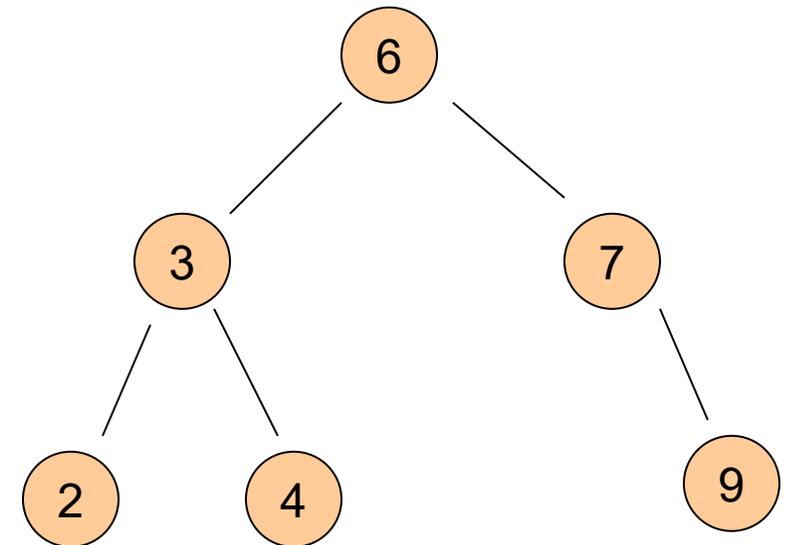
▶ ?



# Properties

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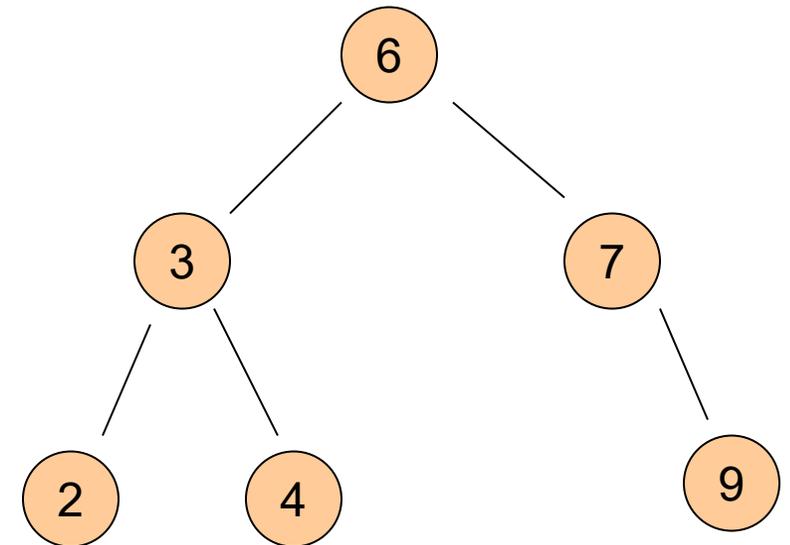
- ▶ **Given the same set of elements**
  - ▶ there are many possible BSTs over them
- ▶ **Given  $n$  nodes,**
  - ▶ Tallest possible BST tree has height  $h =$  \_\_\_\_\_
  - ▶ Shortest possible BST tree has height  $h =$  \_\_\_\_\_



# Properties

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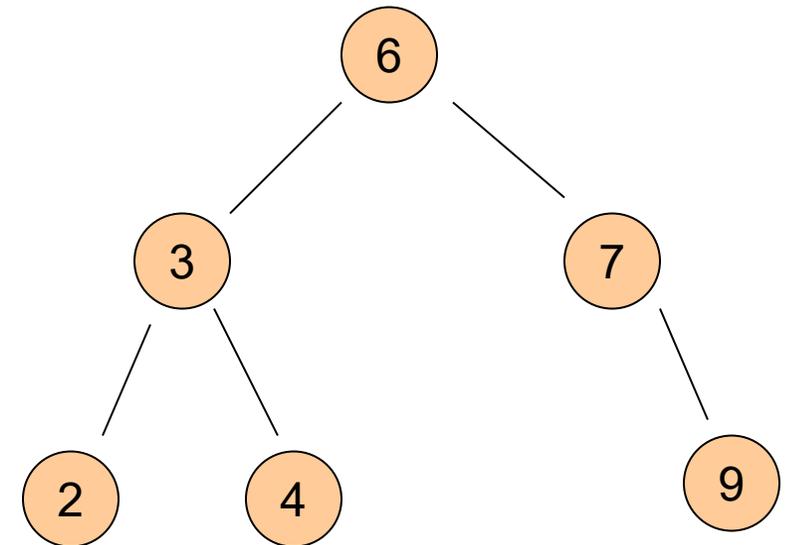
- ▶ Given the same set of elements
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- ▶ Given  $n$  nodes,
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  - ▶ Shortest possible BST tree has height  $h = \log_2 n = \Theta(\lg n)$



# Properties

---

- ▶ **Given the same set of elements**
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- ▶ **Given  $n$  nodes,**
  - ▶ Tallest possible BST tree has height  $h = \frac{n}{1}$
  - ▶ Shortest possible BST tree has height  $h = \log_2 n = \Theta(\lg n)$
- ▶ **Minimum?**
  - ▶ Does it have to be a leaf?
- ▶ **Maximum?**



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Part B:  
Operations in BST

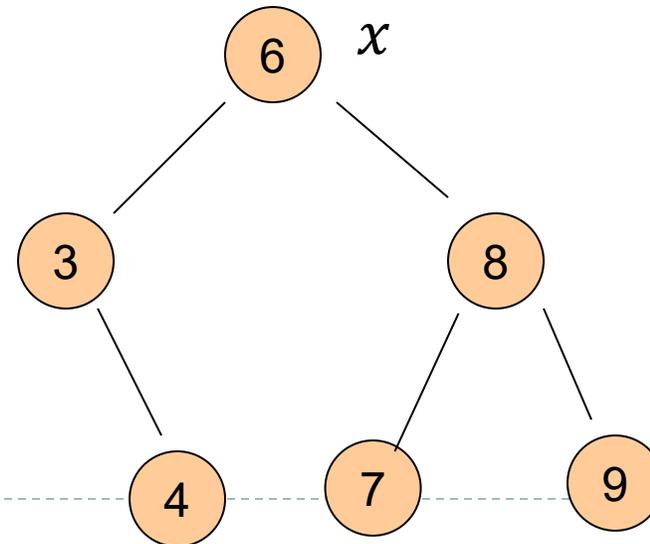
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# Search operation

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- ▶ A BST  $T$  with  $n$  nodes can be viewed as a way to store  $n$  keys in a smart way, so that queries among these keys become easy.
- ▶ Tree-search( $x, k$ )
  - ▶ Input: given a tree node  $x$  and a query key  $k$
  - ▶ Output: search whether  $k$  is in the tree rooted at  $x$ 
    - ▶ if it is in, return a node  $y$  s.t.  $y.key = k$
    - ▶ otherwise, returns  $NIL$



# Search operation

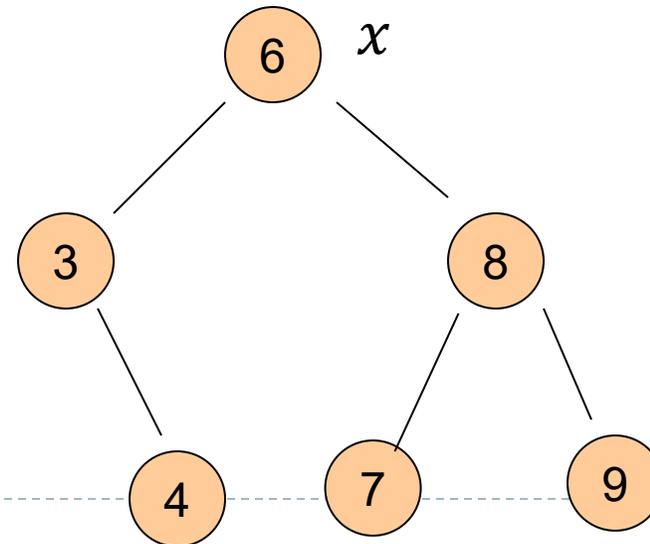
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  - ▶ Output: search whether  $k$  is in the tree rooted at  $x$ 
    - ▶ if it is in, return a node  $y$  s.t.  $y.key = k$
    - ▶ otherwise, returns  $NIL$

Tree-search( $x, 8$ )

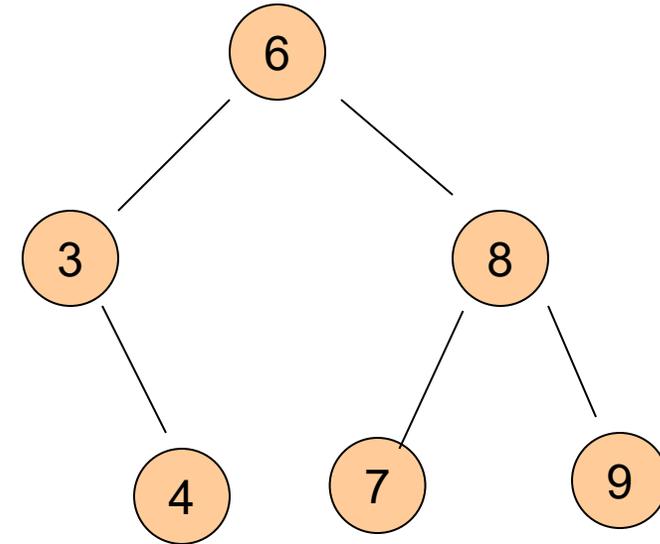
Tree-search( $x, 4$ )

Tree-search( $x, 5$ )



# Tree-search algorithm, recursive version

```
Tree-search (  $x, k$  )  
  if  $x$  is None:  
    return False  
  if  $x.key == k$   
    return  $x$   
  elif  $k < x.key$   
    return Tree-search(  $x.left, k$  )  
  else:  
    return Tree-search(  $x.right, k$  )
```

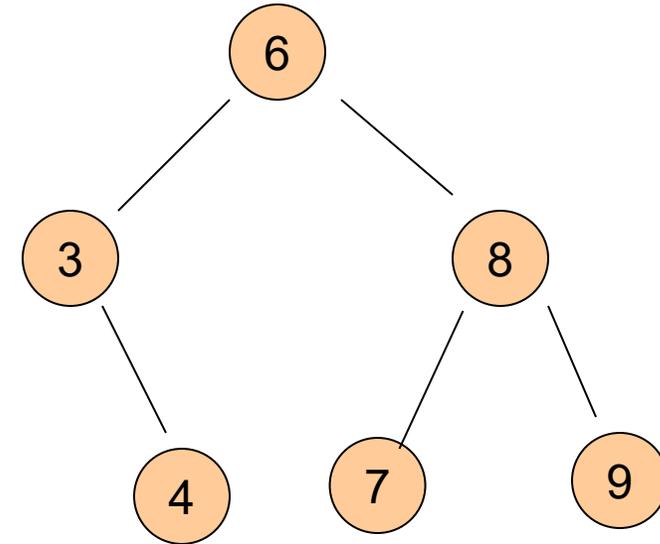


- ▶ Given an input tree  $T$  and a key  $k$ 
  - ▶ we will start by calling `Tree-search( $T.root, k$ )`



# Tree-search algorithm, recursive version

```
Tree-search (  $x, k$  )  
  if  $x$  is None:  
    return False  
  if  $x.key == k$   
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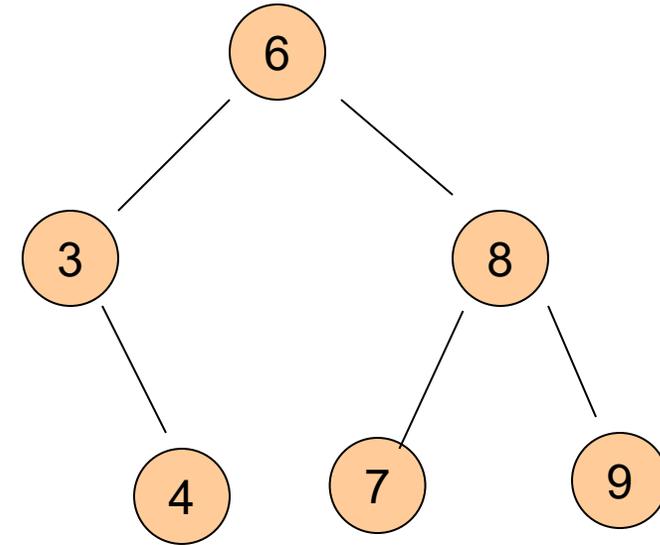
## ▶ Time complexity analysis

- ▶ let  $T(n)$  denote the worst case time complexity of procedure Tree-search() on any tree of  $n$  nodes



# Tree-search algorithm, recursive version

```
Tree-search (  $x, k$  )
  if  $x$  is None:
    return False
  if  $x.key == k$ 
    return  $x$ 
  elif  $k < x.key$ 
    return Tree-search(  $x.left, k$  )
  else:
    return Tree-search(  $x.right, k$  )
```



## ▶ Time complexity analysis

- ▶ Other than recursive call,  $\Theta(1)$  within each Tree-search call. Hence  $T(n)$  is proportional to the number of nodes  $x$  we will call Tree-search on. This implies  $T(n) = \Theta(\text{tree-height}) = O(n)$ .



# Tree-search: Pseudo-code of the iterative version

---

```
procedure IterativeTreeSearch( $x, K$ )  
1 while ( $x = \text{NIL}$ ) and ( $K \neq x.\text{key}$ ) do  
2   |   if ( $K \leq x.\text{key}$ ) then  
3     |    $x \leftarrow x.\text{left};$   
4   |   else  
5     |    $x \leftarrow x.\text{right};$   
6   |   end  
7 end  
8 return ( $x$ );
```

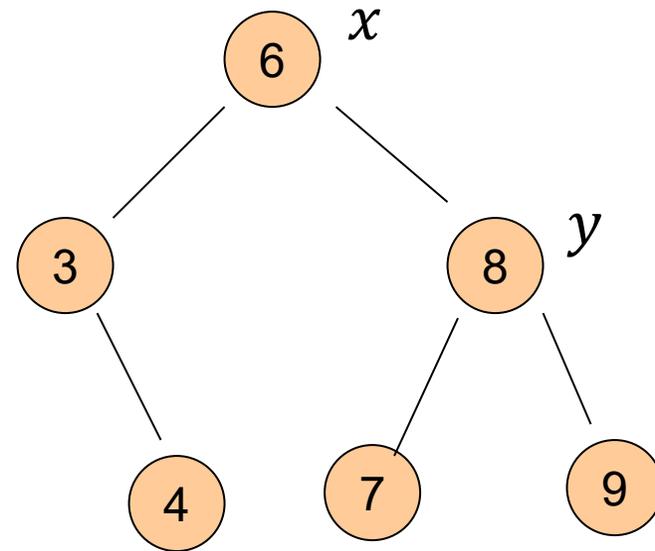


# Minimum / Maximum

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## ▶ Tree-minimum( $x$ )

- ▶ Input: a node  $x$  of a BST  $T$
- ▶ Output: return the node containing minimum key in the subtree rooted at  $x$



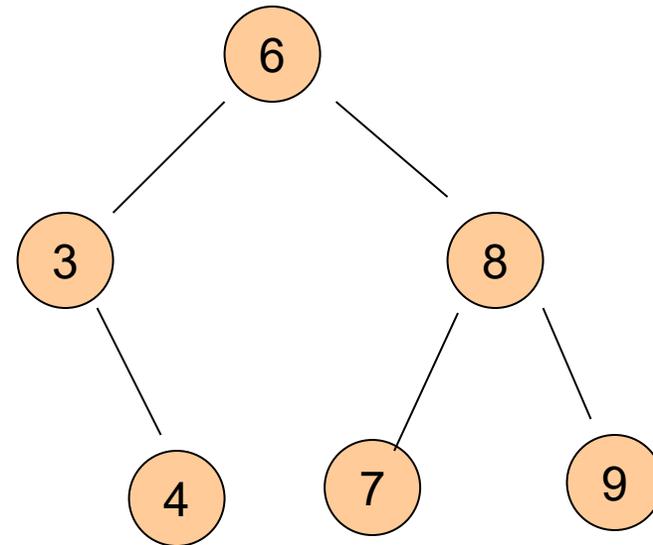
# Minimum / Maximum

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- ▶ Input: a node  $x$  of a BST  $T$
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```
Tree-minimum( $x$ )  
  while  $x$ .left is not None  
     $x = x$ .left;  
  return  $x$ ;
```



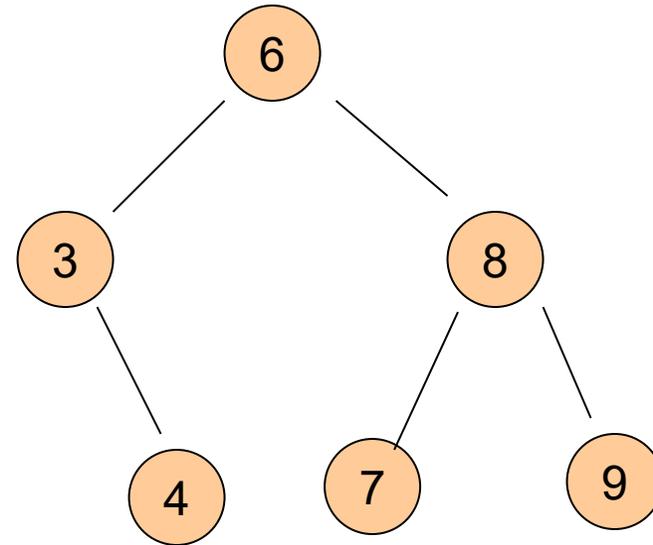
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Tree-minimum( $x$ )  
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    return  $x$ ;
```



## ▶ Time complexity

- ▶  $T(n) = \Theta(h)$  where  $h$  is height of input tree



# Minimum / Maximum

---

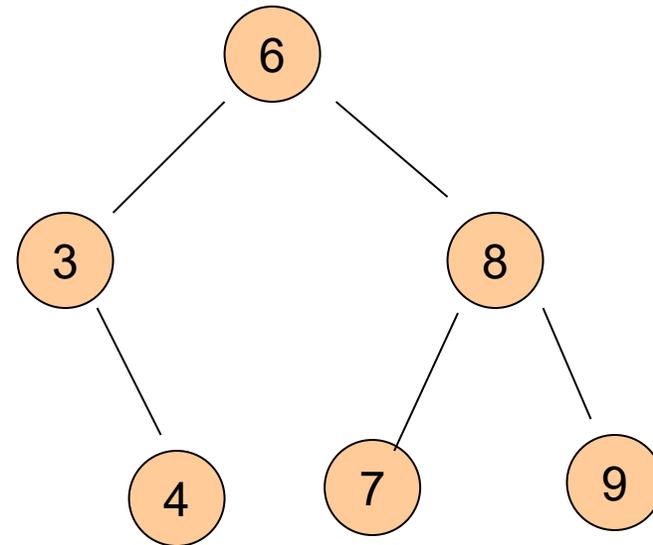
## ▶ Tree-maximum( $x$ )

- ▶ Input: a node  $x$  of a BST  $T$
- ▶ Output: return the node containing a maximum key in the subtree rooted at  $x$

```
Tree-maximum( $x$ )  
    while  $x$ .right is not None  
         $x = x$ .right;  
    return  $x$ ;
```

## ▶ Time complexity

- ▶  $T(n) = \Theta(h)$  where  $h$  is height of input tree



# Tree-insert

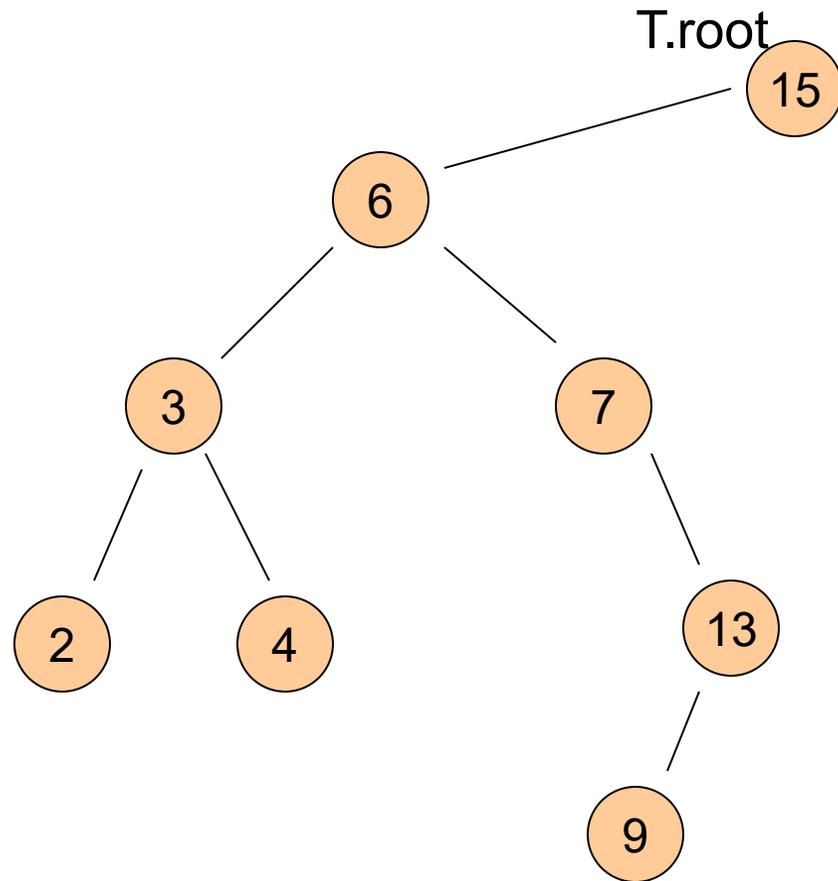
---

- ▶ **Tree-insert( $x, k$ )**
  - ▶ Input: a BST tree node  $x$  and a key  $k$
  - ▶ Output: insert  $k$  to the tree rooted at  $x$  such that the resulting tree is still a binary search tree



# Examples

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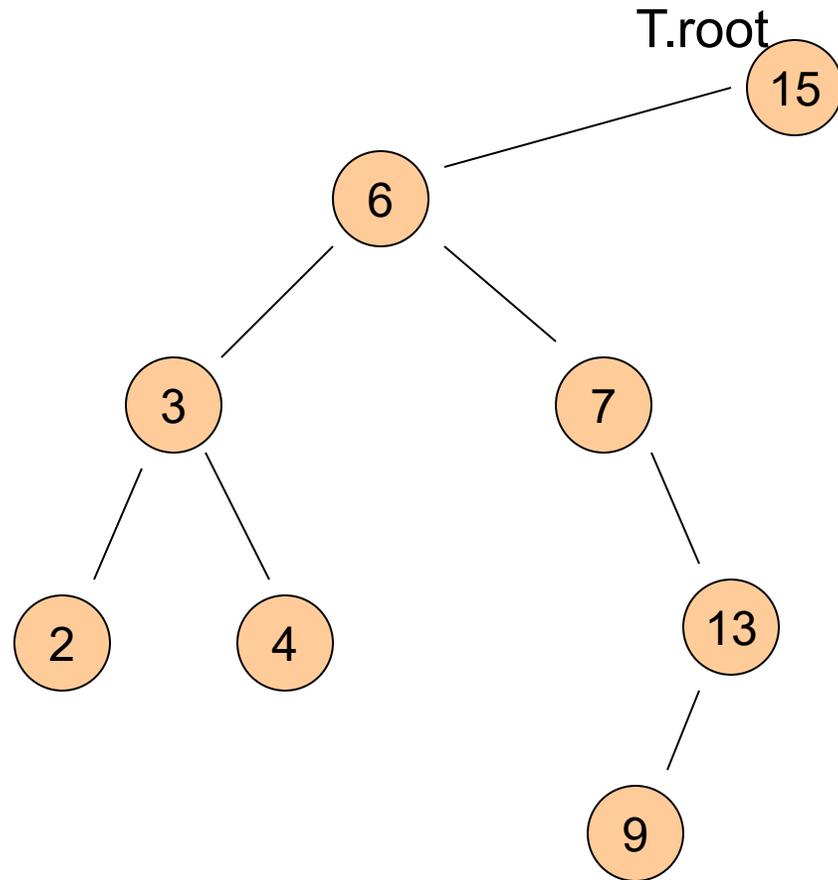
Tree-Insert(T.root, 8)

Tree-Insert(T.root, 6.5)



# Examples

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Tree-Insert(T.root, 8)

Tree-Insert(T.root, 6.5)

Use tree-search !



# Pseudo-code of Tree-insert

---

Tree-insert( $T, k$ )

$y = \text{Nil}; x = T.\text{root}$

$z.\text{key} = k; z.\text{left} = \text{Nil}; z.\text{right} = \text{Nil}$

while ( $x \neq \text{Nil}$ ) do

$y = x$

    if ( $z.\text{key} < x.\text{key}$ )

        then  $x = x.\text{left}$

        else  $x = x.\text{right}$

$z.\text{parent} = y$

if ( $y = \text{Nil}$ ) then  $T.\text{root} = z$

else if ( $z.\text{key} < y.\text{key}$ )

    then  $y.\text{left} = z$

    else  $y.\text{right} = z$



# Tree-insert

## Tree-insert( $T, k$ )

```
 $y = \text{Nil}; x = T.\text{root}$   
 $z.\text{key} = k; z.\text{left} = \text{Nil}; z.\text{right} = \text{Nil}$   
while ( $x \neq \text{Nil}$ ) do  
     $y = x$   
    if ( $z.\text{key} < x.\text{key}$ )  
        then  $x = x.\text{left}$   
        else  $x = x.\text{right}$ 
```

```
 $z.\text{parent} = y$   
if ( $y = \text{Nil}$ ) then  $T.\text{root} = z$ 
```

```
else if ( $z.\text{key} < y.\text{key}$ )  
    then  $y.\text{left} = z$   
    else  $y.\text{right} = z$ 
```

- $z$  is the new node to be inserted
- Locate potential parent  $y$  of  $z$ .

- Set up  $z$  as appropriate child of  $y$

# Tree-insert

```
Tree-insert( $T, k$ )
```

```
   $y = \text{Nil}; x = T.\text{root}$ 
```

```
   $z.\text{key} = k; z.\text{left} = \text{Nil}; z.\text{right} = \text{Nil}$ 
```

```
  while ( $x \neq \text{Nil}$ ) do
```

```
     $y = x$ 
```

```
    if ( $z.\text{key} < x.\text{key}$ )
```

```
      then  $x = x.\text{left}$ 
```

```
      else  $x = x.\text{right}$ 
```

```
   $z.\text{parent} = y$ 
```

```
  if ( $y = \text{Nil}$ ) then  $T.\text{root} = z$ 
```

```
  else if ( $z.\text{key} < y.\text{key}$ )
```

```
    then  $y.\text{left} = z$ 
```

```
    else  $y.\text{right} = z$ 
```

▶ Time complexity

▶  $T(n) = \Theta(h)$ , where  $h$  is height of input tree

# Tree Insert Python Code

---

```
def insert(self, new_key):
    # assume new_key is unique
    current_node = self.root
    parent = None

    # find place to insert the new node
    while current_node is not None:
        parent = current_node
        if current_node.key < new_key:
            current_node = current_node.right
        else: # current_node.key > new_key
            current_node = current_node.left

    # create the new node
    new_node = Node(key=new_key, parent=parent)

    # if parent is None, this is the root. Otherwise, update the
    # parent's left or right child as appropriate
    if parent is None:
        self.root = new_node
    elif parent.key < new_key:
        parent.right = new_node
    else:
        parent.left = new_node
```



# Summary: BST is good for both static and dynamic operations

---

- ▶ Suppose  $n$  input keys are already stored in a BST of height  $h$

Time complexity

▶ Search	$\Theta(h)$
▶ Maximum	$\Theta(h)$
▶ Minimum	$\Theta(h)$
▶ Successor	$\Theta(h)$
▶ Predecessor	$\Theta(h)$
▶ Insert	$\Theta(h)$
▶ Delete	$\Theta(h)$
▶ Extract-Max	$\Theta(h)$
▶ Increase-key	$\Theta(h)$

- However, performance depending on height!
- Height  $h = O(n)$  and  $h = \Omega(\lg n)$

- To have good performance, we want to keep the tree height low!



# Summary: BST is good for both static and dynamic operations

---

- ▶ Suppose  $n$  input keys are already stored in a BST of height  $h$

	Time complexity
--	-----------------

- |                |             |
|----------------|-------------|
| ▶ Search       | $\Theta(h)$ |
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| ▶ Minimum      | $\Theta(h)$ |
| ▶ Successor    | $\Theta(h)$ |
| ▶ Predecessor  | $\Theta(h)$ |
| ▶ Insert       | $\Theta(h)$ |
| ▶ Delete       | $\Theta(h)$ |
| ▶ Extract-Max  | $\Theta(h)$ |
| ▶ Increase-key | $\Theta(h)$ |



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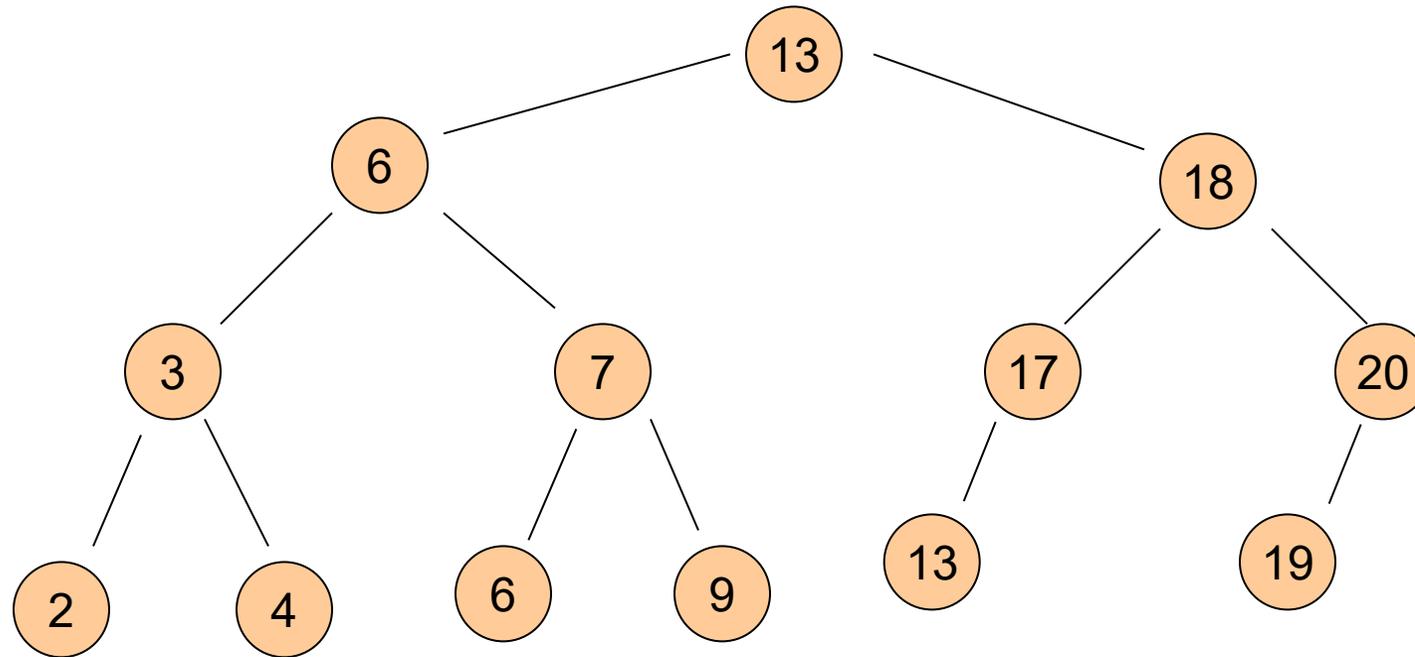
Part C:  
Balanced binary search tree

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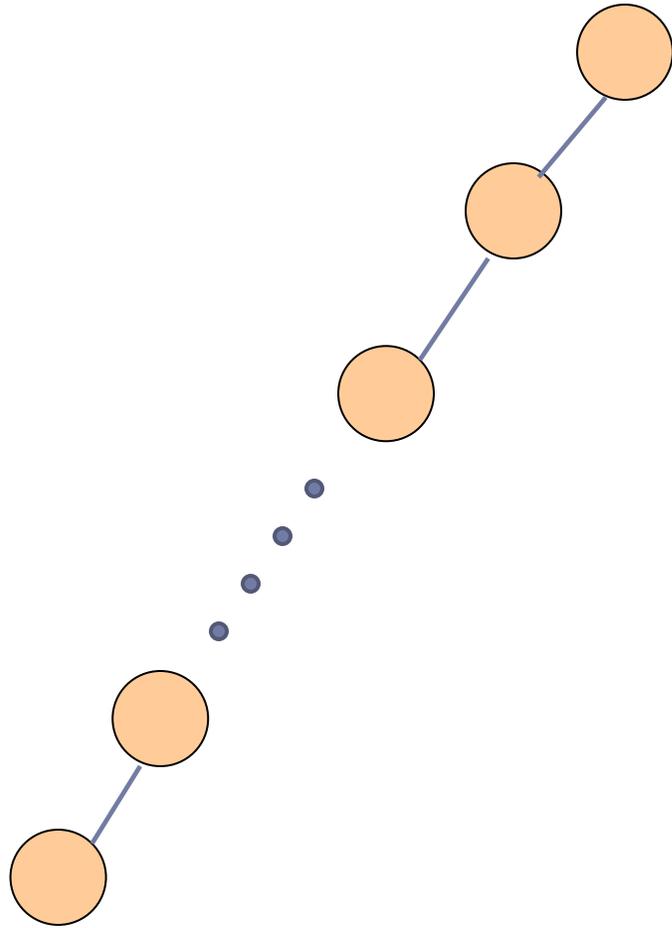
# Good tree

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# Bad Tree

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# Balanced binary search tree

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- ▶ It turns out that there are ways to add extra conditions to binary search trees, so that their height is  $\Theta(\lg n)$ 
  - ▶ E.g, red-black tree, AVL tree, etc
- ▶ Once such a tree is created,
  - ▶ it can support search, minimum, maximum etc in  $\Theta(h) = \Theta(\lg n)$  time using the same algorithms described before
  - ▶ the extra work comes at handling dynamic operations: insertion, deletion, and so on. Re-balancing is needed
  - ▶ however, for standard balanced BSTs, all these operations can be handled in  $\Theta(\lg n)$  time.



# With balanced BST

---

- ▶ Suppose  $n$  input keys are already stored in a balanced BST

	Time complexity
--	-----------------

- |                |                 |
|----------------|-----------------|
| ▶ Search       | $\Theta(\lg n)$ |
| ▶ Maximum      | $\Theta(\lg n)$ |
| ▶ Minimum      | $\Theta(\lg n)$ |
| ▶ Successor    | $\Theta(\lg n)$ |
| ▶ Predecessor  | $\Theta(\lg n)$ |
| ▶ Insert       | $\Theta(\lg n)$ |
| ▶ Delete       | $\Theta(\lg n)$ |
| ▶ Extract-Max  | $\Theta(\lg n)$ |
| ▶ Increase-key | $\Theta(\lg n)$ |

- Height of tree will be  $\Theta(\lg n)$ , where  $n$  is number of nodes in the tree



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Part D:  
*Select* queries  
augmenting data structure

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- 
- ▶ What if we also want to perform Select operation
  - ▶ BST-Select (  $x, k$  ):
    - ▶ Given a list of records whose keys are stored in a tree rooted at  $x$ , return the node whose key has rank  $k$ .
  - ▶ We can use QuickSelect to do this in linear time.. Why are we not satisfied?



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  - ▶ What if we have to do it many times?
    - ▶ Sort it first

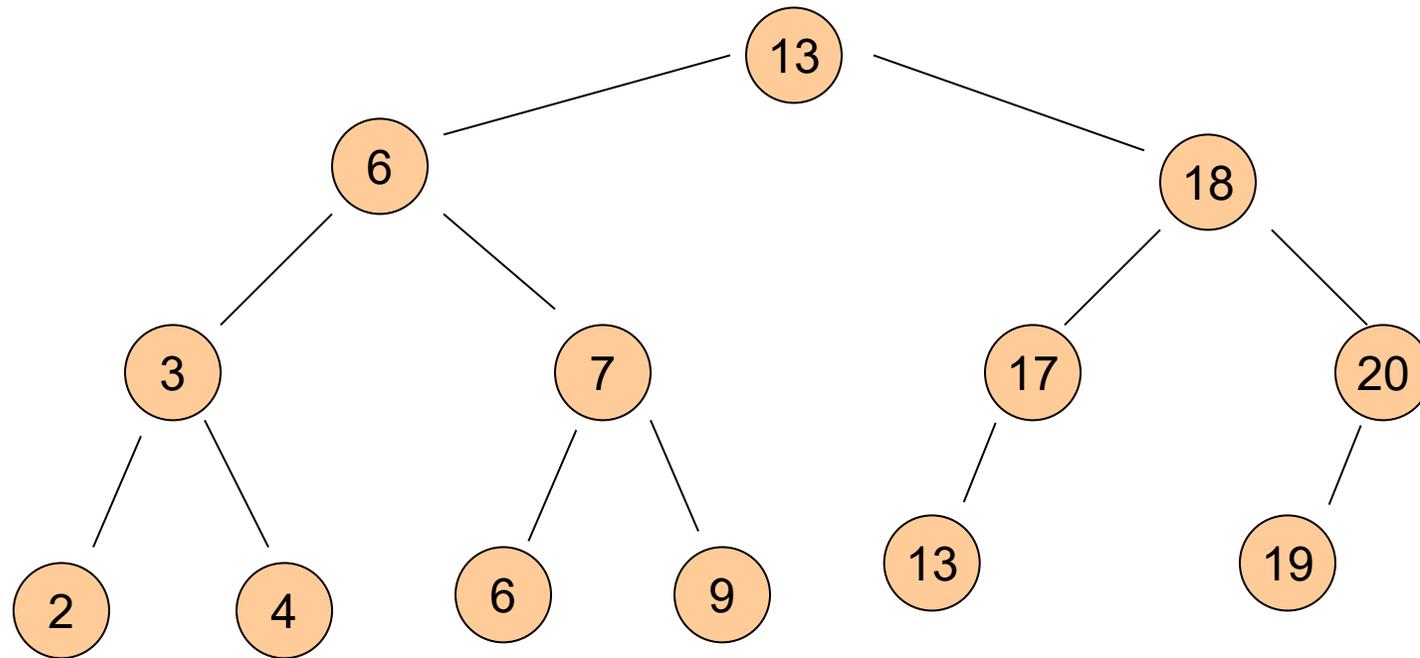


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  - ▶ We can use QuickSelect to do this in linear time.. Why are we not satisfied?
  - ▶ What if we have to do it many times?
    - ▶ Sort it first
  - ▶ But what if we also have **dynamic changes**?
    - ▶ Need a data structure that can support Select under dynamic changes



# BST

---



- 
- ▶ How to perform Select operation over BST?
  - ▶ BST-Select (  $x, k$  ):
    - ▶ Given a list of records whose keys are stored in a tree rooted at  $x$ , return the node whose key has rank  $k$ .
  - ▶ We can do linear search to find it. But can we do better?
  - ▶ Goal:
    - ▶ **Augment** the binary search tree data structure so as to support Select (  $x, k$  ) efficiently
- 



## In particular,

---

- ▶ **BST-Select (  $x, k$  )**
- ▶ **Goal:**
  - ▶ Augment the binary search tree data structure so as to support BST-Select (  $x, k$  ) efficiently
- ▶ **Ordinary binary search tree  $T$** 
  - ▶  $O(h)$  time for  $\text{BST-Select}(T.\text{root}, k)$  where  $h$  is height of tree  $T$
- ▶ **Using balanced search tree)**
  - ▶  $O(\lg n)$  time for  $\text{BST-Select}(T.\text{root}, k)$



# How do we augment a BST $T$ ?

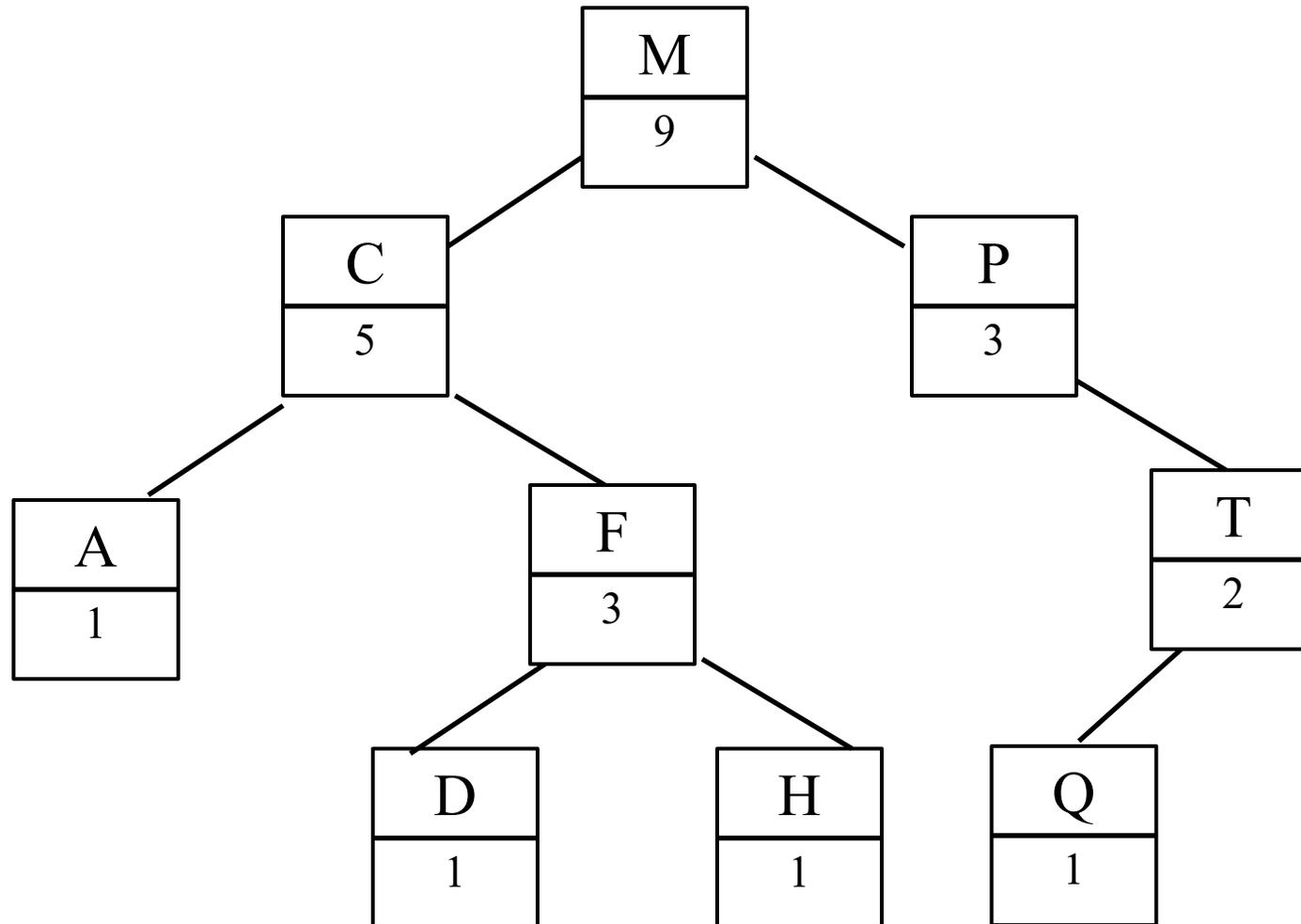
---

- ▶ At each node  $x$  of the tree  $T$ 
  - ▶ store  $x.size = \#$  nodes in the subtree rooted at  $x$ 
    - ▶ Include  $x$  itself
    - ▶ If a node (leaf) is NIL, its size is 0.
- ▶ Space of an augmented tree:
  - ▶  $\Theta(n)$



# An example

---



# How do we augment a BST $T$ ?

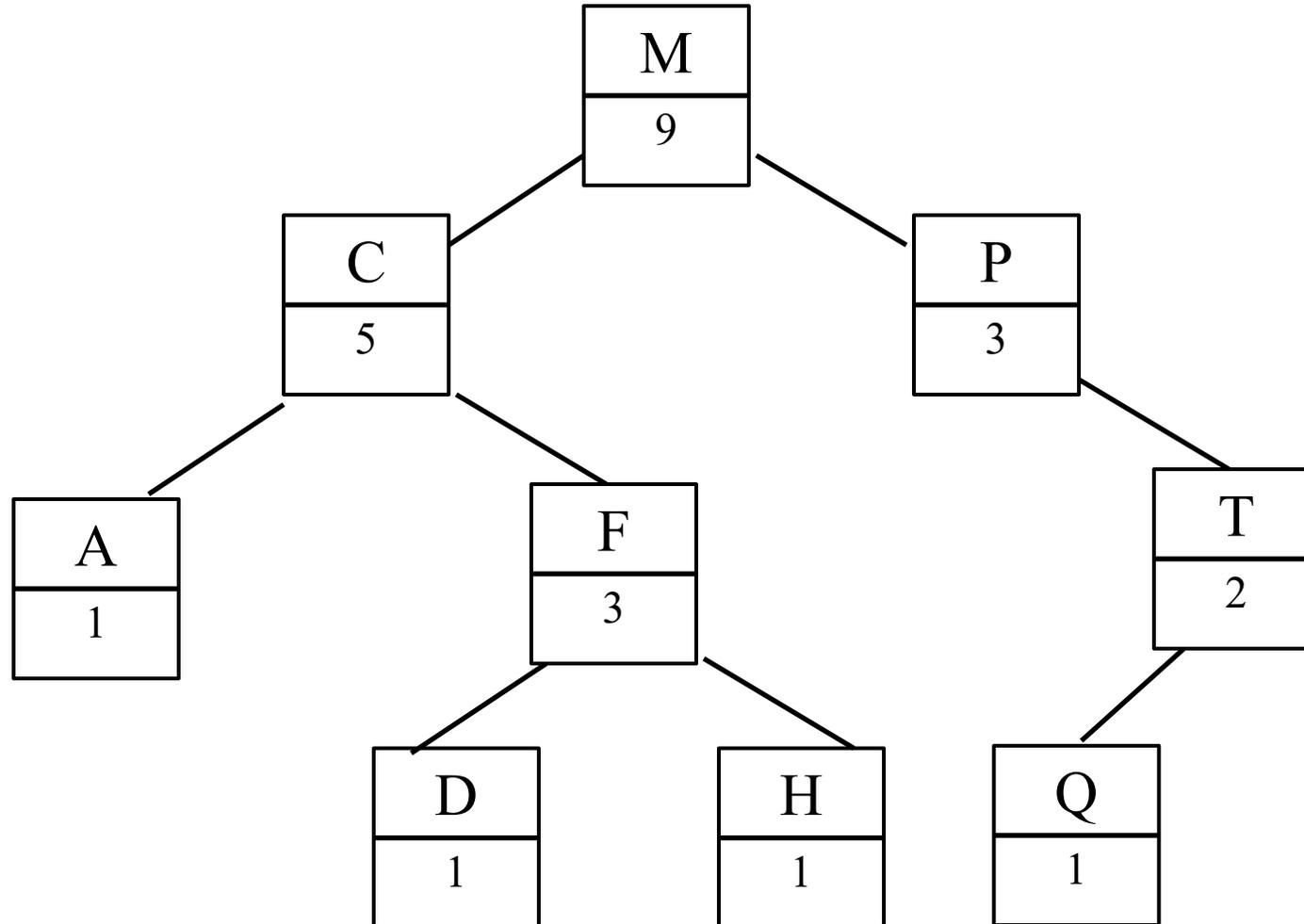
---

- ▶ At each node  $x$  of the tree  $T$ 
  - ▶ store  $x.size = \#$  nodes in the subtree rooted at  $x$ 
    - ▶ Include  $x$  itself
    - ▶ If a node (leaf) is NIL, its size is 0.
- ▶ Space of an augmented tree:
  - ▶  $\Theta(n)$
- ▶ Basic property:
  - ▶  $x.size = x.left.size + x.right.size + 1$



# How to set up size information ?

---



# How to setup size information?

---

▶ procedure *AugmentSize*( *treenode* *x* )

  If (*x* ≠ *NIL* ) then

*Lsize* = *AugmentSize*( *x.left* );

*Rsize* = *AugmentSize*( *x.right*);

*x.size* = *Lsize* + *Rsize* + 1;

    Return( *x.size* );

  end

  Return (0);



# How to setup size information?

---

▶ procedure *AugmentSize*( *treenode x* )

  If (*x* is not *Nil*) then

*Lsize* = *AugmentSize*( *x.left* );

*Rsize* = *AugmentSize*( *x.right*);

*x.size* = *Lsize* + *Rsize* + 1;

    Return( *x.size* );

  end

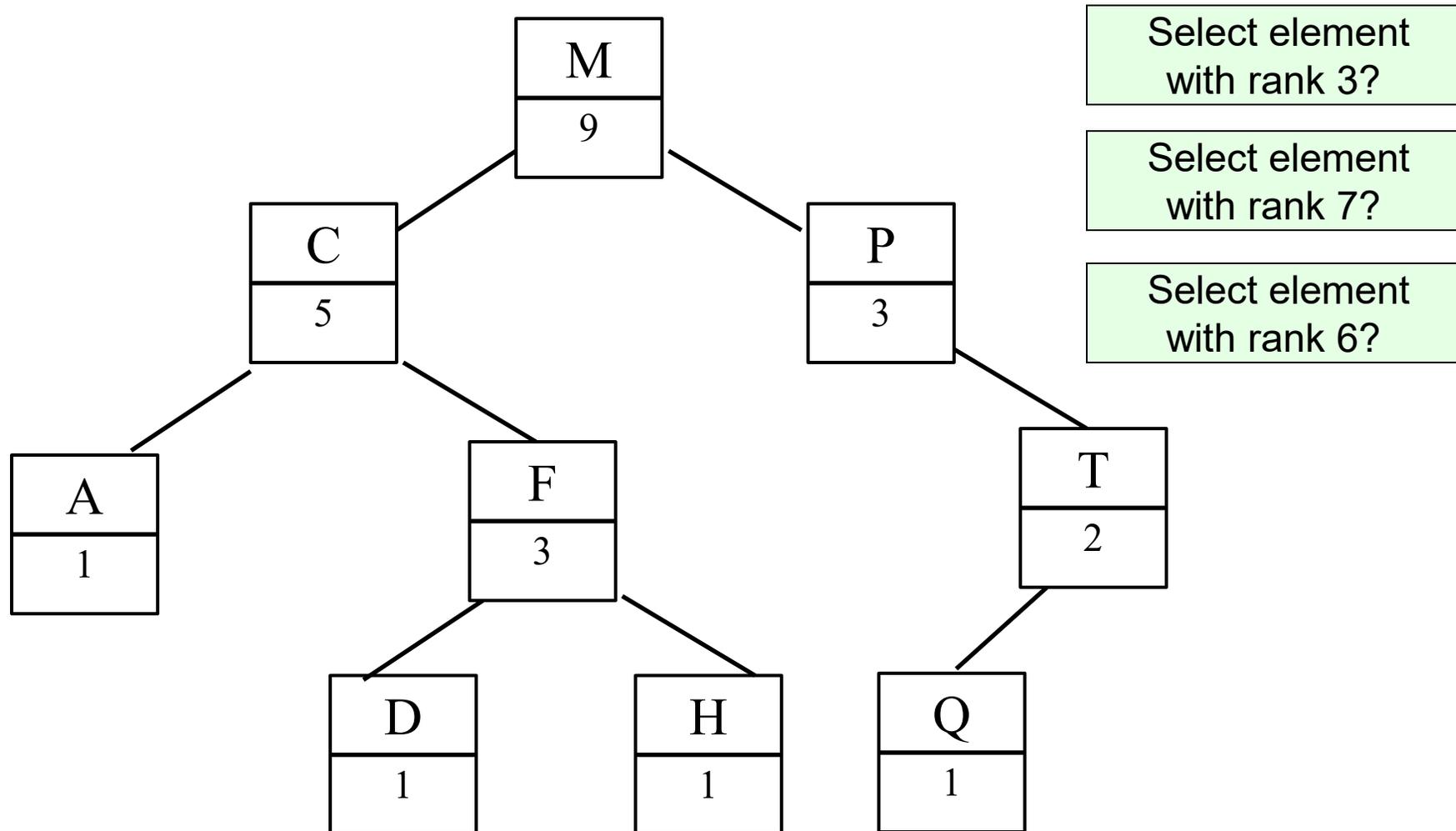
  Return (0);

Postorder traversal  
of the tree !

Time complexity for Augmentsize:  
 $\Theta(\text{size of tree})$



# How to perform select with aug-BST?



- 
- ▶ Let  $T$  be an augmented binary search tree
  - ▶  $BST\text{-}Select(x, k)$ :
    - ▶ Return the  $k$ -th smallest element in the subtree rooted at  $x$
    - ▶  $BST\text{-}Select(T.root, k)$  returns the  $k$ -th smallest elements in the entire tree.
  
  - ▶ Using ideas just described,  $BST\text{-}Select(x, k)$  can be implemented to have  $\Theta(\text{height of tree})$  time complexity
    - ▶ which is  $\Theta(\lg n)$  for a balanced binary tree.
    - ▶ See homework.
- 



# Are we done?

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- ▶ Need to maintain the augmented information under dynamic changes of the tree!
  - ▶ i.e, under insertions / deletions
  - ▶ in this case, just adjusting this size count as we update nodes, or under rotations, and it does not increase asymptotic time complexity of these operations
- ▶ Remark:
  - ▶ Select() in a sorted array can be done in  $\Theta(1)$  time.
  - ▶ However, an array does not support dynamic operations (insert/delete) efficiently. That's augmented BST is a better data structure in this case.



# Summary

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- ▶ Simple example of augmenting data structures
- ▶ In general, the augmented information can be quite complicated
  - ▶ Can be a separate data structure!
- ▶ Need to consider how to maintain such information under dynamic changes



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FIN

