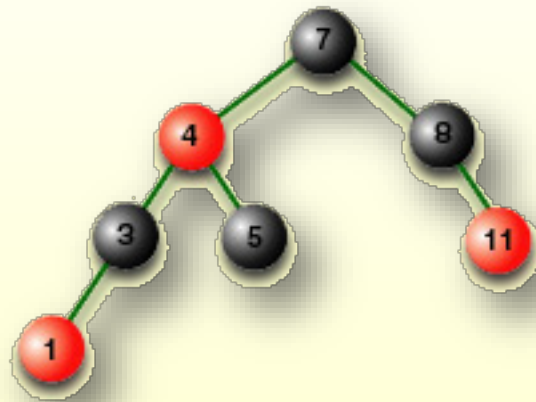


# CS161:

## Design and Analysis of Algorithms

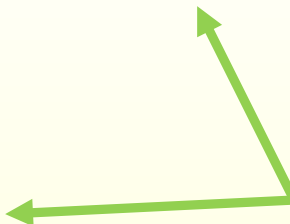


## Lecture 9

### Leonidas Guibas

(Guest lecturer: Mary Wootters)

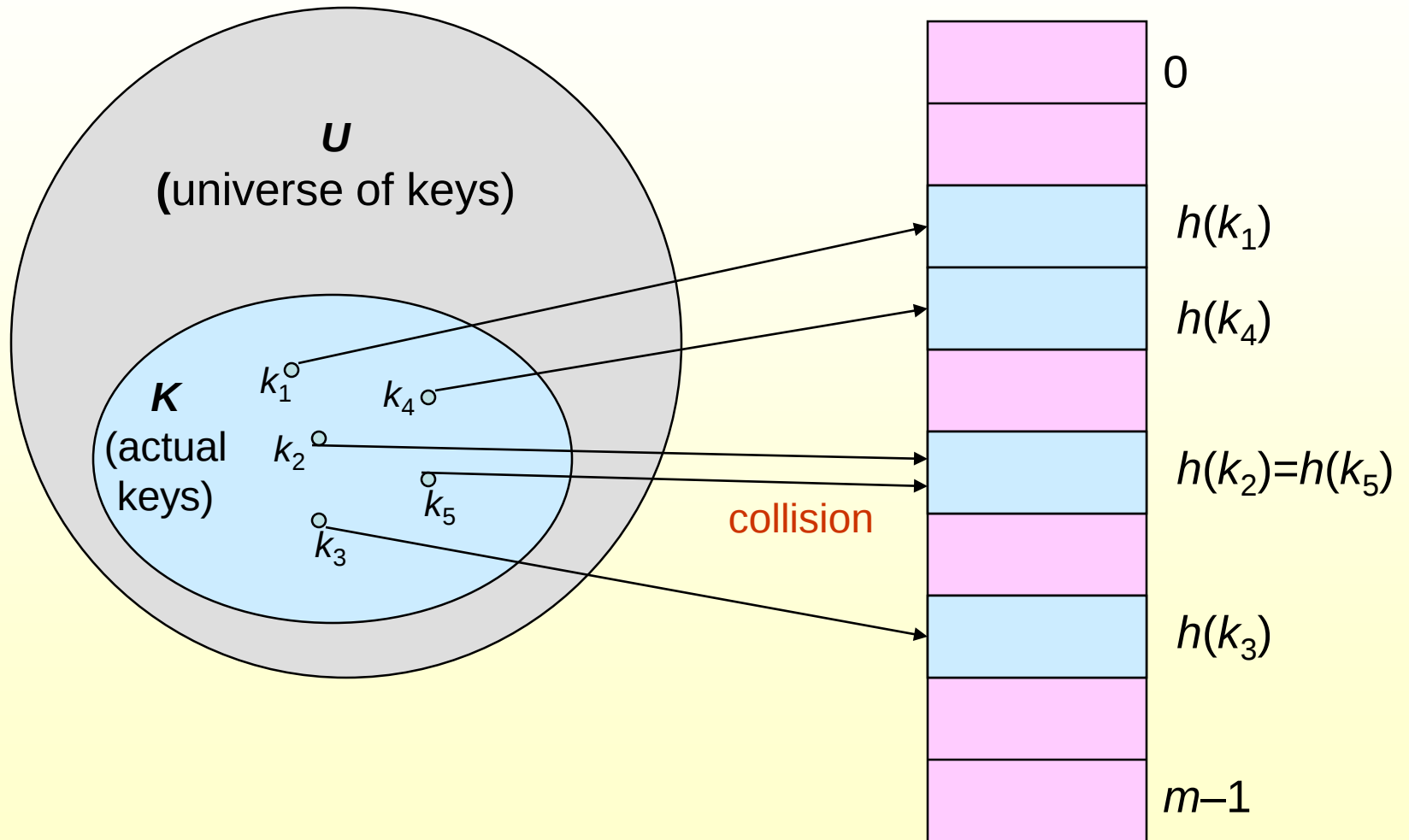
# Outline

- ◆ Review of last lecture: Hashing
  - ◆ Binary Search Trees
    - ◆ Traversals
    - ◆ Search/Insertion/Deletion
    - ◆ TreeSort
    - ◆ Expected height of a tree
- Dictionary Data Structures
- 

Slides modified from

- [www.cse.unr.edu/~bebis/CS477/](http://www.cse.unr.edu/~bebis/CS477/)
- [homes.ieu.edu.tr/cevrendilek/CE221\\_week\\_10\\_Chapter4\\_TreesBST.ppt](http://homes.ieu.edu.tr/cevrendilek/CE221_week_10_Chapter4_TreesBST.ppt)
- <http://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-046j>

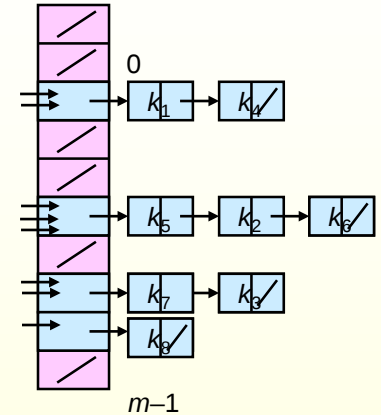
# Hashing



# Methods of Collision Resolution

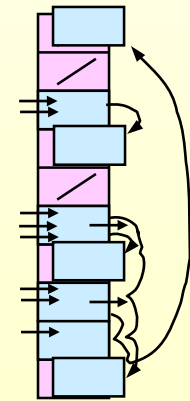
## ◆ Chaining:

- ◆ Store all elements that hash to the same slot in a linked list.
- ◆ Store a pointer to the head of the linked list in the hash table slot.



## ◆ Open Addressing:

- ◆ All elements stored in hash table itself.
- ◆ When collisions occur, use a systematic (consistent) procedure to store (and search for) elements in free slots of the table.



# Choosing A Hash Function

- ◆ Choosing the hash function well is crucial
  - ◆ Bad hash function puts all elements in same slot
  - ◆ A good hash function:
    - ◆ Should distribute keys uniformly into slots
    - ◆ Should not depend on patterns in the data
- ◆ Several ways to do this!

# Motivation for today

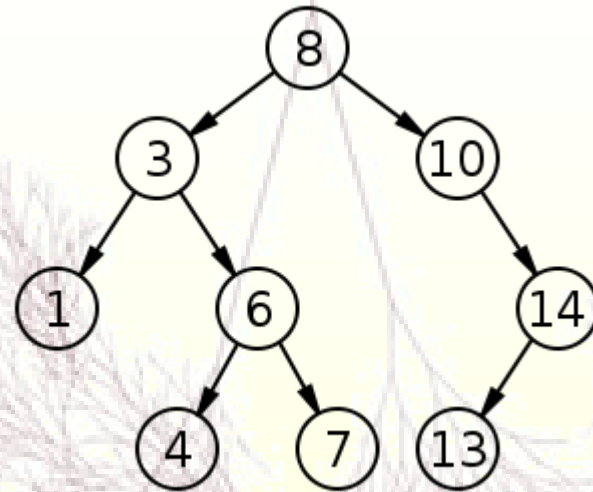
- ◆ Hash tables:

- ◆ Support many operations [SEARCH, INSERT, DELETE, etc...] in expected time  $O(1)$ .
- ◆ But if we pick a bad hash function it may be much worse!

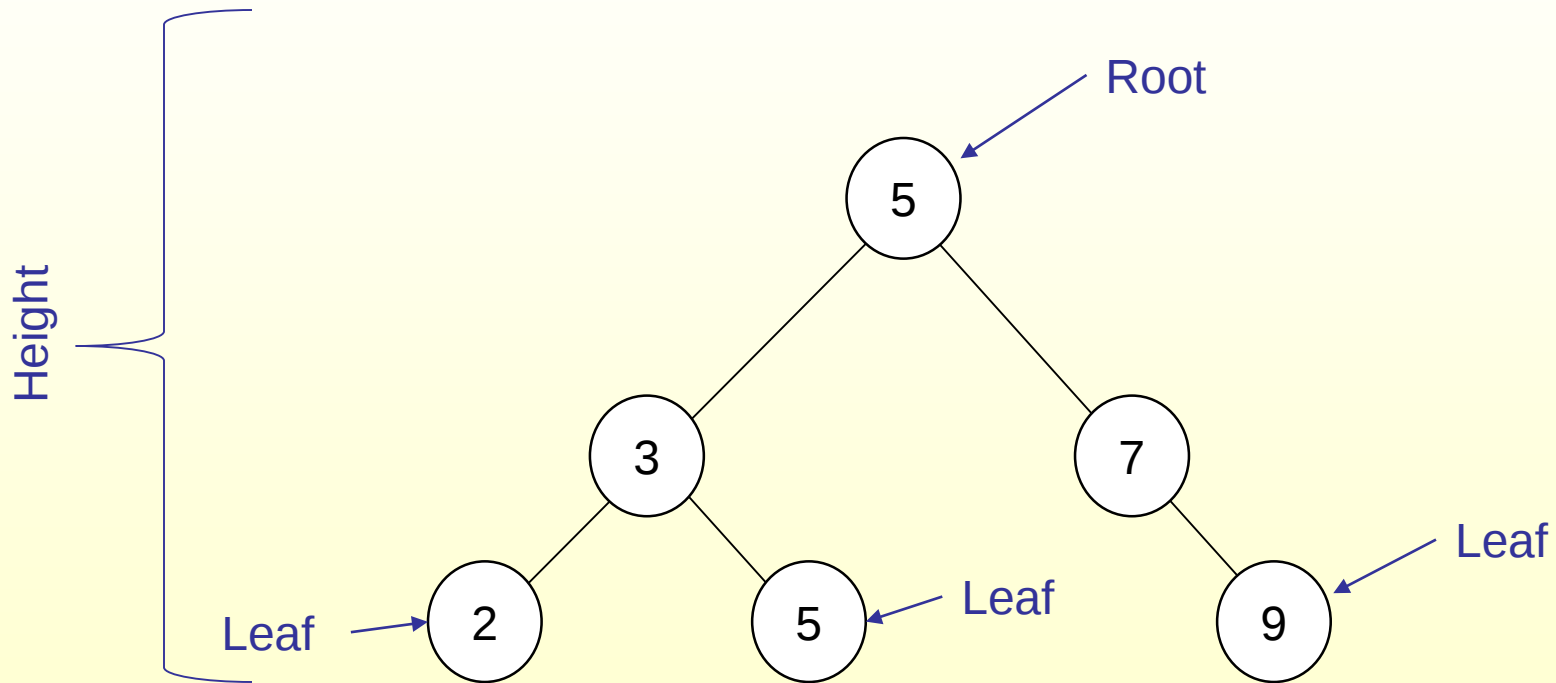
- ◆ Binary Search Trees:

- ◆ Support [SEARCH, INSERT, DELETE, etc...]
- ◆ Eventual goal:  $O(\lg(n))$  for all of them.
- ◆ We won't quite get there today, but we'll get a good start.

# BINARY SEARCH TREES

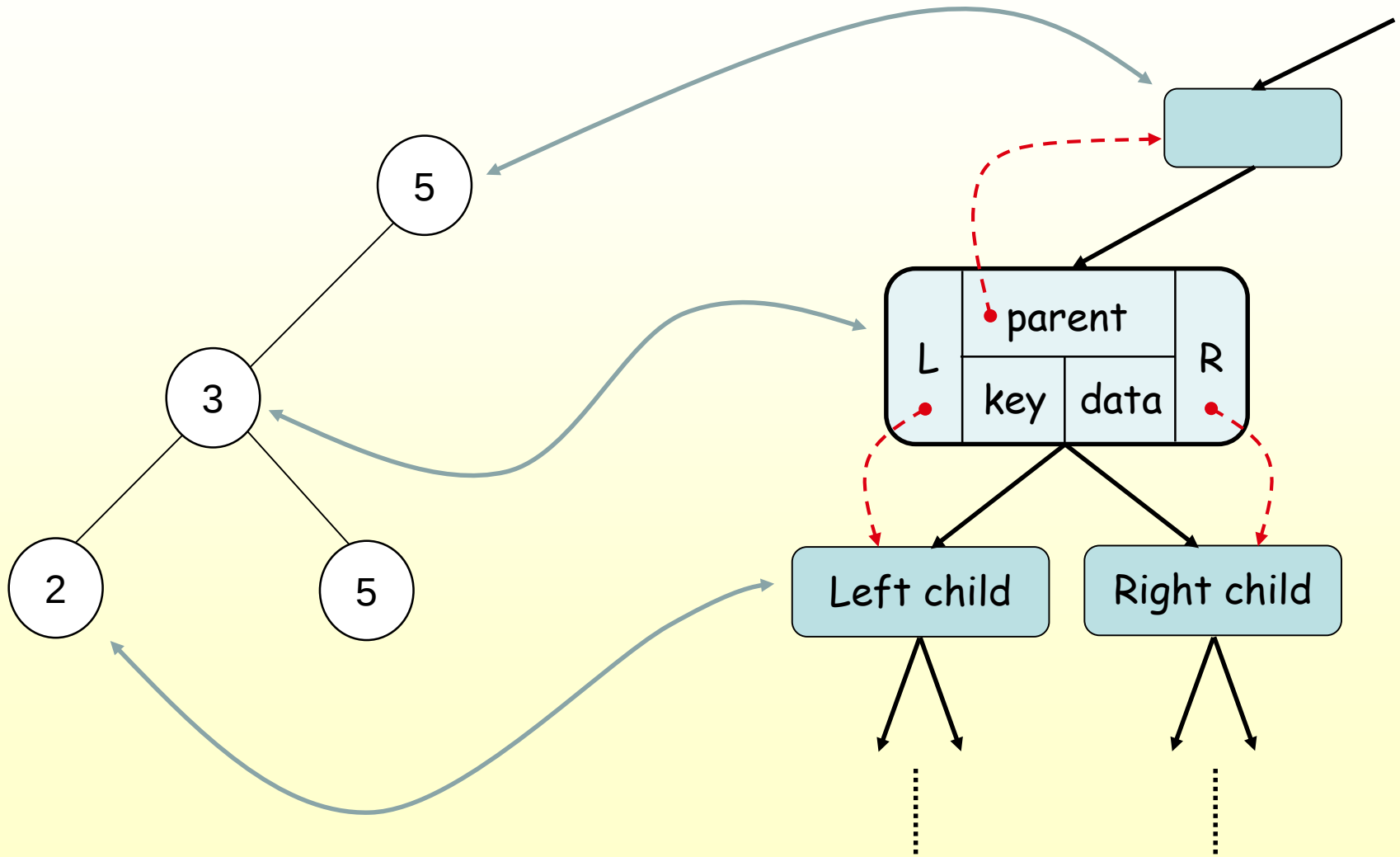


# Binary Tree Terminology



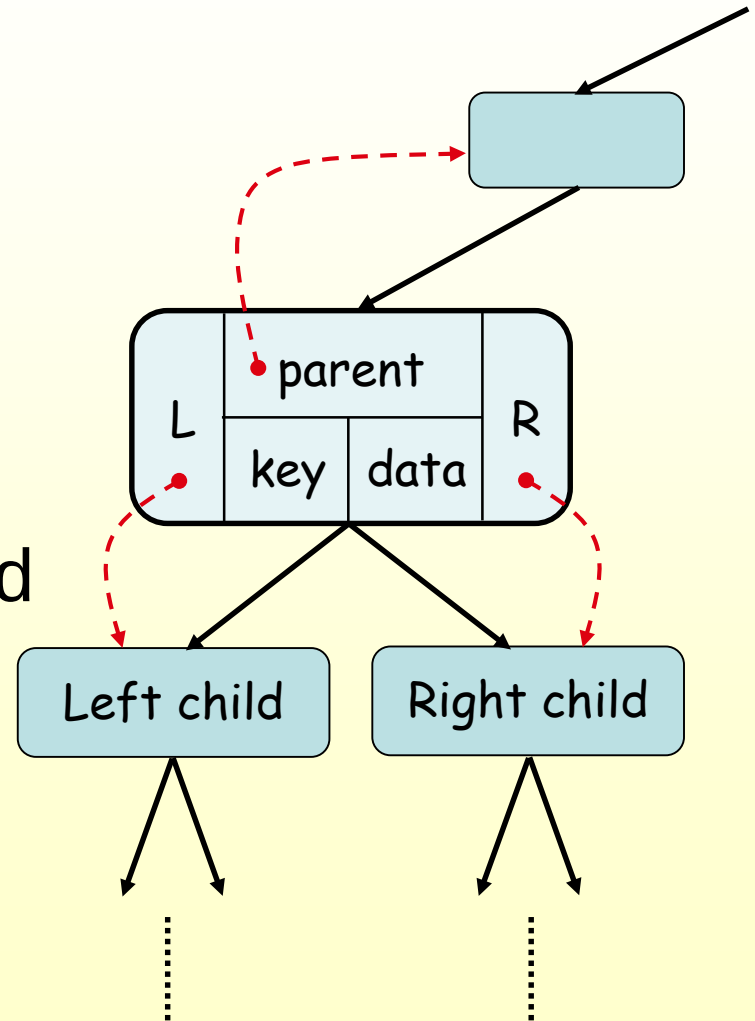
- 3 is the left child of 5
- 7 is the right child of 5
- 5 is the parent of 3

# Binary Tree Implementation



# Binary Tree Implementation

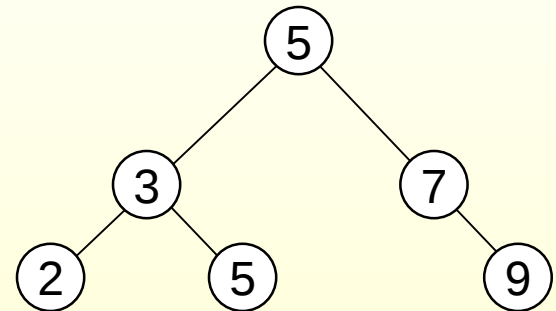
- ◆ Node representation:
  - ◆ Key field
  - ◆ Satellite data
  - ◆ **Left**: pointer to left child
  - ◆ **Right**: pointer to right child
  - ◆ **p**: pointer to parent
    - ◆ ( $p[\text{root}[T]] = \text{NIL}$ )



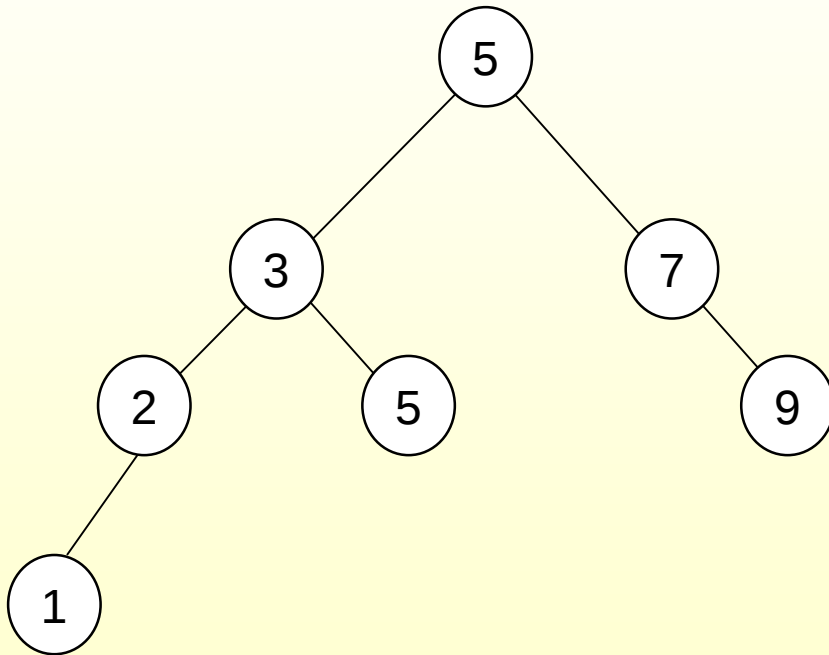
# Binary Search Trees

- ◆ A Binary Search Tree is a binary tree that satisfies the **Binary search tree property**:

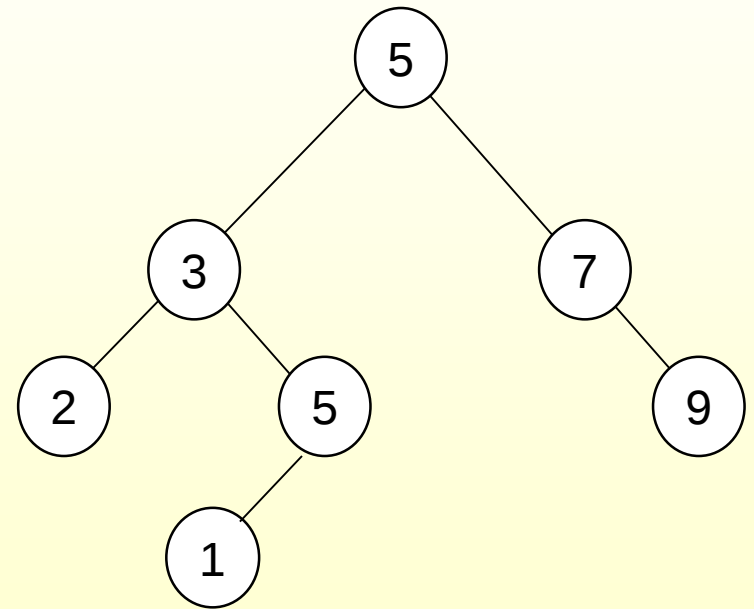
- ◆ If  $y$  is in left subtree of  $x$ ,  
then  $\text{key}[y] \leq \text{key}[x]$
- ◆ If  $y$  is in right subtree of  $x$ ,  
then  $\text{key}[y] \geq \text{key}[x]$



# Example and non-example



Binary Search Tree



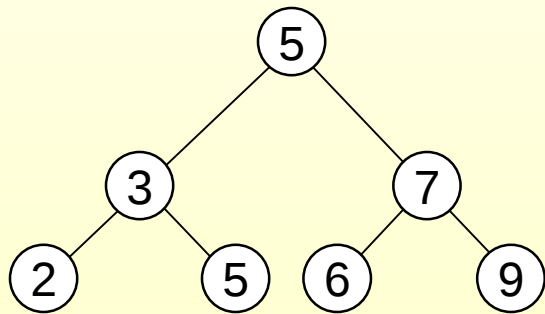
**NOT** a Binary Search Tree

# Binary Search Trees

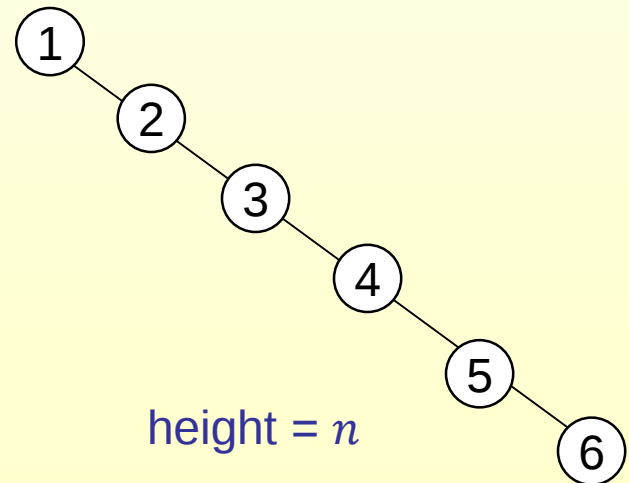
- ◆ Support **many** dynamic set operations:
  - ◆ SEARCH, MINIMUM, MAXIMUM,  
PREDECESSOR, SUCCESSOR, INSERT,  
DELETE
- ◆ In particular, a Binary Search Tree (BST) can implement both the **Dictionary** and the **Priority Queue** abstract data types

# Binary Search Trees

- Running time of basic operations on BSTs:
  - Depends on the **height** of the tree.
- What is the height of a BST on  $n$  nodes?



height =  $\Theta(\log(n))$



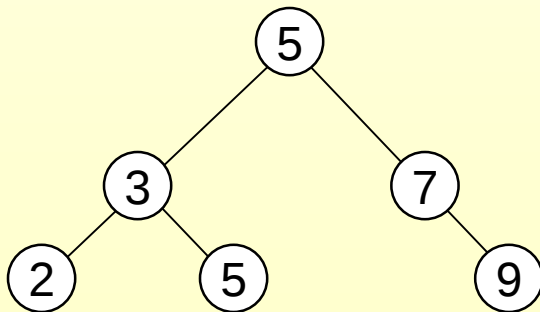
height =  $n$

How to get around a BST?

# TRAVERSALS

# In-Order Traversal

- ◆ When you are at a node x:
  - ◆ Recursively explore x's left child
  - ◆ Visit x
  - ◆ Recursively explore x's right child



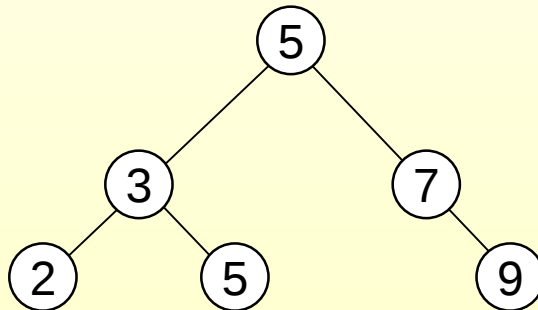
- 2
- 3
- 5
- 5
- 7
- 9

# In-Order Traversal

*Alg:* INORDER-TREE-WALK( $x$ )

1. **if**  $x \neq \text{NIL}$
2.     **then** INORDER-TREE-WALK ( left [ $x$ ] )
3.         visit/print key [ $x$ ]
4.         INORDER-TREE-WALK ( right [ $x$ ] )

◆ *E.g.:*



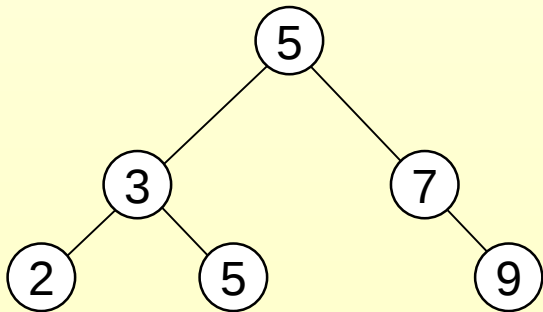
Output: 2 3 5 5 7 9

◆ Running time:

◆  $\Theta(n)$ , where  $n$  is the size of the tree rooted at  $x_{17}$

# Traversing a BST

- ◆ **Inorder** tree walk (traversal):
  - ◆ Root is visited between the visits of its left and right subtrees: **left, root, right**
  - ◆ Keys are visited in **sorted order**
- ◆ **Preorder** tree walk:
  - ◆ root visited first: **root, left, right**
- ◆ **Postorder** tree walk:
  - ◆ root visited last: **left, right, root**



Inorder: 2 3 5 5 7 9

Preorder: 5 3 2 5 7 9

Postorder: 2 5 3 9 7 5

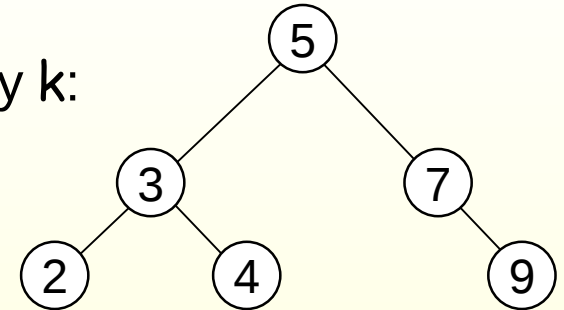
Basic operations on BSTs

**SEARCH, MAX, MIN,  
SUCCESSOR, PREDECESSOR,  
INSERT, DELETE**

# Searching for a Key

- ◆ Goal:

- ◆ Given a pointer to the root of a tree and a key  $k$ :
- ◆ Return a pointer to a node with key  $k$  (if one exists)
- ◆ Otherwise return NIL

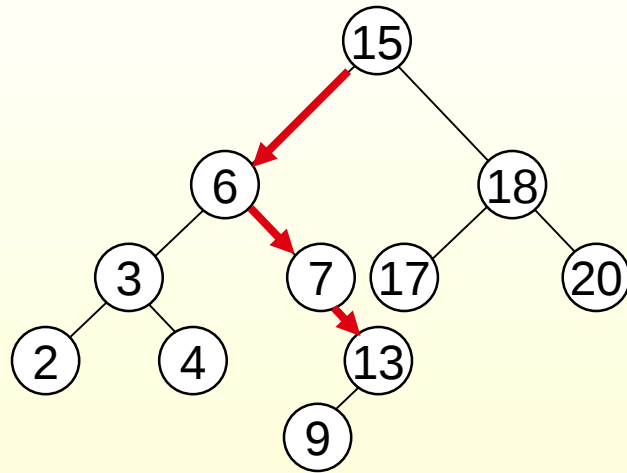


- ◆ Idea:

- ◆ Starting at the root: trace down a path by comparing  $k$  with the key of the current node:
  - ◆ If  $k = \text{key}[x]$ , we have found the key
  - ◆ If  $k < \text{key}[x]$ , search in the left subtree of  $x$
  - ◆ If  $k > \text{key}[x]$ , search in the right subtree of  $x$



# Example: TREE-SEARCH



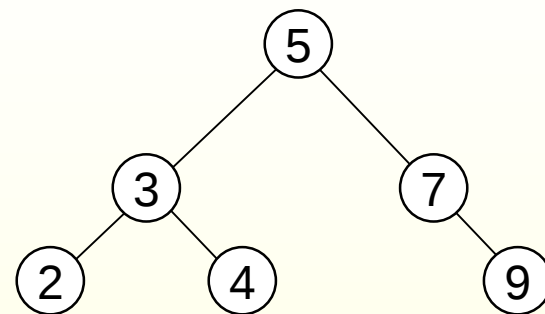
◆ Search for key 13:

◆  $15 \rightarrow 6 \rightarrow 7 \rightarrow 13$

# Searching for a Key

*Alg:* TREE-SEARCH( $x$ ,  $k$ )

1. **if**  $x = \text{NIL}$  or  $k = \text{key}[x]$
2.     **then return**  $x$
3. **if**  $k < \text{key}[x]$
4.     **then return** TREE-SEARCH(left [ $x$ ],  $k$ )
5.     **else return** TREE-SEARCH(right [ $x$ ],  $k$ )



Running Time:  $O(h)$ ,  
 $h$  – the height of the tree

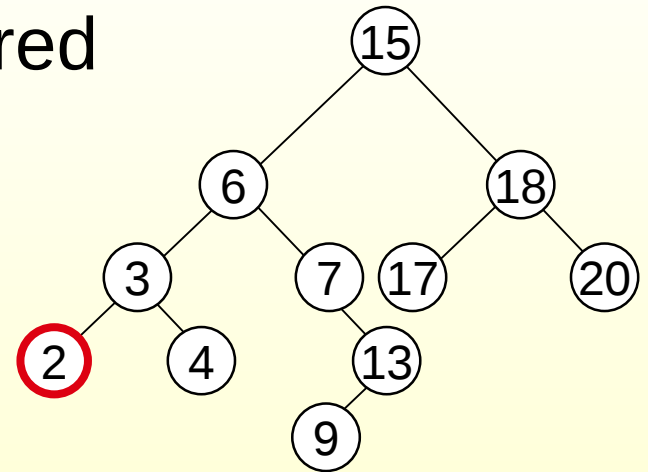
# Find the Minimum in a BST

## ✦ Idea:

- ✦ Follow left child pointers from the root, until a NIL is encountered

*Alg:* TREE-MINIMUM( $x$ )

1. **while**  $\text{left}[x] \neq \text{NIL}$
2.       **do**  $x \leftarrow \text{left}[x]$
3. **return**  $x$



Minimum = 2

Running time:  $O(h)$ ,  $h$  = height of tree

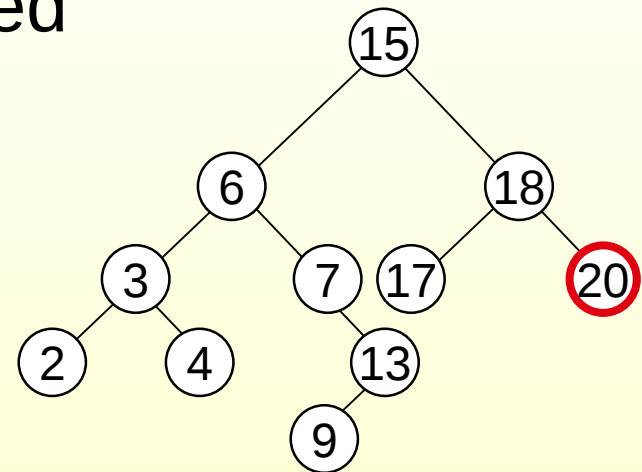
# Find the Maximum in a BST

## ♦ Idea:

- ♦ Follow right child pointers from the root, until a NIL is encountered

*Alg:* TREE-MAXIMUM( $x$ )

1. **while** right [ $x$ ]  $\neq$  NIL
2.       **do**  $x \leftarrow$  right [ $x$ ]
3. **return**  $x$



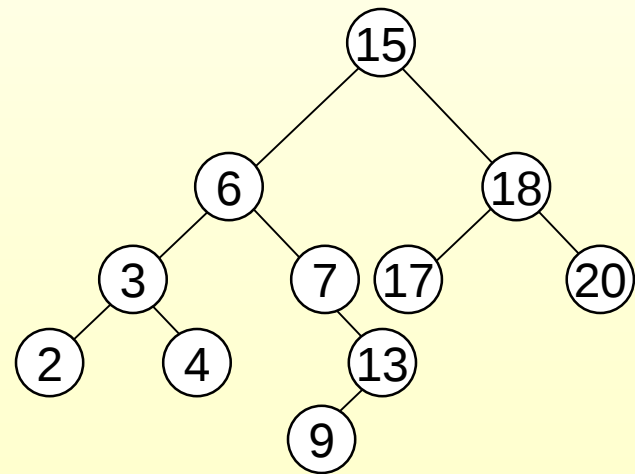
Maximum = 20

Running time:  $O(h)$ ,  $h$  = height of tree

# Successor

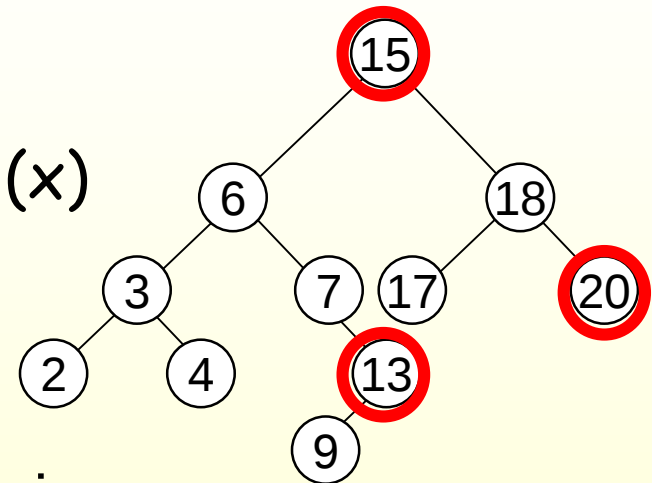
◆ *Def:*  $\text{successor}(x) = y$ , such that  $\text{key}[y]$  is the smallest key with  $\text{key}[y] > \text{key}[x]$

◆ *E.g.:*  $\text{successor}(15) = 17$   
 $\text{successor}(13) = 15$   
 $\text{successor}(9) = 13$



# Finding a Successor

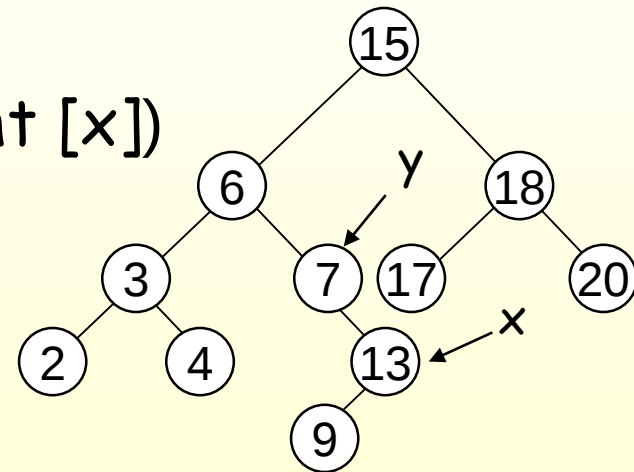
- ◆ Case 1:  $\text{right}(x)$  is non empty
  - ◆ Example:  $\text{successor}(15) = 17$ .
  - ◆  $\text{successor}(x)$  = the minimum in  $\text{right}(x)$
- ◆ Case 2:  $\text{right}(x)$  is empty
  - ◆ Example:  $\text{successor}(13) = 15$ .
  - ◆ go up the tree until the current node is a left child:  $\text{successor}(x)$  is the parent of the current node
  - ◆ Example:  $\text{successor}(20) = \text{NIL}$
  - ◆ if you cannot go further (and you reached the root):  $x$  is the largest element



# Finding a Successor

*Alg:* TREE-SUCCESSOR( $x$ )

1. **if** right [ $x$ ]  $\neq$  NIL
2.     **then return** TREE-MINIMUM(right [ $x$ ])
3.  $y \leftarrow p[x]$
4. **while**  $y \neq$  NIL and  $x =$  right [ $y$ ]
5.     **do**  $x \leftarrow y$
6.      $y \leftarrow p[y]$
7. **return**  $y$

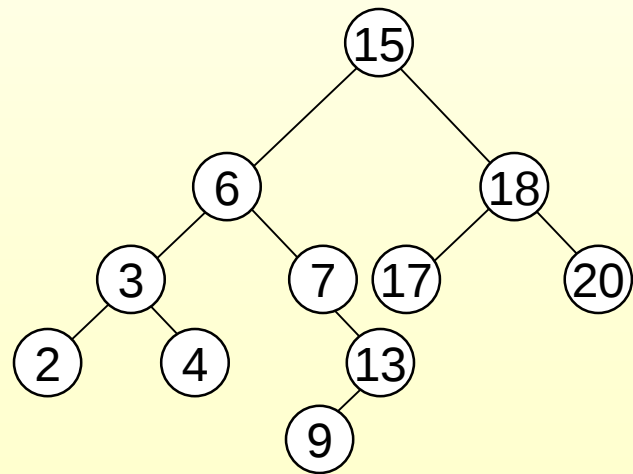


Running time:  $O(h)$ ,  $h$  = height of the tree

# Predecessor

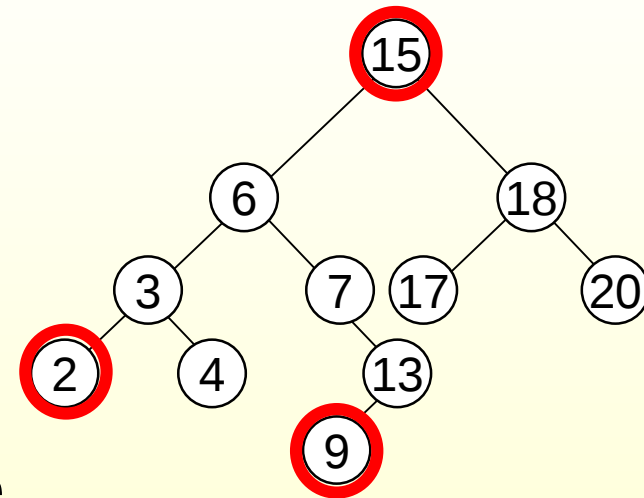
◆ *Def:*  $\text{predecessor}(x) = y$ , such that  $\text{key}[y]$  is the largest key with  $\text{key}[y] < \text{key}[x]$

◆ *E.g.:*  $\text{predecessor}(15) = 13$   
 $\text{predecessor}(9) = 7$   
 $\text{predecessor}(7) = 6$



# Finding a Predecessor

- ◆ Case 1:  $\text{left}(x)$  is non empty
  - ◆ *Example:  $\text{predecessor}(15) = 13$*
  - ◆  $\text{predecessor}(x) = \text{maximum in left}(x)$
- ◆ Case 2:  $\text{left}(x)$  is empty
  - ◆ *Example:  $\text{predecessor}(9) = 7$*
  - ◆ go up the tree until the current node is a right child:  $\text{predecessor}(x)$  is the parent of the current node
  - ◆ *Example:  $\text{predecessor}(2) = \text{NIL}$*
  - ◆ if you cannot go further (and you reached the root):  $x$  is the smallest element



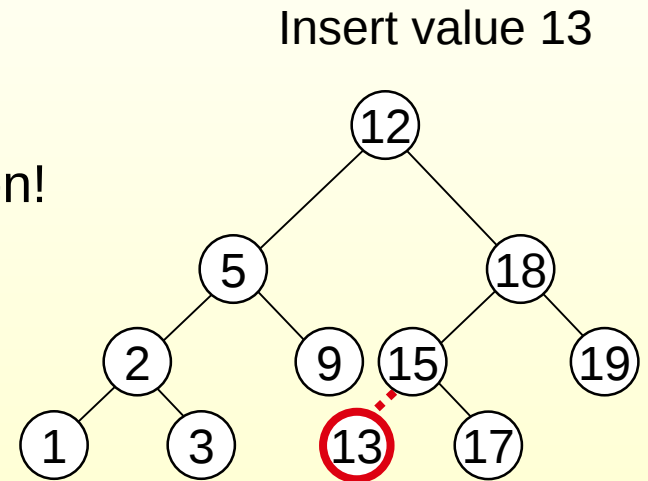
# Insertion

- ◆ Goal:

- ◆ Insert value  $v$  into a binary search tree

- ◆ Idea:

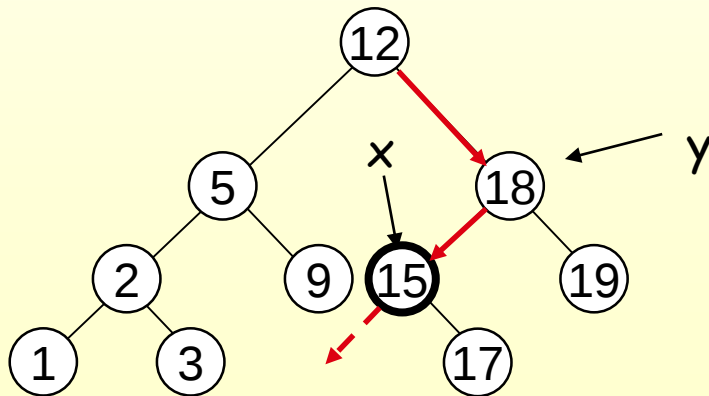
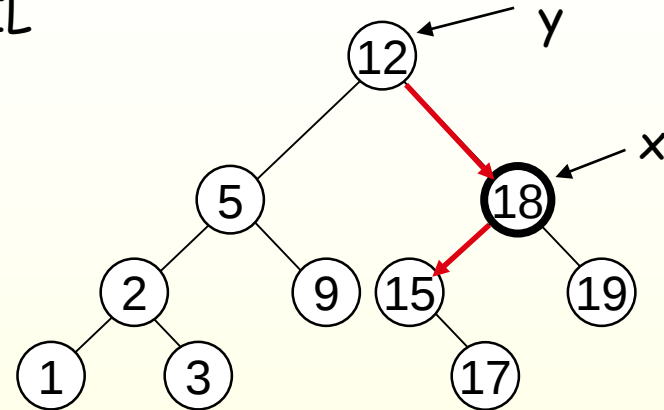
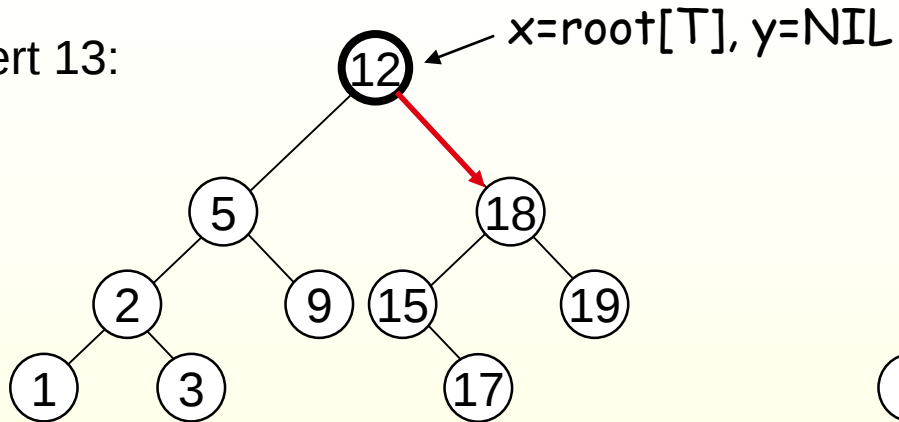
- ◆ If  $\text{key}[x] < v$  move to the right child of  $x$ ,  
else move to the left child of  $x$
- ◆ When  $x$  is NIL, we found the correct position!



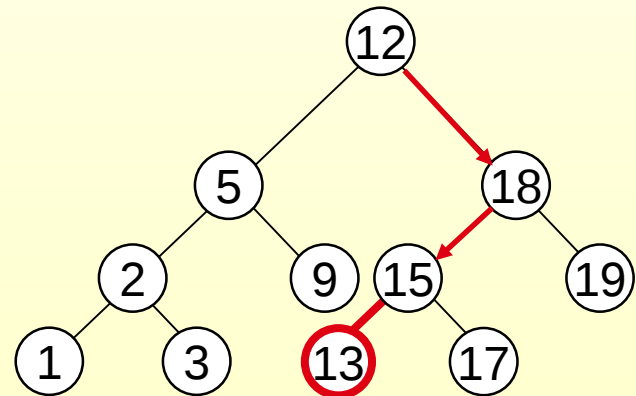
- ◆ Beginning at the root, go down the tree and maintain:
  - ◆ Pointer  $x$  : current node
  - ◆ Pointer  $y$  : parent of  $x$  (“trailing pointer” )

# Example: TREE-INSERT

Insert 13:

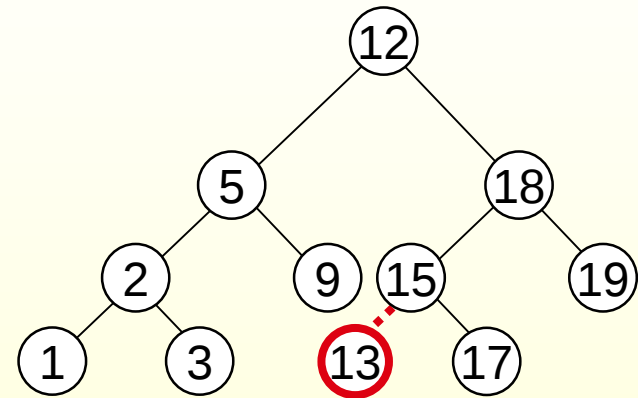


$x = \text{NIL}$   
 $y = 15$



## *Alg:* TREE-INSERT( $T, z$ )

1.  $y \leftarrow \text{NIL}$
2.  $x \leftarrow \text{root}[T]$
3. **while**  $x \neq \text{NIL}$
4.     **do**  $y \leftarrow x$
5.         **if**  $\text{key}[z] < \text{key}[x]$
6.             **then**  $x \leftarrow \text{left}[x]$
7.             **else**  $x \leftarrow \text{right}[x]$
8.  $p[z] \leftarrow y$
9. **if**  $y = \text{NIL}$
10.     **then**  $\text{root}[T] \leftarrow z$
11.     **else if**  $\text{key}[z] < \text{key}[y]$
12.         **then**  $\text{left}[y] \leftarrow z$
13.         **else**  $\text{right}[y] \leftarrow z$



▷ Tree  $T$  was empty

Running time:  $O(h)$

# Deletion

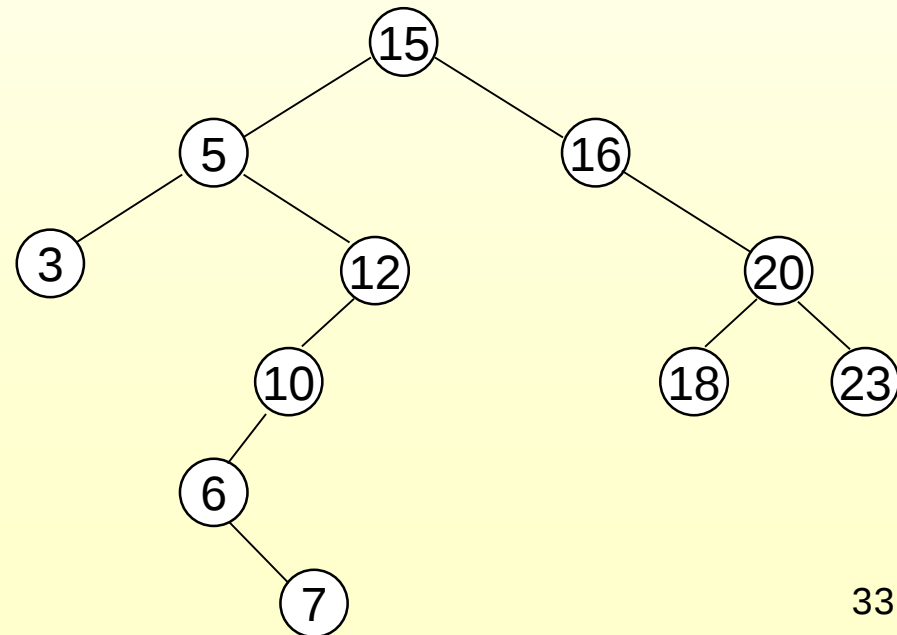
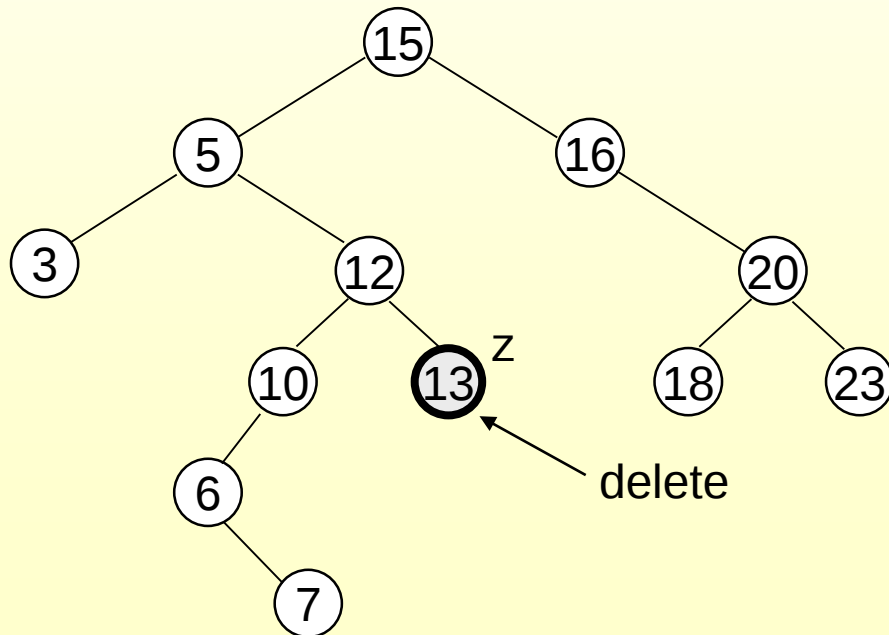
- ◆ Goal:

- ◆ Delete a given node  $z$  from a binary search tree

- ◆ Idea:

- ◆ **Case 1:**  $z$  has no children

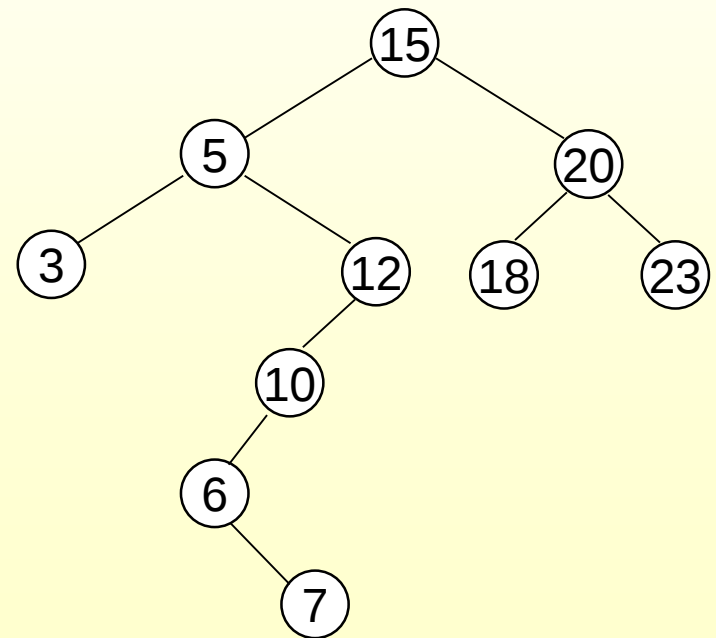
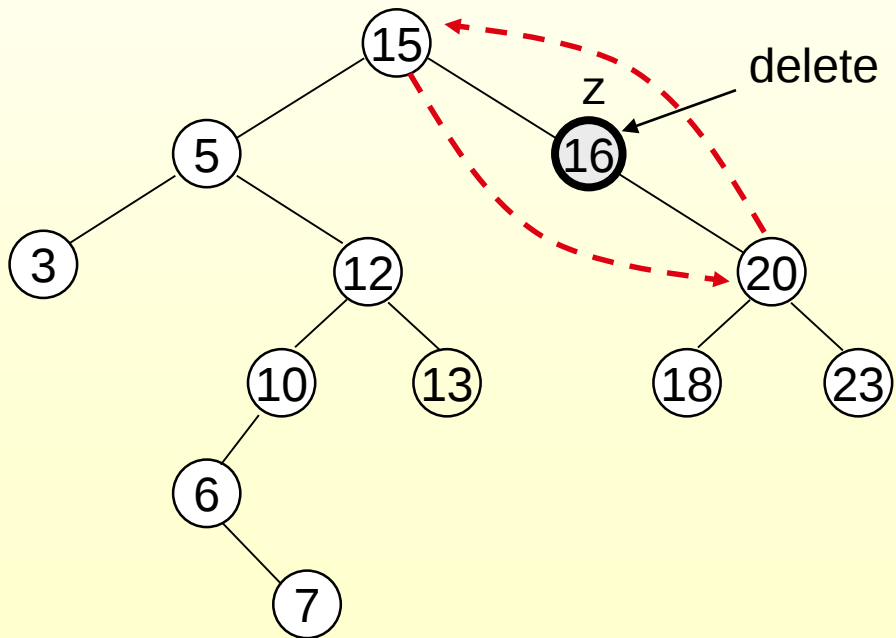
- ◆ Delete  $z$  by making the parent of  $z$  point to NIL



# Deletion

## Case 2: $z$ has one child

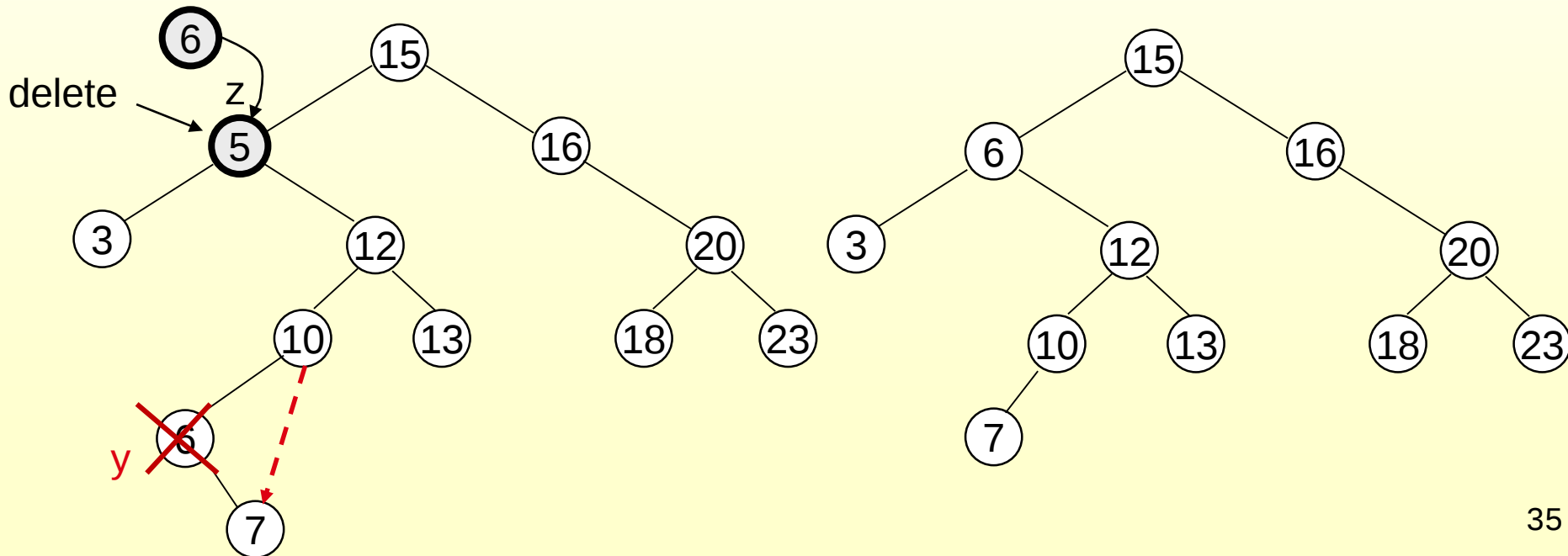
- Delete  $z$  by making the parent of  $z$  point to  $z$ 's child, instead of to  $z$



# Deletion

## ◆ Case 3: z has two children

- ◆ z's successor (y) is the minimum node in z's right subtree
- ◆ y has either no children or one right child (but no left child)
- ◆ Delete y from the tree (via Case 1 or 2)
- ◆ Replace z's key and satellite data with y's.



# TREE-DELETE( $T, z$ )

1. **if**  $\text{left}[z] = \text{NIL}$  or  $\text{right}[z] = \text{NIL}$

2.   **then**  $y \leftarrow z$

$z$  has one child

3.   **else**  $y \leftarrow \text{TREE-SUCCESSOR}(z)$

$z$  has 2 children

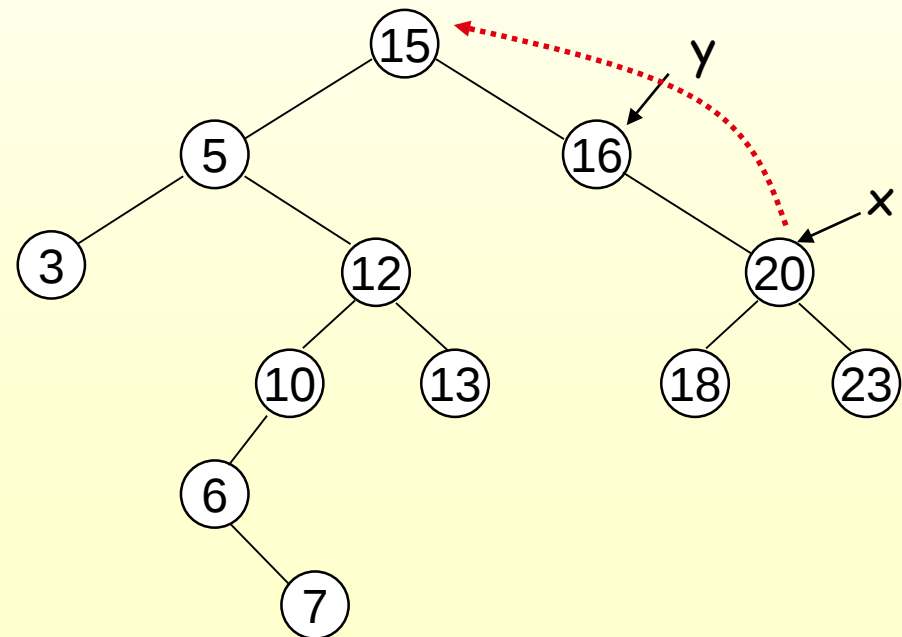
4. **if**  $\text{left}[y] \neq \text{NIL}$

5.   **then**  $x \leftarrow \text{left}[y]$

6.   **else**  $x \leftarrow \text{right}[y]$

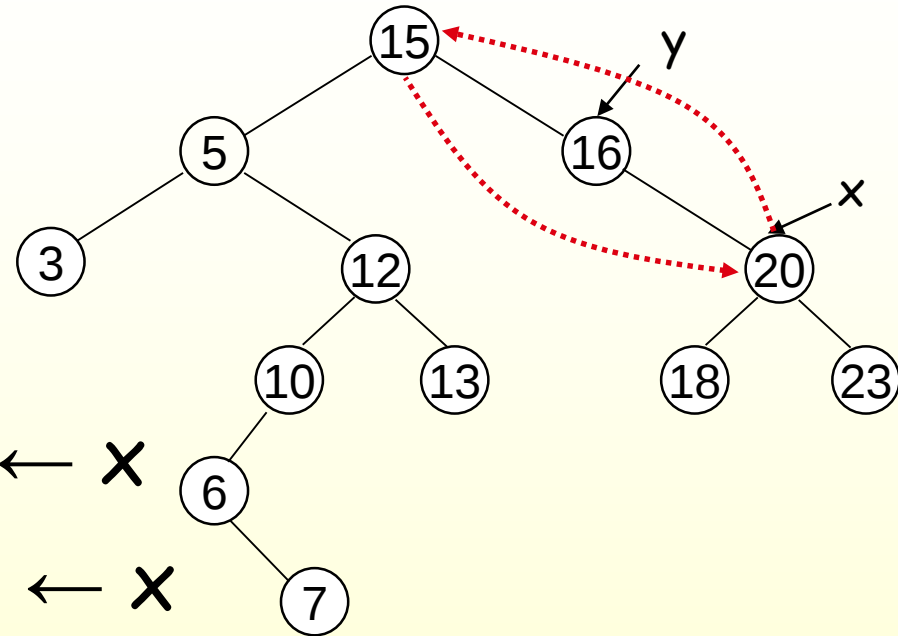
7. **if**  $x \neq \text{NIL}$

8.   **then**  $p[x] \leftarrow p[y]$



# TREE-DELETE( $T, z$ ) – cont.

9. **if**  $p[y] = \text{NIL}$
10.   **then**  $\text{root}[T] \leftarrow x$
11.   **else if**  $y = \text{left}[p[y]]$
12.       **then**  $\text{left}[p[y]] \leftarrow x$
13.       **else**  $\text{right}[p[y]] \leftarrow x$
14. **if**  $y \neq z$
15.   **then**  $\text{key}[z] \leftarrow \text{key}[y]$
16.       copy  $y$ 's satellite data into  $z$
17. **return**  $y$



Running time:  $O(h)$

# Binary Search Trees - Summary

- Operations on binary search trees:

|             |        |
|-------------|--------|
| SEARCH      | $O(h)$ |
| PREDECESSOR | $O(h)$ |
| SUCCESSOR   | $O(h)$ |
| MINIMUM     | $O(h)$ |
| MAXIMUM     | $O(h)$ |
| INSERT      | $O(h)$ |
| DELETE      | $O(h)$ |

- These operations are fast if the height of the tree is **small** – otherwise their performance is similar to that of a linear linked list

Sorting with BSTs

# TREESORT

# Sorting With BSTs

- ◆ Informal code for sorting array  $A$  of length  $n$ :

```
TreeSort(A)
```

```
    for i=1 to n
```

```
        TreeInsert(A[i]);
```

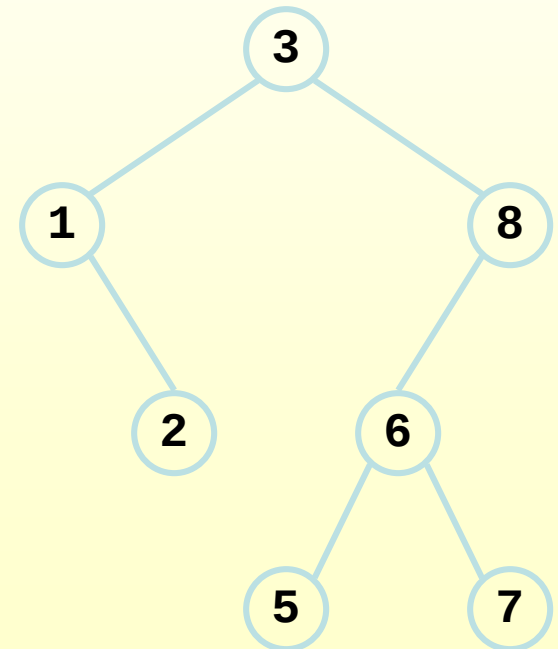
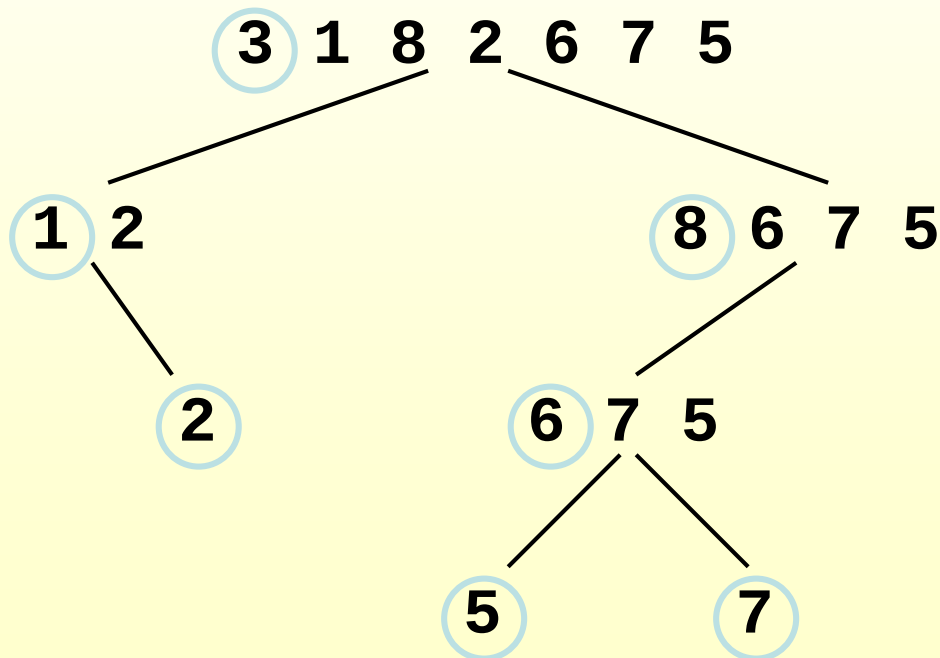
```
    InorderTreeWalk(root);
```

- ◆ *Argue that this is  $\Omega(n \lg n)$*
- ◆ *What will be the running time in the*
  - ◆ *Worst case?*
  - ◆ *Average case? (hint: remind you of anything?)*

# Sorting With BSTs

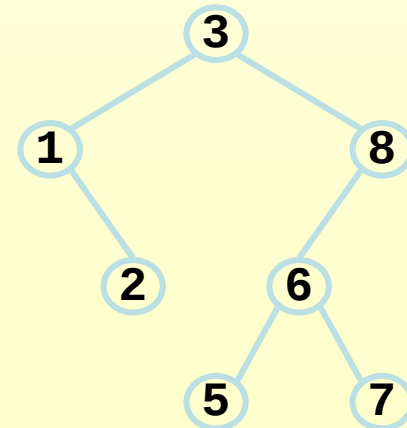
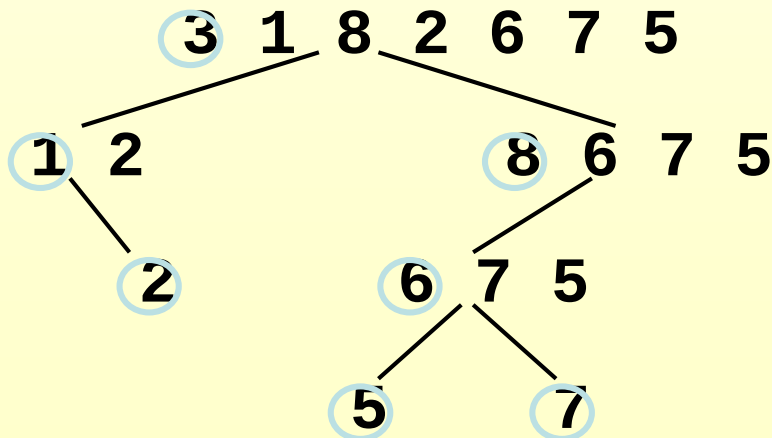
```
for i=1 to n  
    TreeInsert(A[i]);  
InorderTreeWalk(root);
```

✿ A form of QuickSort!



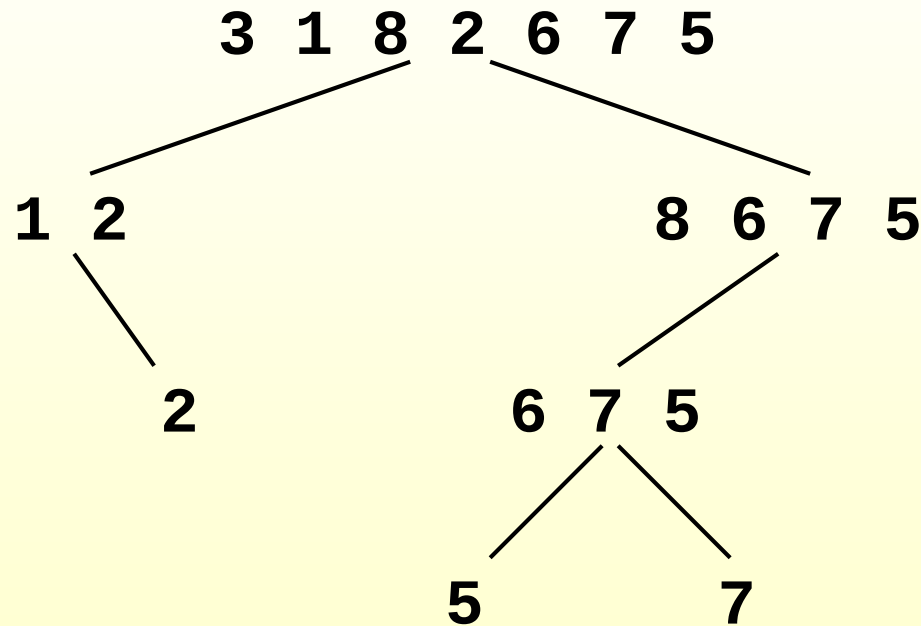
# Sorting with BSTs

- ◆ Same partitions are done as with QuickSort.
- ◆ Same comparisons:
  - ◆ In previous example
    - ◆ Everything was compared to 3 once
    - ◆ Then those items  $< 3$  were compared to 1 once
    - ◆ Etc.
- ◆ But! Comparisons happen in a different order.
  - ◆ Example: consider inserting 5



# Analysis of TreeSort

TreeSort performs the same comparisons as QuickSort, but in a different order.



The expected time to build the tree is asymptotically the same as the running time of QuickSort.

# Sorting with BSTs

- ◆ Since run time is proportional to the number of comparisons, same time as QuickSort:  $O(n \lg n)$
- ◆ *Which do you think is better, QuickSort or TreeSort? Why?*

# Sorting with BSTs

- ◆ Since run time is proportional to the number of comparisons, same time as quicksort:  $O(n \lg n)$
- ◆ *Which do you think is better, QuickSort or TreeSort? Why?*
- ◆ A: QuickSort
  - ◆ Better constants
  - ◆ Sorts in place
  - ◆ Doesn't need to build a data structure

Since everything we've done so far depends on the height of the tree, we'd better analyze...

## **EXPECTED HEIGHT**

# Expected Height

- **Claim:** The expected height of a randomly build BST is  $O(\lg(n))$ .
- **Proof:** (NOT QUITE!)
  - The depth of a node is the number of comparisons made during TREE-INSERT.
  - Average expected node depth:
$$= \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\text{number of comparisons to insert node } i]$$
$$= O\left(\frac{1}{n} \cdot n \lg(n)\right) = O(\lg(n)) \text{ by the QuickSort analysis.}$$
  - So most nodes aren't too deep, so the expected height is small? (NOT NECESSARILY!)

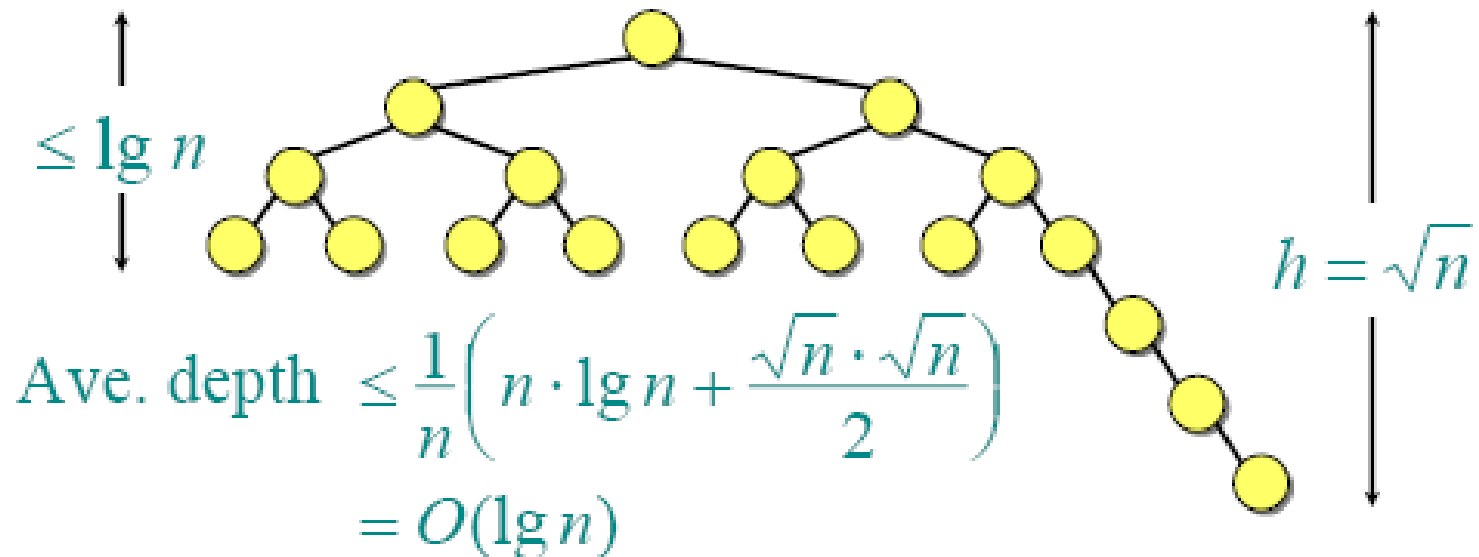
# Expected Node Depth

- ❖ **Claim:** The expected average node depth of a randomly built BST is  $O(\lg(n))$ .
- ❖ **Proof:** previous slide.

# Expected Tree Height

- ❖ The height of a tree might be much larger than the average depth!
- ❖ So we need a more complicated analysis.

## Example.



# The plan

## to estimate Expected Tree Height

- ◆ Review ***Jensen's inequality***:
  - ◆  $f(\mathbb{E}[X]) \leq \mathbb{E}[f(X)]$  for any convex function  $f$  and random variable  $X$ .
- ◆ Analyze  $\exp(\text{height})$  of a randomly built BST:
  - ◆  $Y_n = 2^{X_n}$ , where  $X_n$  is the height of a random tree on  $n$  nodes.
  - ◆ We'll show  $\mathbb{E}[Y_n] = O(n^3)$ , hence
$$2^{\mathbb{E}[X_n]} \leq \mathbb{E}[2^{X_n}] = \mathbb{E}[Y_n] = O(n^3).$$
  - ◆ so  $\mathbb{E}[X_n] = O(\lg(n))$ .

The details will be on the board.  
There are hidden slides after this for reference later.

# Post Mortem

- Q. Does the analysis have to be this hard?
- Q. Why bother with analyzing exponential height?
- Q. Why not just develop the recurrence on

$$X_n = 1 + \max\{X_{k-1}, X_{n-k}\}$$

directly?

# Post Mortem (Continued)

A.

- The inequality

$$\max\{a, b\} \leq a + b$$

provides a poor upper bound, since the RHS approaches the LHS slowly as  $|a - b|$  increases.

- The bound

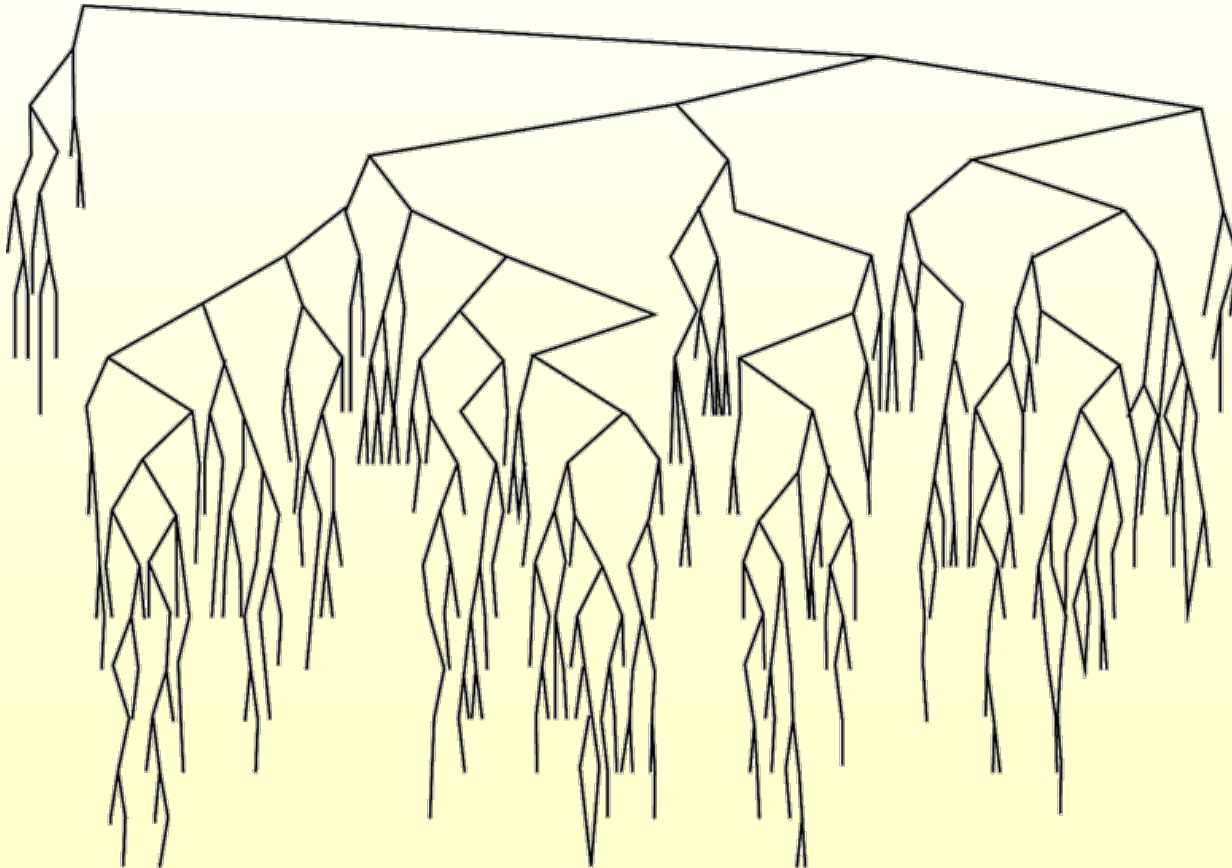
$$\max\{2^a, 2^b\} \leq 2^a + 2^b$$

allows the RHS to approach the LHS more quickly as  $|a - b|$  increases.

- By using the convexity of  $f(x) = 2^x$  via Jensen's inequality, we can manipulate the sum of exponentials, resulting in a tight analysis.

# Example Random BST

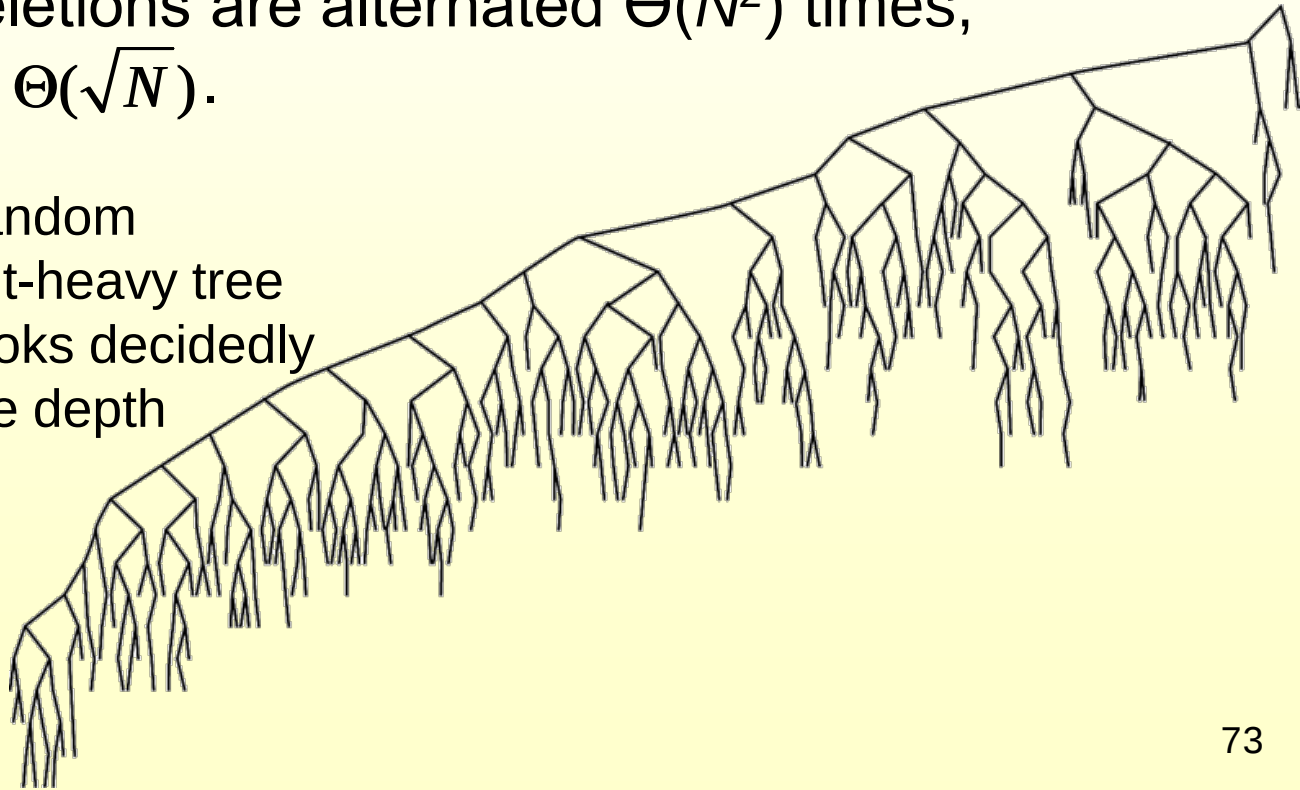
- ◆ An example of a randomly generated 500 node BST
- ◆ nodes at expected depth 9.98, height = 17.



# Not All Tree Operations Preserve Randomness

- Deletion algorithm described favors making left subtrees deeper than right subtrees (a deleted node is replaced with a node from the right). The exact effect of this still unknown, but if insertions and deletions are alternated  $\Theta(N^2)$  times, expected depth is  $\Theta(\sqrt{N})$ .

After a quarter-million random insert/remove pairs, right-heavy tree on the previous slide, looks decidedly unbalanced and average depth becomes 12.51.



# BST Tree Height Summary

- ◆ The height  $h$  of a binary search tree on  $n$  items can be as high as  $n-1$
- ◆ However, if it is built by inserting the elements in random order, then the expected height is  $O(\lg n)$ .

# Wrap-up

## ◆ Binary Search Trees

- ◆ Traversals

- ◆ All sorts of operations!  
Insert/Delete/Search/etc...

- ◆ TreeSort

- ◆ All those things depend on the height of the tree, so it's good that the expected height of a randomly build BST =  $O(\lg(n))$

# Next time

- ◆ Prof. Guibas will be back!
- ◆ You'll see how to make sure that trees stay balanced!