

# Comparison of data structs and algos so far...

	Find/Search Any	Insert	Delete Any	Notes
Stack and Standard Queue	$O(n)$	$O(1)$	$O(n)$	
Unsorted List Priority Queue	$O(n)$	$O(1)$	$O(n)$	
Sorted List Priority Queue	$O(n)$	$O(n)$	$O(n)$	
Heap Priority Queue	$O(n)$	$O(\log n)$	$O(n)$	Designed for operations on "min" in $O(\log n)$
Skip List	$\log n$ high prob.	$\log n$ high prob.	$\log n$ high prob.	Randomized insertion algorithm
Hash Table	1 expected	1 expected	1 expected	$O(n)$ worst case

# Comparison of data structs and algos so far...

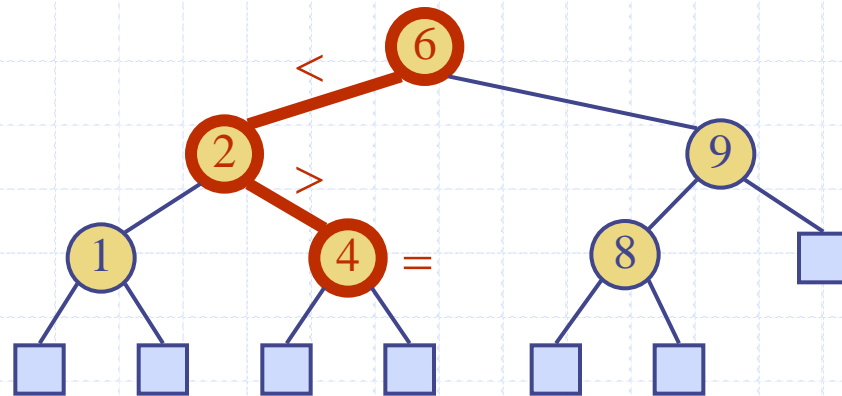
	Find/Search Any	Insert	Delete Any	Notes
Stack and Standard Queue	$O(n)$	$O(1)$	$O(n)$	
Unsorted List Priority Queue	$O(n)$	$O(1)$	$O(n)$	
Sorted List Priority Queue	$O(n)$	$O(n)$	$O(n)$	
Heap Priority Queue	$O(n)$	$O(\log n)$	$O(n)$	Designed for operations on "min" in $O(\log n)$
Skip List	$\log n$ high prob.	$\log n$ high prob.	$\log n$ high prob.	Randomized insertion algorithm
Hash Table	1 expected	1 expected	1 expected	$O(n)$ worst case

# Find/Search

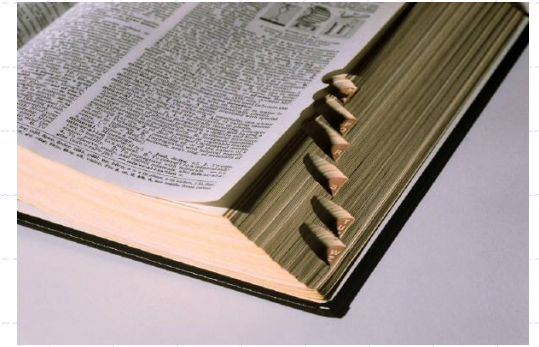
- ◆ Performance of delete depends on find/search
  - (performance of insert might as well...)
- ◆ Thus, we focus on improving find/search
- ◆ This brings us to...

# Binary Search Trees

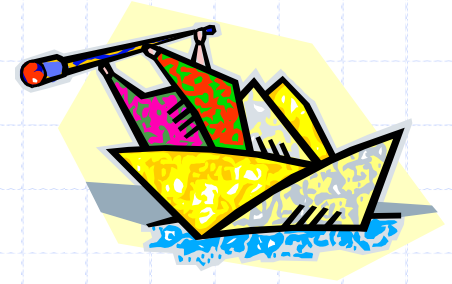
(and several others methods eventually...)



# Assumptions

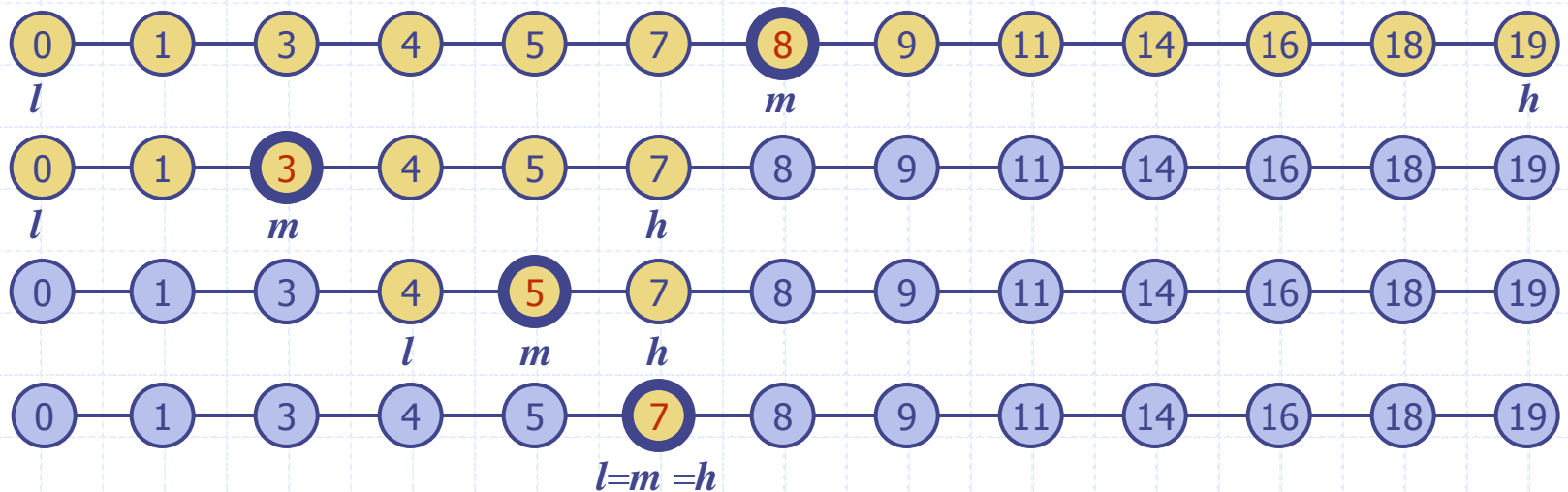


- ◆ We have an “ordered dictionary”
  - Keys are assumed to come from a total order
    - ◆ E.g., you can compare any key to any key and get a proper ordering
      - (this is as opposed to a partial ordering, where only adjacent keys can be compared with other keys)
  - New operations:
    - ◆ `closestBefore(k)`
    - ◆ `closestAfter(k)`

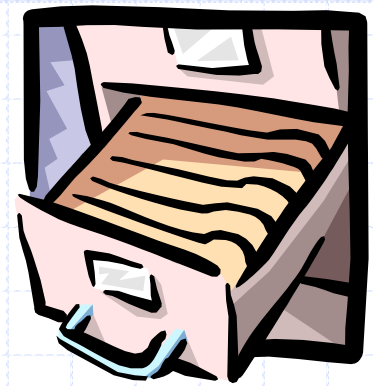


# Binary Search (array-based)

- ◆ Perform operation **find**(k) on a dictionary implemented by means of an array-based sequence, sorted by key
  - at each step, the number of candidate items is halved
  - terminates after  $O(\log n)$  steps
- ◆ Example: **find**(7)

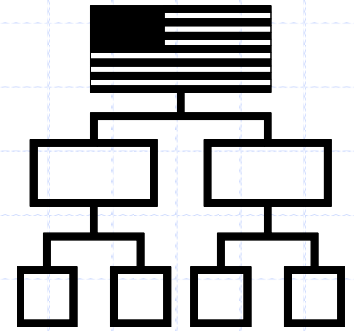


# Lookup Table



- ◆ A lookup table is a dictionary implemented by means of a sorted sequence and uses binary search
  - We store the items of the dictionary in an array-based sequence, sorted by key
  - We use an external comparator for the keys
- ◆ Performance:
  - **Find:**
    - ◆  $O(\log n)$  time (using binary search)
  - **insertItem:**
    - ◆  $O(n)$  time since in the worst case we have to shift  $n/2$  items to make room for the new item
  - **removeElement:**
    - ◆  $O(n)$  time since in the worst case we have to shift  $n/2$  items to compact the items after the removal
- ◆ The lookup table is effective for small dictionaries on which searches are the most common operations, while insertions and removals are rare

# Binary Search Tree



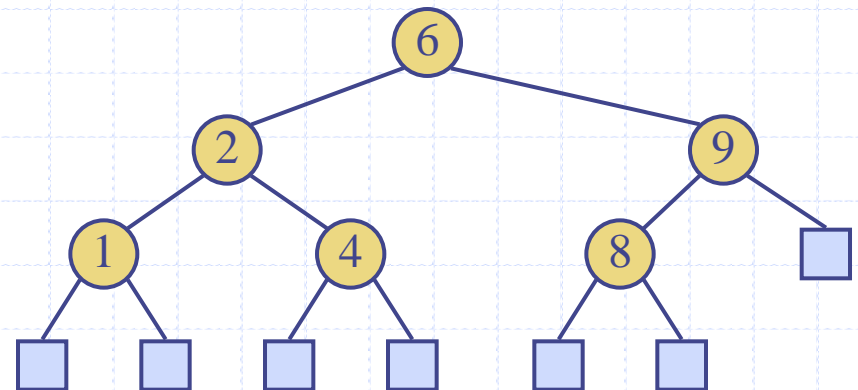
◆ A binary search tree is a binary tree storing key-element pairs at its internal nodes and satisfying the following property:

- Let  $u$ ,  $v$ , and  $w$  be three nodes such that  $u$  is in the left subtree of  $v$  and  $w$  is in the right subtree of  $v$ . We have  $key(u) \leq key(v) \leq key(w)$

◆ External nodes do not store items

◆ Thus, how do you visit all keys in increasing order?

- inorder traversal...





# Find/Search

- ◆ To search for a key  $k$ , we trace a downward path starting at the root
- ◆ The next node visited depends on the outcome of the comparison of  $k$  with the key of the current node
- ◆ If we reach a leaf, the key is not found and we return a null position
- ◆ Example: **find**(4)

**Algorithm** *find* ( $k, v$ )

**if** *T.isExternal* ( $v$ )

**return** *Position*(*null*)

**if**  $k < \text{key}(v)$

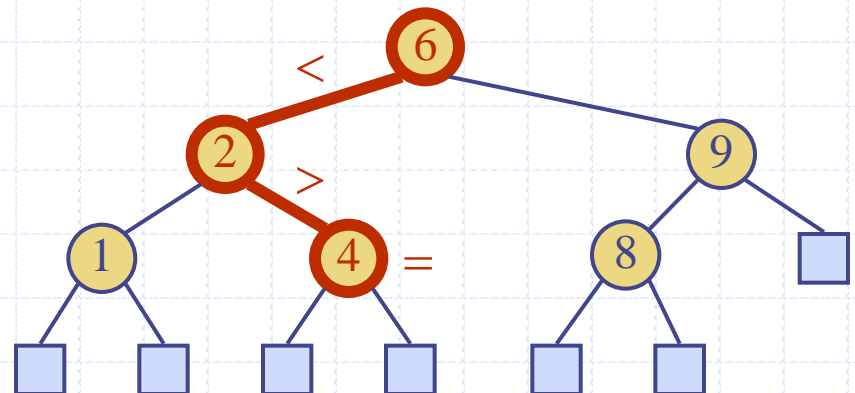
**return** *find*( $k, T.\text{leftChild}(v)$ )

**else if**  $k = \text{key}(v)$

**return** *Position*( $v$ )

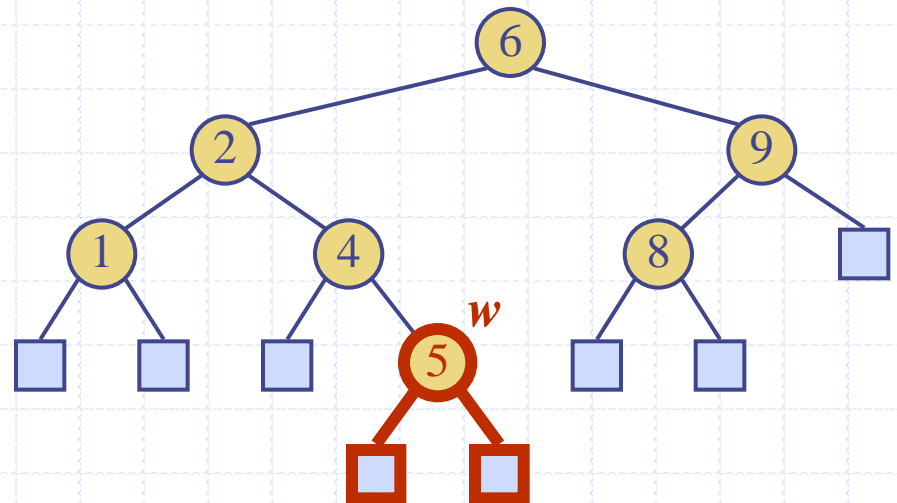
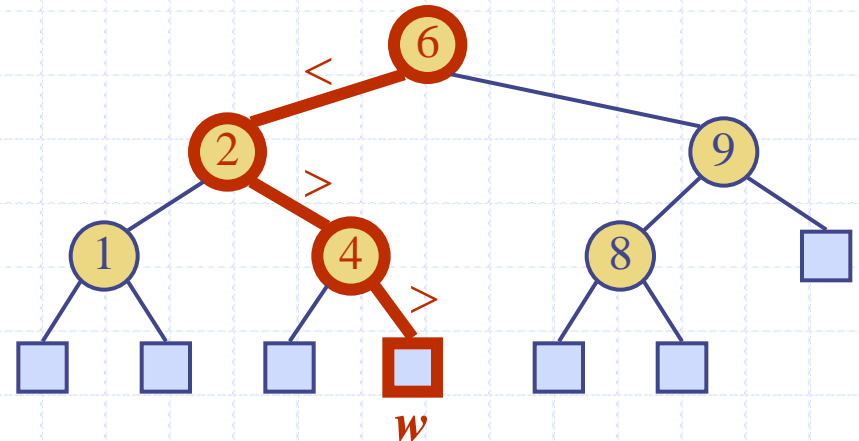
**else** {  $k > \text{key}(v)$  }

**return** *find*( $k, T.\text{rightChild}(v)$ )



# Insertion

- ◆ To perform operation **insertItem**( $k, o$ ), we search for key  $k$
- ◆ Assume  $k$  is not already in the tree, and let  $w$  be the leaf reached by the search
- ◆ We insert  $k$  at node  $w$  and expand  $w$  into an internal node
- ◆ Example: insert 5

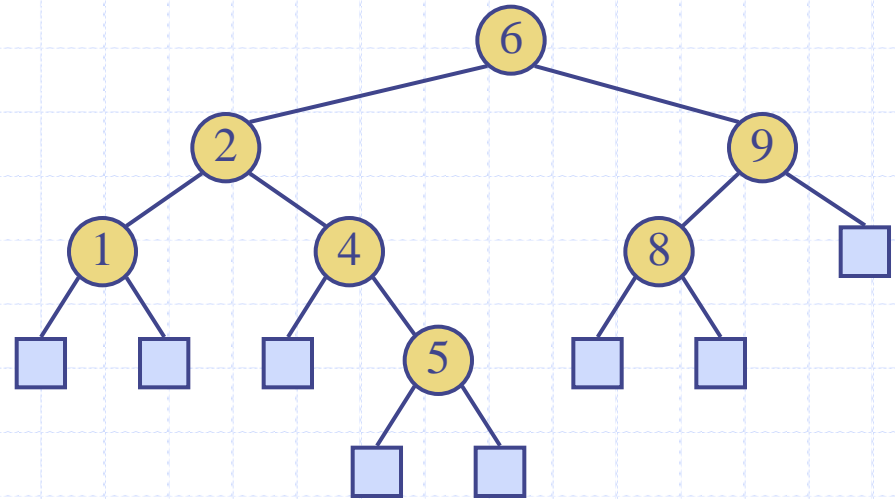


# Deletion

- ◆ Three cases:
  - Zero children
  - One child
  - Two children

# Deletion: zero children

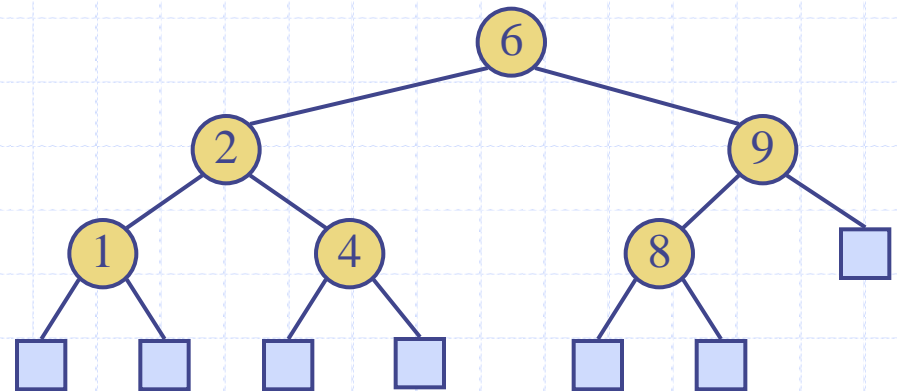
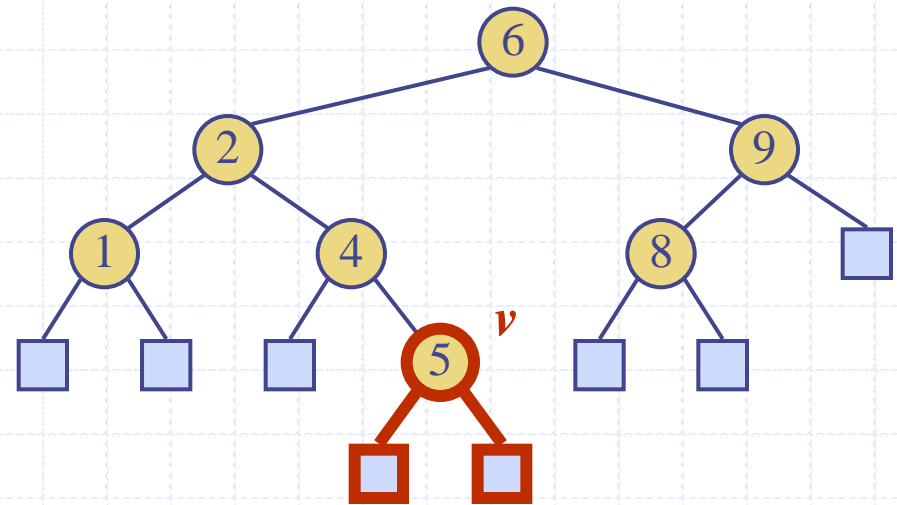
- ◆ Must be a leaf node – simple (e.g., remove 5)



# Deletion: zero children

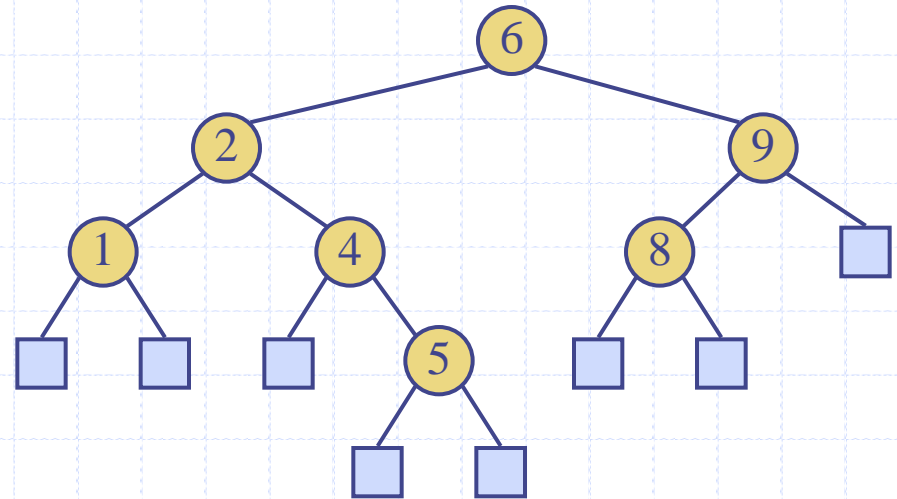
◆ Must be a leaf node – simple (e.g., remove 5)

- Assume key  $k$  is in tree, and let  $v$  be the node storing  $k$
- We search for key  $k$
- Remove node



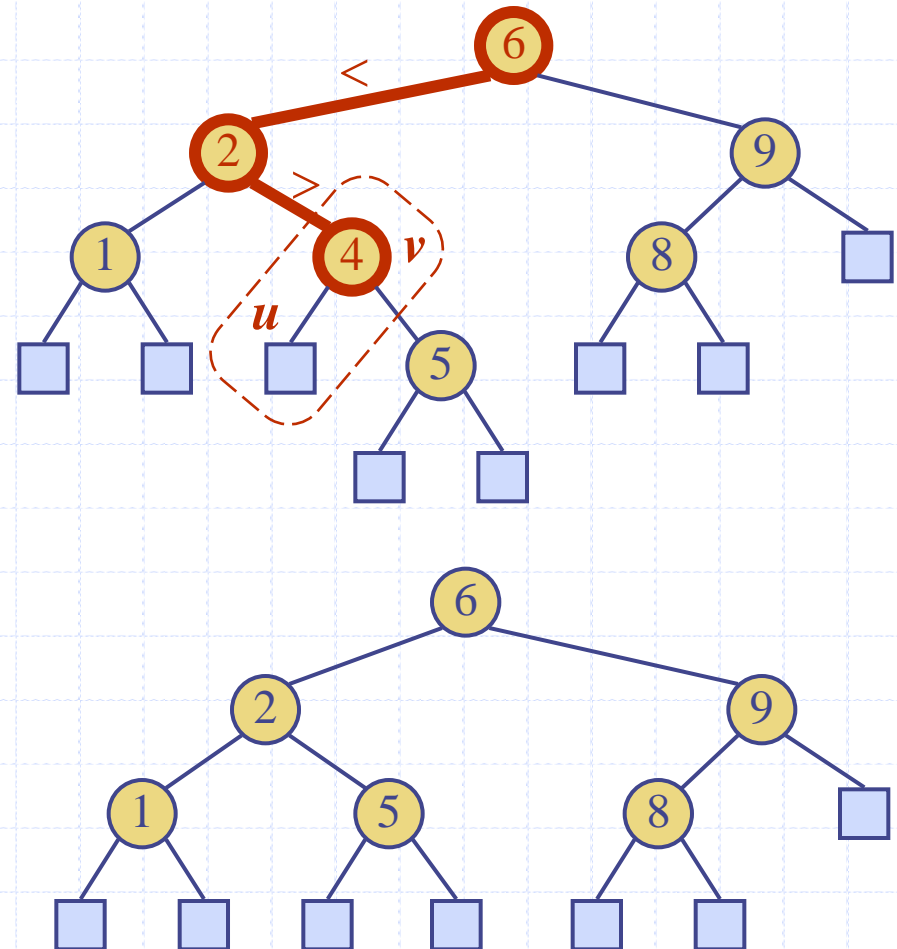
# Deletion: one child

- ◆ To perform operation, we search for key  $k$  (e.g., remove 4)



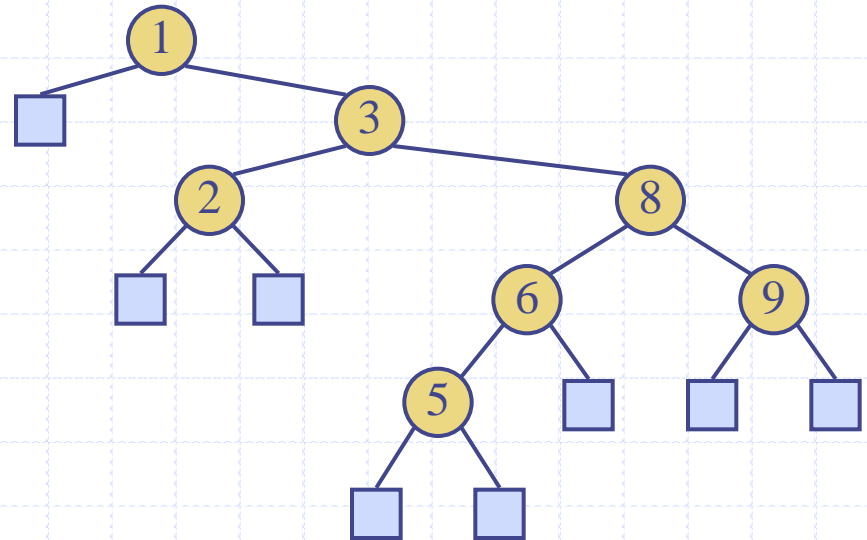
# Deletion: one child

- ◆ To perform operation, we search for key  $k$  (e.g., remove 4)
- ◆ Assume key  $k$  is in tree, and let  $v$  be the node storing  $k$
- ◆ If node  $v$  has **one** leaf child  $u$ , we remove  $v$  and  $u$  from the tree with operation **removeAboveExternal( $u$ )**



# Deletion: two children

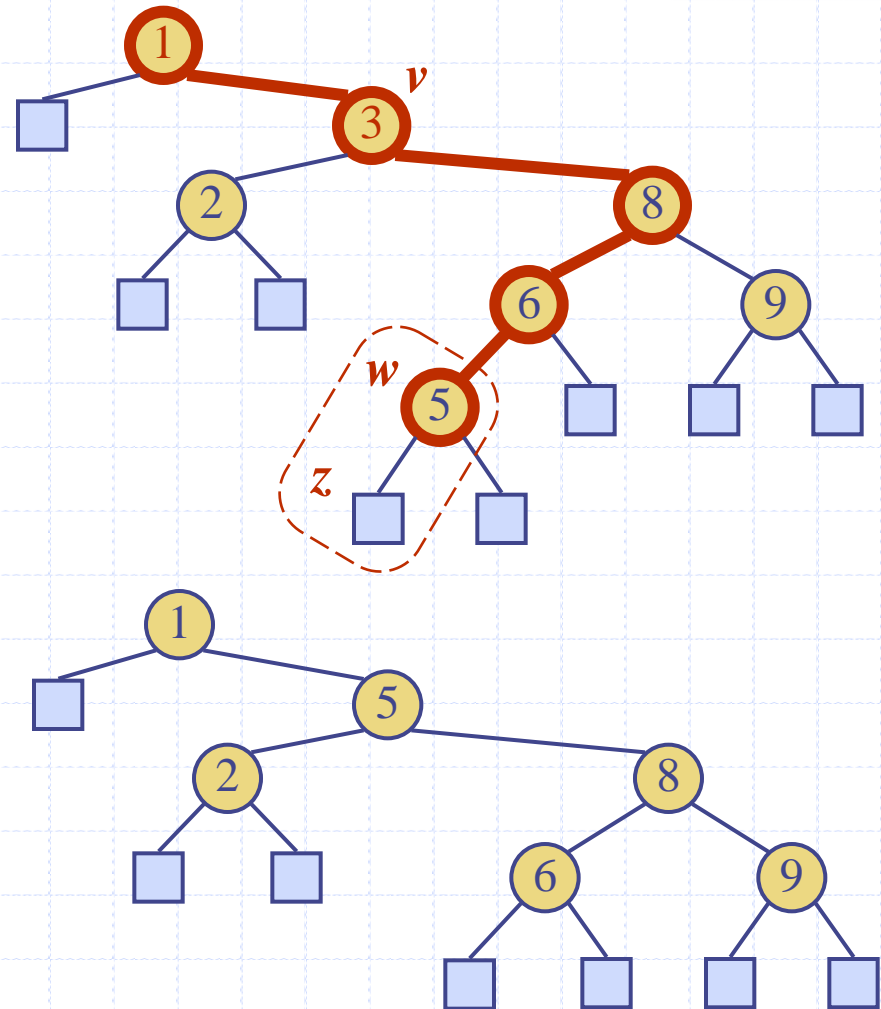
- ◆ What if the key  $k$  to be removed has **two** internal nodes as children, e.g. “remove 3”
  - we find the internal node  $w$  that follows  $v$  in an inorder traversal





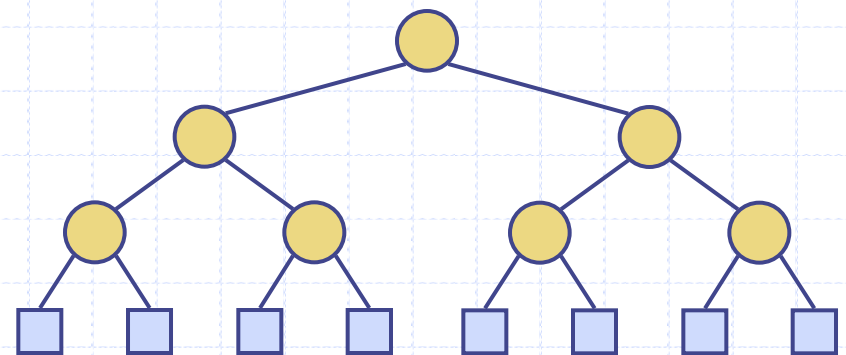
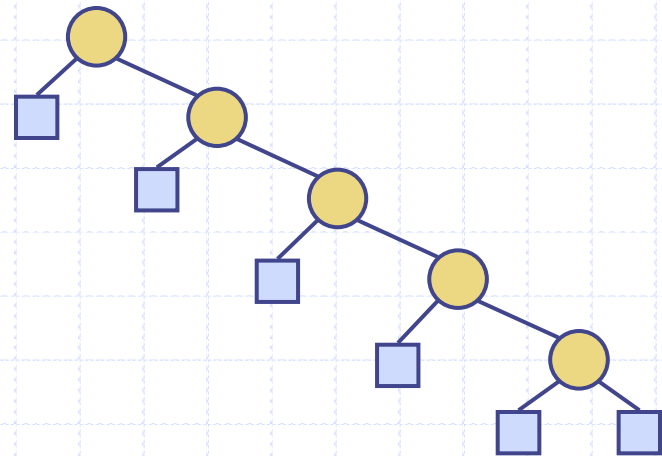
# Deletion: two children

- ◆ What if the key  $k$  to be removed has **two** internal nodes as children, e.g. "remove 3"
  - we find the internal node  $w$  that follows  $v$  in an inorder traversal
  - we copy  $key(w)$  into node  $v$
  - we remove node  $w$  and its left child  $z$  (which must be a leaf) by means of operation **removeAboveExternal( $z$ )**



# Performance

- ◆ A dictionary with  $n$  items implemented with a binary search tree of height  $h$ 
  - Space is:
    - ◆  $O(n)$
  - Time **findElement()** , **insertItem()** and **removeElement()** is:
    - ◆  $O(h)$  time
- ◆ Height  $h$  is  $O(n)$  in the worst case and  $O(\log n)$  in the best case
- ◆ How can we “prevent”  $O(n)$  height?



# Searching++

◆ That brings us to our next (more complex) set of searching algorithms and data structures:

- 2-3-4 trees
- (AVL trees)
- Red-black trees