

CSCI 136  
Data Structures &  
Advanced Programming

Priority Queues  
Introduction & Implementations

# Priority Queues

- Priority Queues
  - Supports Add & Remove (Min) operations
- Heaps
  - A “somewhat-ordered” data structure
    - Conceptual structure
    - Efficient implementations
      - Array Representations of (Binary) Trees

# A New Data Structure

Goal: Design a structure  $S$  to hold items with *priorities*

- $S$  should support operations
  - `add(E item); // add an item`
  - `E remove(); // remove highest priority item`
- $S$  should be designed to make these two operations fast

Such structures are called *Priority Queues*

# Priority Queues

- Priority queues are used for:
  - Scheduling processes in an operating system
    - Priority is function of time waiting + process priority
  - Order services on server
    - Backup is low priority, so don't do when high priority tasks need to happen
  - Scheduling future events in a simulation
  - Medical waiting room
  - Huffman codes - order by tree size/weight
  - A variety of graph/network algorithms

# Priority Queues

- Name is misleading: They are *not* queues
- Always remove object with *highest priority* regardless of when it was enqueued
- Data can be received/inserted in any order, but it is always returned/removed according to priority
- Like ordered structures (i.e., `OrderedVectors` and `OrderedLists`), PQs require comparisons of values

# On Terminology

- In colloquial English, the phrases "highest priority" and "number 1 priority" are used interchangeably
- So keep in mind that, often  
Higher Priority = Smaller Value
- A PQ removes the *smallest* value in an ordering: that is, the *highest priority* value!

# PQ Interface

```
public interface PriorityQueue<E extends Comparable<E>> {  
    public E getFirst(); // peeks at minimum element  
    public E remove(); // removes minimum element  
    public void add(E value); // adds an element  
    public boolean isEmpty();  
    public int size();  
    public void clear();  
}
```

# Notes on PQ Interface

- Unlike previous structures, we do not extend any other interfaces for many reasons
  - Random access is prohibited
  - Removal of arbitrary values is prohibited
- **PriorityQueue uses Comparable**
  - methods use Comparable parameters and
  - methods return Comparable values
- Could be made to use Comparators instead...



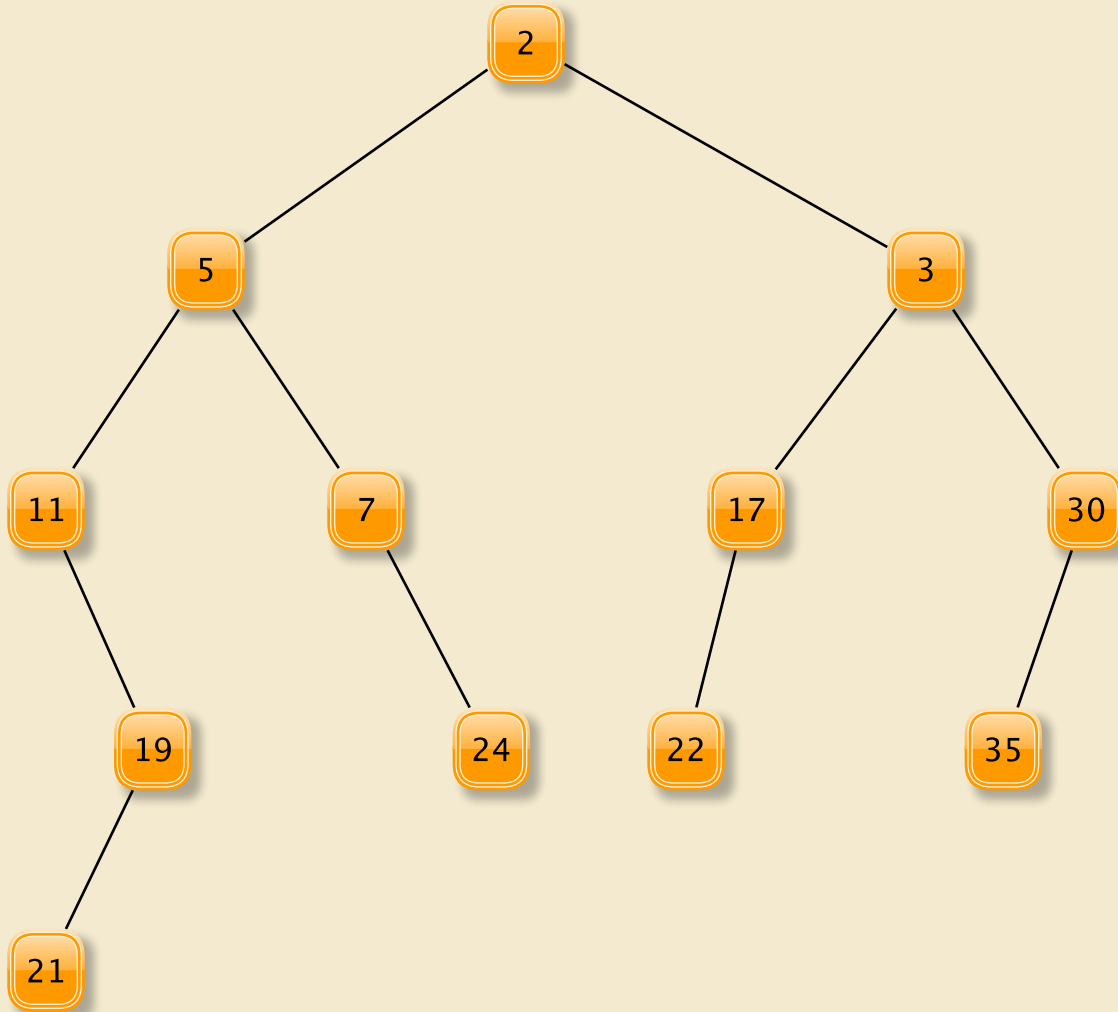
# Implementing PQs

- OrderedVector?
  - Keep ordered vector of objects
  - $O(n)$  to add/remove from vector
  - Can we do better than  $O(n)$ ?
- Binary Search Tree
  - Would need to be balanced for good performance
  - Would get  $O(\log n)$  which is very good
- Could relaxing requirements of total ordering help
  - Overhead of balancing might be avoided
- Heap!
  - Partially ordered binary tree

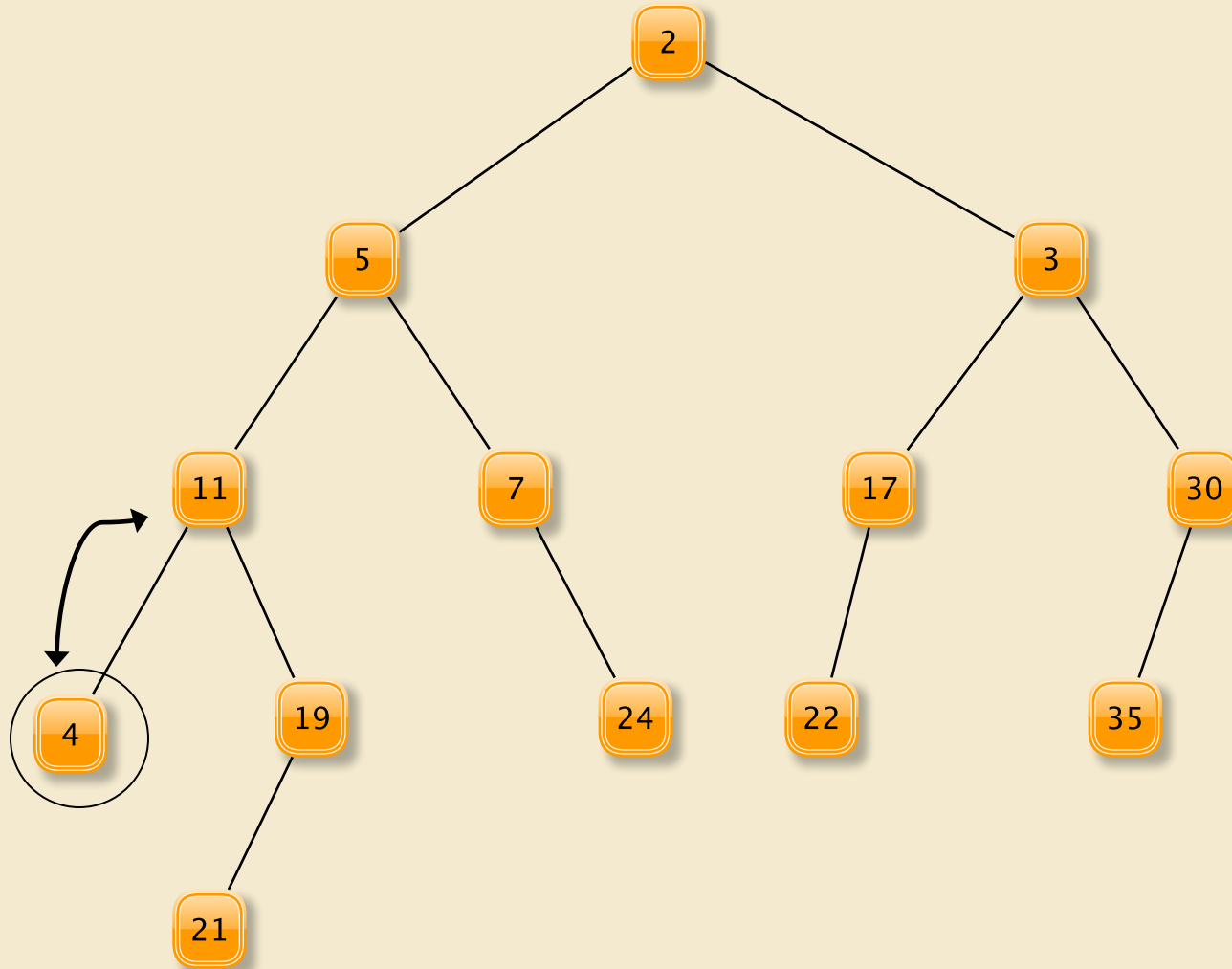
# Heap (aka Min-Heap)

- A heap is a special type of binary tree
- A heap is a binary tree where:
  - Root holds smallest (highest priority) value
  - Subtrees are also heaps (this is crucial!)
- So values increase in priority (decrease in value) from leaves to root (from descendant to ancestor)
- *Alternate definition:* A tree is a heap if and only if
  - For all nodes:  $\text{node.value()} \geq \text{node.parent.value()}$ 
    - This is called the *heap property* or the *heap invariant*
- Several valid heaps for same data set (no unique representation)
  - Note: variants allow more than 2 children per node

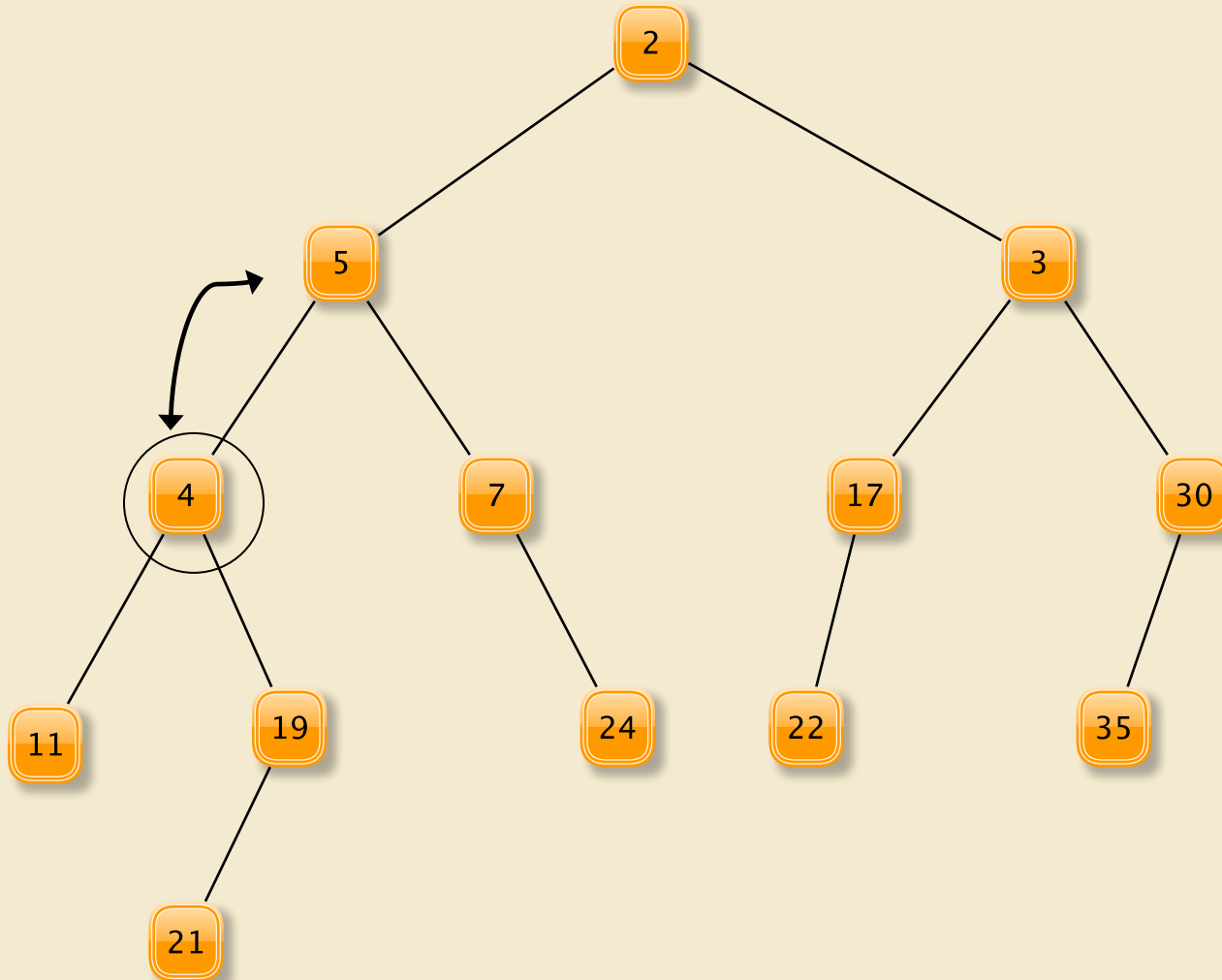
# Inserting into a Heap



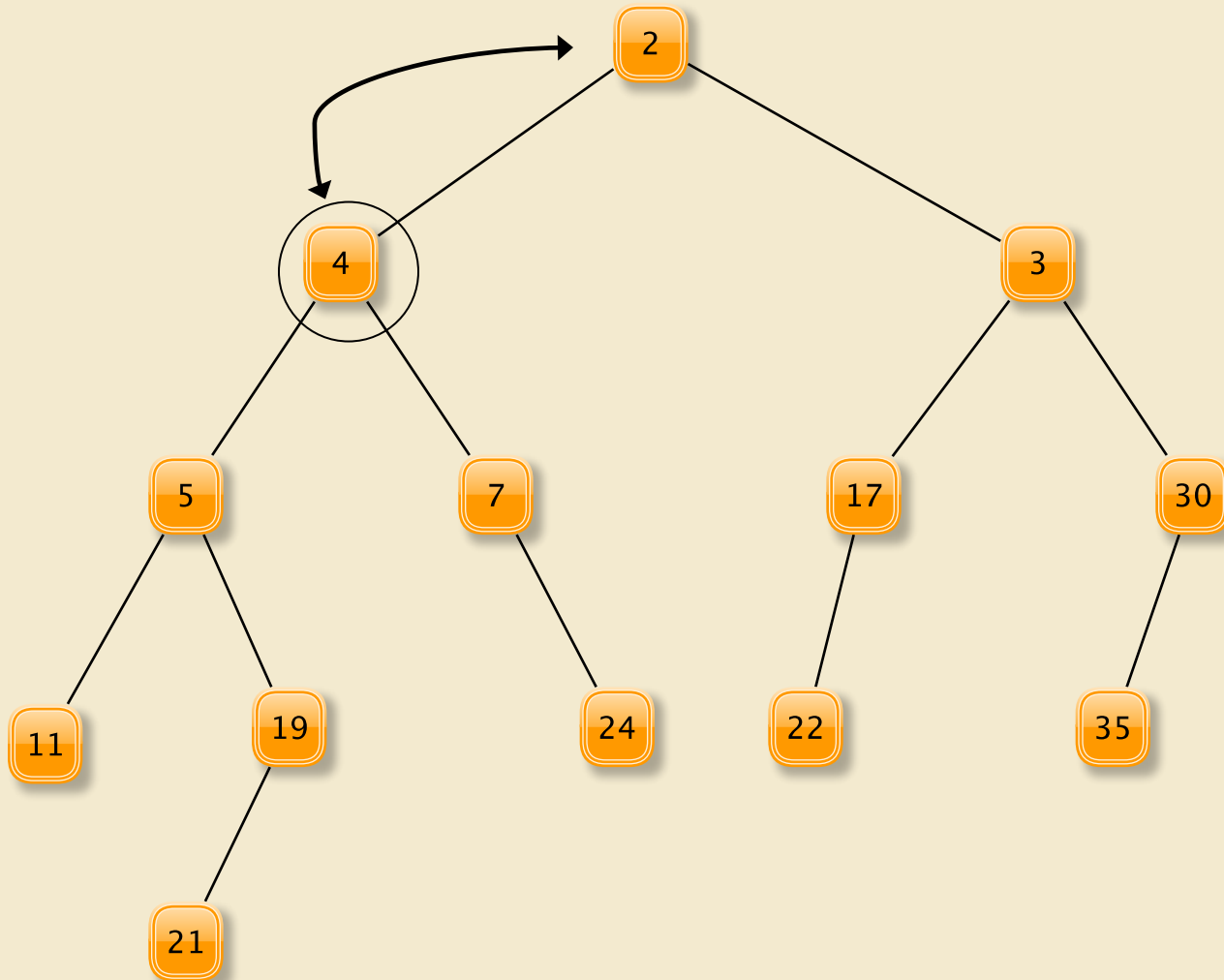
# Inserting into a Heap



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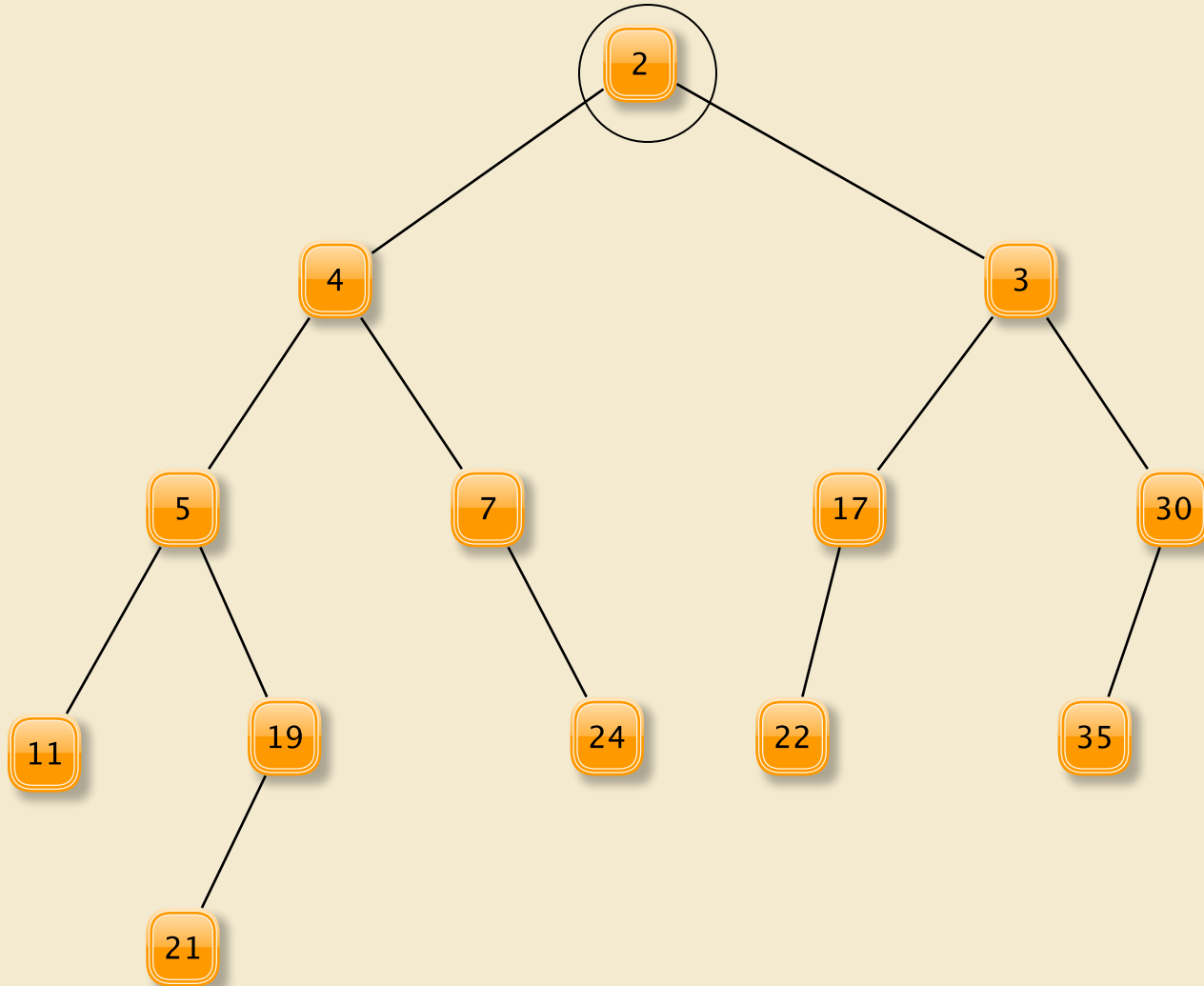
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# Inserting into a Heap

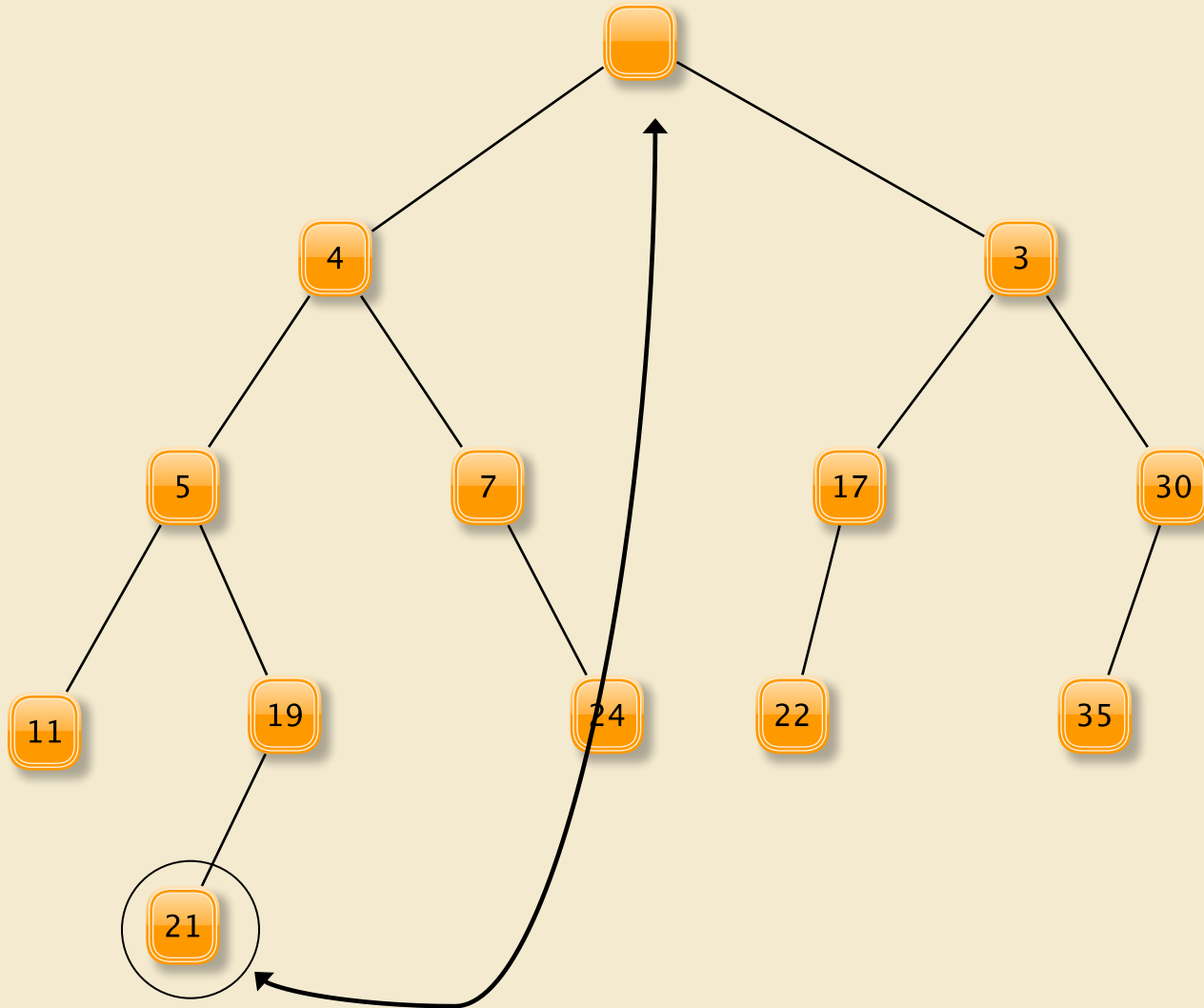
- Add new value as a leaf
- “Percolate” it up the tree
  - while (value < parent’s value) swap with parent
- This operation preserves the heap property since new value was the only one violating heap property
- Efficiency depends upon speed of
  - Finding a node at which to add new child
  - Finding location of parent
  - Tree height

# Removing Min From a PQ

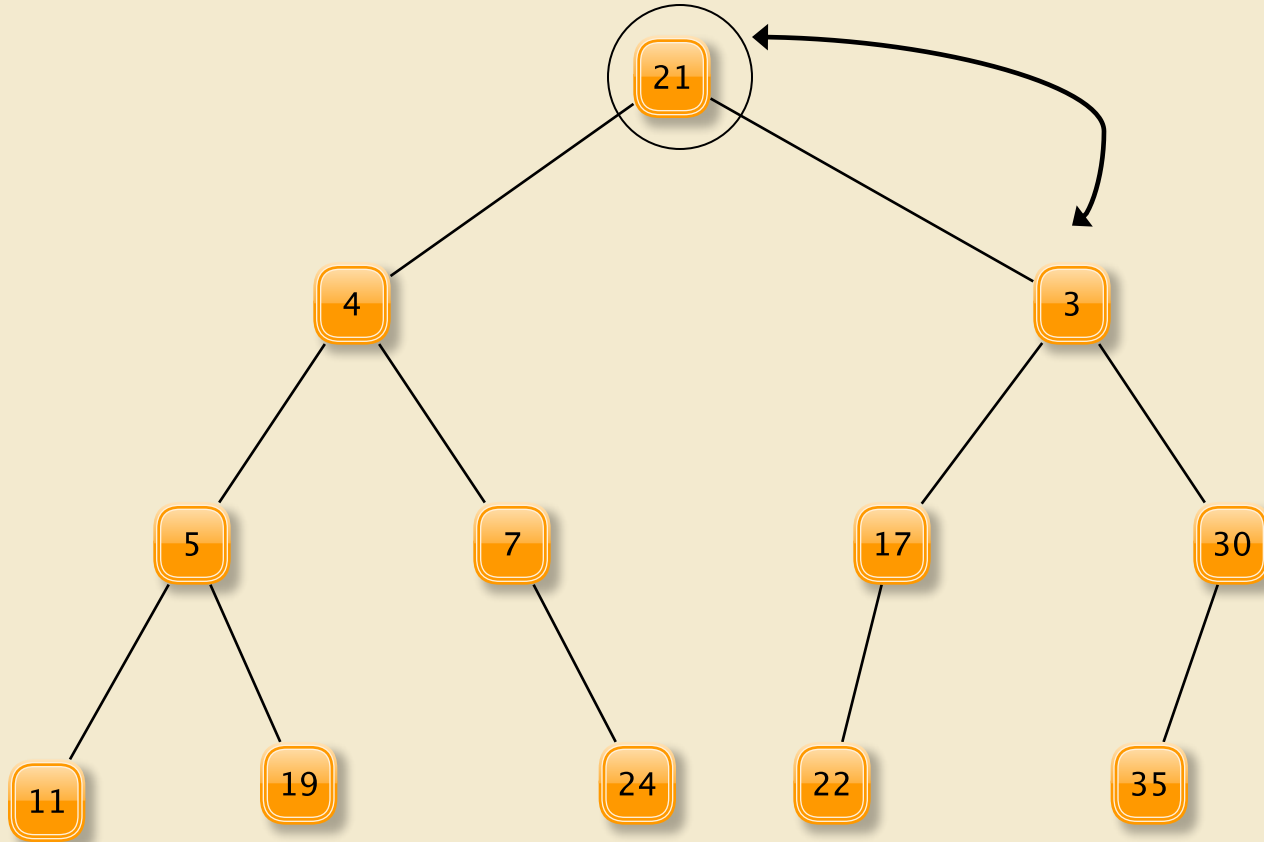




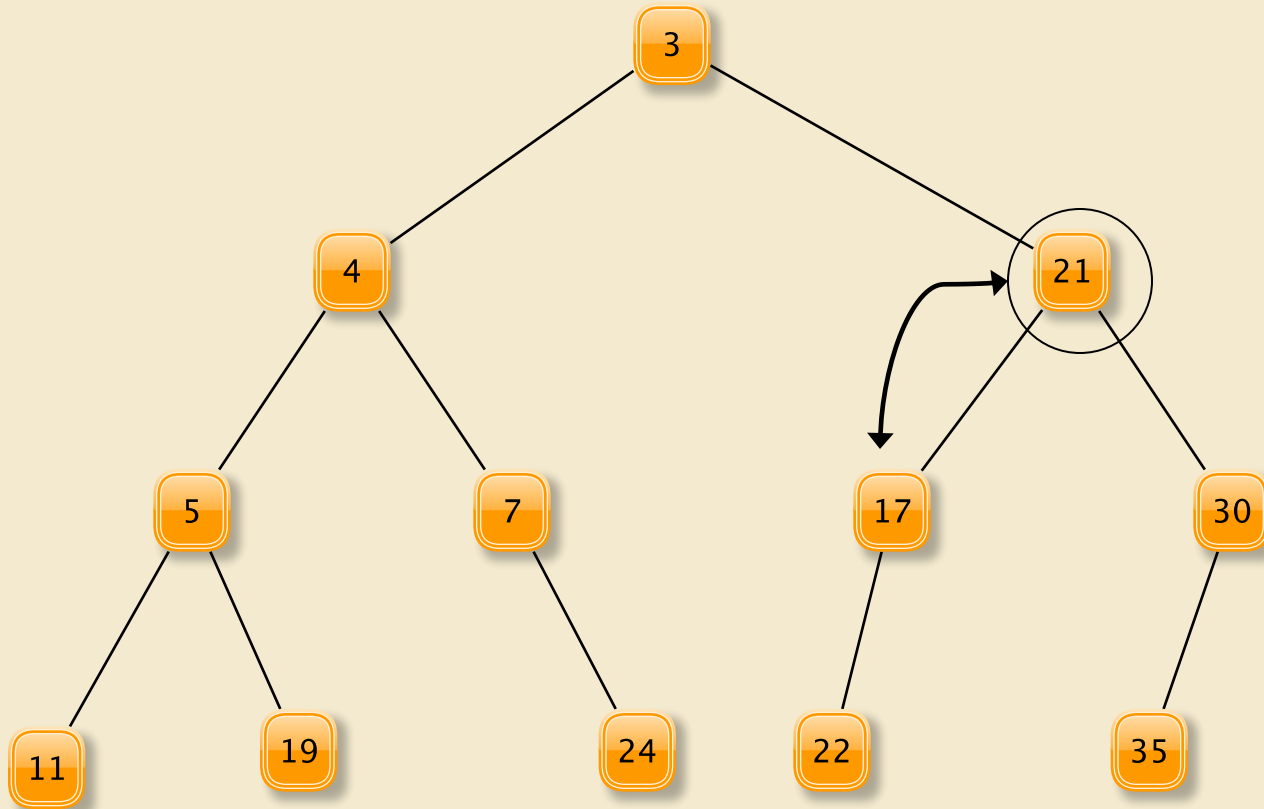
# Removing Min From a PQ



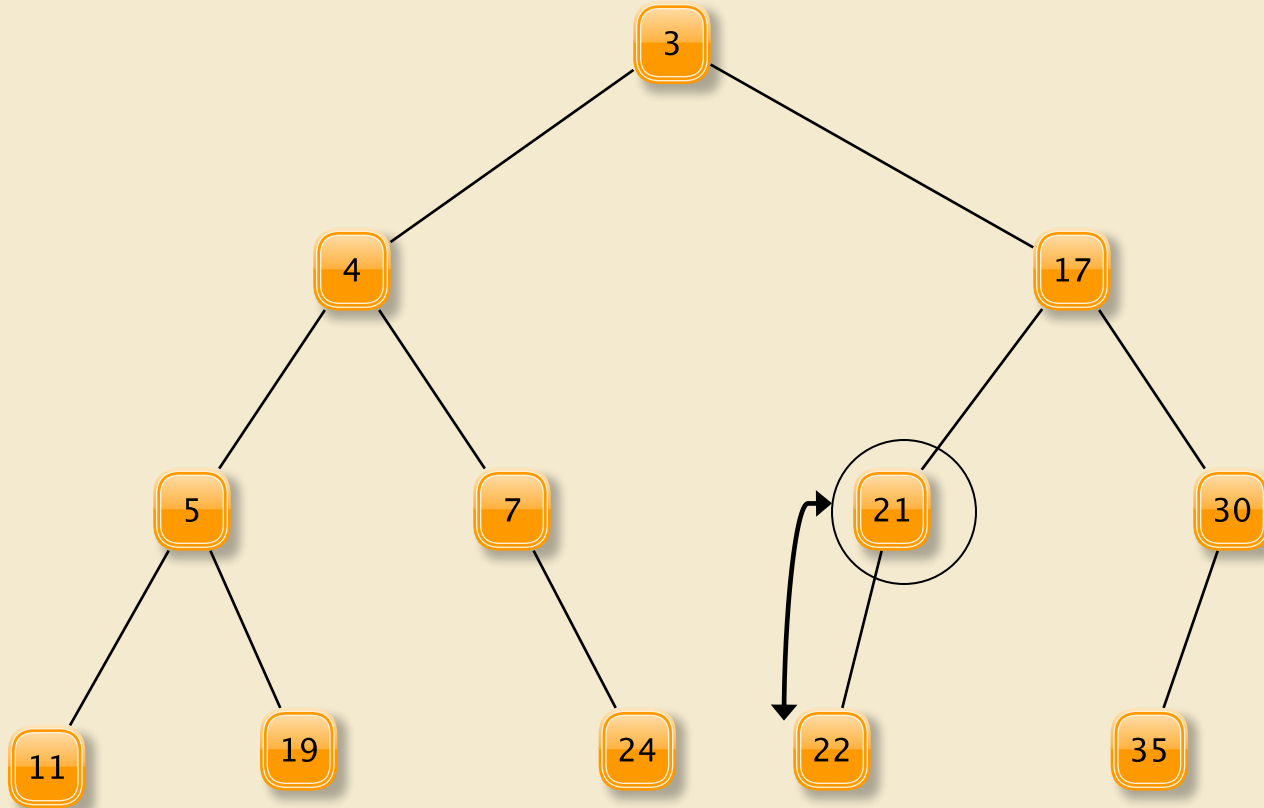
# Removing Min From a PQ



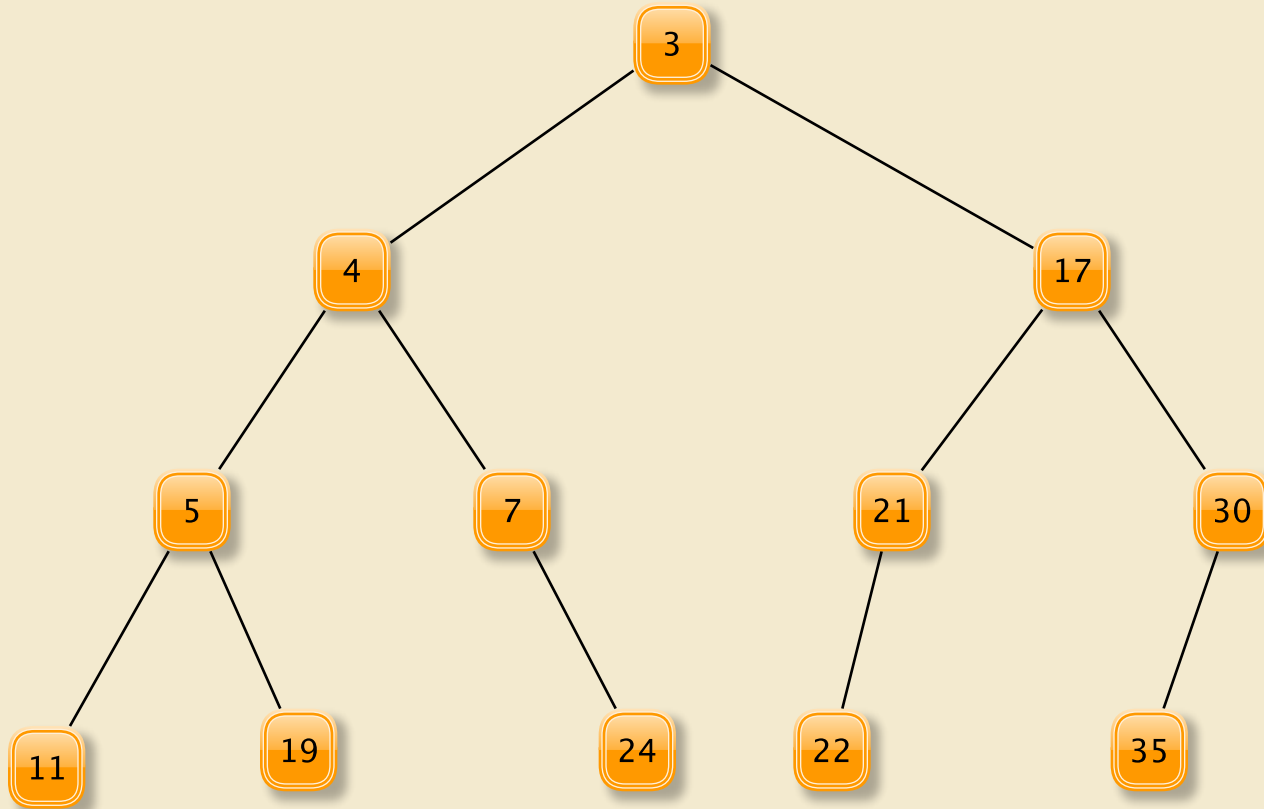
# Removing Min From a PQ



# Removing Min From a PQ



# Removing Min From a PQ



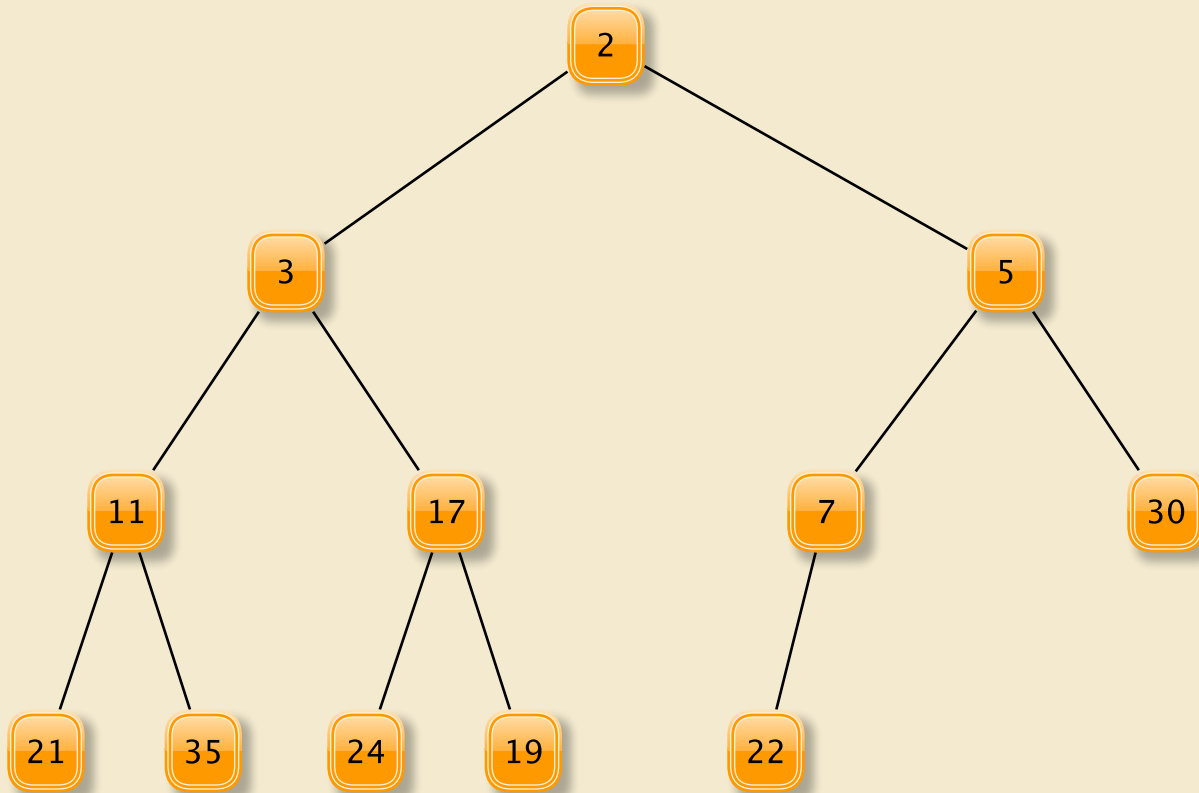
# Removing Root From a PQ

- Copy root value, save it to return
- Find a leaf, delete it, put its *data* in the root
- “Push” *data* down through the tree
  - while ( *data.value* > value of (at least) one child )
    - Swap *data* with data of *smaller* child
- This operation preserves the heap property
- Efficiency depends upon speed of
  - Finding a leaf
  - Finding locations of children
  - Height of tree

# Key Operations/Properties

- Insert efficiency depends upon speed of
  - Finding a node at which to add new child
  - Finding location of parent
  - Tree height
- RemoveMin efficiency depends upon speed of
  - Finding a leaf
  - Finding locations of children
  - Tree Height
- Goal: Find tree structure to optimize these

# Array-Based Binary Trees



2	3	5	11	17	7	30	21	35	24	19	22
0	1	2	3	4	5	6	7	8	9	10	11



# Array-Based Binary Trees

- Encode structure of tree in array indexes
  - Put root at index 0
  - Leave empty slots if no child
- Where are children of node  $i$ ?
  - Children of node  $i$  are at  $2i+1$  and  $2i+2$
  - Look at example
- Where is parent of node  $j$ ?
  - Parent of node  $j$  is at  $(j-1)/2$

# Recall : ArrayTrees

- Why are ArrayTrees good?
  - Save space for links
  - No need for additional memory allocated/garbage collected
  - Works well for *full* or *complete* trees
    - Complete: All levels except last are full and all gaps are at right
    - “A *complete* binary tree of height  $h$  is a full binary tree with 0 or more of the rightmost leaves of level  $h$  removed”
    - No empty slots!
- Why is this an OK assumption for us? Most trees are not complete...
- Insight: We can guarantee that our heap is always a complete tree by smart add/remove choices!

# Implementing Heaps

- VectorHeap
  - Use conceptual array representation of BT (ArrayTree)
  - But use extensible Vector instead of array (makes adding elements easier)
  - Note:
    - Root of tree is location 0 of Vector
    - Children of node in location  $i$  are in locations  $2i+1$  (left) and  $2i+2$  (right)
    - Parent of node  $i$  is in location  $(i-1)/2$

# Implementing Heaps

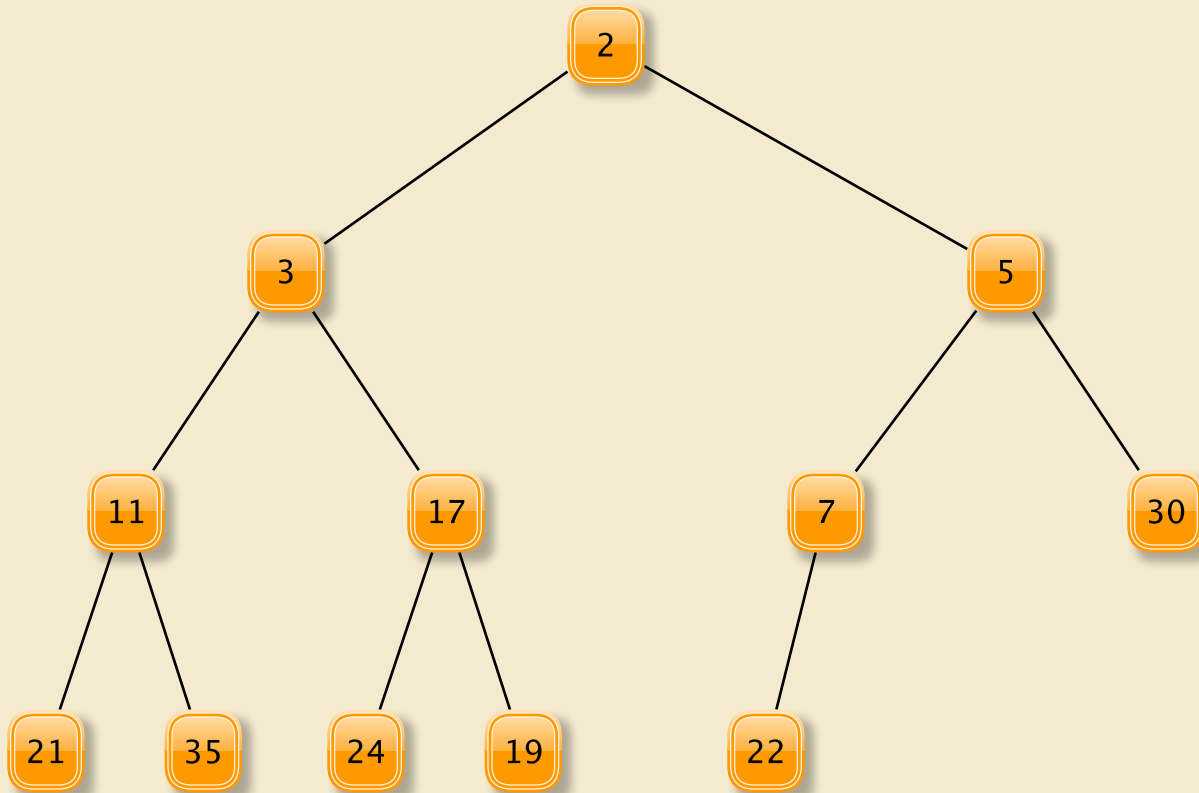
- Features
  - Guarantee no gaps in array (array is *complete*)
    - Always add in next available array slot (left-most available spot in binary tree);
    - Always remove using “right-most” leaf
  - *Heap Invariant* becomes
    - $\text{data}[i] \leq \text{data}[2i+1]$ ;  $\text{data}[i] \leq \text{data}[2i+2]$  (or kids might be null)
  - When elements are added and removed, do small amount of work to “re-heapify”
    - How small? Note: finding a node’s child or parent takes constant time, as does finding “final” leaf or next slot for adding
    - Since this heap corresponds to a full binary tree, the depth of the tree is  $O(\log n)$ , so add/remove take  $O(\log n)$  time!

# Implementing Heaps

## Details

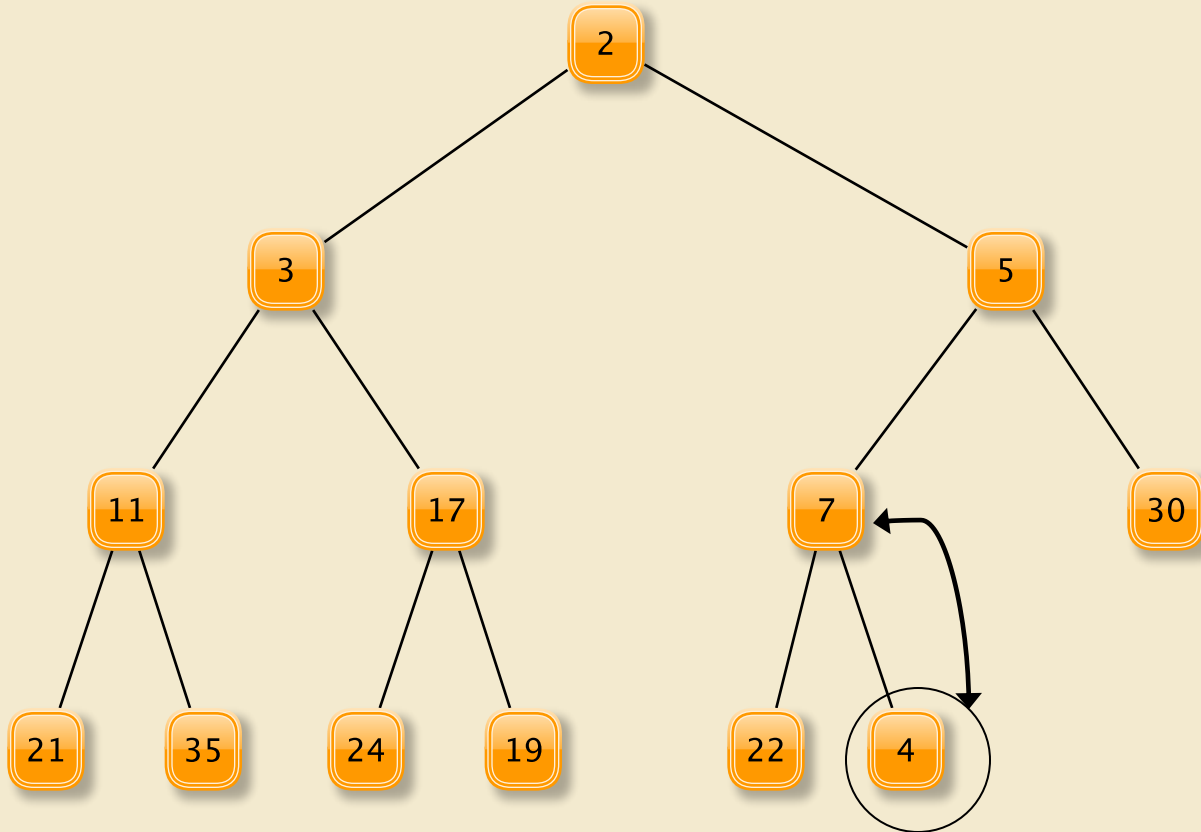
- Add method uses helper `percolateUp(int location)`
  - `percolateUp` moves newly inserted value up the tree until heap property is restored
- Remove method uses helper `pushDownRoot(int root)`
  - Moves value that remove moved from deleted leaf to root down the tree until heap property is restored
- Let's look some examples

# Example : Add(4)



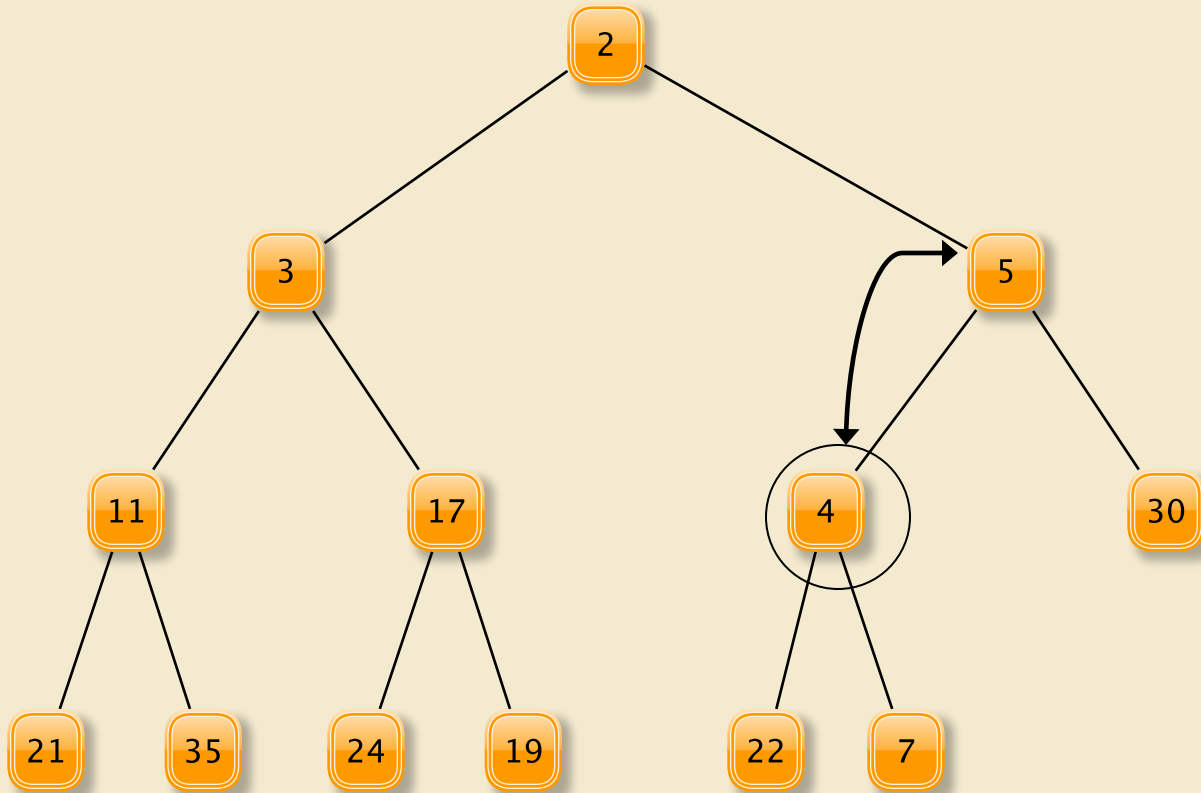
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	3	5	11	17	7	30	21	35	24	19	22	-	-	-

# Example : Add(4)



0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	3	5	11	17	7	30	21	35	24	19	22	4	-	-

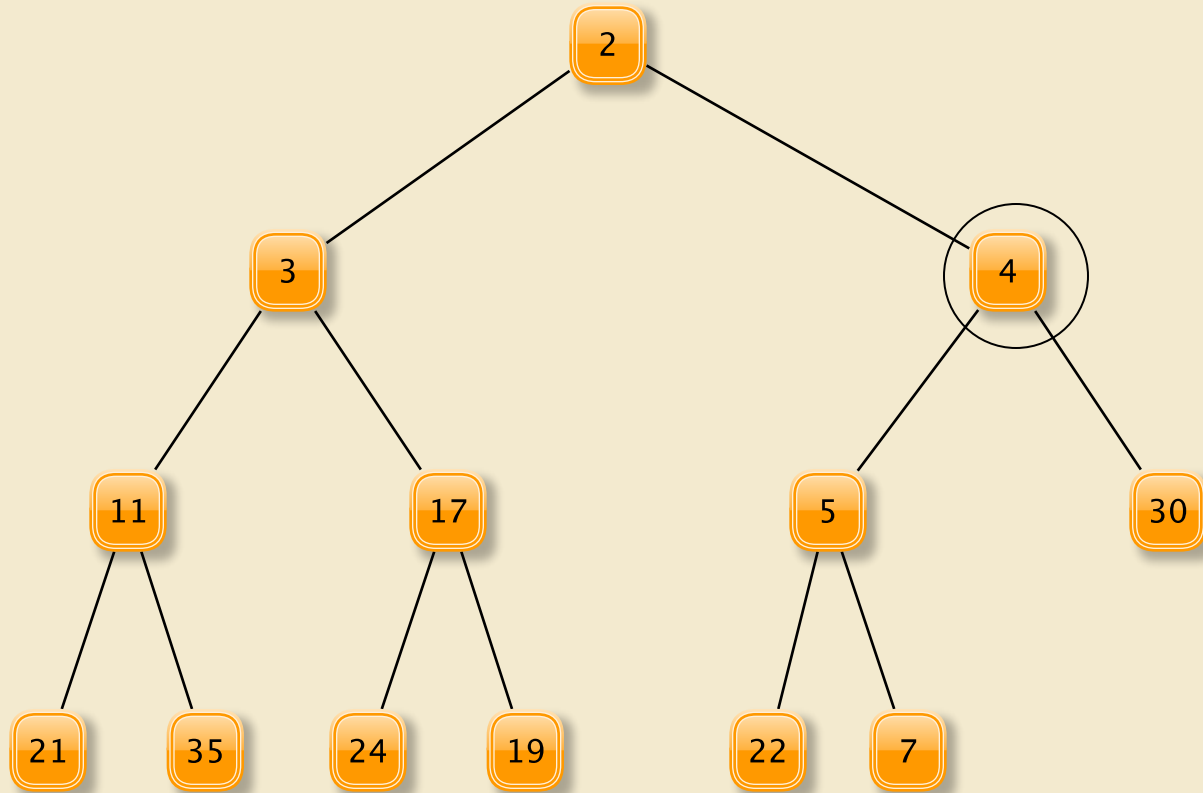
# Example : Add(4)



0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	3	5	11	17	4	30	21	35	24	19	22	7	-	-



# Example : Add(4)



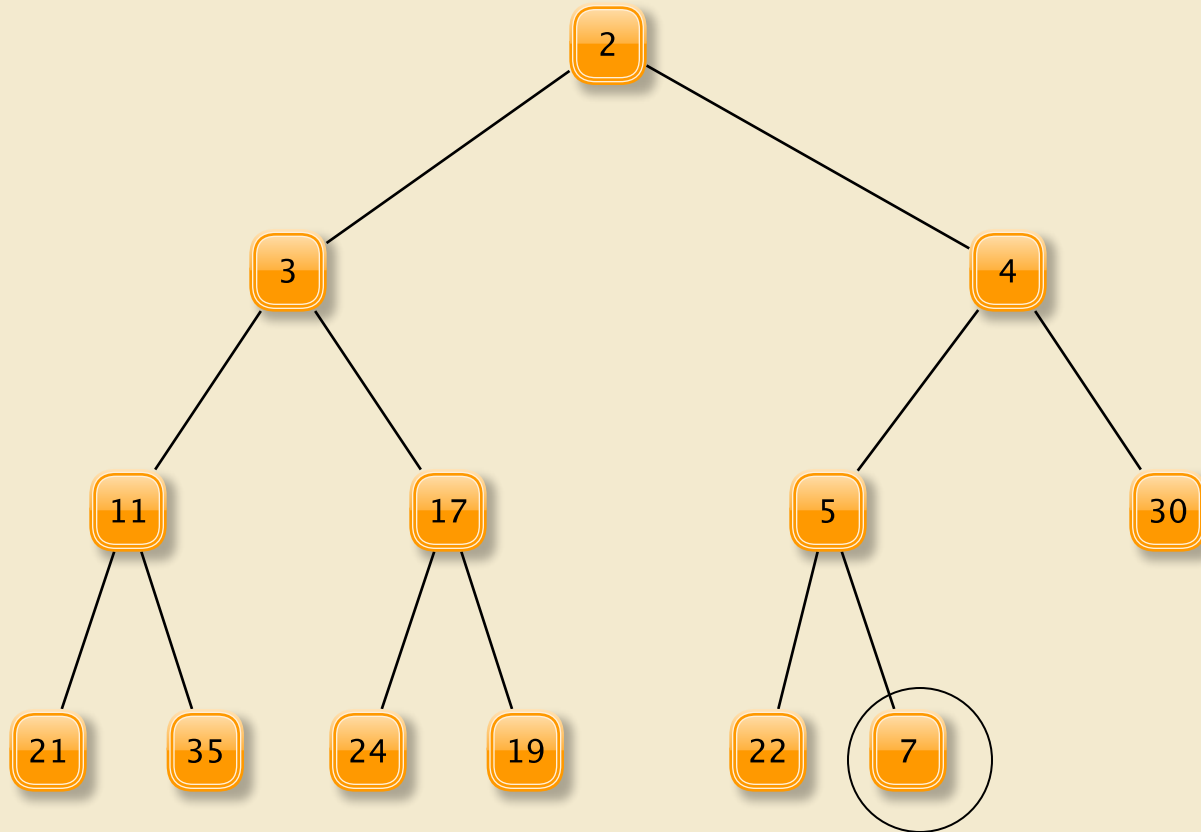
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	3	4	11	17	5	30	21	35	24	19	22	7	-	-

# Add : Uses PercolateUp

```
protected void percolateUp(int leaf) {
    int parent = parent(leaf);
    E value = data.get(leaf);
    while (leaf > 0 &&
        (value.compareTo(data.get(parent)) < 0)) {

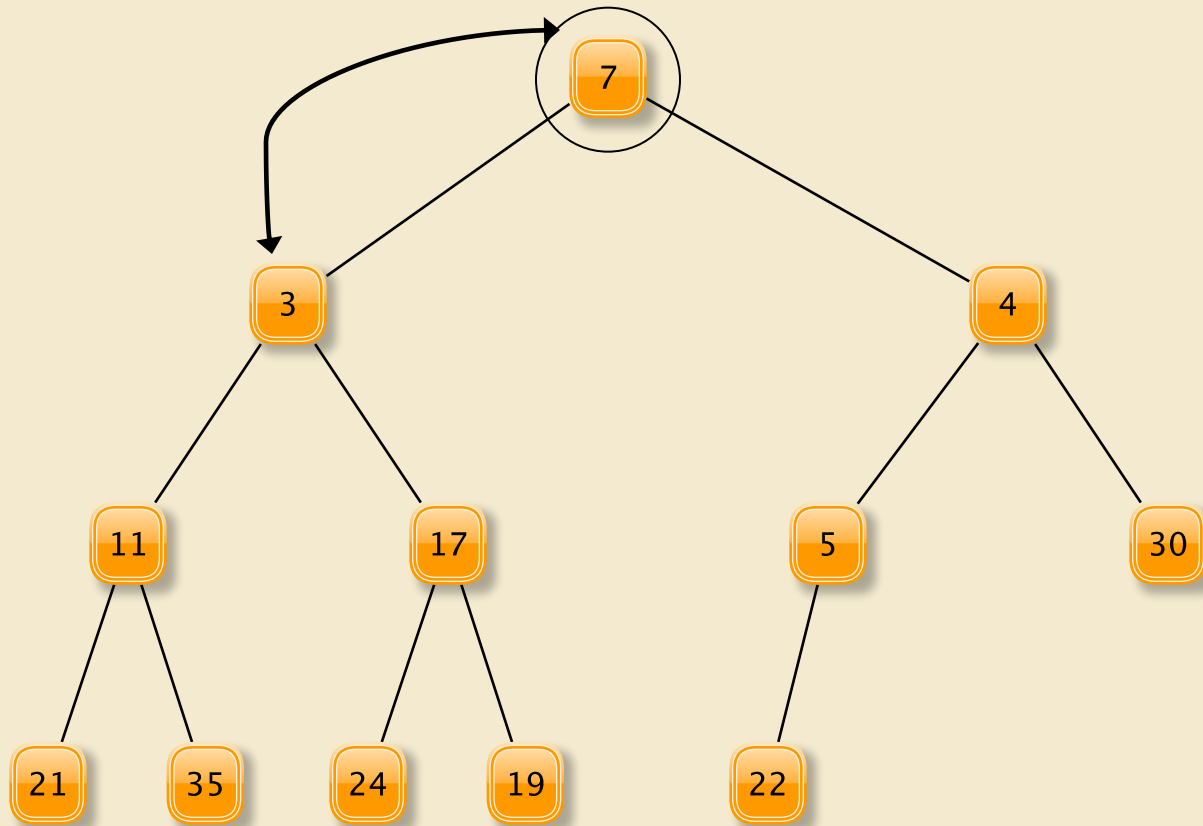
        data.set(leaf, data.get(parent));
        leaf = parent;
        parent = parent(leaf);
    }
    data.set(leaf, value);
}
```

# Example : Remove()



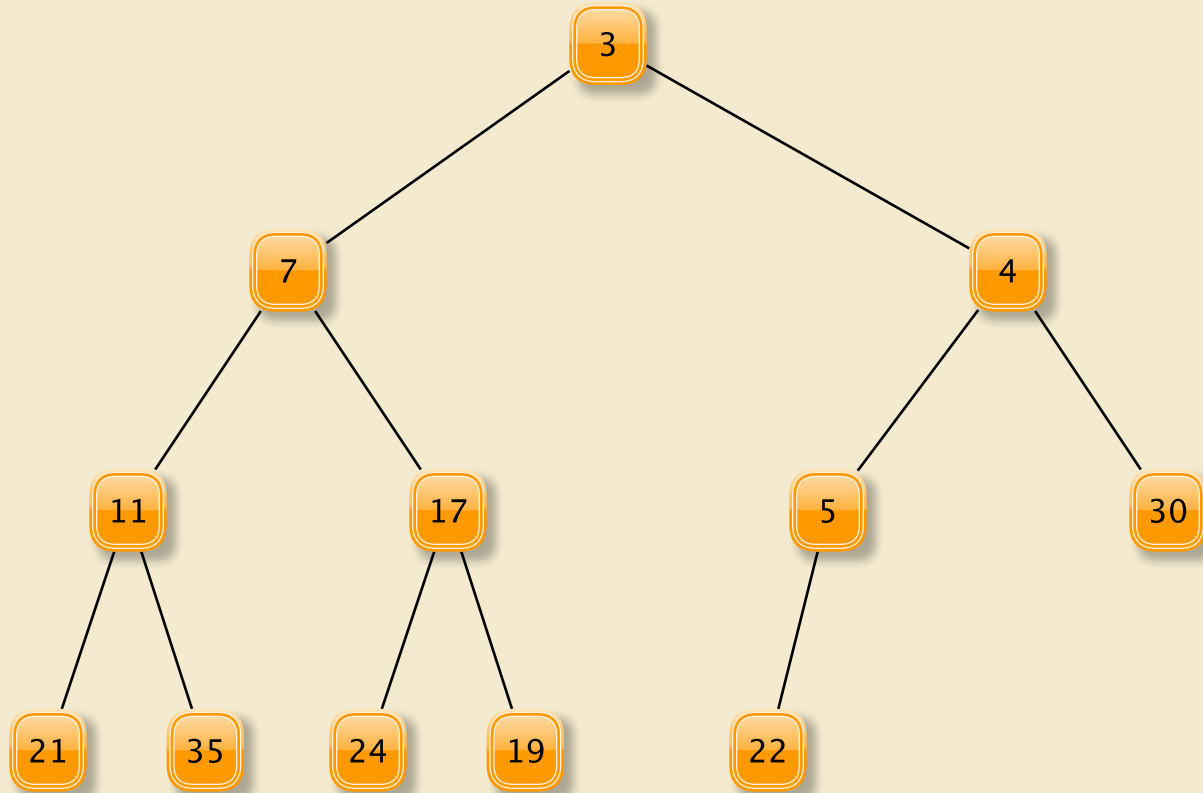
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
2	3	4	11	17	5	30	21	35	24	19	22	7	-	-

# Example : Remove()



0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
7	3	4	11	17	5	30	21	35	24	19	22	-	-	-

# Example : Remove()



0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
3	7	4	11	17	5	30	21	35	24	19	22	-	-	-

# Remove : Uses PushDownRoot

```
protected void pushDownRoot(int root) {
    int heapSize = data.size();
    E value = data.get(root);
    while (root < heapSize) {
        int childpos = left(root);
        // If node has left child
        if (childpos < heapSize) {
            // If right child has smaller value
            if ((right(root) < heapSize) &&
                ((data.get(childpos+1)).compareTo
                 (data.get(childpos)) < 0)) {
                childpos++;
            }
        }
    }
}
```

# Remove : Uses PushDownRoot

```
// Assert: childpos indexes smaller child
// Compare child to value being pushed down
if((data.get(childpos)).compareTo(value)<0){
    data.set(root,data.get(childpos));
    root = childpos; // keep moving down
} else { // found right location
    data.set(root,value);
    return;
}
} else { // at a leaf! insert and halt
    data.set(root,value);
    return;
} } }
```

# VectorHeap Summary

- Add/Remove are both  $O(\log n)$
- Data is not completely sorted
  - “Partial” order is maintained
- Note: `VectorHeap(Vector<E> v)`
  - Takes an unordered Vector and uses it to construct a heap
  - How
    - Uses `VectorHeap add` method to insert elements of `v`
    - This builds the `VectorHeap` in  $O(n \log n)$  time
    - As always, we ask: Can we do better?



# Heapifying A Vector (or array)

- Method I: Top-Down
  - Assume  $V[0..k]$  satisfies the heap property
  - Now call percolate on item in location  $k+1$
  - Then  $V[0..k+1]$  satisfies the heap property
- Method II: Bottom-up
  - Assume  $V[k..n]$  satisfies the heap property
  - Now call pushDown on item in location  $k-1$
  - Then  $V[k-1..n]$  satisfies heap property

# Top-Down vs Bottom-Up

- Top-down heapify: elements at depth  $d$  may be swapped  $d$  times: Total # of swaps is at most

$$\sum_{d=0}^h d2^d = (h - 1)2^{h+1} + 2 = (\log n - 1)2n + 2$$

- This is  $O(n \log n)$
- Some intuition: most of the elements are in the lowest levels of the tree, so each of them might have to move to root:  $O(\log n)$  swaps per element

# Top-Down vs Bottom-Up

- Bottom-up heapify: elements at depth  $d$  may be swapped  $h-d$  times: Total # of swaps is at most

$$\sum_{d=0}^h (h-d)2^d = 2^{h+1} - h - 2 = 2n - \log n + 2$$

- This is  $O(n)$  --- beats top-down!
- Some intuition: most of the elements are in the lowest levels of the tree, so each of them will only be pushed down (swapped) a small number of times **SO COOL!!!**

# HeapSort

- Heaps yield another  $O(n \log n)$  sort method
- To HeapSort a Vector “in place”
  - Perform bottom-up heapify on the reverse ordering: that is: highest rank/lowest priority elements are near the root (low end of Vector)
  - Now repeatedly remove elements to fill in Vector from tail to head
    - For(`int i = v.size() - 1; i > 0; i--`)
      - RemoveMin from `v[0..i]` // `v[i]` is now not in heap
      - Put removed value in location `v[i]`