

# CSE 332: Data Structures and Parallelism

Spring 2022

Richard Anderson

Lecture 4: Priority Queues

# Announcements

- Reading: Weiss, for Monday and Wednesday
  - Priority Queues, 6.1-6.5
- Ex 2 due Friday
- Checkpoint for P1 on Apr 6

# Today

- Leftovers: Algorithm run time and big Oh
- Priority Queues
- Binary Tree Implementation
- Array implementation

# Summary of Friday

- We need a rigorous way of talking about the performance of an algorithm
  - Real world to math world
- Ideas
  - Measure run time on an input by counting “steps”
  - Run time function for an algorithm,  $R(n)$ : Worst case run time on input of size  $n$
  - Only consider the rate of growth of run time functions by ignoring constants with big-Oh

# Asymptotic Analysis

Eliminate low order terms

$$- 4n + 5 \Rightarrow$$

$$- 0.5 n \log n + 2n + 7 \Rightarrow$$

$$- n^3 + 3 \cdot 2^n + 8n \Rightarrow$$

Eliminate coefficients

$$- 4n \Rightarrow$$

$$- 0.5 n \log n \Rightarrow$$

$$- 3 \cdot 2^n \Rightarrow$$

# Formal definition of Big-Oh

$h(n)$  is  $O(f(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $h(n) \leq c f(n)$  for all  $n \geq n_0$

# Big-Oh Example

$h(n)$  is  $O(f(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $h(n) \leq c f(n)$  for all  $n \geq n_0$

Example:

$$100n^2 + 1000 \leq 1 (n^3 + 2n^2) \text{ for all } n \geq 100$$

So  $100n^2 + 1000$  is  $O(n^3 + 2n^2)$

# Asymptotic Lower Bounds

$\Omega( g(n) )$  is used for functions asymptotically greater than or equal to  $g(n)$

$h(n)$  is  $\Omega( g(n) )$  if there exist  $c > 0$  and  $n_0 > 0$  such that  $h(n) \geq c g(n)$  for all  $n \geq n_0$

$h(n)$  is  $\theta( f(n) )$  is  $h(n)$  is  $O( f(n) )$  and  $h(n)$  is  $\Omega(f(n))$

# Logarithms

- $\log_A B$  is the solution of  $A^x = B$
- Default base of log can be 2, e, or 10
- Other bases can be important in CS
  - Height of a d-ary tree

# Properties of Logs

Basic:

- $A^{\log_A B} = B$
- $\log_A A =$

Independent of base:

- $\log(AB) =$
- $\log(A/B) =$
- $\log(A^B) =$
- $\log((A^B)^C) =$

# Change of base

$$\log_a N = \log_a (b^{\log_b N}) = \log_b N \log_a b$$

$$\log_b N = \frac{\log_a N}{\log_a b}$$

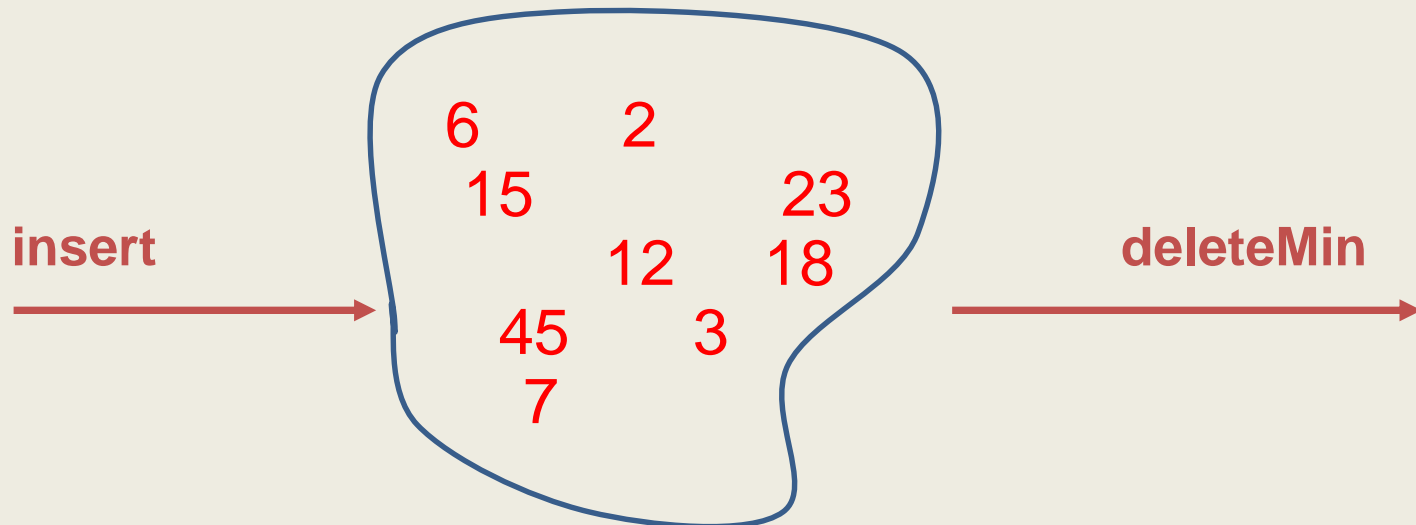
Base of the log can be ignored in big-Oh  
(provided the base is a constant)

# Priority Queues

Manage a set, with insert item and  
get highest priority item

# Priority Queue ADT

- Need a new ADT
- Operations: Insert an Item,  
Remove the “Best” Item



# Priority Queue ADT

1. PQueue data : collection of data with priority
2. PQueue operations
  - insert
  - deleteMin(also: create, destroy, is\_empty)
3. PQueue property: if  $x$  has lower priority than  $y$ ,  $x$  will be deleted before  $y$

# Potential Implementations

	Insert	DeleteMin
Unsorted list (Array)		
Unsorted list (Linked list)		
Sorted list (Array)		
Sorted list (Linked list)		
Binary Search Tree		

# Binary Heap data structure

- **binary heap** (a kind of binary tree) for priority queues:
  - $O(\log n)$  worst case for both insert and deleteMin
  - $O(1)$  average insert
- It's optimized for priority queues. Lousy for some other types of operations (e.g., searching, sorting)

# Tree Review

*root(T):* A

*leaves(T):* D-F, I-N

*children(B):* D-F

*parent(H):* G

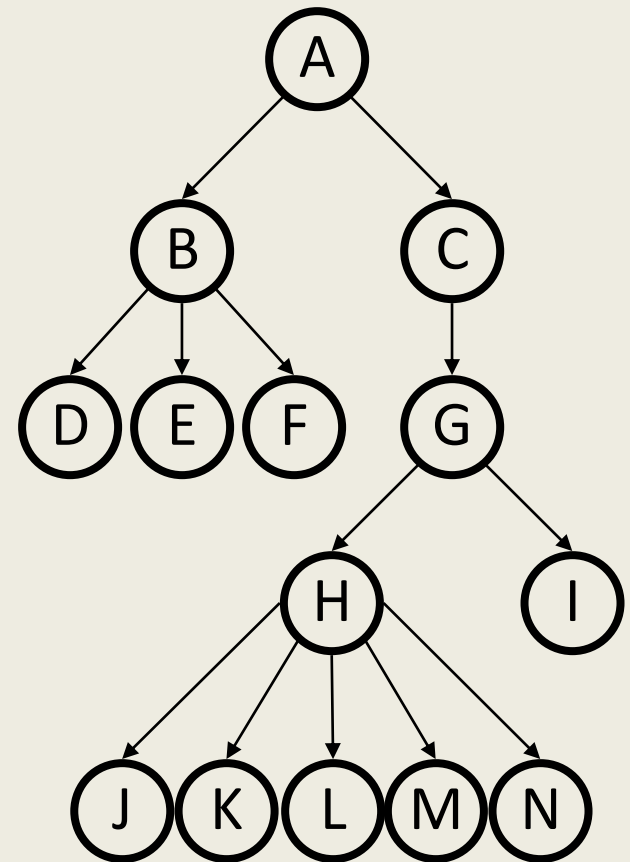
*siblings(E):* D,F

*ancestors(F):*

*descendants(G):*

*subtree(C):*

Tree T



# More Tree Terminology

*depth(B):*

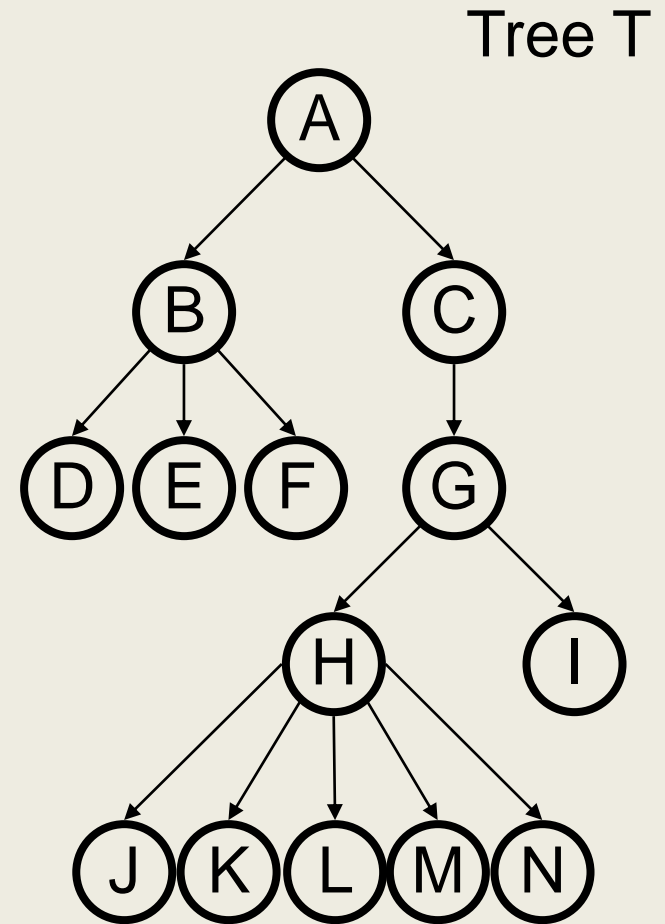
*height(G):*

*height(T):*

*degree(B):*

*branching  
factor(T):*

*n-ary tree:*



# Binary Heap Properties

A binary heap is a binary tree with two important properties that make it a good choice for priority queues:

1. **Completeness**
2. **Heap Order**

**Note:** we will sometimes refer to a binary heap as simply a “heap”.

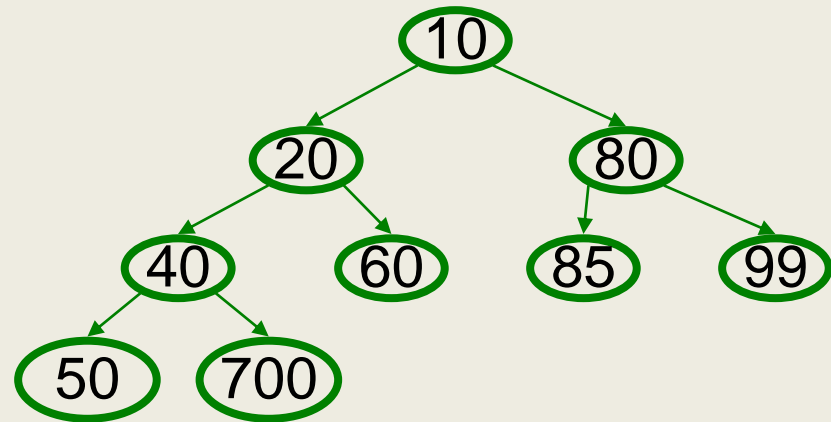
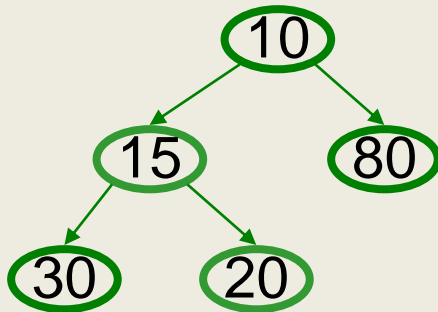
# Completeness

A binary heap is a complete binary tree

All levels are full, except the bottom, which is filled to the right

# Heap Order Property

Heap order property: For every non-root node  $X$ , the value in the parent of  $X$  is less than (or equal to) the value in  $X$ .



# Heap Operations

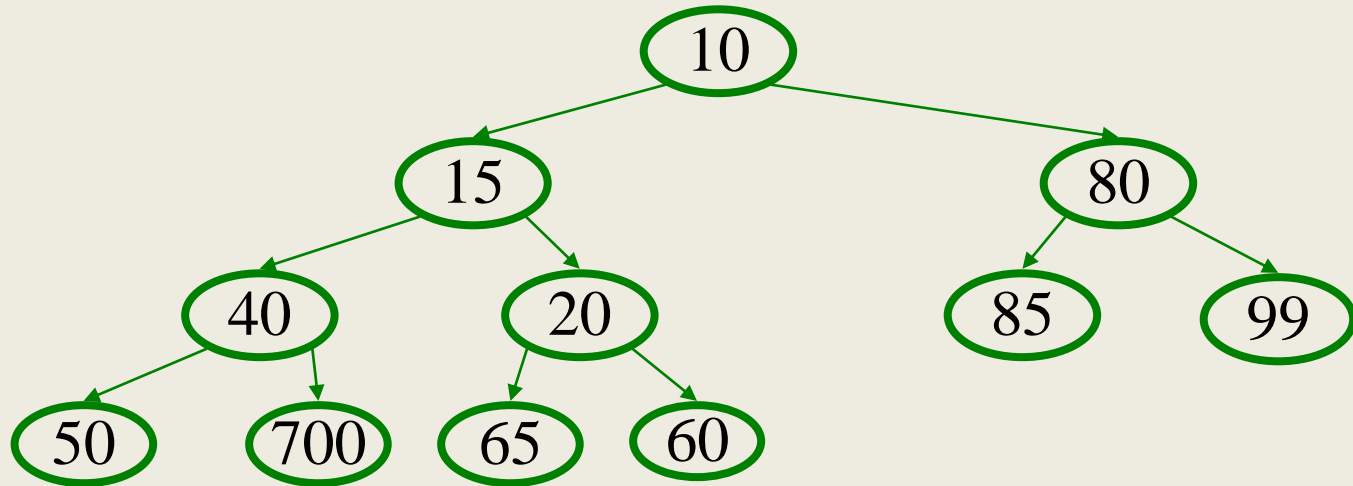
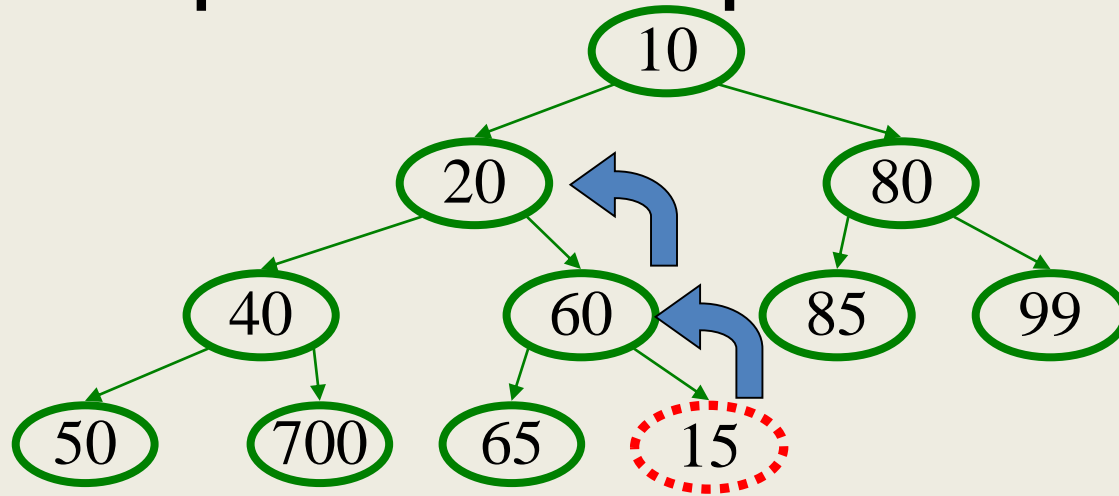
- Main operations: insert, deleteMin
- Key is to maintain
  - Completeness
  - Heap Order
- Basic idea is to propagate changes up/down the tree, fixing order as we go

# Heap – insert(val)

## Basic Idea:

1. Put val at last leaf position
2. Percolate up by repeatedly exchanging node with parent as long as needed

# Insert: percolate up

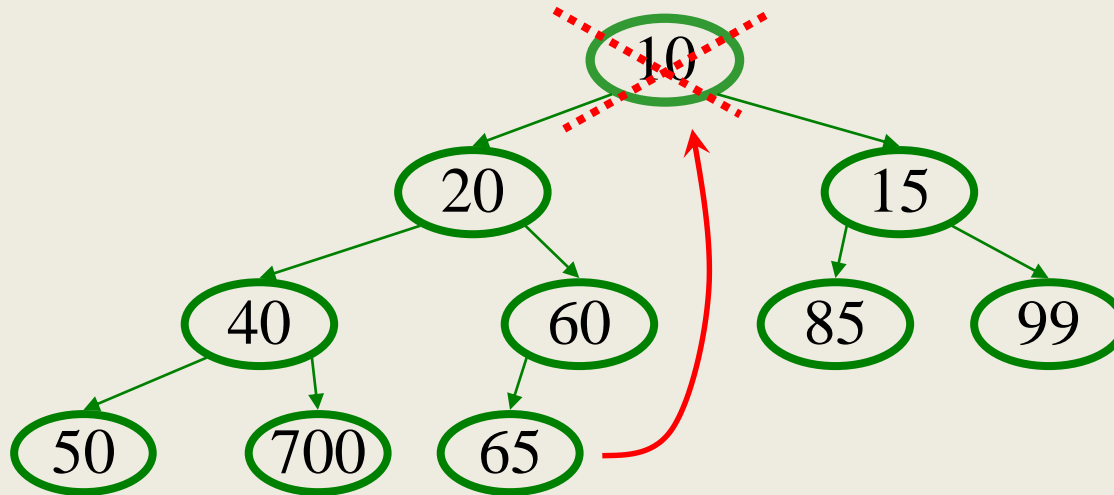


# Heap – deleteMin

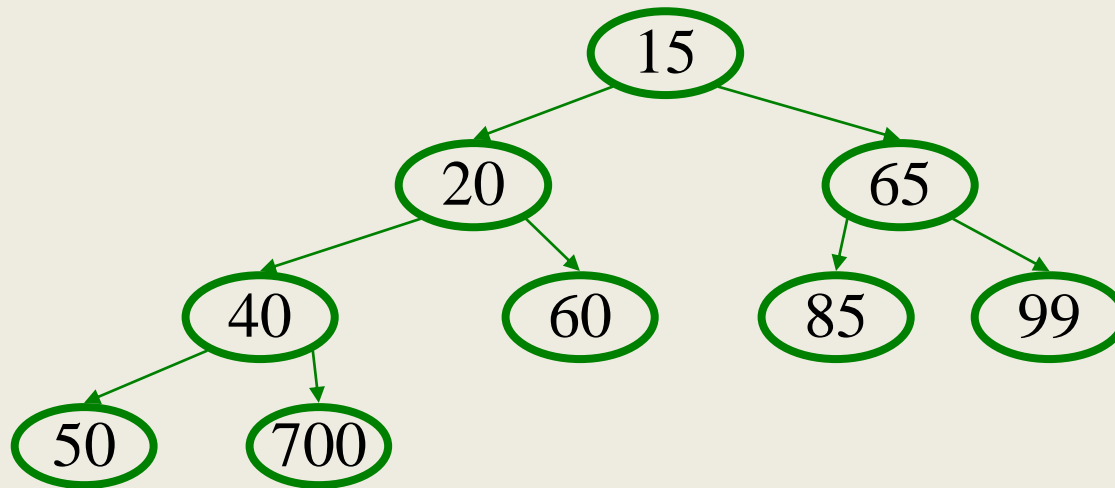
## Basic Idea:

1. Remove min element
2. Put “last” leaf node value at root
3. Find smallest child of node
4. Swap node with its smallest child if needed.
5. Repeat steps 3 & 4 until no swaps needed.

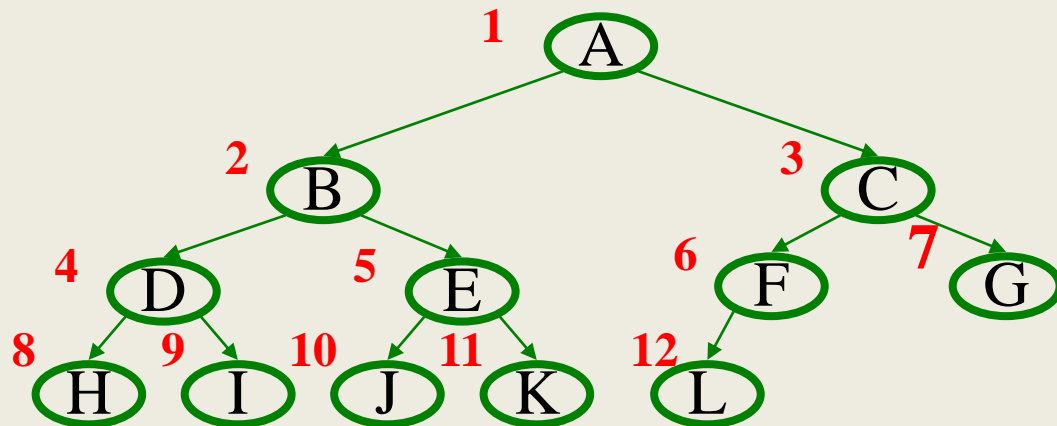
# DeleteMin: percolate down



# DeleteMin: percolate down



# Representing Complete Binary Trees in an Array



From node **i**:

left child:  
right child:  
parent:

	A	B	C	D	E	F	G	H	I	J	K	L	
0	1	2	3	4	5	6	7	8	9	10	11	12	13

# Why use an array?

# DeleteMin Code

```
Object deleteMin() {
    assert(!isEmpty());
    returnVal = Heap[1];
    size--;
    newPos =
        percolateDown(1,
            Heap[size + 1]);
    Heap[newPos] =
        Heap[size + 1];
    return returnVal;
}
```

*runtime:*

(Java code in book)

```
int percolateDown(int hole,
                  Object val) {
    while (2*hole <= size) {
        left = 2*hole;
        right = left + 1;
        if (right <= size &&
            Heap[right] < Heap[left])
            target = right;
        else
            target = left;

        if (Heap[target] < val) {
            Heap[hole] = Heap[target];
            hole = target;
        }
        else
            break;
    }
    return hole;
}
```

# Insert Code

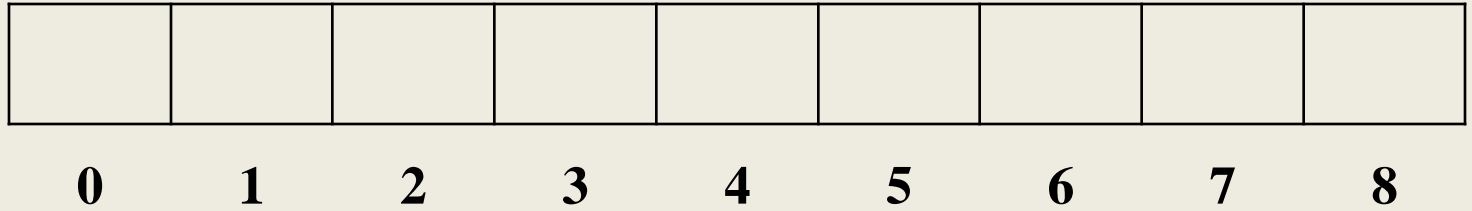
```
void insert(Object o) {
    assert(!isFull());
    size++;
    newPos =
        percolateUp(size, o);
    Heap[newPos] = o;
}

int percolateUp(int hole,
                Object val) {
    while (hole > 1 &&
           val < Heap[hole/2])
        Heap[hole] = Heap[hole/2];
        hole /= 2;
    }
    return hole;
}
```

*runtime:*

(Java code in book)

Insert: 16, 32, 4, 69, 105, 43, 2



# More Priority Queue Operations

## **decreaseKey(nodePtr, amount):**

given a pointer to a node in the queue, reduce its priority

Binary heap: change priority of node and \_\_\_\_\_

## **increaseKey(nodePtr, amount):**

given a pointer to a node in the queue, increase its priority

Binary heap: change priority of node and \_\_\_\_\_

**Why do we need a *pointer*? Why not simply data value?**

**Worst case running times?**

# More Priority Queue Operations

## **remove(objPtr):**

given a pointer to an object in the queue, remove it

Binary heap: \_\_\_\_\_

## **findMax( ):**

Find the object with the highest value in the queue

Binary heap: \_\_\_\_\_

## **Worst case running times?**

# More Binary Heap Operations

## **expandHeap( ):**

If heap has used up array, copy to new, larger array.

- Running time:

## **buildHeap(objList):**

Given list of objects with priorities, fill the heap.

- Running time:

We do better with **buildHeap...**

# Building a Heap: Take 1

12	5	11	3	10	6	9	4	8	1	7	2
----	---	----	---	----	---	---	---	---	---	---	---

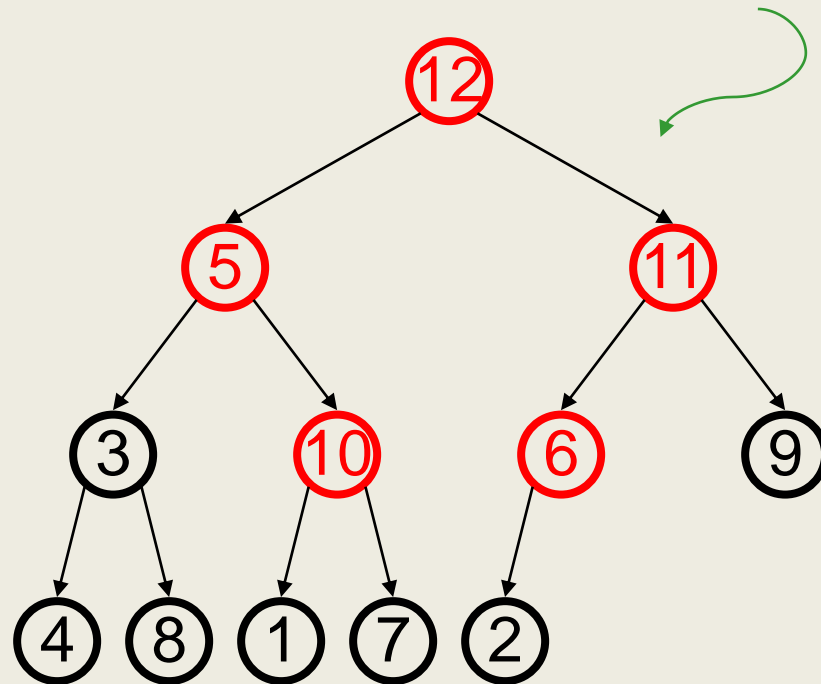
# BuildHeap: Floyd's Method

12	5	11	3	10	6	9	4	8	1	7	2
----	---	----	---	----	---	---	---	---	---	---	---

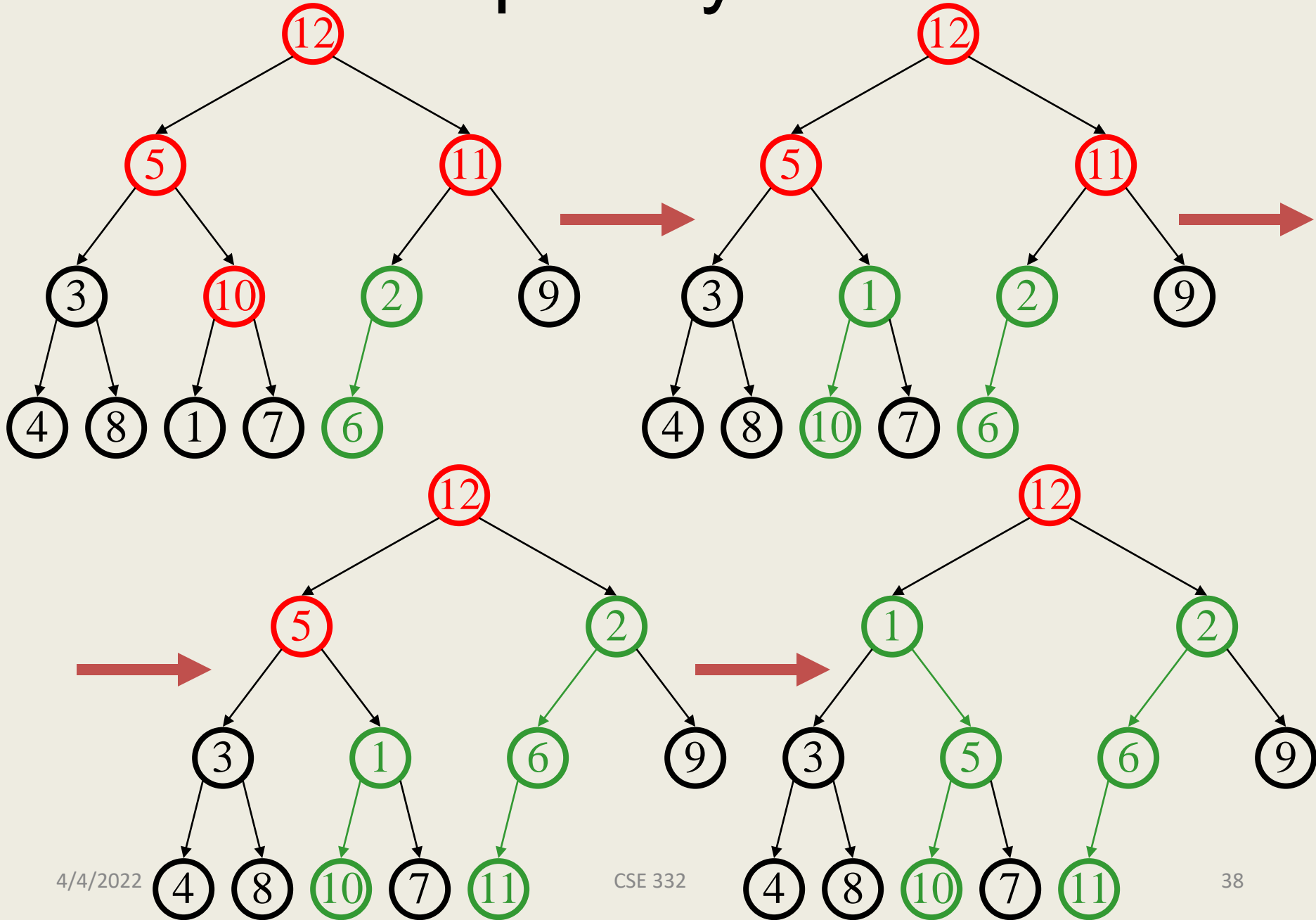
Add elements arbitrarily to form a complete tree.  
Pretend it's a heap and fix the heap-order property!

Red nodes need  
to percolate  
down

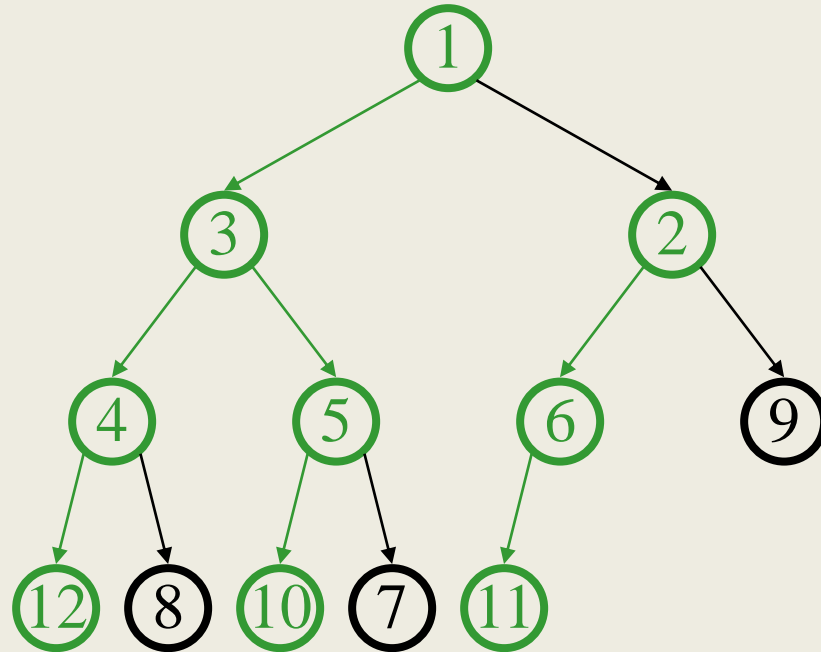
**Key idea:** fix red  
nodes from  
**bottom-up**



# BuildHeap: Floyd's Method



# Finally...



# Buildheap pseudocode

```
private void buildHeap() {  
    for ( int i = currentSize/2; i > 0; i-- )  
        percolateDown( i );  
}
```

*runtime:*

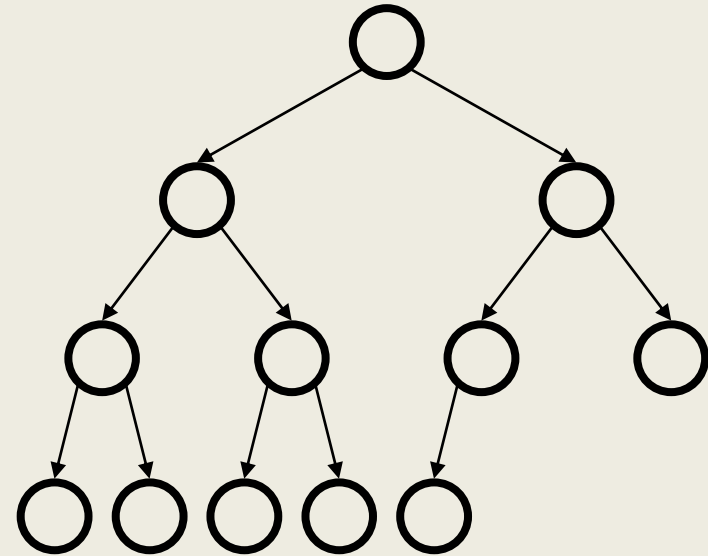
# Buildheap Analysis

$n/4$  nodes percolate at most 1 level

$n/8$  percolate at most 2 levels

$n/16$  percolate at most 3 levels

...



*runtime:*