



# CSE 332: Data Structures & Parallelism

## Lecture 3: Priority Queues

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# *Today*

- Finish up Asymptotic Analysis
- New ADT! Priority Queues

# *Scenario*

What is the difference between waiting for service at a pharmacy versus an ER?

Pharmacies usually follow the rule  
First Come, First Served

Emergency Rooms assign priorities  
based on each individual's need

# *A new ADT: Priority Queue*

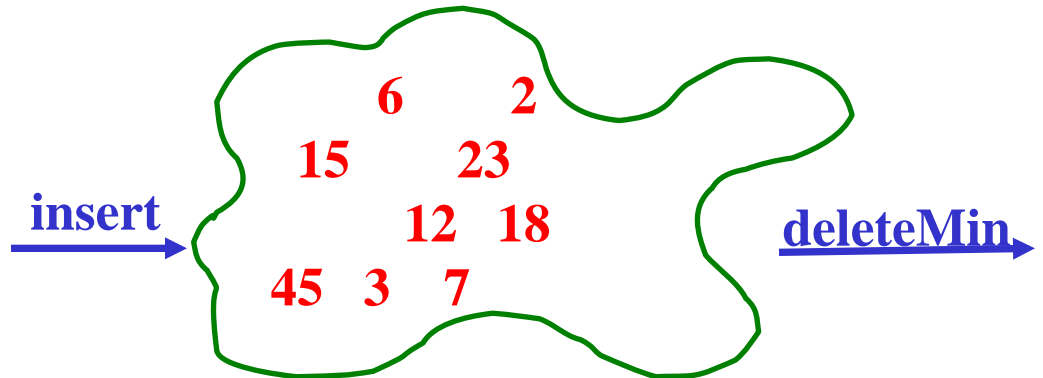
- Textbook Chapter 6
  - We will go back to binary search trees (ch4) and hash tables (ch5) later
  - Nice to see a new and surprising data structure first
- A **priority queue** holds *compare-able data*
  - Unlike stacks and queues need to *compare items*
    - Given  $x$  and  $y$ , is  $x$  less than, equal to, or greater than  $y$
    - What this means can depend on your data
    - Much of course will require comparable data: e.g. sorting
  - Integers are comparable, so will use them in examples
    - But the priority queue ADT is much more general
    - Typically two fields, the *priority* and the *data*

# Priority Queue ADT

- Assume each item has a “priority”
  - The *lesser* item is the one with the *greater* priority
  - So “priority 1” is more important than “priority 4”
  - Just a convention, could also do a maximum priority

- Main Operations:

- **insert**
- **deleteMin**



- Key property: **deleteMin** returns and deletes from the queue the item with greatest priority (lowest priority value)
  - Can resolve ties arbitrarily

## *Aside: We will use ints as data and priority*

For simplicity in lecture, we'll often suppose items are just `ints` and the `int` is also the priority

- So an operation sequence could be
  - `insert 6`
  - `insert 5`
  - `x = deleteMin // Now x = 5.`
- `int` priorities are common, but really just need comparable
- Not having “other data” is very rare
  - Example: print job has a priority *and* the file to print is the data

# Priority Queue Example

To simplify our examples,  
we will just use the priority  
values from now on

`insert` *a* with priority *5*

`insert` *b* with priority *3*

`insert` *c* with priority *4*

*w* = `deleteMin`

*x* = `deleteMin`

`insert` *d* with priority *2*

`insert` *e* with priority *6*

*y* = `deleteMin`

*z* = `deleteMin`

**after execution:**

**Analogy: `insert` is like enqueue, `deleteMin` is like dequeue**  
**But the whole point is to use priorities instead of FIFO**

# *Applications*

Like all good ADTs, the priority queue arises often

- Sometimes “directly”, sometimes less obvious
- Run multiple programs in the operating system
  - “critical” before “interactive” before “compute-intensive”
  - Maybe let users set priority level
- Treat hospital patients in order of severity (or triage)
- Select print jobs in order of decreasing length?
- Forward network packets in order of urgency
- Select most frequent symbols for data compression (cf. CSE143)
- Sort: **insert** all, then repeatedly **deleteMin**



# *More applications*

- “Greedy” algorithms
  - Select the ‘best-looking’ choice at the moment
  - Will see an example when we study graphs in a few weeks
- Discrete event simulation (system modeling, virtual worlds, ...)
  - Simulate how state changes when events fire
  - Each event  $e$  happens at some time  $t$  and generates new events  $e_1, \dots, e_n$  at times  $t+t_1, \dots, t+t_n$
  - Naïve approach: advance “clock” by 1 unit at a time and process any events that happen then
  - Better:
    - *Pending events* in a priority queue (priority = time happens)
    - Repeatedly: **deleteMin** and then **insert** new events
    - Effectively, “set clock ahead to next event”

# Preliminary Implementations of Priority Queue ADT

|                             | insert | deleteMin |
|-----------------------------|--------|-----------|
| Unsorted Array              |        |           |
| Unsorted Linked-List        |        |           |
| Sorted Circular Array       |        |           |
| Sorted Linked-List          |        |           |
| Binary Search Tree<br>(BST) |        |           |

## *Aside: More on possibilities*

- Note: If priorities are inserted in random order, binary search tree will likely do better than  $O(n)$ 
  - $O(\log n)$  **insert** and  $O(\log n)$  **deleteMin** on average
  - Could get same performance from a *balanced* binary search tree (e.g. AVL tree we will study later)
- One more idea: if priorities are  $0, 1, \dots, k$  can use array of lists
  - **insert**: add to front of list at **arr[priority]**,  $O(1)$
  - **deleteMin**: remove from lowest non-empty list  $O(k)$

# *Our Data Structure: The Heap*

## **The Heap:**

- Worst case:  $O(\log n)$  for insert
- Worst case:  $O(\log n)$  for deleteMin
- If items arrive in random order, then the average-case of insert is  $O(1)$
- Very good constant factors

## **Key idea:** Only pay for functionality needed

- We need something better than scanning unsorted items
- But we do not need to maintain a full sorted list
- We will *visualize* our heap as a tree, so we need to review some tree terminology

# Q: Reviewing Some Tree Terminology

*root(T):*

*leaves(T):*

*children(B):*

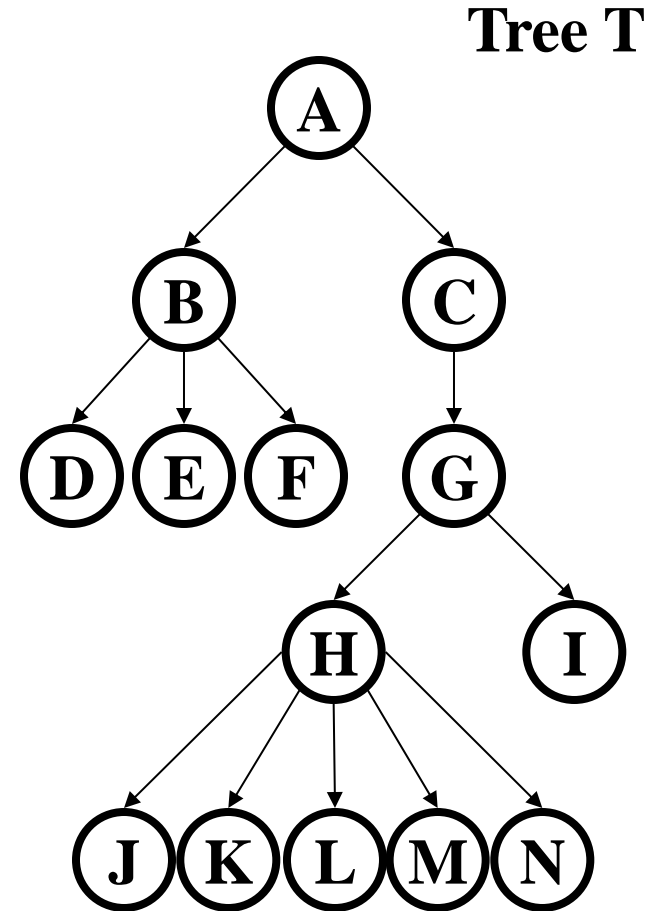
*parent(H):*

*siblings(E):*

*ancestors(F):*

*descendants(G):*

*subtree(G):*



## Q: Some More Tree Terminology

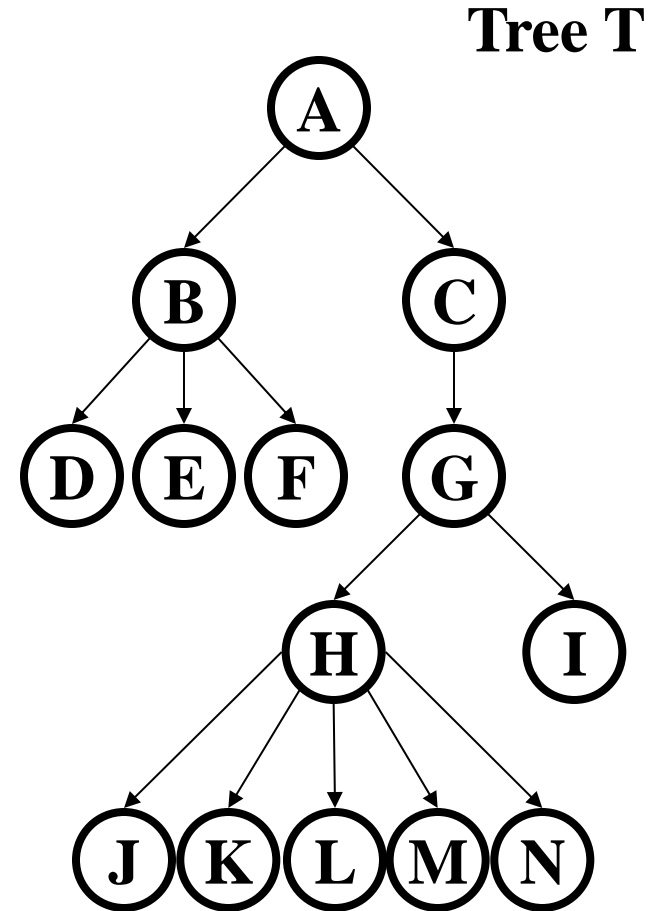
*depth(B):*

*height(G):*

*height(T):*

*degree(B):*

*branching factor(T):*



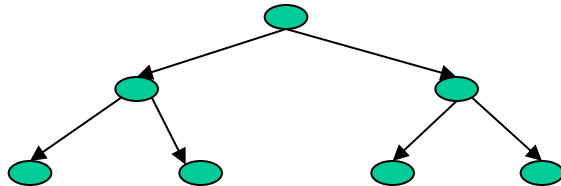
# Types of Trees

Binary tree: Every node has  $\leq 2$  children

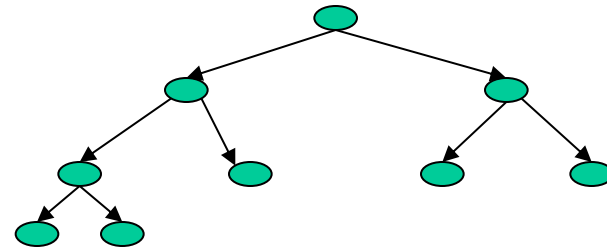
n-ary tree: Every node has  $\leq n$  children

Perfect tree: Every row is completely full

Complete tree: All rows except possibly the bottom are completely full, and it is filled from left to right



Perfect Tree



Complete Tree

# *Some Basic Tree Properties*

*Nodes* in a perfect binary tree of height  $h$ ?

*Leaf nodes* in a perfect binary tree of height  $h$ ?

Height of a perfect binary tree with  $n$  nodes?

Height of a complete binary tree with  $n$  nodes?



# *Properties of a Binary Min-Heap*

More commonly known as a **binary heap** or simply a **heap**

- **Structure Property:**

A complete [binary] tree

- **Heap Property:**

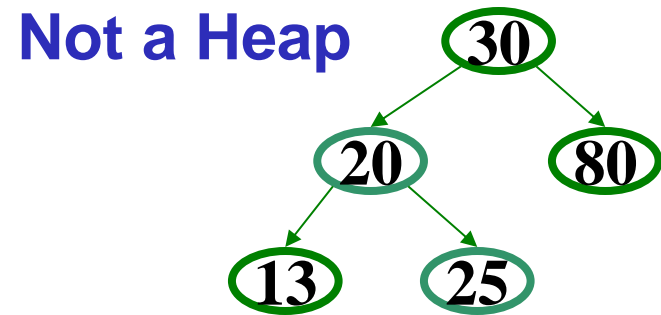
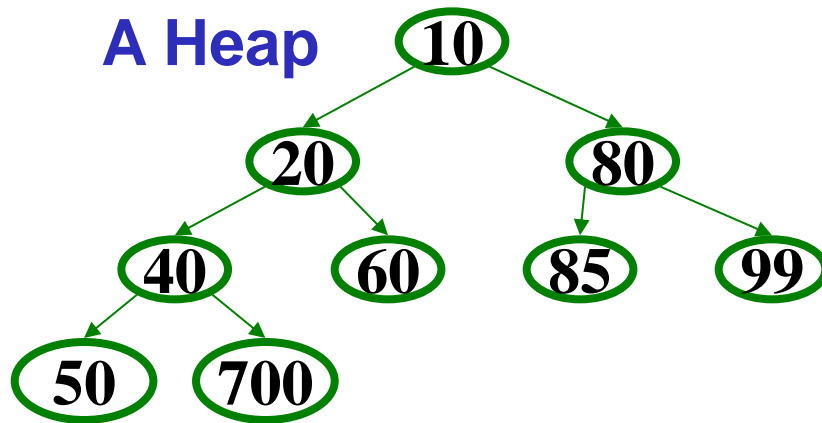
The priority of every non-root node is greater than (or possibly equal to) the priority of its parent

How is this different from a binary search tree?

# *Properties of a Binary Min-Heap*

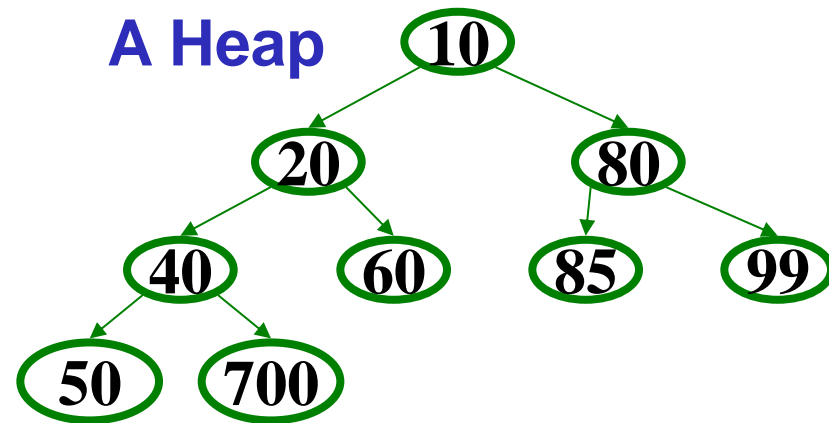
More commonly known as a **binary heap** or simply a **heap**

- **Structure Property:**  
A complete [binary] tree
- **Heap Order Property:**  
The priority of every non-root node is greater than the priority of its parent



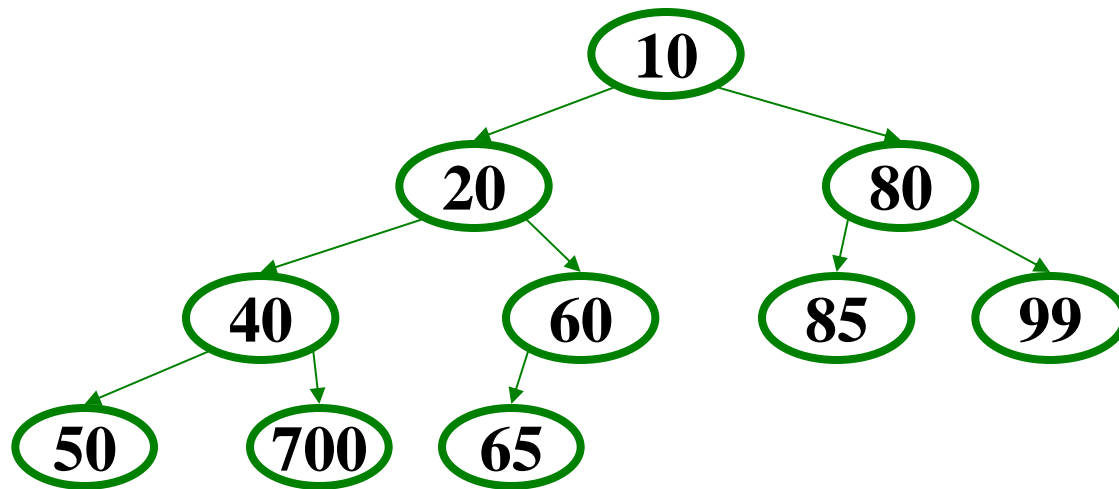
# *Properties of a Binary Min-Heap*

- Where is the minimum priority item?
- What is the height of a heap with  $n$  items?



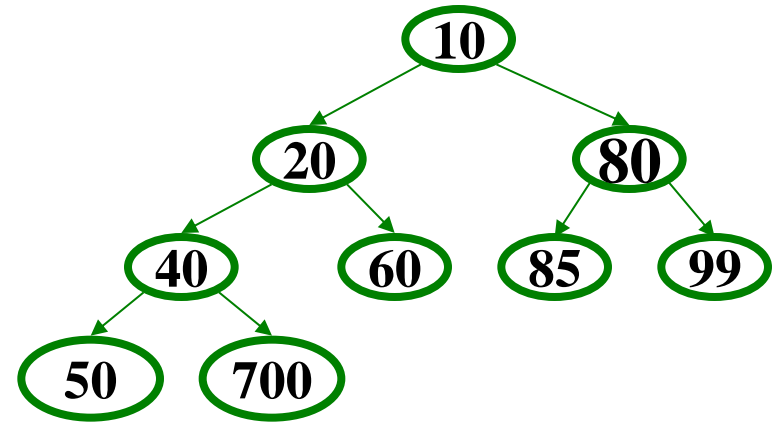
# Heap Operations

- findMin:
- deleteMin: percolate down.
- insert(val): percolate up.



# Operations: basic idea

- **findMin:**  
    return `root.data`
- **deleteMin:**
  1. `answer = root.data`
  2. Move right-most node in last row to root to restore structure property
  3. “Percolate down” to restore heap order property
- **insert:**
  1. Put new node in next position on bottom row to restore structure property
  2. “Percolate up” to restore heap order property

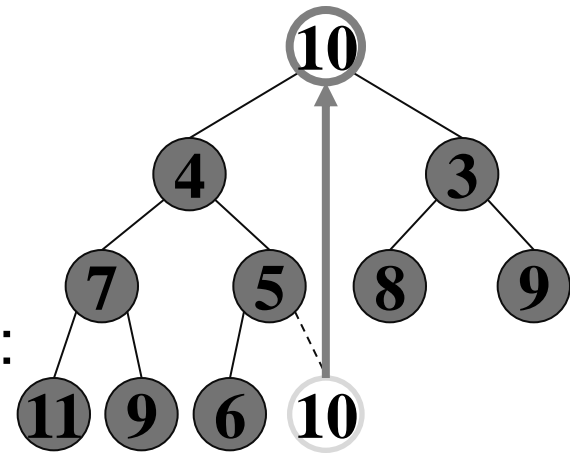
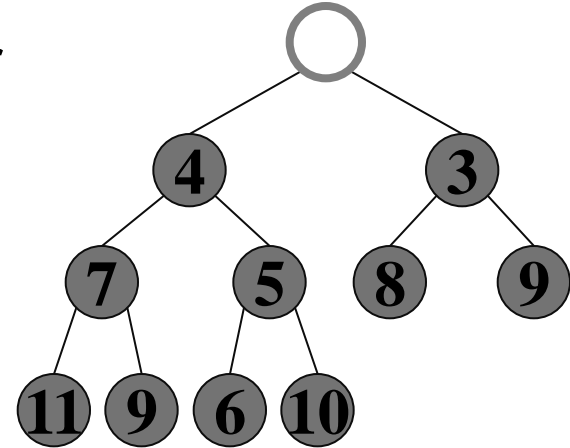


## *Overall strategy:*

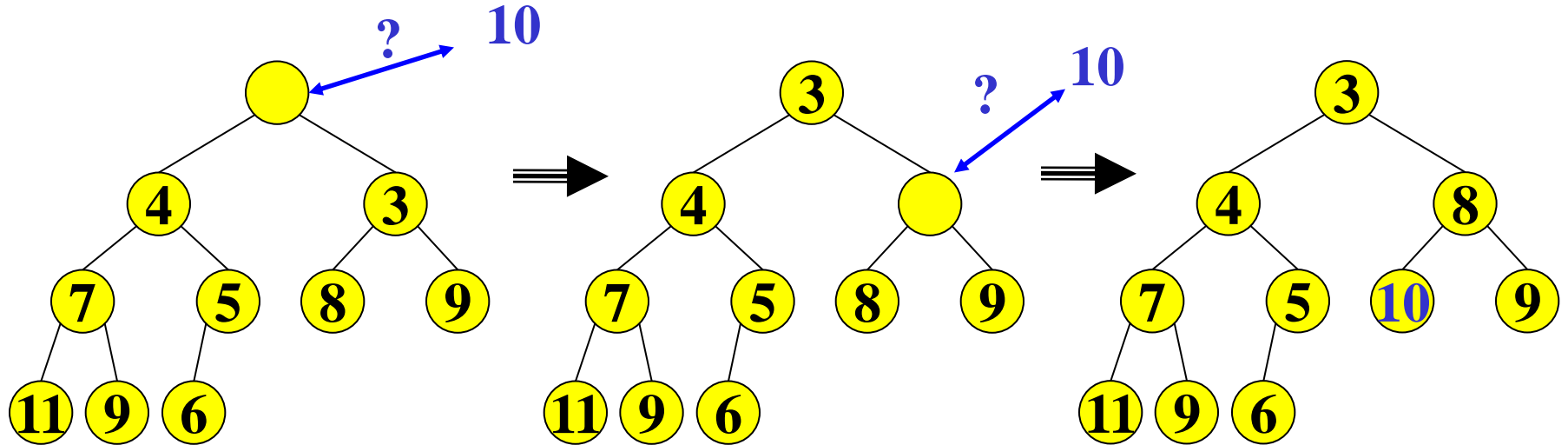
- *Preserve complete tree structure property*
- *This may break heap order property*
- *Percolate to restore heap order property*

# DeleteMin Implementation

1. Delete value at root node (and store it for later return)
2. There is now a "hole" at the root. We must "fill" the hole with another value, must have a tree with one less node, and it must still be a complete tree
3. The "last" node is the obvious choice, but now the heap order property is violated
4. We **percolate down** to fix the heap order:  
While greater than either child  
    Swap with smaller child



# Percolate Down



Percolate down:

- Keep comparing with both children
- Move smaller child up and go down one level
- Done if both children are  $\geq$  item or reached a leaf node
- Why does this work? What is the run time?

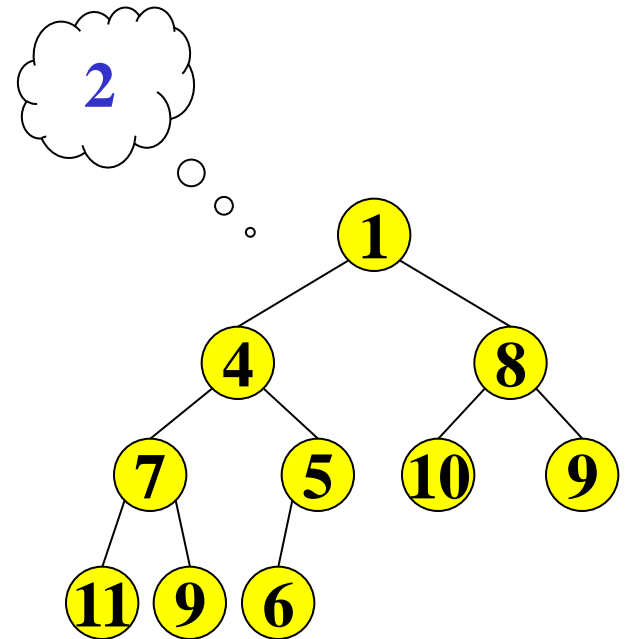
# *DeleteMin: Run Time Analysis*

- Run time is  $O(\text{height of heap})$
- A heap is a complete binary tree
- Height of a complete binary tree of  $n$  nodes?
  - $\text{height} = \lfloor \log_2(n) \rfloor$
- Run time of **deleteMin** is  $O(\log n)$



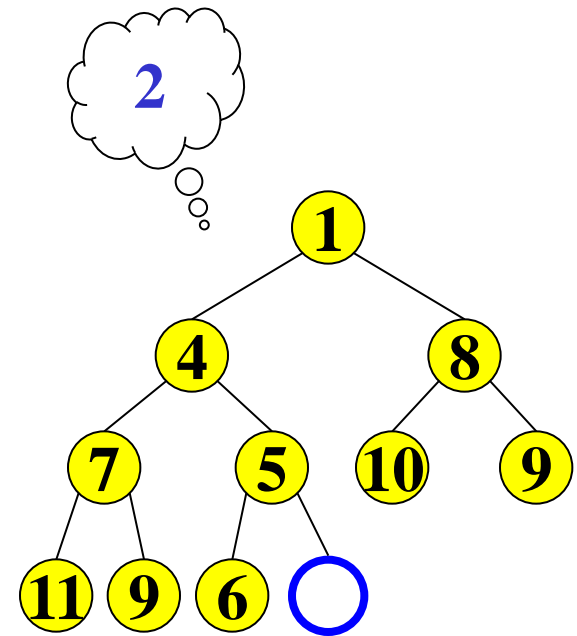
# *Insert*

- Add a value to the tree
- Structure and heap order properties must still be correct afterwards

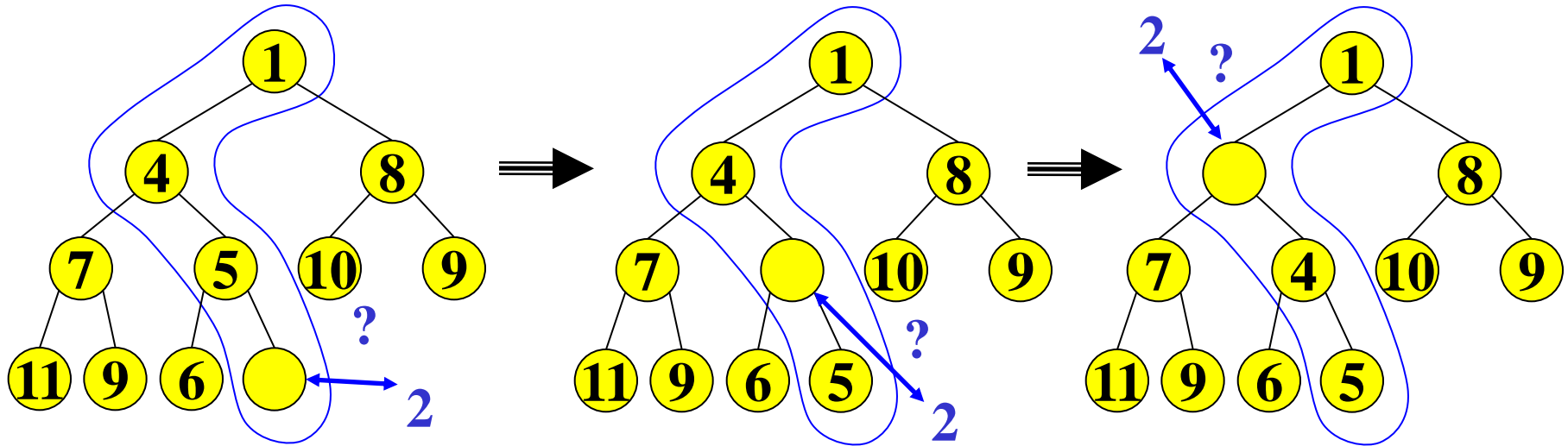


# *Insert: Maintain the Structure Property*

- There is only **one** valid tree shape after we add one more node!
- So put our new data there and then focus on restoring the heap order property



# *Maintain the heap order property*



Percolate up:

- Put new data in new location
- If parent larger, swap with parent, and continue
- Done if parent  $\leq$  item or reached root
- Why does this work? What is the run time?

# *A Clever Trick for Storing the Heap...*

Clearly, insert and deleteMin are worst-case  $O(\log n)$

- But we promised average-case  $O(1)$  insert (how??)

Insert requires access to the “next to use” position in the tree

- Walking the tree from root to leaf requires  $O(\log n)$  steps
- Insert and Deletemin would have to update the “next to use” reference each time:  $O(\log n)$

We should only pay for the functionality we need!!

- Why have we insisted the tree be complete? ☺

All complete trees of size  $n$  contain the same edges

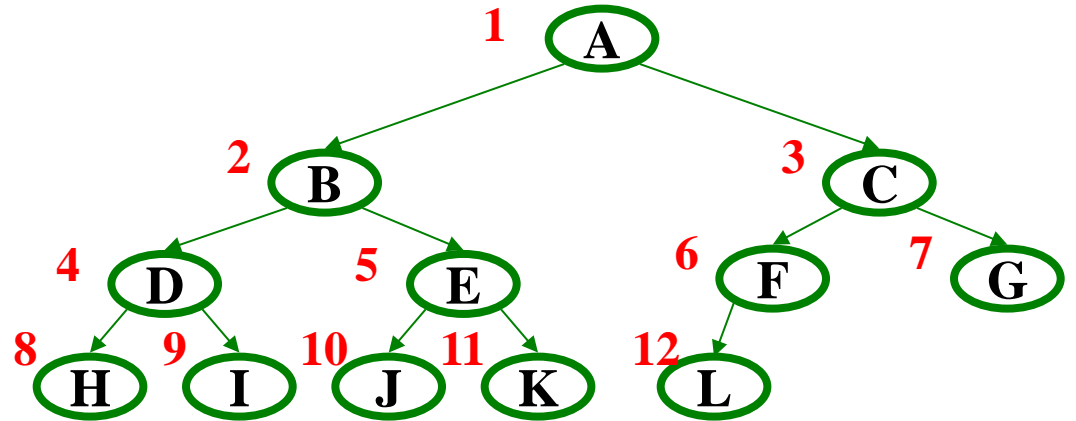
- So why are we even representing the edges?

**Here comes the really clever bit about implementing heaps!!!**

# Array Representation of a Binary Heap

From node i:

- left child:
- right child:
- parent:



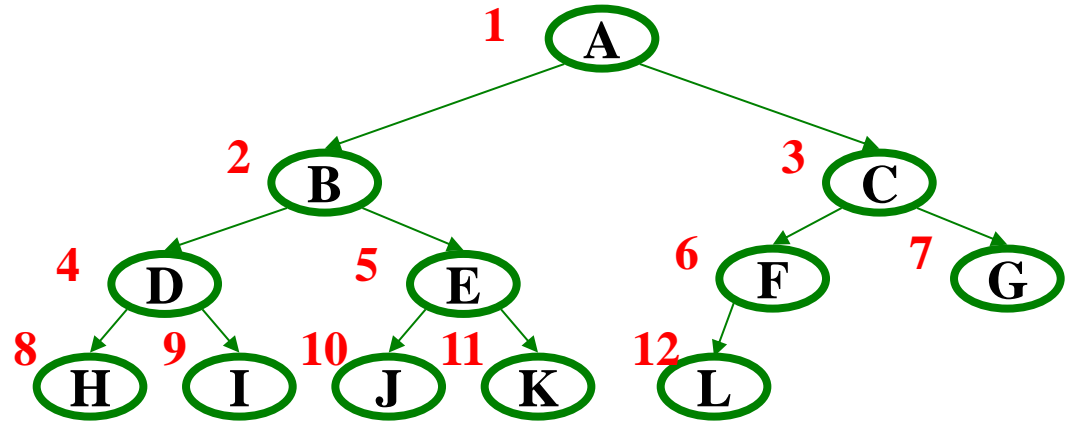
|   |   |   |   |   |   |   |   |   |   |    |    |    |    |
|---|---|---|---|---|---|---|---|---|---|----|----|----|----|
|   | A | B | C | D | E | F | G | H | I | J  | K  | L  |    |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 |

- We skip index 0 to make the math simpler
- Actually, it can be a good place to store the current size of the heap

# Array Representation of a Binary Heap

From node  $i$ :

- left child:  $2i$
- right child:  $2i+1$
- parent:  $i / 2$



|   |   |   |   |   |   |   |   |   |   |    |    |    |    |
|---|---|---|---|---|---|---|---|---|---|----|----|----|----|
|   | 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 |

- We skip index 0 to make the math simpler
- Actually, it can be a good place to store the current size of the heap