



# CSE332: Data Abstractions

## Lecture 4: Priority Queues

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# A new ADT: Priority Queue

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- ▶ Textbook Chapter 6: Priority Queues
  - ▶ Will go back to binary search trees (4) and hashtables (5) later
- ▶ 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
      - Numbers: numeric ordering
      - Strings: lexicon ordering
      - Employee profile: lexicon ordering on name? Id?
  - ▶ Much of course will require comparable items:
    - ▶ Sorting
    - ▶ Binary Search Trees
  - ▶ Integers are comparable, so will use them in examples
    - ▶ But the priority queue ADT is much more general

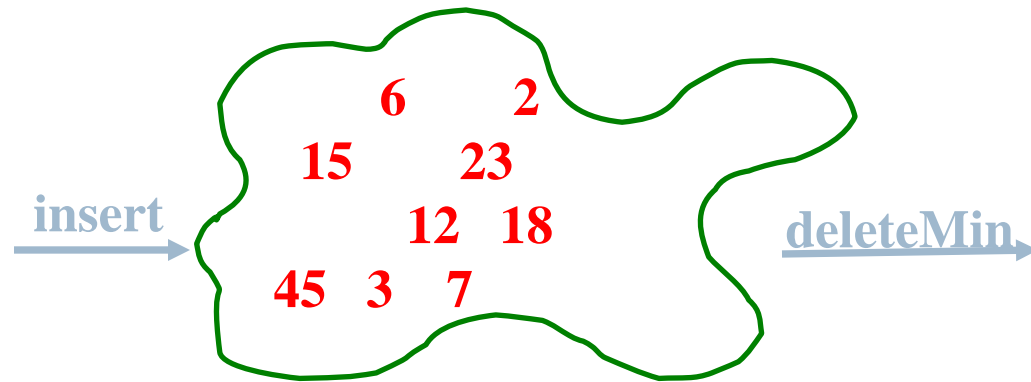
# Priority Queues

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- ▶ Assume each item has a “priority”
  - ▶ The *lesser value* item is the one with the *greater* priority
  - ▶ So “priority 1” is more important than “priority 4”
  - ▶ (Just a convention)

- ▶ Operations:

- ▶ `insert`
- ▶ `deleteMin`
- ▶ `create`, `is_empty`, `destroy`



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

# Focusing on the numbers

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- ▶ For simplicity in lecture, we'll often suppose items are just `ints` and the `int` is the priority
  - ▶ So an operation sequence could be

```
insert 6
insert 5
x = deleteMin
```
  - `int` priorities are common, but really just need comparable
  - ▶ Not having “other data” is very rare
    - ▶ Example: print job is a priority *and* the file

# Example

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`insert 5`

`insert 3`

`insert 4`

`a = deleteMin`                    `3`

`b = deleteMin`                    `4`

`insert 2`

`insert 6`

`c = deleteMin`                    `2`

`d = deleteMin`                    `5`

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

# Applications

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Like all good ADTs, the priority queue arises often

- ▶ 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**
  - ▶ Much like Project 1 uses a stack to implement reverse

# More applications for Priority Queues

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- ▶ “Greedy” algorithms
  - ▶ Perform 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”

# Need a good data structure!

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- ▶ Will show an efficient, non-obvious data structure for this ADT
  - ▶ But first let's analyze some "obvious" ideas for  $n$  data items
  - ▶ All times worst-case; but assume arrays "have room"

<i>data</i>	<i>insert algorithm / time</i>		<i>deleteMin algorithm / time</i>	
unsorted array	add at end	$O(1)$	search	$O(n)$
unsorted linked list	add at front	$O(1)$	search	$O(n)$
sorted circular array	search / shift	$O(n)$	move front	$O(1)$
sorted linked list	put in right place	$O(n)$	remove at front	$O(1)$
binary search tree	put in right place	$O(n)$	leftmost	$O(n)$



# More on possibilities

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- ▶ If priorities are random, binary search tree will likely do better
  - ▶  $O(\log n)$  `insert` and  $O(\log n)$  `deleteMin` on average
- ▶ 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  $O(k)$
  - ▶ Only really feasible for small  $k$
- ▶ But we are about to see a data structure called a “binary heap”
  - ▶  $O(\log n)$  `insert` and  $O(\log n)$  `deleteMin` worst-case
  - ▶ Very good constant factors
  - ▶ If items arrive in random order, then `insert` is  $O(1)$  on average!

# Tree terms (review?)

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The binary heap data structure implementing the priority queue ADT will be a *tree*, so worth establishing some terminology

*root*(tree)

*children*(node)

*parent*(node)

*leaves*(tree)

*siblings*(node)

*ancestors*(node)

*descendents*(node)

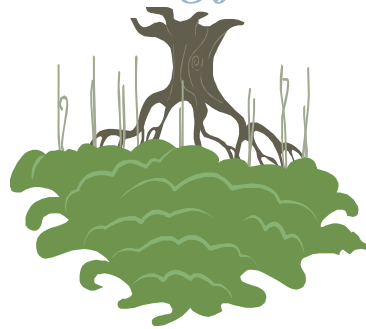
*subtree*(node)

*depth*(node)

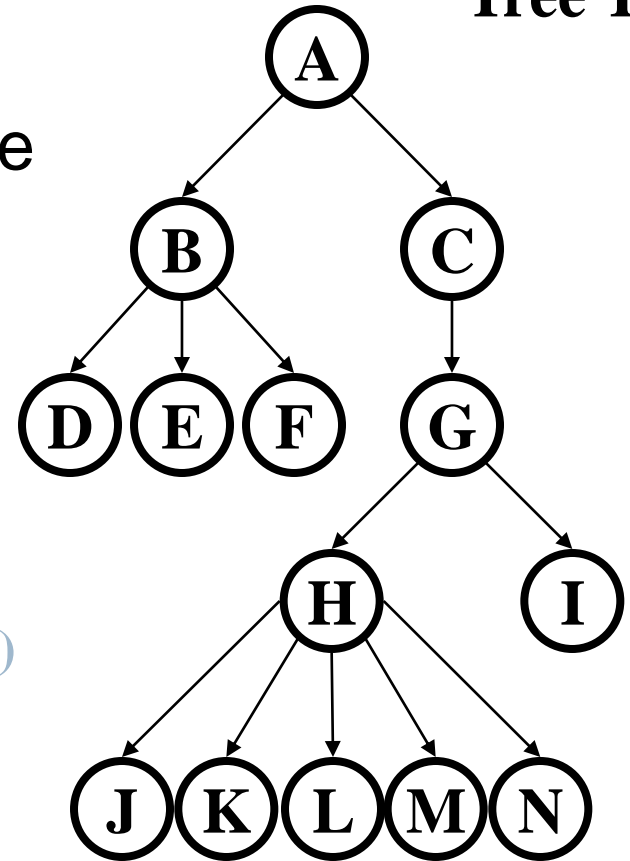
*height*(tree)

*degree*(node)

*branching factor*(tree)



**Tree T**

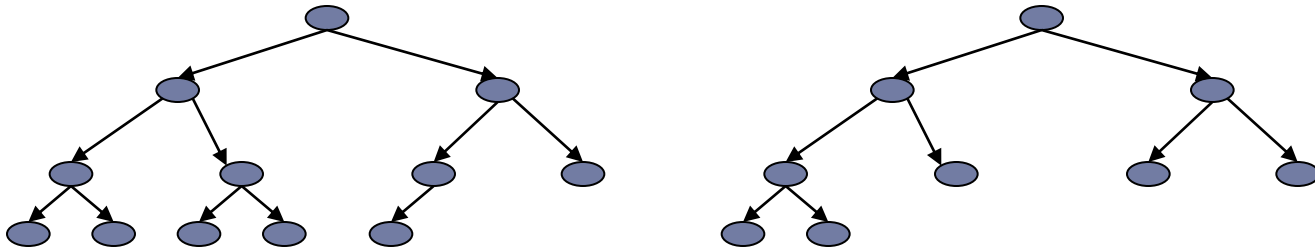


# Kinds of trees

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Certain terms define trees with specific structure

- ▶ **Binary tree:** Each node has at most 2 children
- ▶  **$n$ -ary tree:** Each node has at most  $n$  children
- ▶ **Complete tree:** Each row is completely full except maybe the bottom row, which is filled from left to right

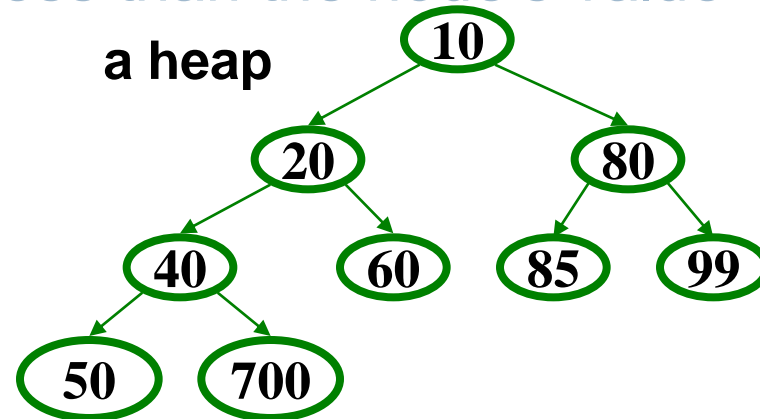
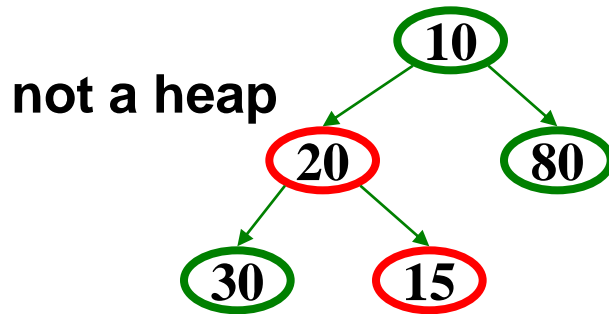


Later we'll learn a **tree** is a kind of **directed graph** with specific structure

# Binary Heap: Priority Queue DS

Finally, then, a *binary min-heap* (aka *binary heap* or just *heap*) has the following 2 properties:

- ▶ Structure property : A complete tree
- ▶ Heap ordering property: For every (non-root) node the parent node's value is less than the node's value



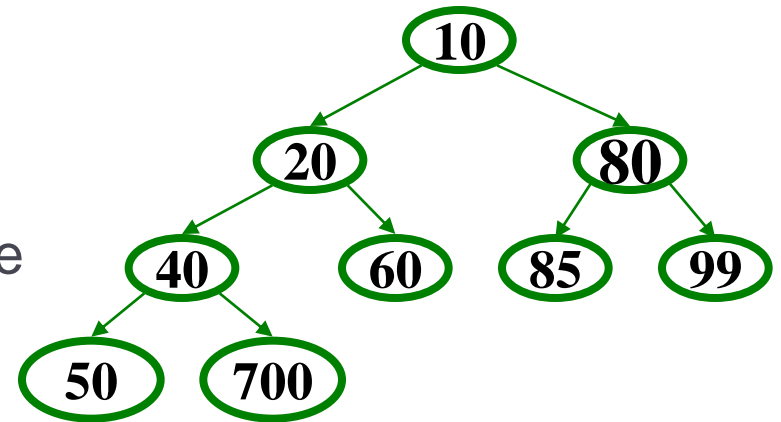
So:

- Where is the highest-priority item? **root**
- What is the height of a heap with  $n$  items?  **$O(\log n)$**

# Operations: basic idea

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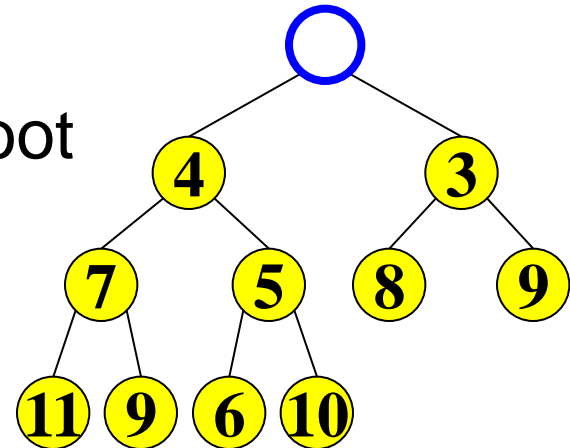
- ▶ **findMin:** return `root.data`
- ▶ **deleteMin:**
  1. `answer = root.data`
  2. Move right-most node in last row to root to restore structure property
  3. “Percolate down” to restore heap property
- ▶ **insert:**
  1. Put new node in next position on bottom row to restore structure property
 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



# DeleteMin

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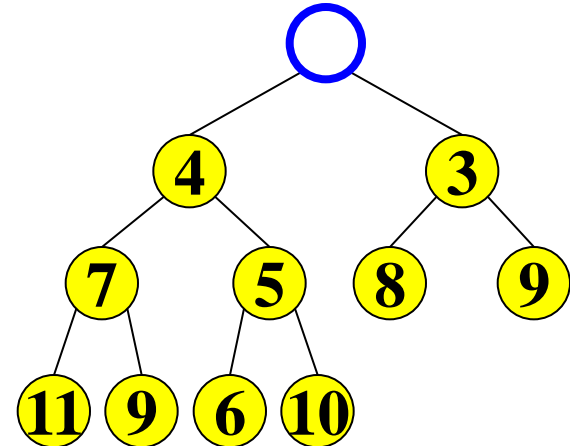
1. Delete (and return) value at root node



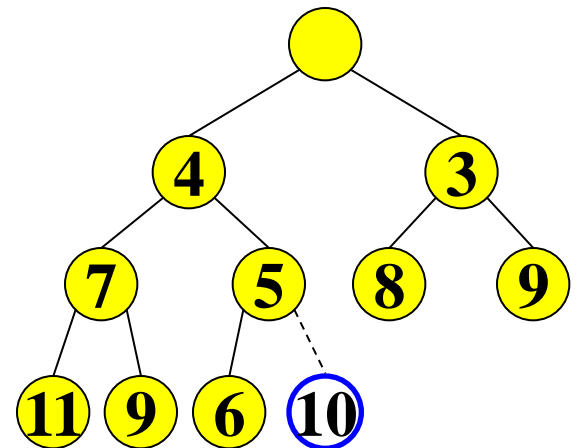
## 2. Restore the Structure Property

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- ▶ We now have a “hole” at the root
  - ▶ Need to fill the hole with another value

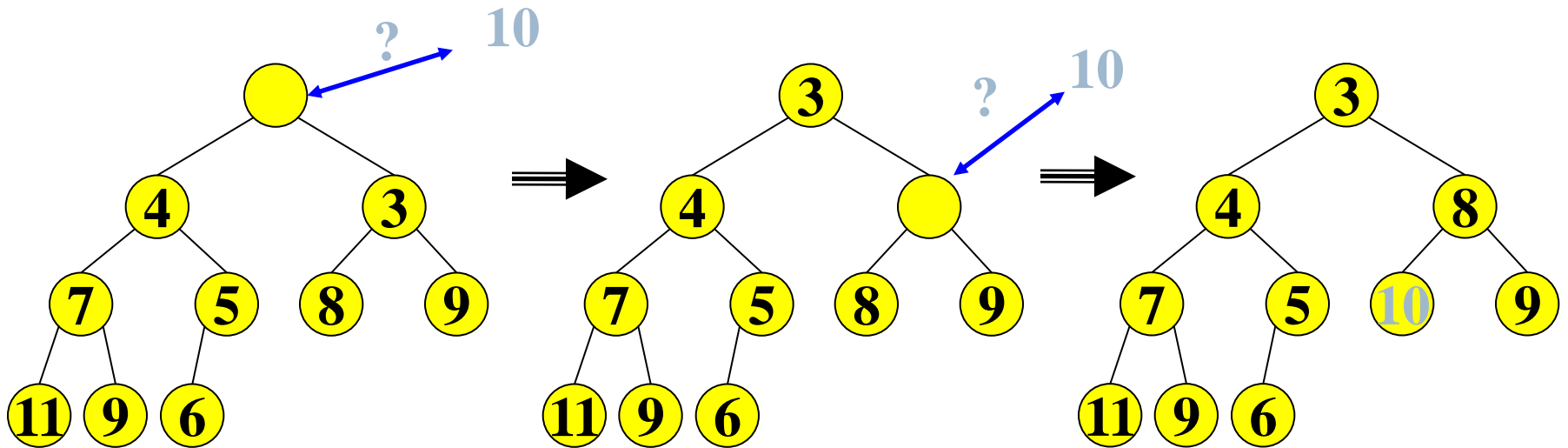


- ▶ When we are done, the tree will have one less node and must still be complete



### 3. Restore the Heap Property

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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
- What is the run time?

**$O(\log n)$**

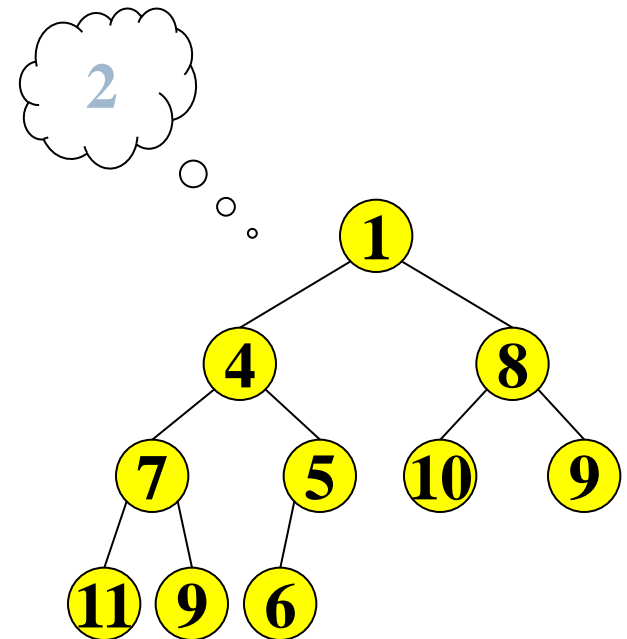
- Why not swap with larger of children, if it's smaller than both?



# Insert

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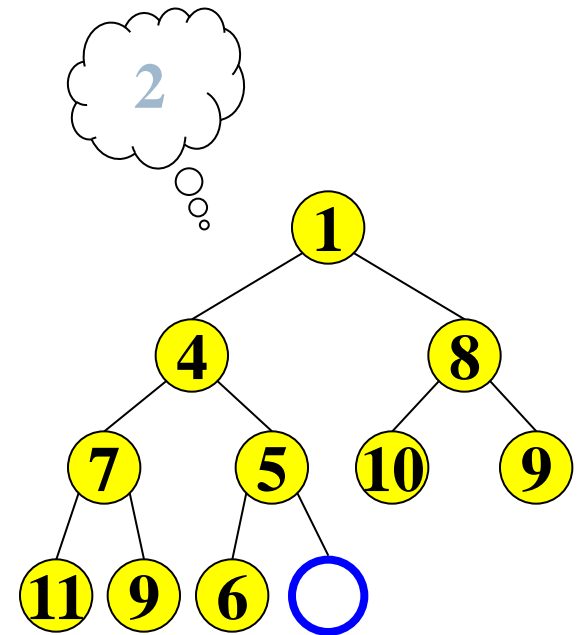
- ▶ Add a value to the tree
- ▶ Structure and heap order properties must still be correct afterwards



# Insert: Maintain the Structure Property

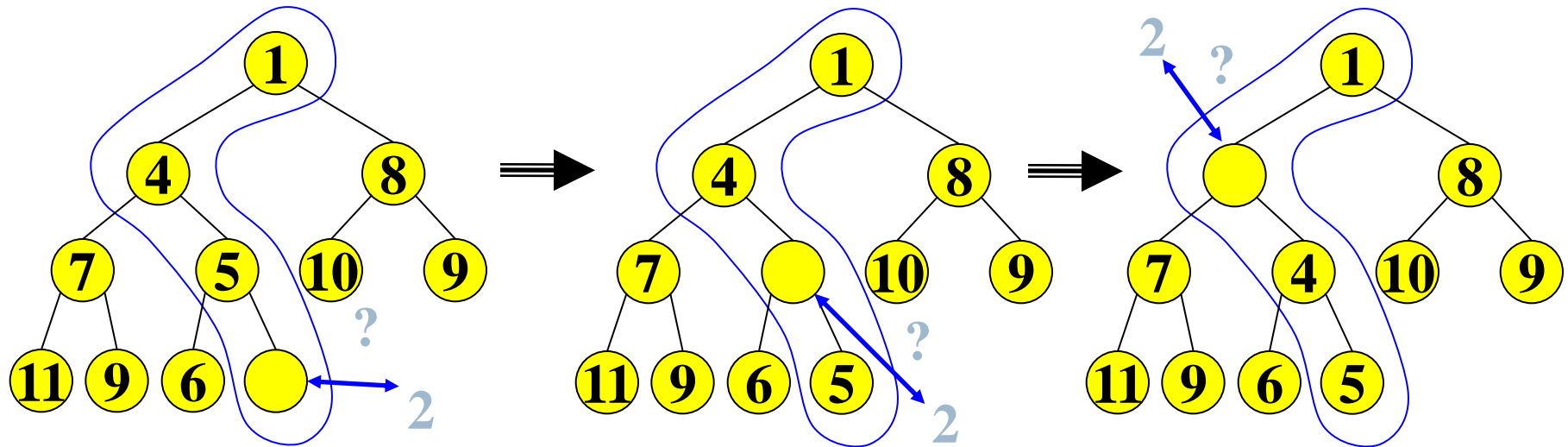
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- ▶ There is only one valid tree shape after we add one more node
- ▶ So put our new data there and then focus on restoring the heap property



# Maintain the heap property

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Percolate up:

- Put new data in new location
- If parent larger, swap with parent, and continue
- Done if parent  $\leq$  item or reached root
- Run time?

At the end, how do we know 2 is going to be less than its left child (here, 7) which it wasn't compared against?

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# Insert: Run Time Analysis

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- ▶ Like **deleteMin**, worst-case time proportional to tree height
  - ▶  $O(\log n)$
- ▶ But... **deleteMin** needs the “last used” complete-tree position and **insert** needs the “next to use” complete-tree position
  - ▶ If “keep a reference to there” then **insert** and **deleteMin** have to adjust that reference:  $O(\log n)$  in worst case
  - ▶ Could calculate how to find it in  $O(\log n)$  from the root given the size of the heap
    - ▶ But it’s not easy
    - ▶ And then **insert** is always  $O(\log n)$ ; what about the promised  $O(1)$  on average (assuming random arrival of items)?
- ▶ There’s a “trick”: don’t represent complete trees as nodes with pointers to children