

LEC 13

CSE 373

# Heaps II, Interviews

## BEFORE WE START

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# Announcements

- P2 due TONIGHT 11:59pm PDT
- P3 will come out today. Is also due in two weeks.
  - Very related to lectures this week!

# Learning Objectives

After this lecture, you should be able to...

1. (Continued) Trace the `removeMin()`, and `add()` methods, including `percolateDown()` and `percolateUp()`
2. Describe how a heap can be stored using an array, and compare that implementation to one using linked nodes
3. Recall the runtime to build a heap with Floyd's `buildHeap` algorithm, and describe how the algorithm proceeds

# Lecture Outline

- **Heaps II**
  - **Operations & Implementation** 
  - Building a Heap
- Technical Interviews

# Review Priority Queue ADT



We'll assume this one in 373!

- If a Queue is “First-In-First-Out” (FIFO), Priority Queues are “Most-Important-Out-First”
- Items in Priority Queue must be **comparable** – The data structure will maintain some amount of internal sorting, in a sort of similar way to BSTs/AVLs
- We'll talk about “Min Priority Queues” (lowest priority is most important), but “Max Priority Queues” are almost identical

## MIN PRIORITY QUEUE ADT

### State

Set of comparable values (*ordered based on “priority”*)

### Behavior

**add(value)** – add a new element to the collection

**removeMin()** – returns the element with the smallest priority, removes it from the collection

**peekMin()** – find, but do not remove the element with the smallest priority

## MAX PRIORITY QUEUE ADT

### State

Set of comparable values (*ordered based on “priority”*)

### Behavior

**add(value)** – add a new element to the collection

**removeMin()** – returns the element with the largest priority, removes it from the collection

**peekMin()** – find, but do not remove the element with the largest priority



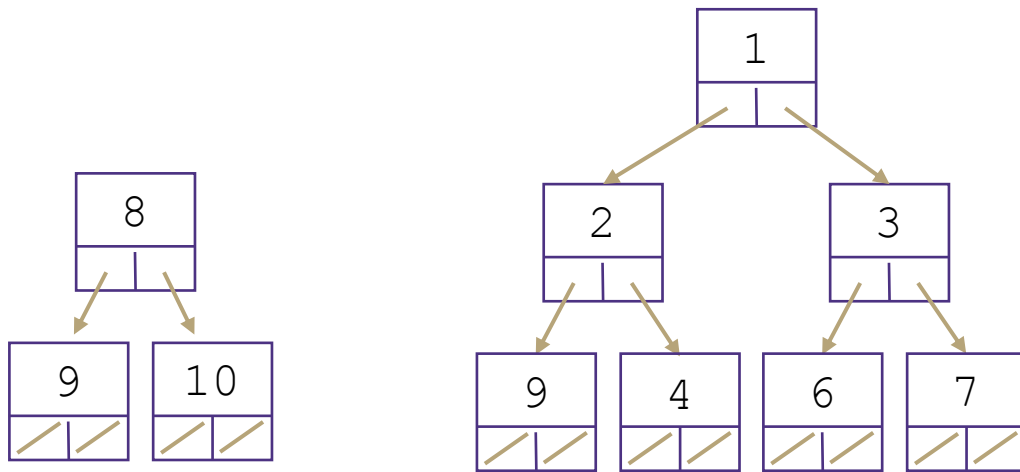
# Review Binary Heap Invariants Summary

- One flavor of heap is a **binary heap**, which is a Binary Tree with the heap invariants (*NOT* a Binary Search Tree)

INVARIANT

## Binary Tree

Every node has at most 2 children



INVARIANT

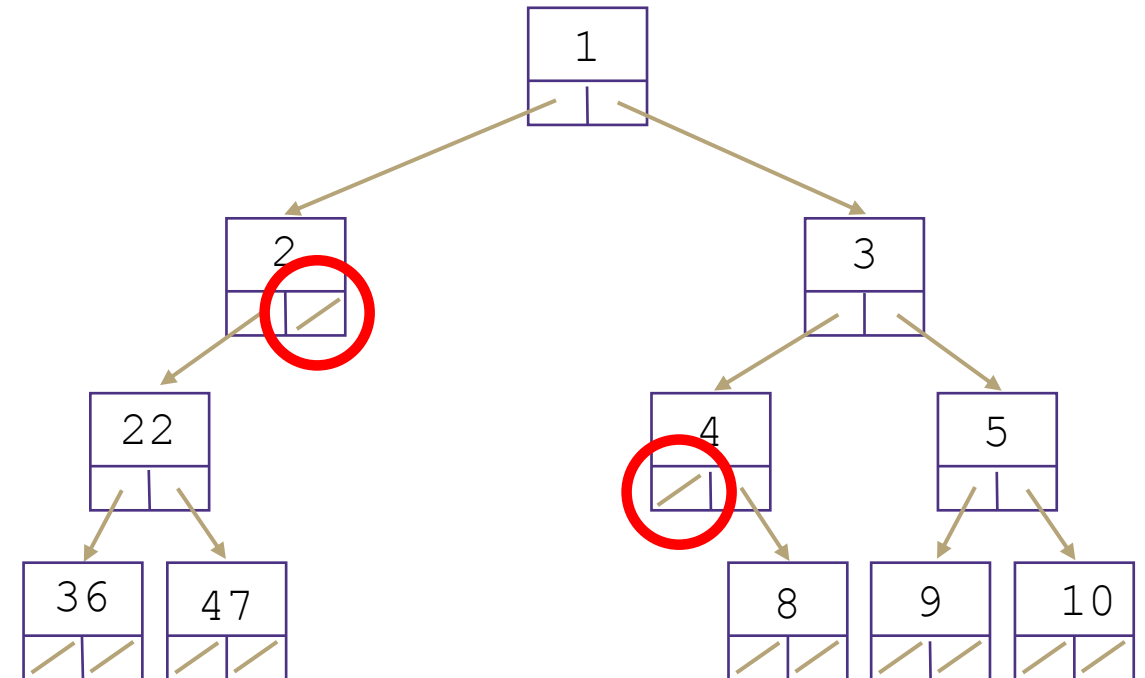
## Heap Invariant

Every node is less than or equal to all of its children.

INVARIANT

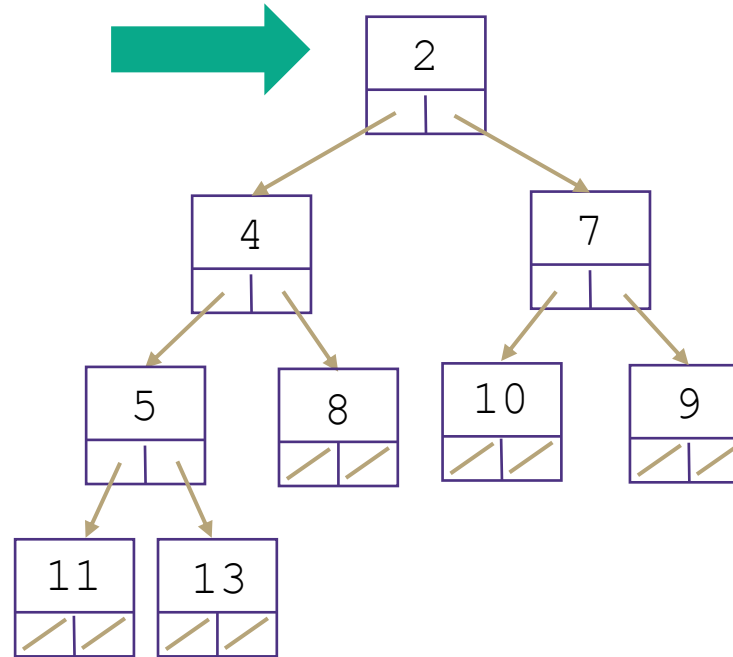
## Heap Structure Invariant

A heap is always a *complete* tree.



# Review Implementing peekMin()

Runtime:  $\Theta(1)$



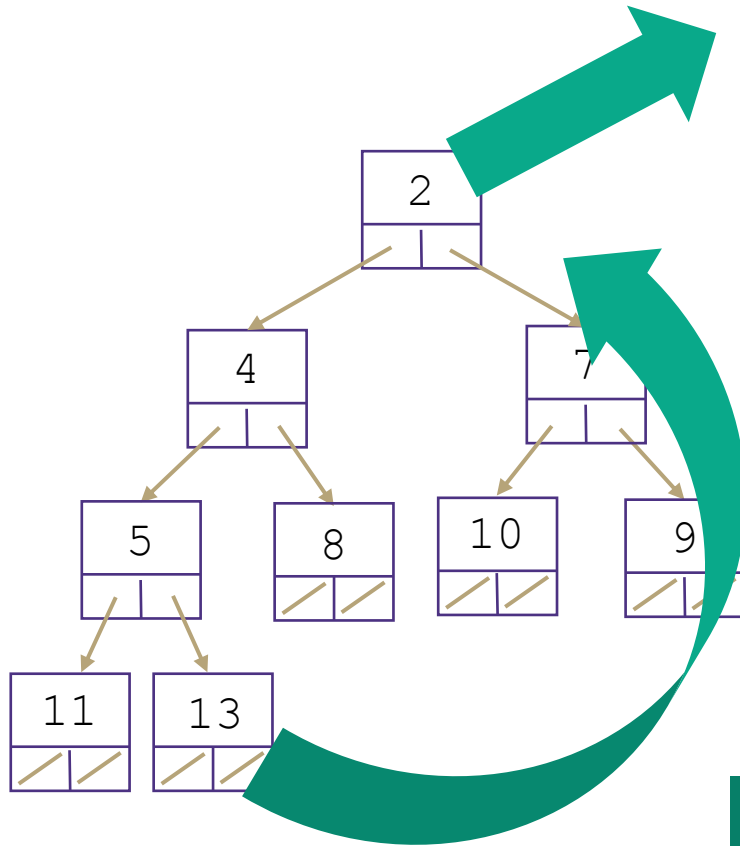
Simply return the value at the root!  
That's a constant-time operation if  
we've ever seen one 😊

# Review Implement removeMin()

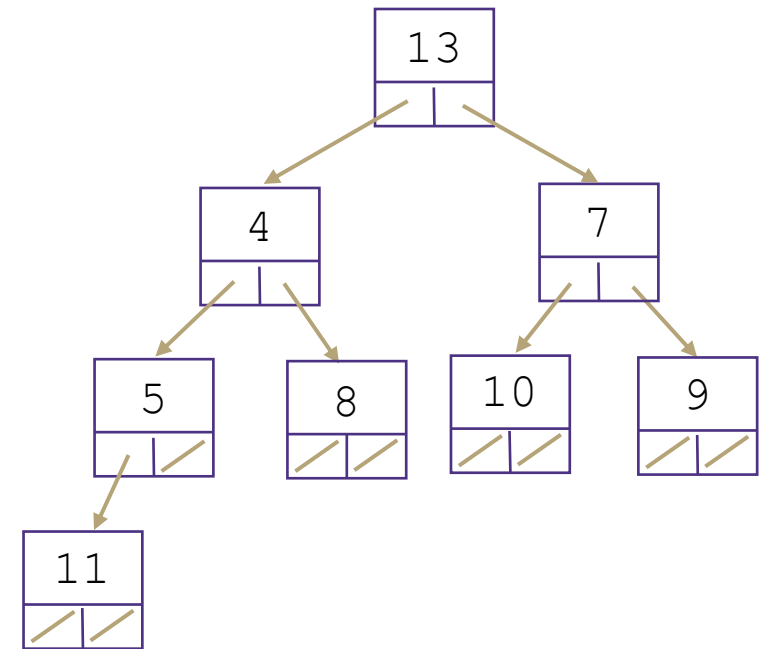
INVARIANT

## Heap Structure Invariant

A heap is always a *complete* tree.



- 1) Remove min to return
- 2) Structure Invariant broken: replace with bottom level right-most node (the only one that can be moved)



INVARIANT

## Heap Invariant

Every node is less than or equal to all of its children.

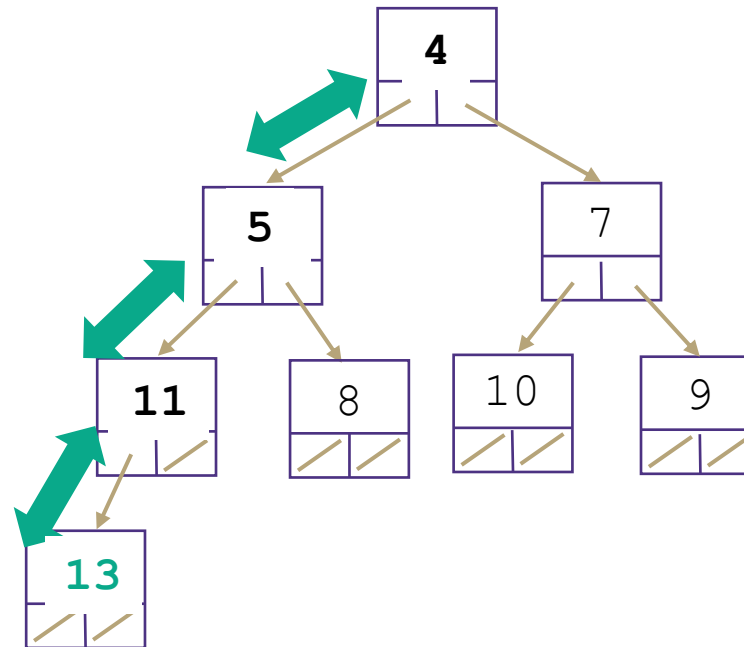
Structure Invariant restored, but **Heap Invariant now broken**



# Review Implement removeMin(): percolateDown

## 3) percolateDown

Recursively swap parent with **smallest** child until parent is smaller than both children (or at a leaf).



What's the worst-case running time?

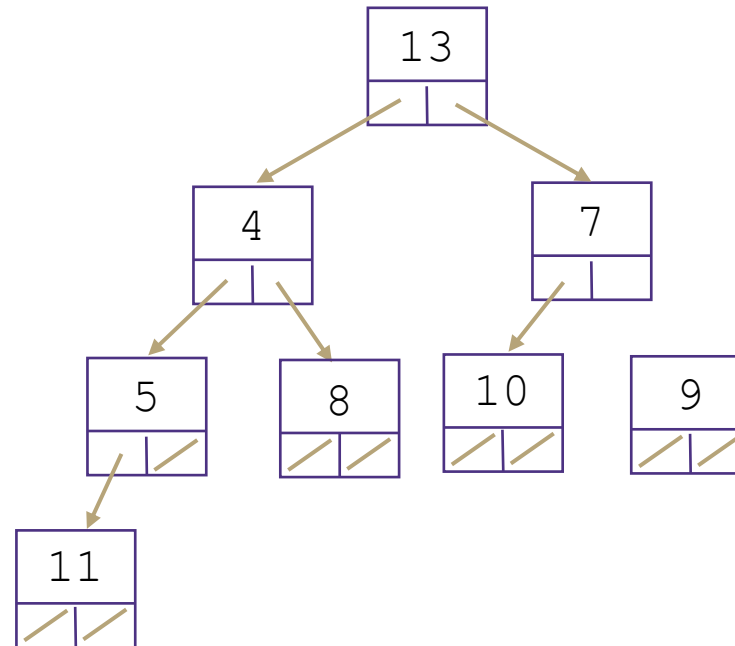
- Find last element
- Move it to top spot
- Swap until invariant restored

*This is why we want to keep the height of the tree small! The height of these tree structures (BST, AVL, heaps) directly correlates with the worst case runtimes*

Structure invariant restored, heap invariant restored

# *Review* percolateDown: Why Smallest Child?

- Why does percolateDown swap with the smallest child instead of just any child?

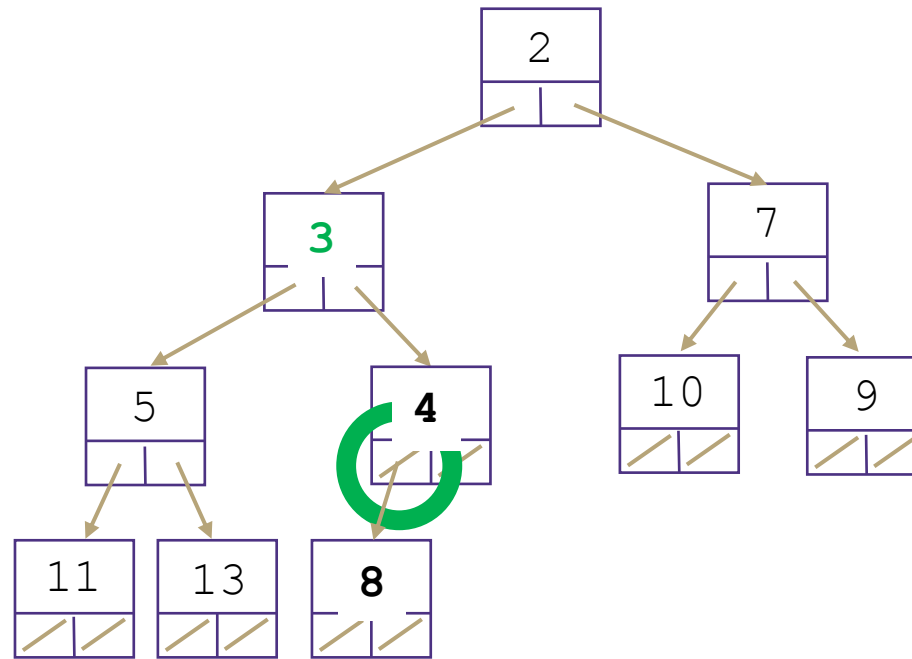


- If we swap 13 and 7, the heap invariant isn't restored!
- 7 is greater than 4 (it's not the smallest child!) so it will violate the invariant.

# Review Implement add(key): percolateUp!

## ADD ALGORITHM

- Insert node on the bottom level to ensure no gaps (**Heap Structure Invariant**)
- Fix **Heap Invariant** with new technique: **percolateUp**
  - Swap with parent, until your parent is smaller than you (or you're the root).



Worst case runtime similar to removeMin and percolateDown

- might have to do  $\log(n)$  swaps, so the worst-case runtime is  $\Theta(\log(n))$

# MinHeap Runtimes

removeMin():

- 1 1. Remove root node
- log n 2. Find last node in tree and swap to top level
- log n 3. Percolate down to fix heap invariant

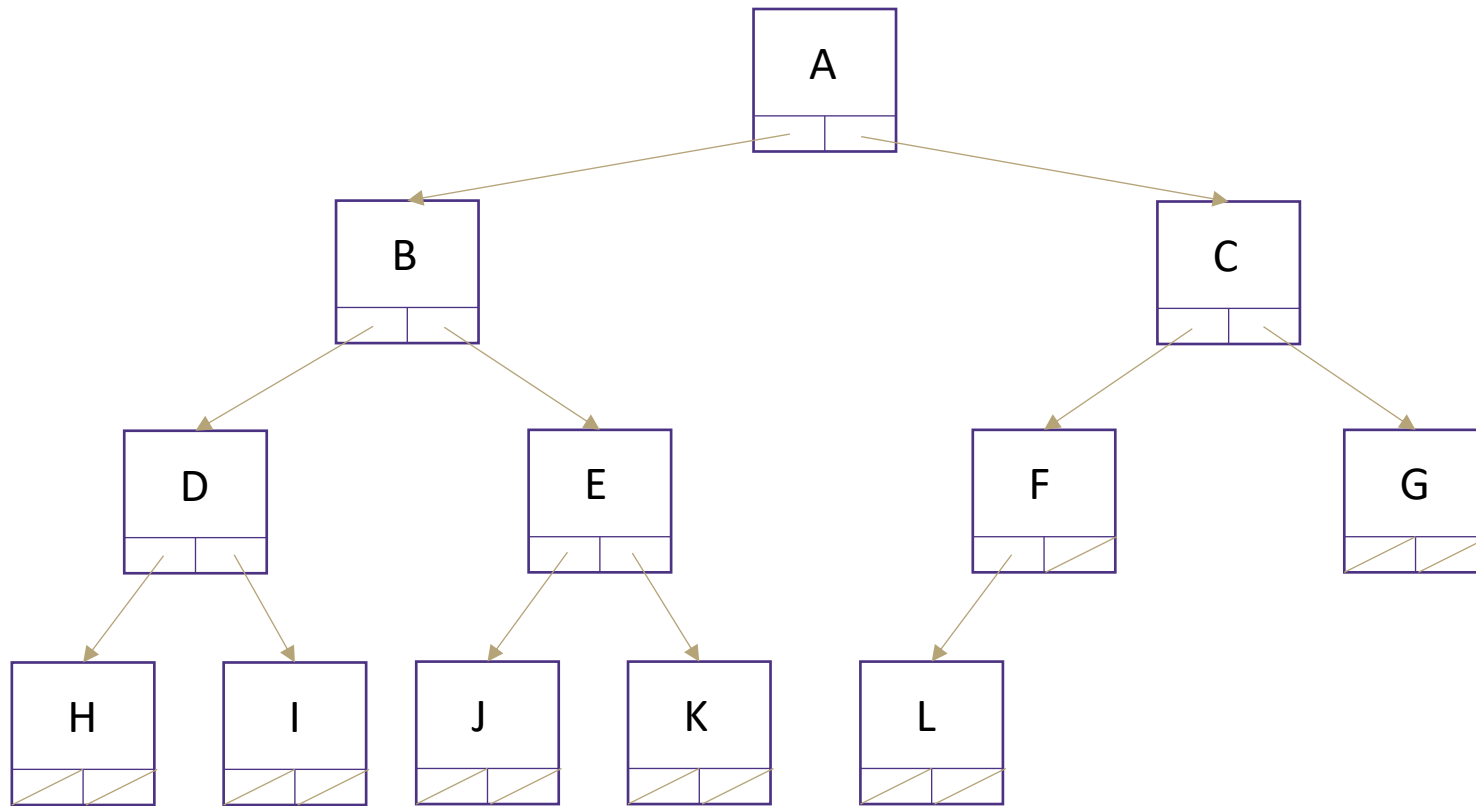
add(key):

- log n 1. Find next available spot and insert new node
- log n 2. Percolate up to fix heap invariant

Operation	Case	Runtime
removeMin()	best	$\Theta(1)$
	worst	$\Theta(\log n)$
add(key)	best	$\Theta(1)$
	worst	$\Theta(\log n)$

- Finding the "end" of the heap is hard!
  - Can do it in  $\Theta(\log n)$  time on complete trees with extra class variables by walking down
  - Fortunately, there's a better way 😊

# Implementing Heaps with an Array

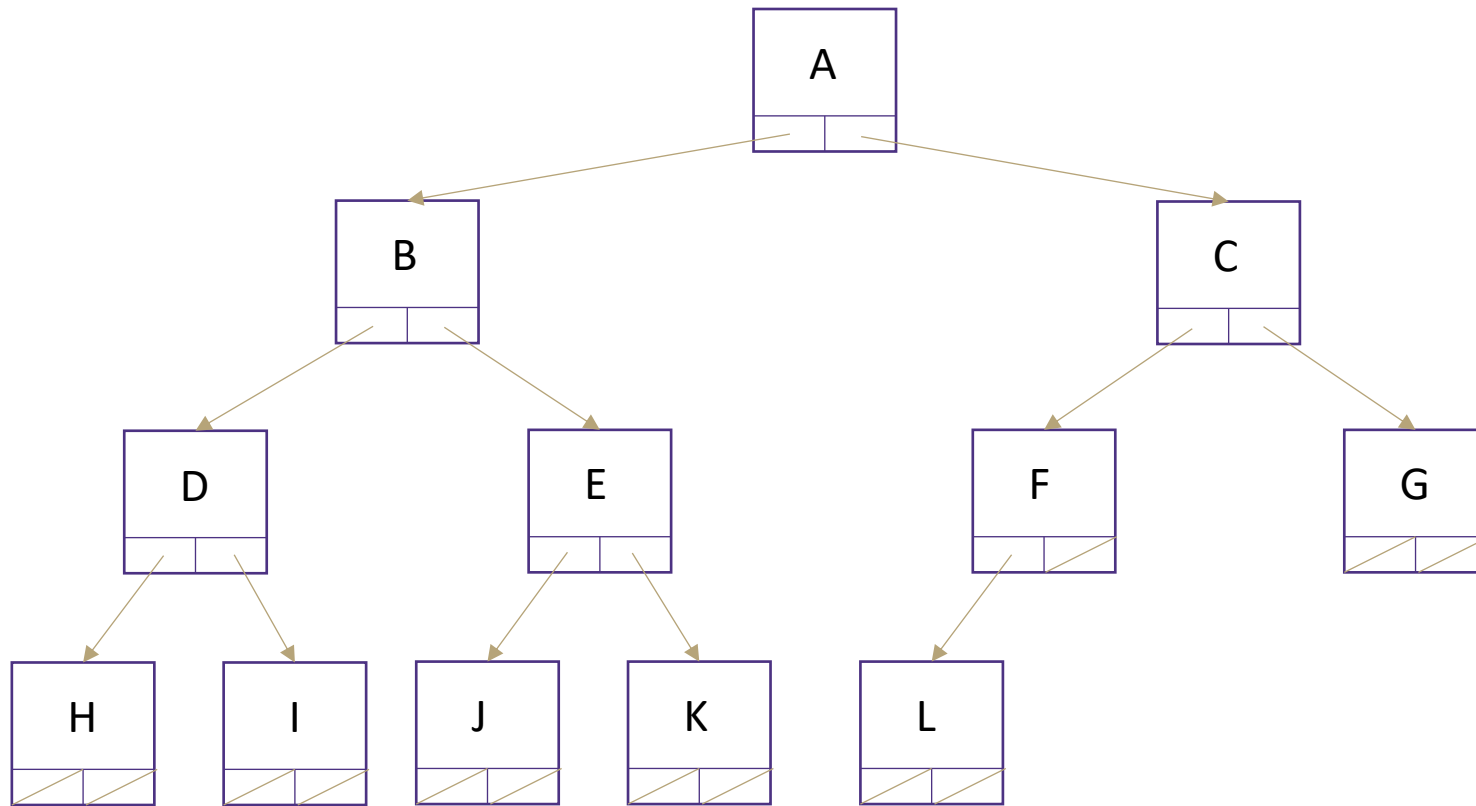


Fill array in **level-order** from left to right

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

- Map our binary tree heap representation into an array
  - Fill in the array in level order from left to right
  - Remember, heaps are complete trees – very predictable number of nodes on each level!
- Note: array implementation is how people almost always implement a heap
  - But tree drawing is good way to think of it conceptually!
  - **Everything we've discussed about the tree is still true** – these are just different ways of looking at the same thing

# Implementing Heaps with an Array



Fill array in **level-order** from left to right

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

Calculations to navigate array:

How do we find the minimum node?

$$\text{peekMin}() = \text{arr}[0]$$

How do we find the last node?

$$\text{lastNode}() = \text{arr}[\text{size} - 1]$$

How do we find the next open space?

$$\text{openSpace}() = \text{arr}[\text{size}]$$

How do we find a node's left child?

$$\text{leftChild}(i) = 2i + 1$$

How do we find a node's right child?

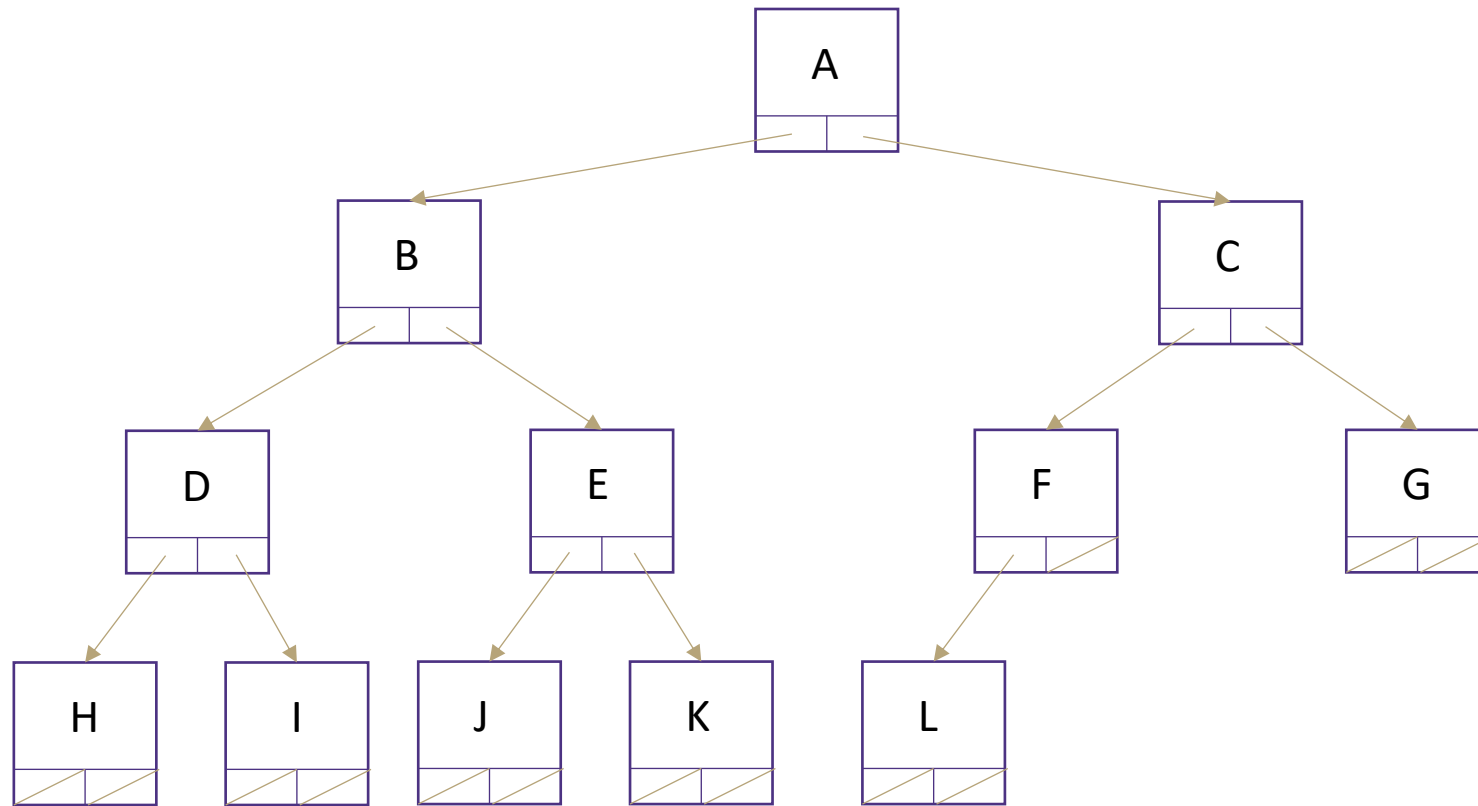
$$\text{rightChild}(i) = 2i + 2$$

How do we find a node's parent?

$$\text{parent}(i) = \frac{(i - 1)}{2}$$



# Implementing Heaps with an Array



Fill array in **level-order** from left to right

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

*Simplified* calculations to navigate array, **if we skip index 0**:

How do we find the minimum node?

$$\text{peekMin}() = \text{arr}[1]$$

How do we find the last node?

$$\text{lastNode}() = \text{arr}[\text{size}]$$

How do we find the next open space?

$$\text{openSpace}() = \text{arr}[\text{size} + 1]$$

How do we find a node's left child?

$$\text{leftChild}(i) = 2i$$

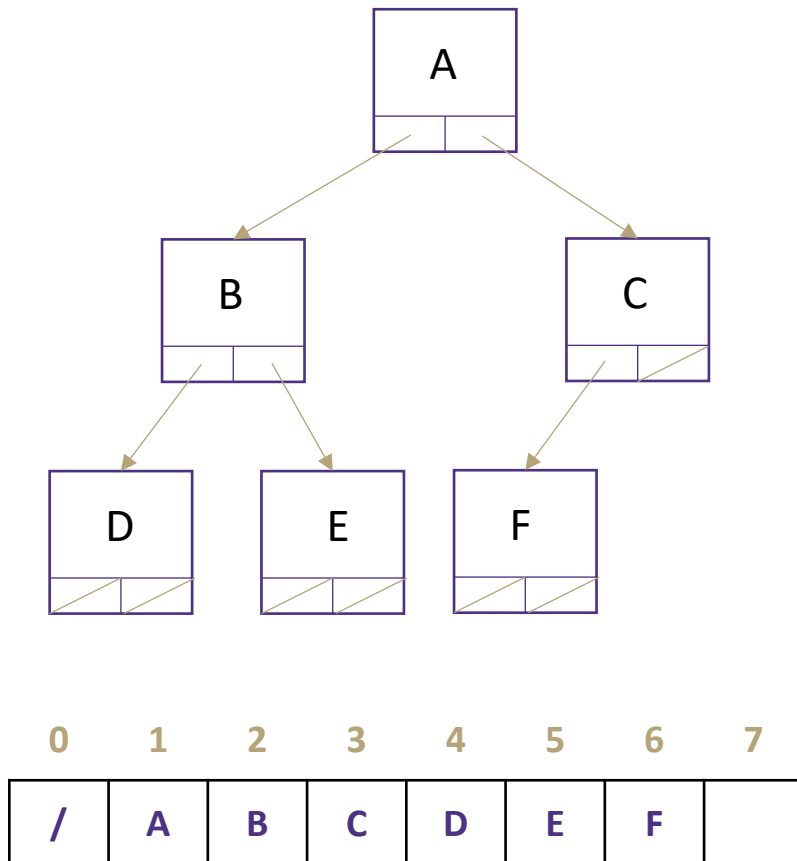
How do we find a node's right child?

$$\text{rightChild}(i) = 2i + 1$$

How do we find a node's parent?

$$\text{parent}(i) = \frac{i}{2}$$

# Array-Implemented MinHeap Runtimes



Operation	Case	Runtime
removeMin()	best	$\Theta(1)$
	worst	$\Theta(\log n)$
	in practice	$\Theta(\log n)$
add(key)	best	$\Theta(1)$
	worst	$\Theta(\log n)$
	in practice	$\Theta(1)$
peekMin()	all cases	$\Theta(1)$

- With array implementation, heaps match runtime of finding min in AVL trees
- But better in many ways!
  - Constant factors: array accesses give contiguous memory/spatial locality, tree constant factor shorter due to stricter height invariant
  - In practice, add doesn't require many swaps
  - WAY simpler to implement!

# AVL vs Heaps: Good For Different Situations

## HEAPS

- `removeMin`: much better constant factors than AVL Trees, though asymptotically the same
- `add`: in-practice, sweet sweet  $\Theta(1)$  (few swaps usually required)



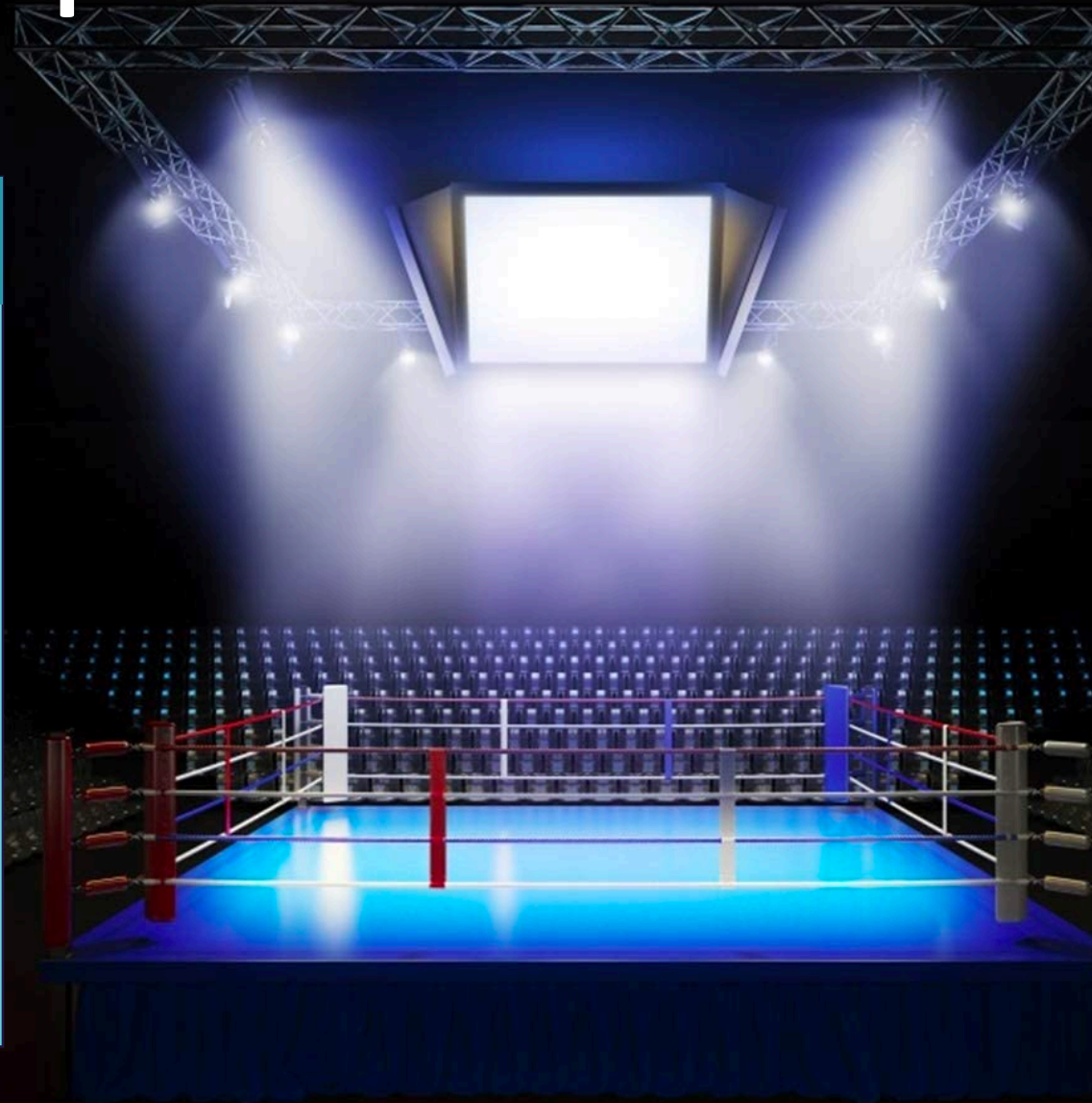
PriorityQueue

## AVL TREES

- `get`, `containsKey`: worst-case ( $\log n$ ) time (unlike Heap, which has to do a linear scan of the array)



Map/Set



# Lecture Outline

- **Heaps II**
  - Operations & Implementation
  - **Building a Heap** 
- Design Decisions
- Technical Interviews

# Building a Heap

- **buildHeap**(elements  $e_1, \dots, e_n$ ) – Given  $n$  elements, create a heap containing exactly those  $n$  elements.
- Idea 1: Call add  $n$  times.
  - Worst case runtime?
    - Each call takes logarithmic time, and there are  $n$  calls
    - $\Theta(n \log n)$
    - (Technically, the worst case is not this simple – you're not always going to hit logarithmic runtime because many insertions happen in a pretty empty tree – but this intuition is good enough)
  - Could we do better?

Operation	Case	Runtime
removeMin()	best	$\Theta(1)$
	worst	$\Theta(\log n)$
	in practice	$\Theta(\log n)$
add(key)	best	$\Theta(1)$
	worst	$\Theta(\log n)$
	in practice	$\Theta(1)$
peekMin()	all cases	$\Theta(1)$

# Can We Do Better?

- What's causing the  $n$  add strategy to take so long?
  - Most nodes are near the bottom, and might need to percolate all the way up.
- Idea 2: Dump everything in the array, and percolate things down until the heap invariant is satisfied
  - Intuition: this could be faster!
  - The bottom two levels of the tree have  $\Omega(n)$  nodes, the top two have 3 nodes
  - Maybe we can make “most of the nodes” go only a constant distance

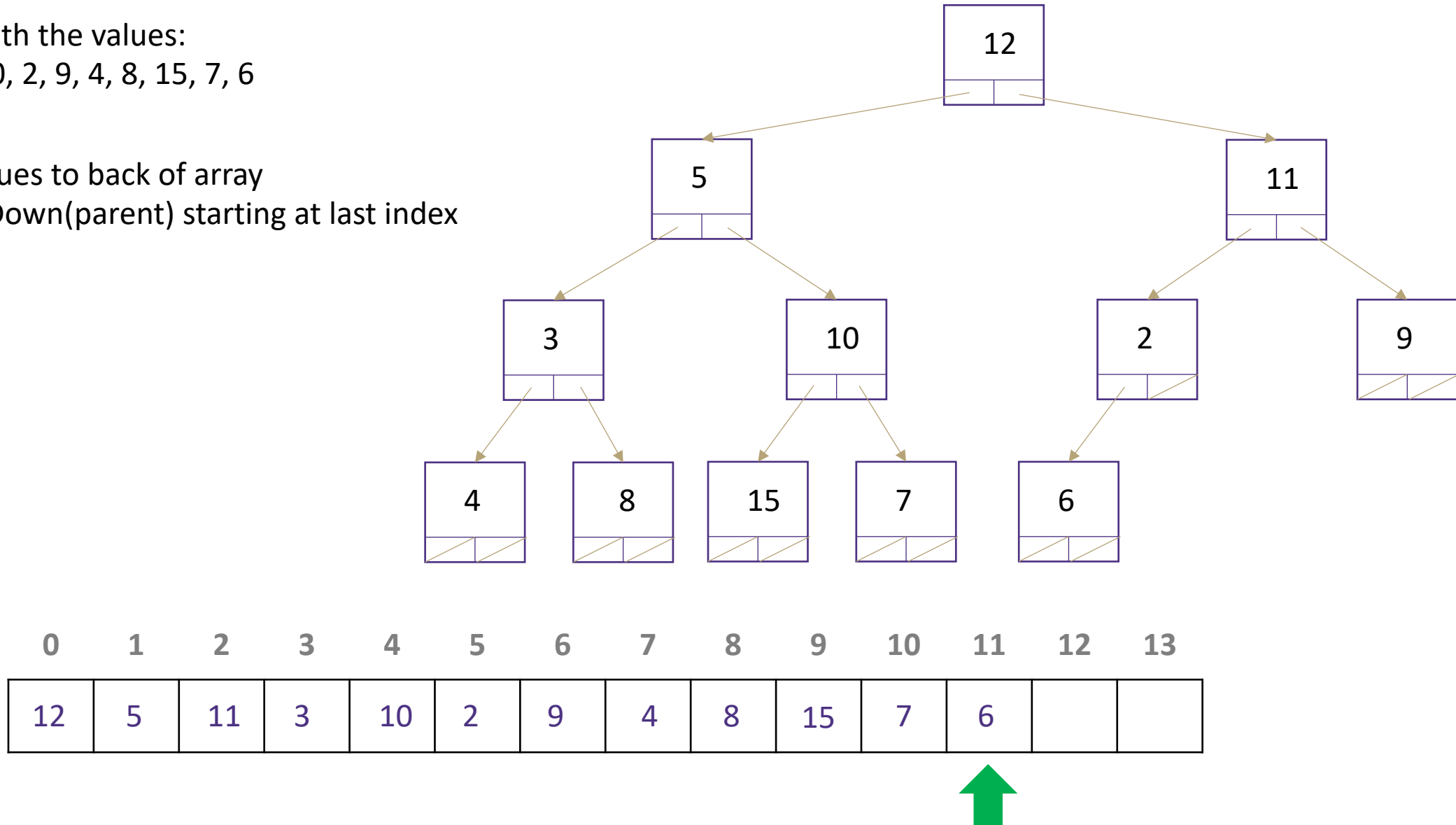


# Floyd's buildHeap algorithm

Build a tree with the values:

12, 5, 11, 3, 10, 2, 9, 4, 8, 15, 7, 6

1. Add all values to back of array
2. percolateDown(parent) starting at last index

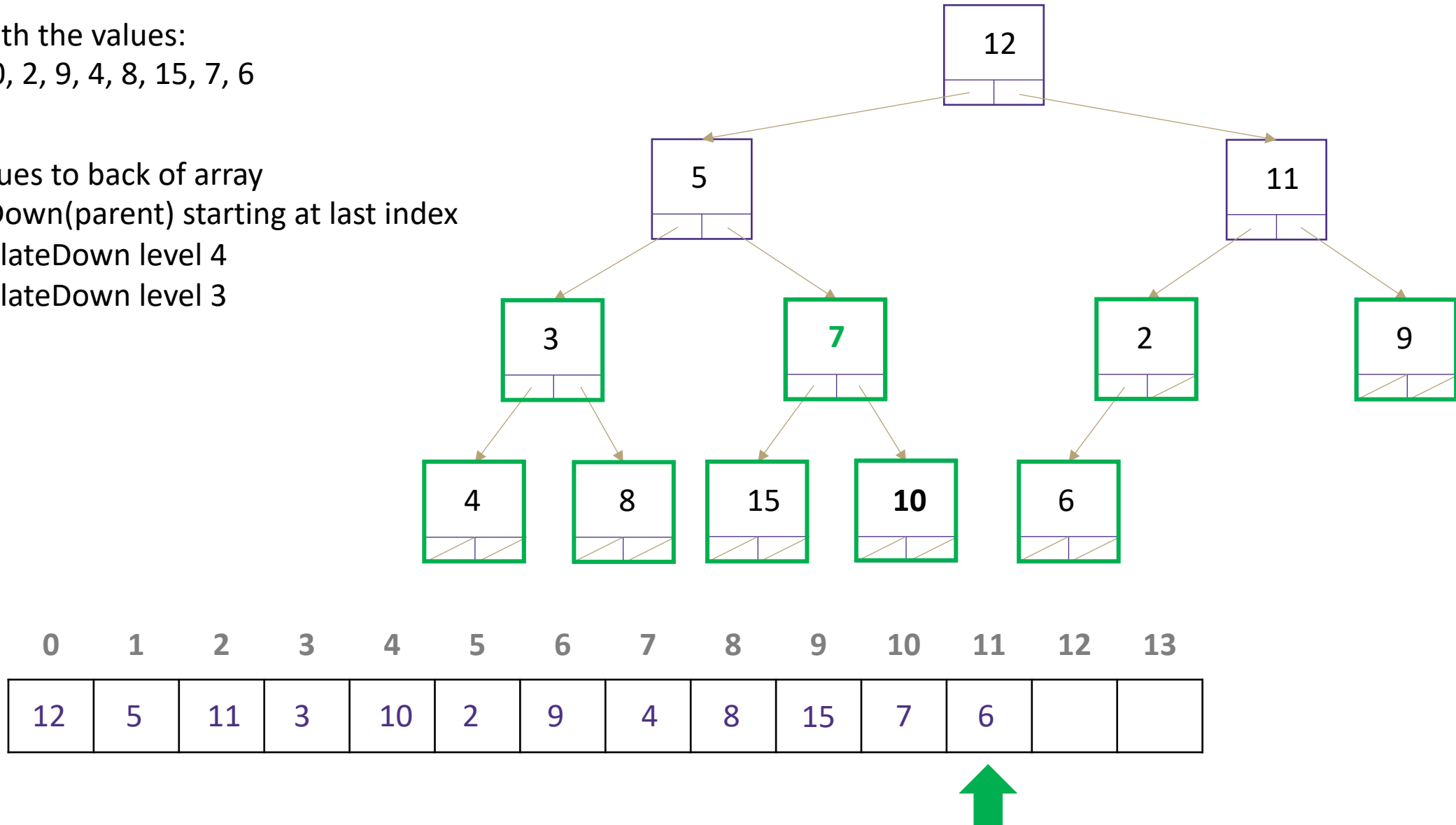


# Floyd's buildHeap algorithm

Build a tree with the values:

12, 5, 11, 3, 10, 2, 9, 4, 8, 15, 7, 6

1. Add all values to back of array
2. percolateDown(parent) starting at last index
  1. percolateDown level 4
  2. percolateDown level 3



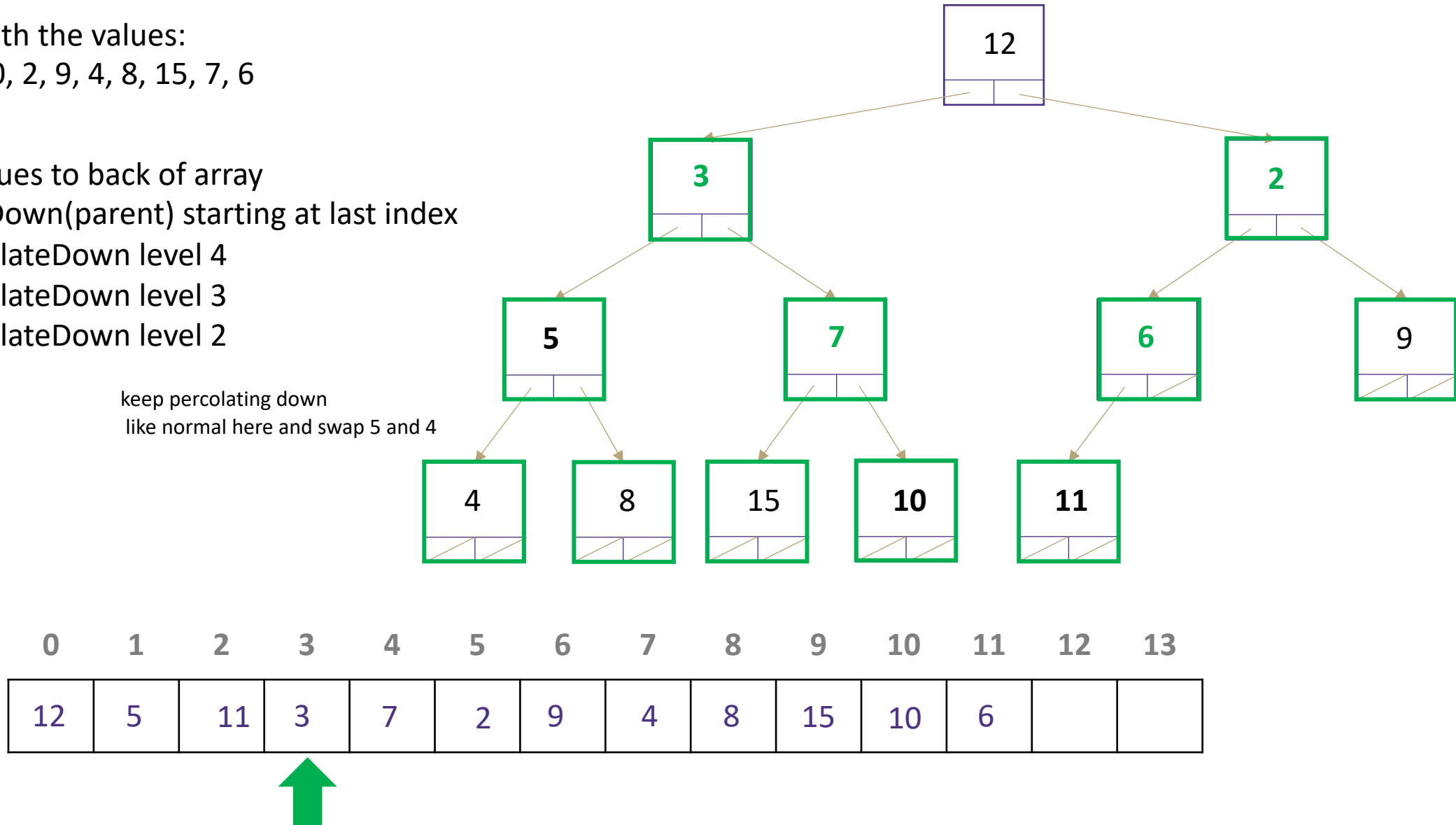
# Floyd's buildHeap algorithm

Build a tree with the values:

12, 5, 11, 3, 10, 2, 9, 4, 8, 15, 7, 6

1. Add all values to back of array
2. `percolateDown(parent)` starting at last index
  1. `percolateDown` level 4
  2. `percolateDown` level 3
  3. `percolateDown` level 2

keep percolating down  
like normal here and swap 5 and 4

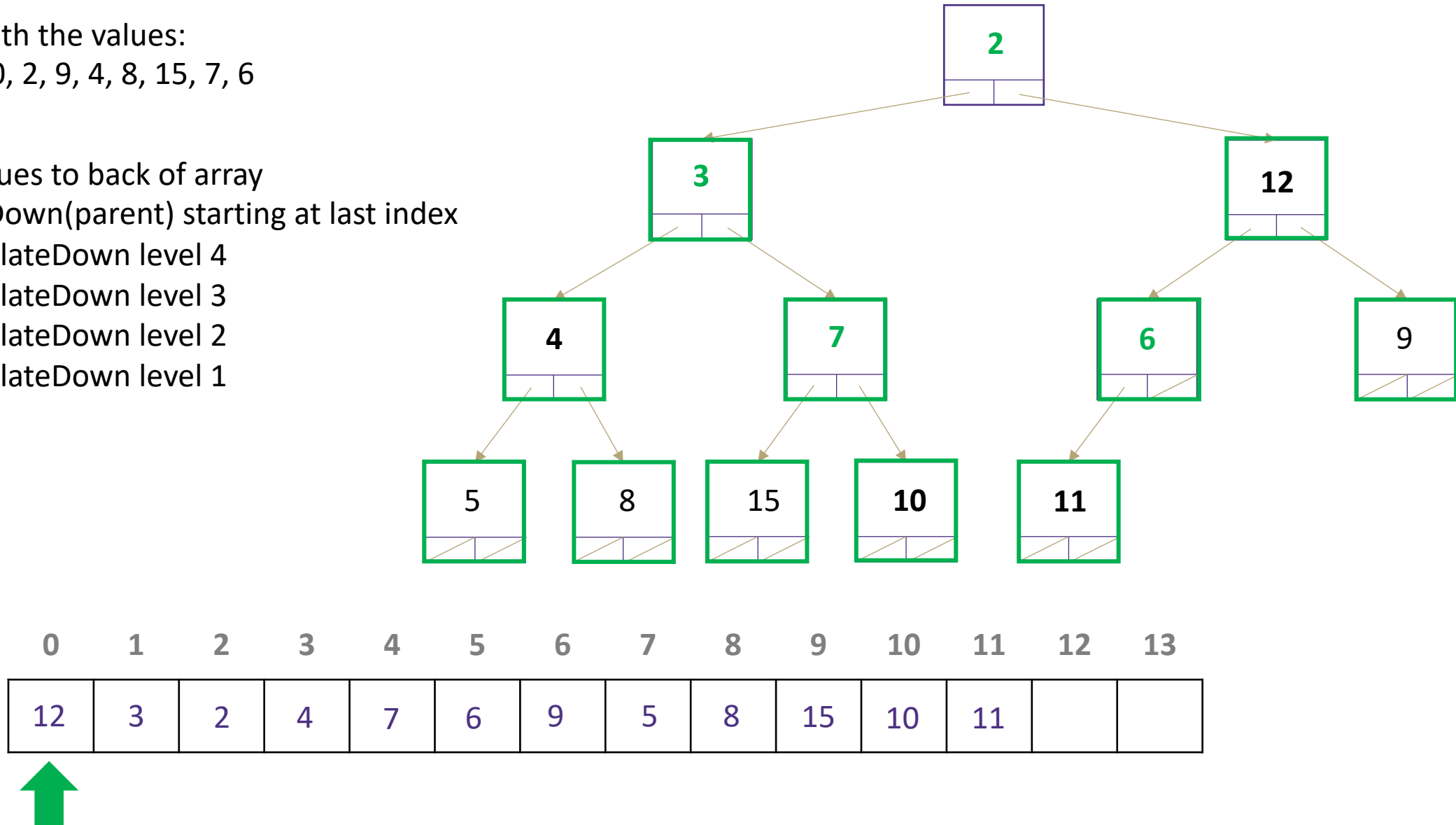


# Floyd's buildHeap algorithm

Build a tree with the values:

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1. Add all values to back of array
2. `percolateDown(parent)` starting at last index
  1. `percolateDown` level 4
  2. `percolateDown` level 3
  3. `percolateDown` level 2
  4. `percolateDown` level 1

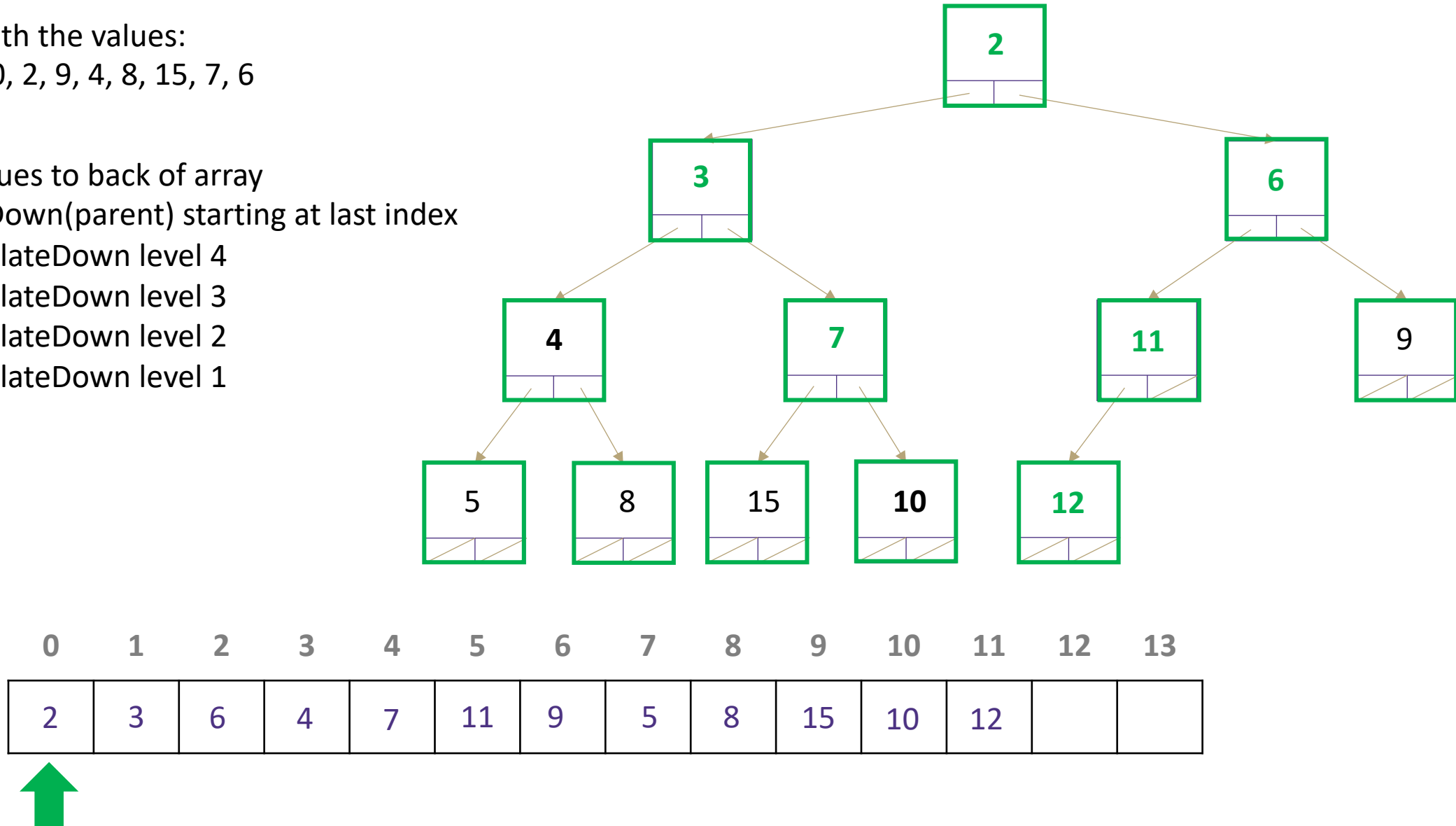


# Floyd's buildHeap algorithm

Build a tree with the values:

12, 5, 11, 3, 10, 2, 9, 4, 8, 15, 7, 6

1. Add all values to back of array
2. percolateDown(parent) starting at last index
  1. percolateDown level 4
  2. percolateDown level 3
  3. percolateDown level 2
  4. percolateDown level 1



# Is It Really Faster?

Floyd's buildHeap runs in  $O(n)$  time!

- percolateDown() has worst case  $\log n$  in general, but for most of these nodes, it has a much smaller worst case!
  - $n/2$  nodes in the tree are leaves, have 0 levels to travel
  - $n/4$  nodes have at most 1 level to travel
  - $n/8$  nodes have at most 2 levels to travel
  - etc...

- $$\text{worst-case-work}(n) \approx \underbrace{\frac{n}{2} \cdot 1}_{\text{much of the work}} + \underbrace{\frac{n}{4} \cdot 2}_{\text{a little less}} + \underbrace{\frac{n}{8} \cdot 3}_{\text{a little less}} + \cdots + \underbrace{1 \cdot (\log n)}_{\text{barely anything}}$$

- Intuition: Even though there are  $\log n$  levels, each level does a smaller and smaller amount of work. Even with infinite levels, as we sum smaller and smaller values (think  $\frac{1}{2^i}$ ) we converge to a constant factor of  $n$ .



# Optional Slide Floyd's buildHeap Summation

- $n/2 \cdot 1 + n/4 \cdot 2 + n/8 \cdot 3 + \dots + 1 \cdot (\log n)$

factor out n

$$\text{work}(n) \approx n \left( \frac{1}{2} + \frac{2}{4} + \frac{3}{8} + \dots + \frac{\log n}{n} \right) \text{ find a pattern } \rightarrow \text{powers of 2} \quad \text{work}(n) \approx n \left( \frac{1}{2^1} + \frac{2}{2^2} + \frac{3}{2^3} + \dots + \frac{\log n}{2^{\log n}} \right) \text{ Summation!}$$

$$\text{work}(n) \approx n \sum_{i=1}^? \frac{i}{2^i} \quad ? = \text{upper limit should give last term}$$

We don't have a summation for this! Let's make it look more like a summation we do know.

Infinite geometric series

$$\text{work}(n) \leq n \sum_{i=1}^{\log n} \frac{\left(\frac{3}{2}\right)^i}{2^i} \quad \text{if } -1 < x < 1 \text{ then } \sum_{i=0}^{\infty} x^i = \frac{1}{1-x} = x \quad \text{work}(n) \approx n \sum_{i=1}^{\log n} \frac{i}{2^i} \leq n \sum_{i=0}^{\infty} \left(\frac{3}{4}\right)^i = n * 4$$

Floyd's buildHeap runs in O(n) time!

# Project 3

- Build a heap! Alongside hash maps, heaps are one of the most useful data structures to know – and pop up many more times this quarter!
  - You'll also get practice using multiple data structures together to implement an ADT!
  - Directly apply the invariants we've talked so much about in lecture! Even has an invariant checker to verify this (a *great* defensive programming technique!)

## MIN PRIORITY QUEUE ADT

### State

Set of comparable values (*ordered based on "priority"*)

### Behavior

**add(value)** – add a new element to the collection

**removeMin()** – returns the element with the smallest priority, removes it from the collection

**peekMin()** – find, but do not remove the element with the smallest priority

***changePriority(item, priority)*** – *update the priority of an element*

***contains(item)*** – *check if an element exists in the priority queue*

# Project 3 Tips

- Project 3 adds `changePriority` and `contains` to the `PriorityQueue` ADT, which aren't efficient on a heap alone
- **You should utilize an extra data structure for `changePriority`!**
  - Doesn't affect *correctness* of PQ, just runtime. Please use a built-in Java collection instead of implementing your own (although you could in theory).
- **`changePriority` Implementation Strategy:**
  - implement without regards to efficiency (without the extra data structure) at first
  - analyze your code's runtime and figure out which parts are inefficient
  - reflect on the data structures we've learned and see how any of them could be useful in improving the slow parts in your code

## MIN PRIORITY QUEUE ADT

### State

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### Behavior

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***`changePriority(item, priority)`*** – *update the priority of an element*

***`contains(item)`*** – *check if an element exists in the priority queue*

# Lecture Outline

- Heaps II
  - Operations & Implementation
  - Building a Heap
- **Technical Interviews** 

# Beyond CSE 373: Industry

- Many people take CSE 373 because they're interested in a career related to software engineering or more broadly CS
  - If not, that's totally okay too! There are so many good reasons to take this class and ways to apply it; you don't have to be interested in software engineering specifically
  - Perhaps you've become more interested over the course of the quarter
- If this sounds like you, it's never too early to start thinking about preparing for a job/internship hunt!
  - We'll talk about this a few more times throughout the course
  - But a few highlights now to get you thinking!

# The Technical Interview Process

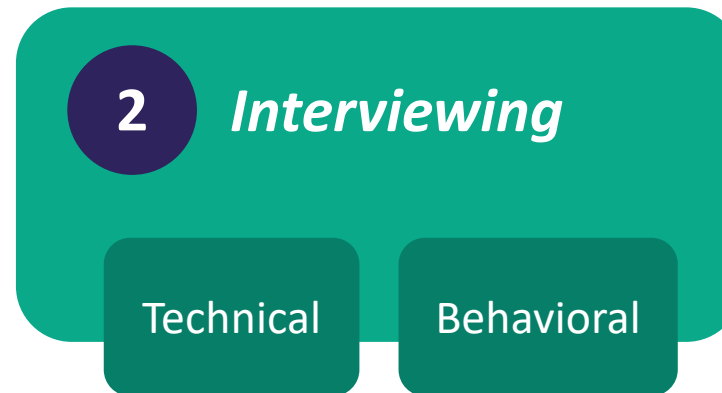
PREPARATION

- Putting together your resume
- Personal projects/other experience to help you stand out
- Identifying where to apply, and when

## CSE 373 MATERIAL

- Practicing with interview problems & design decisions
- Identifying common patterns in interview questions

PROCESS



*Job/Internship Offer*



## 1

# Applying

***Start Here:***

<http://bit.ly/cseresumeguide>

An incredibly useful guide to fleshing out your resume, highlighting your experience to make you stand out, Do's and Don'ts of what to include

- <http://bit.ly/csestorycrafting> – how to turn your experience into a cohesive story that stands out to recruiters
- [UW Career & Internship Center](#) – tons of helpful articles, *especially* for online job hunting!
- [College of Engineering Career Center](#) – schedule an appointment to talk to a career counselor

## 2

# Interviewing

## This is where CSE 373 comes in!

- Technical interviews *love* to ask about maps, trees, heaps, recursive algorithms, and **especially algorithmic analysis!**
- Making design decisions and determining tradeoffs between data structures is a crucial skill! Fortunately, you've been practicing all quarter 😊

- [Leetcode](#) and [Hackerrank](#) – *tons* of practice interview questions. If you're feeling nervous, it always helps to practice 😊
  - Check out #career-prep on Discord! Your amazing TA Joyce has been highlighting example interview problems, and more and more options available as we learn more this quarter!
- [The Secrets No One Told You About Technical Interviews](#) – fantastic article (& by previous 373 instructor Kasey Champion!)
- [Approaching Technical Interview Questions](#) – when you get asked a question that stumps you, what should you do?

# A+ Advice for Getting a Software Job

- A guest lecture by the amazing Kim Nguyen, former Career Coach for UW CSE, **specifically for 373 students!**
  - Back when a “lecture” was something that happened in person...
  - If you're overwhelmed with all these resources, **highly recommend you start here!** Kim is the absolute best!

