# CS101 Data Structure

Heaps and Priority Queues
Textbook Ch 6

## Outline

- Priority queue
- Binary heap
- Heapsort

### **Definition**

#### Queues

The order may be summarized by first in, first out

#### Priority queues

- Each object is associated with a priority
  - The value 0 has the *highest* priority, and
  - The higher the number, the lower the priority
- We pop the object which has the highest priority

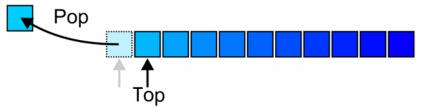
## **Operations**

The top of a priority queue is the object with highest priority

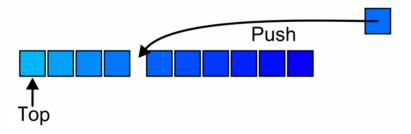


Popping from a priority queue removes the current highest priority

object:



Push places a new object into the appropriate place



# **Application**

#### Process priority in operation systems

- In Unix, you may set the priority of a process, e.g.,

```
% nice +15 ./a.out
```

reduces the priority of the execution of the routine a.out by 15

### **Implementations**

Our goal is to make the run time of each operation as close to  $\Theta(1)$  as possible

We will look at an implementation using a data structure we already know:

Multiple queues — one for each priority

Then we will introduce a more appropriate data structure: *heap* 

## Multiple Queues

Assume there is a fixed number of priorities, say M

- Create an array of M queues
- Push a new object onto the queue corresponding to the priority
- Top and pop find the first non-empty queue with highest priority

# Multiple Queues

#### The run times are reasonable:

- Push is  $\Theta(1)$
- Top and pop are both O(M)

#### Problems:

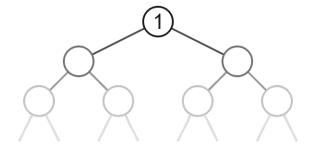
- It restricts the range of priorities
- The memory requirement is  $\Theta(M+n)$

## Heaps

#### Can we do better?

#### We need a heap

- A tree with the top object at the root
- We will look at binary heaps
- Numerous other heaps exists:
  - *d*-ary heaps
  - Leftist heaps
  - Skew heaps
  - Binomial heaps
  - Fibonacci heaps
  - Bi-parental heaps



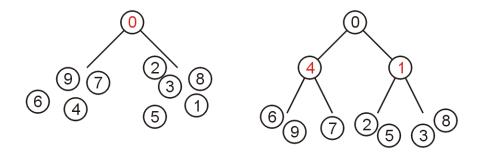
### Outline

- Priority queue
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- Heapsort

#### **Definition**

A non-empty tree is a min-heap if

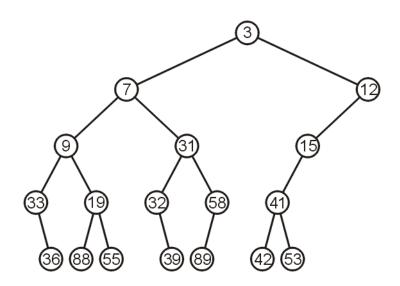
- The key associated with the root is less than or equal to the keys associated with the sub-trees (if any)
- The sub-trees (if any) are also min-heaps



There is no other relationship between the elements in the subtrees!

# Example

This is a (naive) binary min-heap:



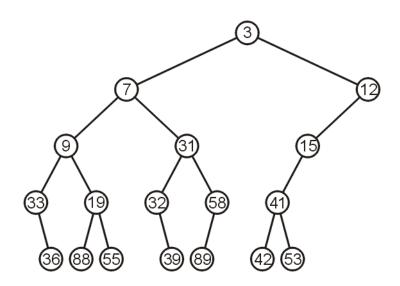
# Operations

We will consider three operations:

- Тор
- Pop
- Push

# Example

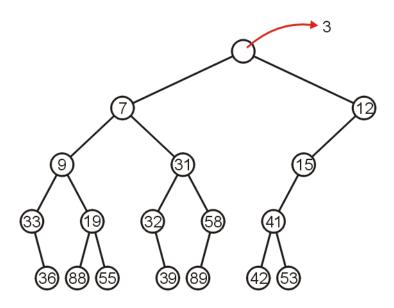
We can find the top object in  $\Theta(1)$  time: 3



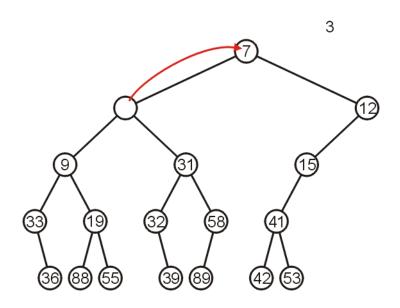
#### To remove the minimum object:

- Promote the node of the sub-tree which has the least value
- Recursively process the sub-tree from which we promoted the least value

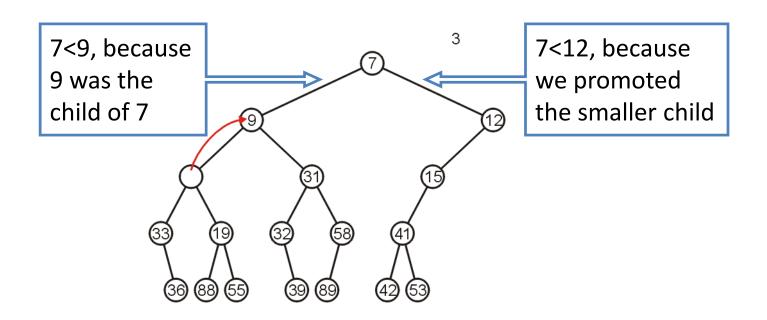
Using our example, we remove 3:



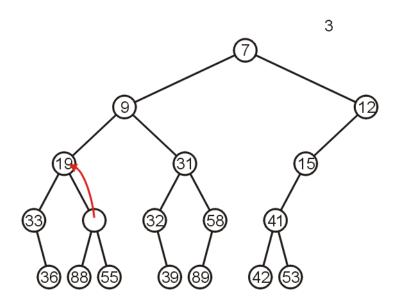
We promote 7 (the minimum of 7 and 12) to the root:



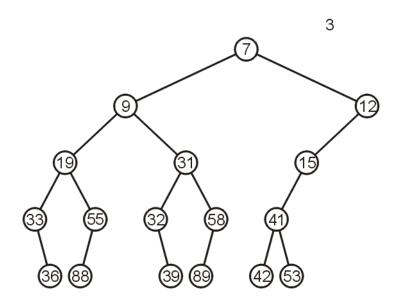
In the left sub-tree, we promote 9:



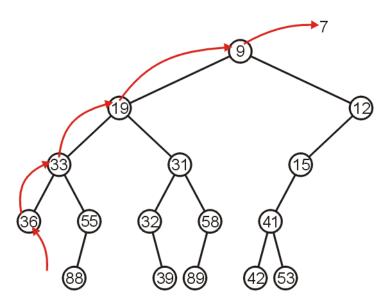
Recursively, we promote 19:



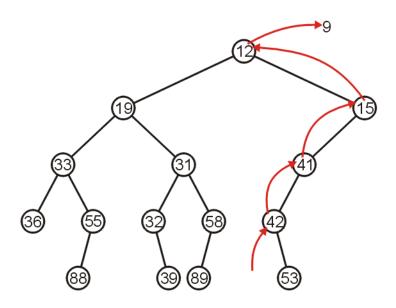
Finally, 55 is a leaf node, so we promote it and delete the leaf



Repeating this operation again, we can remove 7:



If we remove 9, we must now promote from the right sub-tree:



Inserting into a heap may be done either:

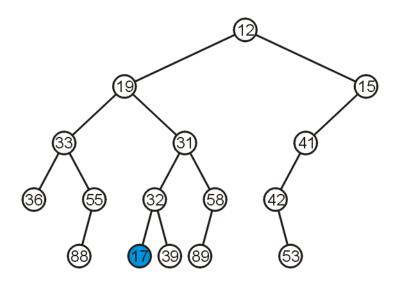
- At a leaf (move it up if it is smaller than the parent)
- At the root (insert the larger object into one of the subtrees)

We will use the first approach with binary heaps

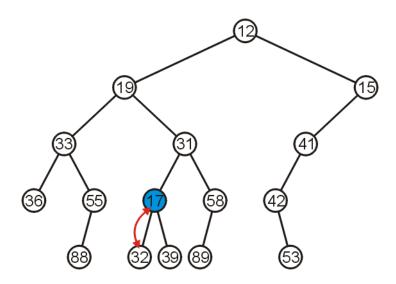
Other heaps use the second

### Inserting 17 into the last heap

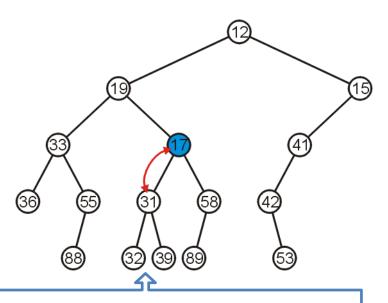
Select an arbitrary node to insert a new leaf node:



The node 17 is less than the node 32, so we swap them

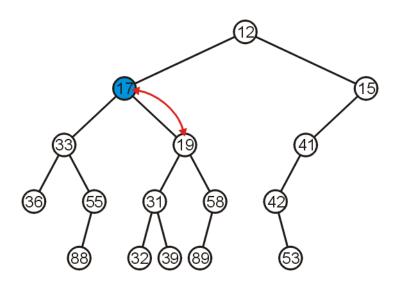


The node 17 is less than the node 31; swap them

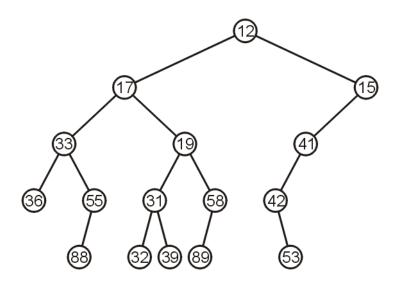


31 is smaller than 32 and 39 because 31 was the ancestor of 32 and 39

The node 17 is less than the node 19; swap them



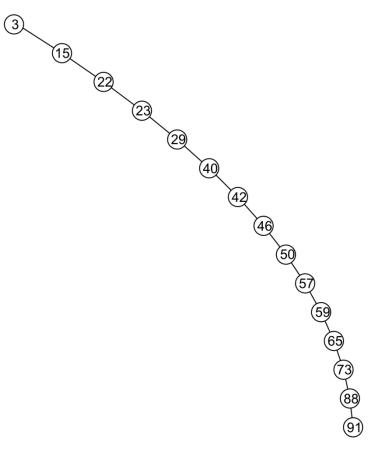
The node 17 is greater than 12 so we are finished



This process is called *percolation*, that is, the lighter (smaller) objects move up from the bottom of the min-heap

# Time Complexity

- Time complexity of pop and push?
  - O(n)
  - Worst case: the binary tree is highly unbalanced
- Can we do better?
  - Keep balance of the binary tree



#### Balance

There are multiple means of keeping balance with binary heaps:

- Complete binary trees
- Leftist heaps
- Skew heaps

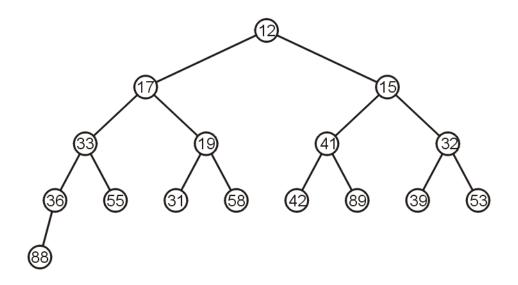
This defines the actual "binary heap"

We will look at using complete binary trees

It has optimal memory characteristics but sub-optimal run-time characteristics

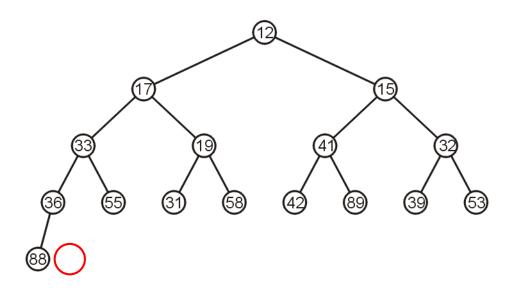
# Complete Trees

For example, the previous heap may be represented as the following (non-unique!) complete tree:



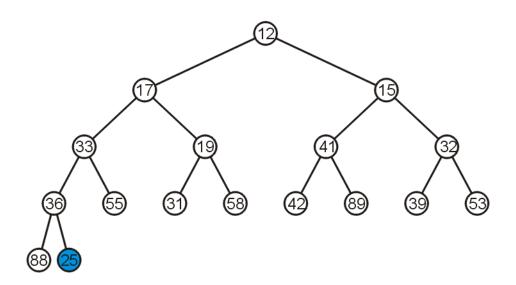
# Complete Trees: Push

If we insert into a complete tree, we need only place the new node as a leaf node in the appropriate location and percolate up



# Complete Trees: Push

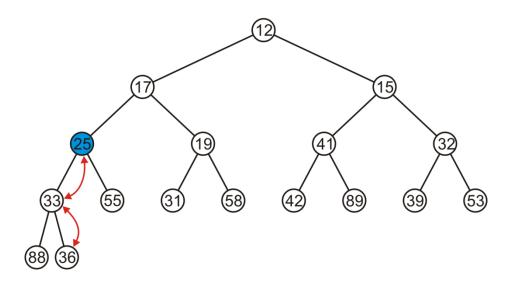
For example, push 25:



## Complete Trees: Push

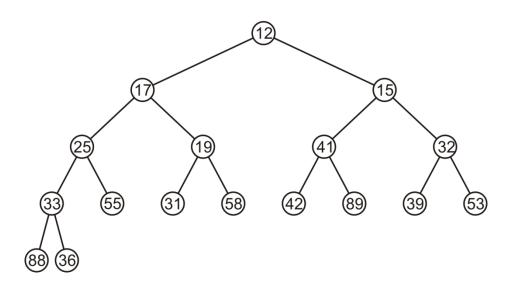
We have to percolate 25 up into its appropriate location

The resulting heap is still a complete tree

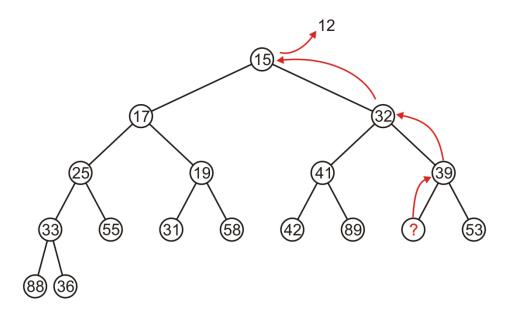


# Complete Trees: Pop

Suppose we want to pop the top entry: 12

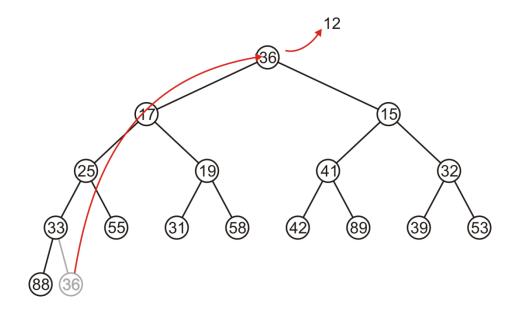


Percolating up creates a hole leading to a non-complete tree



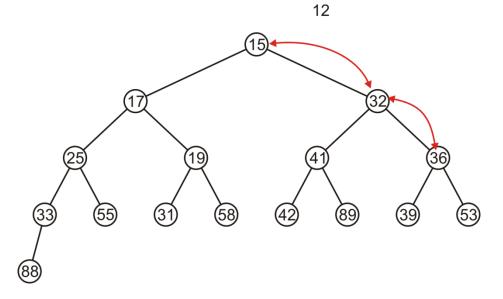
What's wrong?

Instead, copy the last entry in the heap to the root

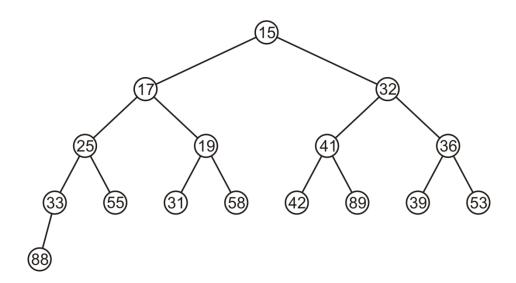


Now, percolate 36 down swapping it with the smallest of its children

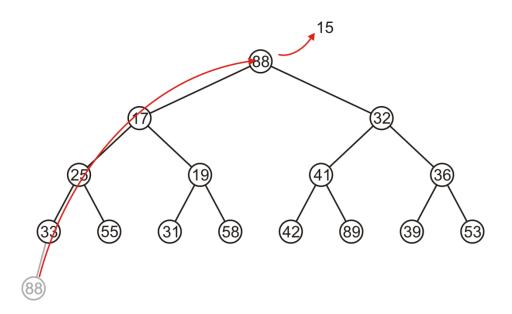
We halt when both children are larger



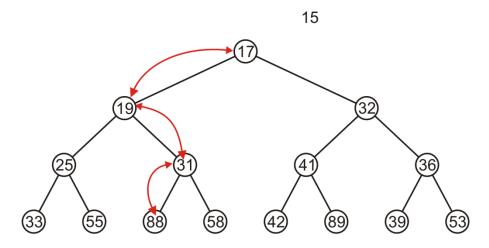
The resulting tree is now still a complete tree:



Again, popping 15, copy up the last entry: 88



This time, it gets percolated down to the point where it has no children



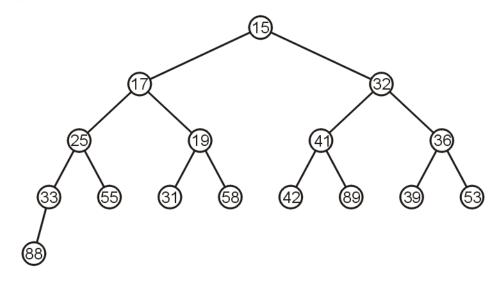
## Complete Tree

Therefore, we can maintain the complete-tree shape of a heap

#### We may store a complete tree using an array:

The array is filled using breadth-first traversal on the tree

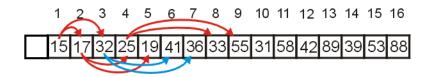
For the heap



a breadth-first traversal yields:

15 17 32 25 19 41 36 33 55 31 58 42 89 39 53 88

We start at index 1 when filling the array.



Given the entry at index k, it follows that:

```
- The parent of node is a k/2 parent = k >> 1;
```

```
- the children are at 2k and 2k+1 left_child = k << 1; right_child = left_child | 1;
```

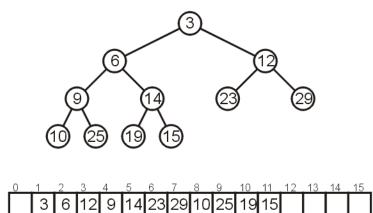
If the heap-as-array has **count** entries, then the next empty node in the corresponding complete tree is at location **posn** = **count** + **1** 

We compare the item at location posn with the item at posn/2

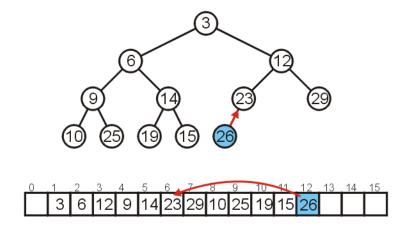
If they are out of order

- Swap them
- Set posn /= 2 and repeat

Consider the following heap, both as a tree and in its array representation

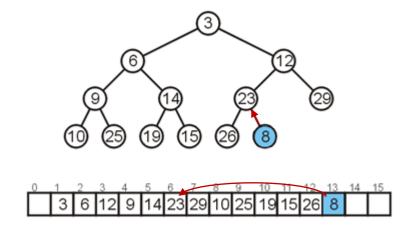


Inserting 26 requires no changes

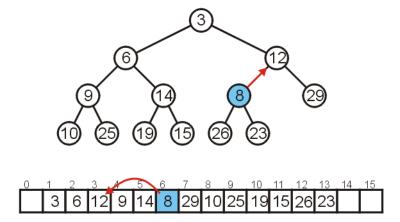


Inserting 8 requires a few percolations:

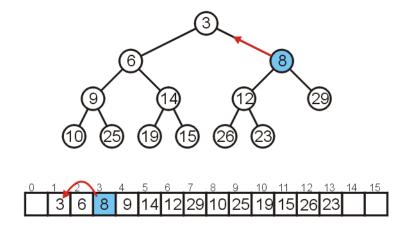
Swap 8 and 23



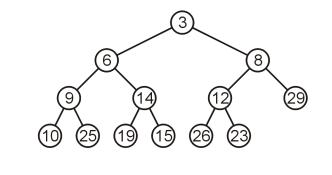
Swap 8 and 12



At this point, it is greater than its parent, so we are finished



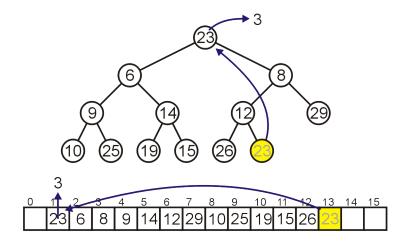
As before, popping the top has us copy the last entry to the top



 1
 2
 3
 4
 5
 6
 7
 8
 9
 10
 11
 12
 13

 3
 6
 8
 9
 14
 12
 29
 10
 25
 19
 15
 26
 23

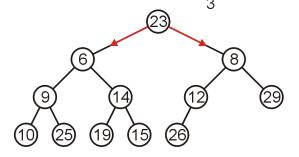
As before, popping the top has us copy the last entry to the top



Now percolate down

Compare Node 1 with its children: Nodes 2 and 3

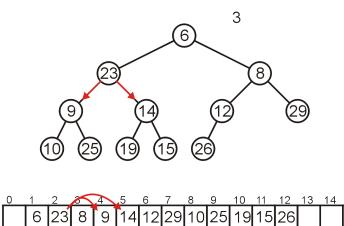
- Swap 23 and 6



0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
	23	6	8	9	14	12	29	10	25	19	15	26			

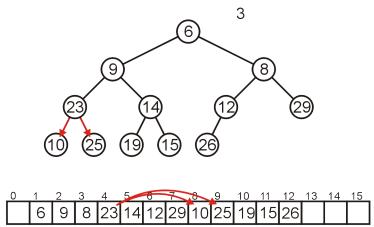
Compare Node 2 with its children: Nodes 4 and 5

Swap 23 and 9



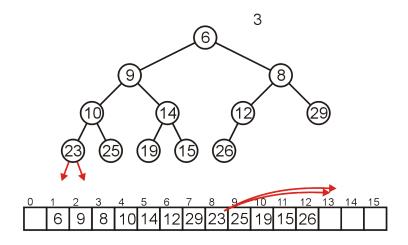
Compare Node 4 with its children: Nodes 8 and 9

- Swap 23 and 10



The children of Node 8 are beyond the end of the array:

Stop



#### Run-time Analysis

Accessing the top object is  $\Theta(1)$ 

Popping the top object is  $O(\ln(n))$ 

 We copy something that is already in the lowest depth—it will likely be moved back to the lowest depth

Pushing an object is also  $O(\ln(n))$ 

If we insert an object less than the root, it will be moved up to the top

Space complexity O(n)

So binary heap is a better implementation of priority queue

## Run-time Analysis

If we are inserting an object less than the root (at the front), then the run time will be  $\Theta(\ln(n))$ 

If we insert at the back (greater than any object) then the run time will be  $\Theta(1)$ 

How about an arbitrary insertion?

- It will be  $O(\ln(n))$ ? Could the average be less?

## Run-time Analysis

With each percolation, it will move an object past half of the remaining entries in the tree

 Therefore after one percolation, it will probably be past half of the entries, and therefore on average will require no more percolations

$$\frac{1}{n} \sum_{k=0}^{h} (h-k) 2^{k} = \frac{2^{h+1} - h - 2}{n}$$
$$= \frac{n-h-1}{n} = \mathbf{\Theta}(1)$$

Therefore, we have an average run time of  $\Theta(1)$ 

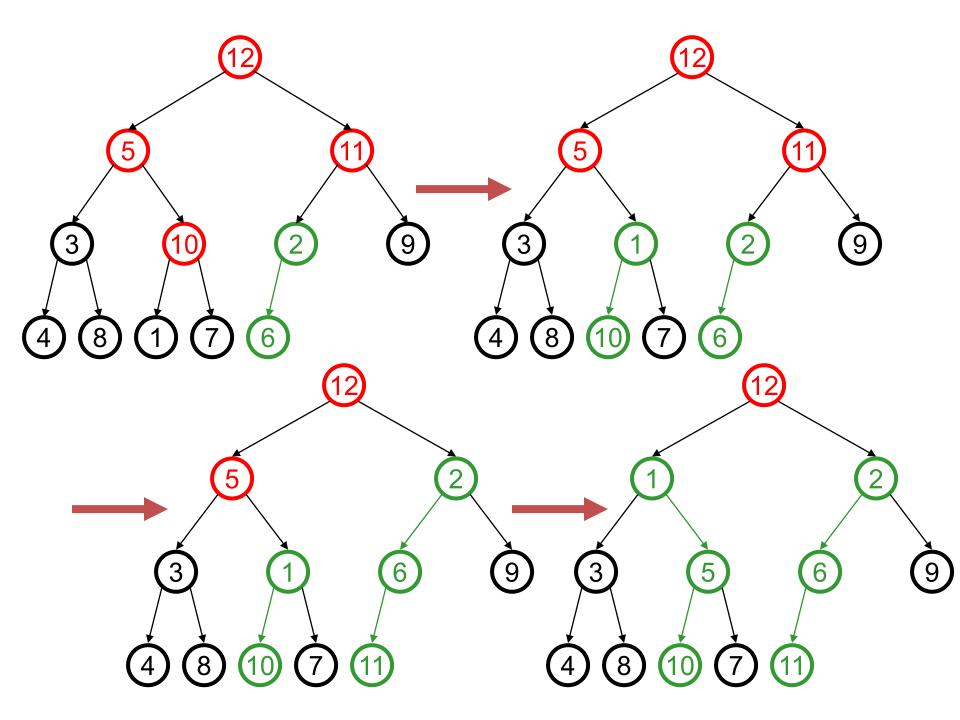
## **Build Heap**

- Task: Given a set of n keys, build a heap all at once
- Approach 1
  - Repeatedly perform push
- Complexity
  - $O(n \ln(n))$

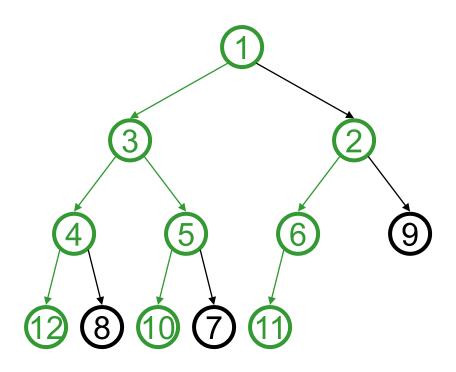
## Floyd's Method

Put the keys in a binary tree and fix the heap property!

```
buildHeap() {
  for (i=size/2; i>0; i--)
     percolateDown(i);
          5
                   3
                             6
                       10
                                 9
                                      4
```



Finally...



#### Complexity of Build Heap

- No percolation for the leaf nodes (n/2 nodes)
- At most n/4 nodes percolate down 1 level at most n/8 nodes percolate down 2 levels at most n/16 nodes percolate down 3 levels

. . .

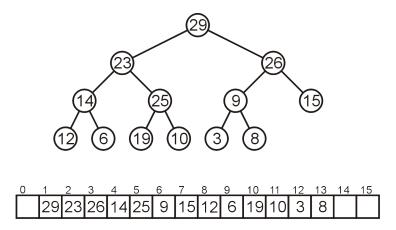
$$1\frac{n}{4} + 2\frac{n}{8} + 3\frac{n}{16} + \dots = \sum_{i=1}^{\log n} i \frac{n}{2^{i+2}} = \frac{n}{4} \sum_{i=1}^{\log n} \frac{i}{2^i} = \frac{n}{4} 2 = \frac{n}{2}$$

 $\Theta(n)$ 

## Binary Max Heaps

A binary max-heap is identical to a binary min-heap except that the parent is always larger than either of the children

For example, the same data as before stored as a max-heap yields



#### Outline

- Priority queue
- Binary heap
- Heapsort

## Heapsort

- Sorting
  - take a list of objects  $(a_0, a_1, ..., a_{n-1})$
  - return a reordering  $(a'_0, a'_1, ..., a'_{n-1})$  such that  $a'_0 \le a'_1 \le \cdots \le a'_{n-1}$
- Heapsort
  - Place the objects into a heap
    - O(*n*) time
  - Repeatedly popping the top object until the heap is empty
    - O(*n* In(*n*)) time
  - Time complexity:  $O(n \ln(n))$

#### Problem:

- This solution requires additional memory: a min-heap of size n
- This requires  $\Theta(n)$  memory

If the unsorted objects are <u>stored in an array</u>, is it possible to perform a heap sort in place, that is, require at most  $\Theta(1)$  memory (a few extra variables)?

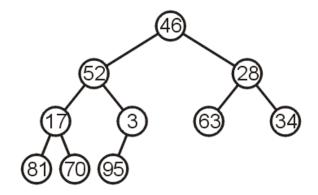
Instead of implementing a min-heap, consider a max-heap:

The maximum element is at the top of the heap

We then repeatedly pop the top object and move it to the end of the array.

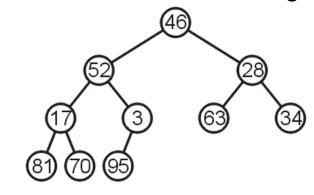
Now, consider this unsorted array:

This array represents the following complete tree:



Now, consider this unsorted array:

Because we start at 0 (instead of 1 as in array storage of complete trees), we need different formulas for finding the children and parent



Children

$$2*k + 1 2*k + 2$$

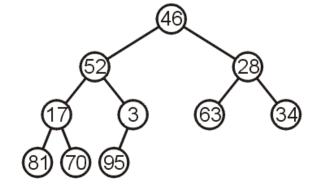
**Parent** 

$$(k + 1)/2 - 1$$

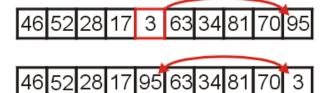
First, we must convert the unordered array with n = 10 elements into a max-heap

None of the leaf nodes need to be percolated down, and the last non-leaf node is in position n/2-1

Thus we start with position 10/2-1=4



We compare 3 with its child and swap them



We compare 17 with its two children and swap it with the maximum child (81)



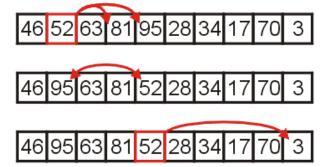
We compare 28 with its two children, 63 and 34, and swap it with the largest child



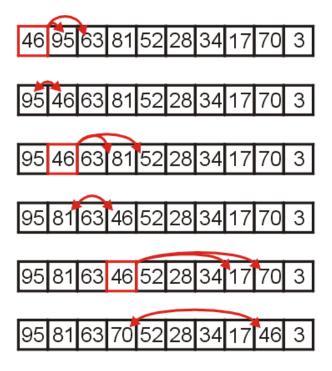
46 52 63 81 95 28 34 17

We compare 52 with its children, swap it with the largest

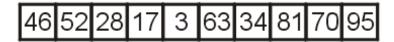
Recursing, no further swaps are needed



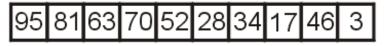
Finally, we swap the root with its largest child, and recurse, swapping 46 again with 81, and then again with 70

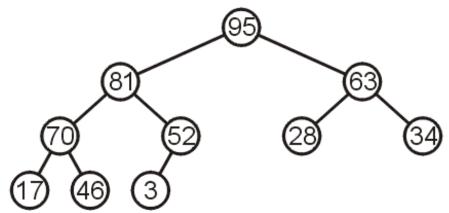


We have now converted the unsorted array



into a max-heap:

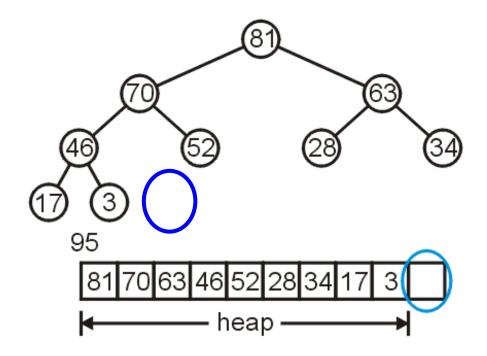




We pop the maximum element of this heap

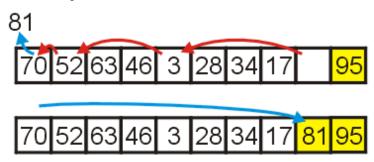


This leaves a gap at the back of the array:



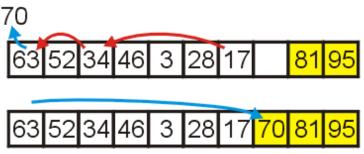
This is the last entry in the array, so why not fill it with the largest element?

Repeat this process: pop the maximum element, and then insert it at the end of the array:

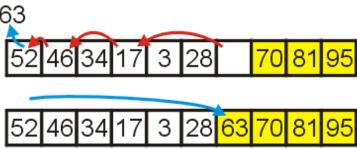


#### Repeat this process

Pop and append 70

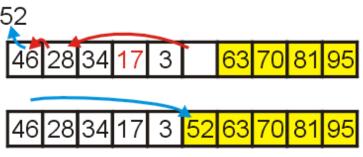


Pop and append 63

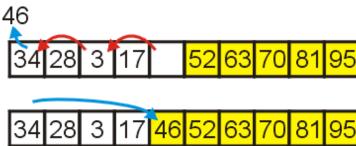


We have the 4 largest elements in order

Pop and append 52

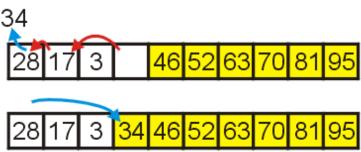


Pop and append 46

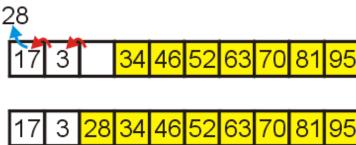


#### Continuing...

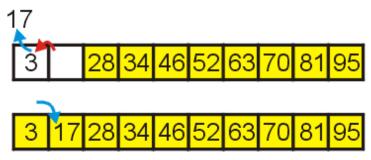
Pop and append 34



Pop and append 2828



Finally, we can pop 17, insert it into the 2<sup>nd</sup> location, and the resulting array is sorted



#### Summary

- Priority queue
  - pop the object with the highest priority
- Binary heap
  - Operations

<ul> <li>Top</li> </ul>	$\Theta(1)$
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- Push  $O(\ln(n))$
- Pop  $O(\ln(n))$
- Build O(n)
- Implementation using arrays
- Heapsort
  - Time:  $O(n \ln(n))$
  - Space: O(1)