

Heaps



6.006 Lecture 4

Image from www.telegraph.co.uk

Today's plan

- Priority Queues
- Heaps
 - Heapsort: Another $O(n \log n)$ sorting algorithm!

Priority Queue

An *abstract data structure* implementing a set S of *elements*, each associated with a *key*, supporting the following operations:

$\text{insert}(S, x)$: insert element $x = (\text{name}(x), \text{key}(x))$ into set S

$\text{max}(S)$: return element of S with largest key

$\text{extract_max}(S)$: return element of S with largest key and remove it from S

$\text{increase_key}(S, x, k)$: Change key-value of x to k
(assumed to be increase)

Tons of applications!

e.g. triage in hospital based on how bad things are going

Priority Queue

$\text{insert}(S, x)$: insert element $x = (\text{name}(x), \text{key}(x))$ into set S

$\text{max}(S)$: return element of S with largest key

$\text{extract_max}(S)$: return element of S with largest key and remove it from S

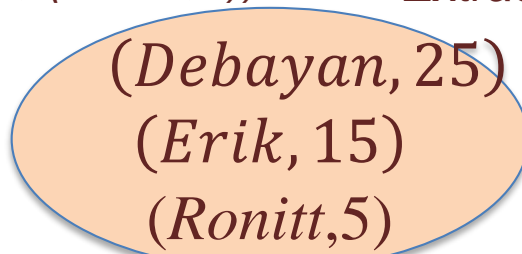
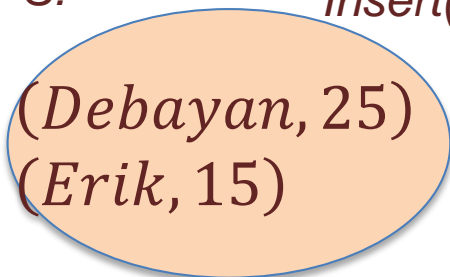
$\text{increase_key}(S, x, k)$: increase the value of element x 's key to new value k
(assumed to be as large as current value)

S :

$\text{Insert}(S, (\text{Ronitt}, 5))$:

$\text{Extract_max}(S)$:

$\text{Increase_key}(S, (\text{Ronitt}, 33))$:



How do we best implement it?

Priority Queue: First (slow) idea

Maintain an (unsorted) array?

$\text{insert}(S, x)$: insert x into S $O(1)$

$\text{max}(S)$: return largest element $O(n)$ to find

$\text{extract_max}(S)$: return largest element and remove
 $O(n)$ to find

$\text{increase_key}(S, x, k)$: increase the value of element x 's key to
new value k
(assumed to be as large as current value)

$O(n)$ to find

Priority Queue: Second (slow) idea

Maintain a **sorted array** (based on keys)?

$\text{insert}(S, x)$: insert x into S $O(n)$ to shift others right

$\text{max}(S)$: return largest element $O(1)$ to find

$\text{extract_max}(S)$: return largest element and remove

$O(1)$ to find, then $O(n)$ to shift others left

$\text{increase_key}(S, x, k)$: increase the value of element x 's key to new value k
(assumed to be as large as current value)

$O(n)$ to find, then $O(n)$ to shift left

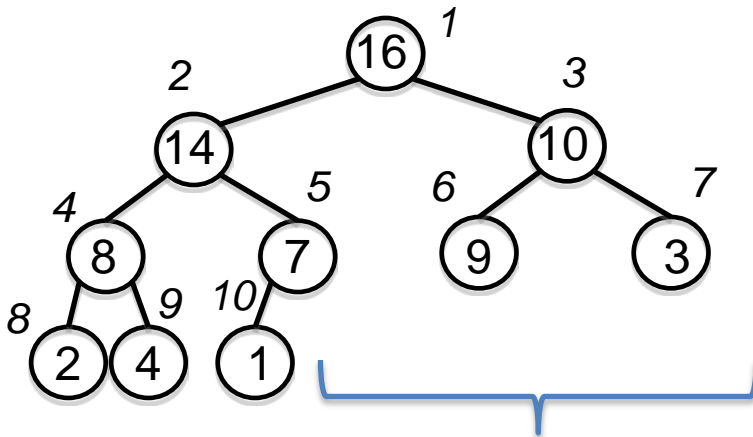
Let's apply our 6.006 **data**
structure superpowers!



New data structure

(Max) Heap

- A nearly complete **binary tree**
- **Max Heap Property (MHP)**: key of a node \geq keys of its children (**Min Heap** defined analogously)



Missing only rightmost nodes on bottom level

Important fact:
Height of the tree is
always $O(\log n)$

Heap Operations

insert:

max:

extract_max:

increase_key:

build_max_heap : produce a max-heap from an unordered array

max_heapify : correct a **single** violation of the heap property at the root (rest of heap is fine)

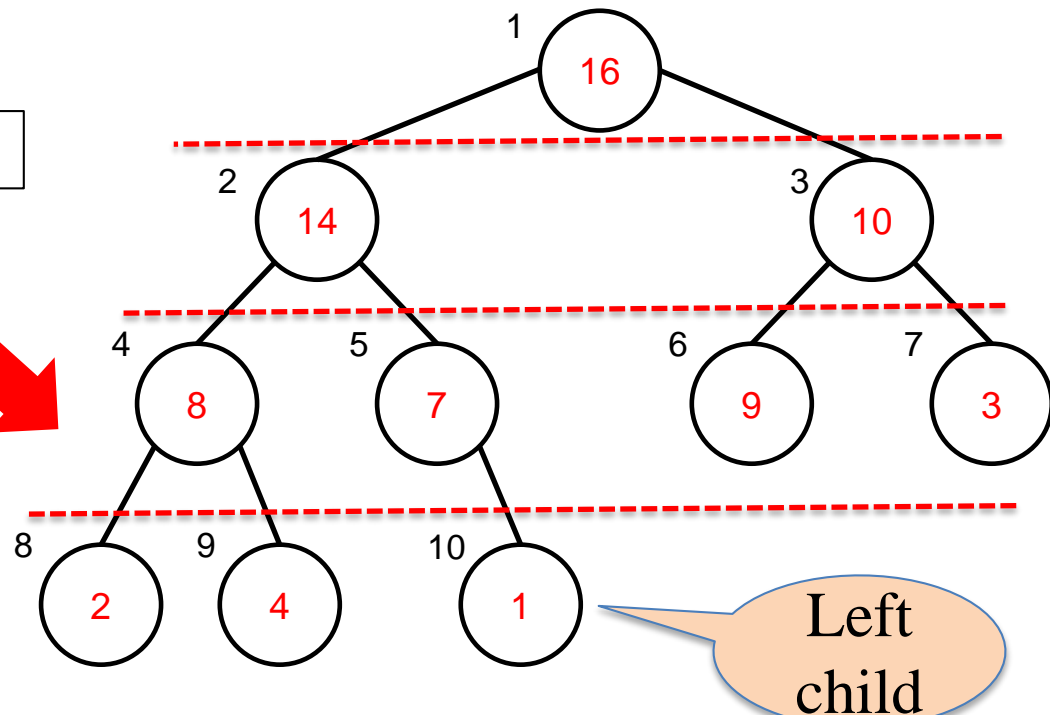
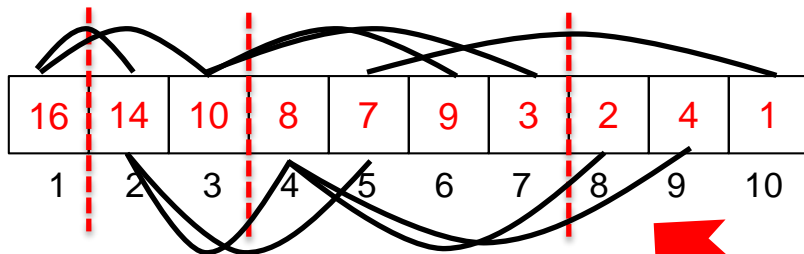
Heap operations can be used to
implement priority queue!

Representation via *array*

- root: first element in the array (**$i=1$**)
- parent(i): **$\text{floor}(i/2)$** returns index of node's parent
- left(i): **$2i$** returns index of node's *left* child
- right(i): **$2i+1$** returns index of node's *right* child

(Why? No pointers needed!)

Important detail: index elements starting from **$i=1$**

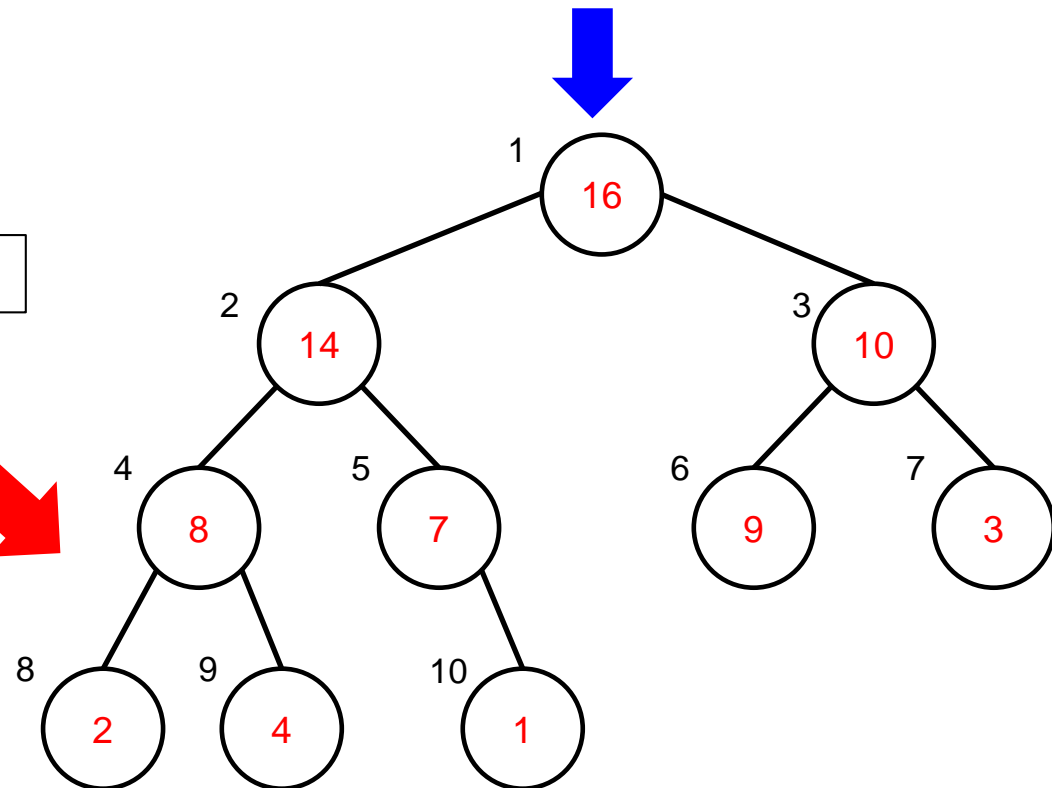
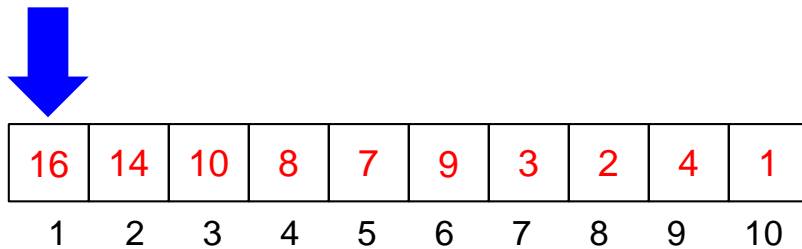


Why Heaps?

Key consequence of Max Heap Property:

Root/first element is always the max \rightarrow can do $\max(S)$ in $\Theta(1)$ time!
(**Note:** the array is not sorted though!)

But: How to maintain the Max Heap Property property after `insert/extract_max/increase_key`?



In fact:

How to build a Max Heap to begin with?

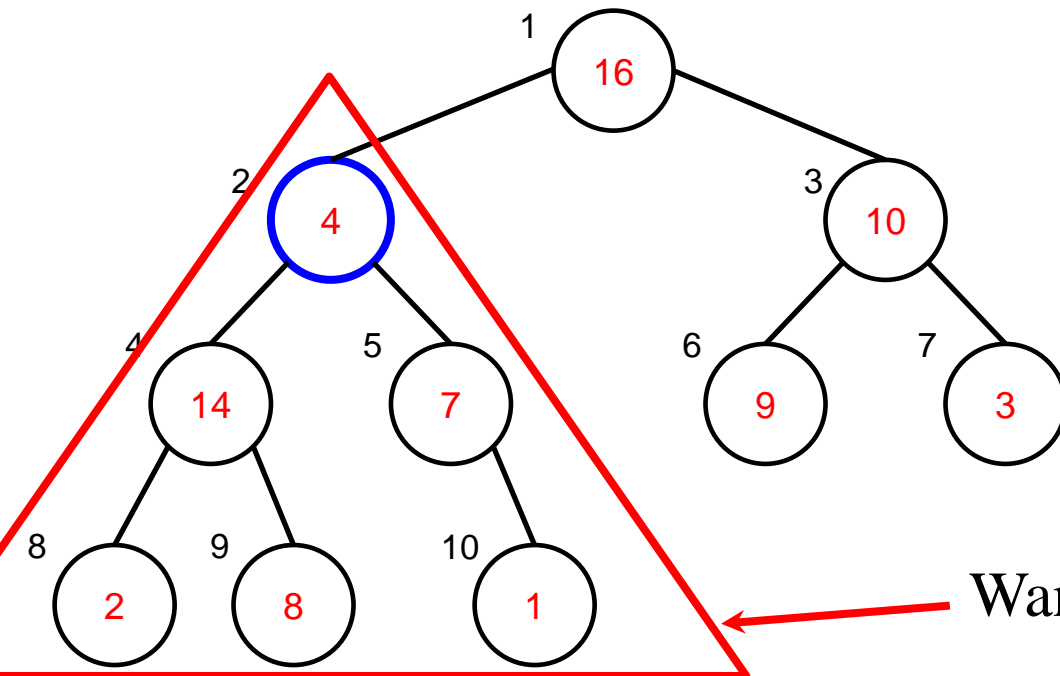
Key Primitive

`max_heapify(A[i]):` Corrects a **single** violation of Max Heap Property in a subtree rooted at **i only**

How to implement it?

➔ Assume that the trees rooted at `left(i)` and `right(i)` are Max Heaps

➔ If element `A[i]` violates the MHP, correct violation by “trickling” this element down the tree, making the subtree rooted at `i` a Max Heap (**Important:** Always swap with the larger of two children. **Why?**)



Want this to be a Max Heap

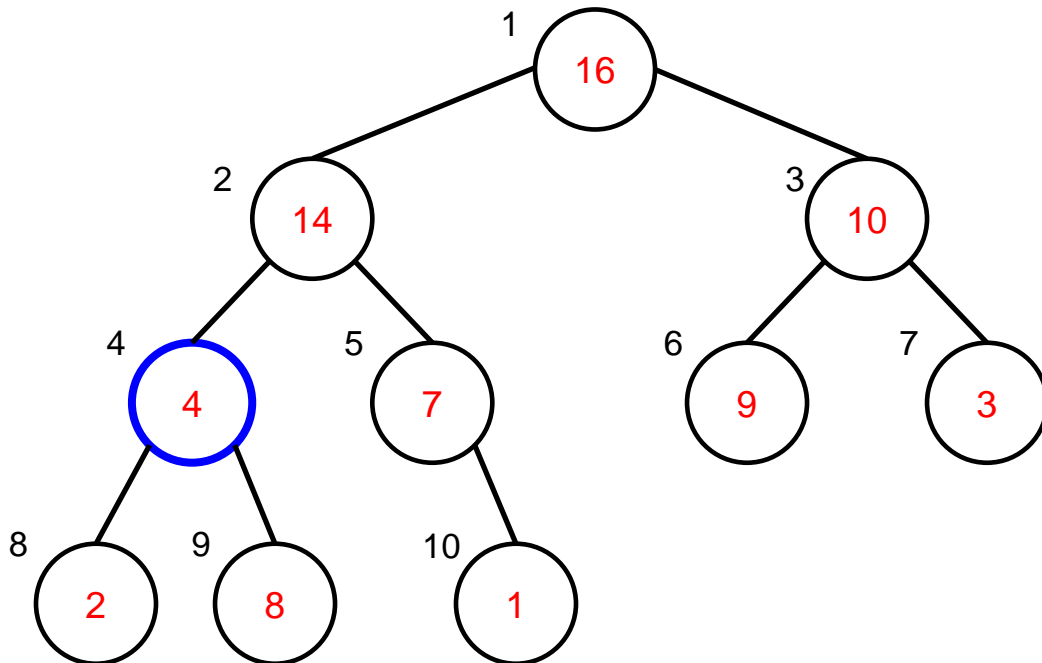
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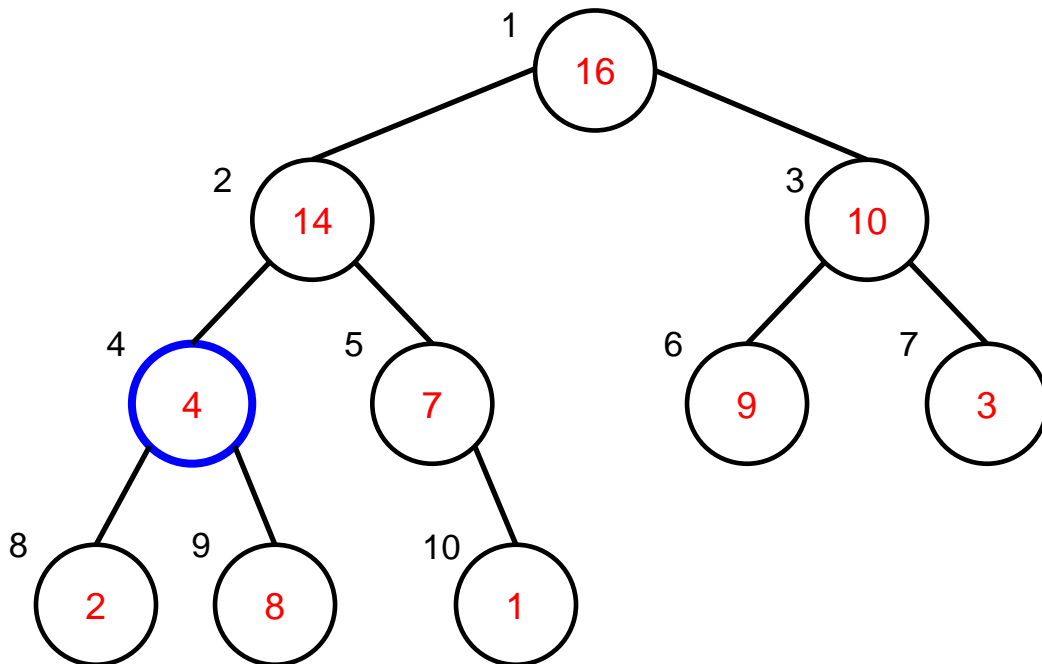
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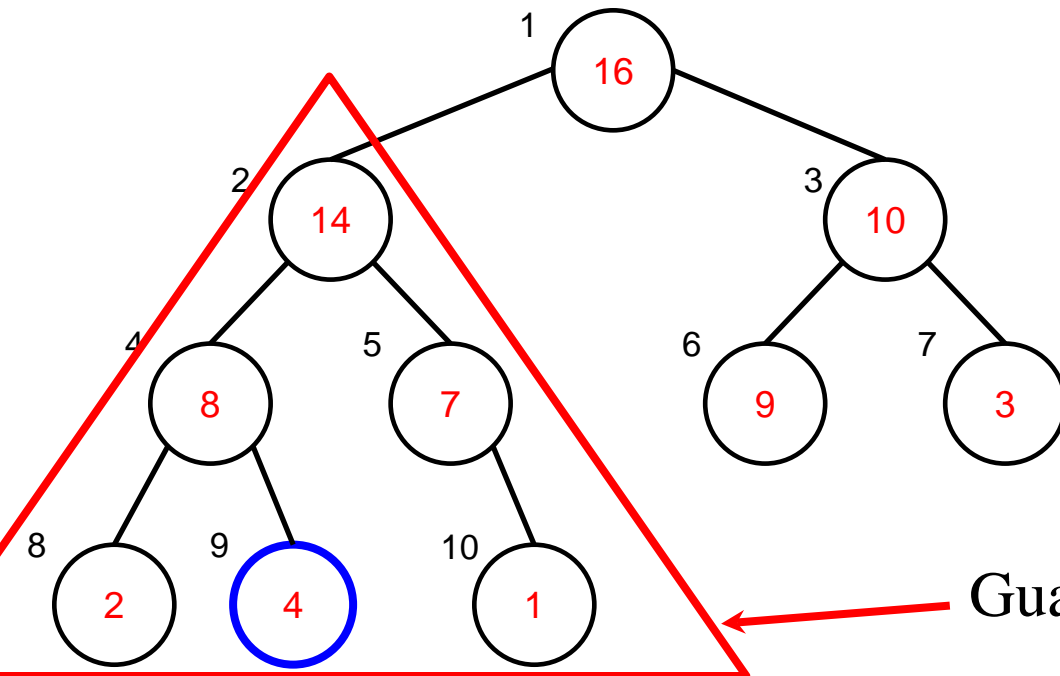
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Run time?

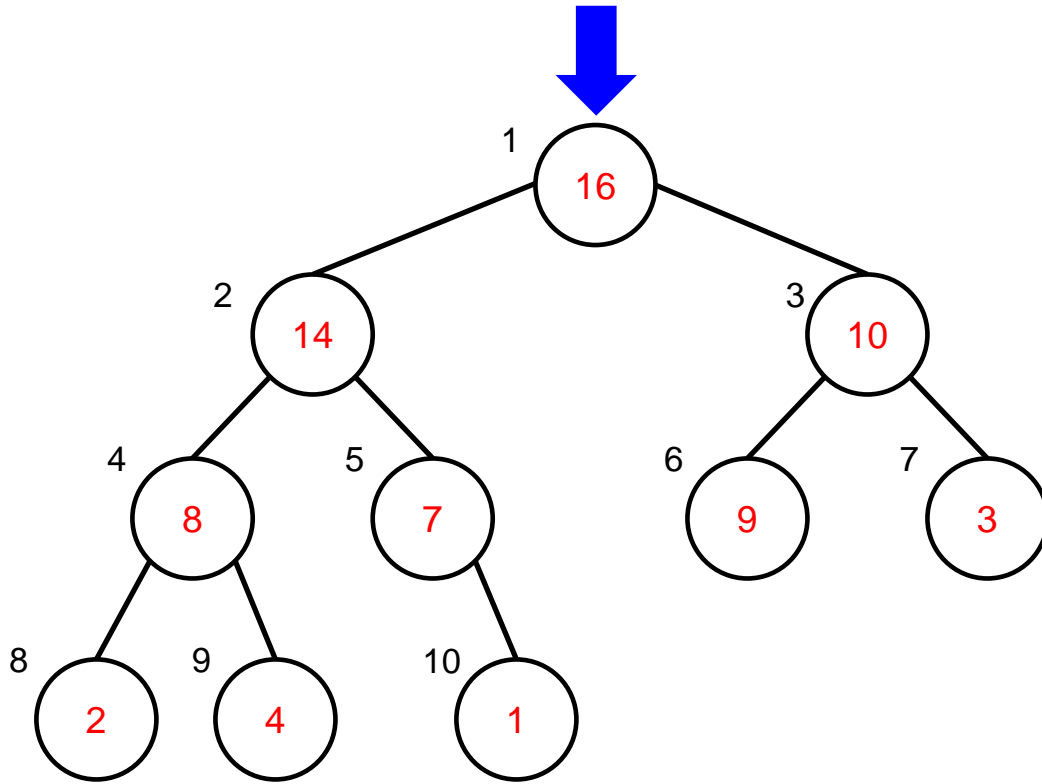
➔ $\Theta(1)$ at each level

➔ total: $O(\text{subtree height})$
 $= \Theta(\log n)$ (worst case)

(**Recall:** tree is balanced)

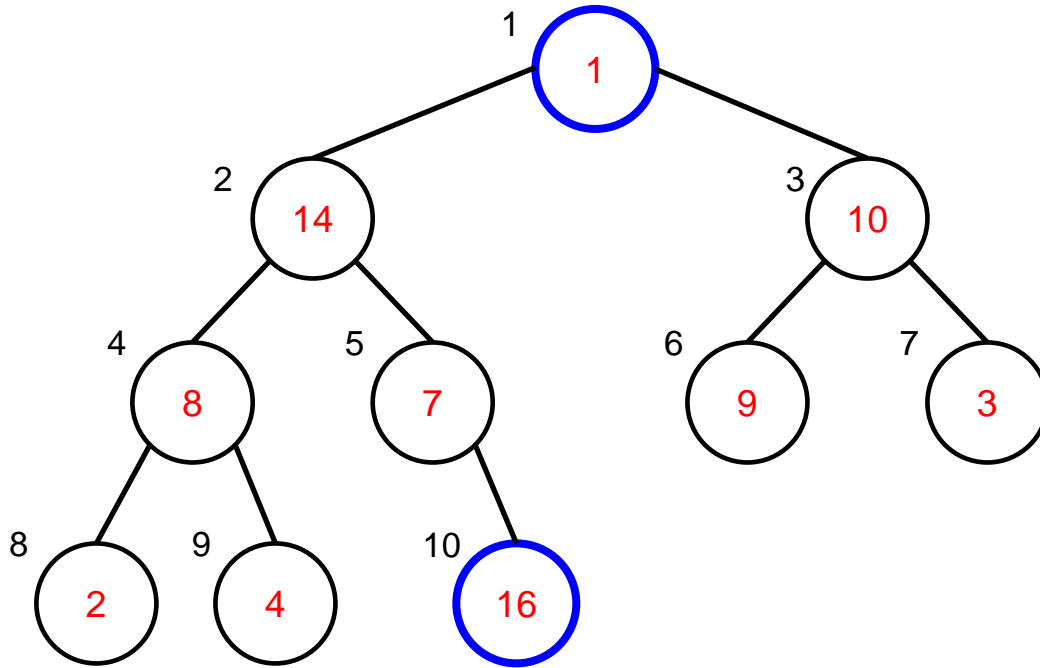
Guaranteed to be a Max Heap now

Implementing Extract_Max

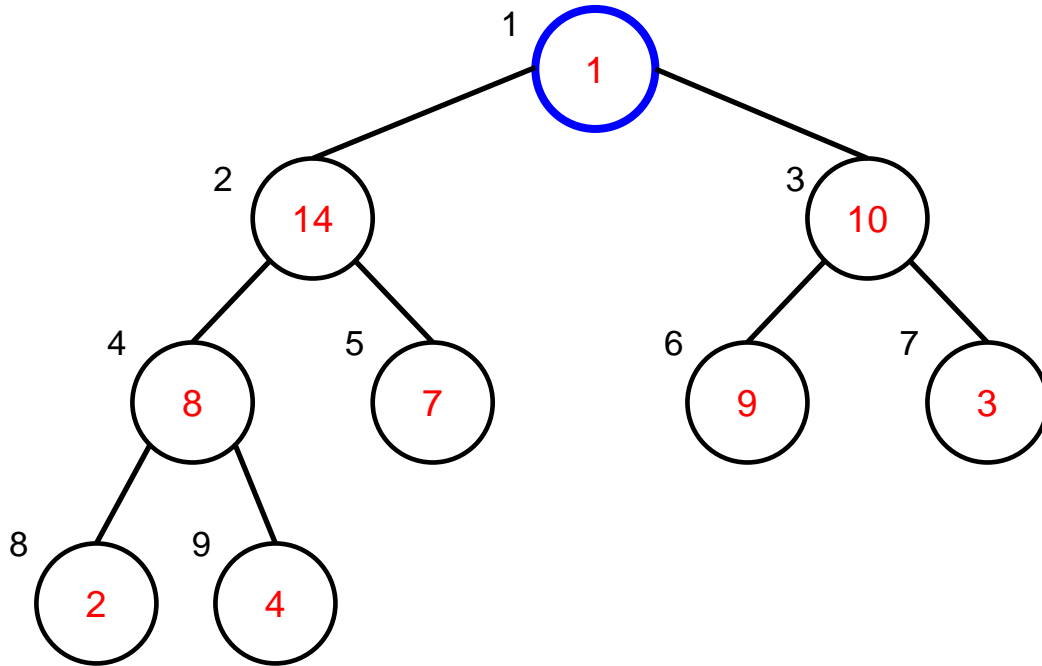


Implementing Extract_Max

- Swap the root with the last element of the heap

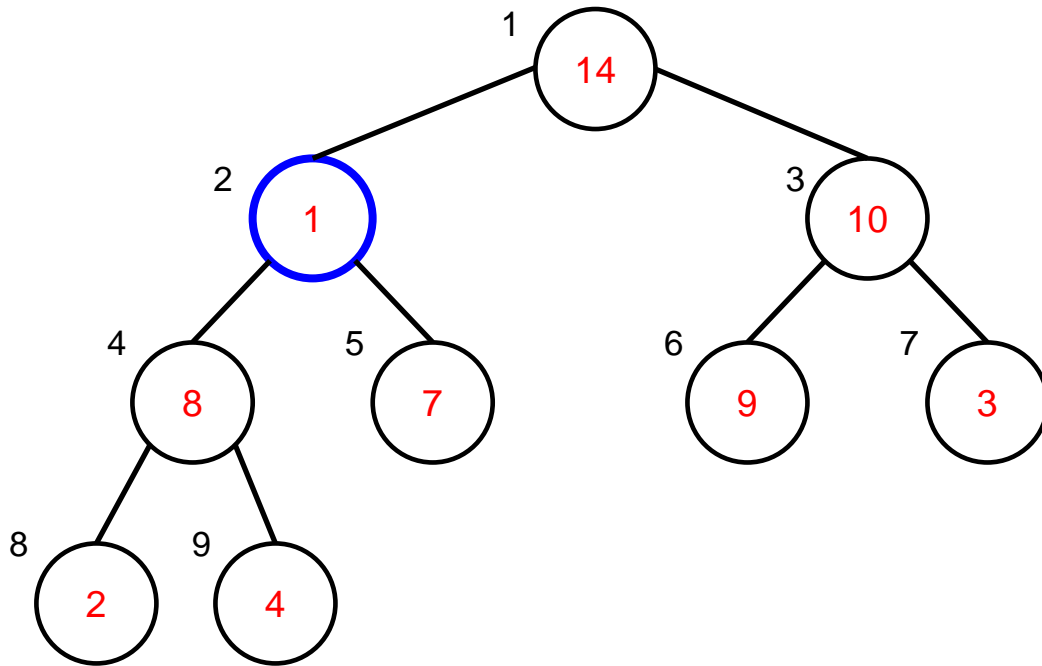


Implementing Extract_Max



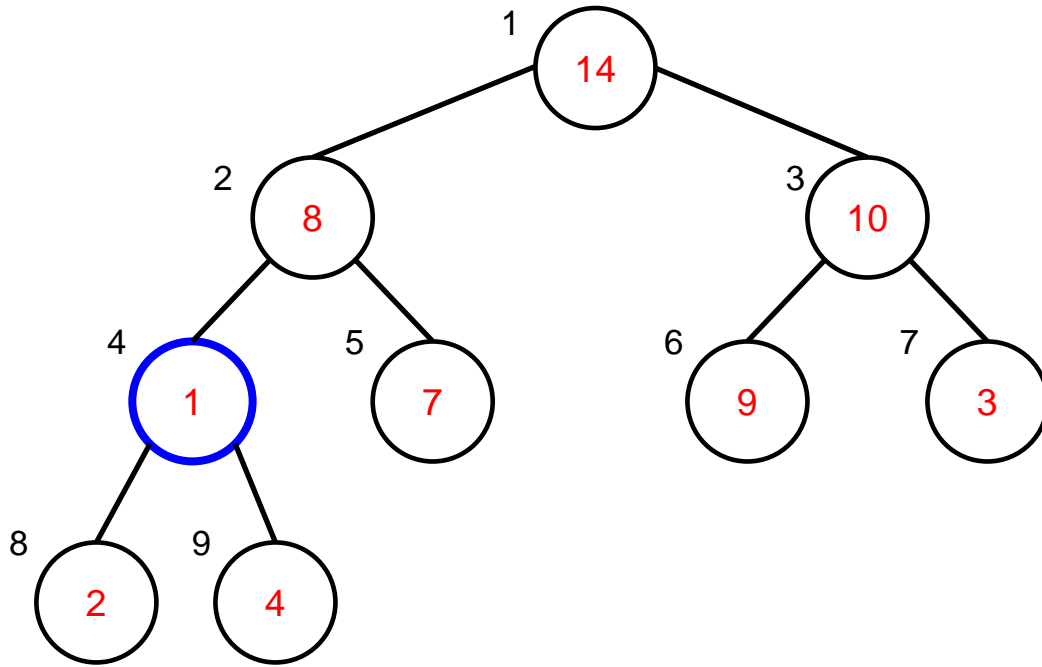
- Swap the root with the last element of the heap
- Now we can remove it from the heap (decrease the heap size by one)
- **To fix MHP:**
Max_heapify the root

Implementing Extract_Max



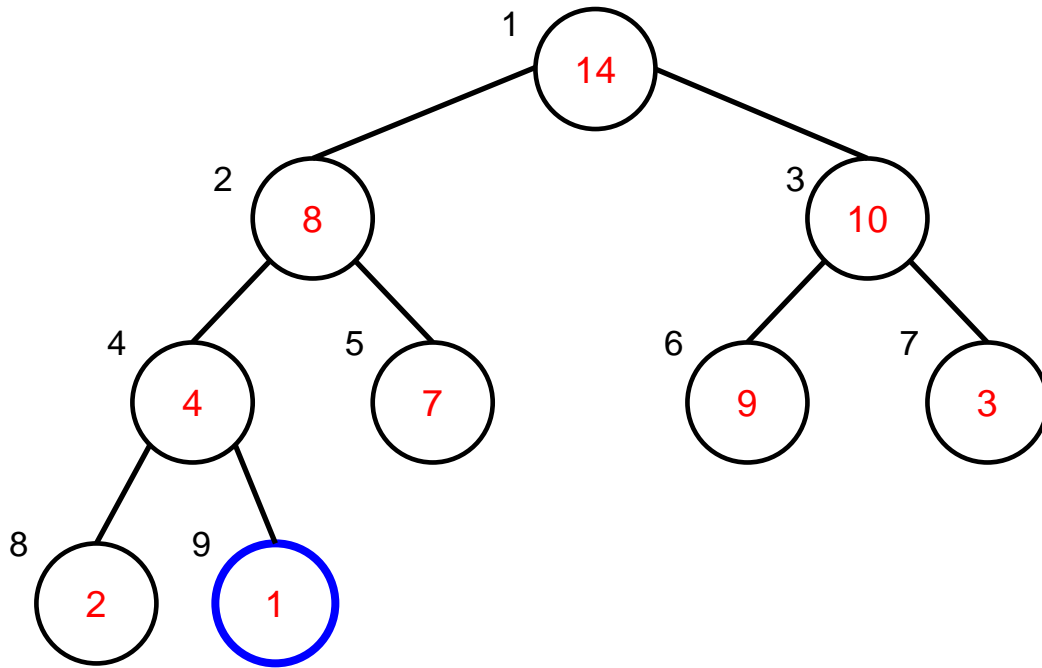
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Implementing Extract_Max



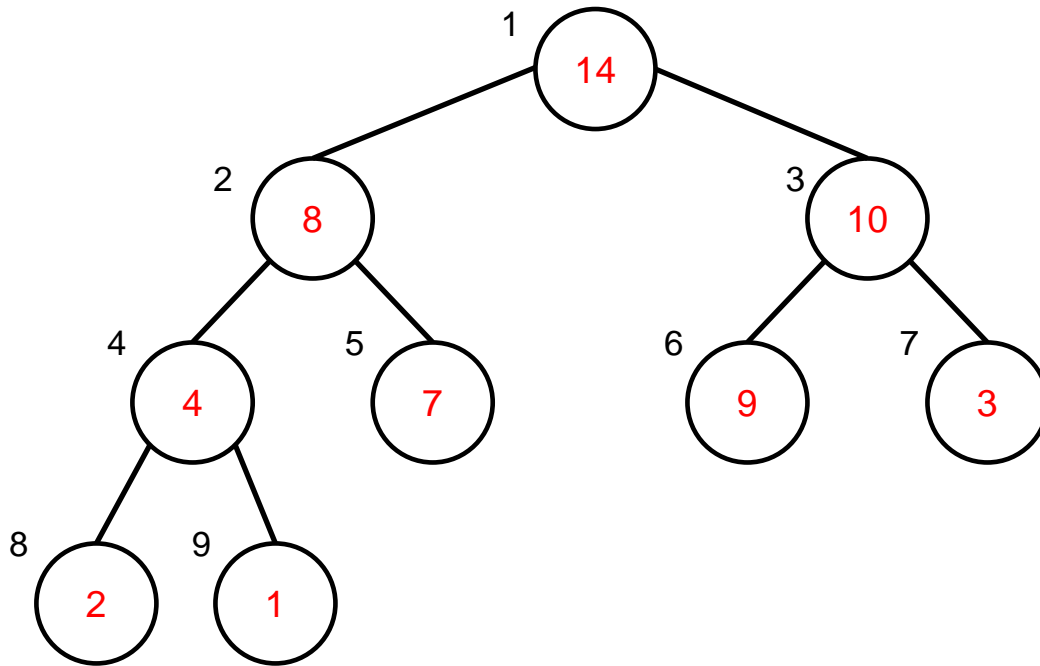
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Implementing Extract_Max



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Implementing Extract_Max



- Swap the root with the last element of the heap
- Now we can remove it from the heap (decrease the heap size by one)
- **To fix MHP:**
Max_heapify the root
- Done!

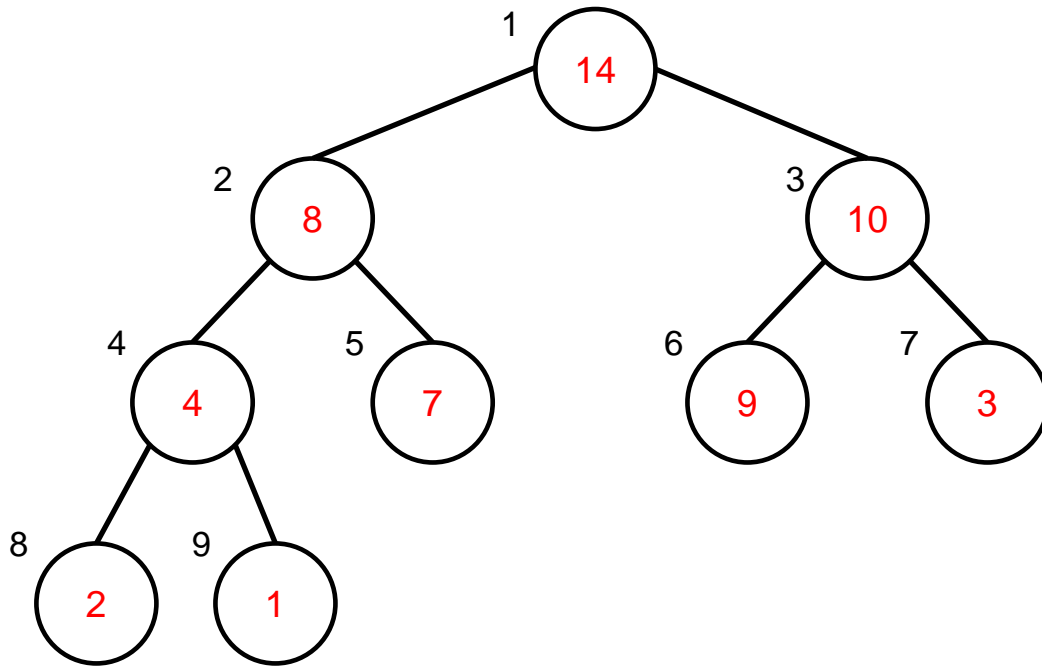
Run time?

➔ **$\Theta(1)$** (swapping) + **$\Theta(1)$** (removal) + **$O(\log n)$** (max_heapify)

➔ total: **$\Theta(\log n)$** (worst case)

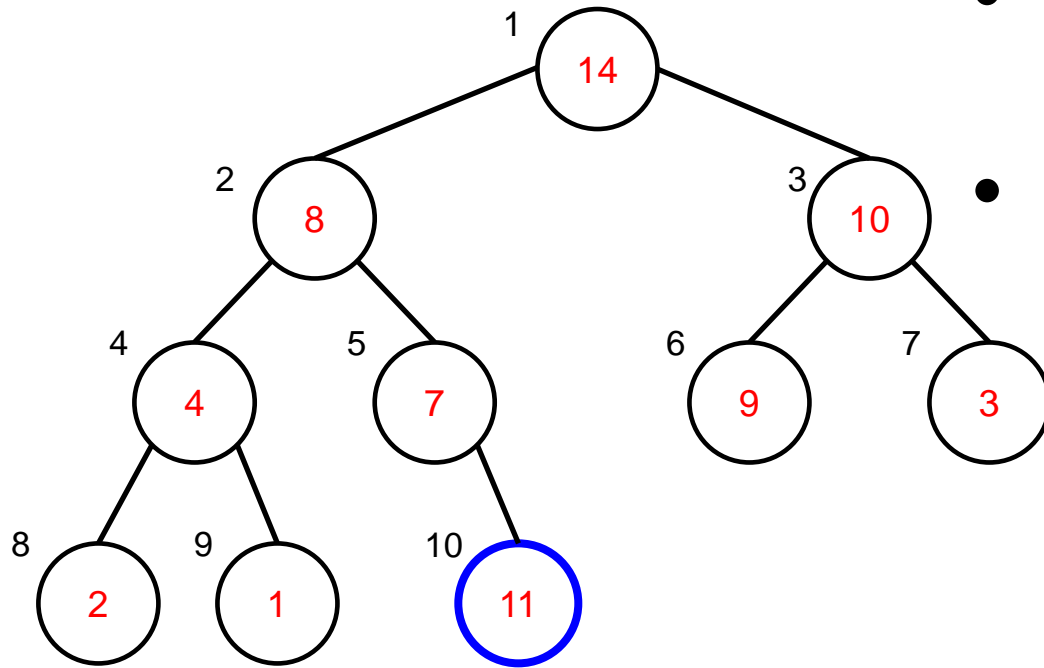
Implementing Insert

(in a sense: “reversing” the extract_max)



Implementing Insert

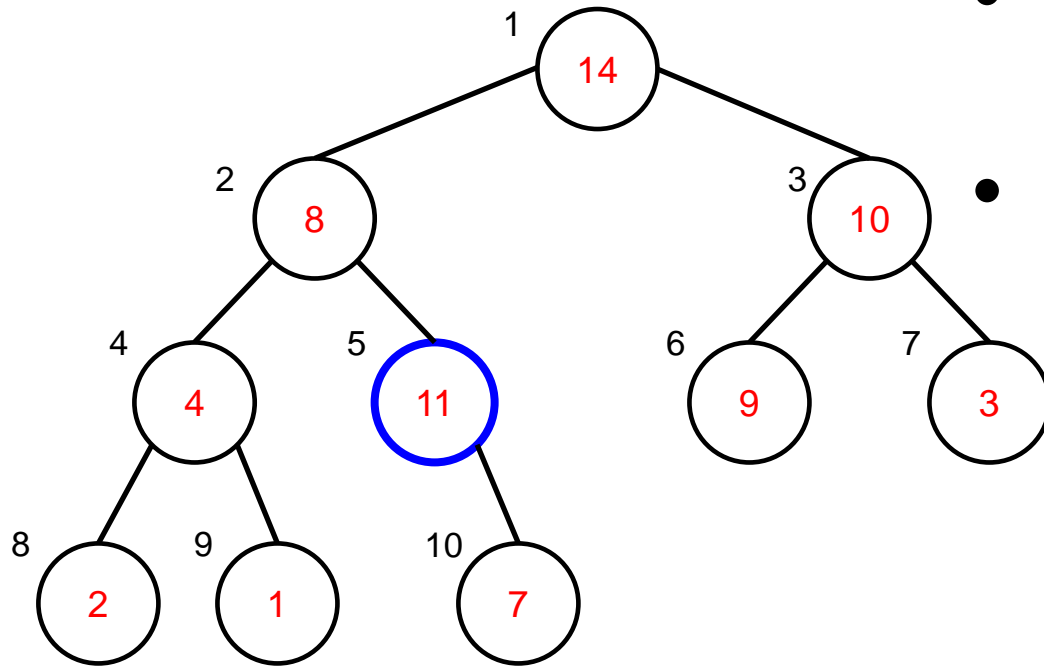
(in a sense: “reversing” the extract_max)



- Add the new element as the last one
- **To fix MHP:**
“Promote” the new element up the tree
 (“reversed” max_heapify)

Implementing Insert

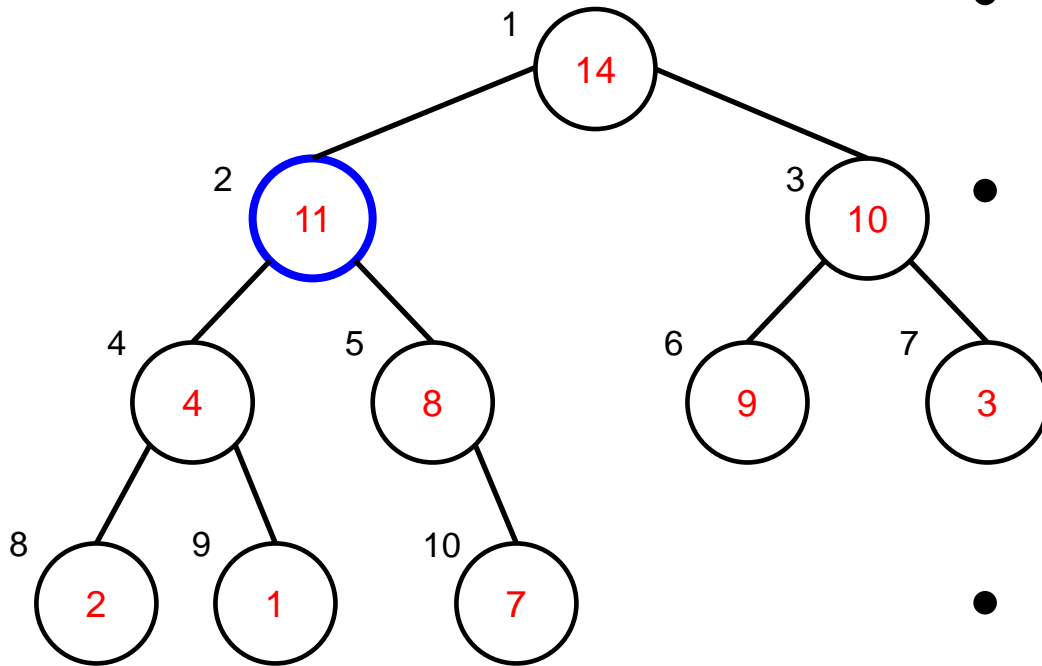
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Implementing Insert

(in a sense: “reversing” the extract_max)

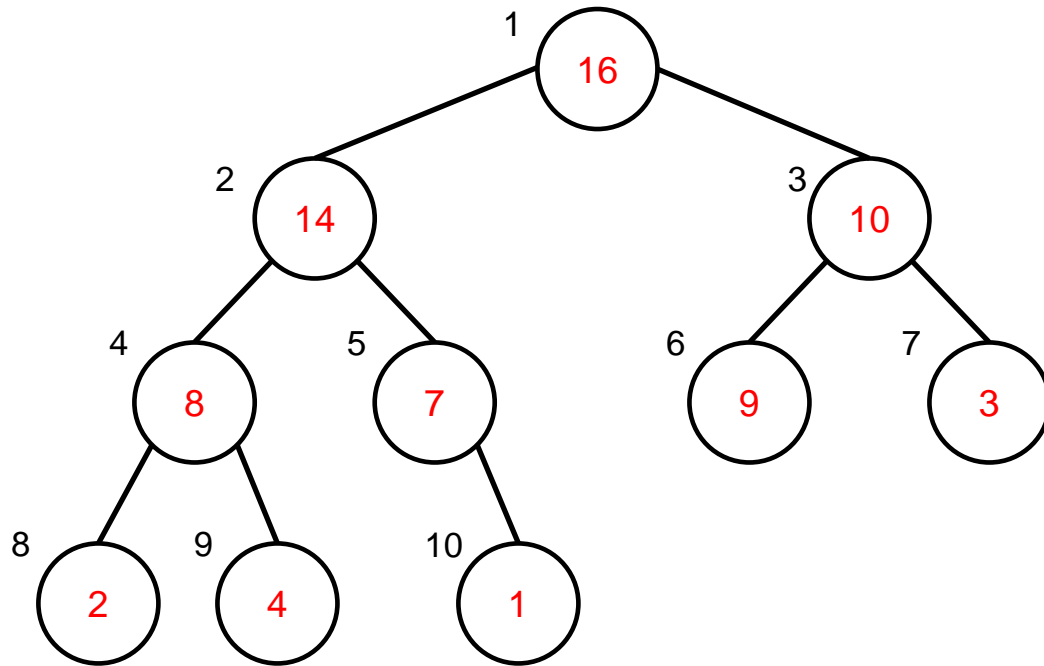


- Add the new element as the last one
- **To fix MHP:**
“Promote” the new element up the tree
 (“reversed” max_heapify)
- Done!

Run time?

- ➔ $\Theta(1)$ (addition) + $O(\log n)$ (promotion up the tree)
- ➔ total: $\Theta(\log n)$ (worst case)

Implementing Increase_key (Similar to Insert)



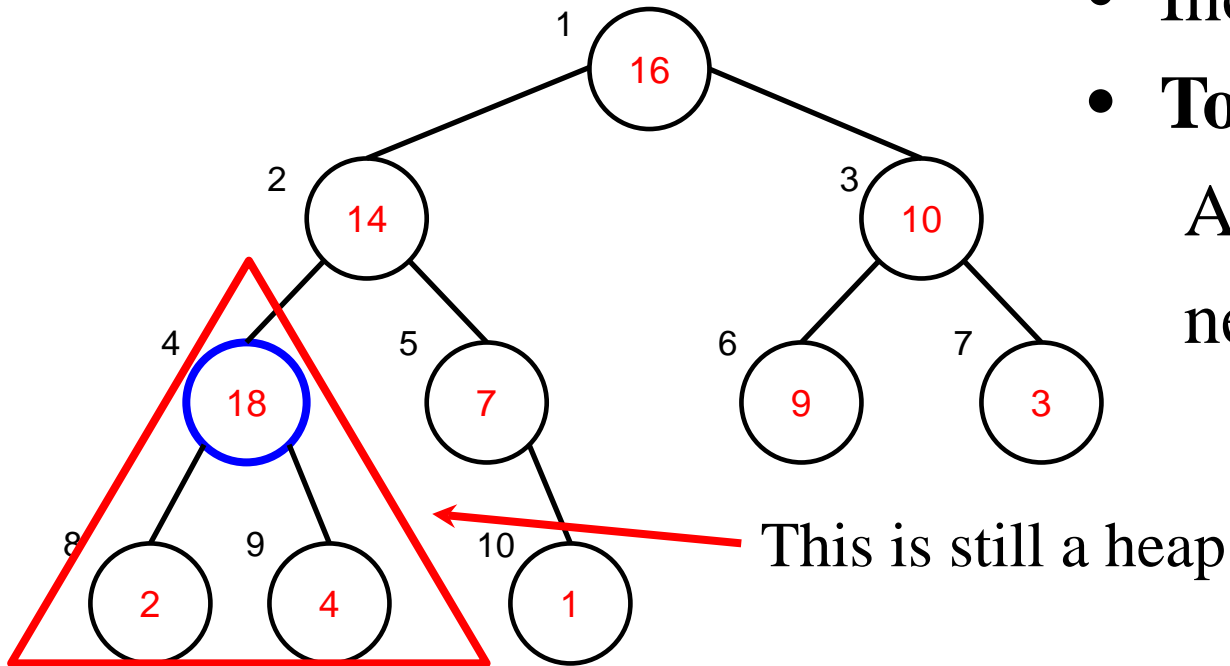
Implementing Increase_key

(Similar to Insert)

- Increase the key value

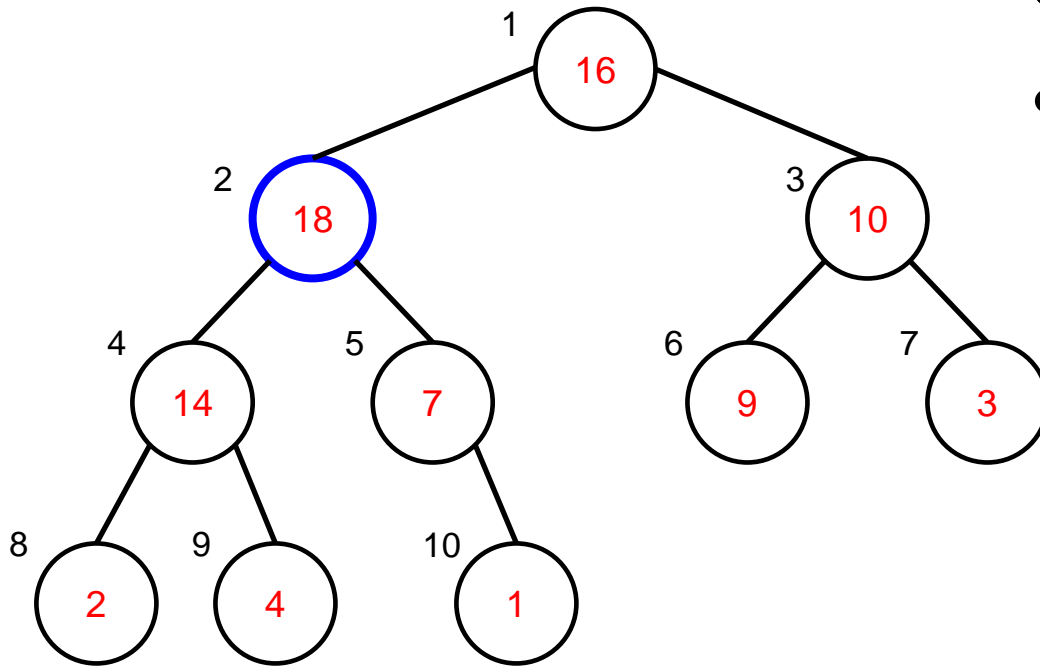
- **To fix MHP:**

Again, “promote” the new element up the tree



Implementing Increase_key

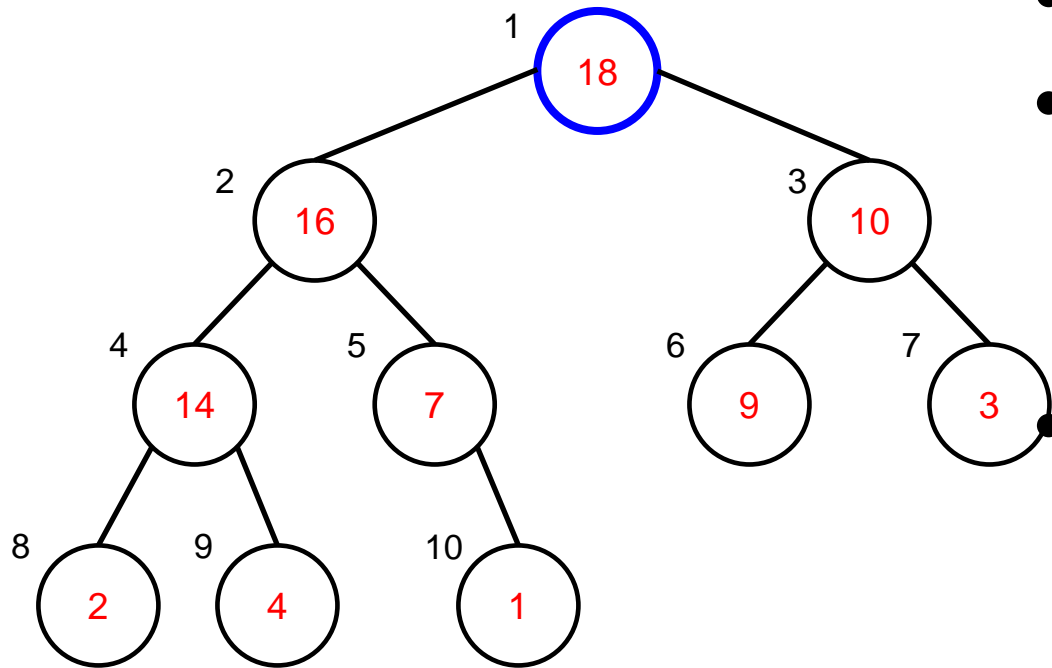
(Similar to Insert)



- Increase the key value
- **To fix MHP:**
Again, “promote” the new element up the tree

Implementing Increase_key

(Similar to Insert)



- Increase the key value
- **To fix MHP:**
Again, “promote” the new element up the tree
Done!

Run time?

- ➔ $\Theta(1)$ (key value increase) + $O(\log n)$ (promotion up the tree)
- ➔ total: $\Theta(\log n)$ (worst case)

How to build a heap from a scratch?

Simple way:

- Start with an empty heap
- Insert all the n elements into it
- Total time: $\Theta(n \log n)$ (worst-case)

Iterative (and in-place) way:

build_max_heap(A):

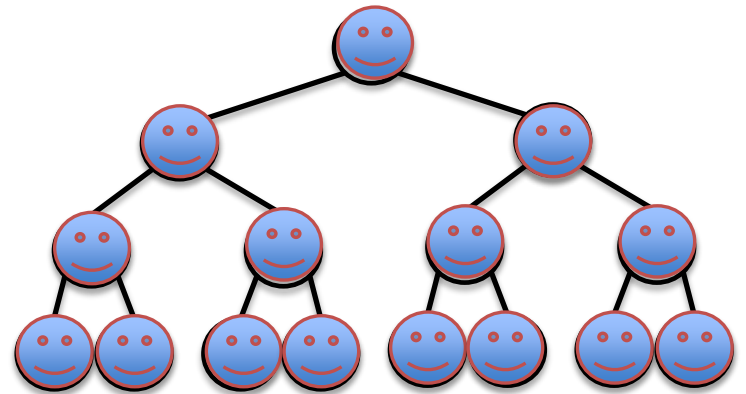
 for $i=n$ downto 1

 do max_heapify(A,i)

$O(n \log n)$

→ Total time? At first glance: ~~$\Theta(n \log n)$~~ (cost of n max_heapify)

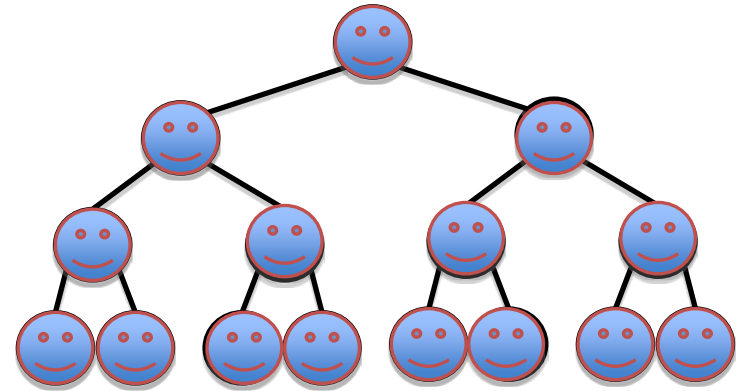
Actually: $\Theta(n)$



Build_Max_Heap Analysis

Converts $A[1 \dots n]$ to a max heap

```
Build_Max_Heap(A):  
  for  $i = n/2$  downto 1  
    do Max_Heapify(A, i)
```



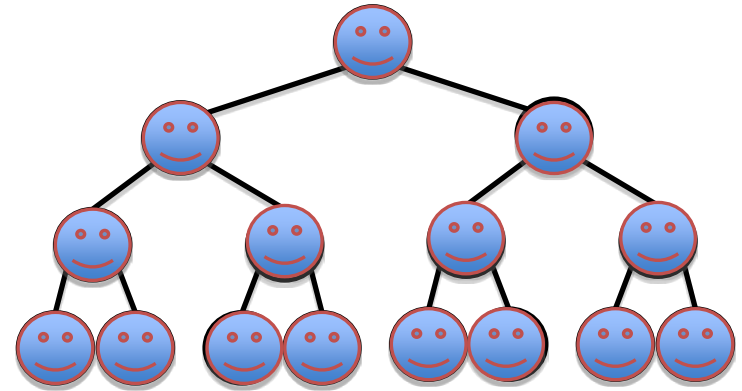
$O(l)$ time for nodes that are l levels above leaves.

We have $n/4$ nodes with level 1, $n/8$ with level 2,...

Build_Max_Heap Analysis

Converts $A[1 \dots n]$ to a max heap

```
Build_Max_Heap(A):  
  for i=n/2 downto 1  
    do Max_Heapify(A, i)
```



Total amount of work in the for loop can be summed as:

$$n/4 (1 c) + n/8 (2 c) + n/16 (3 c) + \dots + 1 (\lg n c)$$

Setting $n/4 = 2^k$ and simplifying we get:

$$c 2^k (1/2^0 + 2/2^1 + 3/2^2 + \dots (k+1)/2^k)$$

The term in brackets is bounded by a constant!

This means that Build_Max_Heap is $O(n)$

Cool application: Sorting

Heapsort: Sorting using a heap/priority queue

- ➔ Build a heap out of all elements
- ➔ Extract_max all elements one-by-one in an (inversely) sorted order!
- ➔ Total time: $\Theta(n \log n)$ (worst-case)

This is a different algorithm than Merge sort!

In particular: Heapsort is actually an in-place algorithm (once we unravel the implementation of the heap)

More applications of heaps:

