

# Heap

## A priority queue data structure

### Sources:

David Matuszek, University Of Pennsylvania: [www.cis.upenn.edu/~matuszek/](http://www.cis.upenn.edu/~matuszek/)

Dr. George Bebis, University of Nevada: Course page: [www.cse.unr.edu/~bebis](http://www.cse.unr.edu/~bebis)

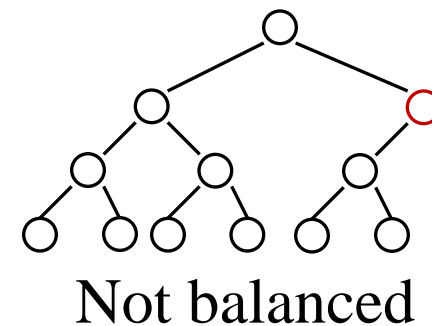
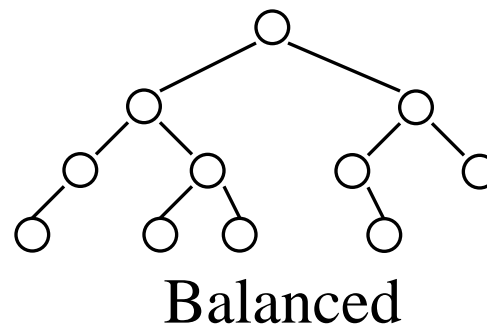
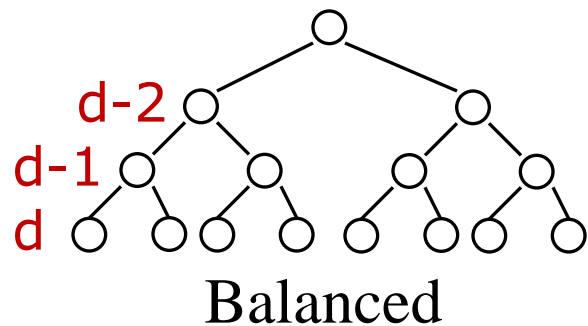
Slides and figures have been collected from various publicly available Internet sources for preparing the lecture slides of IT2001 course. I acknowledge and thank all the original authors for their contribution to prepare the content.

# What is a “Heap”?

- Definitions of **heap**:
  1. A large area of memory from which the programmer can allocate blocks as needed, and deallocate them (or allow them to be garbage collected) when no longer needed
  2. A balanced, left-justified binary tree in which no node has a value greater than the value in its parent
- These two definitions have little in common
- Heap data structure uses the **second definition**

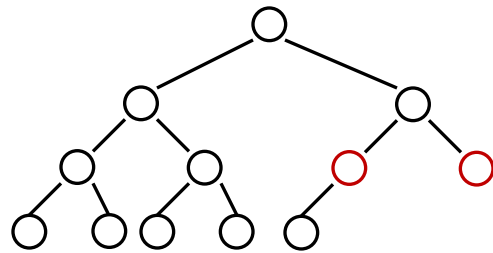
# Balanced binary trees

- Recall:
  - The **depth of a node** is its distance from the root
  - The **depth of a tree** is the depth of the deepest node
- A binary tree of depth  **$d$**  is **balanced** if all the nodes at depths  **$0$**  through  **$d-2$**  have two children

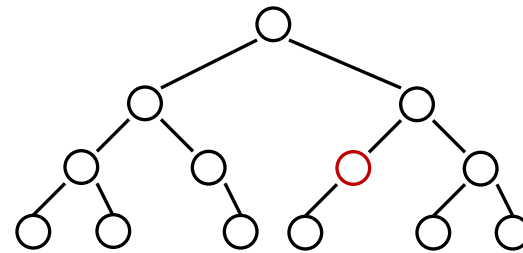


# Left-justified binary trees

- A balanced binary tree is **left-justified** if:
  - all the leaves are at the same depth, or
  - all the leaves at depth **d** are to the left of all the nodes at depth **d-1**



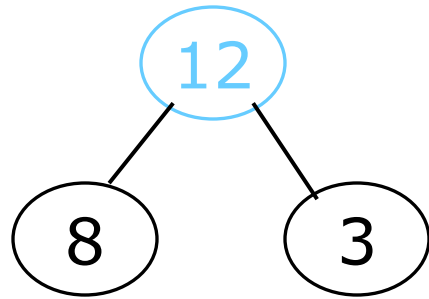
## Left-justified



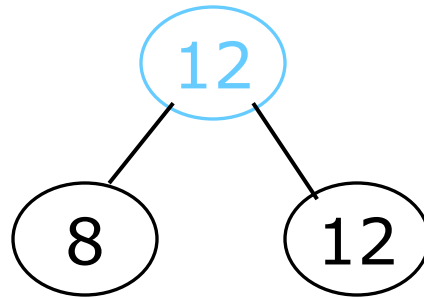
Not left-justified

# The heap property

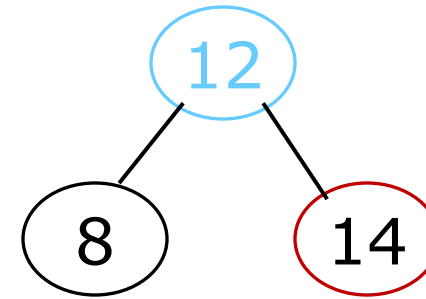
- A node has the **heap property** if the value in the node is as **large as or larger** than the values in its children



Blue node has  
heap property



Blue node has  
heap property



Blue node does not  
have heap property

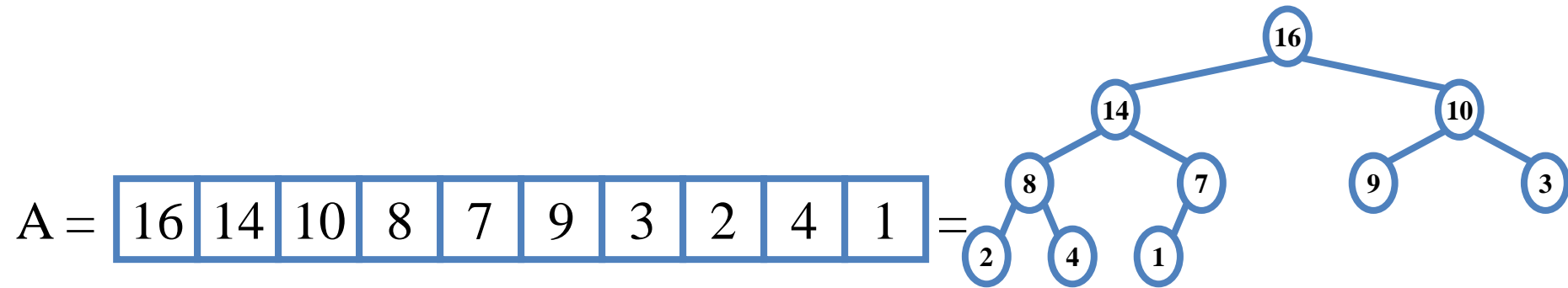
- All leaf nodes automatically have the heap property
- A binary tree is a **heap** if *all* nodes in it have the heap property

# Heap Types

- **Max-heaps** (largest element at root), have the *max-heap property*:
    - for all nodes  $i$ , excluding the root:
$$A[\text{PARENT}(i)] \geq A[i]$$
  - **Min-heaps** (smallest element at root), have the *min-heap property*:
    - for all nodes  $i$ , excluding the root:
$$A[\text{PARENT}(i)] \leq A[i]$$
- We will consider Max-heaps only.

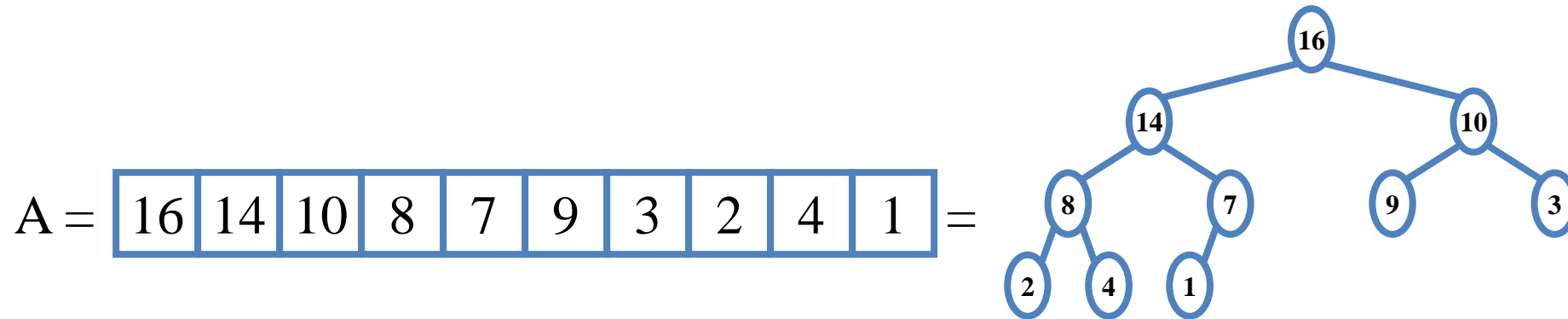
# Heaps

- In practice, heaps are usually implemented as arrays:



# Heaps

- To represent a complete binary tree as an array:
  - The root node is  $A[1]$
  - Node  $i$  is  $A[i]$
  - The parent of node  $i$  is  $A[i/2]$  (note: integer divide)
  - The left child of node  $i$  is  $A[2i]$
  - The right child of node  $i$  is  $A[2i + 1]$





# Referencing Heap Elements

- So...

```
Parent(i) { return  $\lfloor i/2 \rfloor$ ; }
```

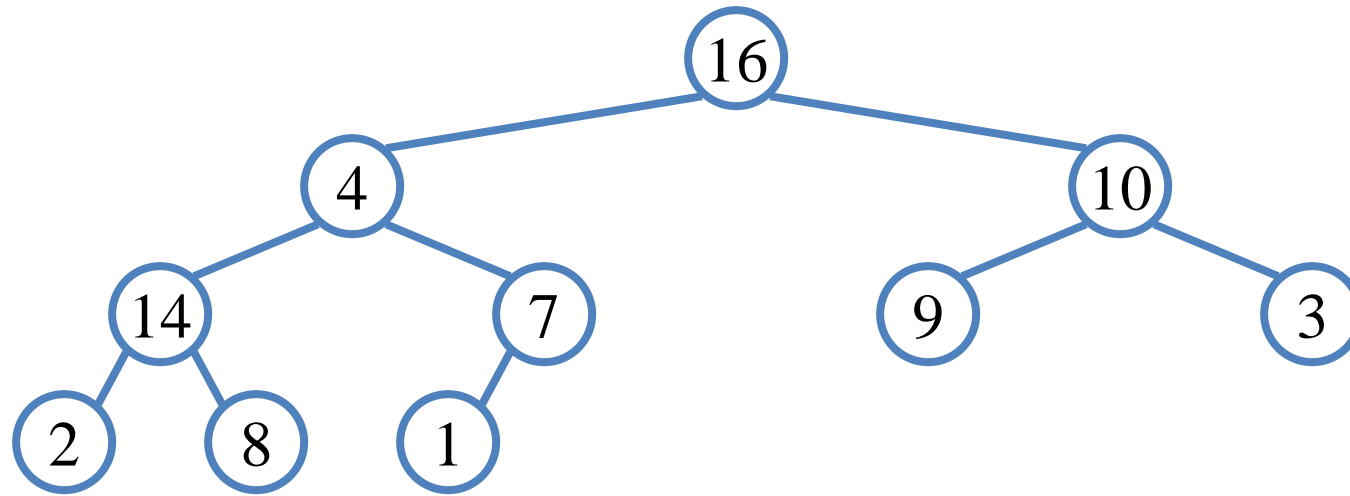
```
Left(i) { return 2*i; }
```

```
right(i) { return 2*i + 1; }
```

# Heap Operations: Heapify()

- **Heapify()** : maintain the heap property
  - Given: a node  $i$  in the heap with children  $l$  and  $r$
  - Given: two subtrees rooted at  $l$  and  $r$ , assumed to be heaps
  - Problem: The subtree rooted at  $i$  may violate the heap property (*How?*)
  - Action: let the value of the parent node “float down” so subtree at  $i$  satisfies the heap property
    - *What do you suppose will be the basic operation between  $i$ ,  $l$ , and  $r$ ?*

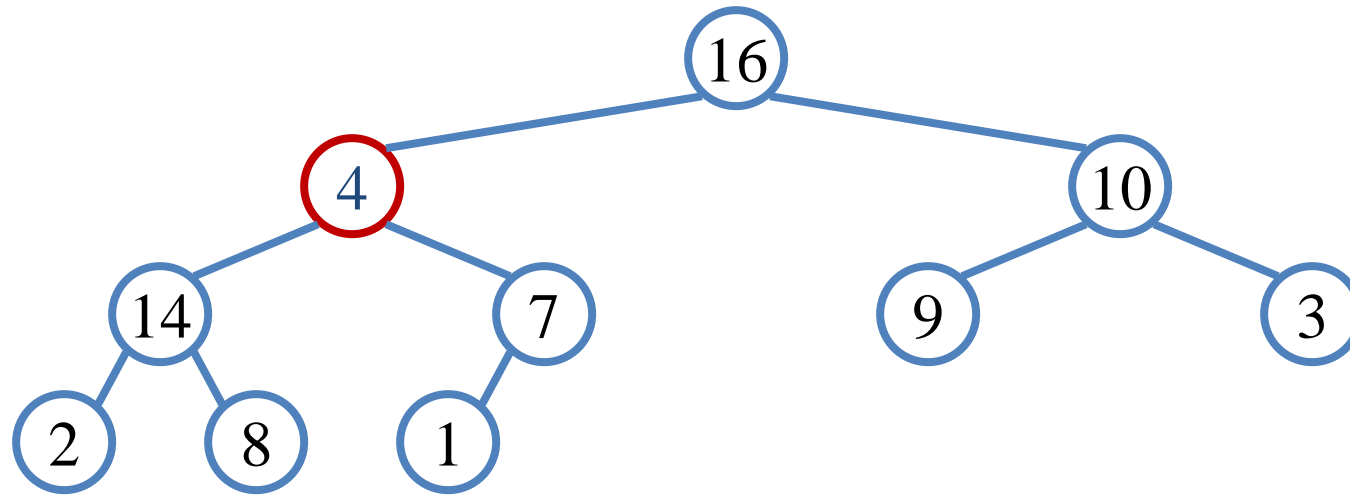
# Heapify() Example



A = 

16	4	10	14	7	9	3	2	8	1
----	---	----	----	---	---	---	---	---	---

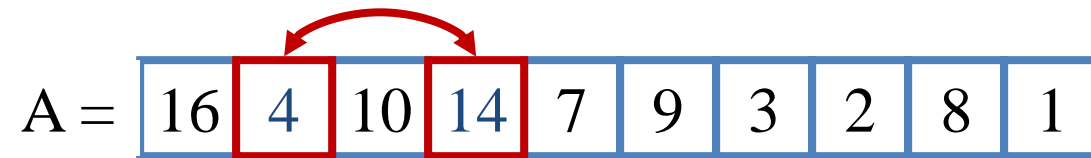
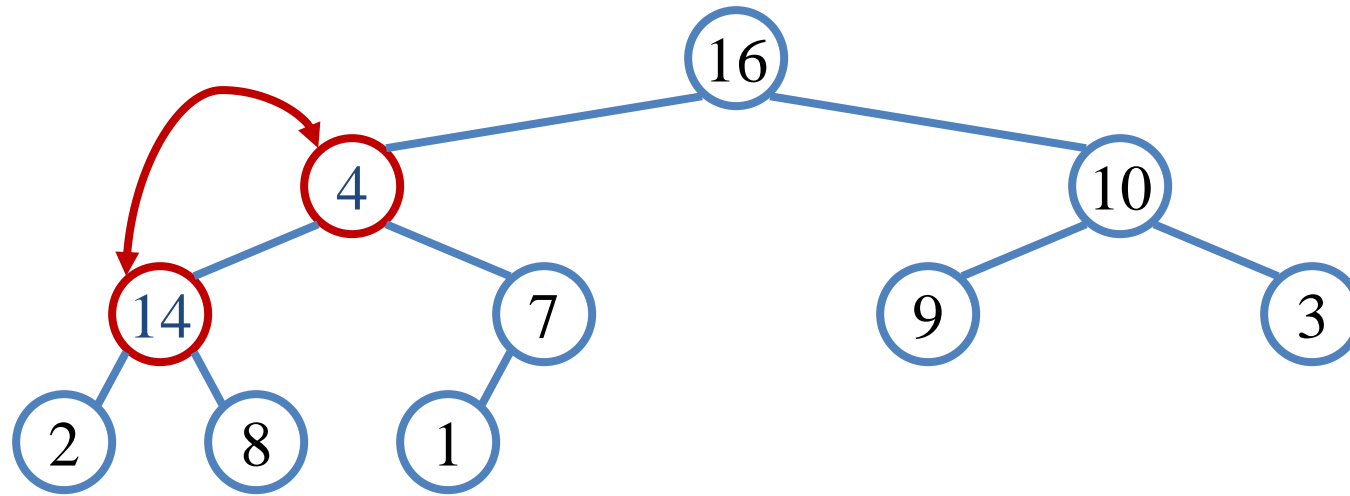
# Heapify() Example



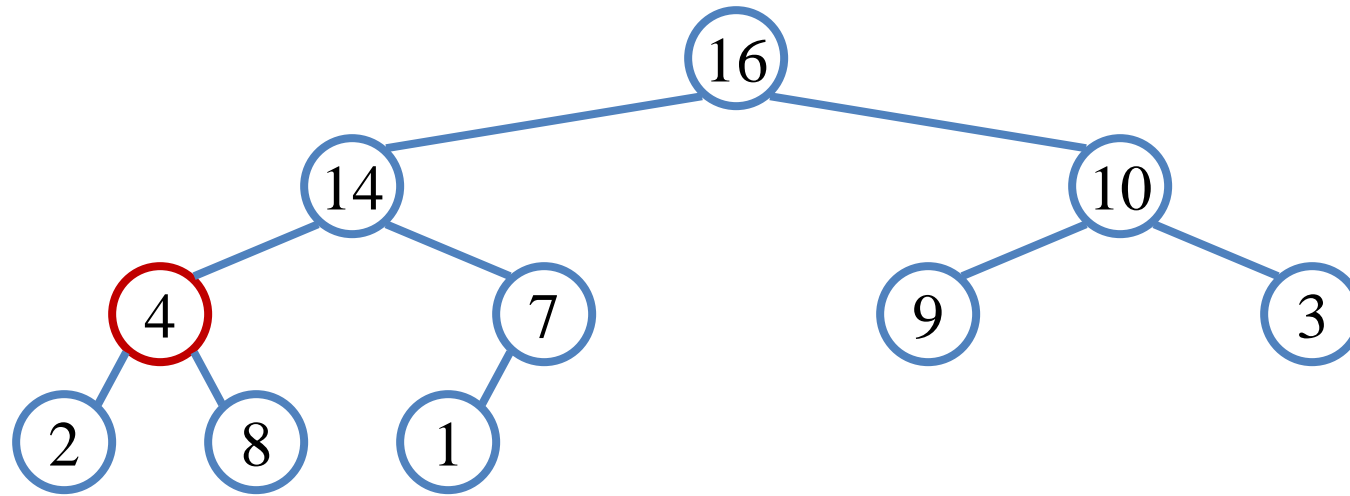
A = 

16	4	10	14	7	9	3	2	8	1
----	---	----	----	---	---	---	---	---	---

# Heapify() Example



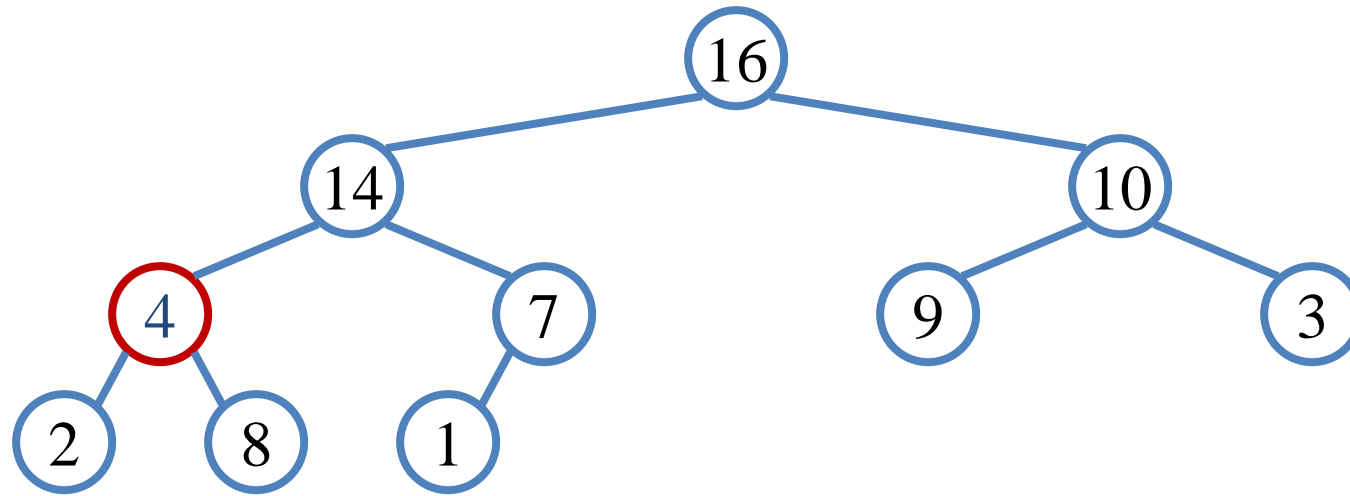
# Heapify() Example



A = 

16	14	10	4	7	9	3	2	8	1
----	----	----	---	---	---	---	---	---	---

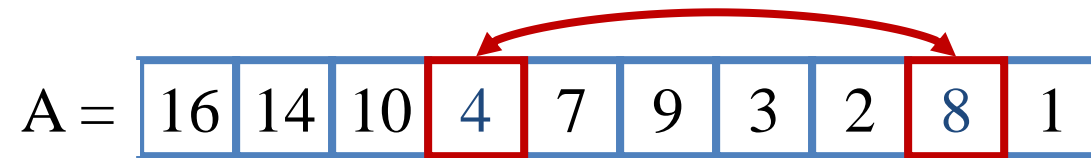
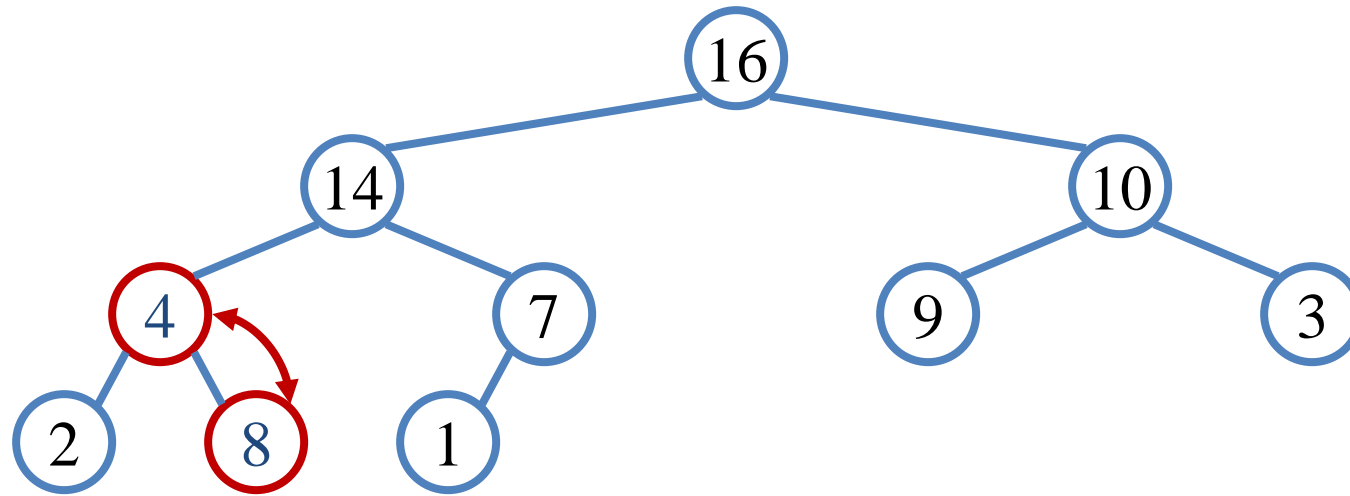
# Heapify() Example



A = 

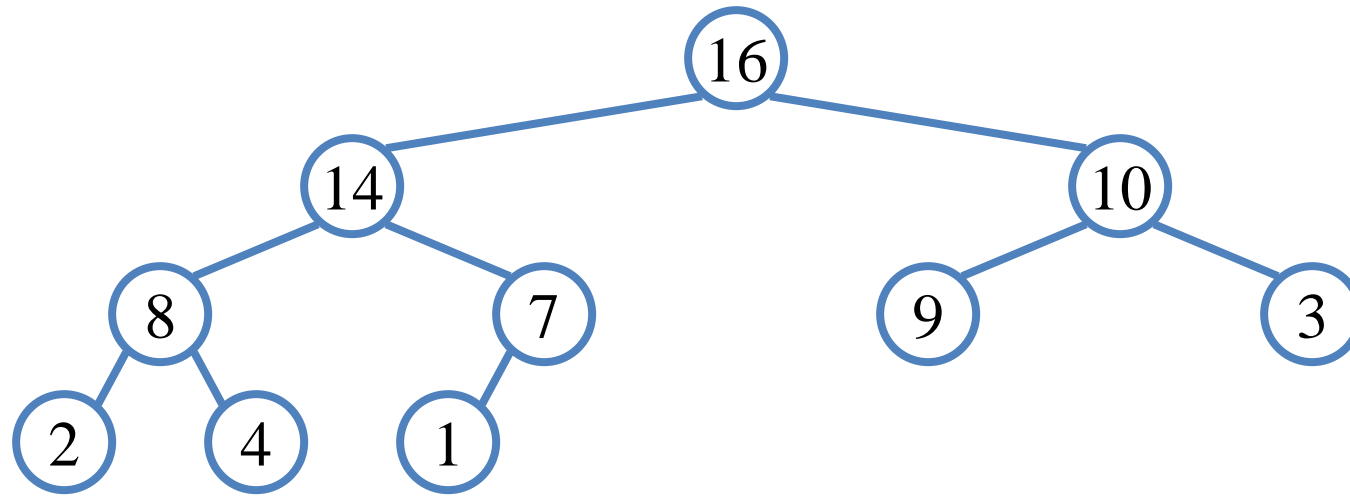
16	14	10	4	7	9	3	2	8	1
----	----	----	---	---	---	---	---	---	---

# Heapify() Example





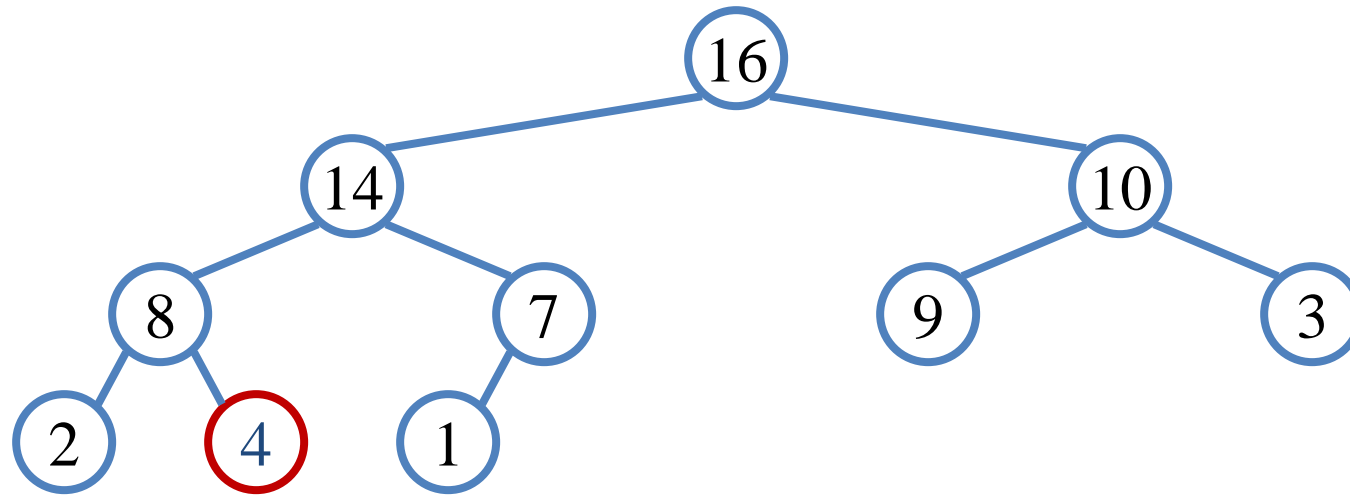
# Heapify() Example



A = 

16	14	10	8	7	9	3	2	4	1
----	----	----	---	---	---	---	---	---	---

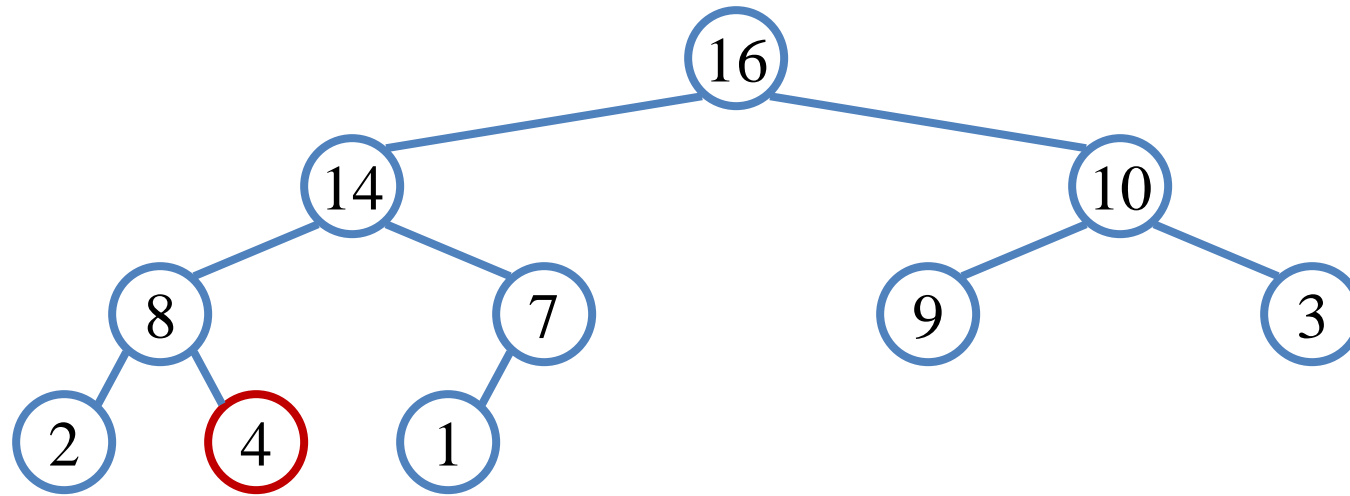
# Heapify() Example



A = 

16	14	10	8	7	9	3	2	4	1
----	----	----	---	---	---	---	---	---	---

# Heapify() Example



A = 

16	14	10	8	7	9	3	2	4	1
----	----	----	---	---	---	---	---	---	---

# Heap Operations: Heapify()

```
MAX-Heapify(A, i, Heap_size) // MAX-Heapify operation for MAX-Heap
{
    l = Left(i); r = Right(i);
    if (l <= heap_size(A) && A[l] > A[i])
        largest = l;
    else //find the largest between A[l] and A[r]
        largest = i;
    if (r <= heap_size(A) && A[r] > A[largest])
        largest = r;
    if (largest != i)
        Swap(A, i, largest);
    MAX-Heapify(A, largest, Heap_size);
}
```

# Analyzing Heapify(): Informal

- Aside from the recursive call, what is the running time of **Heapify()**?
- How many times can **Heapify()** recursively call itself?
- What is the worst-case running time of **Heapify()** on a heap of size  $n$ ?

# Analyzing Heapify(): Formal

- Fixing up relationships between  $i$ ,  $l$ , and  $r$  takes  $\Theta(1)$  time
- If the heap at  $i$  has  $n$  elements, how many elements can the subtrees at  $l$  or  $r$  have?
  - Draw it
- Answer:  $2n/3$  (worst case: bottom row 1/2 full,  $(2n/3+n/3=n)$ )
- So time taken by **Heapify** ( ) is given by
$$T(n) \leq T(2n/3) + \Theta(1)$$

# Analyzing Heapify(): Formal

- So we have

$$T(n) \leq T(2n/3) + \Theta(1)$$

- By case 2 of the Master Theorem,

$$T(n) = O(\lg n)$$

- Thus, **Heapify()** takes logarithmic time

# Heap Operations: BuildHeap()

- We can build a heap in a bottom-up manner by running **Heapify()** on successive subarrays
  - Fact: for array of length  $n$ , all elements in range  $A[\lfloor n/2 \rfloor + 1 .. n]$  are heaps (*Why?*)
  - The elements in the subarray  $A[(\lfloor n/2 \rfloor + 1) .. n]$  are leaves
  - So:
    - Walk backwards through the array from  $n/2$  to 1, calling **Heapify()** on each node.
    - Order of processing guarantees that the children of node  $i$  are heaps when  $i$  is processed

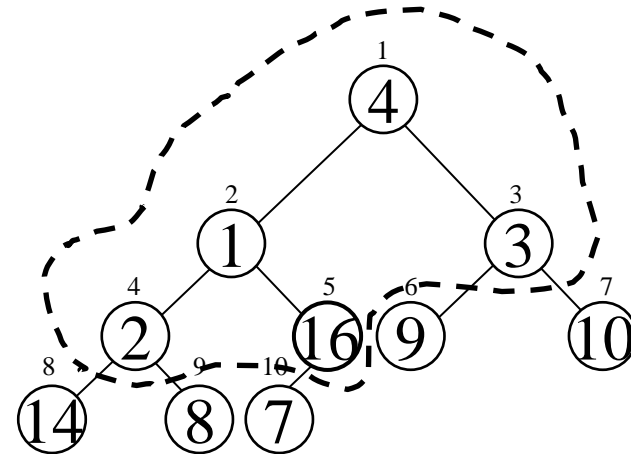


# Building a Heap

- Convert an array  $A[1 \dots n]$  into a max-heap ( $n = \text{length}[A]$ )
- The elements in the subarray  $A[(\lfloor n/2 \rfloor + 1) \dots n]$  are leaves
- Apply MAX-HEAPIFY on elements between 1 and  $\lfloor n/2 \rfloor$

*Alg:* BUILD-MAX-HEAP( $A$ )

1.  $n = \text{length}[A]$
2. **for**  $i \leftarrow \lfloor n/2 \rfloor$  **downto** 1
3.     **do** MAX-HEAPIFY( $A, i, n$ )



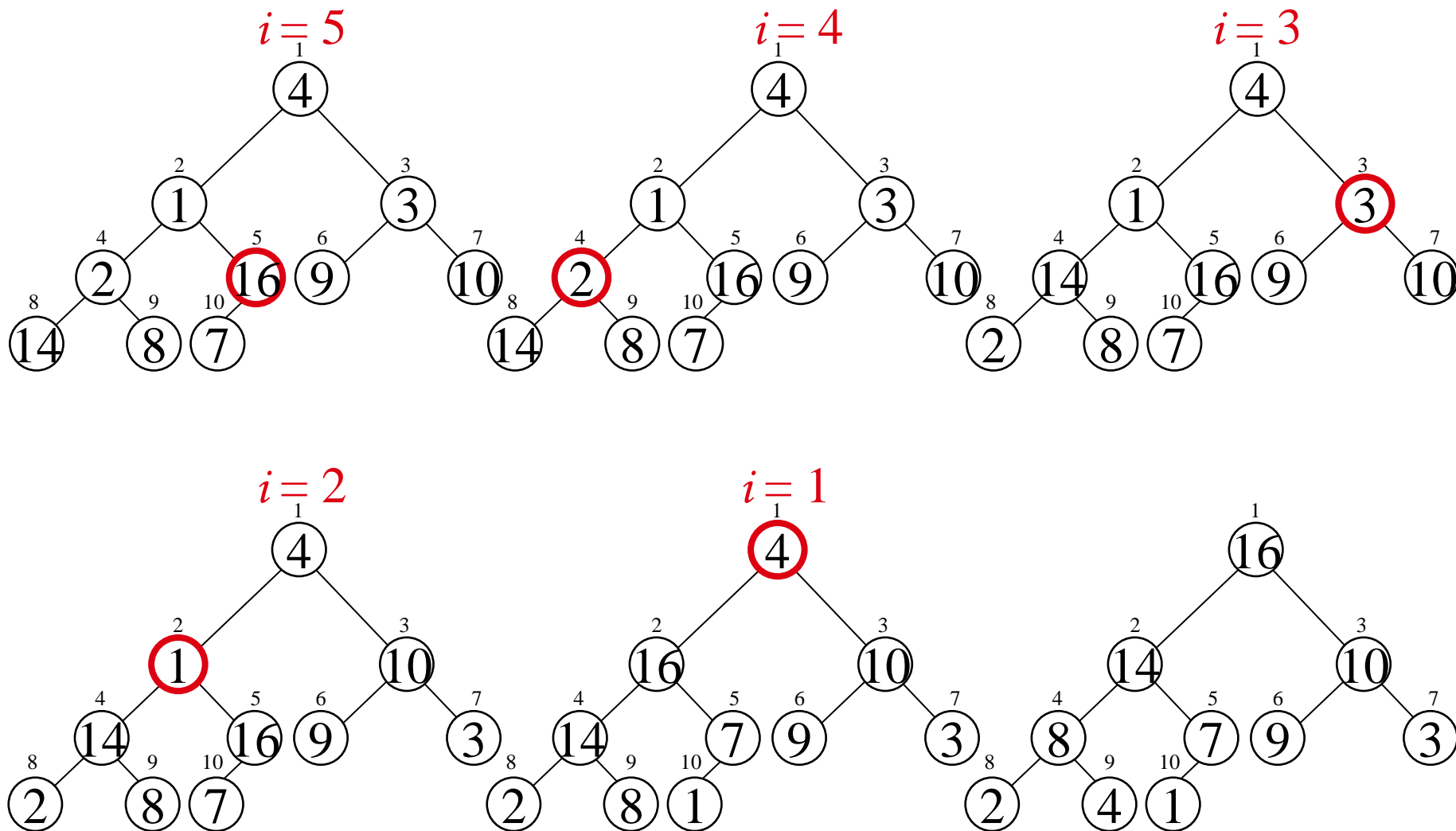
A: 

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---

# Example:

# A

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---



# Running Time of BUILD MAX HEAP

*Alg:* BUILD-MAX-HEAP( $A$ )

1.  $n = \text{length}[A]$
  2. **for**  $i \leftarrow \lfloor n/2 \rfloor$  **downto** 1
  3.     **do** MAX-HEAPIFY( $A, i, n$ )      $O(\lg n)$
- }  $O(n)$

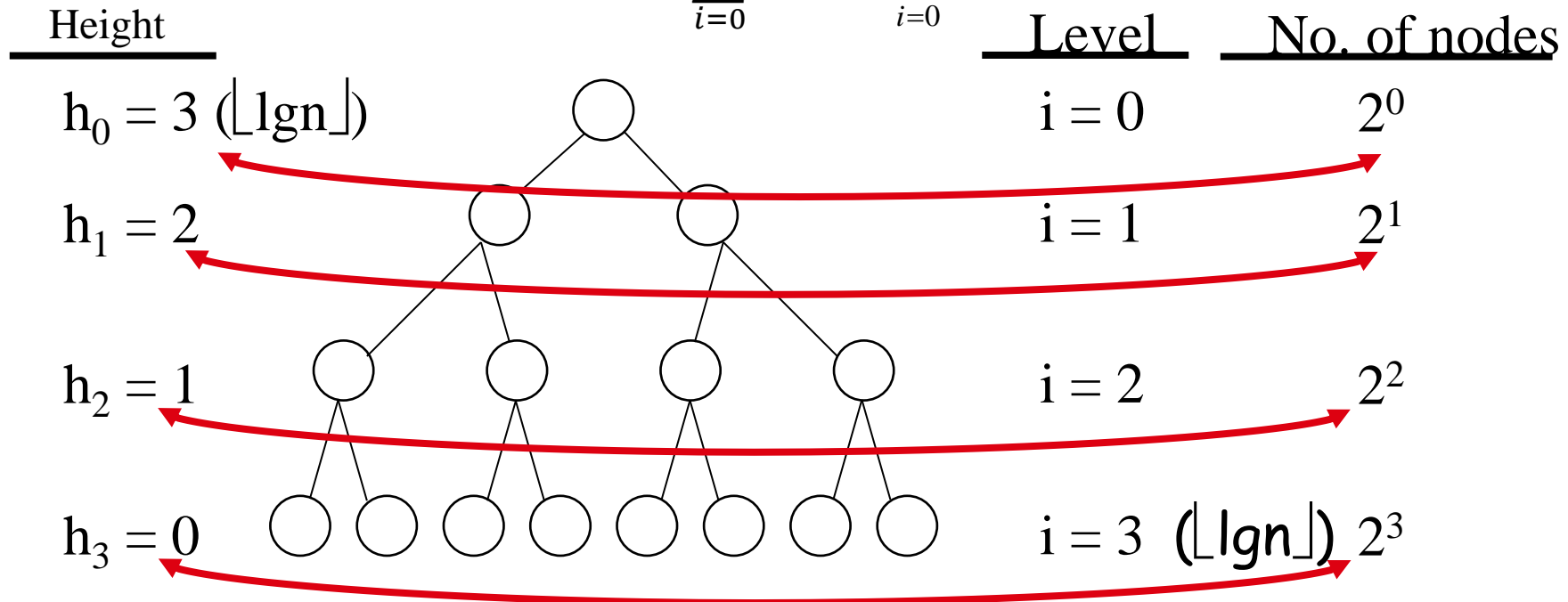
$\Rightarrow$  Running time:  $O(n \lg n)$

- This is not an **asymptotically tight** upper bound

# Running Time of BUILD MAX HEAP

- HEAPIFY takes  $O(h) \Rightarrow$  the cost of HEAPIFY on a node  $i$  is proportional to the height of the node  $i$  in the tree

$$\Rightarrow T(n) = \sum_{i=0}^h n_i h_i = \sum_{i=0}^h 2^i (h-i) = O(n)$$



$h_i = h - i$  height of the heap rooted at level  $i$

$n_i = 2^i$  number of nodes at level  $i$

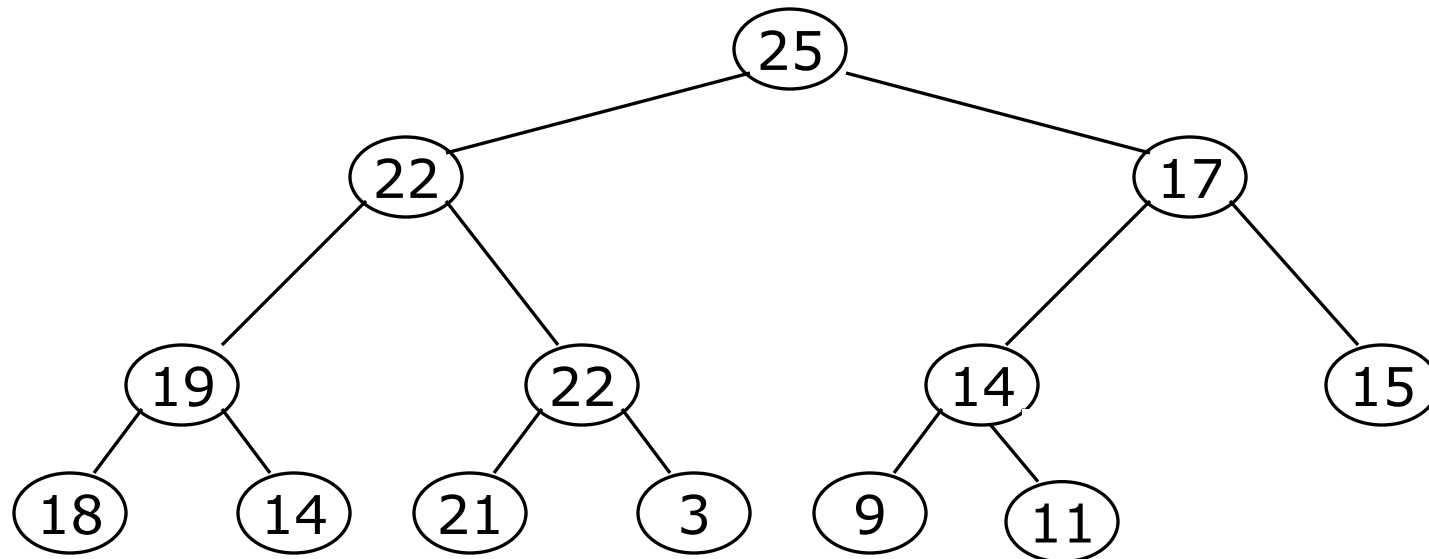
# Running Time of BUILD MAX HEAP

$T(n) = \sum_{i=0}^h n_i h_i$	Cost of HEAPIFY at level i * number of nodes at that level
$= \sum_{i=0}^h 2^i (h - i)$	Replace the values of $n_i$ and $h_i$ computed before
$= \sum_{i=0}^h \frac{h-i}{2^{h-i}} 2^h$	Multiply by $2^h$ both at the nominator and denominator and write $2^i$ as $\frac{1}{2^{-i}}$
$= 2^h \sum_{k=0}^h \frac{k}{2^k}$	Change variables: $k = h - i$
$\leq n \sum_{k=0}^{\infty} \frac{k}{2^k}$	The sum above is smaller than the sum of all elements to $\infty$ and $h = \lg n$
$= O(n)$	The sum above is smaller than 2

Running time of BUILD-MAX-HEAP:  $T(n) = O(n)$

# Removing the root

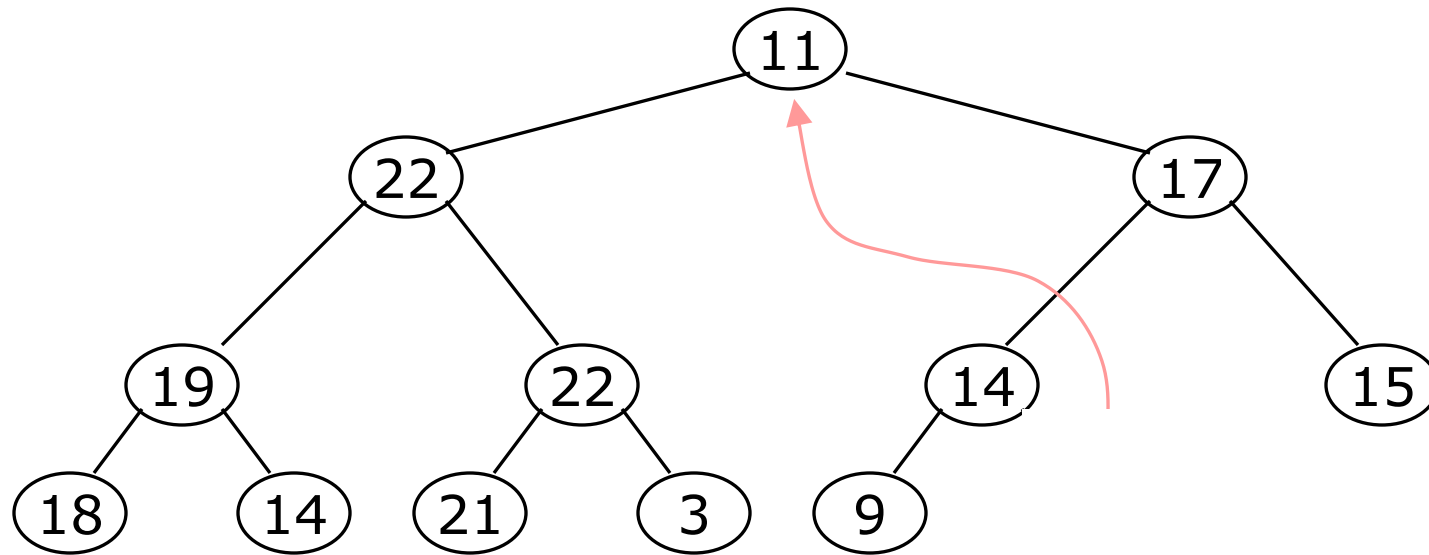
- Notice that the **largest** number is now in the root
- Suppose we *discard* the root:



- How can we fix the binary tree so that it is once again *balanced and left-justified*?
- **Solution: remove the rightmost leaf at the deepest level and use it for the new root**

# Removing the root

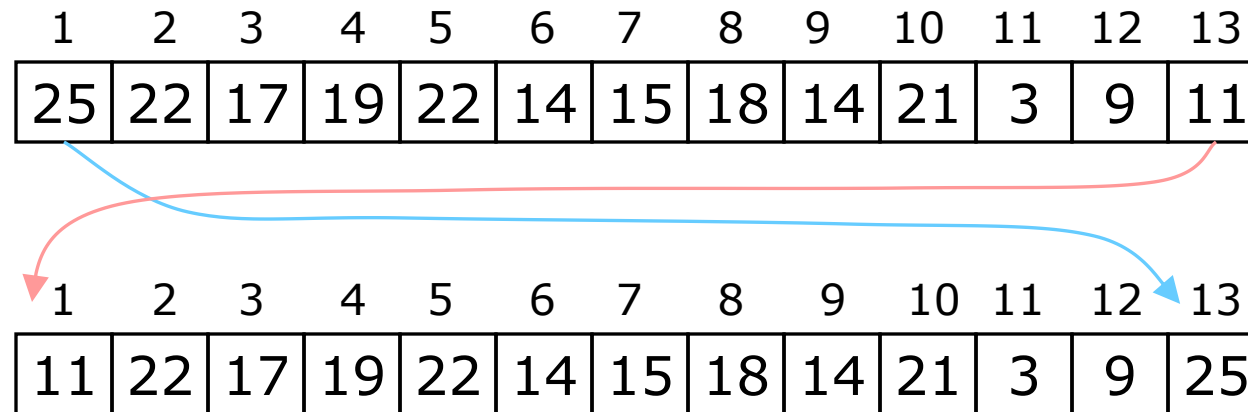
- Notice that the **largest** number is now in the root
- Suppose we *discard* the root:



- How can we fix the binary tree so it is once again *balanced and left-justified*?
- **Solution: remove the rightmost leaf at the deepest level and use it for the new root**

# Removing and replacing the root

- The “root” is the first element in the array
- The “rightmost node at the deepest level” is the last element
- Swap them...



- ...And pretend that the last element in the array no longer exists—that is, the “last index” is **12 (9)**



# Removing and replacing the root

DEL\_HEAP(A, N)

// the procedure deletes the max from heap A[1:N] and stores it in ITEM

{

if N = 0 then “heap is empty”

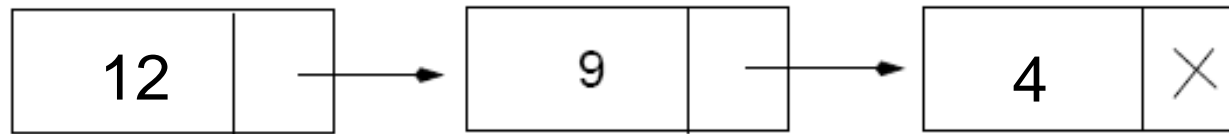
SWAP (A, 1, N)

MAX\_HEAPIFY(A,1,N-1)

}

# Priority Queues

- Each element is associated with a value (Priority)
- A priority queue is different from a "normal" queue, because instead of being a "first-in-first-out" data structure, values come out in order by priority.
- The key with highest (Lowest) priority is dequeued first



# Operations on Priority Queues

- Max-priority queues support the following operations:
  - $\text{INSERT}(S, x)$ : inserts element  $x$  into set  $S$
  - $\text{MAXIMUM}(S)$ : returns element of  $S$  with largest key
  - $\text{EXTRACT-MAX}(S)$ : removes and returns element of  $S$  with largest key
  - $\text{INCREASE-KEY}(S, x, k)$ : increases value of element  $x$ 's key to  $k$  (Assume  $k \geq x$ 's current key value)

# HEAP-MAXIMUM

Goal:

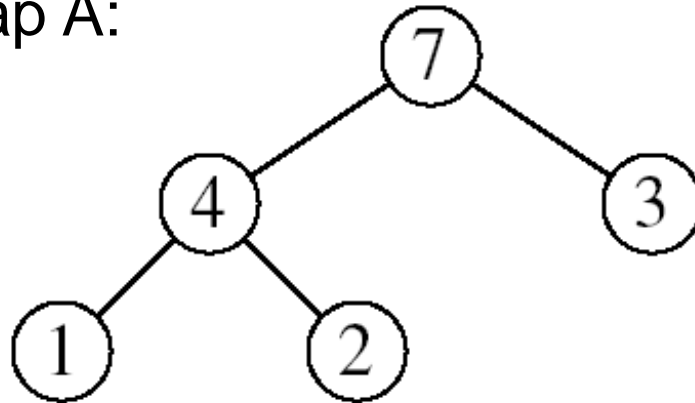
- Return the largest element of the heap

*Alg:* HEAP-MAXIMUM( $A$ )

Running time:  $O(1)$

1.      **return**  $A[1]$

Heap A:



Heap-Maximum( $A$ ) returns 7

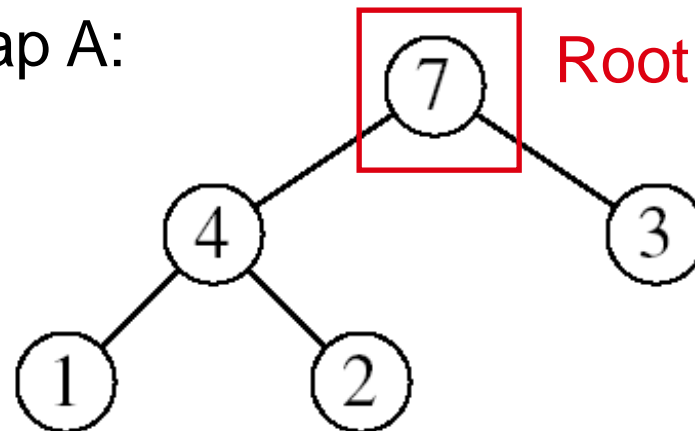
# HEAP-EXTRACT-MAX

Goal: Extract the largest element of the heap (i.e., return the max value and also remove that element from the heap)

Idea:

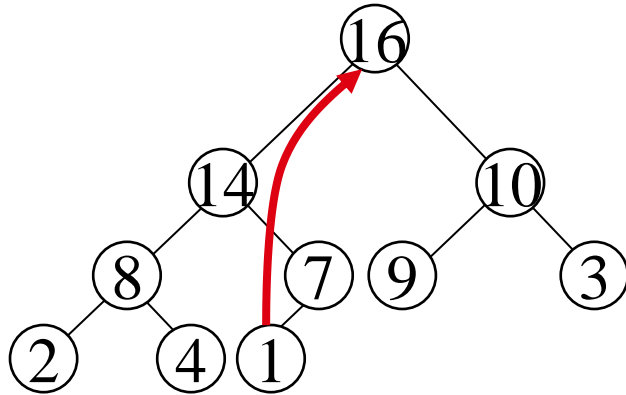
- Exchange the root element with the last
- Decrease the size of the heap by 1 element
- Call MAX-HEAPIFY on the new root, on a heap of size  $n-1$

Heap A:

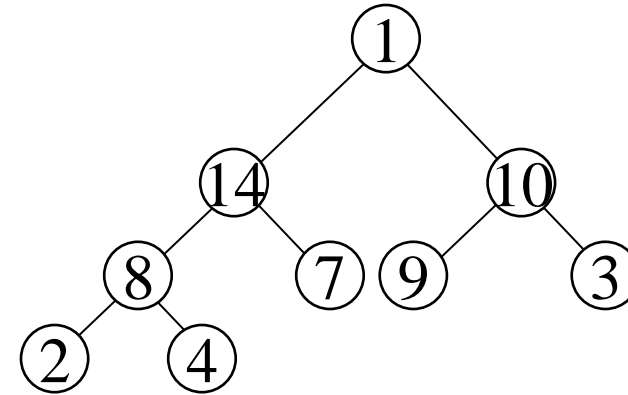


Root is the largest element

# Example: HEAP-EXTRACT-MAX

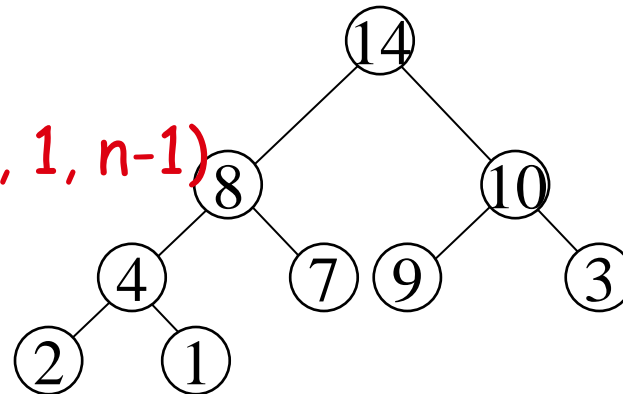


max = 16



Heap size decreased by 1

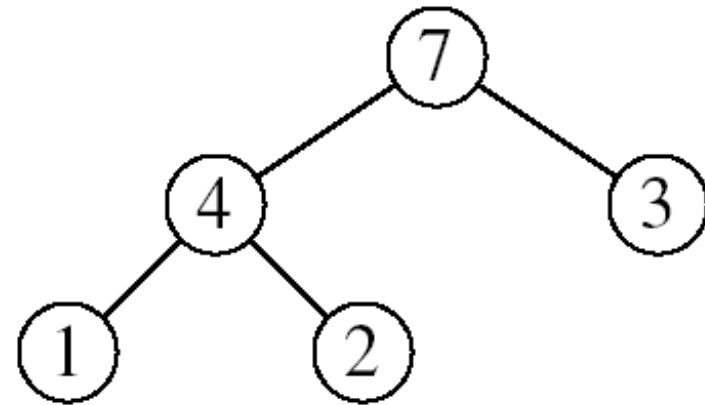
Call MAX-HEAPIFY(A, 1, n-1)



# HEAP-EXTRACT-MAX

*Alg:* HEAP-EXTRACT-MAX( $A, n$ )

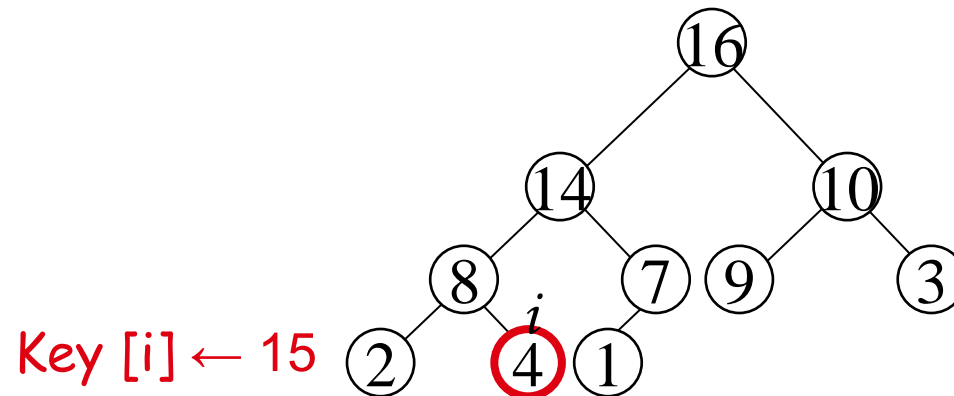
1. if  $n < 1$
2.     **then error** “heap underflow”
3.      $\text{max} \leftarrow A[1]$
4.      $A[1] \leftarrow A[n]$
5.     MAX-HEAPIFY( $A, 1, n-1$ )     *//remakes heap*
6.     **return** max



Running time:  $O(\lg n)$

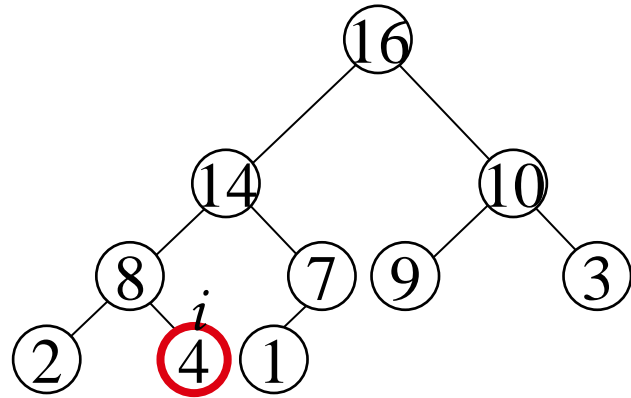
# HEAP-INCREASE-KEY

- Goal:
  - Increases the key of an element  $i$  in the heap
- Idea:
  - Increment the key of  $A[i]$  to its new value
  - If the max-heap property does not hold anymore: traverse a path toward the root to find the proper place for the newly increased key

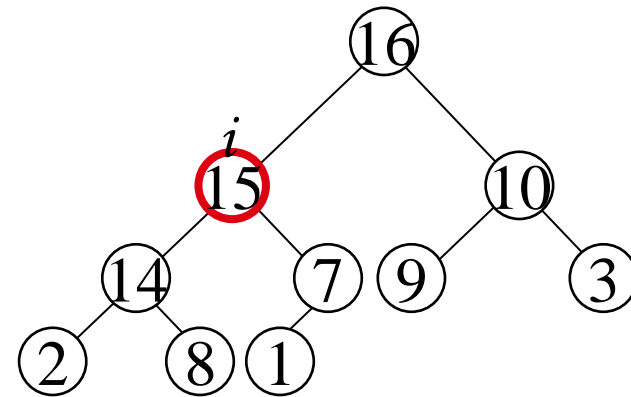
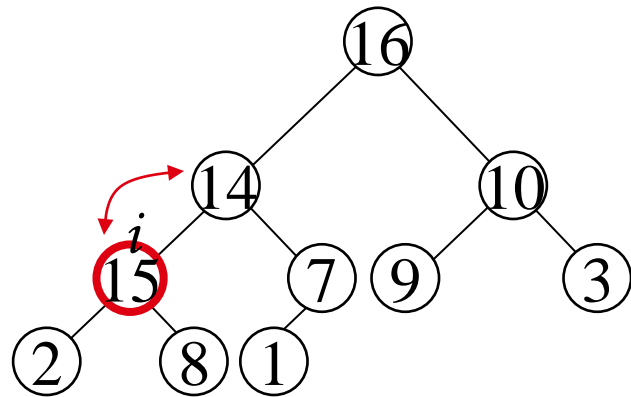
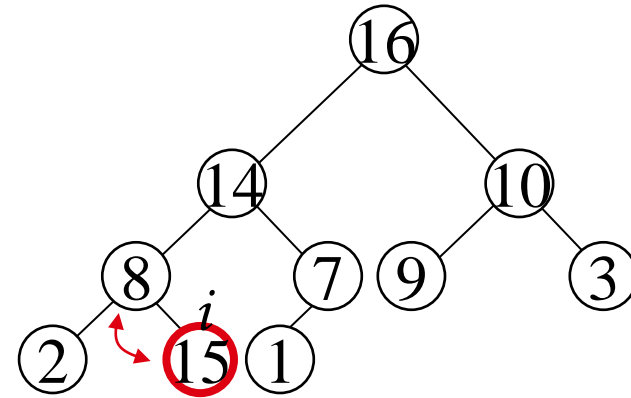




# Example: HEAP-INCREASE-KEY



$Key[i] \leftarrow 15$

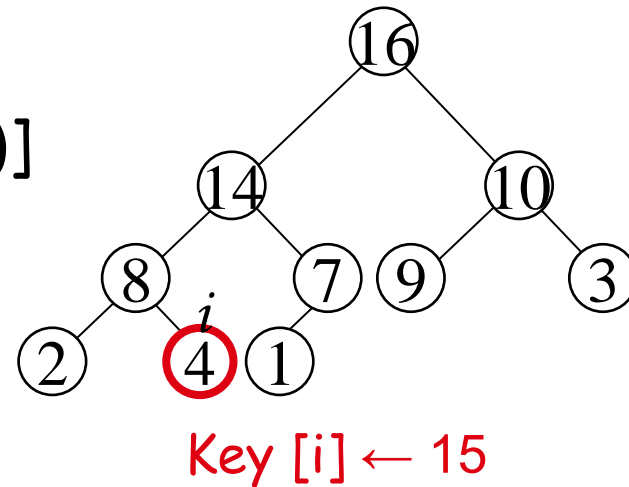


# HEAP-INCREASE-KEY

*Alg:* HEAP-INCREASE-KEY( $A, i, \text{key}$ )

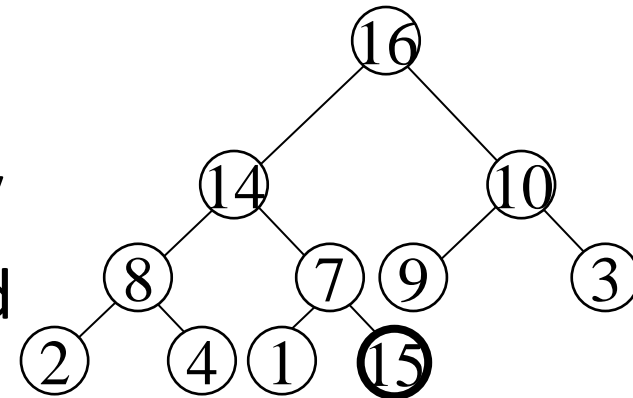
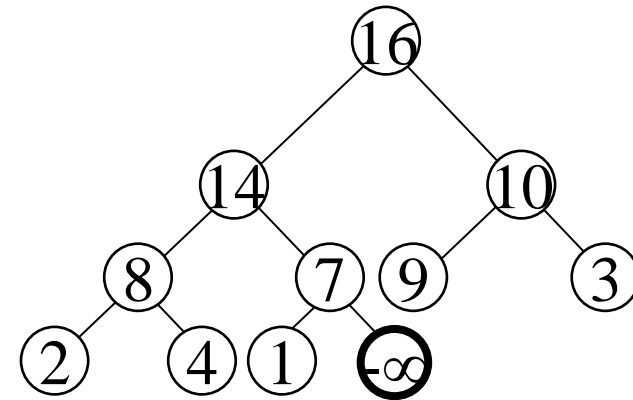
1. **if**  $\text{key} < A[i]$
2.     **then error** “new key is smaller than current key”
3.      $A[i] \leftarrow \text{key}$
4.     **while**  $i > 1$  and  $A[\text{PARENT}(i)] < A[i]$
5.         **do** exchange  $A[i] \leftrightarrow A[\text{PARENT}(i)]$
6.          $i \leftarrow \text{PARENT}(i)$

- Running time:  $O(\lg n)$



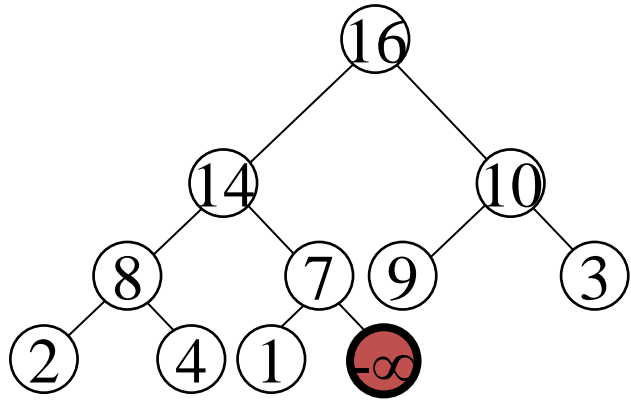
# MAX-HEAP-INSERT

- Goal:
  - Inserts a new element into a max-heap
- Idea:
  - Expand the max-heap with a new element whose key is  $-\infty$
  - Calls HEAP-INCREASE-KEY to set the key of the new node to its correct value and maintain the max-heap property

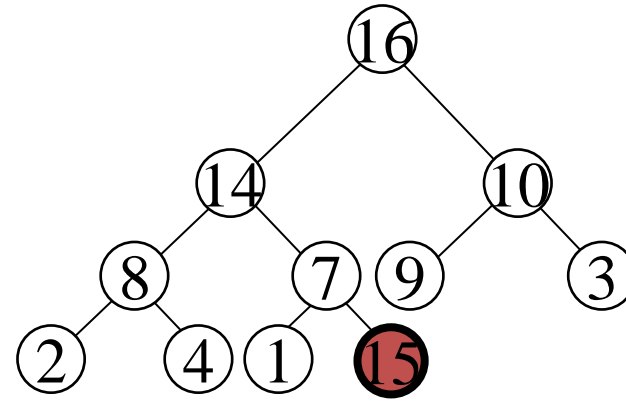


# Example: MAX-HEAP-INSERT

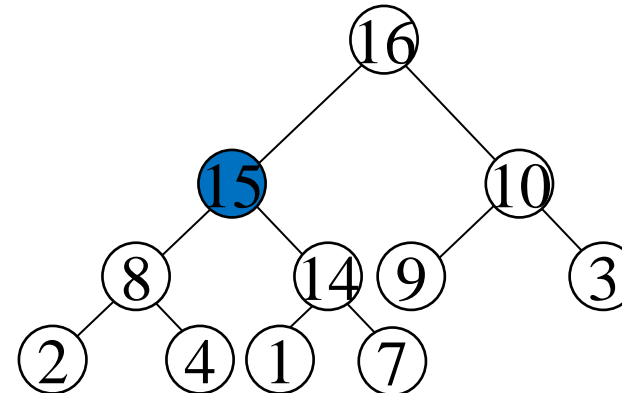
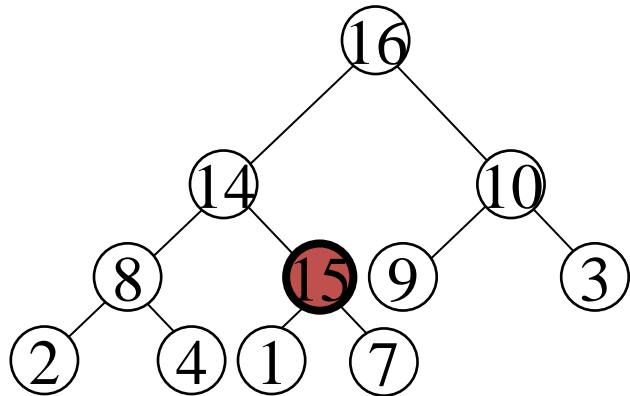
Insert value 15:  
- Start by inserting  $-\infty$



Increase the key to 15  
Call HEAP-INCREASE-KEY on  $A[11] = 15$



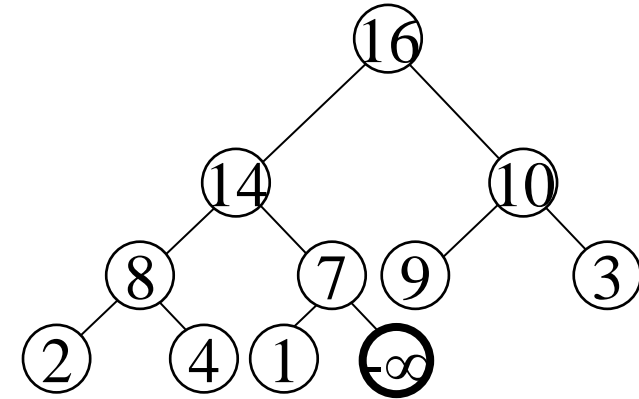
The restored heap containing  
the newly added element



# MAX-HEAP-INSERT

*Alg:* MAX-HEAP-INSERT( $A$ ,  $key$ ,  $n$ )

1.  $heap-size[A] \leftarrow n + 1$
2.  $A[n + 1] \leftarrow -\infty$
3. HEAP-INCREASE-KEY( $A$ ,  $n + 1$ ,  $key$ )



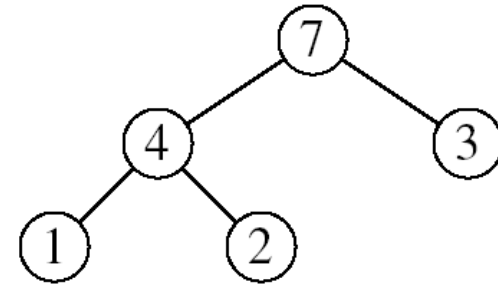
Running time:  $O(\lg n)$

# Why study Heapsort?

- It is a well-known, traditional sorting algorithm you will be expected to know
- Heapsort is *always*  $O(n \log n)$ 
  - Quicksort is usually  $O(n \log n)$  but in the worst case slows to  $O(n^2)$
  - Quicksort is generally faster, but Heapsort is better in time-critical applications

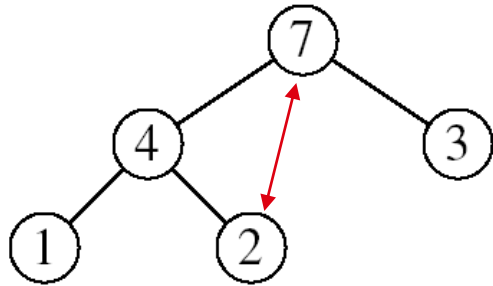
# Heapsort

- Goal:
  - Sort an array using heap representations
- Idea:
  - Build a **max-heap** from the array
  - Swap the root (the maximum element) with the last element in the array
  - “Discard” this last node by decreasing the heap size
  - Call MAX-HEAPIFY on the new root
  - Repeat this process until only one node remains

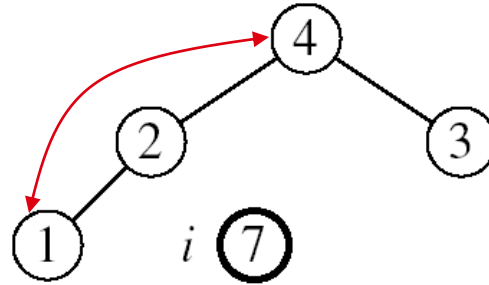


Example:

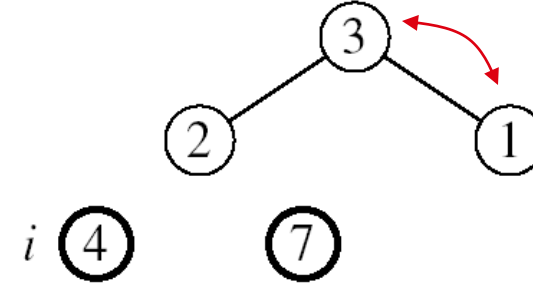
$A=[7, 4, 3, 1, 2]$



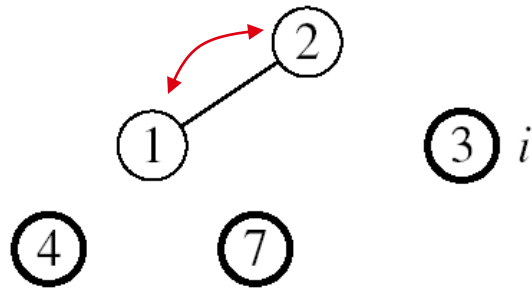
MAX-HEAPIFY(A, 1, 4)



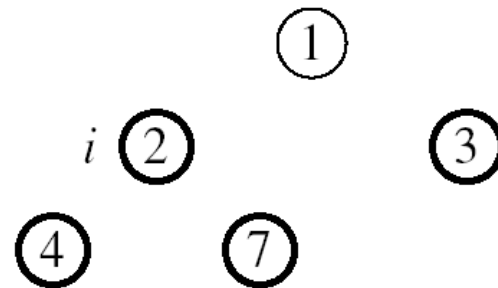
MAX-HEAPIFY(A, 1, 3)



MAX-HEAPIFY(A, 1, 2)



MAX-HEAPIFY(A, 1, 1)



A

1	2	3	4	7
---	---	---	---	---



# *Alg*: HEAPSORT(*A*)

1. BUILD-MAX-HEAP(*A*)  $O(n)$
  2. **for**  $i \leftarrow \text{length}[A]$  **downto** 2
  3.     **do** exchange  $A[1] \leftrightarrow A[i]$
  4.     MAX-HEAPIFY(*A*, 1,  $i - 1$ )  $O(\lg n)$
- }  $n-1$  times

- Running time:  $O(n \lg n)$

# Summary

- We can perform the following operations on heaps:
  - MAX-HEAPIFY  $O(\lg n)$
  - BUILD-MAX-HEAP  $O(n)$
  - HEAP-SORT  $O(n \lg n)$
  - MAX-HEAP-INSERT  $O(\lg n)$
  - HEAP-EXTRACT-MAX  $O(\lg n)$
  - HEAP-INCREASE-KEY  $O(\lg n)$
  - HEAP-MAXIMUM  $O(1)$

# References

- David Matuszek, University Of Pennsylvania
  - [www.cis.upenn.edu/~matuszek/](http://www.cis.upenn.edu/~matuszek/)
- Dr. George Bebis, University of Nevada
  - Course page: [www.cse.unr.edu/~bebis](http://www.cse.unr.edu/~bebis)