

Chapter 10



Heapsort

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Lecture Outline

- Heap
- Binary Heap
- Heap Property
- Building A Heap
- The HeapSort Algorithm



Introduction

- Heapsort

- Running time: $O(n \lg n)$
 - Like merge sort
- Sort in place: only a constant number of array elements are stored outside the input array at any time
 - Like insertion sort

- Heap

- A data structure used by Heapsort to manage information during the execution of the algorithm
- Can be used as an efficient priority queue



Binary Heap

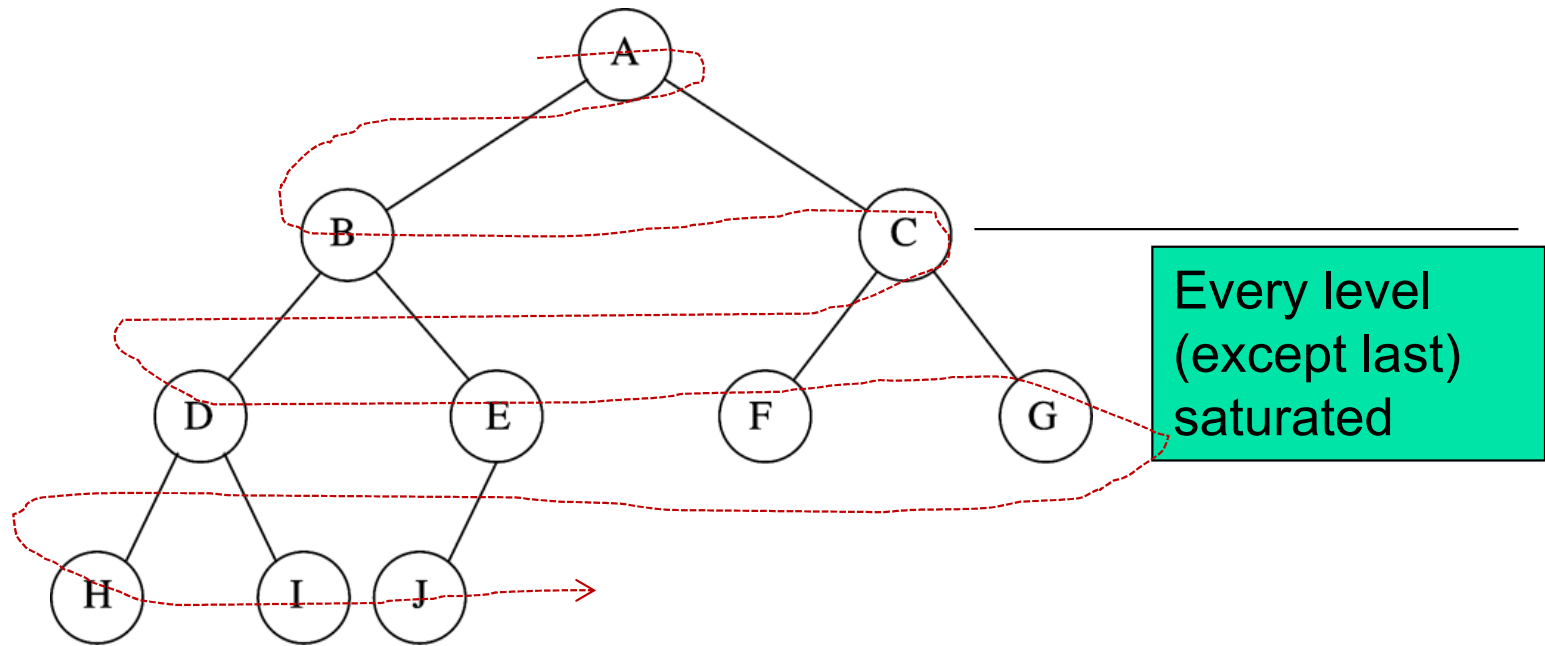
- A heap can be seen as a complete binary tree
- The tree is completely filled on all levels except possibly the lowest.
- In practice, heaps are usually implemented as arrays
- An array A that represent a heap is an object with two attributes:
 $A[1 \dots \text{length}[A]]$
 - $\text{length}[A]$: # of elements in the array
 - $\text{heap-size}[A]$: # of elements in the heap stored within array A , where $\text{heap-size}[A] \leq \text{length}[A]$
 - No element past $A[\text{heap-size}[A]]$ is an element of the heap
- max-heap and min-heap

$A =$

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

Binary Heap Example

N=10



Array representation:

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

A Max-Heap

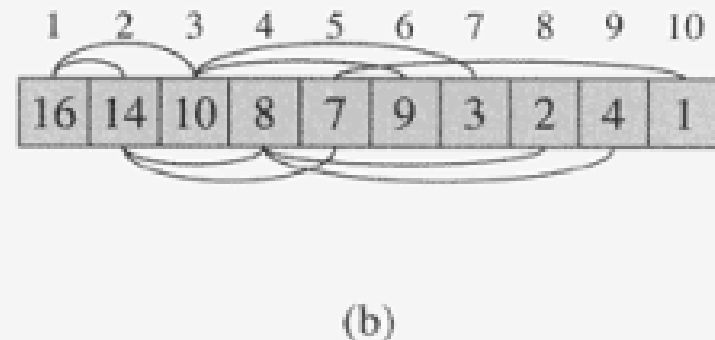
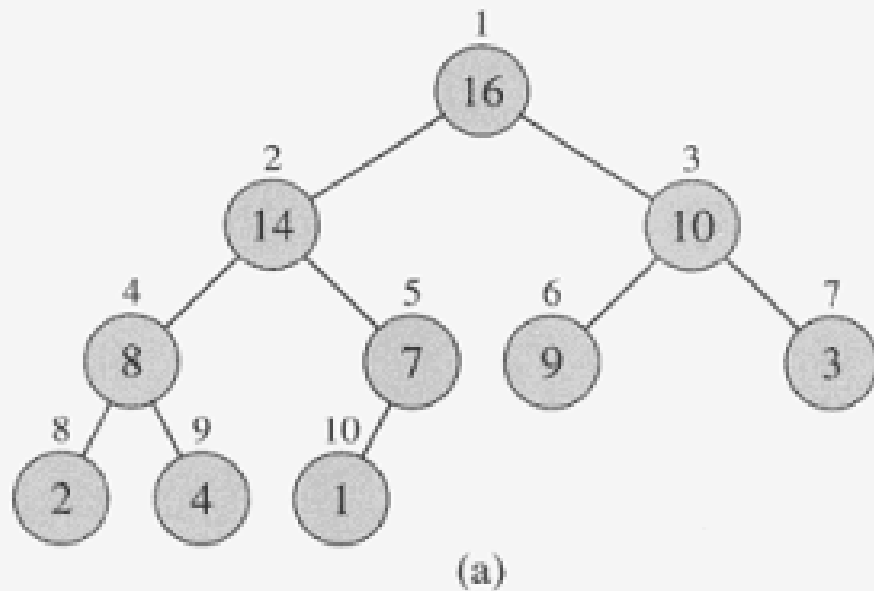


Figure 6.1 A max-heap viewed as (a) a binary tree and (b) an array. The number within the circle at each node in the tree is the value stored at that node. The number above a node is the corresponding index in the array. Above and below the array are lines showing parent-child relationships; parents are always to the left of their children. The tree has height three; the node at index 4 (with value 8) has height one.

Referencing Heap Elements

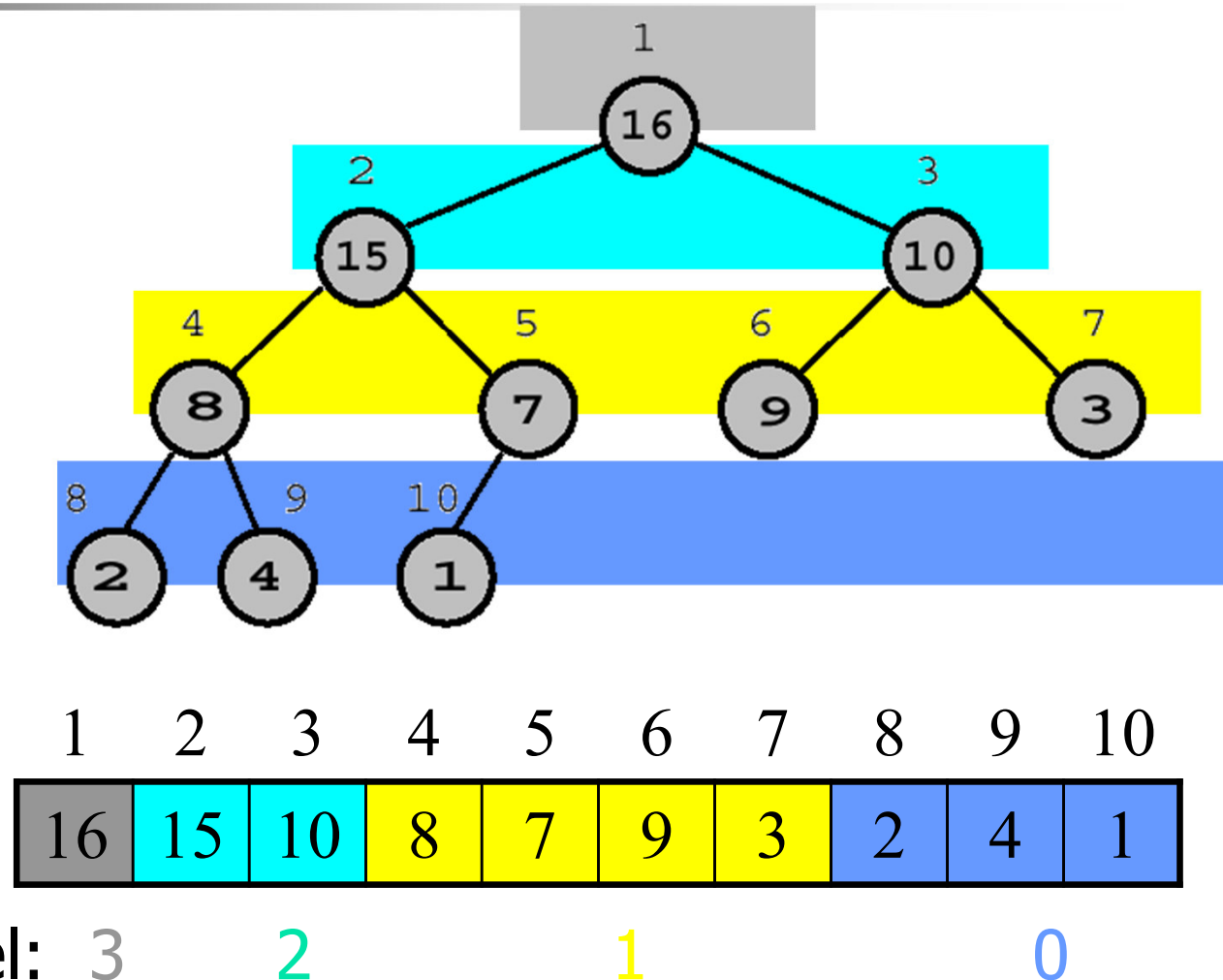
- The root node is $A[1]$

- Node i is $A[i]$

- Parent(i)**
 - return $\lfloor i/2 \rfloor$

- Left(i)**
 - return $2*i$

- Right(i)**
 - return $2*i + 1$



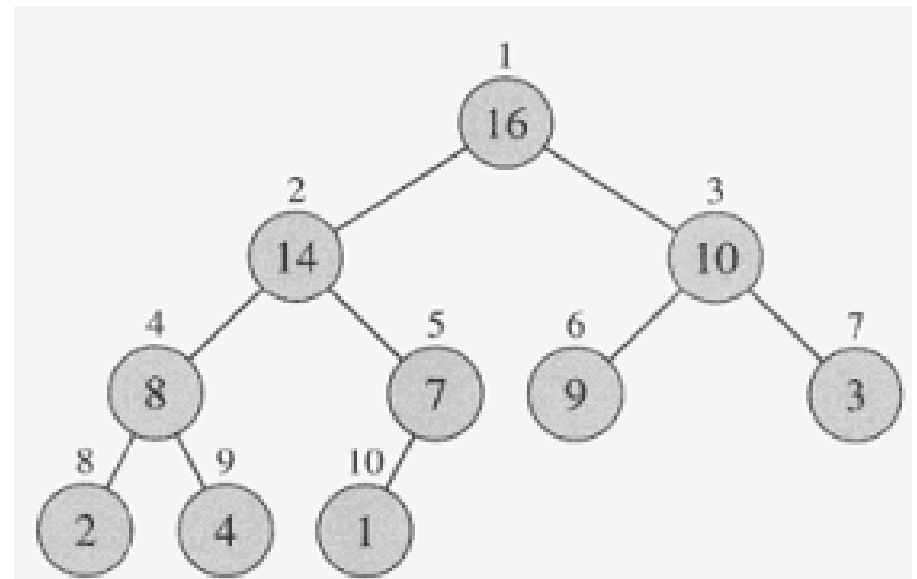
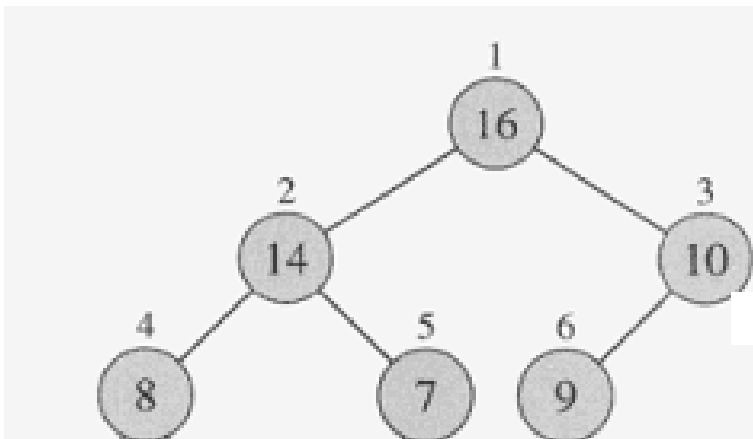


Heap Property

- Heap property – the property that the values in the node must satisfy
- Max-heap property: for every node i other than the root
 - $A[\text{PARENT}(i)] \geq A[i]$
 - The value of a node is at most the value of its parent
 - The largest element in a max-heap is stored at the root
 - The subtree rooted at a node contains values no larger than that contained at the node itself
- Min-heap property: for every node i other than the root
 - $A[\text{PARENT}(i)] \leq A[i]$

Heap Height

- The **height** of a node in a heap is the number of edges on the longest simple downward path from the node to a leaf
- The height of a heap is the height of its root
 - The height of a heap of n elements is $\Theta(\lg n)$





of nodes in each level

- Fact: an n -element heap has at most 2^{h-k} nodes of level k , where h is the height of the tree
- for $k = h$ (root level) $\rightarrow 2^{h-h} = 2^0 = 1$
- for $k = h-1$ $\rightarrow 2^{h-(h-1)} = 2^1 = 2$
- for $k = h-2$ $\rightarrow 2^{h-(h-2)} = 2^2 = 4$
- for $k = h-3$ $\rightarrow 2^{h-(h-3)} = 2^3 = 8$
- ...
- for $k = 1$ $\rightarrow 2^{h-1} = 2^{h-1}$
- for $k = 0$ (leaves level) $\rightarrow 2^{h-0} = 2^h$



Heap Height

- A heap storing n keys has height $h = \lfloor \lg n \rfloor = \Theta(\lg n)$
- Due to heap being **complete**, we know:
 - The maximum # of nodes in a heap of height h
 - $2^h + 2^{h-1} + \dots + 2^2 + 2^1 + 2^0 =$
 - $\sum_{i=0 \text{ to } h} 2^i = (2^{h+1} - 1) / (2 - 1) = 2^{h+1} - 1$
 - The minimum # of nodes in a heap of height h
 - $1 + 2^{h-1} + \dots + 2^2 + 2^1 + 2^0 =$
 - $\sum_{i=0 \text{ to } h-1} 2^i + 1 = (2^{h-1+1} - 1) / (2 - 1) + 1 = 2^h$
 - Therefore
 - $2^h \leq n \leq 2^{h+1} - 1$
 - $h \leq \lg n$ & $\lg(n+1) - 1 \leq h$
 - $\lg(n+1) - 1 \leq h \leq \lg n$
 - which in turn implies:
 - $h = \lfloor \lg n \rfloor = \Theta(\lg n)$



Heap Procedures

- MAX-HEAPIFY: maintain the max-heap property
 - $O(\lg n)$
- BUILD-MAX-HEAP: produces a max-heap from an unordered input array
 - $O(n)$
- HEAPSORT: sorts an array in place
 - $O(n \lg n)$
- MAX-HEAP-INSERT, HEAP-EXTRACT, HEAP-INCREASE-KEY, HEAP-MAXIMUM: allow the heap data structure to be used as a **priority queue**
 - $O(\lg n)$



Maintaining the Heap Property

■ MAX-HEAPIFY

- Inputs: an array A and an index i into the array
- Assume the binary tree rooted at $\text{LEFT}(i)$ and $\text{RIGHT}(i)$ are max-heaps, but $A[i]$ may be smaller than its children (**violate the max-heap property**)
- MAX-HEAPIFY let the value at $A[i]$ floats down in the max-heap



MAX-HEAPIFY

MAX-HEAPIFY(A, i)

1 $l \leftarrow \text{LEFT}(i)$

2 $r \leftarrow \text{RIGHT}(i)$

} Extract the indices of LEFT and RIGHT children of i

3 **if** $l \leq \text{heap-size}[A]$ and $A[l] > A[i]$

4 **then** $\text{largest} \leftarrow l$

5 **else** $\text{largest} \leftarrow i$

Choose the largest of $A[i]$, $A[l]$, $A[r]$

6 **if** $r \leq \text{heap-size}[A]$ and $A[r] > A[\text{largest}]$

7 **then** $\text{largest} \leftarrow r$

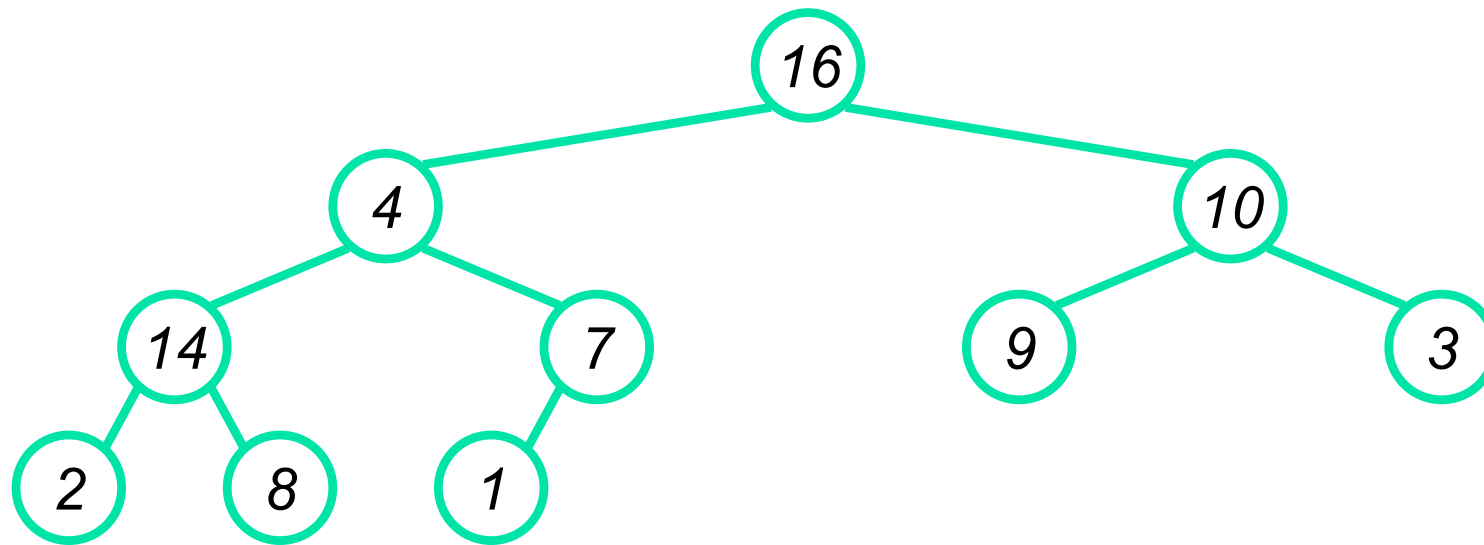
8 **if** $\text{largest} \neq i$

Float down $A[i]$ recursively

9 **then** exchange $A[i] \leftrightarrow A[\text{largest}]$

10 MAX-HEAPIFY($A, \text{largest}$)

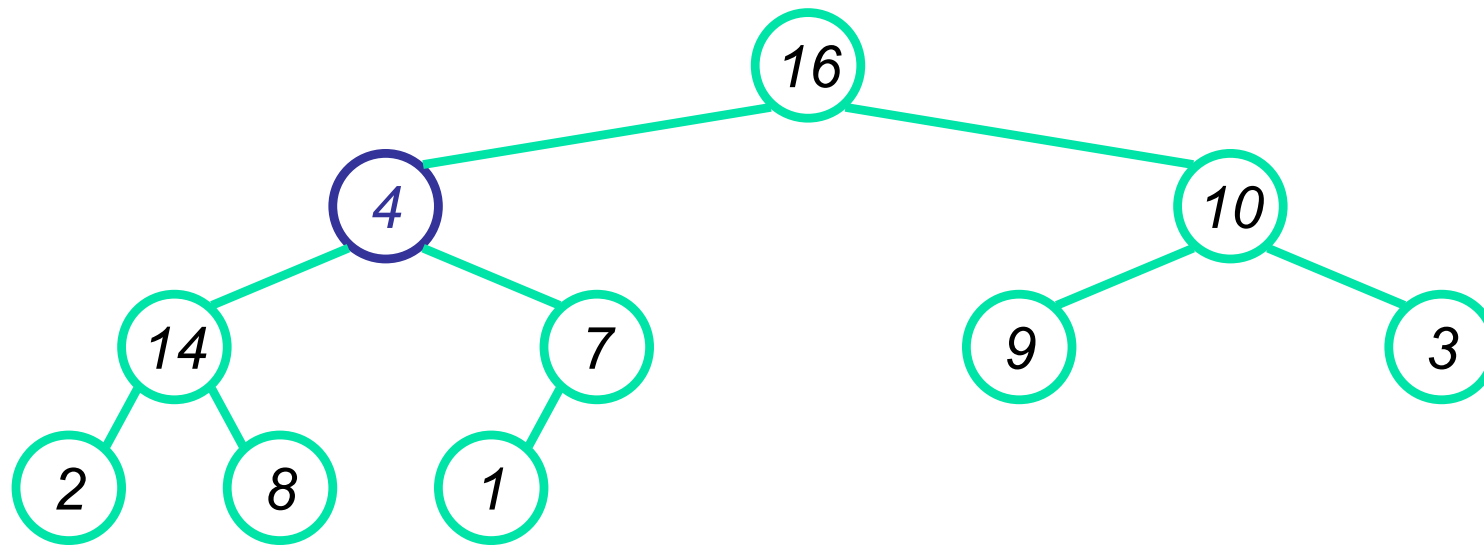
MAX-HEAPIFY Example



$A =$

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

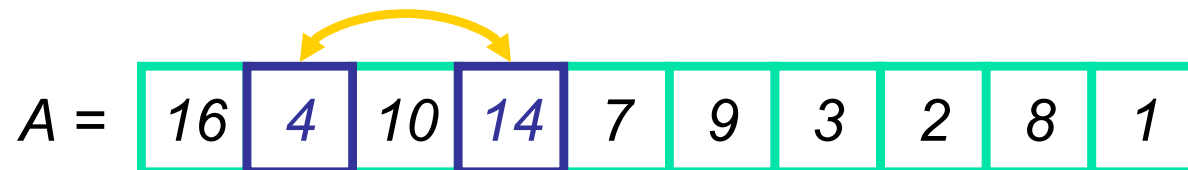
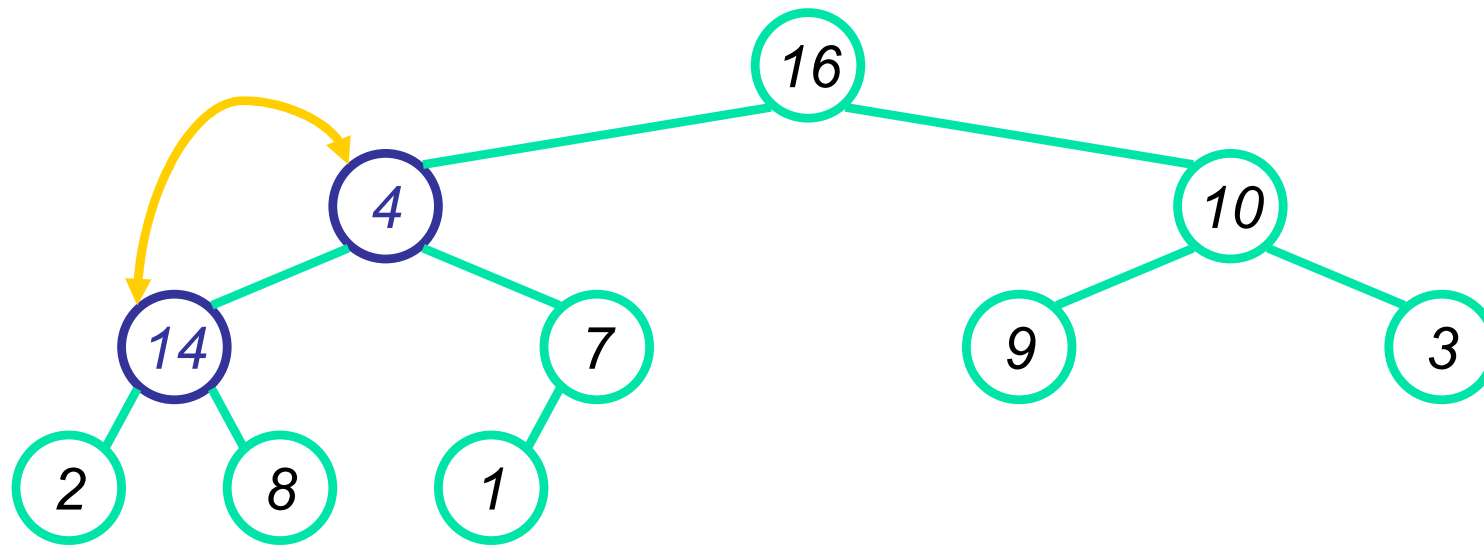
MAX-HEAPIFY Example



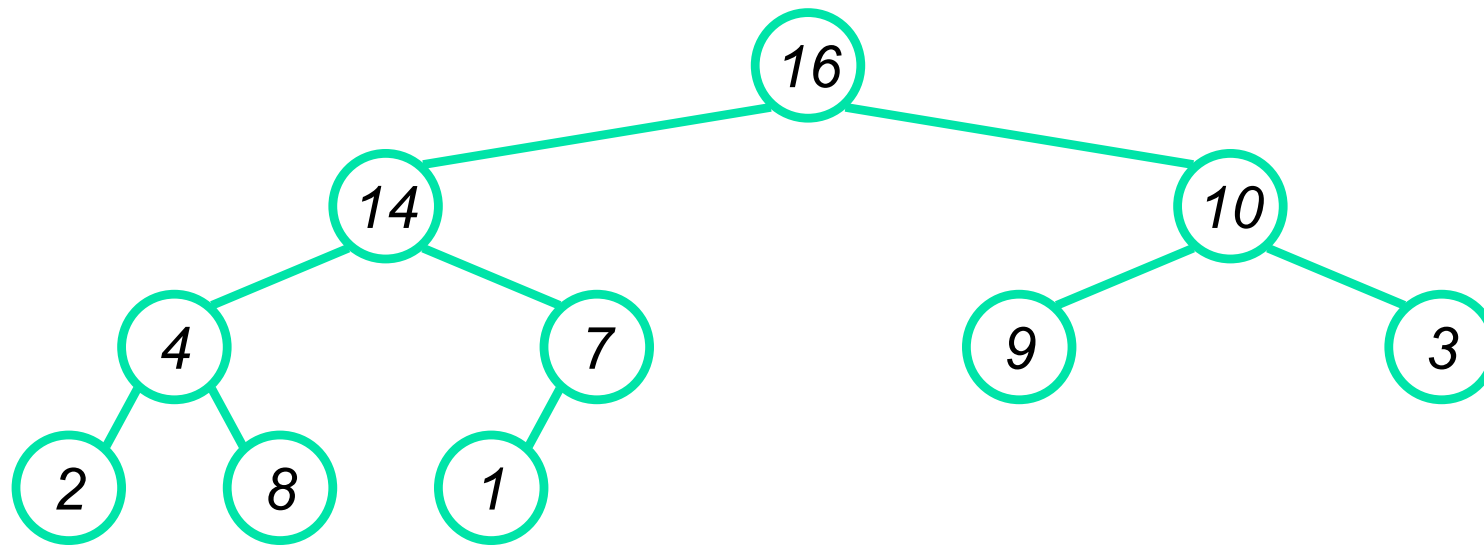
$A =$

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

MAX-HEAPIFY Example



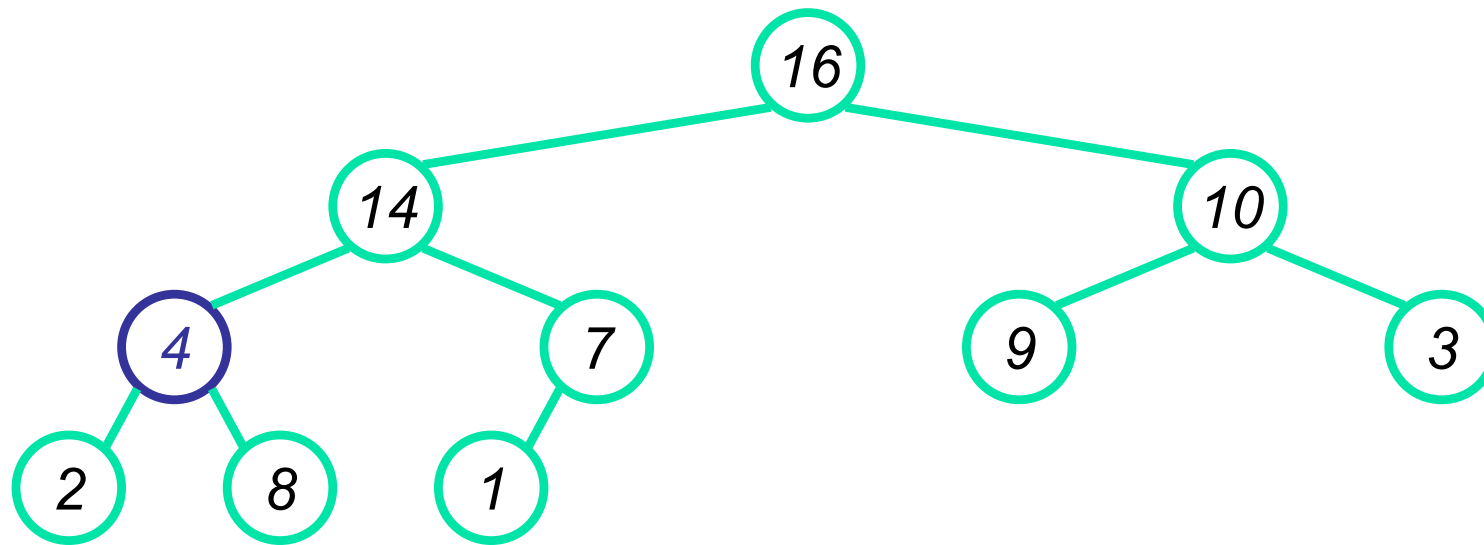
MAX-HEAPIFY Example



$A =$

| | | | | | | | | | |
|----|----|----|---|---|---|---|---|---|---|
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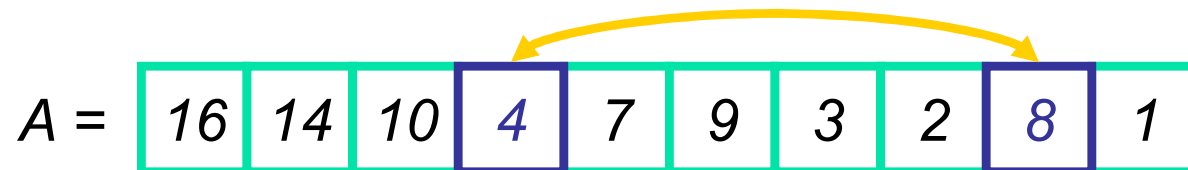
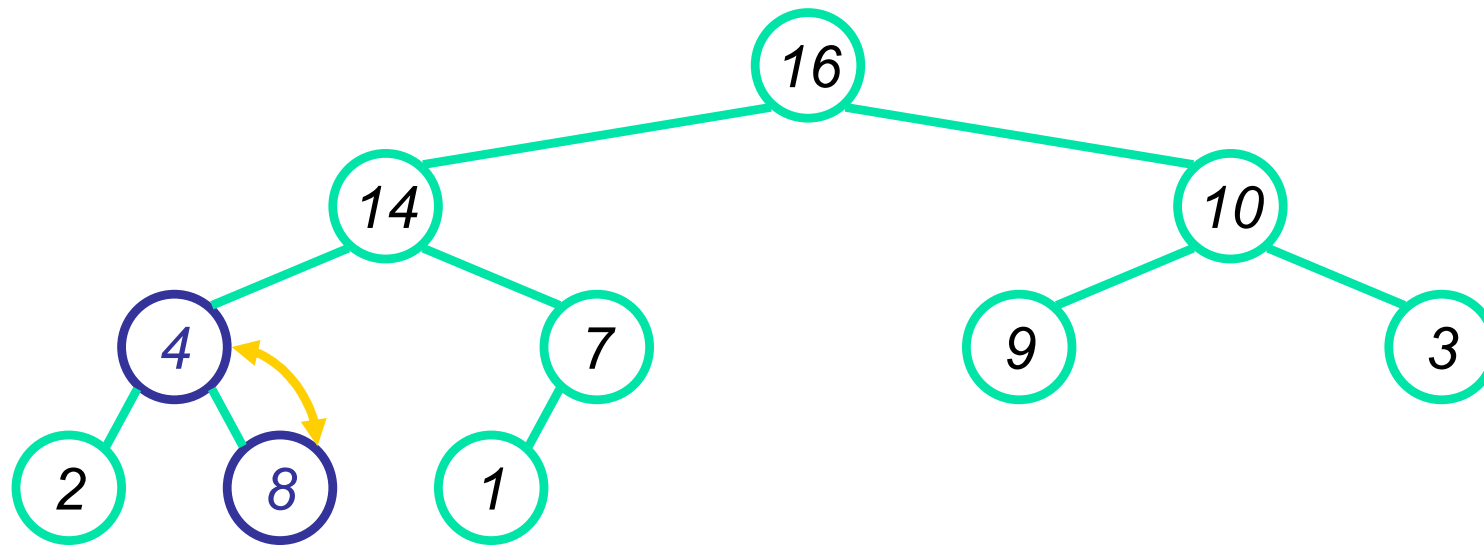
MAX-HEAPIFY Example



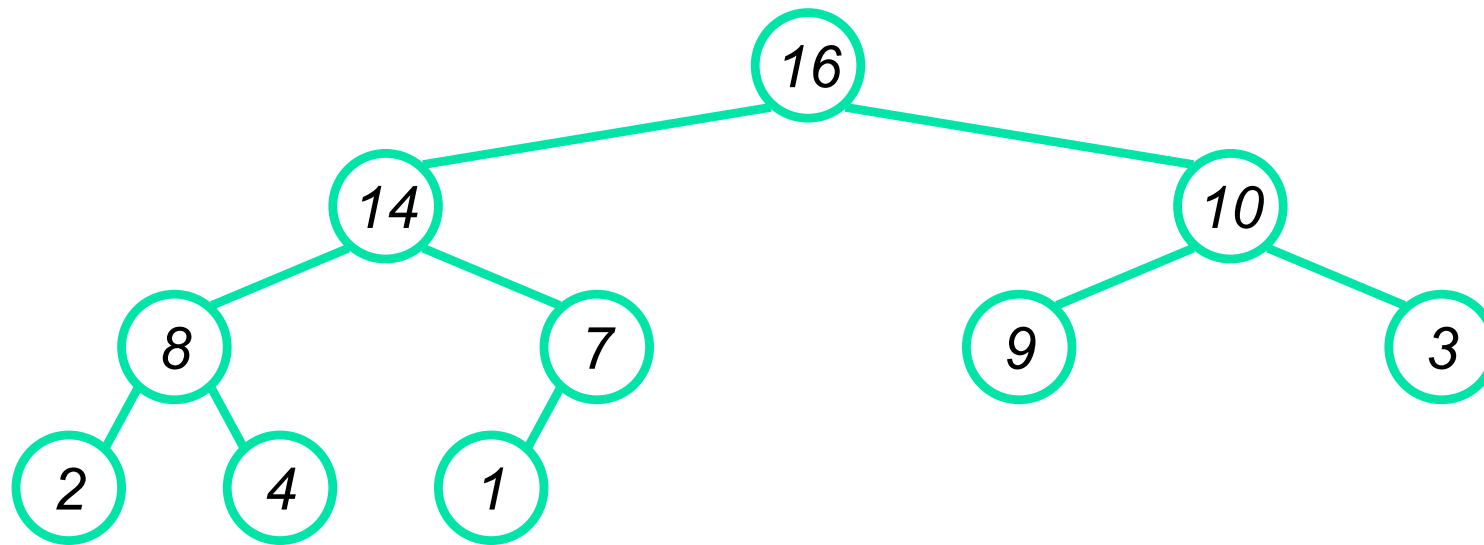
$A =$

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

MAX-HEAPIFY Example



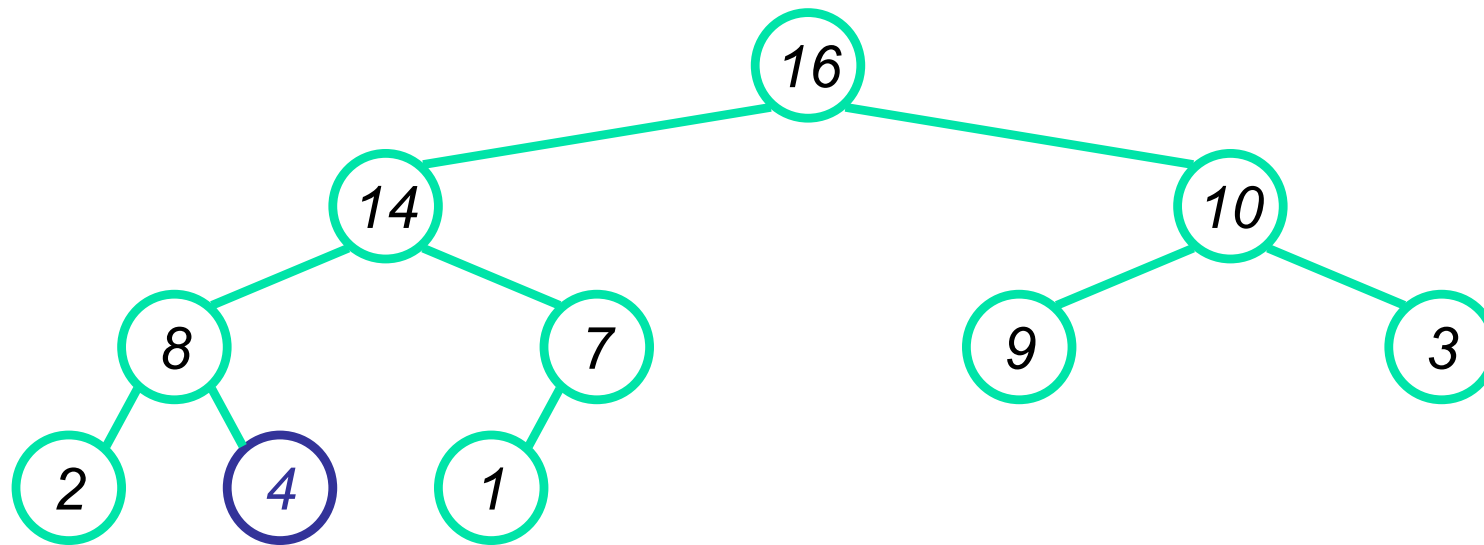
MAX-HEAPIFY Example



$A =$

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

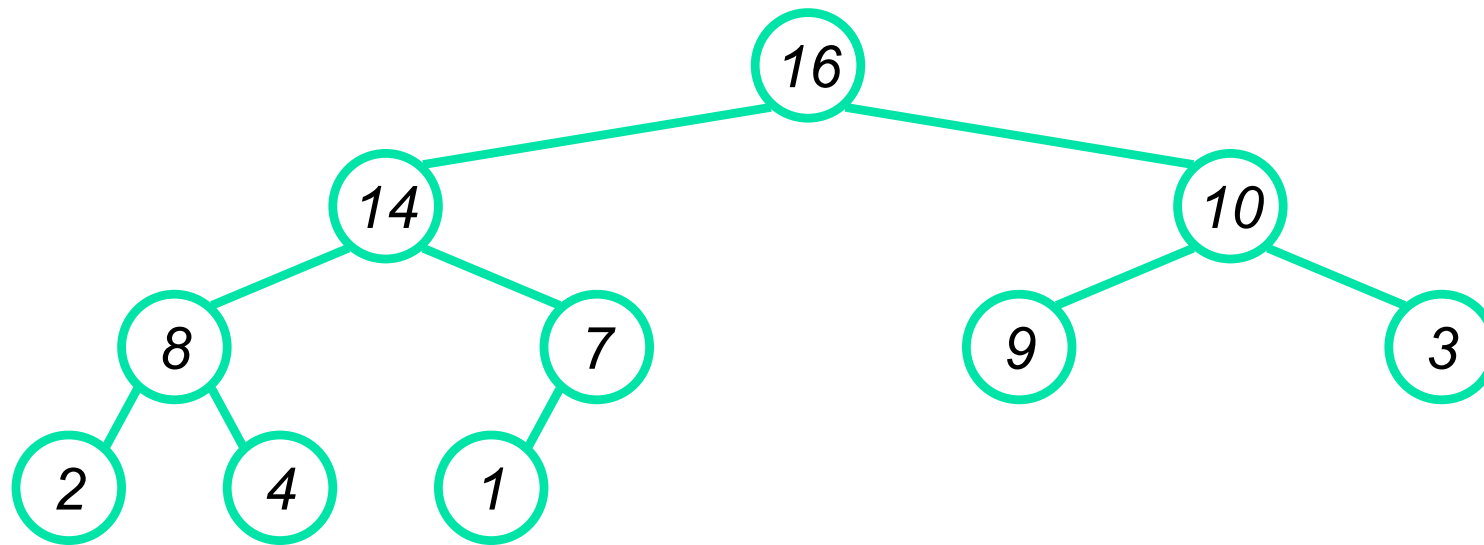
MAX-HEAPIFY Example



$A =$

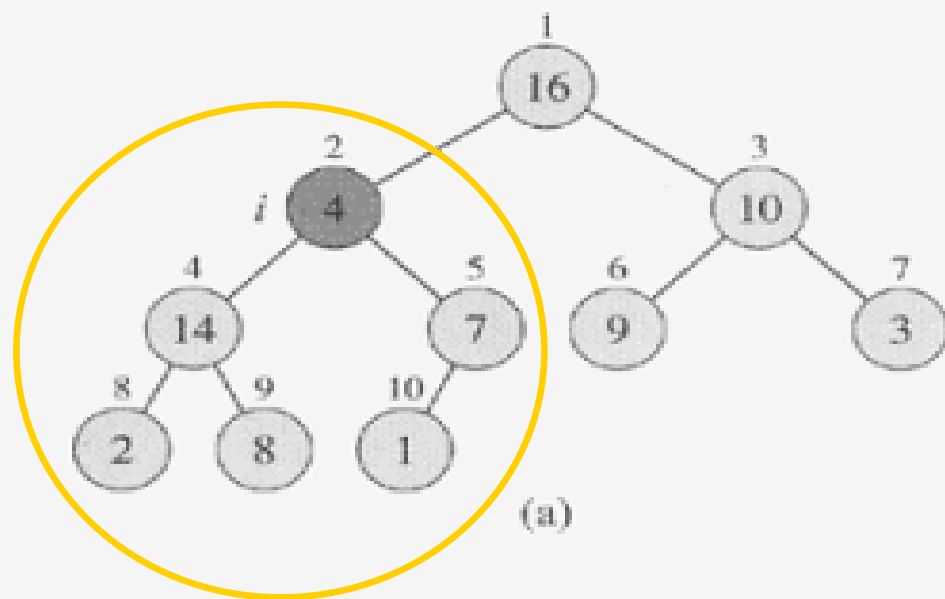
| | | | | | | | | | |
|----|----|----|---|---|---|---|---|---|---|
| 16 | 14 | 10 | 8 | 7 | 9 | 3 | 2 | 4 | 1 |
|----|----|----|---|---|---|---|---|---|---|

MAX-HEAPIFY Example

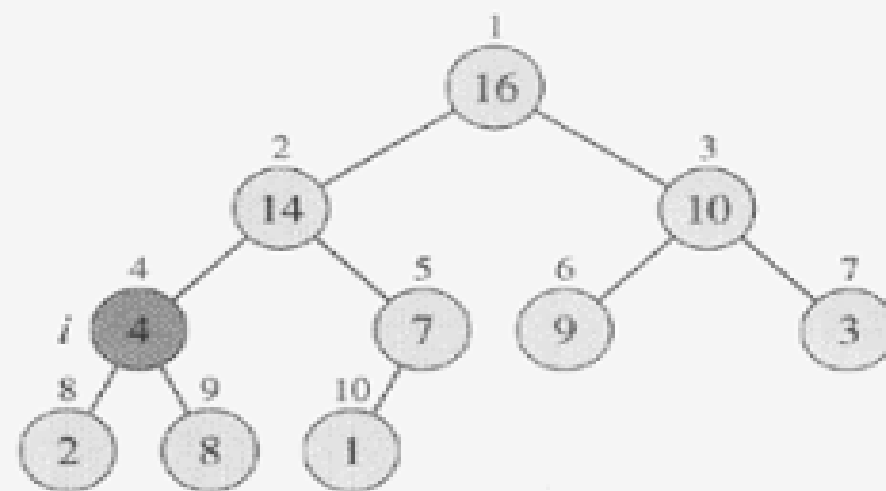


$A =$

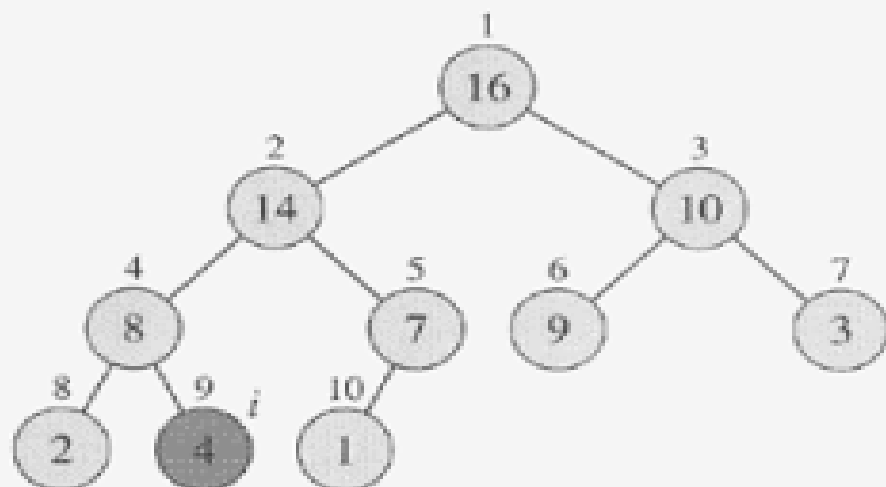
| | | | | | | | | | |
|----|----|----|---|---|---|---|---|---|---|
| 16 | 14 | 10 | 8 | 7 | 9 | 3 | 2 | 4 | 1 |
|----|----|----|---|---|---|---|---|---|---|



(a)



(b)



(c)

Example of MAX-HEAPIFY

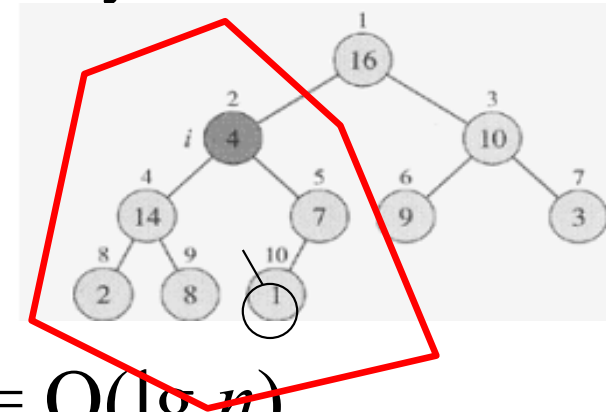
Figure 6.2 The action of $\text{MAX-HEAPIFY}(A, 2)$, where $\text{heap-size}[A] = 10$. (a) The initial configuration, with $A[2]$ at node $i = 2$ violating the max-heap property since it is not larger than both children. The max-heap property is restored for node 2 in (b) by exchanging $A[2]$ with $A[4]$, which destroys the max-heap property for node 4. The recursive call $\text{MAX-HEAPIFY}(A, 4)$ now has $i = 4$. After swapping $A[4]$ with $A[9]$, as shown in (c), node 4 is fixed up, and the recursive call $\text{MAX-HEAPIFY}(A, 9)$ yields no further change to the data structure.

Analyzing MAX-HEAPIFY (1/2)

- $\Theta(1)$ to find out the largest among $A[i]$, $A[\text{LEFT}(i)]$, and $A[\text{RIGHT}(i)]$
- Plus the time to run MAX-HEAPIFY on a subtree rooted at one of the children of node i
 - The children's subtrees each have size at most $2n/3$ – the worst case occurs when the last row of the tree is exactly half full

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

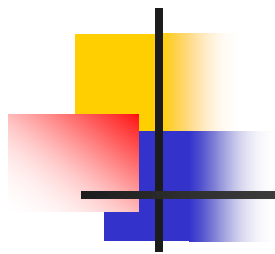
- By case 2 of the master theorem: $T(n) = O(\lg n)$





Analyzing MAX-HEAPIFY (2/2)

- Alternately, Heapify takes $T(n) = \Theta(h)$
 - $h = \text{height of heap} = \lg n$
 - $T(n) = \Theta(\lg n)$



Building A Heap



Build-Max-Heap (1/2)

- We can build a heap in a bottom-up manner by running Build-Max-Heap() on successive subarrays
 - Fact: for array of length n , all elements in range $A[\lfloor n/2 \rfloor + 1, \lfloor n/2 \rfloor + 2 .. n]$ are heaps (*Why?*)
 - These elements are **leaves**, they do not have children
 - We also know that the leave-level has at most 2^h nodes = $\lceil n/2 \rceil$ nodes
 - and other levels have a total of $\lfloor n/2 \rfloor$ nodes
- Walk backwards through the array from $n/2$ to 1, calling Build-Max-Heap() on each node.



Build-Max-Heap (2/2)

// given an unsorted array A, make A a heap

```
BUILD-MAX-HEAP(A)
```

```
1  heap-size[A]  $\leftarrow$  length[A]
```

```
2  for  $i \leftarrow \lfloor \text{length}[A]/2 \rfloor$  downto 1
```

```
3      do MAX-HEAPIFY(A,  $i$ )
```

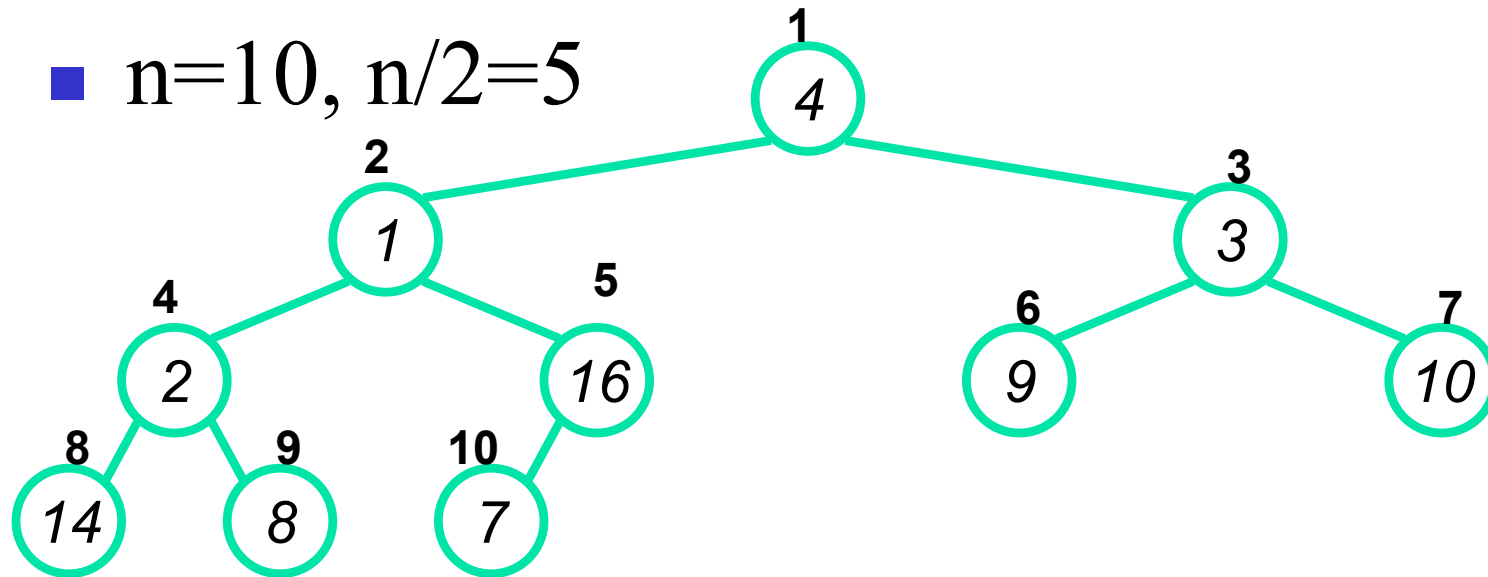
- The **Build-Max-Heap** () procedure, which runs in linear time, produces a *max-heap* from an unsorted input array.
- However, the **MAX-HEAPIFY**() procedure, which runs in $O(\lg n)$ time, is the key to maintaining the heap property.

Build-Max-Heap Example

- Work through example

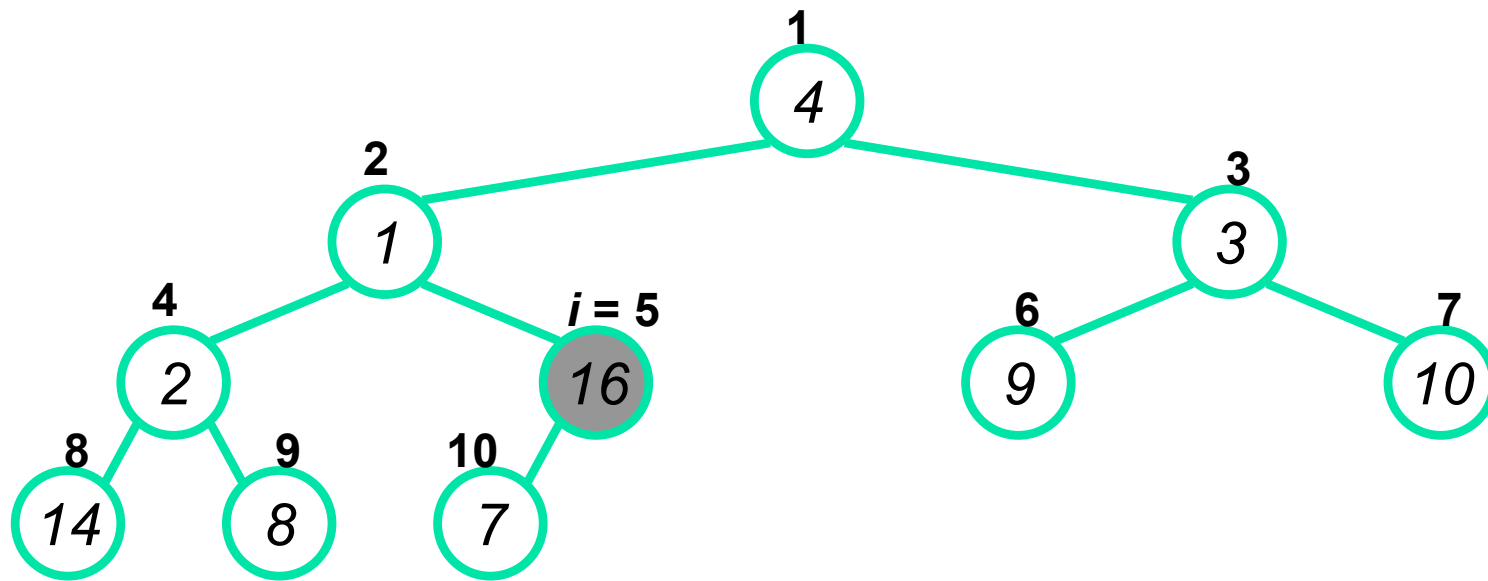
$A = \{4, 1, 3, 2, 16, 9, 10, 14, 8, 7\}$

- $n=10, n/2=5$



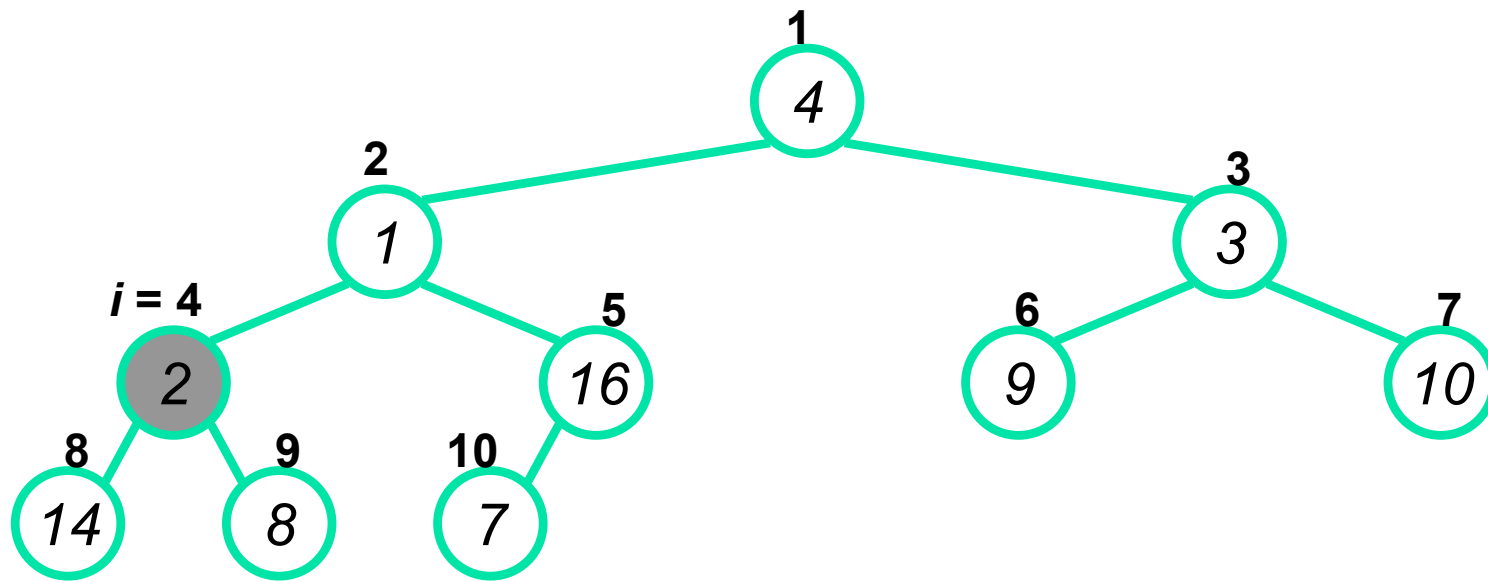
Build-Max-Heap Example

- $A = \{4, 1, 3, 2, 16, 9, 10, 14, 8, 7\}$



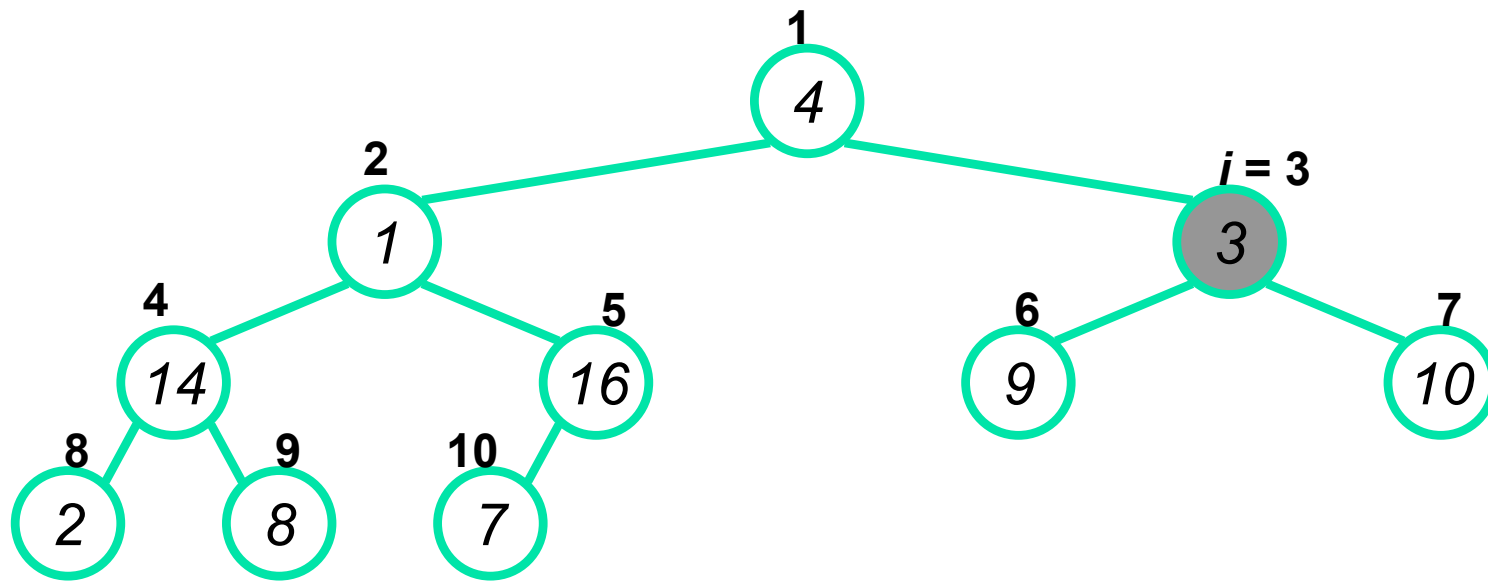
Build-Max-Heap Example

- $A = \{4, 1, 3, 2, 16, 9, 10, 14, 8, 7\}$



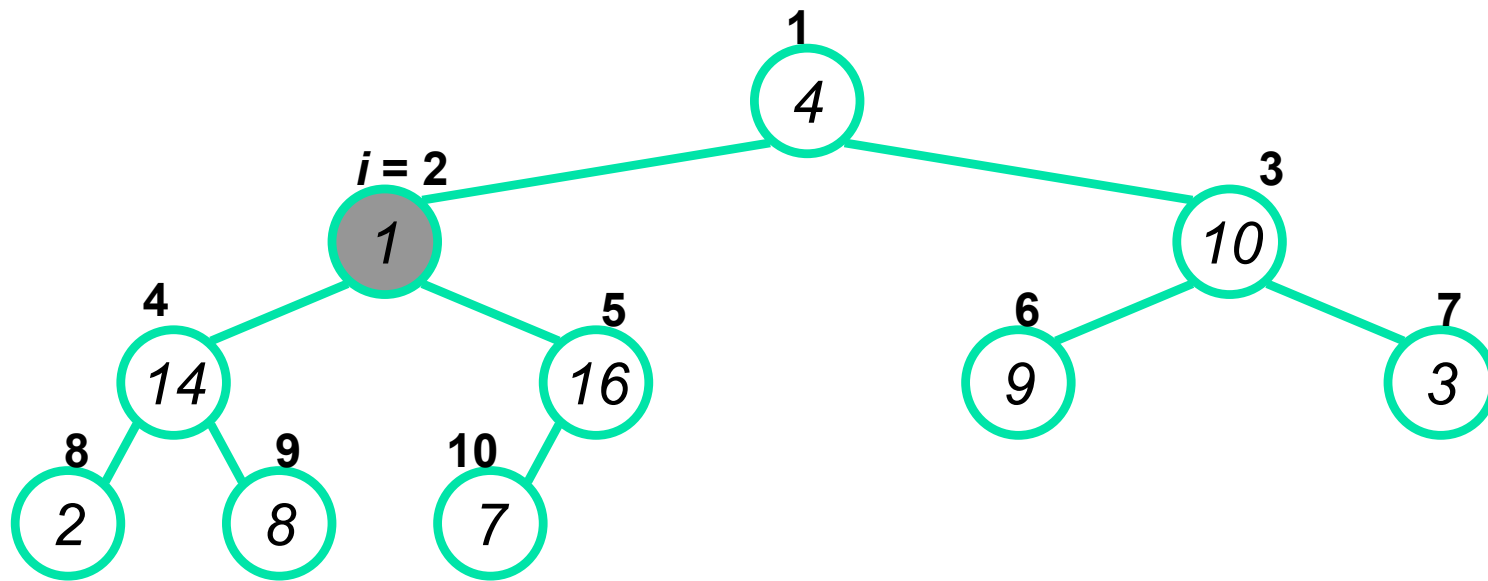
Build-Max-Heap Example

- $A = \{4, 1, 3, 14, 16, 9, 10, 2, 8, 7\}$



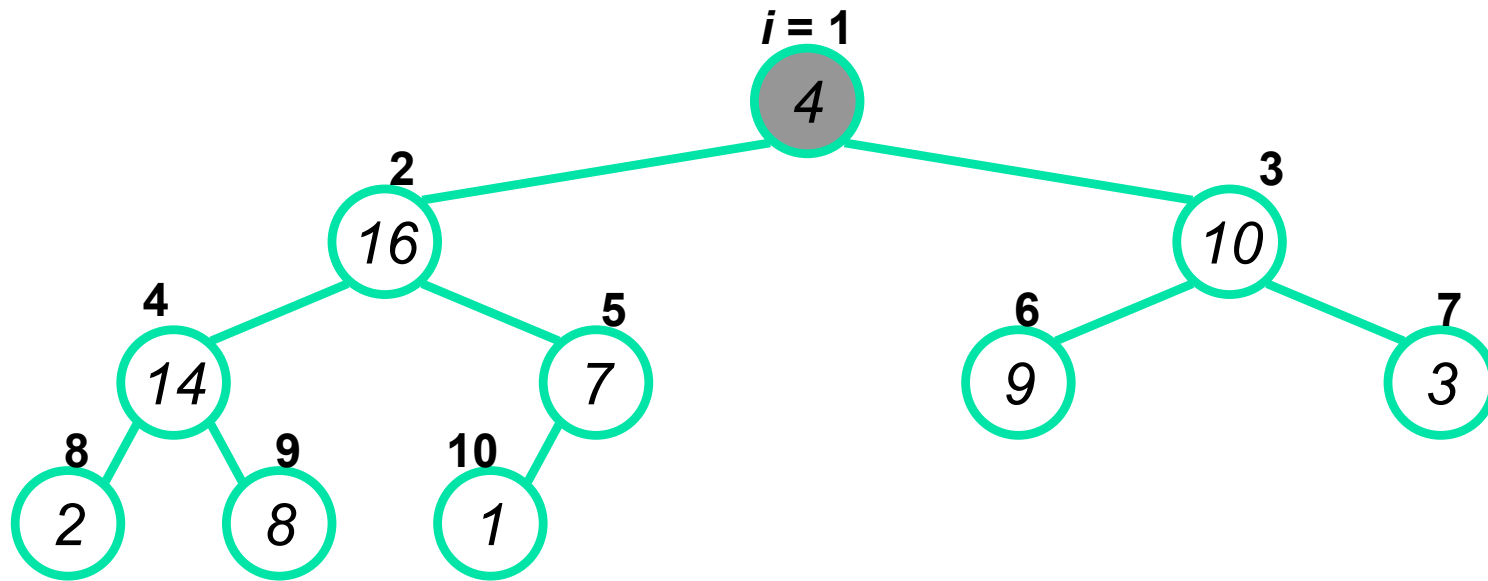
Build-Max-Heap Example

- $A = \{4, 1, \mathbf{10}, 14, 16, 9, \mathbf{3}, 2, 8, 7\}$



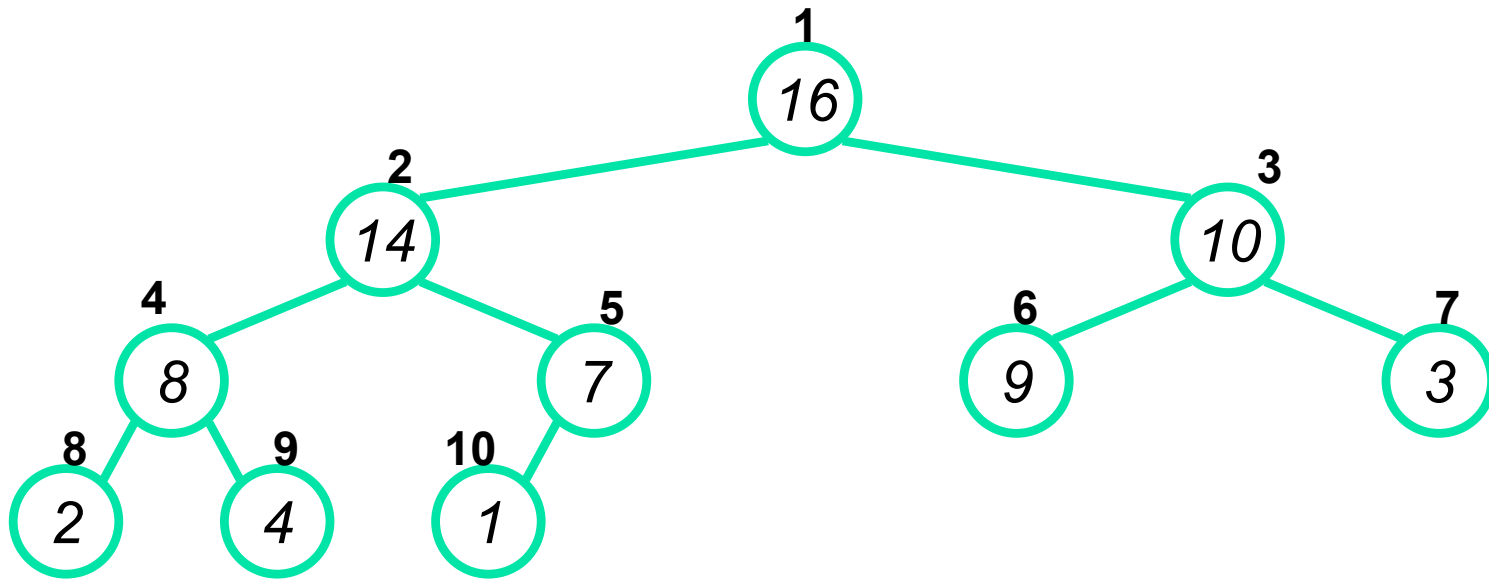
Build-Max-Heap Example

- $A = \{4, 16, 10, 14, 7, 9, 3, 2, 8, 1\}$

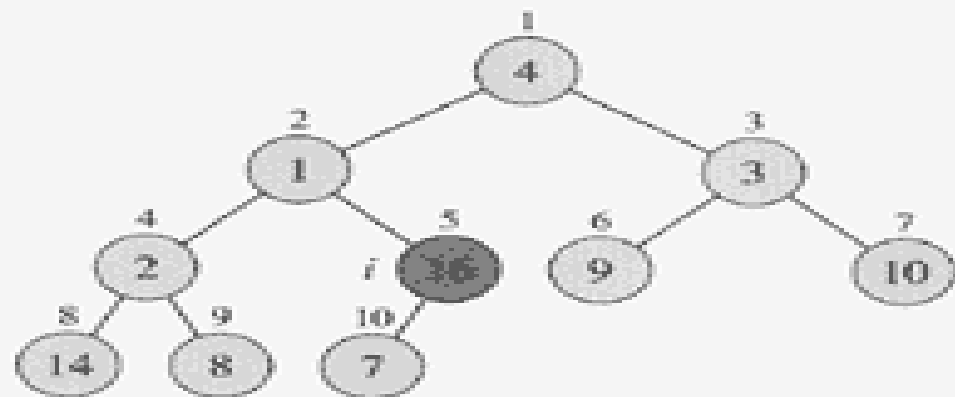


Build-Max-Heap Example

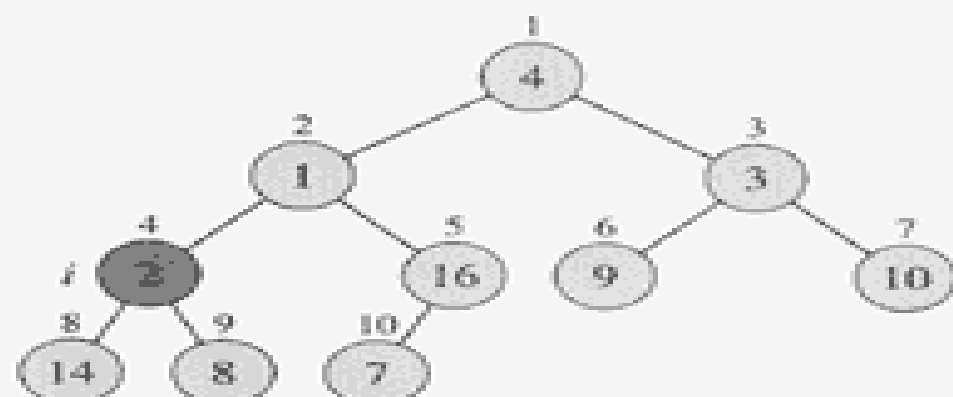
- $A = \{16, 14, 10, 8, 7, 9, 3, 2, 4, 1\}$



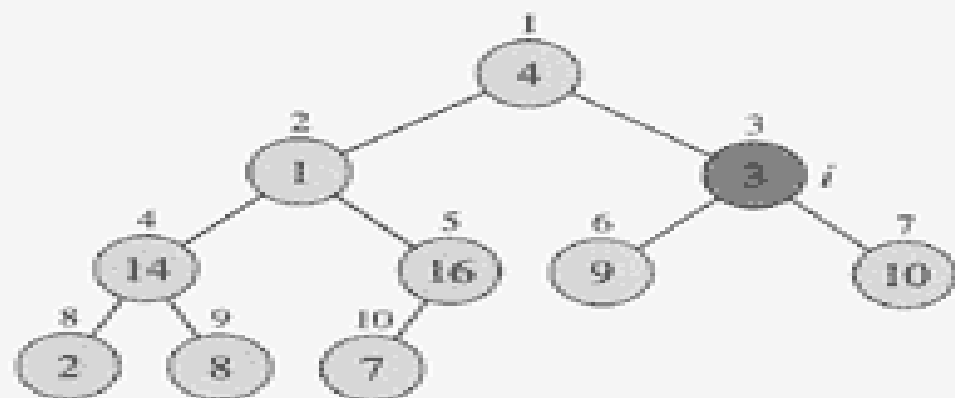
$A = [4, 1, 3, 2, 16, 9, 10, 14, 8, 7]$



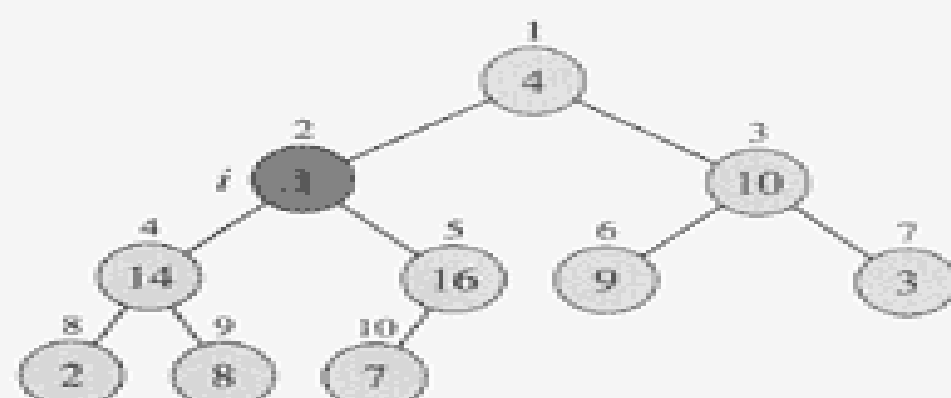
(a)



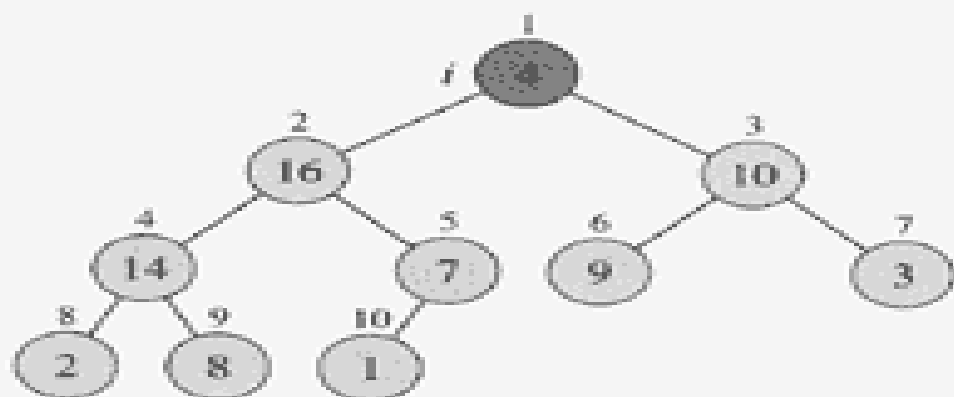
(b)



(c)

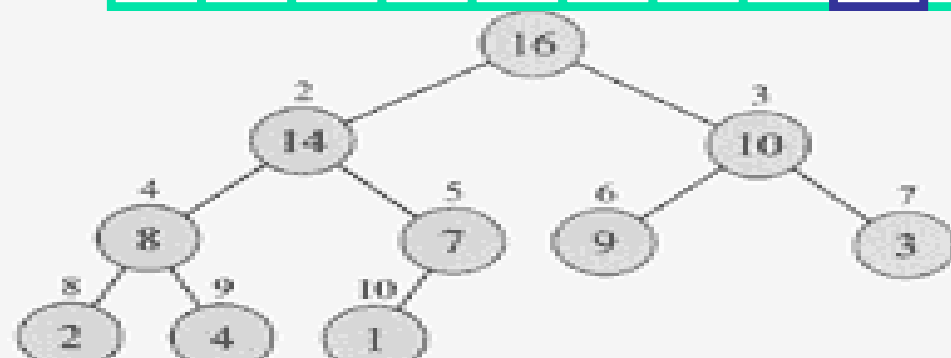


(d)



(e)

$A = [16, 14, 10, 8, 7, 9, 3, 2, 4, 1]$



(f)



Analyzing Build-Max-Heap (1/2)

- Each call to **MAX-HEAPIFY** takes $O(\lg n)$ time
- There are $O(n)$ such calls (specifically, $\lfloor n/2 \rfloor$)
- Thus the running time is $O(n \lg n)$
 - *Is this a correct asymptotic upper bound?*
 - **YES**
 - *Is this an asymptotically **tight** bound?*
 - **NO**
- A tighter bound is $O(n)$
 - *How can this be? Is there a flaw in the above reasoning?*
 - *We can derive a tighter bound by observing that the time for MAX-HEAPIFY to run at a node **varies** with the height of the node in the tree, and the heights of most nodes are small.*
- Fact: an n -element heap has at most 2^{h-k} nodes of level k , where h is the height of the tree.



Analyzing Build-Max-Heap (2/2)

- The time required by *MAX-HEAPIFY* on a node of height k is $O(k)$. So we can express the total cost of *Build-Max-Heap* as

$$\begin{aligned}\sum_{k=0 \text{ to } h} 2^{h-k} O(k) &= O(2^h \sum_{k=0 \text{ to } h} k/2^k) \\ &= O(n \sum_{k=0 \text{ to } h} k(1/2)^k)\end{aligned}$$

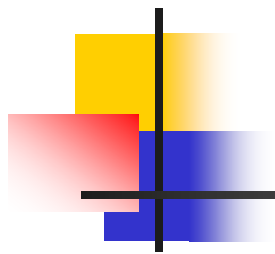
From: $\sum_{k=0 \text{ to } \infty} k x^k = x/(1-x)^2$ where $x = 1/2$

So, $\sum_{k=0 \text{ to } \infty} k/2^k = (1/2)/(1 - 1/2)^2 = 2$

Therefore, $O(n \sum_{k=0 \text{ to } h} k/2^k) = O(n)$

- So, we can bound the running time for building a heap from an unordered array in linear time

IA, P-135



The HeapSort Algorithm



Heapsort

- Given Build-Max-Heap an **in-place** sorting algorithm is easily constructed:
 - Maximum element is at $A[1]$
 - Discard by swapping with element at $A[n]$
 - Decrement $\text{heap_size}[A]$
 - $A[n]$ now contains correct value
 - Restore heap property at $A[1]$ by calling MAX-HEAPIFY
 - Repeat, always swapping $A[1]$ for $A[\text{heap_size}(A)]$



Heapsort

HEAPSORT(A)

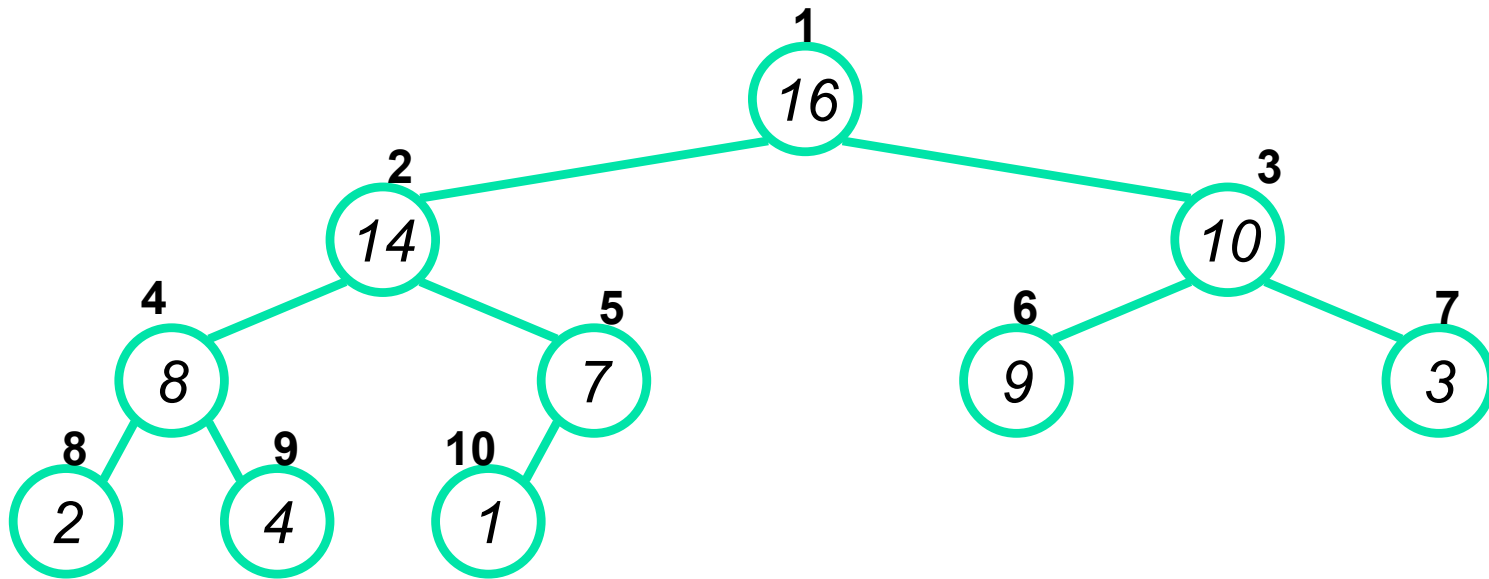
{

1. Build-MAX-Heap(A)
2. **for** $i \leftarrow \text{length}[A]$ downto 2
3. **do** exchange $A[1] \leftrightarrow A[i]$
4. $\text{heap-size}[A] \leftarrow \text{heap-size}[A] - 1$
5. MAX-Heapify($A, 1$)

}

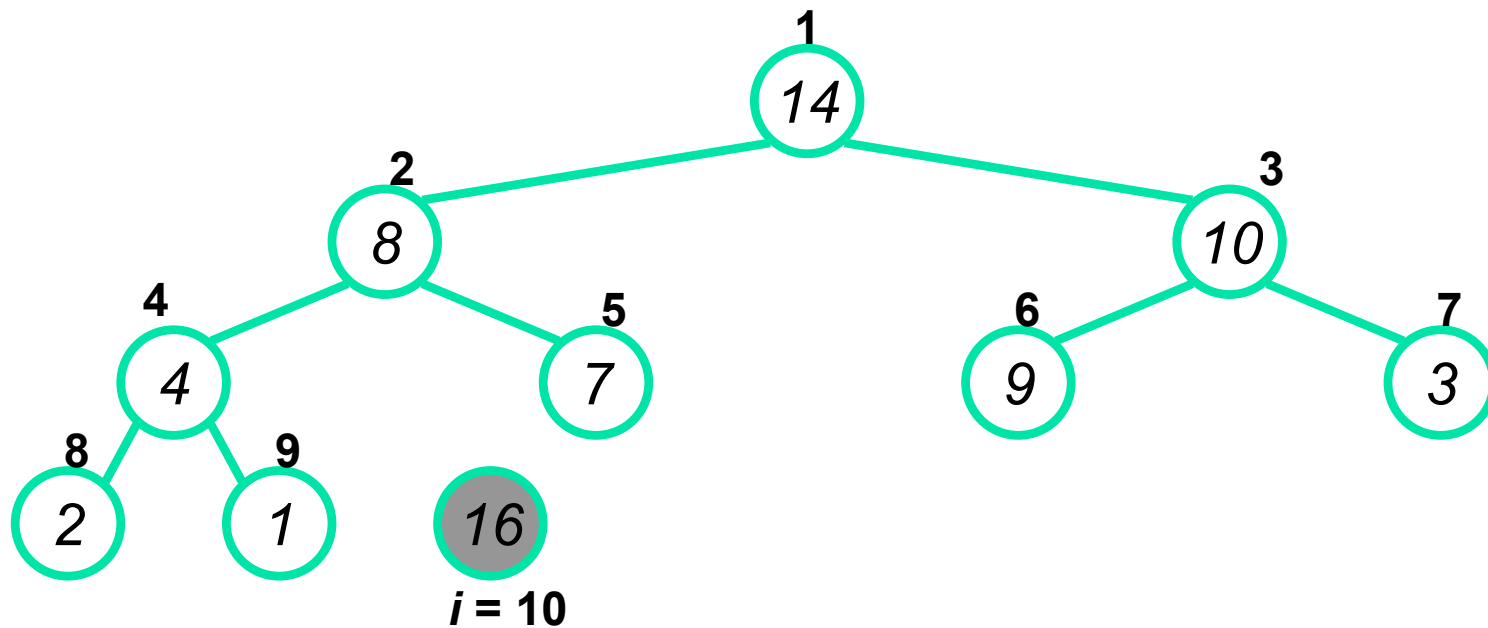
HeapSort Example

- $A = \{16, 14, 10, 8, 7, 9, 3, 2, 4, 1\}$



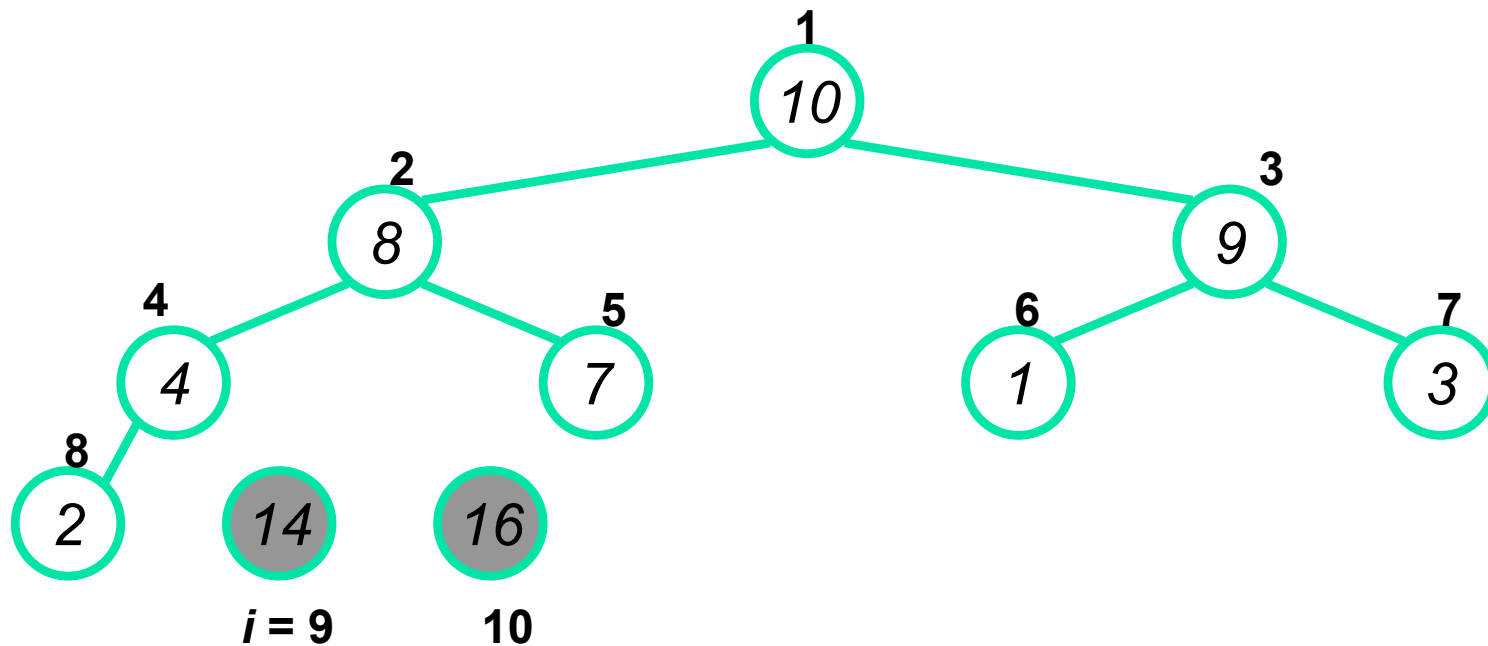
HeapSort Example

- $A = \{14, 8, 10, 4, 7, 9, 3, 2, 1, \mathbf{16}\}$



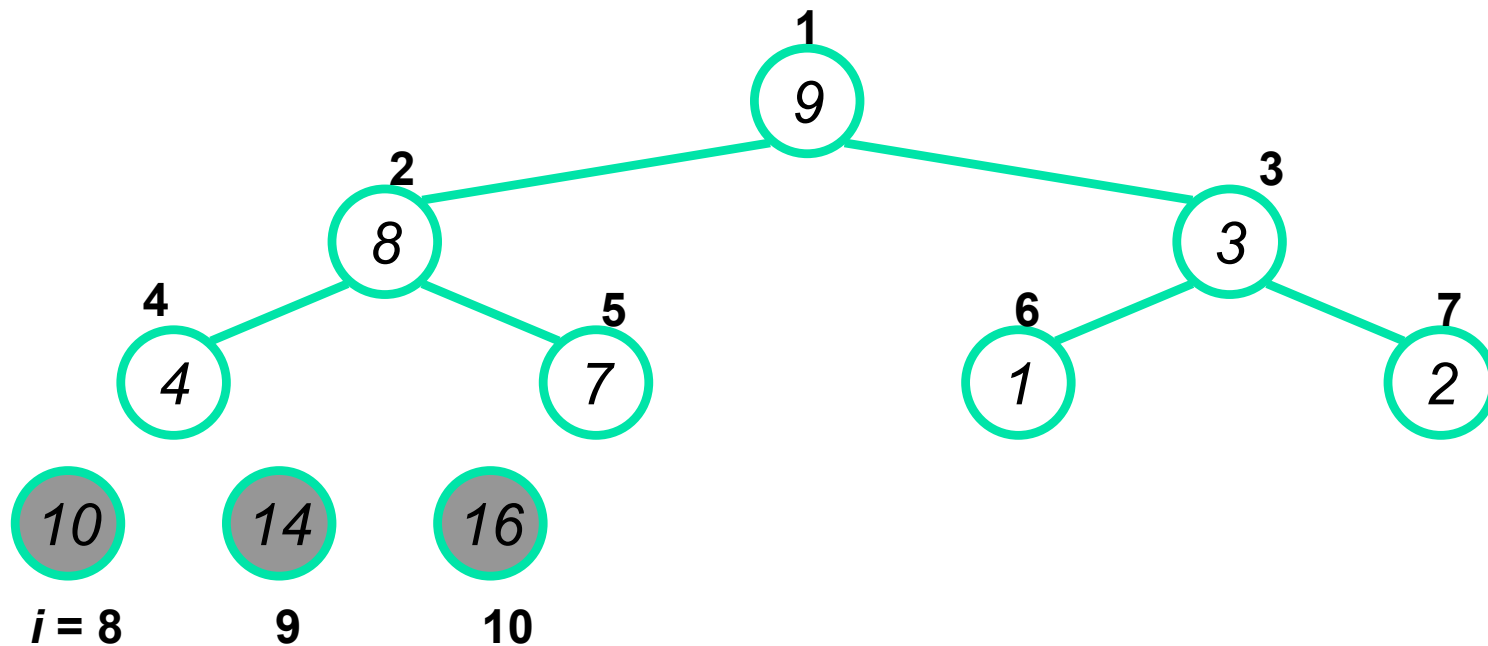
HeapSort Example

- $A = \{10, 8, 9, 4, 7, 1, 3, 2, \mathbf{14}, \mathbf{16}\}$



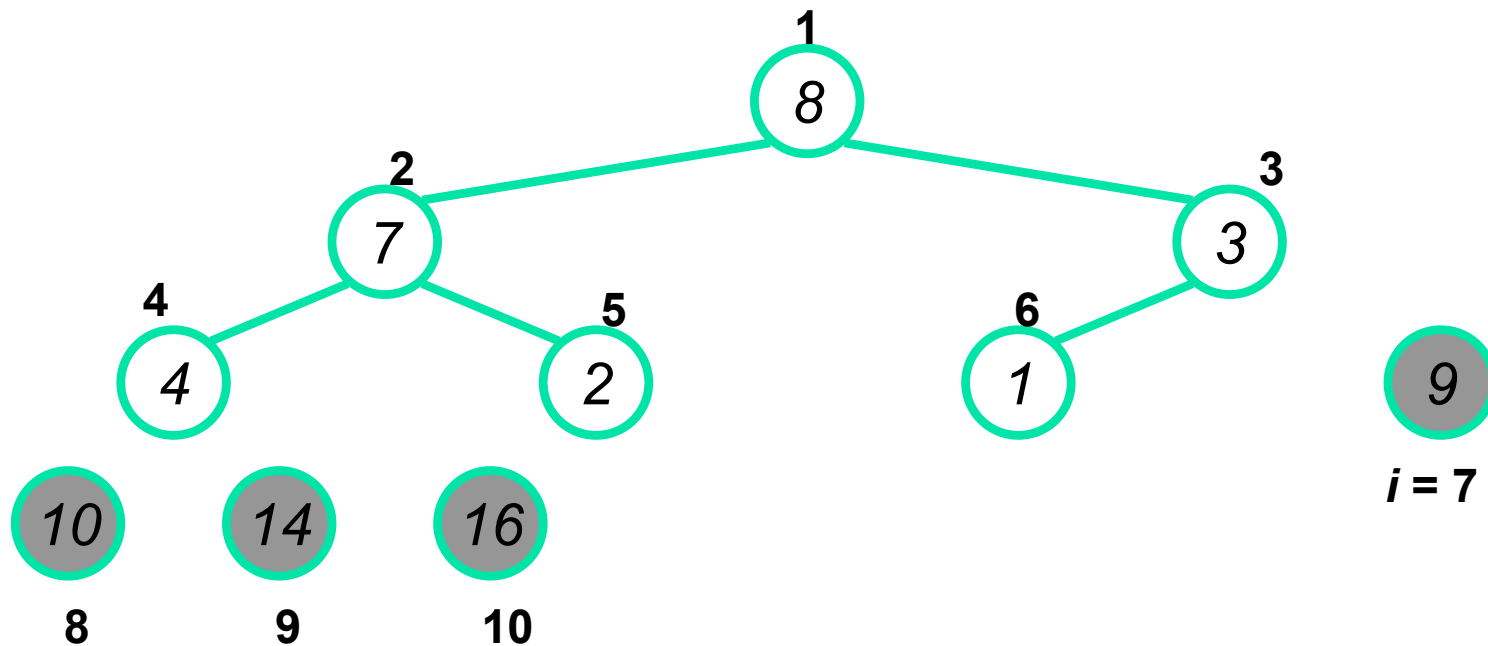
HeapSort Example

- $A = \{9, 8, 3, 4, 7, 1, 2, \mathbf{10}, \mathbf{14}, \mathbf{16}\}$



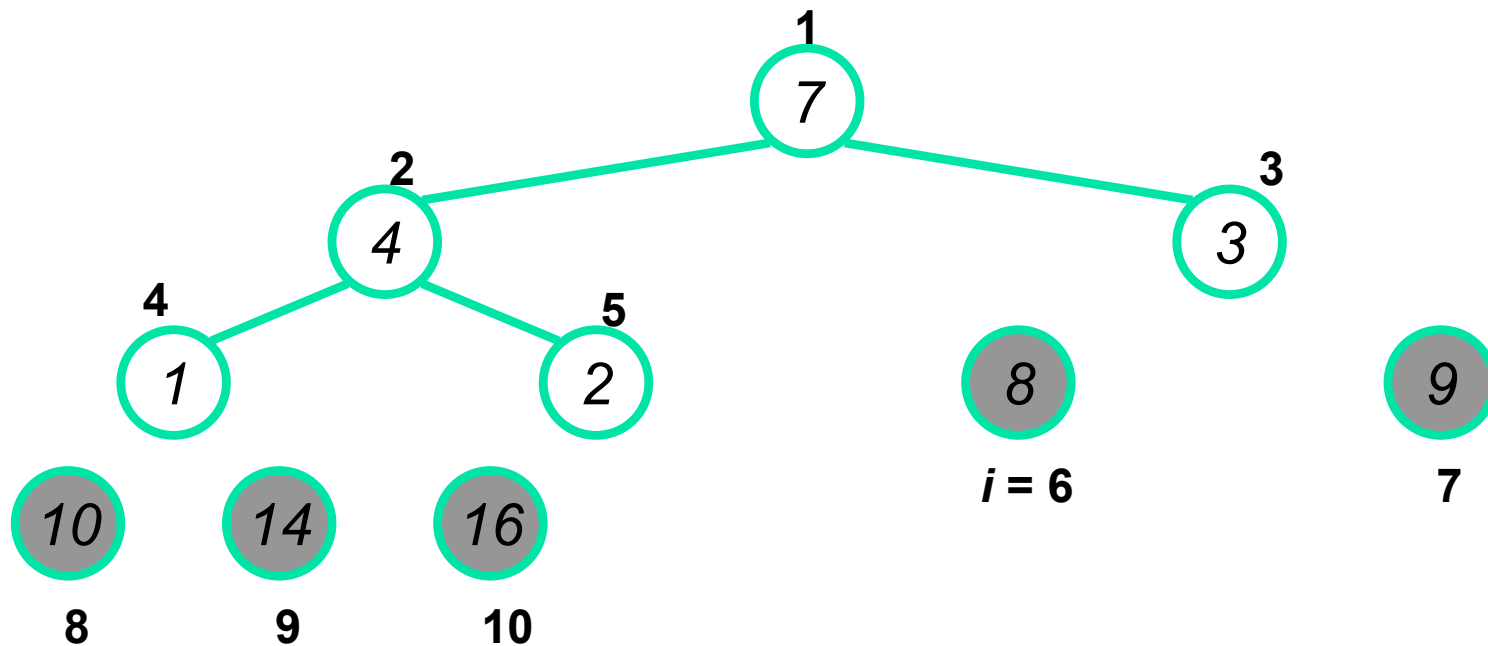
HeapSort Example

- $A = \{8, 7, 3, 4, 2, 1, 9, 10, 14, 16\}$



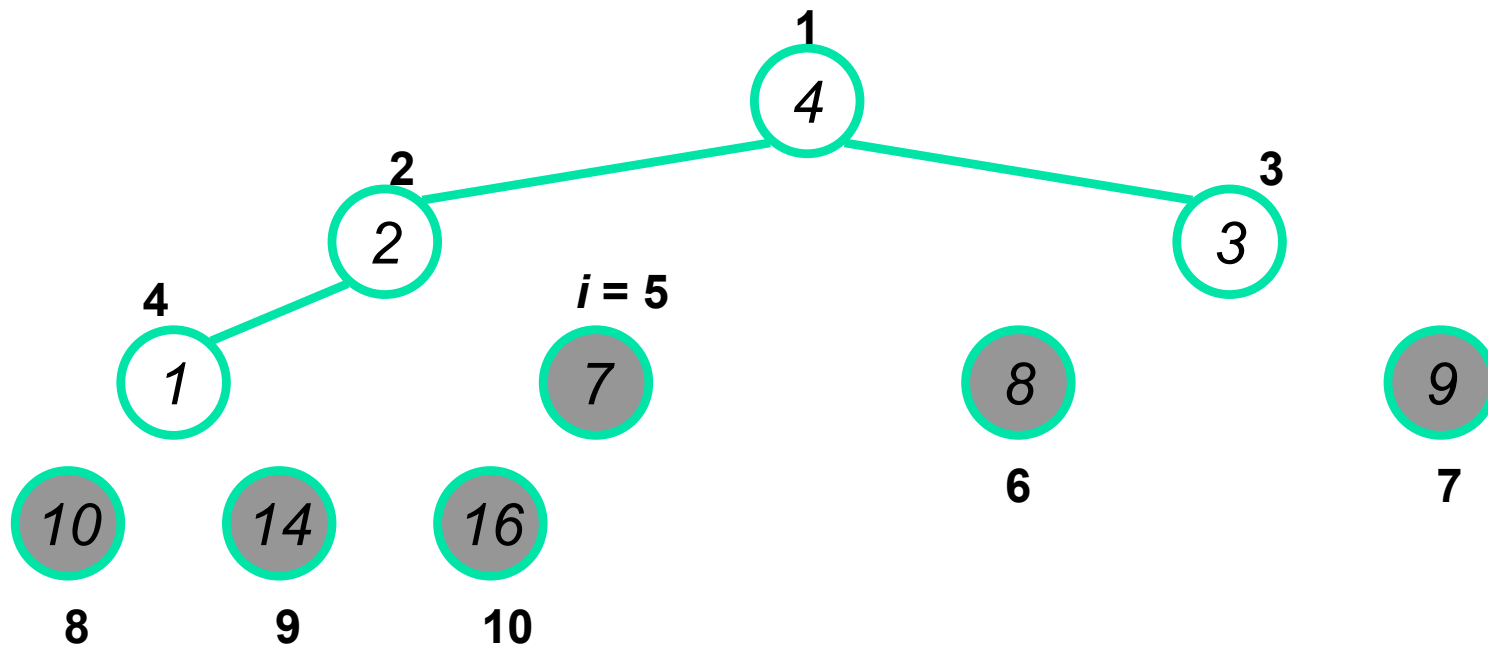
HeapSort Example

- $A = \{7, 4, 3, 1, 2, 8, 9, 10, 14, 16\}$



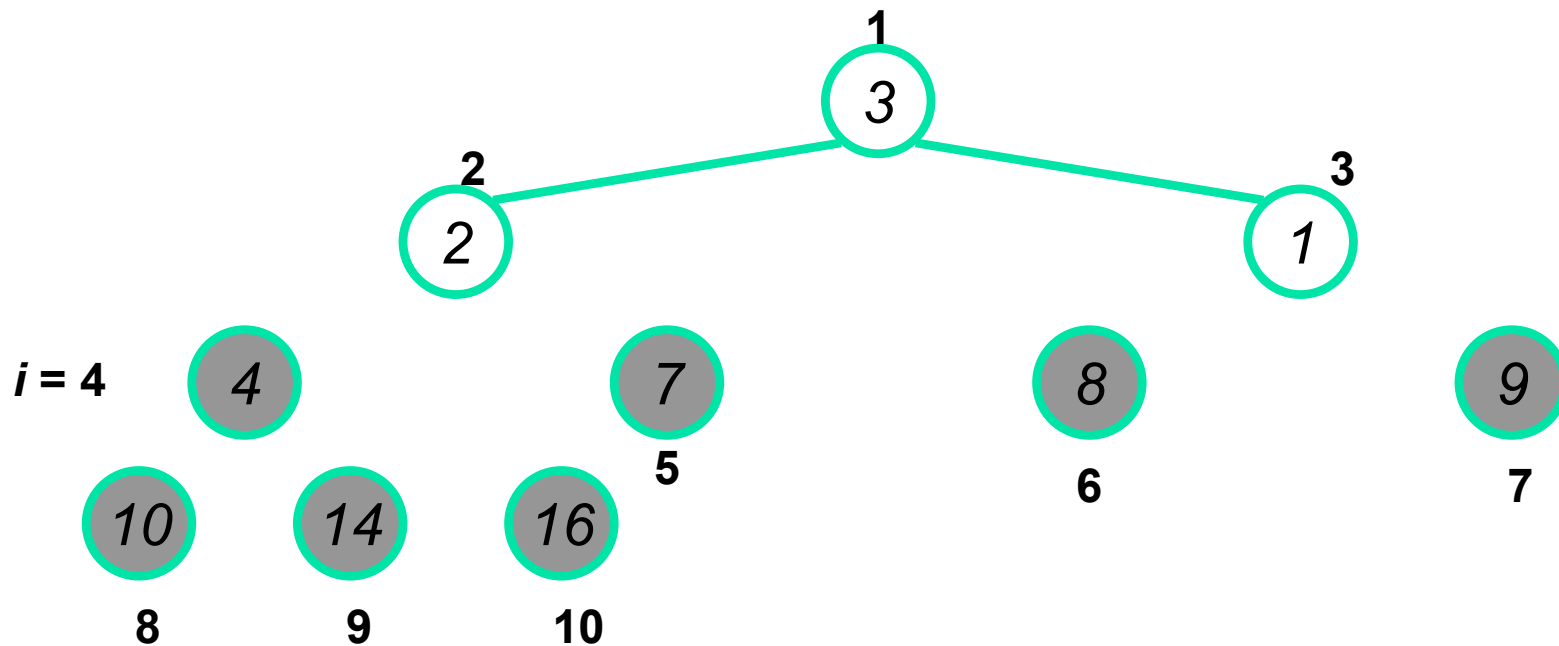
HeapSort Example

- $A = \{4, 2, 3, 1, 7, 8, 9, 10, 14, 16\}$



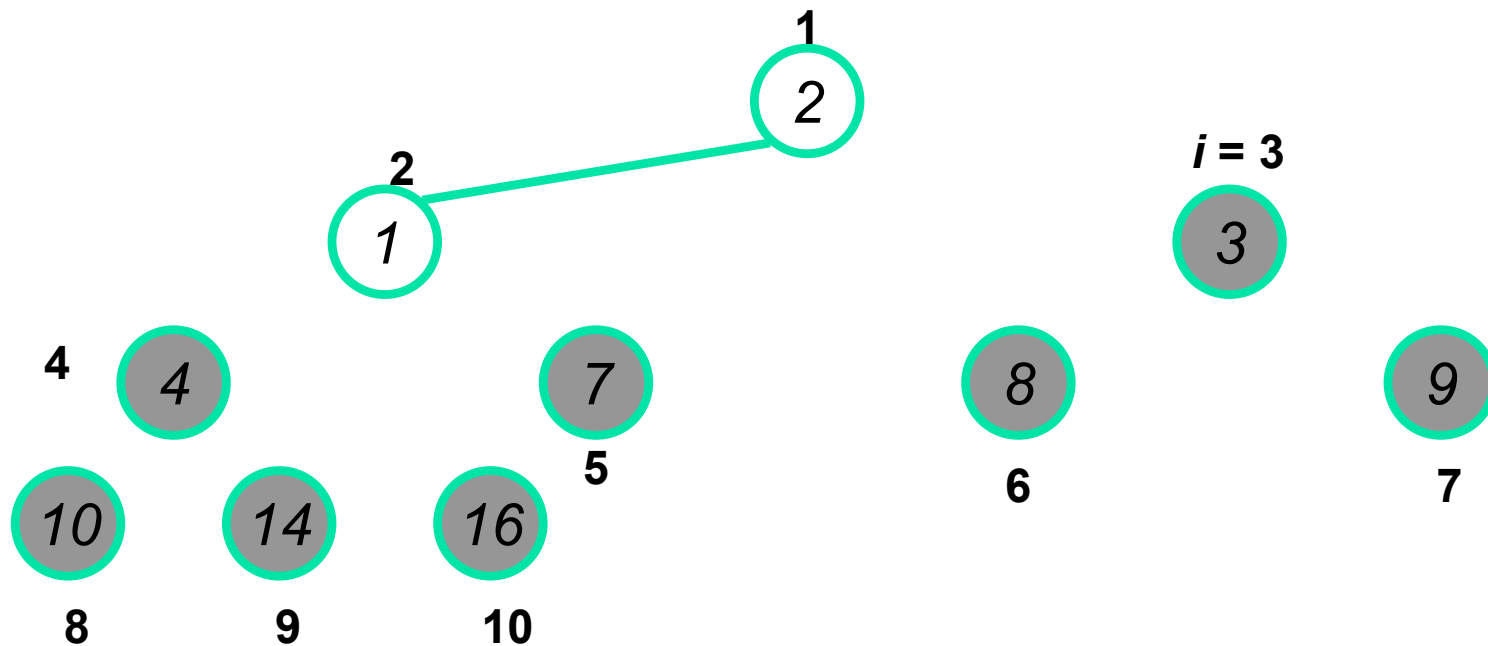
HeapSort Example

- $A = \{3, 2, 1, 4, 7, 8, 9, 10, 14, 16\}$



HeapSort Example

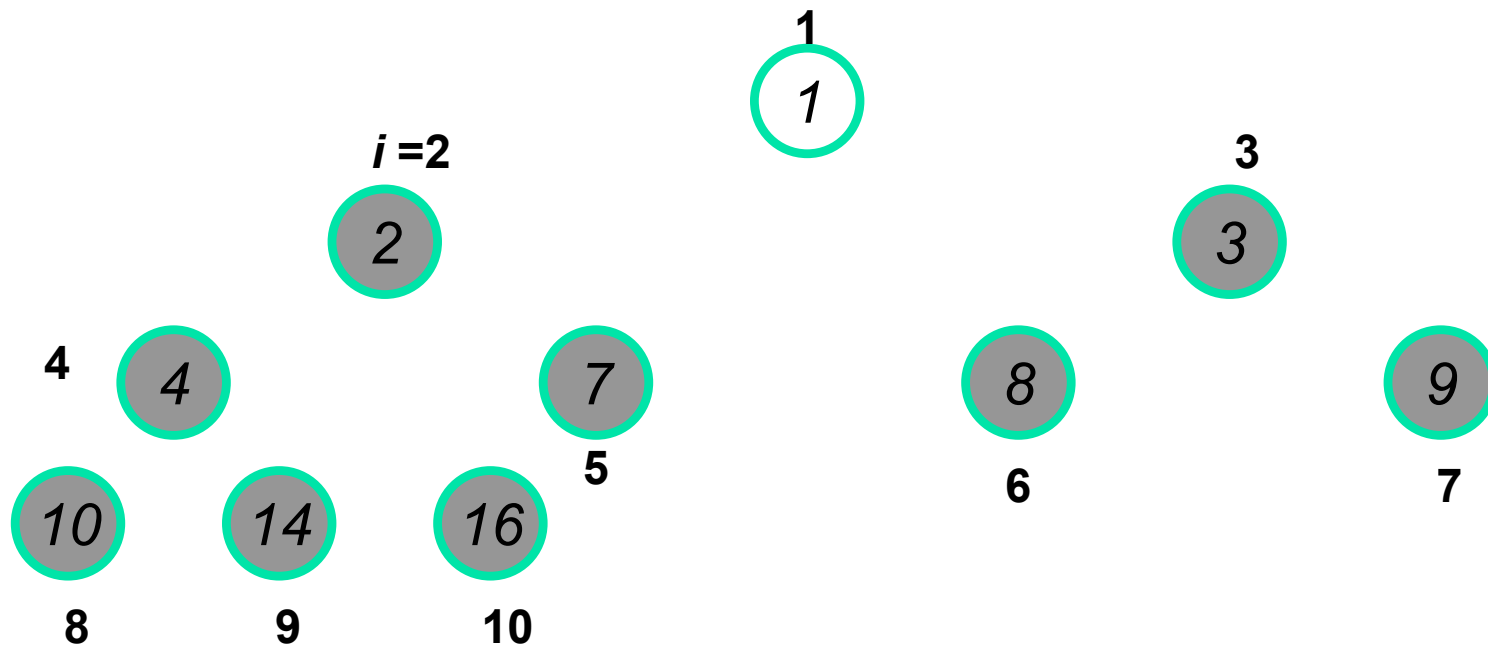
- $A = \{2, 1, 3, 4, 7, 8, 9, 10, 14, 16\}$





HeapSort Example

- $A = \{1, 2, 3, 4, 7, 8, 9, 10, 14, 16\}$





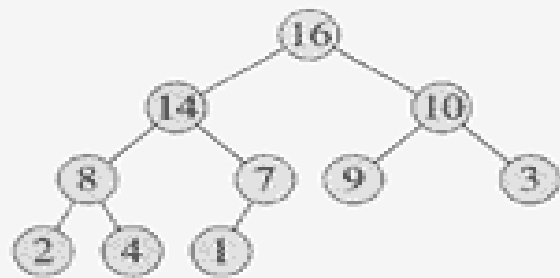
Analyzing Heapsort (1/2)

- The call to **BUILD-MAX-HEAP ()** takes $O(n)$ time
- Each of the $n - 1$ calls to **MAX-HEAPIFY()** takes $O(\lg n)$ time
- Thus the total time taken by **HEAPSORT ()**
$$= O(n) + (n - 1) O(\lg n)$$
$$= O(n) + O(n \lg n)$$
$$= O(n \lg n)$$

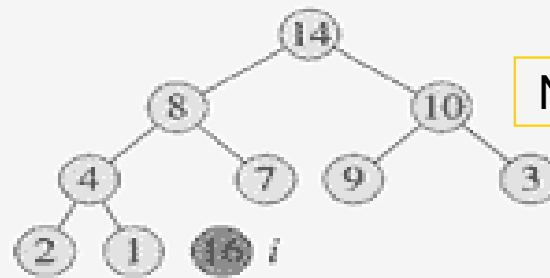


Analyzing Heapsort (2/2)

- The $O(n \log n)$ run time of heap-sort is much better than the $O(n^2)$ run time of selection and insertion sort
- Although, it has the same run time as Merge sort, but it is better than Merge Sort regarding memory space
 - Heap sort is **in-place** sorting algorithm
 - But **not stable**
 - **Does not preserve the relative order of elements with equal keys**
 - **Sorting algorithm (stable) if 2 records with same key stay in original order**

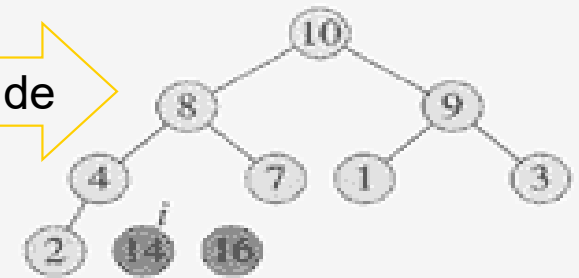


(a)

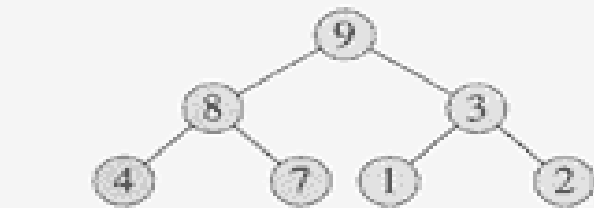


(b)

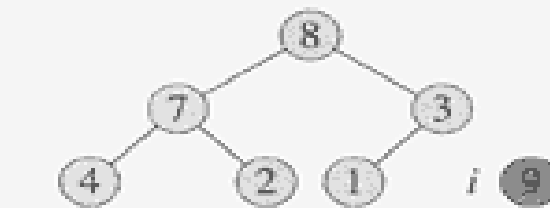
Next Slide



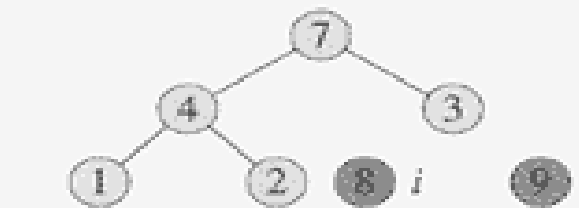
(c)



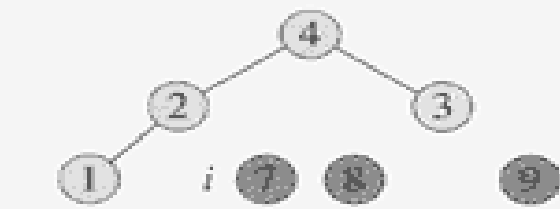
(d)



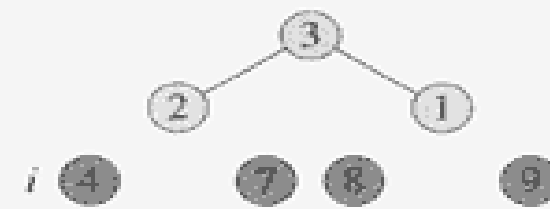
(e)



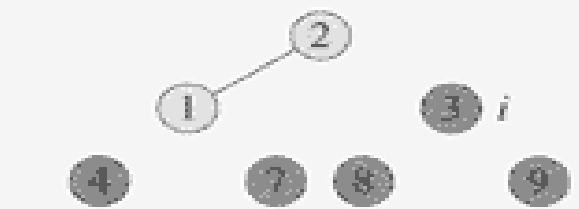
(f)



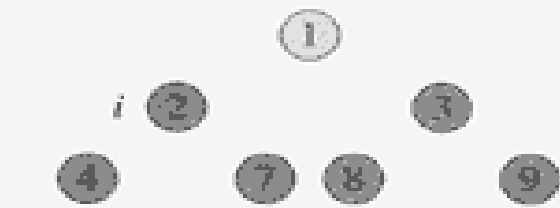
(g)



(h)



(i)



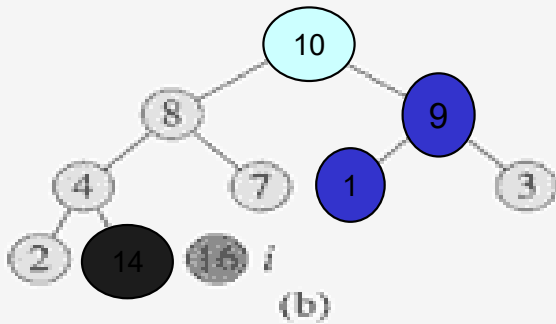
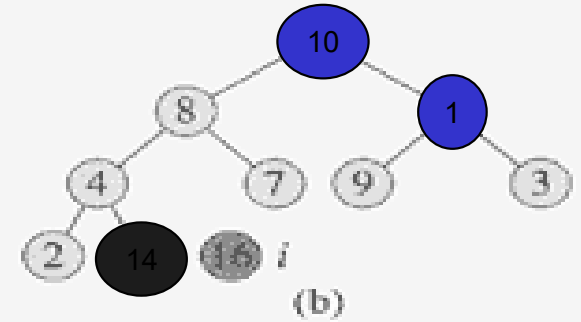
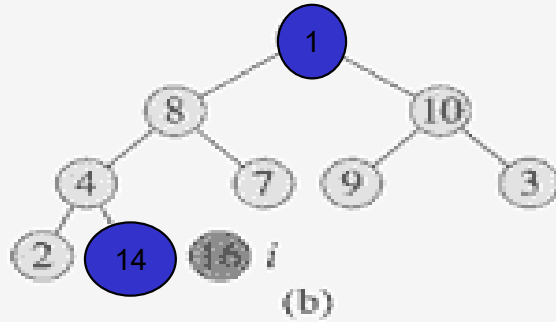
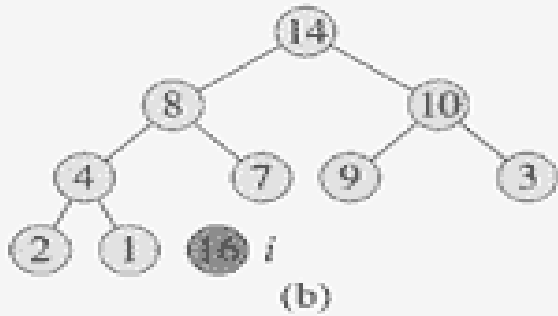
(j)

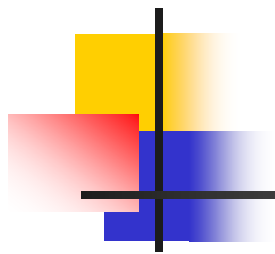
A [1 | 2 | 3 | 4 | 7 | 8 | 9 | 10 | 14 | 16]

(k)

Example of HeapSort

Example of HeapSort (Cont.)





Priority Queues



Max-Priority Queues

- A data structure for maintaining a set S of elements, each with an associated value called a *key*.
- Applications:
 - scheduling jobs on a shared computer
 - prioritizing events to be processed based on their predicted time of occurrence.
 - Printer queue
- Heap can be used to implement a max-priority queue

Max-Priority Queue: Basic Operations

- $\text{Maximum}(S)$: \longrightarrow *return* $A[1]$
 - returns the element of S with the largest key (value)
- $\text{Extract-Max}(S)$:
 - removes and returns the element of S with the largest key
- $\text{Increase-Key}(S, x, k)$:
 - increases the value of element x 's key to the new value k ,
 $x.\text{value} \leq k$
- $\text{Insert}(S, x)$:
 - inserts the element x into the set S , i.e. $S \rightarrow S \cup \{x\}$



HEAP-MAXIMUM

```
HEAP-MAXIMUM(A)  
1  return A[1]
```

$\Theta(1)$



HEAP-Extract-Max(A)

1. if $heap-size[A] < 1$ // zero elements
2. **then error** “heap underflow”
3. $max \leftarrow A[1]$ // max element in first position
4. $A[1] \leftarrow A[heap-size[A]]$
 // value of last position assigned to first position
5. $heap-size[A] \leftarrow heap-size[A] - 1$
6. Heapify($A, 1$)
7. return max

Running time : Dominated by the running time of MaxHeapify
= $O(\lg n)$



HEAP-Extract-MIN(A)

- Write a pseudo-code code similar to HEAP-Extract-MIN(A)

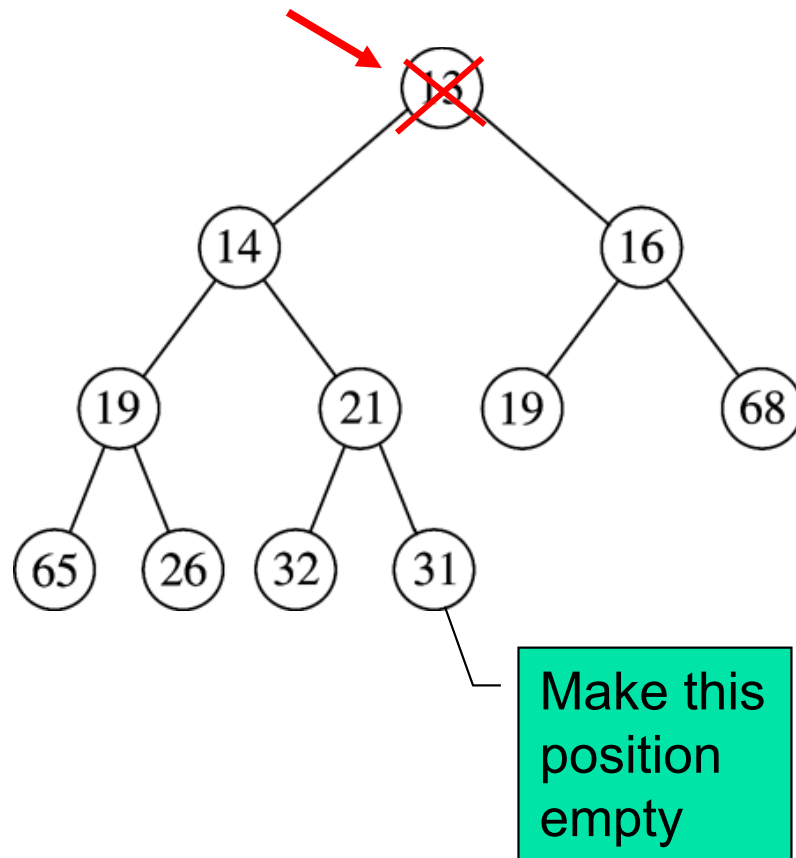


HEAP-Extract-MIN

- Minimum element is always at the root in min-heap
- Heap decreases by one in size
- Move last element into hole at root
- *Percolate down* while heap-order property not satisfied

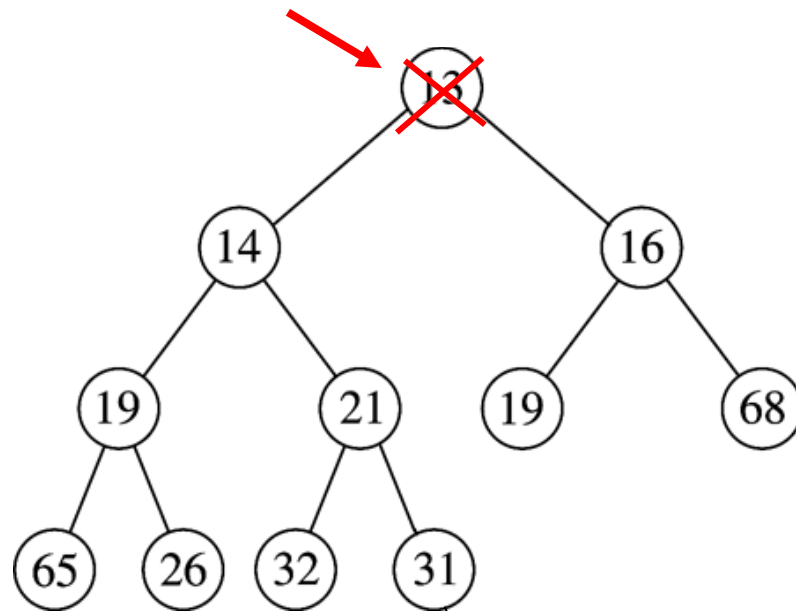
Percolating down...

HEAP-Extract-MIN : Example

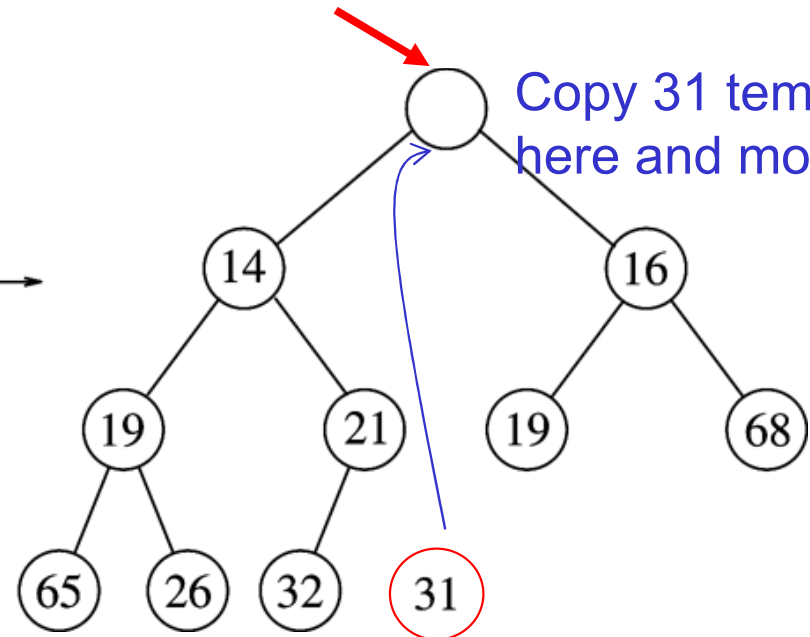


Percolating down...

HEAP-Extract-MIN : Example



Make this position empty



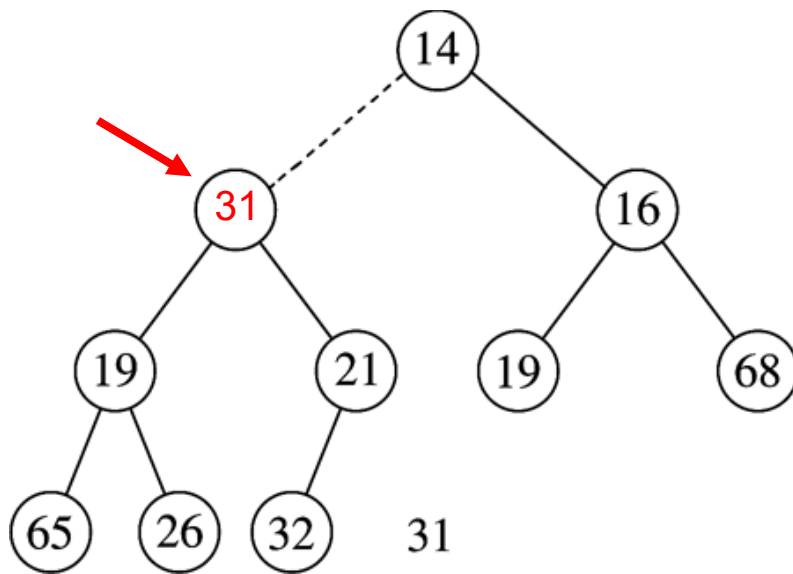
Copy 31 temporarily here and move it down

Is $31 > \min(14, 16)$?

• Yes - swap 31 with $\min(14, 16)$

Percolating down...

HEAP-Extract-MIN : Example

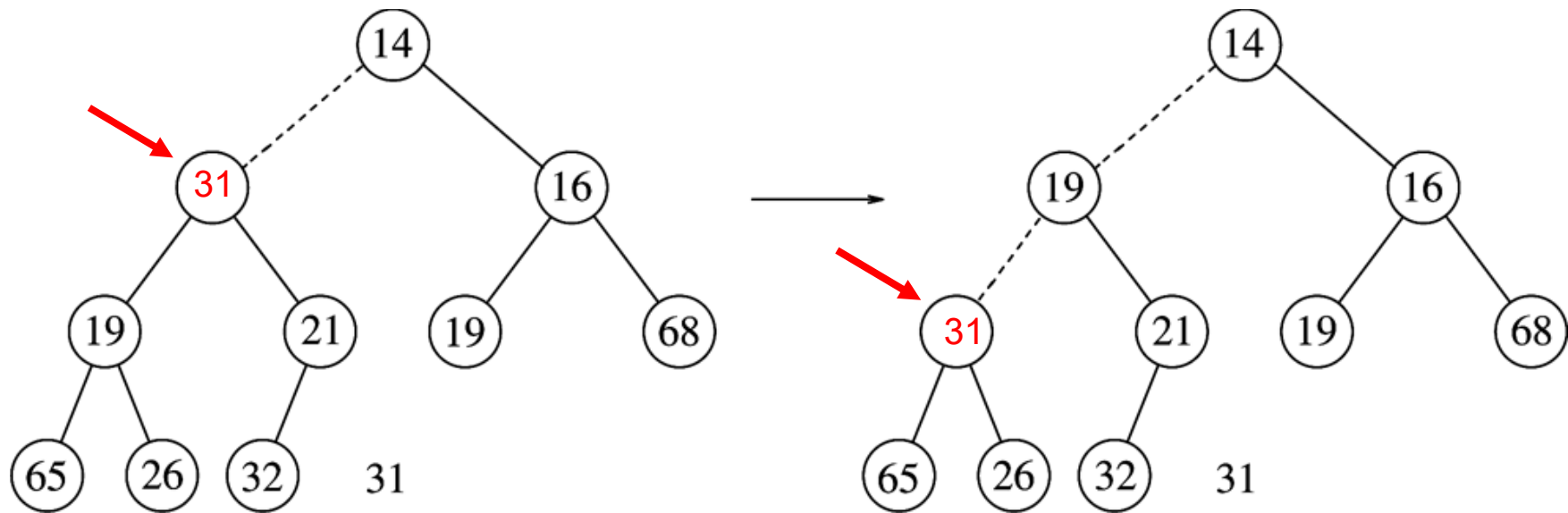


Is $31 > \min(19, 21)$?

• Yes - swap 31 with $\min(19, 21)$

Percolating down...

HEAP-Extract-MIN : Example



Is $31 > \min(19, 21)$?

• Yes - swap 31 with $\min(19, 21)$

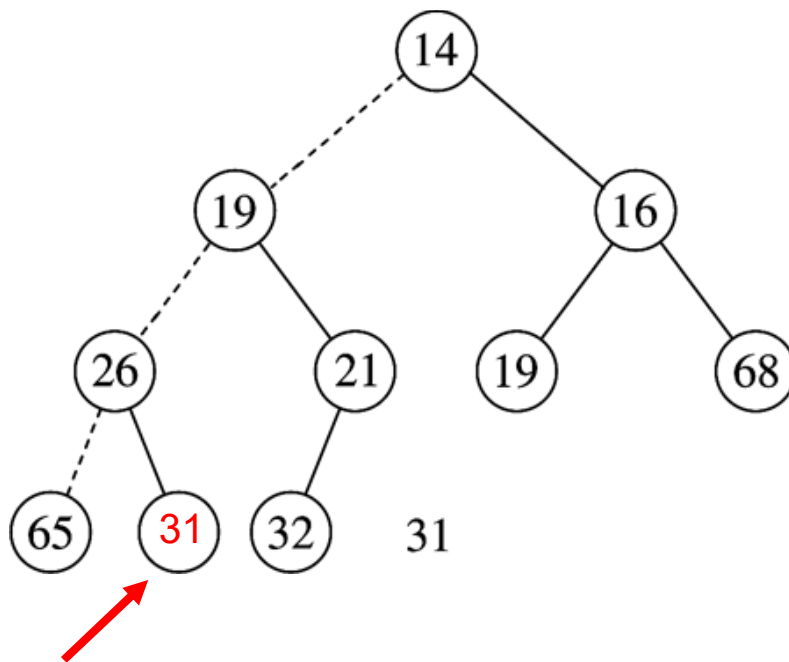
Is $31 > \min(65, 26)$?

• Yes - swap 31 with $\min(65, 26)$

Percolating down...

Percolating down...

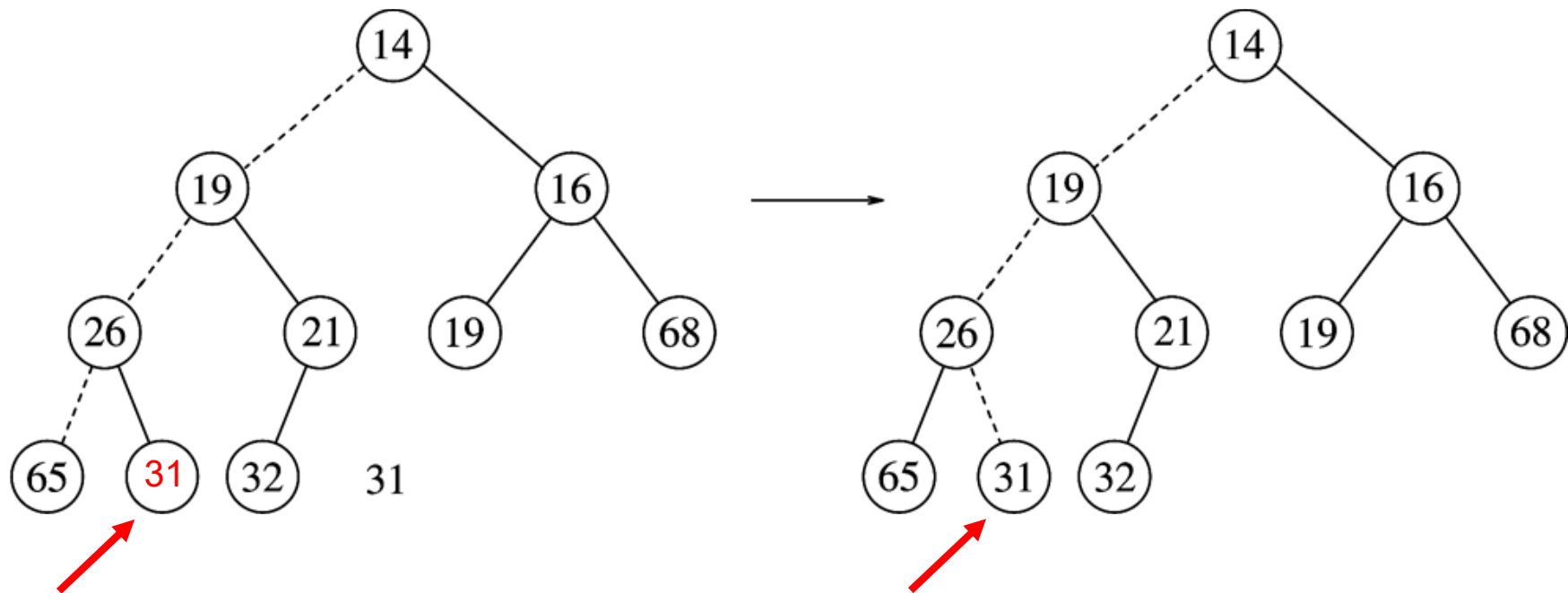
HEAP-Extract-MIN : Example



Percolating down...

Percolating down...

HEAP-Extract-MIN : Example



- ✓ Heap order prop
- ✓ Structure prop



HEAP-INCREASE-KEY

- Increase the job priority
- Steps
 - Update the key of $A[i]$ to its new value
 - May violate the max-heap property
 - Traverse a path from $A[i]$ toward the root to find a proper place for the newly increased key



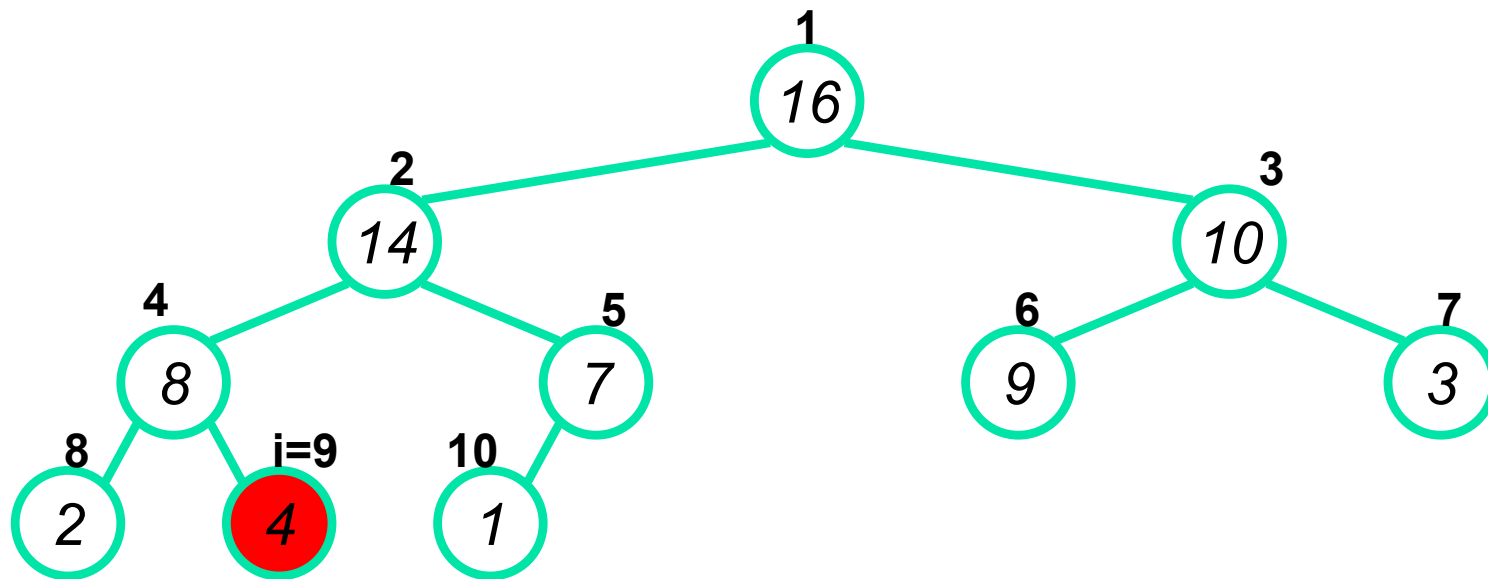
HEAP-INCREASE-KEY(A, i, key)

// increase a value (key) in the array

1. **if** $key < A[i]$
 2. **then error** “new key is smaller than current key”
 3. $A[i] \leftarrow key$
 4. **while** $i > 1$ and $A[\text{Parent}(i)] < A[i]$
 5. do exchange $A[i] \leftrightarrow A[\text{Parent}(i)]$
 6. $i \leftarrow \text{Parent}(i)$ // move index up to parent
- $O(\lg n)$

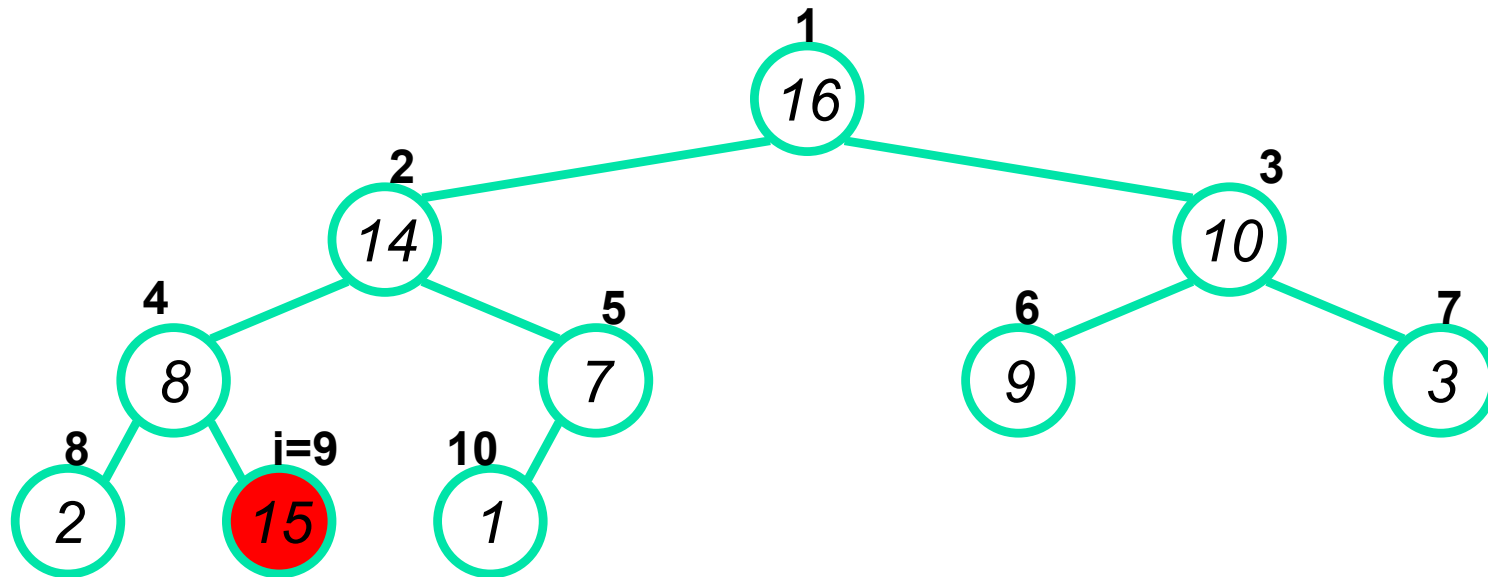
HEAP-INCREASE-KEY() Example

- $A = \{16, 14, 10, 8, 7, 9, 3, 2, 4, 1\}$



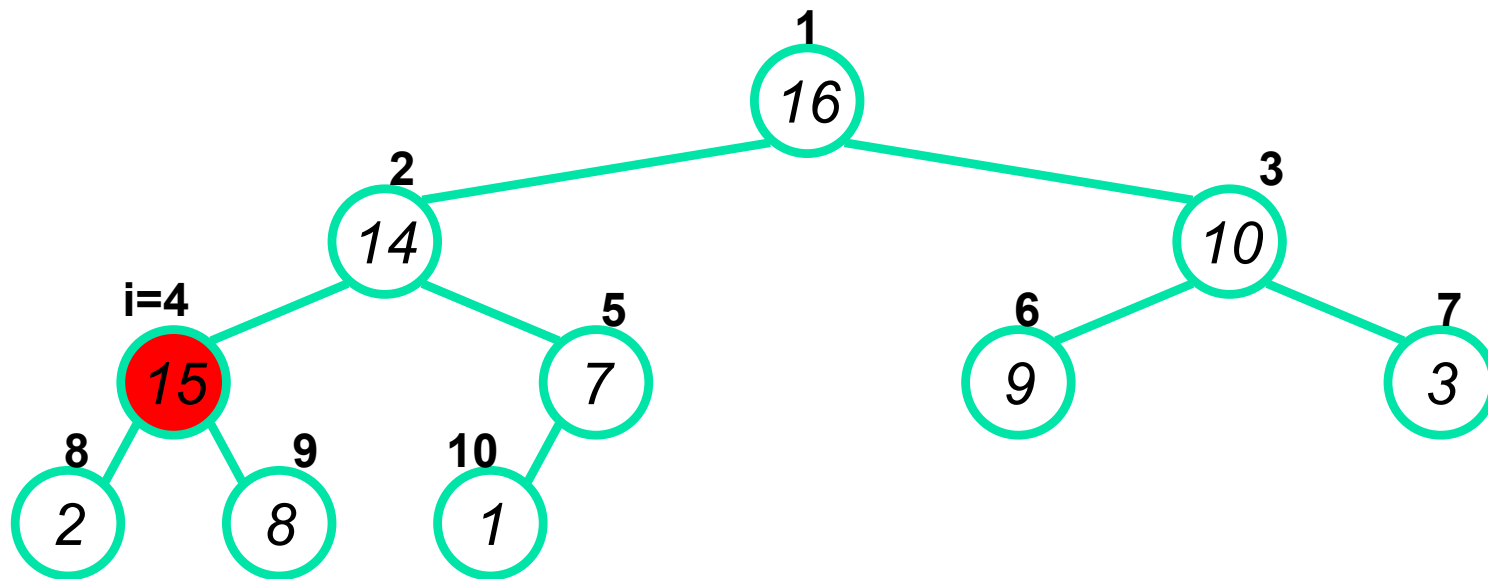
HEAP-INCREASE-KEY() Example

- $A = \{16, 14, 10, 8, 7, 9, 3, 2, 15, 1\}$
- The index $i=9$ increased to 15.



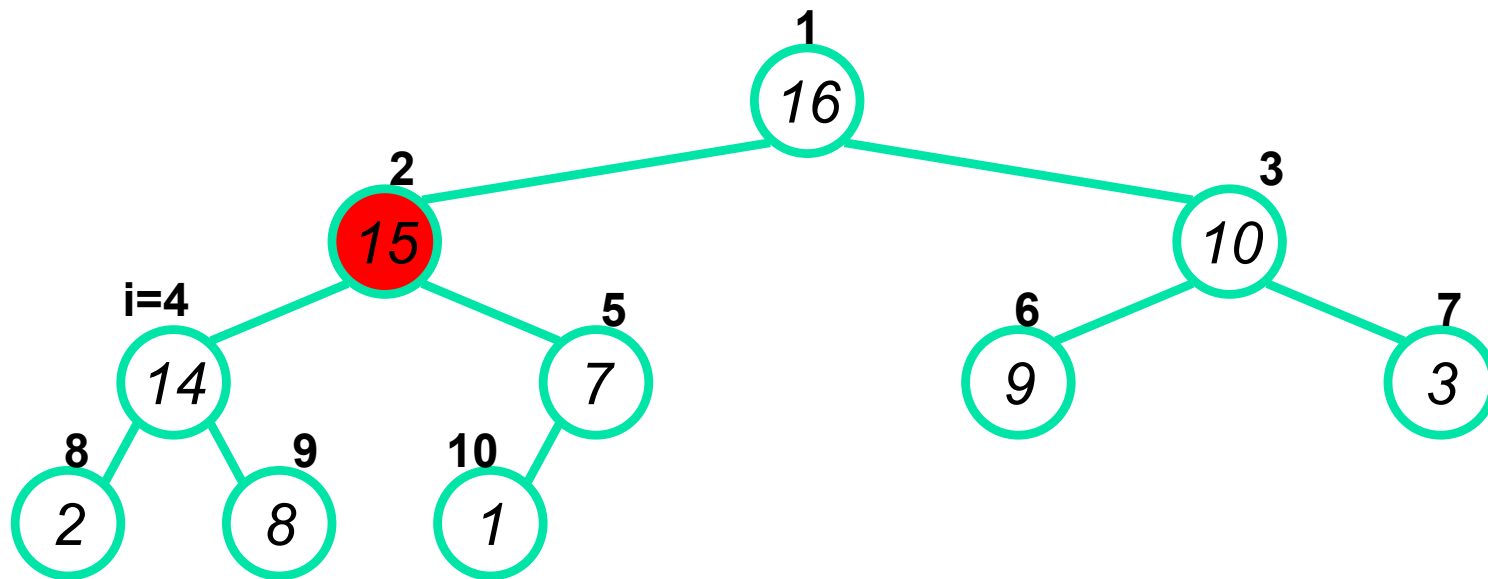
HEAP-INCREASE-KEY() Example

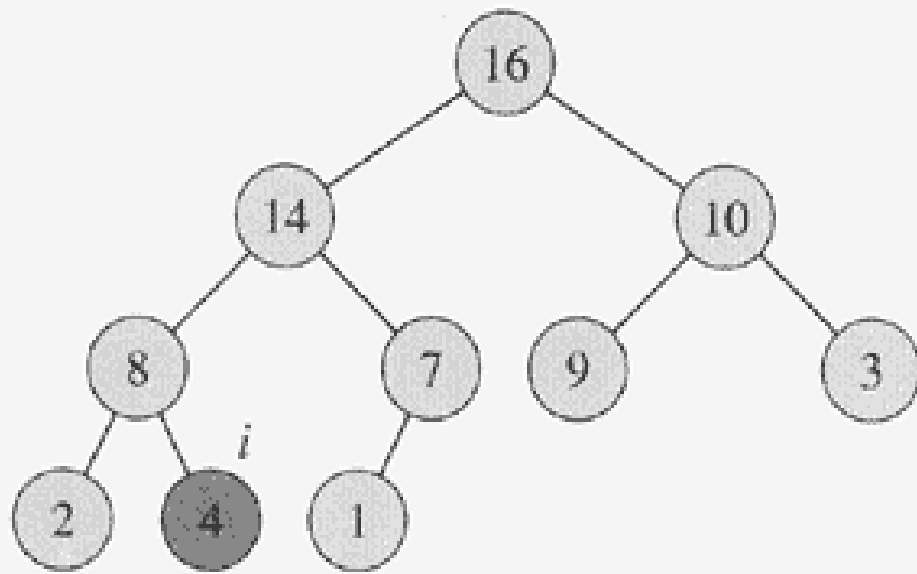
- $A = \{16, 14, 10, 15, 7, 9, 3, 2, 8, 1\}$
- After one iteration of the while loop of lines 4-6, the node and its parent have exchanged keys (values), and the index i moves up to the parent.



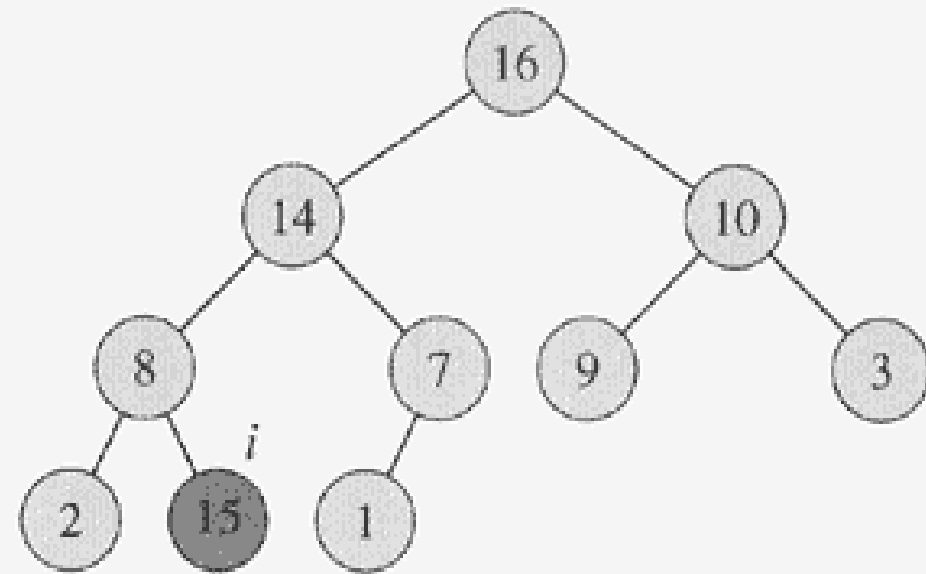
HEAP-INCREASE-KEY() Example

- $A = \{16, 15, 10, 14, 7, 9, 3, 2, 8, 1\}$
- After one more iteration of the while loop.
- The max-heap property now holds and the procedure terminates.

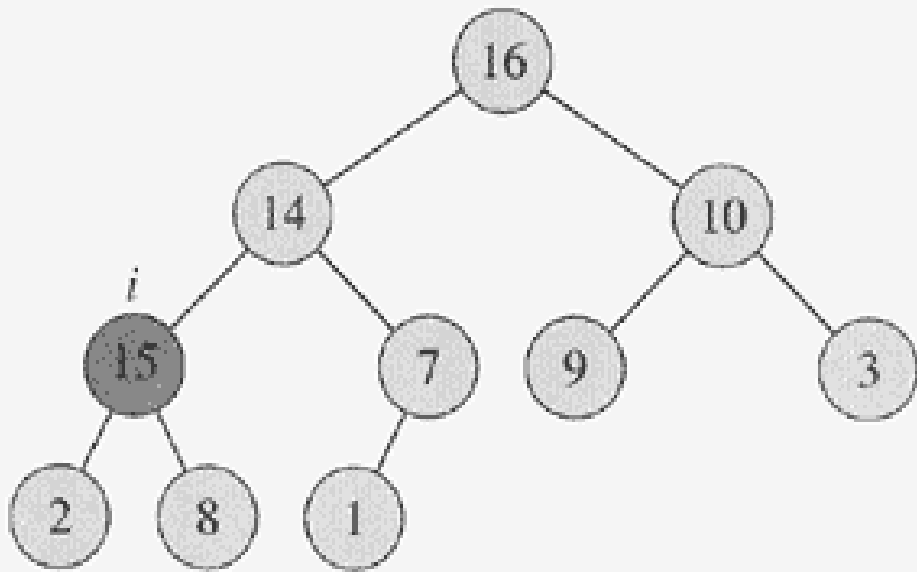




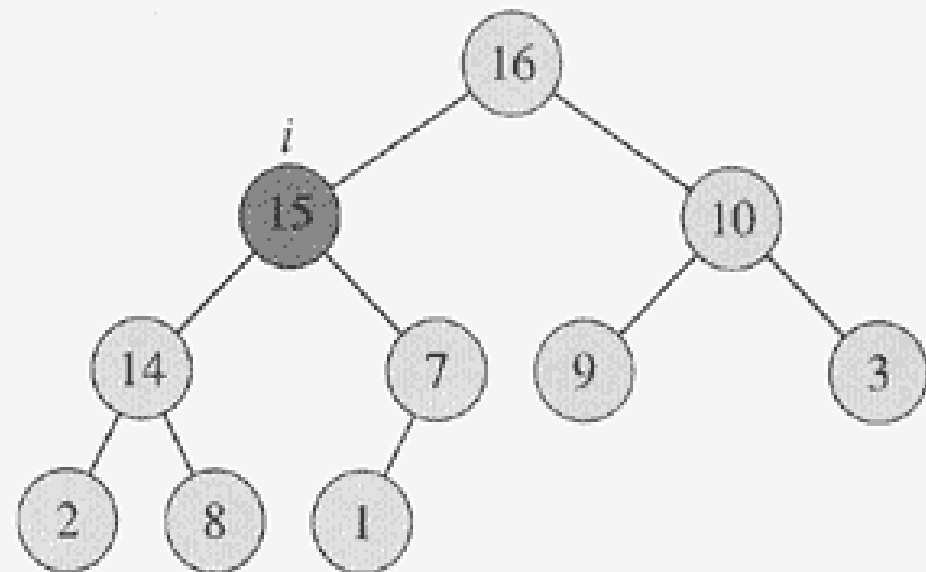
(a)



(b)



(c)



(d)

Example of HEAP-INCREASE-KEY



MAX-HEAP-INSERT

MAX-HEAP-INSERT(A, key)

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

Running time is $O(\lg n)$

The path traced from the new leaf to the root has
length $O(\lg n)$



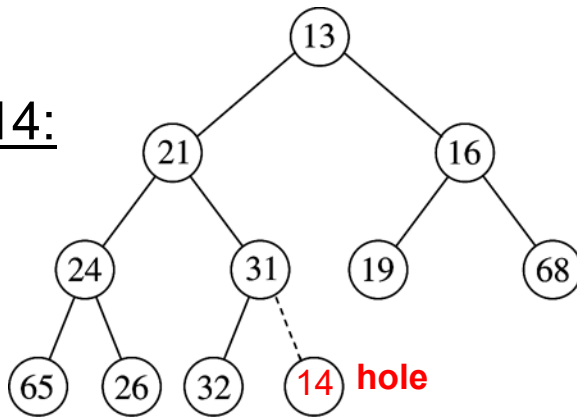
MIN-HEAP-INSERT

- Insert new element into the heap at the next available slot (“hole”)
 - According to maintaining a complete binary tree
- Then, “percolate” the element up the heap while heap-order property not satisfied

Percolating Up

MIN-HEAP-INSERT : Example

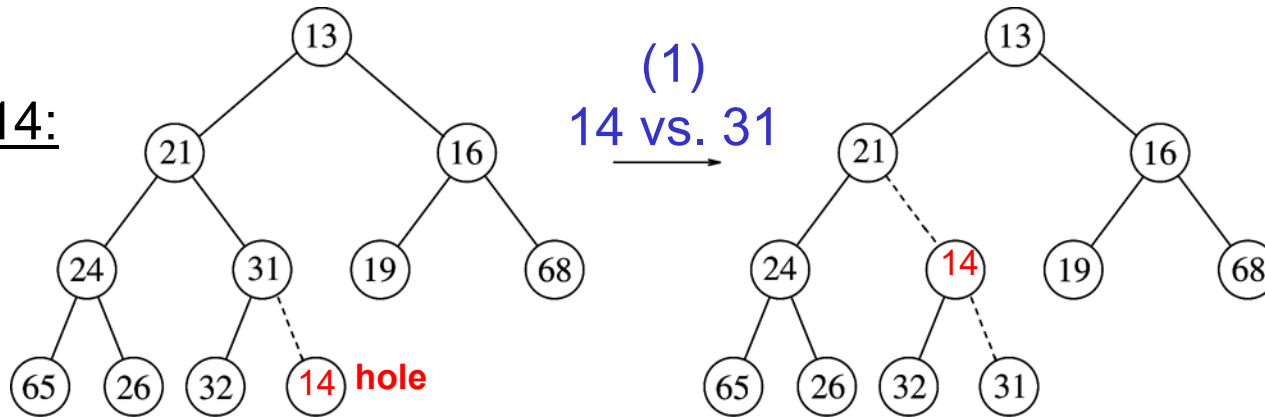
Insert 14:



Percolating Up

MIN-HEAP-INSERT : Example

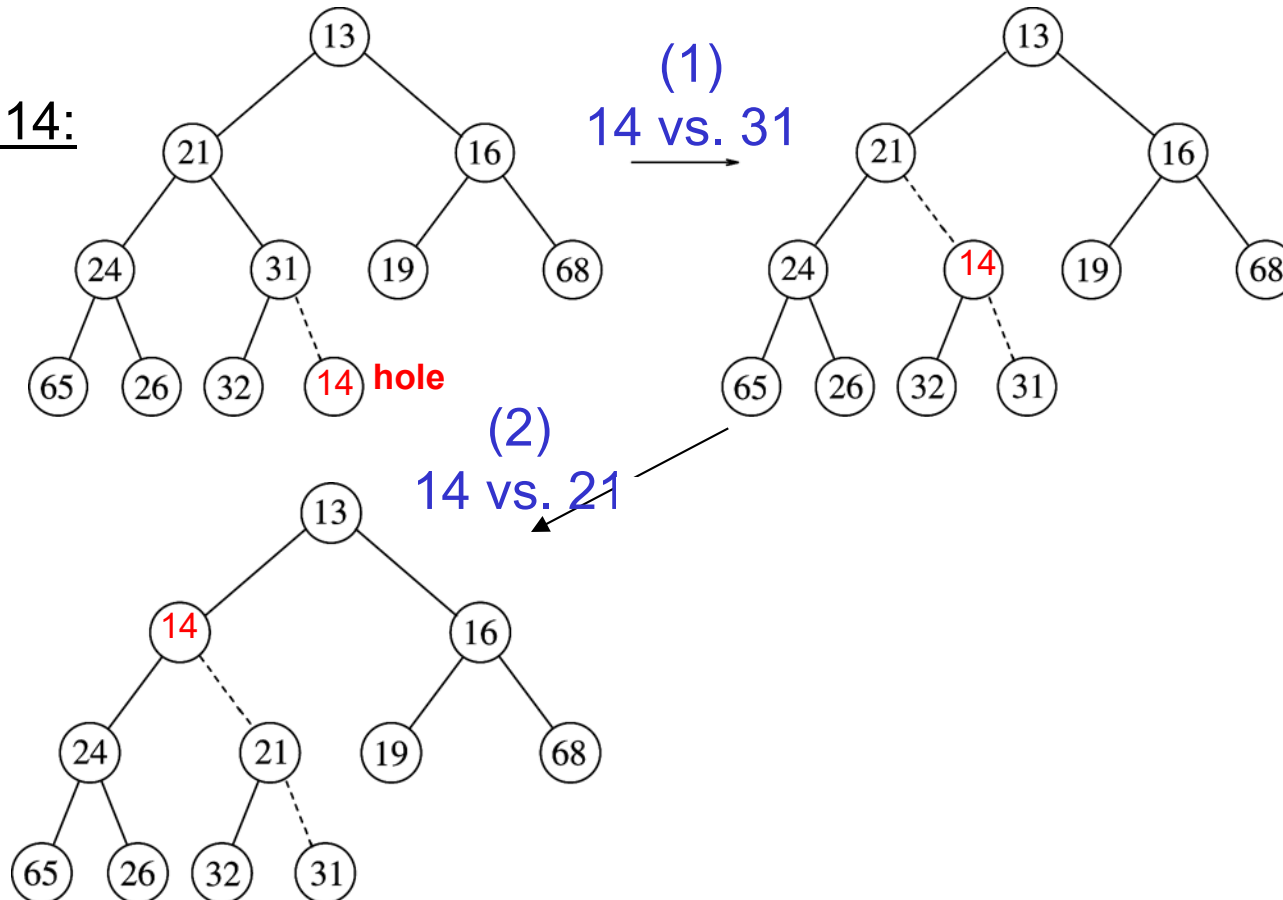
Insert 14:



Percolating Up

MIN-HEAP-INSERT : Example

Insert 14:



Percolating Up

MIN-HEAP-INSERT : Example

Insert 14:

