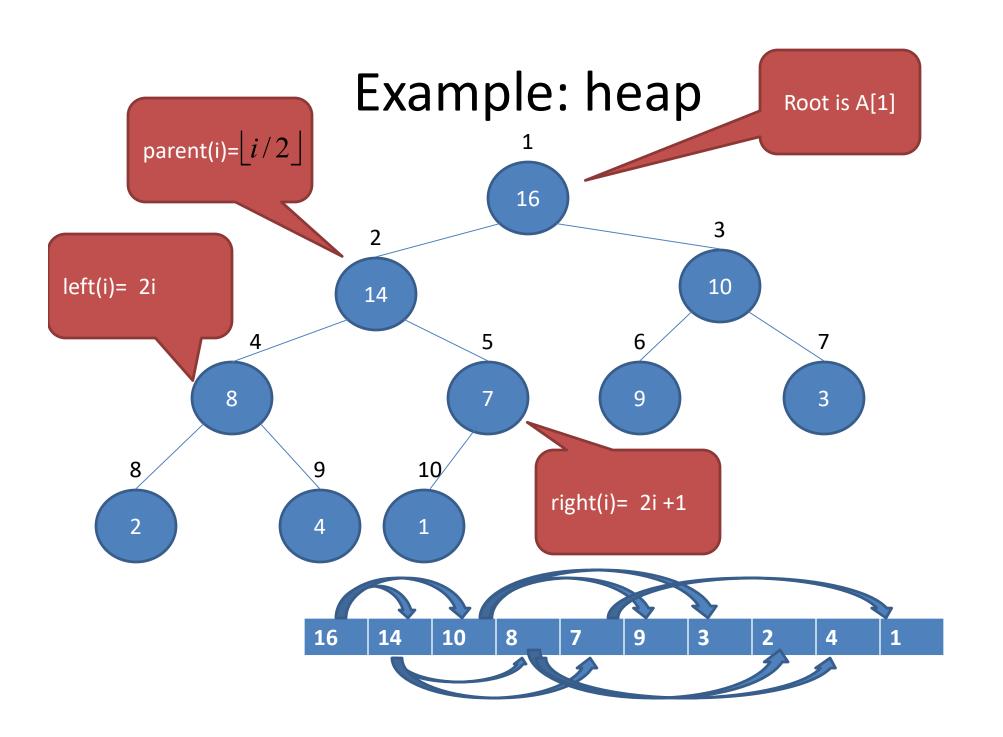
Ch11: Heap and Heap Sort

305234
Algorithm Analysis and Design
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What is the (binary) heap?

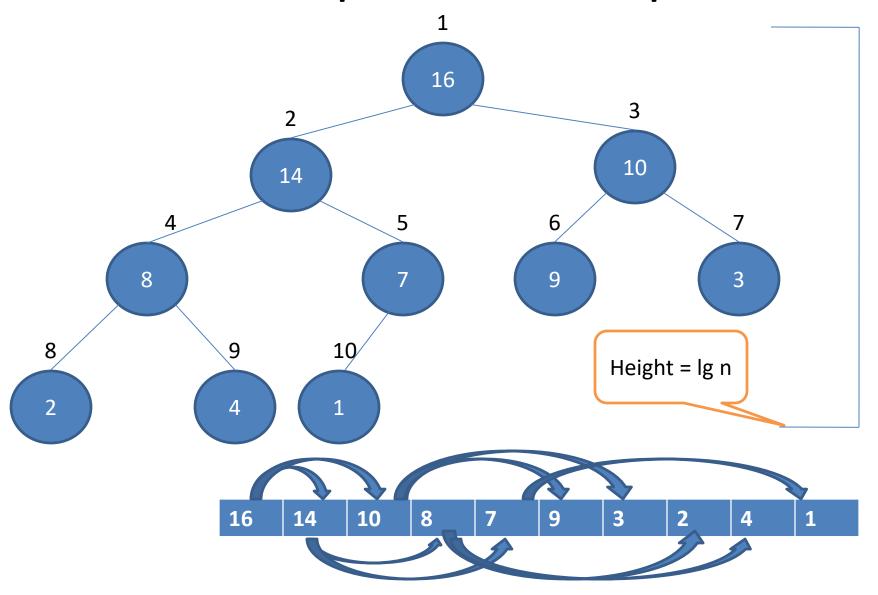
- The (binary) heap data structure is an array object that can be viewed as a nearly complete binary tree. Each node of the tree corresponds to an element of the array that stores the value in the node.
- An array A that represents a heap is an object with two attributes:
 - length[A] is the number of elements in the array
 - heap-size[A] is the number of elements in the heap stored within array A.



Binary Heap

- There are two kinds of binary heaps:
- Max-heaps
 - For every node i other than the root,A[parent(i)] ≥ A[i]
- Min-heaps
 - For every node i other than the root,A[parent(i)] ≤ A[i]

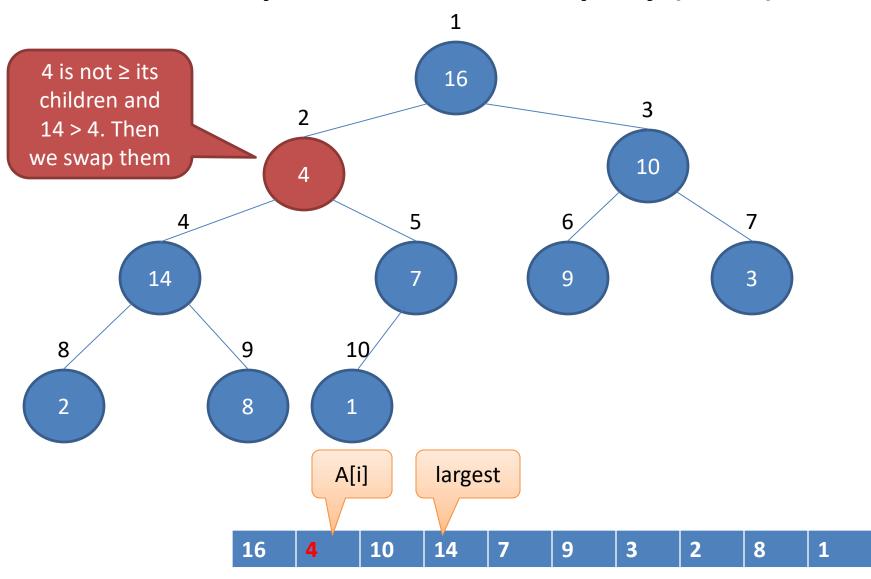
Example: Max-heap



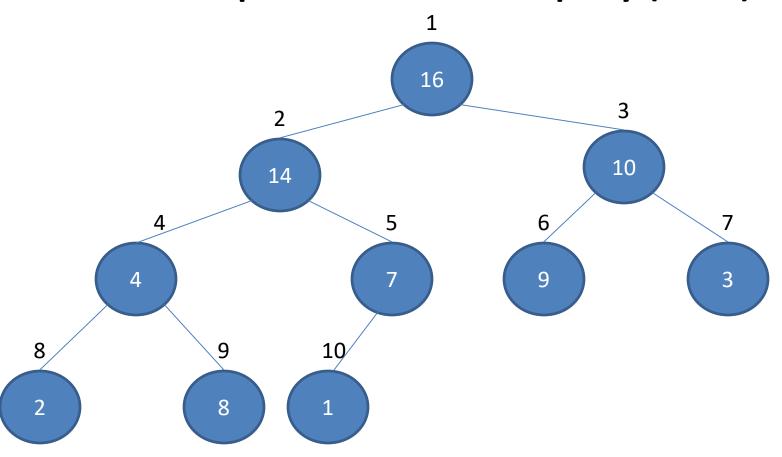
Max-Heapify(A,i)

```
I = left(i)
r = right(i)
if I \leq \text{heap-size}[A] and A[I] > A[i]
       then largest = I
else largest = i
if r \le heap\text{-size}[A] and A[r] > A[largest]
       then largest = r
If largest != i
       then exchanged A[i] and A[largest]
             Max-Heapify(A, largest)
```

Example: Max-Heapify(A,2)

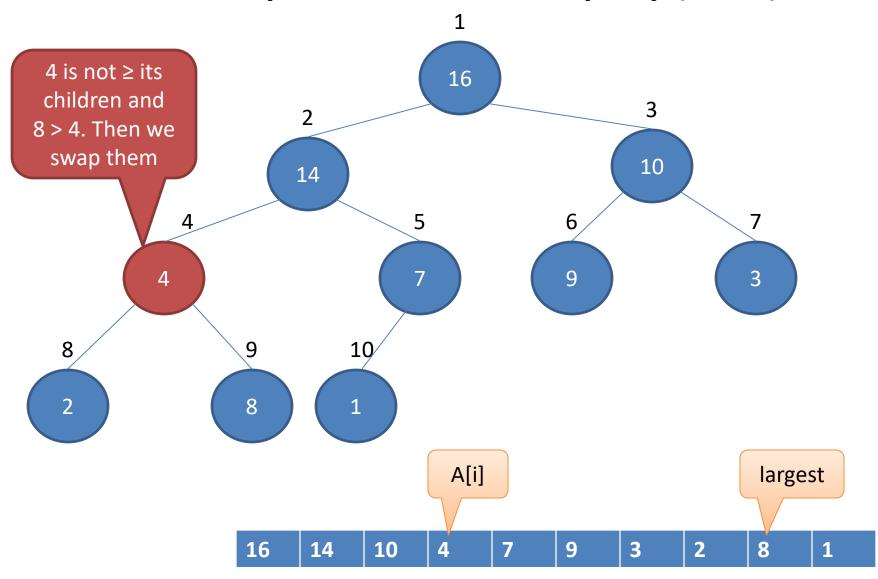


Example: Max-Heapify(A,2)

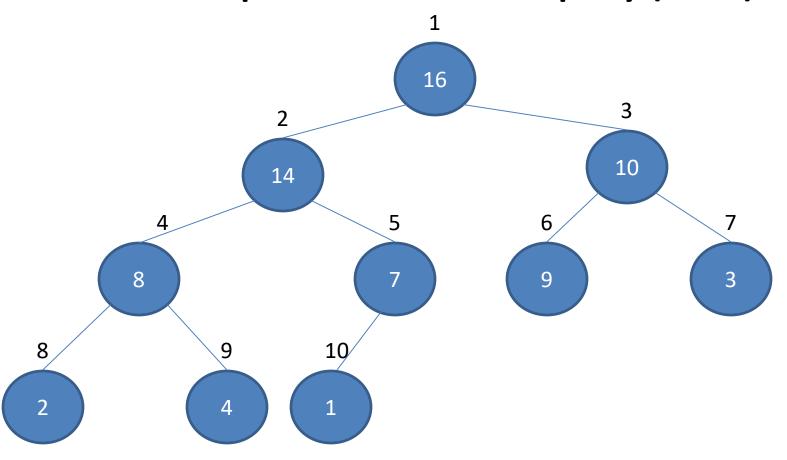


 16
 14
 10
 4
 7
 9
 3
 2
 8
 1

Example: Max-Heapify(A,4)

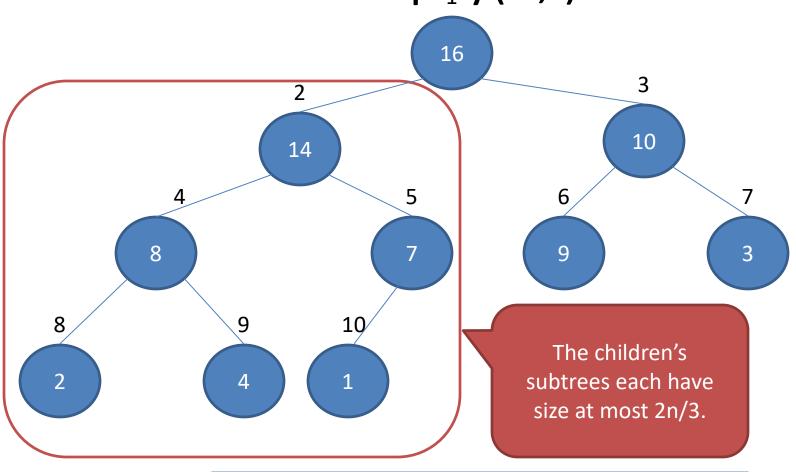


Example: Max-Heapify(A,4)



 16
 14
 10
 8
 7
 9
 3
 2
 4
 1

Analyze: Running time of Max-Heapify(A,i)



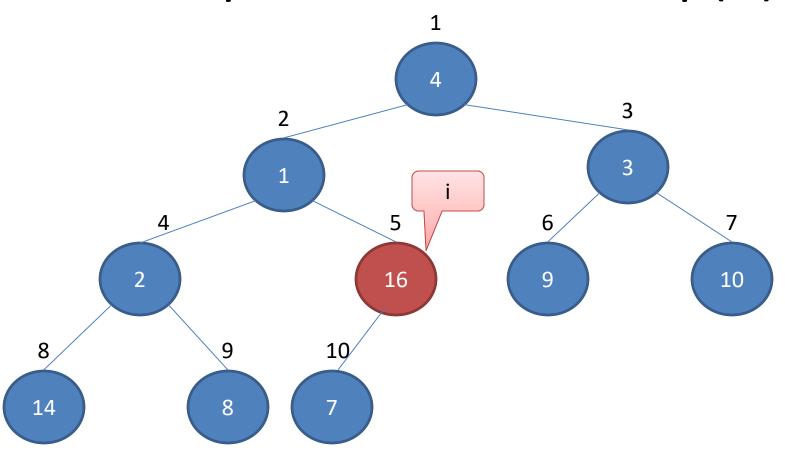
$$T(n) \le T(2n/3) + \Theta(1)$$

Analyze: Running time of Max-Heapify(A,i)

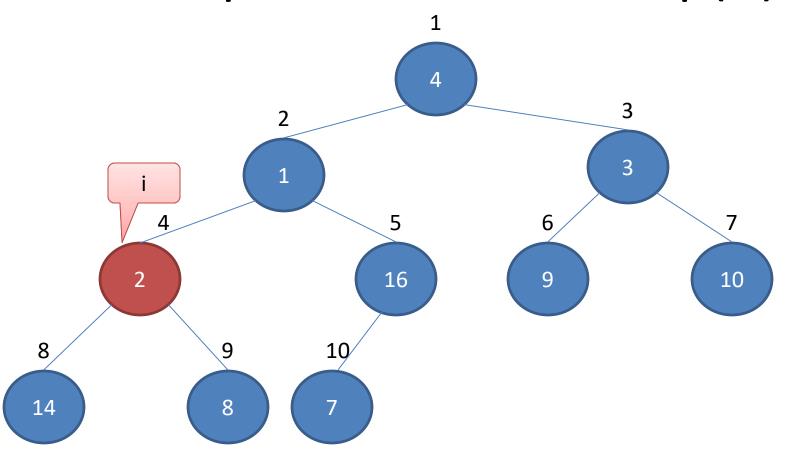
- We have T(n) = T(2n/3) + 1
- Determine which case of the master theorem applies:
- We have a=1, b=3/2, f(n)=1
- Thus we have $n^{\log_b a} = n^{\log_{3/2} 1} = n^0 = 1$
- Since $f(n) = \Theta(n^0) = \Theta(1)$ we can apply case 2 of the master theorem and conclude that the solution is $T(n) = \Theta(n^{\log_b a} \lg n) = \Theta(\lg n)$

Build-Max-Heap(A)

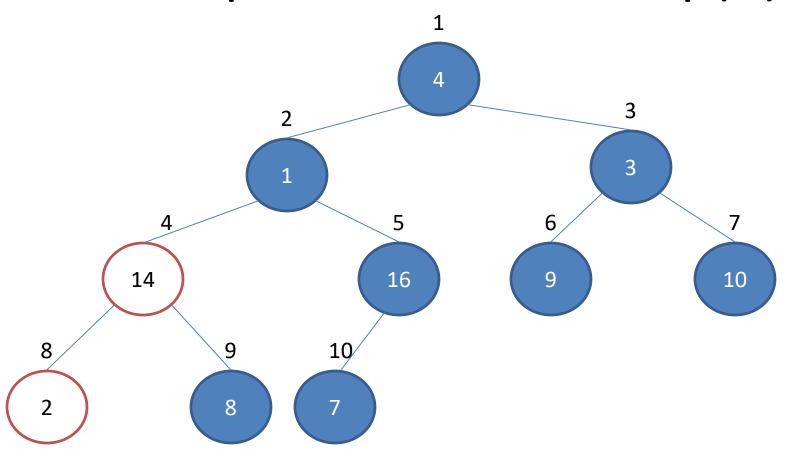
```
heap-size[A] = length[A]
for i = \lfloor length[A]/2 \rfloor downto 1
do Max-Heapify(A,i)
```

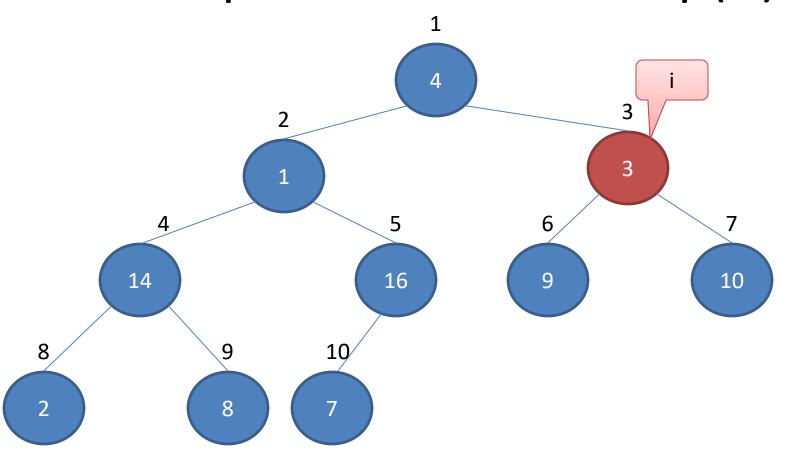


4 1 3 2 16 9 10 14 8 7

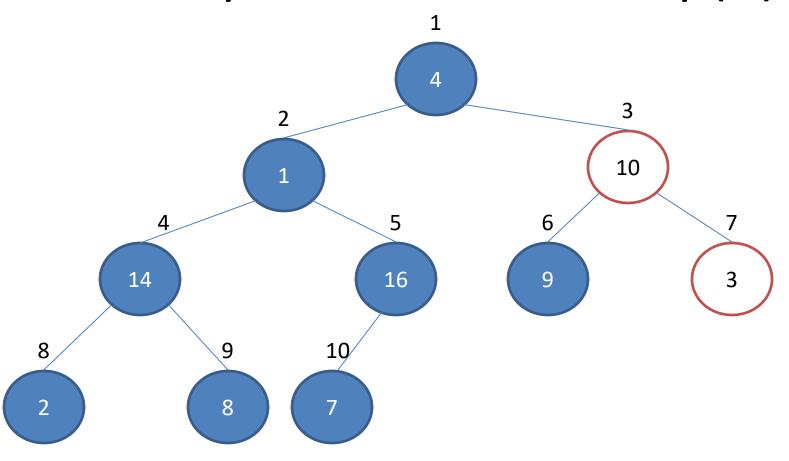


4 1 3 2 16 9 10 14 8 7

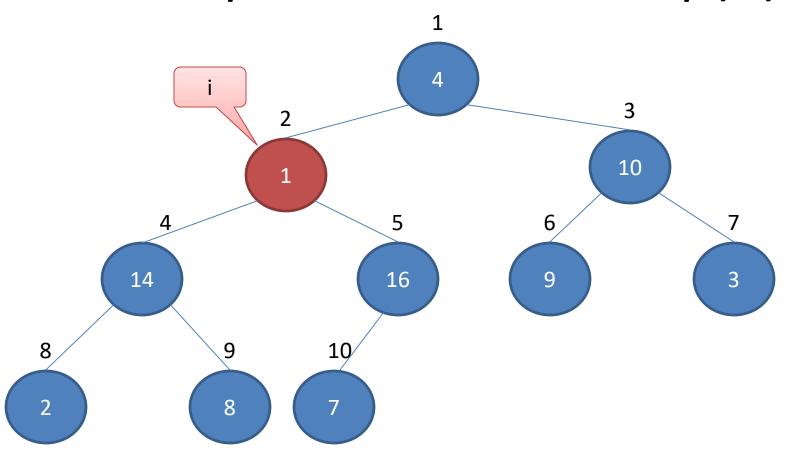




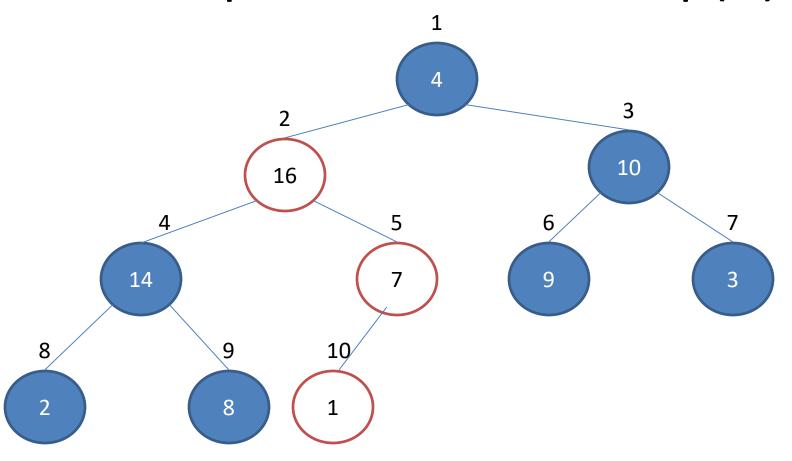
4 1 3 14 16 9 10 2 8 7



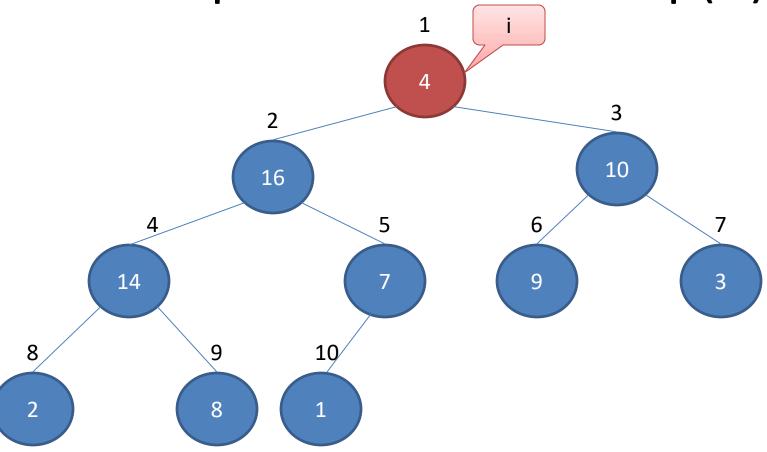
4 1 10 14 16 9 3 2 8 7

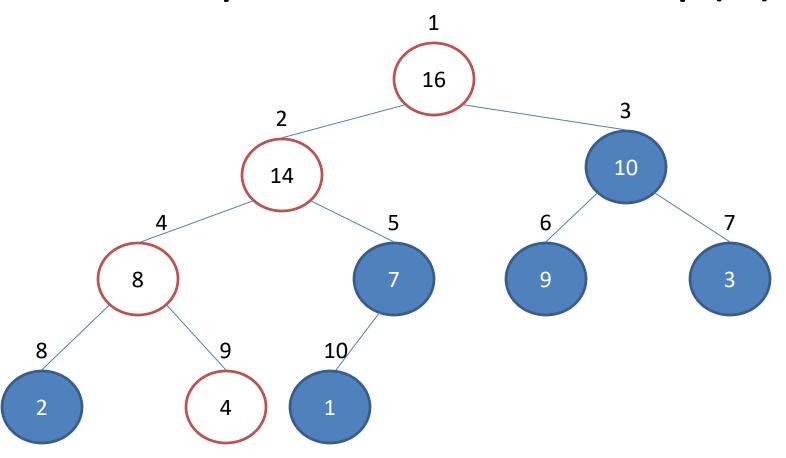


4 1 10 14 16 9 3 2 8 7



4 16 10 14 7 9 3 2 8 1





 16
 14
 10
 8
 7
 9
 3
 2
 4
 1

Analyze: Build-Max-Heap(A)

heap-size[A] = length[A]
for i =
$$\lfloor length[A]/2 \rfloor$$
 downto 1
do Max-Heapify(A,i)

loop invariant =

at the start of each iteration of the for loop of lines 2-3, each node i+1, i+2,...,n is the root of a max-heap.

Initialization:

Before running loop 1, i = $\lfloor n/2 \rfloor$. Each node $\lfloor n/2 \rfloor +1$, $\lfloor n/2 \rfloor +2$, ..., n is a leaf and is the root of a trivial max-heap. (True!!)

Maintenance:

if children of node i are numbered higher than I, they are both roots of max-heaps.

The condition required for the call Max-Heapify(A,i) to make node I a max-heap root.

Decrementing in the for loop update reestablishes the loop invariant for the next loop. (True!!)

Termination: at termination, i =0. each node 1,2,....,n Is the root of a max-heap. Node 1 is. (True!!)

Analyze running time: Build-Max-Heap(A)

Times

heap-size[A] = length[A] 1 for $i = \lfloor length[A]/2 \rfloor$ downto 1 n/2+1 do Max-Heapify(A,i) n/2.O(lg n)

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

Tight Analysis: an n-element heap has height = $\lfloor \lg n \rfloor$ and at most $\lceil n/2^{h+1} \rceil$ nodes of any height h. The time required by Max-Heapify when called on a node of height h is O(h). Thus running time can be bounded as O(n)

Heapsort(A)

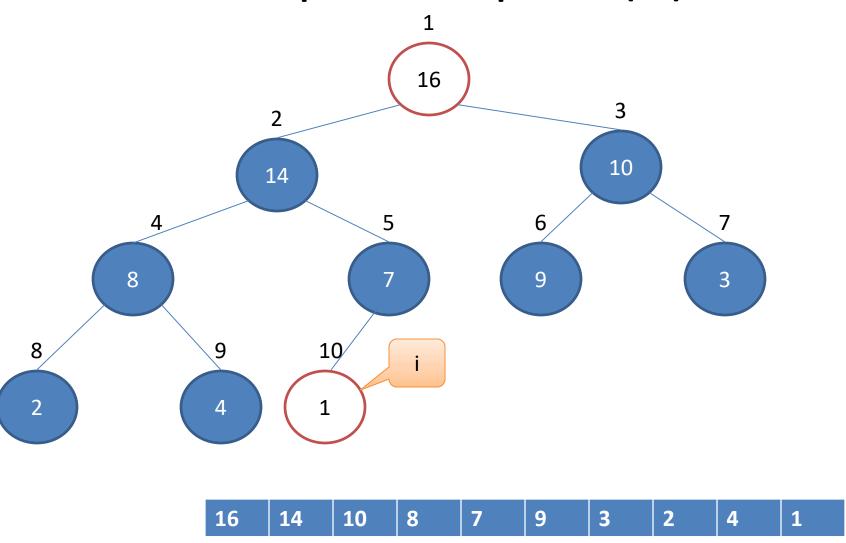
```
Build-Max-Heap(A)

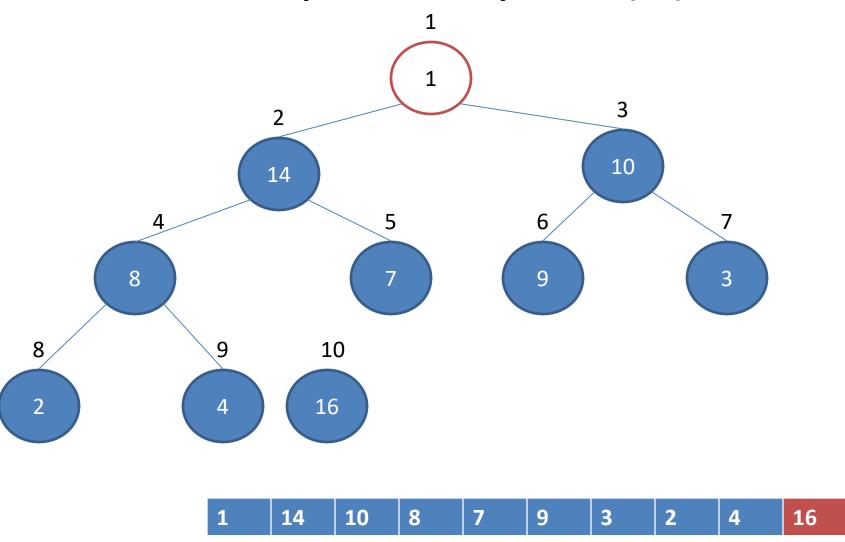
for i = length[A] downto 2

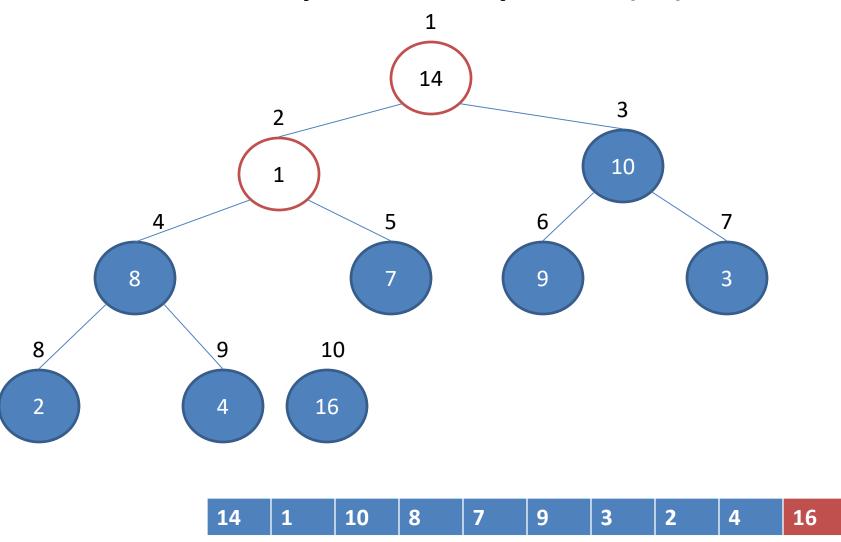
do exchange A[1] and A[i]

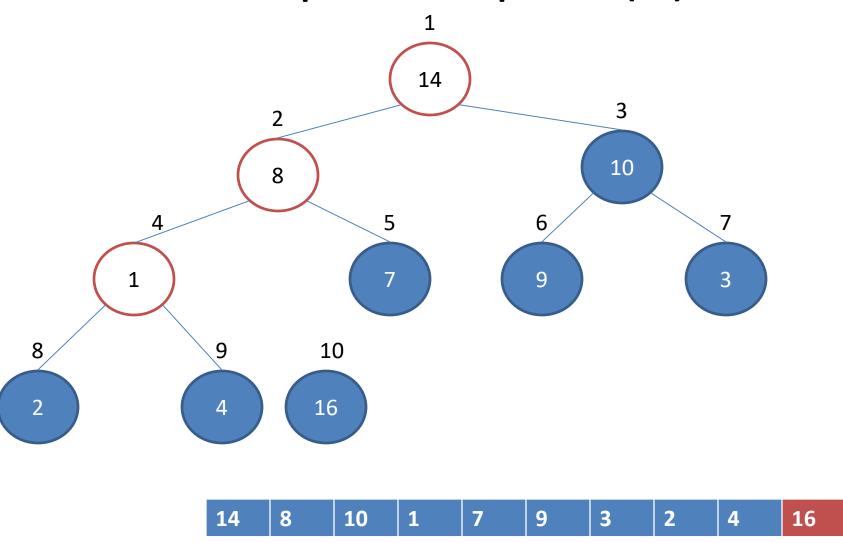
heap-size[A] = heap-size[A] -1

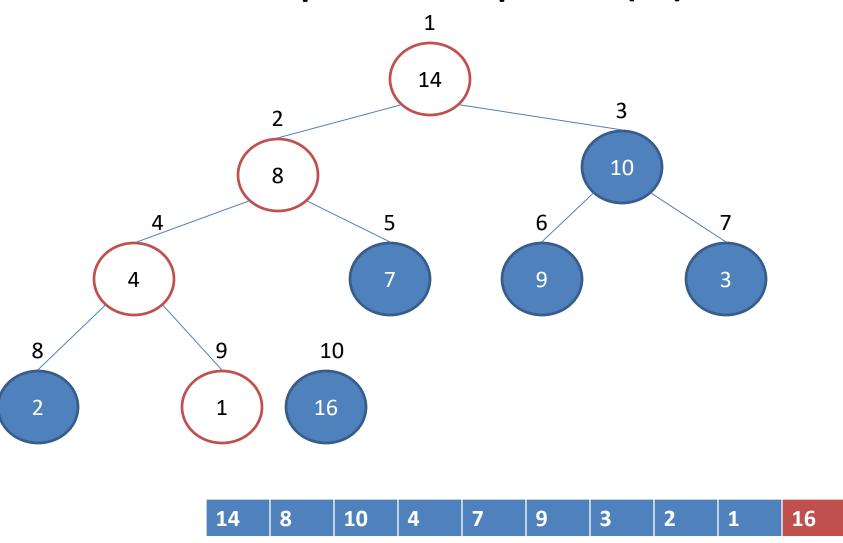
Max-Heapify(A,1)
```

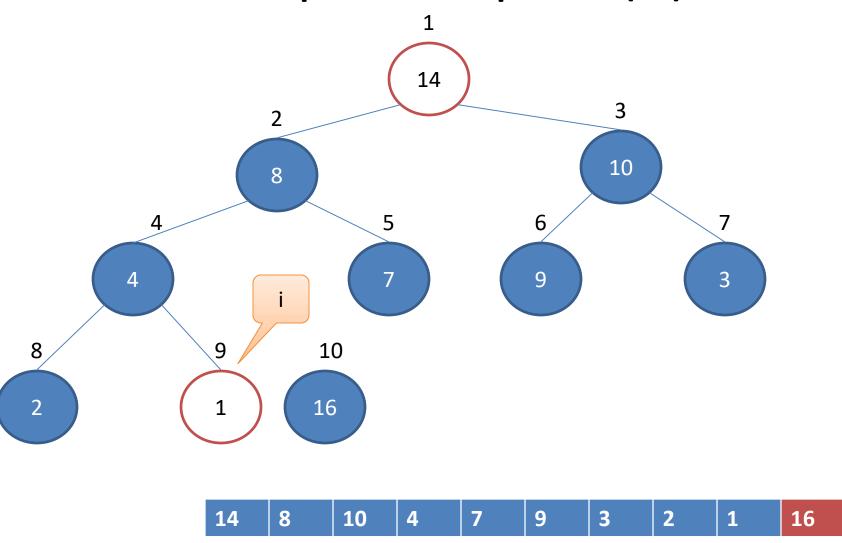


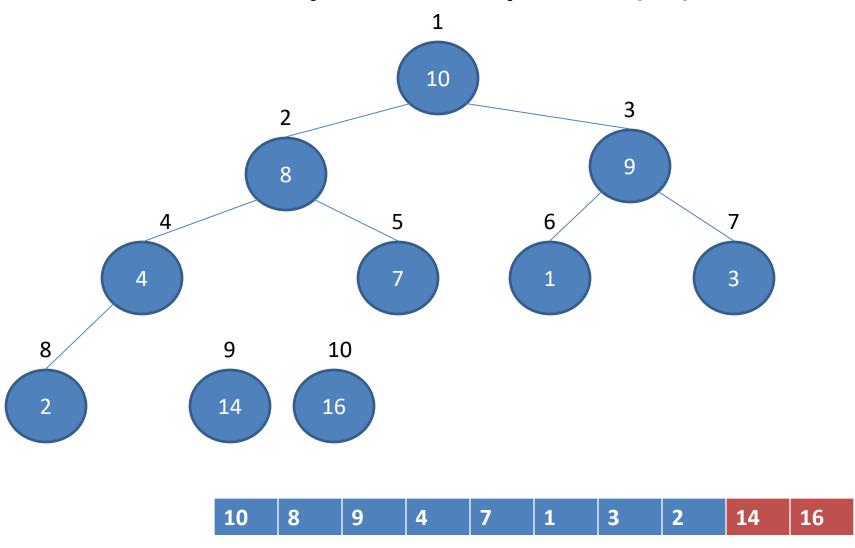


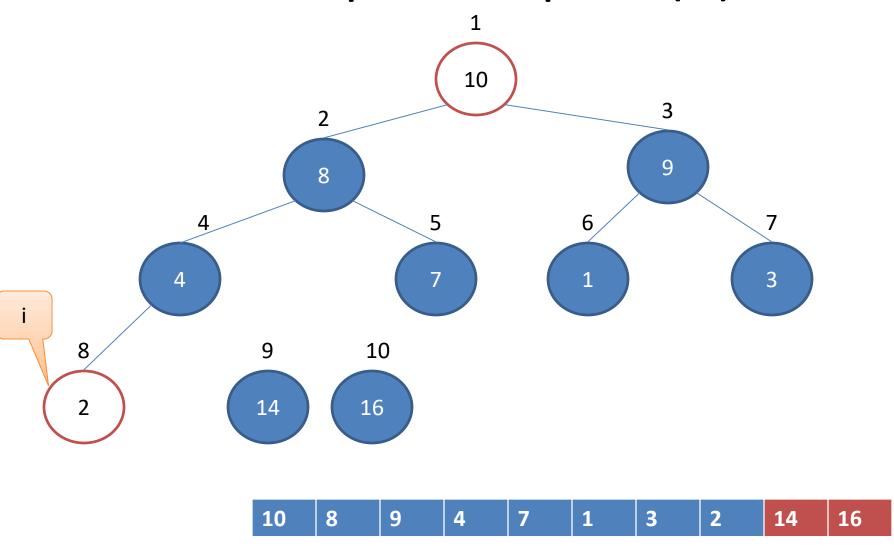


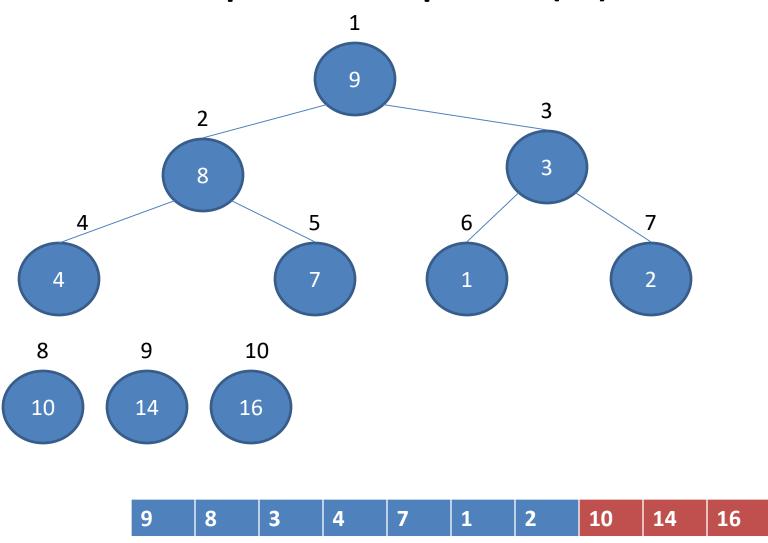


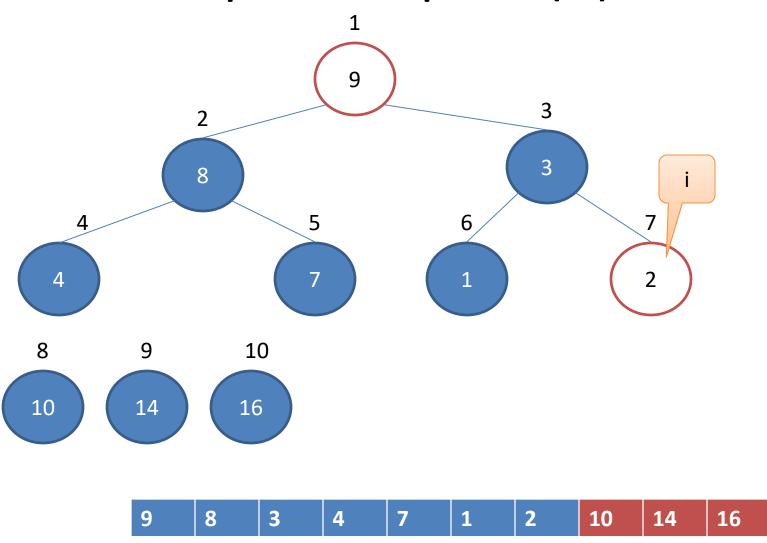


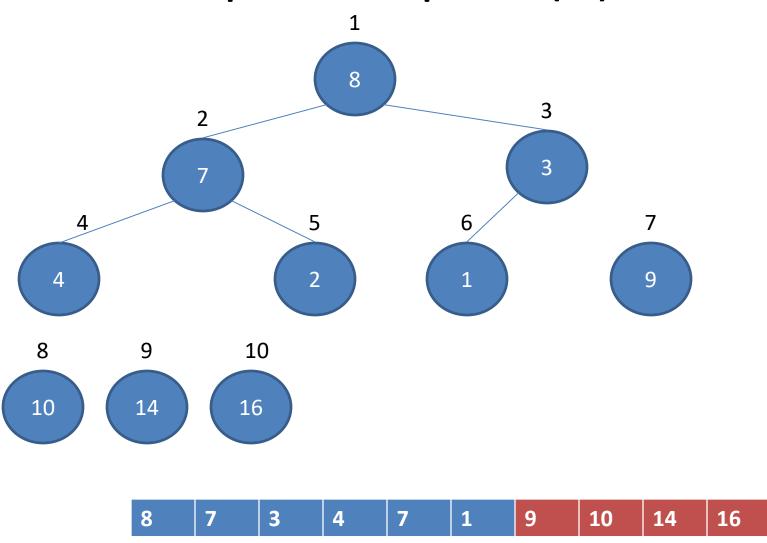


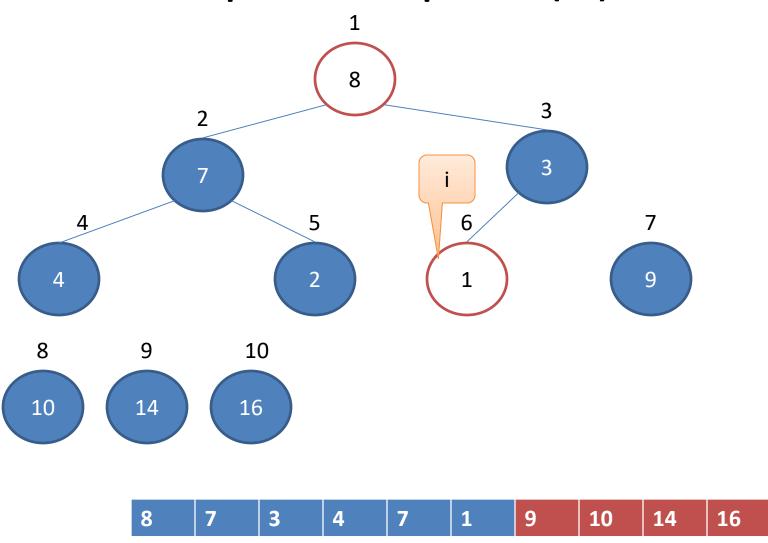


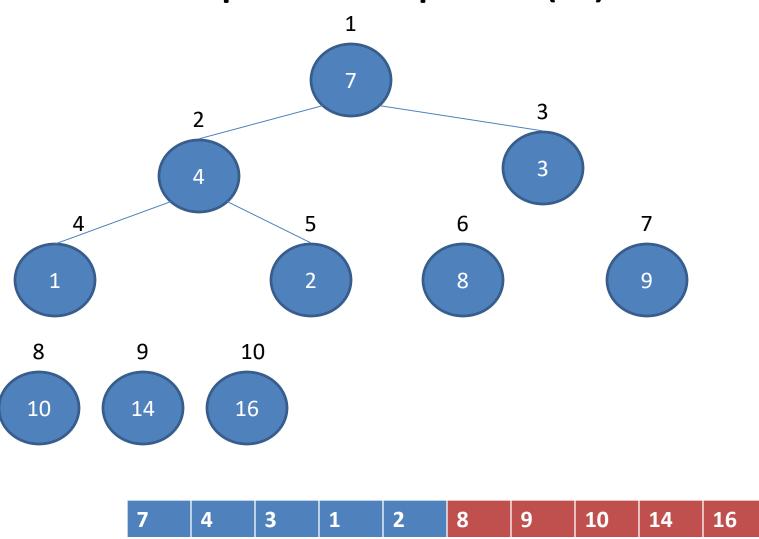


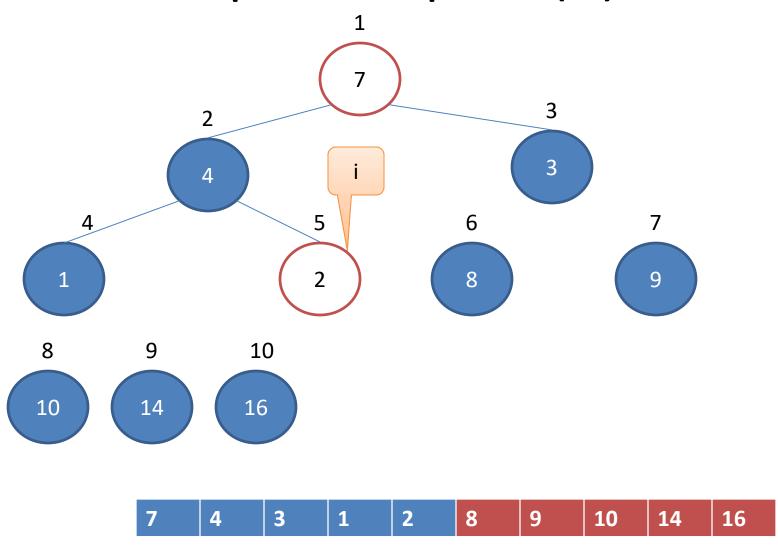


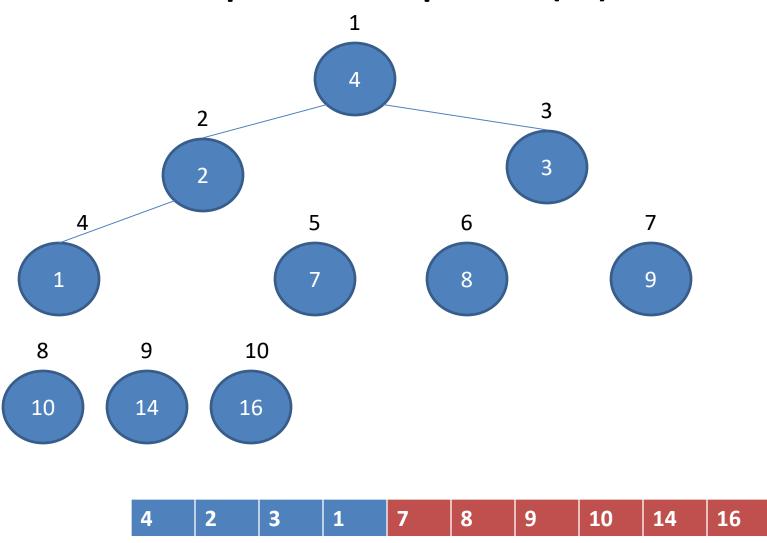


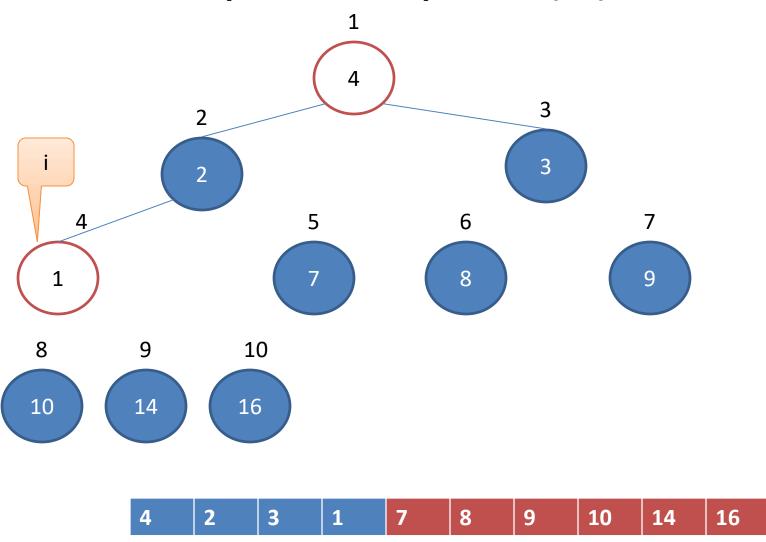


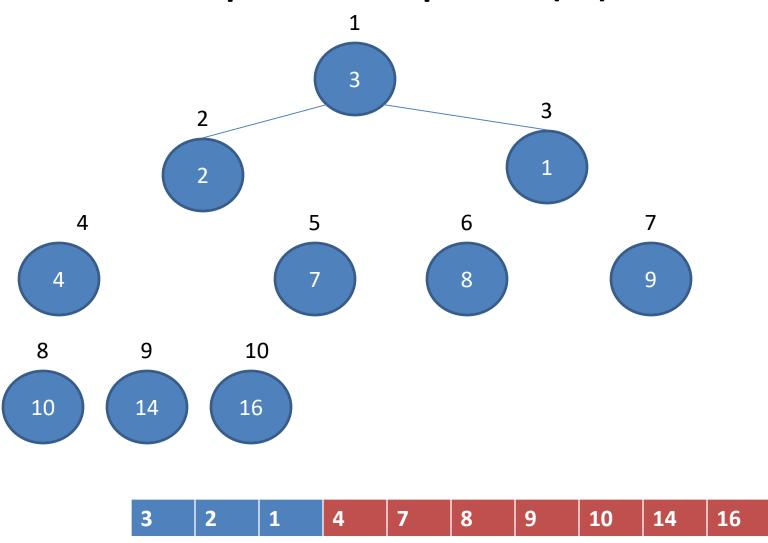


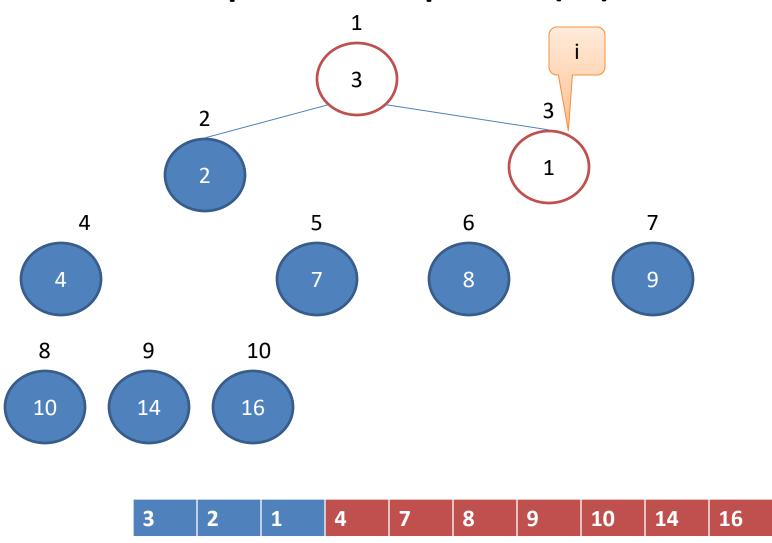


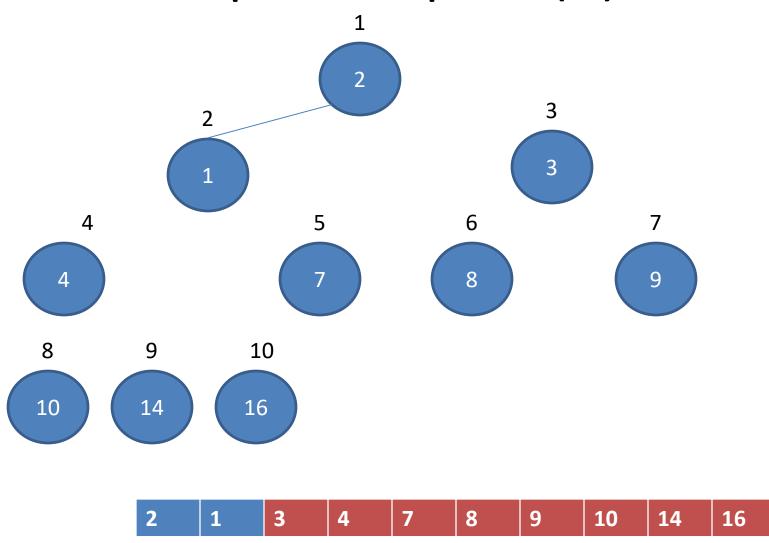


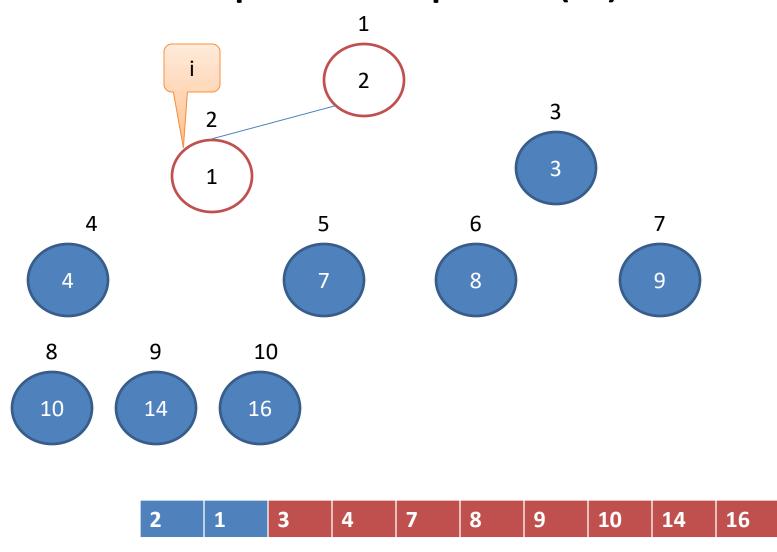


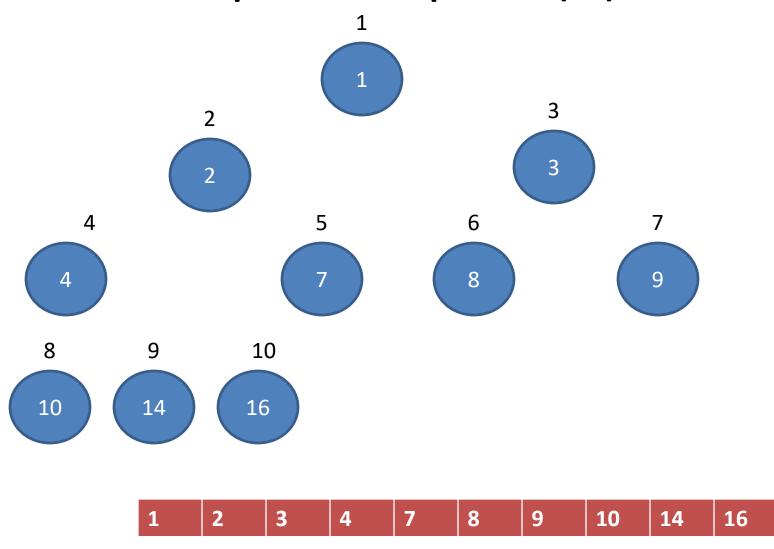












Analyze: Heapsort(A)

```
Build-Max-Heap(A)

for i = length[A] downto 2

do exchange A[1] and A[i]

heap-size[A] = heap-size[A] -1

Max-Heapify(A,1)
```

Times
O(n)
n
n-1
n-1
n-1 O(lg n)

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

Practice: Heapsort

8 | 17 | 12 | 15 | 92 | 16 | 11 | 52 | 41

Priority Queues

- A priority queue is a data structure for maintaining a set S of elements, each with an associated value called a key.
- There are two kinds of priority queues:
- A max priority queue supports these operations:
 - Insert(S,x) -> inserts the element x into the set S.
 - Maximum(S) -> returns the element of S with the largest key.
 - Extract-Max(S) -> removes and returns the element of S with the largest key.
 - Increase-Key(S,x,k) -> increases the value of element x's key to the new value k which is assumed to be as large as x's current key value.
- A min priority queue supports these operations:
 - Insert(S,x) , Minimum(S) , Extract-Min(S), Decrease-Key(S,x,k)

Pseudo-code: Heap-Maximum(A)

return A[1]

Running time = O(1)

Running time = O(lg n)

```
Pseudo-code: Heap-Increase-Key(A,I,Key)
if key < A[i]
  then error "new key is smaller than current key"

A[i] = key
while i > 1 and A[parent(i)] < A[i]
      do exchange A[i] and A[parent(i)]
      i = parent(i)</pre>
```

Running time = O(lg n)

Pseudo-code: Max-Heap-Insert(A,key)

heap-size(A) = heap-size(A)+1

A[heap-size[A]] = $-\infty$

Heap-Increase-Key(A, heap-size[A], key)

Running time = O(lg n)

