

## Heapsort

Heapsort introduces another algorithm design technique using a data structure "heap" to manage information.

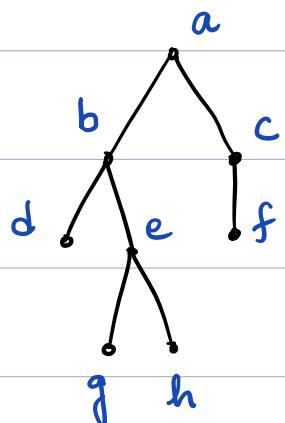
Running time :  $O(n \log n)$

## Plan

- Review of Some basic definitions
- Heaps
- Heap Sort.

## Binary tree:

It is a tree in which every node/vertex has at most two children.



Refer to any standard text book for the terminology.

'a' : root node

b is the Parent of d

d and e are children of b

d, g, h, f are leaves [leaf nodes] [Nodes with no child]

a, b, c, d, e are internal nodes [Nodes with at least one child]

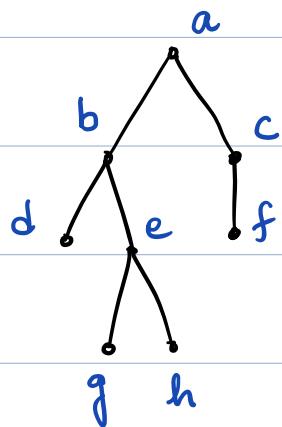
b is an ancestor of g

f is a descendant of a

Def:

The height of a node in a binary tree is the number of edges on the longest simple downward Path from the node to a leaf. Height of the tree is height of the root.

Example:



- ① Height of the node g is zero and b is two
- ② Height of the tree is Three.

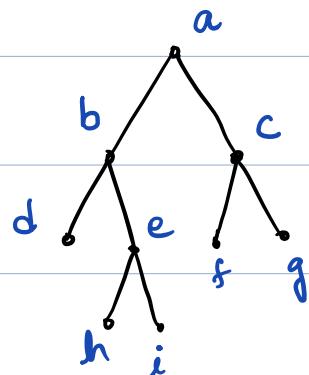
The depth of a node is the number edges from the node to the tree's root.

- ① Depth of the node e is two.
- ② Depth of root node is zero.

$$\text{Level of node} = \text{Its depth} + 1$$

## Complete Binary tree :

A complete binary tree is a binary tree  
in which every level, except possibly the last level.  
is completely filled.



Heap (or Binary Heap) : It is a Complete binary

tree with two additional constraints.

- ① A heap is complete binary tree that is all levels of the tree, except possibly last are filled and if the last level of the tree is not complete, the nodes of that level are filled from left to right.

- ② Heap Property : The key (or value) stored at each node is either greater than or equal to or less than or equal to the keys stored at the children.

There are two kinds of binary heaps

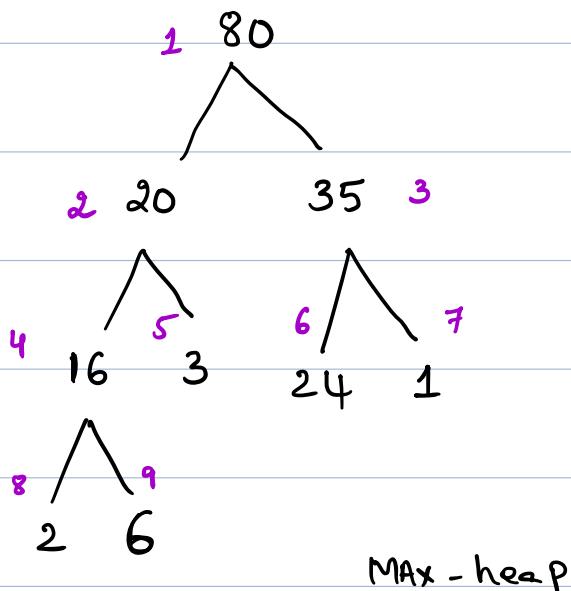
① Max-heaps : For every node  $i$ , other than the root

$$A[\text{Parent}(i)] \geq A[i]$$

i.e., the value of a node is at most the value  
of its parent.

The maximum element in a max-heap is at the root.

Example :

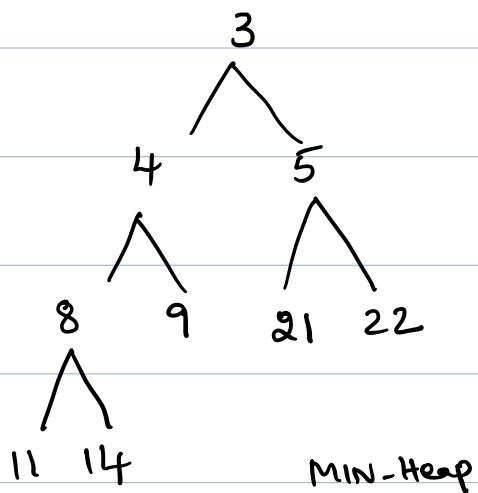


② Min-heaps: For every node  $i$ , other than the root

$$A[\text{Parent}(i)] \leq A[i]$$

The minimum element in a min-heap is at the root.

Example:



We mainly focus on Max-heaps in this class.

### Short Quiz:

① What are the minimum and maximum number of elements in a complete binary tree of height  $h$ ?

$$2^h \leq \# \text{ of elements} \leq 2^{h+1} - 1$$

② What is the height of a complete binary tree which has  $n$ -nodes?

$$\lfloor \log n \rfloor$$

③ In an  $n$ -element heap, What are the indices of the leaf nodes?

Ans  $I = \left\{ \left\lfloor \frac{n}{2} \right\rfloor + 1 \dots n \right\}$

Suppose  $i$  is an index in  $I$

then its children will be at  $2i$  and  $2i+1$

$$2(i) = 2 \left\lfloor \frac{n}{2} \right\rfloor + 2 > n, \text{ not possible.}$$

Suppose  $i$  is an index with no children

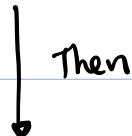
i.e.  $2i$  and  $2i+1$  are  $> n$  i.e.,  $i > \frac{n}{2}$

$$\text{i.e., } i \in \left\{ \left\lfloor \frac{n}{2} \right\rfloor + 1 \dots n \right\}$$

## Overview of the Algorithm - HEAP SORT

Q] How to use Heap (max-heap) to  
Sort the elements of an array.

If we can make a max-heap from the given array elements. Then we can easily get the maximum element. (Present at root)



Then  
Maybe again if we can make max-heap from the remaining elements we can get 2<sup>nd</sup> largest element



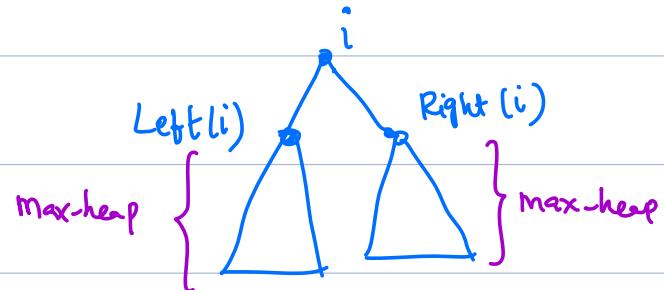
Then

the  
Repeat above steps

## Maintaining the heap Property

We describe Procedure The Max-Heapify  
to maintain the max-heap Property.

Given input an array A and an index i  
of the array. The max-heapify assumes that  
binary trees rooted at Left(i) and Right(i) are  
max-heaps, but  $A[i]$  might be smaller than its  
children ( hence violates max-heap property ).



Our goal is to maintain the heap Property.

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### MAX-HEAPIFY( $A, i$ )

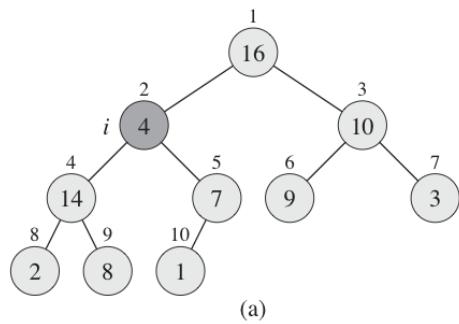
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```
1   $l = \text{LEFT}(i)$ 
2   $r = \text{RIGHT}(i)$ 
3  if  $l \leq A.\text{heap-size}$  and  $A[l] > A[i]$ 
4       $\text{largest} = l$ 
5  else  $\text{largest} = i$ 
6  if  $r \leq A.\text{heap-size}$  and  $A[r] > A[\text{largest}]$ 
7       $\text{largest} = r$ 
8  if  $\text{largest} \neq i$ 
9      exchange  $A[i]$  with  $A[\text{largest}]$ 
10     MAX-HEAPIFY( $A, \text{largest}$ )
```

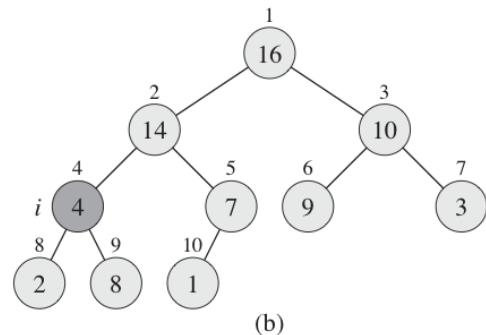
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Example: The action of Max-Heapify ( $A, 2$ )

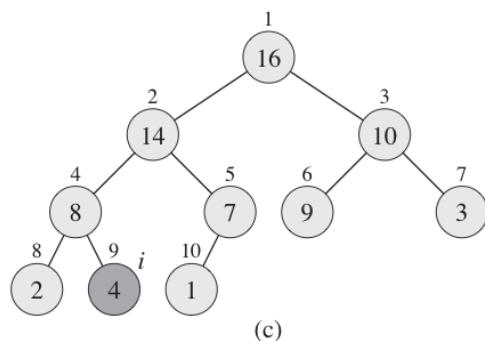
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(a)



(b)



(c)

### Running time:

① To compare  $A[i]$ ,  $A[Left(i)]$  and  $A[Right(i)]$

take  $\Theta(1)$  time at given node  $i$ .

② We have to repeat the above task at most

$O(h)$  time,  $h$  is the <sup>height</sup> of the node  $i$ .

$\therefore$  the running time of MAX-HEAPIFY is  $O(h) = O(\log n)$

## Building a heap:

Q: Given an array  $A[1,..n]$ , how to build a max-heap representing  $A$ ?

idea is to use Max-heapify in a bottom-up manner

to convert an array into a max-heap.

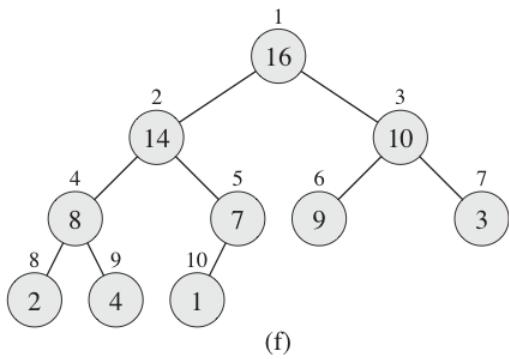
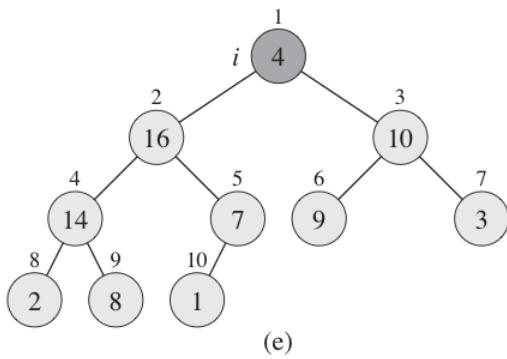
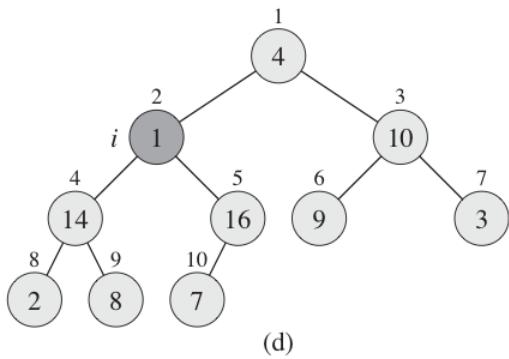
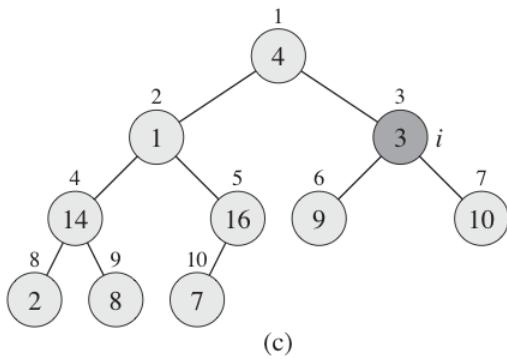
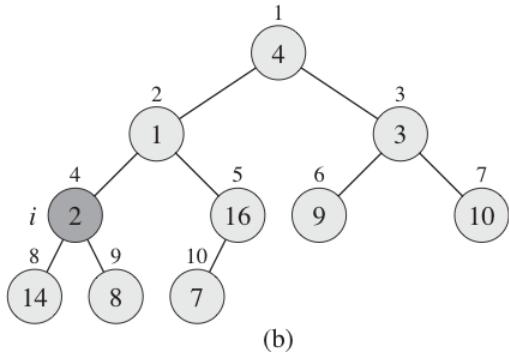
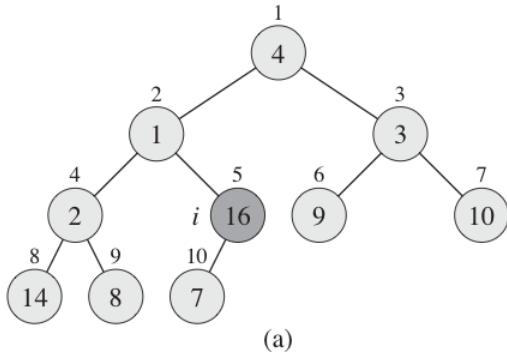
Note: The elements in the Subarray  $A[\lceil \frac{n}{2} \rceil + 1, .. n]$  are all leaves of the tree, so each is a 1-element heap.

BUILD-MAX-HEAP( $A$ )

```
1  $A.\text{heap-size} = A.\text{length}$ 
2 for  $i = \lfloor A.\text{length}/2 \rfloor$  downto 1
3   MAX-HEAPIFY( $A, i$ )
```

Example

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



Running time:

BUILD-MAX-HEAP( $A$ )

- 1  $A.\text{heap-size} = A.\text{length}$
- 2 **for**  $i = \lfloor A.\text{length}/2 \rfloor$  **downto** 1 →  $O(n)$
- 3     MAX-HEAPIFY( $A, i$ ) →  $O(\log n)$

From the above, the running time of the algorithm  
is  $O(n\log n)$ . As we will see next this is  
not asymptotically tight.

We will derive a better upperbound on running time.

The time for Max-Heapify at a node depends

on the height of the node in the tree.

Recall:

① An  $n$ -element heap has height  $\lfloor \log n \rfloor$

② An  $n$ -element heap has at most  $\left\lceil \frac{n}{2^{h+1}} \right\rceil$  nodes of any height  $h$ .

The time required for max-heapify when called on a node of height  $h$  is  $O(h)$ , so the running time of Build-Max-heap is

$$\sum_{h=0}^{\lfloor \log n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h) = O\left(n \sum_{h=0}^{\lfloor \log n \rfloor} \frac{h}{2^h}\right) \text{---(1)}$$

$$\sum_{h=0}^{\infty} \frac{h}{2^h} = \frac{\gamma_2}{(1-\gamma_2)^2} = 2$$

$$\left\{ \begin{array}{l} \sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2} \\ |x| < 1 \end{array} \right\}$$

From ① we get

$$O\left(n \sum_{h=0}^{\lfloor \log n \rfloor} \frac{h}{2^h}\right) = O\left(n \sum_{h=0}^{\infty} \frac{h}{2^h}\right)$$
$$= O(n).$$

i.e., We can build a max-heap from an  
unordered array in linear time.

## Heapsort algorithm

IDEA: Given an array  $A$ , we build a max-heap and then maximum element of the array is stored at the root  $A[1]$ . We exchange  $A[1]$  with  $A[n]$  and discard node  $n$  from the heap.

Now the <sup>new</sup> root element may violate the max-heap property. We use max-heapify to restore the max-heap property.

The heapsort algorithm repeats the above process for the max-heap of size  $n-1$  to heap of size 2.

HEAPSORT( $A$ )

- 1 BUILD-MAX-HEAP( $A$ )  $\rightarrow O(n)$
- 2 for  $i = A.length$  down to 2  $\rightarrow n-1$  times
- 3 exchange  $A[1]$  with  $A[i]$
- 4  $A.heap-size = A.heap-size - 1$
- 5 MAX-HEAPIFY( $A, 1$ )  $\rightarrow O(\log n)$

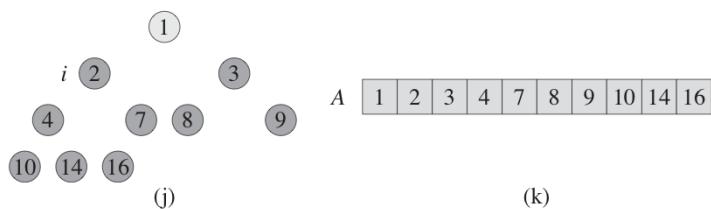
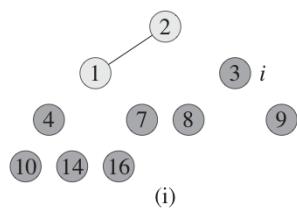
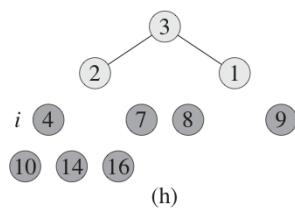
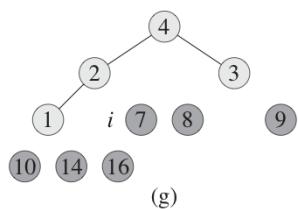
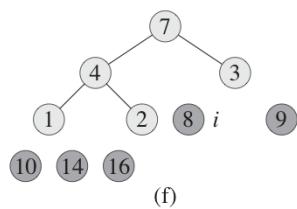
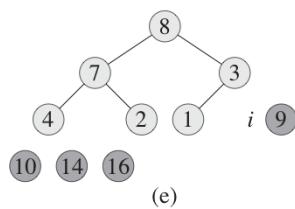
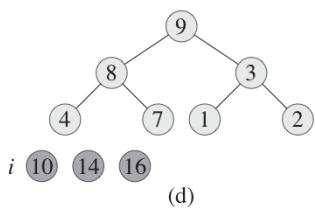
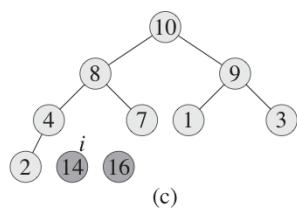
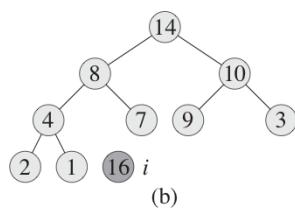
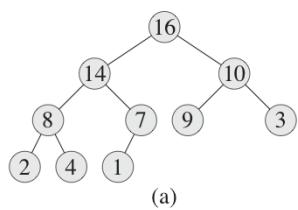
Running time:

Build-max-Heap Procedure takes  $O(n)$  time.

Max-heapify called  $(n-1)$  times.

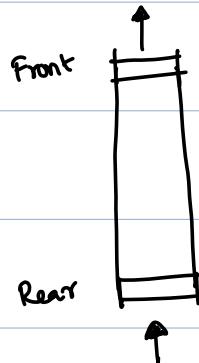
$$\begin{aligned} \text{So total running time} &= O(n) + (n-1) O(\log n) \\ &= O(n \log n) \end{aligned}$$

Example



## Priority Queue: (max-priority queue)

Queue: First in first out



A Priority queue is different from queue, instead of being a first-in first-out, values come out in order by priority.

Priority queues are used in scheduling processes in a computer and they are used in some algorithms like Prim's and Huffman codes

Some implementation:

First Idea

$n$  - # of elements

Array of unordered elements:

Insert: Constant time, we only have insert it at  $n$  and increment  $n$ .

Maximum: returns the highest priority element.

This will take  $O(n)$  time { we have to scan the whole array }

Extract-Max: Removes and returns the highest priority element.

Takes  $O(n)$  time : first find the highest priority element  
then swap it with last element  
and decrement  $n$ .

Second Idea:-

lowest to highest Priority elements.

Sorted array: Array is Sorted based on Priorities.

Insert:  $O(n)$  time: ①  $O(\log n)$  steps to find the place  $i$  where it belongs,

② We need to Shift elements to make space for the insertion, This takes  $O(n)$  time.

Maximum:  $O(1)$  time

Extract max:  $O(1)$  time ( Since highest Priority element is at the end of the array )

Next idea to use HEAPS. (

Heaps are used to implement a Priority queue

### Operations in Priority queue

① Insert - Insert a new element in the Queue

② Maximum (minimum) - gives the maximum element from the Priority Queue

③ Extract maximum (minimum) - To remove and return the maximum element from the Priority Queue

④ Increase | Decrease key

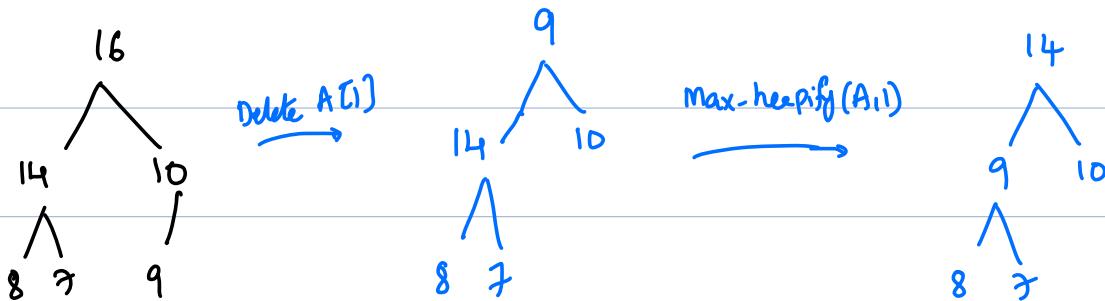
To increase / decrease key or value of any element in the Queue.

Heap as Priority Queue : [max-heap]

Maximum :  $O(1)$  time

Extract max : Swap the root  $A[1]$  with  $A[A.heap-size]$

and decrease the size then call  $\text{max-Heapify}(A, 1)$



HEAP-EXTRACT-MAX( $A$ )

```
1 if  $A.heap-size < 1$ 
2   error "heap underflow"
3  $max = A[1]$ 
4  $A[1] = A[A.heap-size]$ 
5  $A.heap-size = A.heap-size - 1$ 
6 MAX-HEAPIFY( $A, 1$ )
7 return  $max$ 
```

} Constant  
 $O(\log n)$

Running time:  $O(\log n)$

Increase key: increases the key of element  $A[i]$

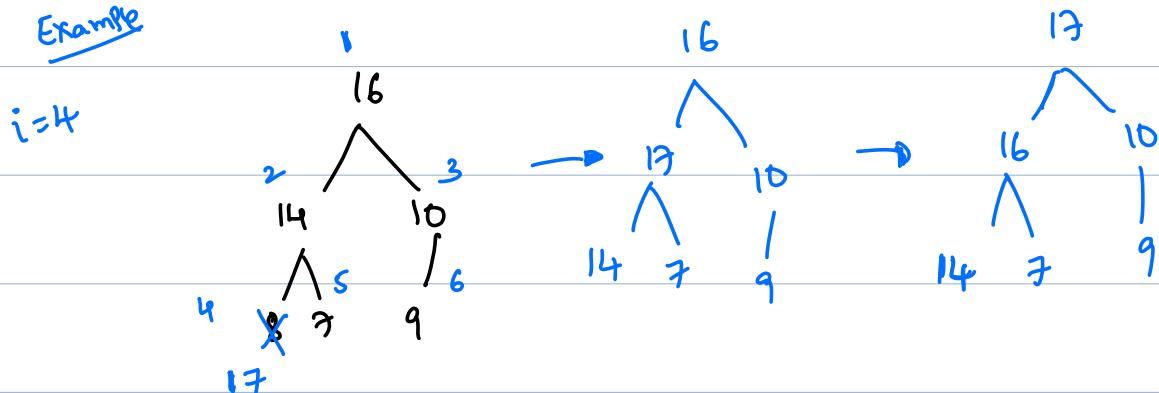
to its new value.

Increasing the key of  $A[i]$  might violate  
the max-heap Property.

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

- 1 **if**  $key < A[i]$
- 2       **error** "new key is smaller than current key"
- 3        $A[i] = key$
- 4 **while**  $i > 1$  and  $A[\text{PARENT}(i)] < A[i]$
- 5       exchange  $A[i]$  with  $A[\text{PARENT}(i)]$
- 6        $i = \text{PARENT}(i)$

Example



Running time:  $O(\log n)$

Similarly we can implement Decrease key.

Insert:

MAX-HEAP-INSERT( $A, key$ )

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

Running time:  $O(\log n)$