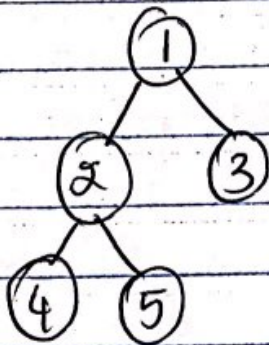


# Heaps

- It is a tree with some special properties.
- Basic requirements of a heap is
  - \* \* \*  $\rightarrow$  ① value of a node must be  $\geq$  (or  $\leq$ ) to the values of its children (Applies to all nodes)
  - \* \* \* ② This is also called heap property.
  - \* \* \* ③ All leaves should be at  $h$  or  $h-1$  levels i.e., complete binary trees.
- \* heap should form a Complete BT

## Valid heaps

①

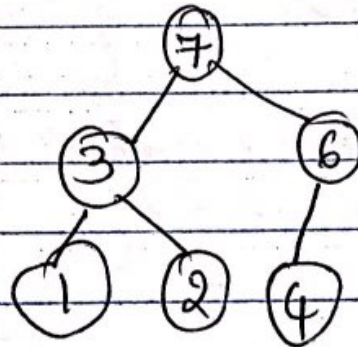


$$1 < 2, 3$$

$$2 < 4, 5$$

leaves at  $h$  &  $h-1$ .

②



$$7 > 3, 6$$

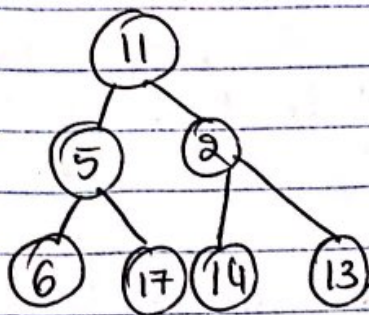
$$3 > 1, 2$$

$$6 > 4$$

leaves at  $h$  &  $h-1$ .



## Invalid heap

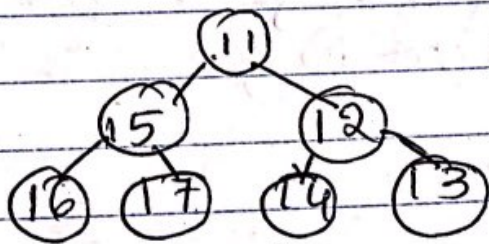


diff operations (should be same).  
11 > 5, 2 } All parent nodes must follow the same property.  
5 < 6, 17 }  
2 < 14, 13 }  
∴ ~~Not~~ Not a valid heap.

## Types of heaps

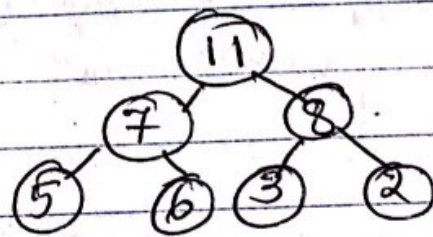
### 1) Min heap

Value of a node must be  $\leq$  to values of child ren



### 2) Max heap

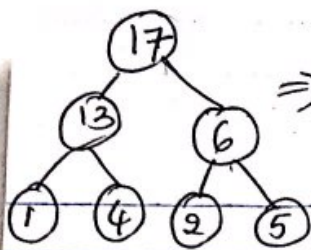
parent node  $\geq$  child ren



## Binary Heaps

- Each node can have up to 2 children.
- Implemented using arrays [unlike BT's where BT's are impl using list [stacks]].
- Heaps are complete BT's, so there won't be any wastage of locations.





⇒ In a heap, this is represented as

0	1	2	3	4	5	6
17	13	6	1	4	2	5

## Declaration of Heap -

```

public class Heap {
    public int[] arr;
    // int Count;
    // int Capacity;
    // int heap-type;
    public Heap (int Capacity, int heap-type);
    public Parent (int Capacity, int heap-type);
    public int LeftChild (int i);
    public int RightChild (int i);
    public int GetMax (int i);
    public int getMin (int i);
}
  
```

## Parent of a Node .

For Node at  $i^{\text{th}}$  location,  
its parent is at  $\left(\frac{i-1}{2}\right)^{\text{th}}$  location

$$\boxed{\text{Node at } i = \text{parent at } \left(\frac{i-1}{2}\right)^{\text{th}} \text{ location}}$$

In the above eg, Node 6 is at location 2,  
∴ its parent (17) is at  $\frac{2-1}{2} = \frac{1}{2} \approx 0$   
0<sup>th</sup> location



main operations :- 1) getMin() / getMax()  
2) extractMin() / extractMax()  
3) insert() <sup>delete</sup>

### Children of a Node

For parent at  $i^{th}$  location,

- \* Left child will be at  $(2i+1)^{th}$  location
- \* Right child will be at  $(2i+2)^{th}$  location

In the above ex.

Node 6 at location 2

- Left child (2) is at  $2 \times 2 + 1 = 5$
- Right child (5) is at  $2 \times 2 + 2 = \underline{\underline{6}}$

### Getting the Max element

- \* Max element in max heap is always at root, i.e., at array[0].

### Heapifying an Element

- The process of adjusting the location of the inserted element into heap & make it a heap again is called heapifying.

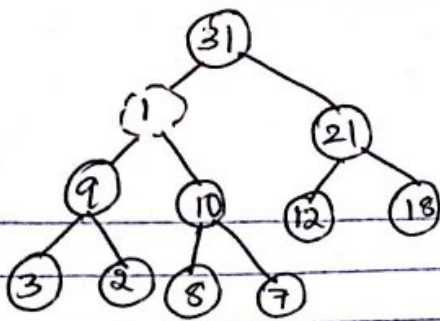
In max heap,

• To heapify an element, ~~find~~

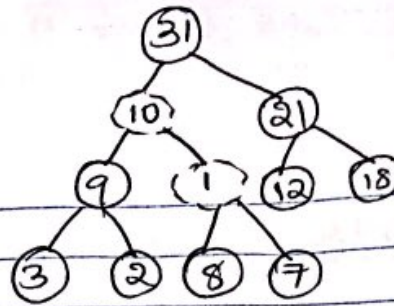
① find the maximum of its children & swap it with current element

② Continue until heap property is satisfied at every node in the heap.

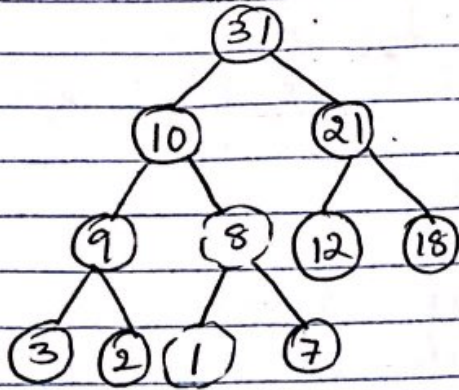




⇒



⇒



In heapifying process,  
1) Moving from top to bottom  
is called Percolate down.

2) Moving from bottom to top  
is called Percolate up.

TC of heapifying is  $O(\log n)$   
SC  $O(1)$

## Deleting an element

- 1) Copy first element into some variable
- 2) Copy last element into first element location
- 3) Percolate down [top to bottom heapify] the first element.

TC ⇒  $O(\log n)$

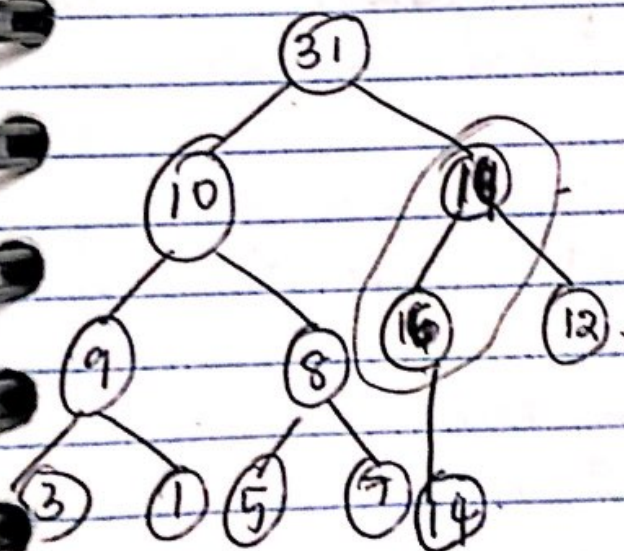
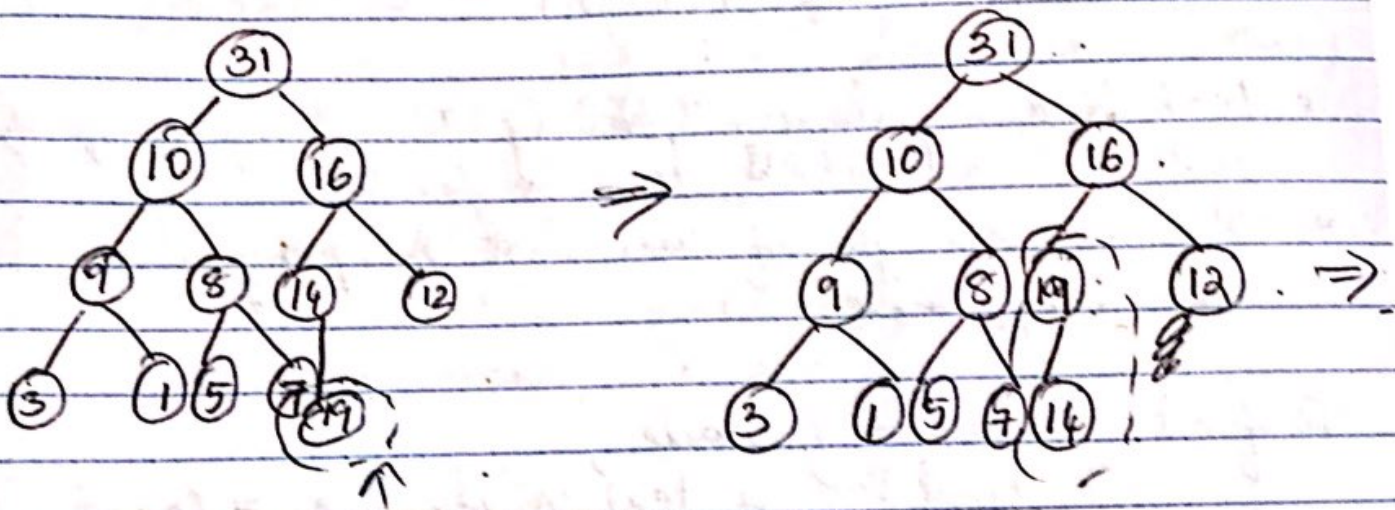
1<sup>st</sup> point above takes  $O(1)$  TC.  
2<sup>nd</sup> point — " —  $O(1)$  TC  
3<sup>rd</sup> point — " —  $O(\log n)$  TC,  
for heapifying



Deleting an element uses percolate down  
Inserting an element uses percolate up.

### Inserting an element

- 1) Increase heap size.
- 2) Keep new element at the end of heap.
- 3) Heapify the element from bottom to top (percolate up).



Insertion takes

$$O(1) + O(1) + O(\log n) \\ = O(\log n)$$



## Destroying heap.

Set, count = 0 & array = null.

## Heapifying the array.

- 1) Take  $n$  input items & place them into an empty heap.
  - 2) Takes  $O(n \log n)$  time in worst case,  $\because$   $n$  inserts at  $O(\log n)$  time each.
- Leaf nodes always satisfy the heap property and do not need to care for them.
  - It is enough if we just heapify the non-leaf nodes.

To find non-leaf node,  
 $\Rightarrow$  Find the a leaf node  $[h = \text{count} - 1]$   
 $\Rightarrow$  parent of that leaf node.  
$$= \frac{h-1}{2}$$

TC :



# HeapSort

- Heap Sort algo, inserts all elements (from an unsorted array) into a heap, then removes them from the root of that heap until the heap is empty.
- Heap Sort can be done in place with the array to be sorted.
- An Alternative - Instead of deleting an ~~element~~ element,
  - \* exchange the first element (max) with last element & reduce heap size (array size)
  - \* ~~Then~~ Then we heapify 1<sup>st</sup> element
  - \* Continue until no. of remaining elements is one.

TC :  $O(\log n)$

As we remove elements from the heap, values become sorted

- \* Why heaps are implemented using arrays & not trees? [like BT using stacks]
- 1) Lower mem usage (no need to store 3 pointers)
  - 2) Easier mem management (just one object allocated, rather than  $N$ )
  - 3) Better locality of reference (Hence in heap are relatively closer together in mem)



~~But~~ Using arrays is a good choice if we know that the data is constant.

- But in reality, not all BT's are completely full & perfectly balanced.

→ If array-bound BT is <sup>mostly</sup> empty, array space is wasted

→ If the tree needs to grow deeper, then we ~~as~~ this array has to be copied to a larger array, this is time expensive.