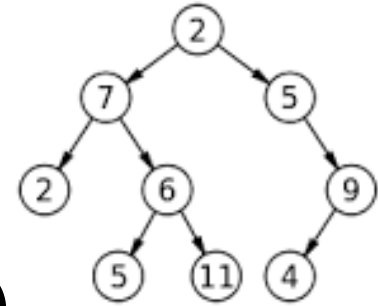


Binary Heap

- Binary Tree
- Strictly Binary Tree (Full Binary Tree)
- Complete Binary Tree

Binary Heap



➤ Binary Tree

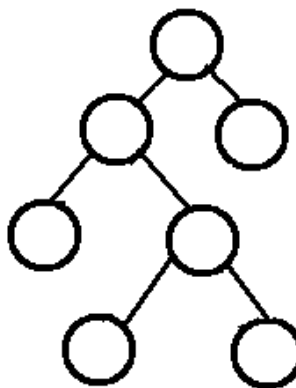
- Any node can have at most 2 children

➤ Strictly Binary Tree (Full Binary Tree)

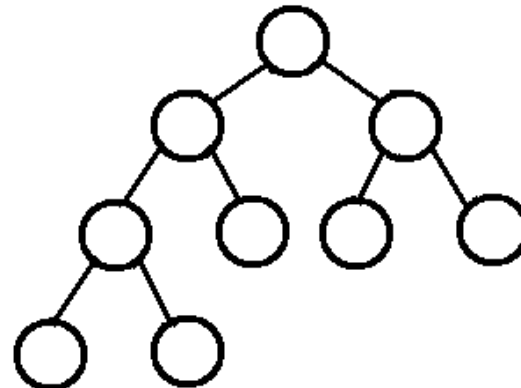
- A binary tree is a full binary tree if every node has 0 or 2 children.

➤ Complete Binary Tree

- a binary tree in which every level, except possibly the last, is completely filled, and all nodes are as far left as possible.



full tree



complete tree

Binary Heap

(Binary) Heaps

A binary tree that stores priorities (or priority-element) pairs at nodes

Structural property:
All levels except last are full. Last level is left-filled.

Heap property:
Priority of node is at least as large as that of its parent.



Binary Heap

Examples of non-Heaps

□ Heap property violated



Binary Heap

Example of non-heap

□ Last level not left-filled



Binary Heap

Finding the minimum element

- The element with smallest priority always sits at the root of the heap.
- This is because if it was elsewhere, it would have a parent with larger priority and this would violate the heap property.
- Hence minimum() can be done in $O(1)$ time.

Binary Heap

Height of a heap

- Suppose a heap of n nodes has height h .
- Recall: complete binary tree of height h has $2^{h+1}-1$ nodes.
- Hence $2^h-1 < n \leq 2^{h+1}-1$.
- ~~$n = \lfloor \log_2 h \rfloor$~~ $\leftarrow \text{Error, } h = \lfloor \log_2 n \rfloor$

Binary Heap

Implementing Heaps

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

Left (i)
return $2i$

Right (i)
return $2i+1$



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

Binary Heap

Implementing Heaps (2)

- Notice the implicit tree links; children of node i are $2i$ and $2i+1$
- Why is this useful?
 - In a binary representation, a multiplication/division by two is left/right shift
 - Adding 1 can be done by adding the lowest bit

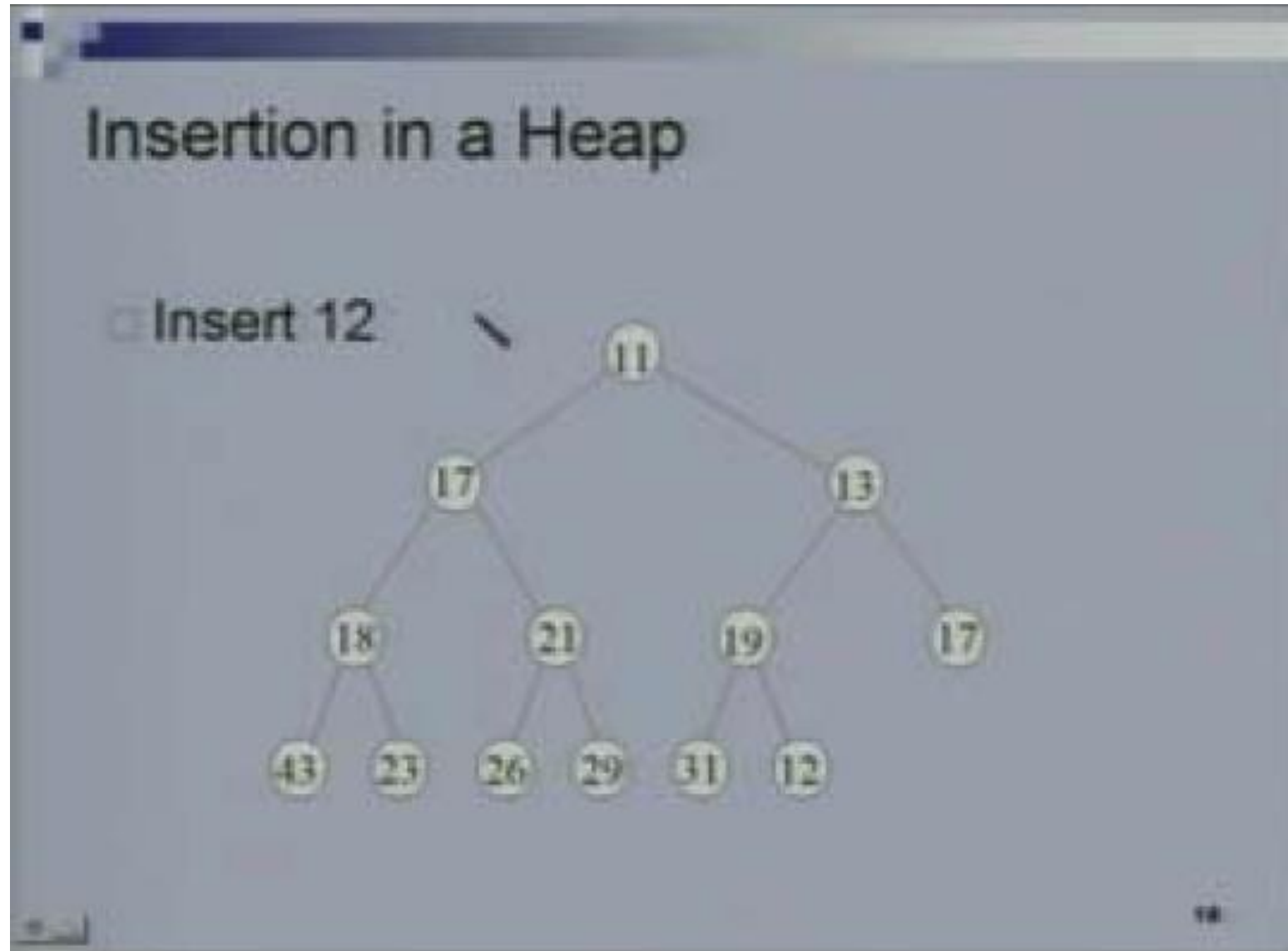
Binary Heap

Insertion in a Heap

Insert 12



Binary Heap



Binary Heap

Insertion in a Heap

□ Insert 12



Binary Heap

Insertion in a Heap

Insert 12



Binary Heap

Insertion in a Heap

- Insert 12
- Insert 8



Binary Heap

Insertion in a Heap

□ Insert 12

□ Insert 8



Binary Heap

Heapify

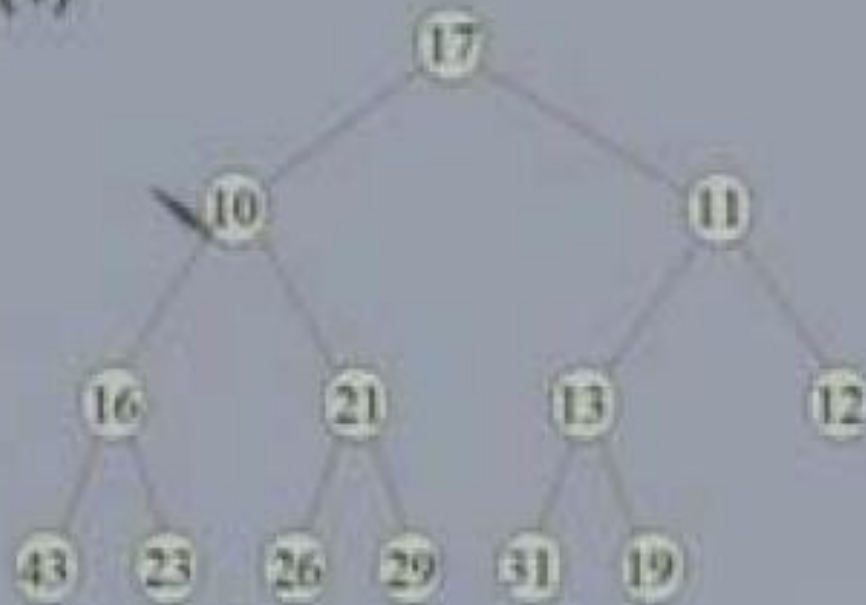
- i is index into the array A
- Binary trees rooted at $\text{Left}(i)$ and $\text{Right}(i)$ are heaps
- But, $A[i]$ might be smaller than its children, thus violating the heap property
- The method **Heapify** makes binary tree rooted at i a heap by moving $A[i]$ down the heap.

larger

Binary Heap

Heapify

- Heap property violated at node with index 1 but subtrees rooted at 2, 3 are heaps.
- `heapify(1)`



Binary Heap

Running time Analysis

- A heap of n nodes has height $O(\log n)$.
- While **inserting** we might have to move the element all the way to the top.
- Hence at most $O(\log n)$ steps required.
- In **Heapify**, the element might be moved all the way to the last level.
- Hence **Heapify** also requires $O(\log n)$ time.

Binary Heap

Binary Heaps (A Few More Operations)

- Delete-min
- Building a heap in $O(n)$ time
- Heap Sort

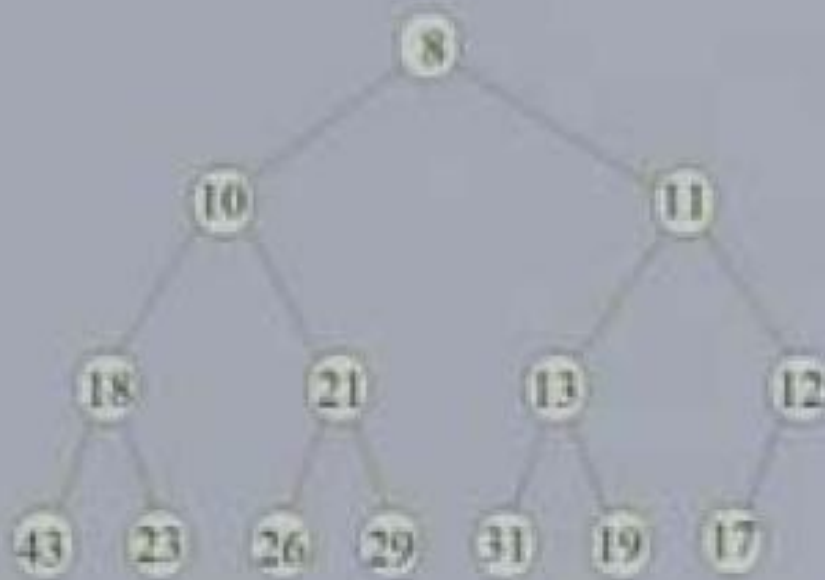
Binary Heap

Delete-min

- ❑ The minimum element is the one at the top of the heap.
- ❑ We can delete this and move one of its children up to fill the space.
- ❑ Empty location moves down the tree.
- ❑ Might end up at any position on last level.
- ❑ Resulting tree would not be left filled.

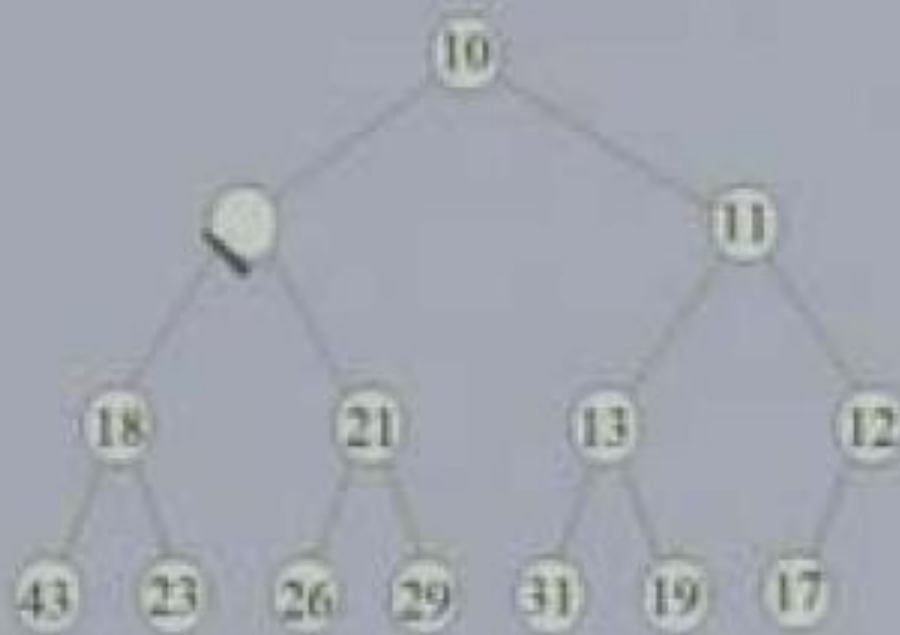
Binary Heap

Delete-min in a Heap



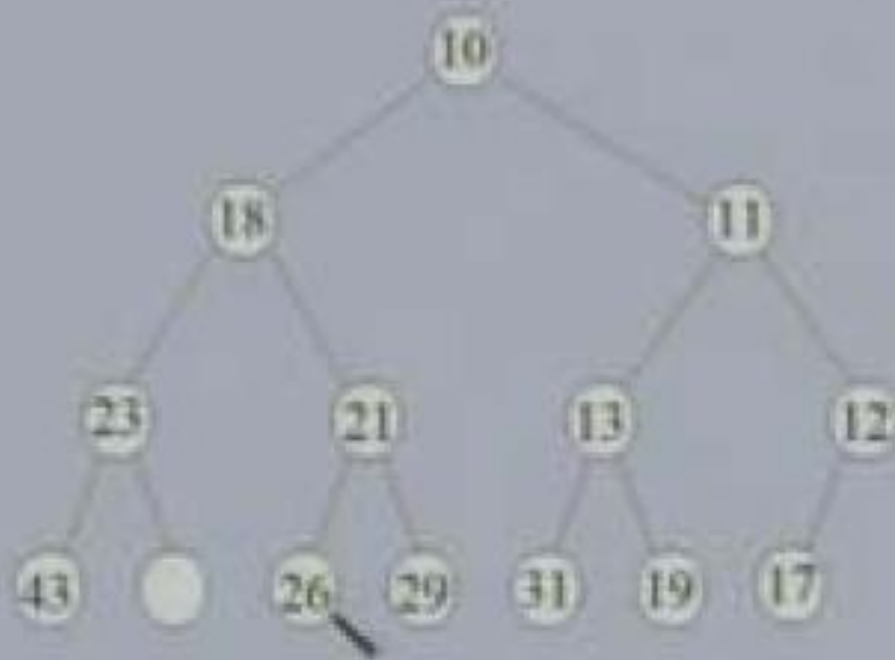
Binary Heap

Delete-min in a Heap



Binary Heap

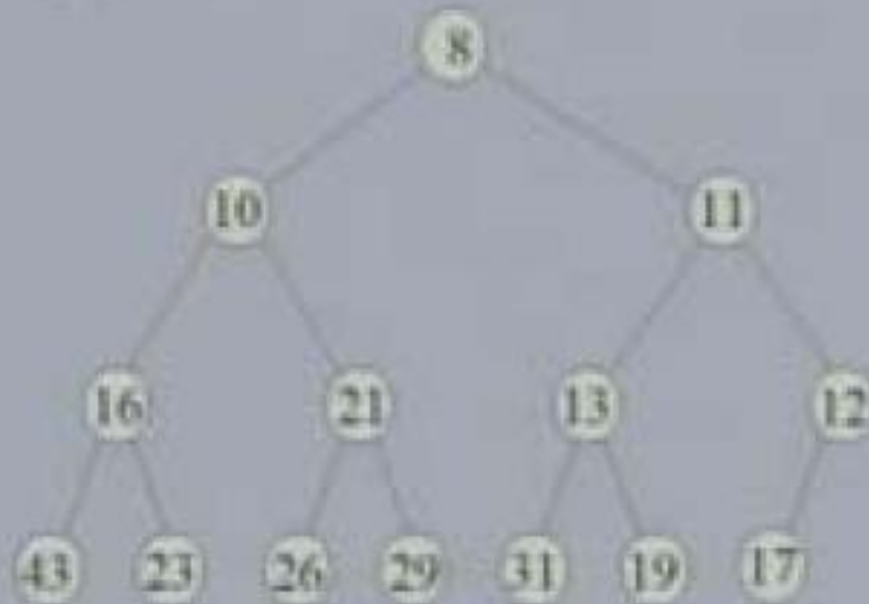
Delete-min in a Heap



Binary Heap

Delete-min in a Heap (2)

- Replace top element with last element of heap.
- And then Heapify(1)



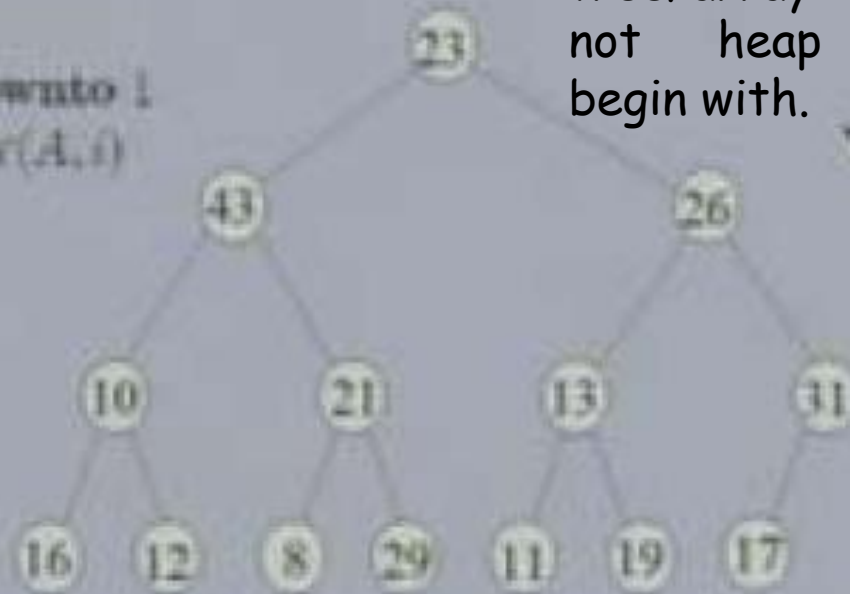
Binary Heap

Building a heap

- We start from the bottom and move up
- All leaves are heaps to begin with

BUILD-HEAP(A)


```
1 for  $i \leftarrow \lfloor n/2 \rfloor$  downto 1  
2   do HEAPIFY( $A, i$ )
```



Note that given tree/array is not heap to begin with.

Binary Heap

Building a Heap: Analysis

- Correctness: induction on i , all trees rooted at $m > i$ are heaps
- Running time: n calls to Heapify = $n O(\lg n) = O(n \lg n)$  $\left\lfloor \frac{n}{2} \right\rfloor$
- We can provide a better $O(n)$ bound.
 - Intuition: for most of the time Heapify works on smaller than n element heaps

Binary Heap

- We can provide a better $O(n)$ bound
 - Intuition
 - We can derive a tighter bound by observing that the time for HEAPIFY to run at a node varies with the height of the node in the tree, and the heights of most nodes are small.
 - Our tighter analysis relies on the properties that an n -element heap has height $\lceil \lg n \rceil$ and at most $\lceil n/2^{h+1} \rceil$ nodes of any height h .
 - The time required by HEAPIFY when called on a node of height h is $O(h)$, and so we can express the total cost of BUILD-HEAP as being bounded from above by

Binary Heap

➤ Total Number of swaps required

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h) = O\left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h}\right) \rightarrow \sum_{h=0}^{\infty} \frac{h}{2^h} = \frac{1/2}{(1 - 1/2)^2} \quad (\because \sum_{k=0}^{\infty} k x^k = \frac{x}{(1-x)^2} \text{ for } |x| < 1.)$$

$$= 2.$$

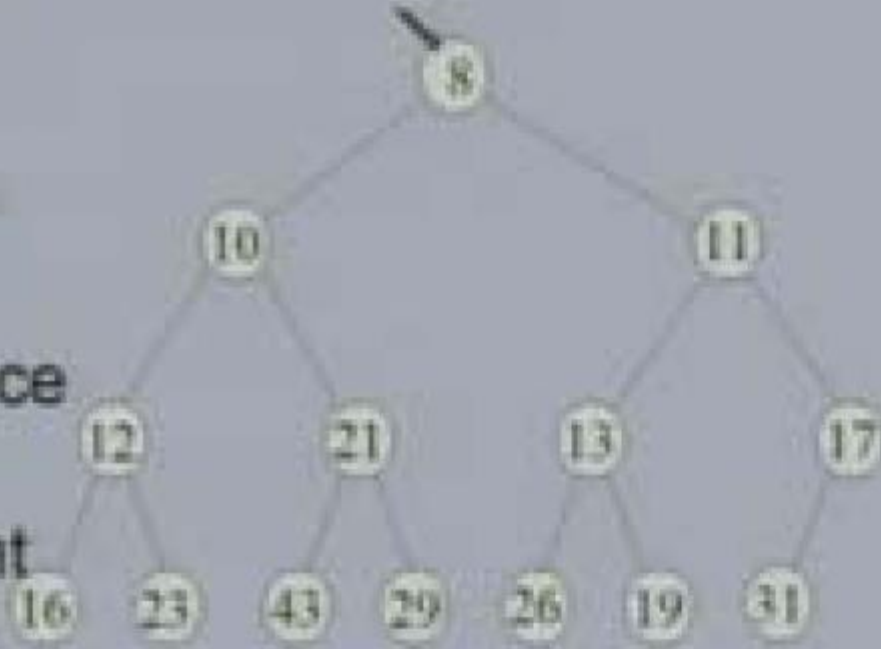
$$O\left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h}\right) = O\left(n \sum_{h=0}^{\infty} \frac{h}{2^h}\right)$$

$$= O(n).$$

Binary Heap

Heap Sort

- Create a heap.
- Do delete-min repeatedly till heap becomes empty.
- To do an in place sort, we move deleted element to end of heap.



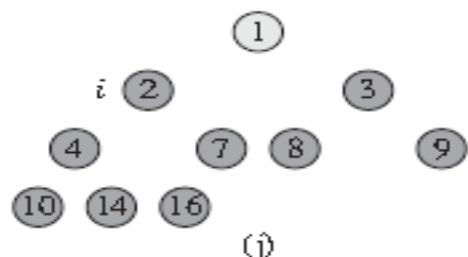
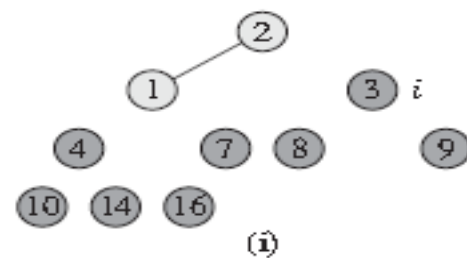
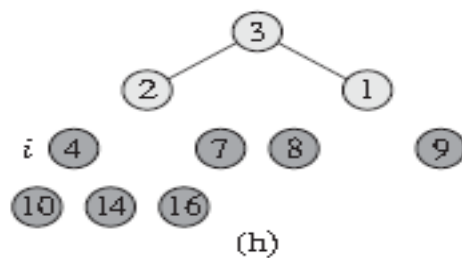
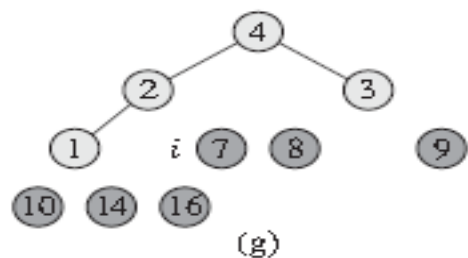
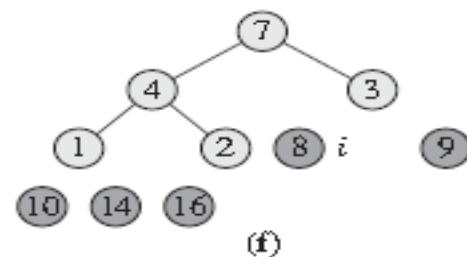
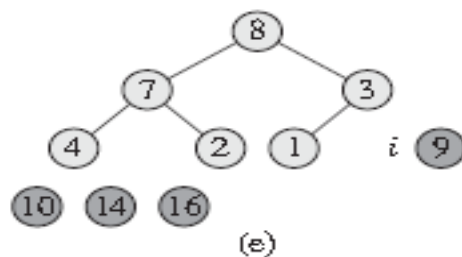
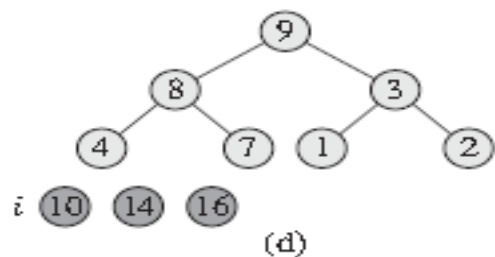
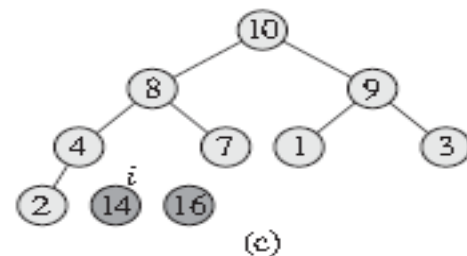
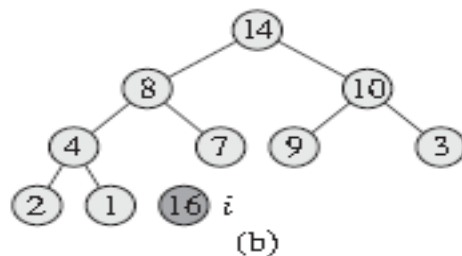
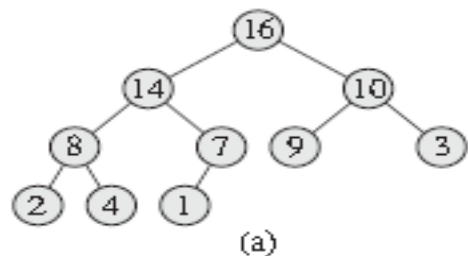
Remember that swap can be performed without temporary variable $a = a + b, b = a - b, a = a - b$

Binary Heap

HEAPSORT(A)

```
1  BUILD-MAX-HEAP( $A$ )
2  for  $i = A.length$  downto 2
3      exchange  $A[1]$  with  $A[i]$ 
4       $A.heap-size = A.heap-size - 1$ 
5      MAX-HEAPIFY( $A, 1$ )
```

Binary Heap



A

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

(k)

Binary Heap

MAX-HEAPIFY(A, i)

```
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( $A, \text{largest}$ )
```


Binary Heap

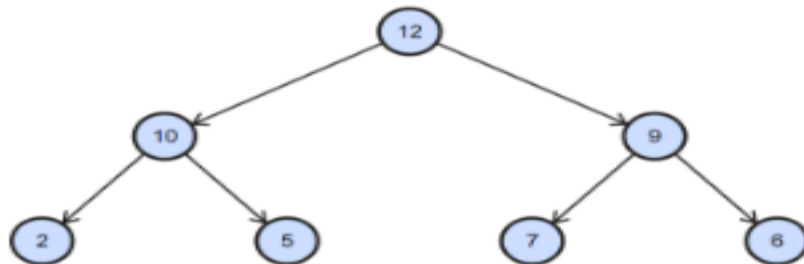
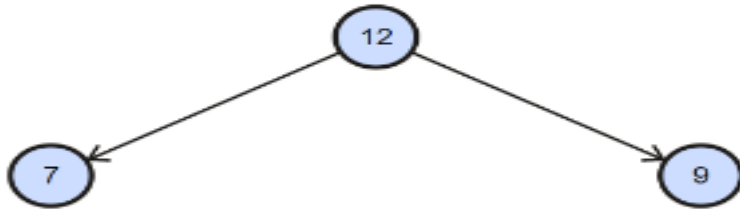
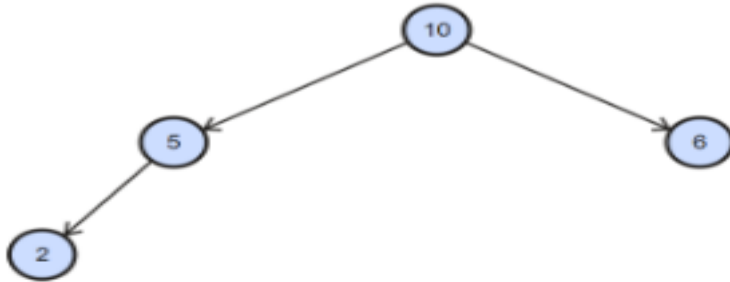
Running times of heap operations

- Insert: $O(\log n)$
- Heapify: $O(\log n)$
- Find minimum: $O(1)$
- Delete-min: $O(\log n)$
- Building a heap: $O(n)$
- Heap Sort: $O(n \log n)$

Binary Heap

➤ Merging two Binary Heaps

```
Input  : a = {10, 5, 6, 2},  
        b = {12, 7, 9}  
Output : {12, 10, 9, 2, 5, 7, 6}
```



Just put the two arrays together and create a new heap out of them which takes $O(n)$.

Binomial Heap

- The main application of Binary Heap is in implementation of priority queue.
- Binomial Heap is an extension of Binary Heap that provides faster union or merge operation together with other operations provided by Binary Heap.



| Operation | Binary ^[1] | Binomial ^[1] | Fibonacci ^{[1][2]} |
|--------------|-----------------------|-------------------------|-----------------------------|
| find-min | $\Theta(1)$ | $\Theta(\log n)$ | $\Theta(1)$ |
| delete-min | $\Theta(\log n)$ | $\Theta(\log n)$ | $O(\log n)^{[b]}$ |
| insert | $O(\log n)$ | $\Theta(1)^{[b]}$ | $\Theta(1)$ |
| decrease-key | $\Theta(\log n)$ | $\Theta(\log n)$ | $\Theta(1)^{[b]}$ |
| merge | $\Theta(n)$ | $O(\log n)^{[d]}$ | $\Theta(1)$ |

Binomial Heap

- A Binomial Heap is a collection of Binomial Trees