

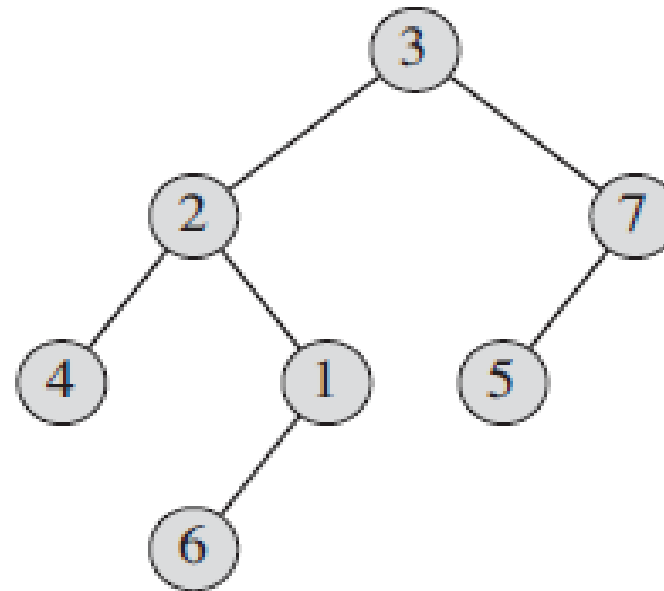
Trees – degree of a node

► The *degree* of a node x is the number of children of x

► Degree of node 3: 2

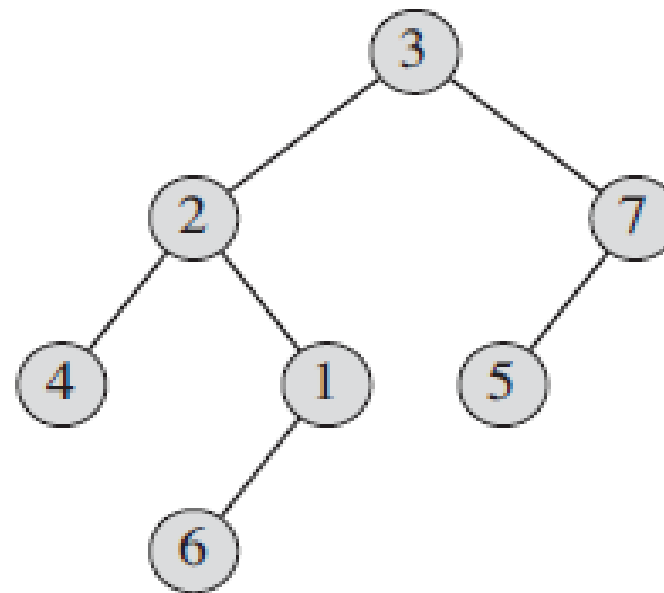
► Degree of node 7: 1

► Degree of node 6: 0



Trees – Depth of a node

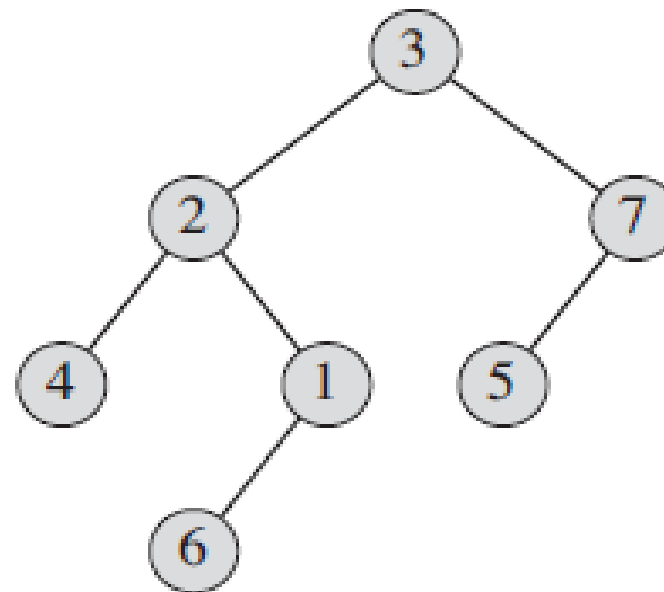
- Depth of node x - The length of the simple path from the root r to a node x
- Depth of node 6: 3
 - Depth of nodes 1, 4, and 5: 2
 - Depth of nodes 2 and 7: 1
 - Depth of node 3: 0



Trees – Levels

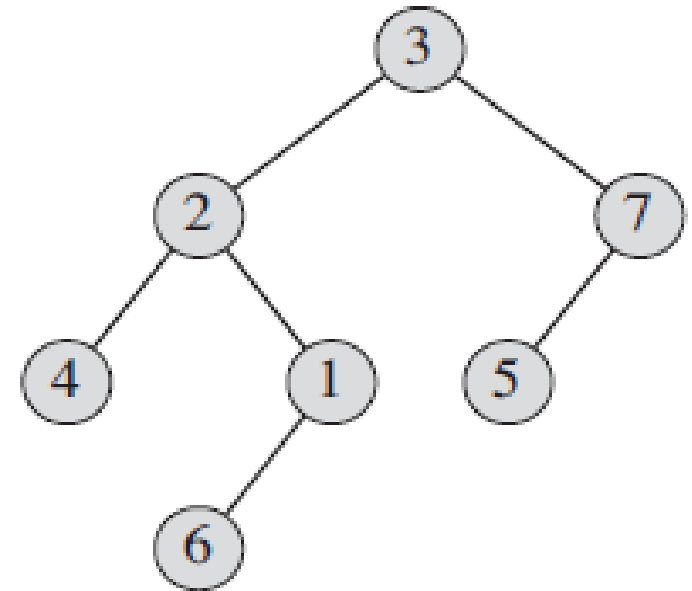
► A level of a tree consists of all nodes at the same depth.

- Root node at level 0
- Nodes at level 2: 4, 1, 5
- Nodes at levels 1 ?
- Nodes at levels 3 ?



Trees – Height of a node

- ▶ **Height of a node**
 - ▶ number of edges on the longest simple downward path from the node to a leaf
- ▶ **Height of a tree** is the height of its root
 - ▶ Height of the tree shown is 3 = length of the path from root to node the node labeled 6



Complete Binary Trees

- A binary tree in which all **leaves** have the **same depth** and all internal nodes have **degree 2**.

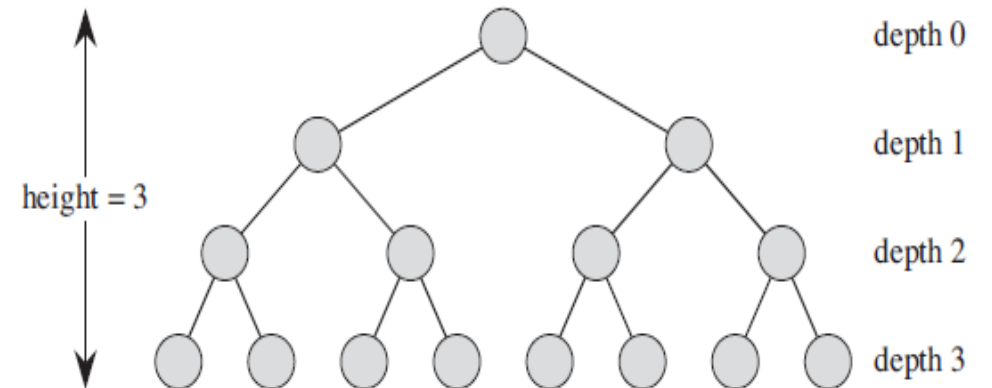
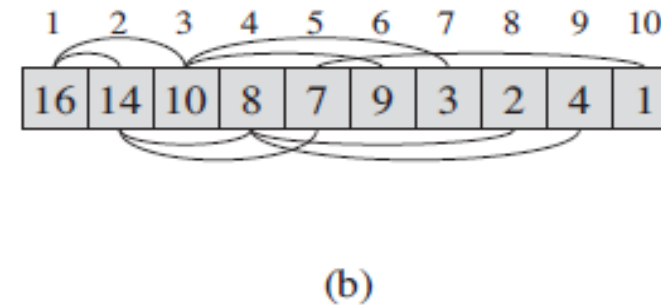
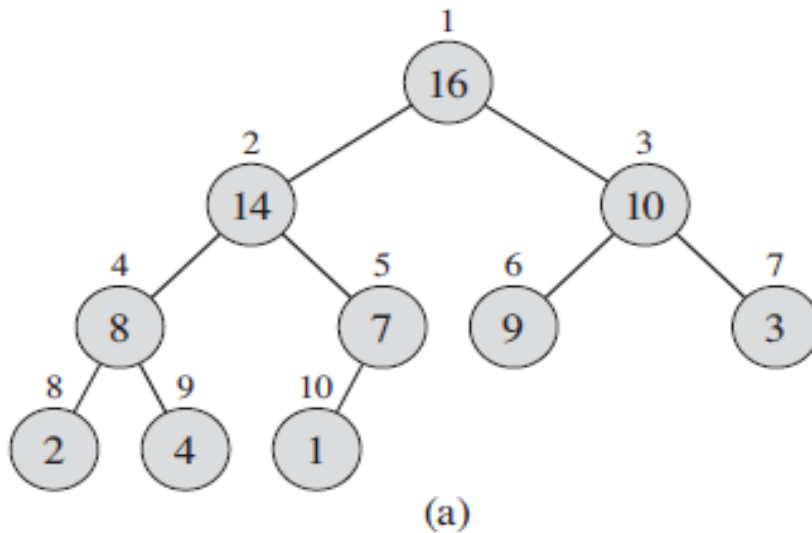


Figure B.8 A complete binary tree of height 3 with 8 leaves and 7 internal nodes.

Nearly Complete Binary Trees

- Tree is completely filled on all levels except possibly the lowest, which is filled from the left up to a point



Binary heap

- ▶ Nearly Complete Binary Tree
- ▶ Each node satisfying the Heap property
- ▶ Height of a node in a heap?
- ▶ Height of a heap?
- ▶ Heap of n elements has height $\Theta(\lg n)$

MAX-HEAPIFY

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


Running time of Max-Heapify

Running time of **MAX-HEAPIFY**

- 1) Time to fix up the relationships between **$A[i]$** , **$A[\text{LEFT}(i)]$** and **$A[\text{RIGHT}(i)]$**
- 2) Time to run **MAX-HEAPIFY** on a subtree rooted at one of the children of node **i** (assuming that the recursive call occurs)

Running time of Max-Heapify

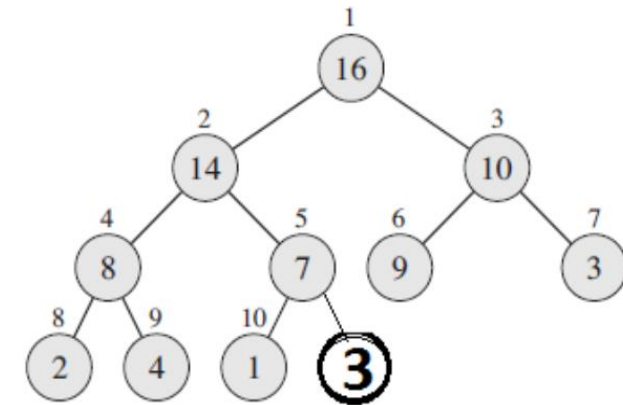
Running time of **MAX-HEAPIFY**

- 1) Time to fix up the relationships between **A[i]**, **A[LEFT(i)]** and **A[RIGHT(i)]** - $\Theta(1)$
- 2) Time to run **MAX-HEAPIFY** on a subtree rooted at one of the children of node **i** - ?

Running time of MAX-HEAPIFY

Running time of MAX-HEAPIFY on a **subtree** rooted at one of the **children of node i**:

- ▶ Need to know the size of the subtree (# nodes) rooted at one of the children of node i.
- ▶ What is the worst case size of the subtree?
 - ▶ Since heap is a nearly complete binary tree, the worst case occurs when the bottom level of the tree is **exactly half full**
 - ▶ Hence, find the maximum **number of nodes in the left subtree** of the **nearly complete binary tree**



Total number of nodes in a complete binary tree of height h is $2^{h+1} - 1$

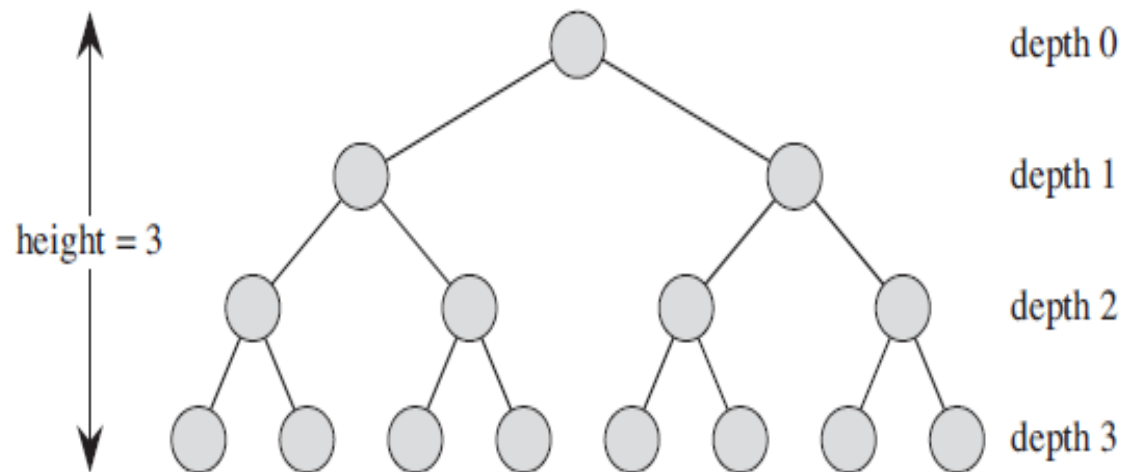
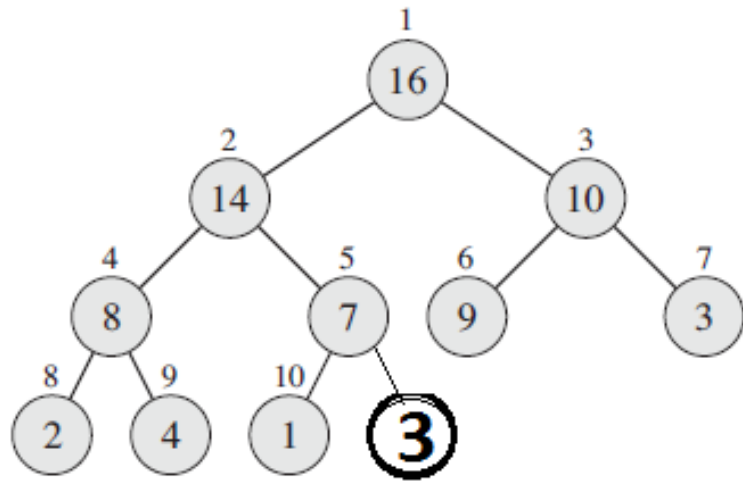


Figure B.8 A complete binary tree of height 3 with 8 leaves and 7 internal nodes.



- Total number of nodes in a nearly complete binary tree (if as in figure) of height h
 $(2^{h+1} - 1) - (2^{h-1})$
- Why do we subtract 2^{h-1} from the total number of nodes.....?

- ▶ If **h** is the **height** of the subtree, number of **leaf nodes** is 2^h

- ▶ **Eg: h=3, #leaves=8**

- ▶ **Since it is a binary tree, $2^h/2$** is the number of leaf nodes in each of the **left and right subtree**

i.e $2^h/2 = 2^{h-1}$

- ▶ Hence, total number of nodes in a nearly complete binary tree
 $n = (2^{h+1} - 1) - (2^{h-1})$

(Writing 2^{h+1} in terms of 2^{h-1})

$$n = (4 * 2^{h-1} - 1) - (2^{h-1})$$

$$= 3 * 2^{h-1} - 1$$

$$2^{h-1} = (n + 1)/3 \text{ -----(1)}$$

- 
- Maximum number of nodes in the left subtree of a nearly complete binary tree of **height h**,

$$= 2^h - 1 \text{ (writing in terms of } 2^{h-1})$$

$$= 2 * 2^{h-1} - 1 \text{ -----(2)}$$

Substitute (1) in (2)

$$= 2 ((n + 1)/3) - 1$$

$$= (2n - 1) / 3$$

$$\leq 2n/3$$

Hence, the worst case size of the subtree rooted at i, is at most $2n/3$

Children's subtrees each have size at most $2n/3$

The Running time of Max-Heapify ,

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

How to solve $T(n) = T(2n/3) + \Theta(1)$

$T(n) = c + T(2/3 * n)$ ----- (1) (Replacing $\Theta(1)$ by constant c)

$T(2/3 * n) = c + T(2/3 * 2/3 * n)$ ---(2)

Substitute (2) in (1)

$$T(n) = c + c + T(2/3 * 2/3 * n)$$

$$T(n) = 2c + T(2^2/3^2 * n)$$



Further :

$$T(n) = 3c + T(2^3/3^3 * n)$$

....

$$T(n) = kc + T(2^k/3^k * n) \text{ -----(3)}$$

$$\text{Assume } n = (3/2)^k \text{ -----(4)}$$


Substitute (4) in (3)

$$T(n) = kc + T(1)$$

$$T(n) = c \log_{3/2} n + T(1)$$

$$k \text{ is } \log_{3/2} n$$

k can be written as $\log_2 n$. How?


$$n = (3/2)^k \text{ -----(5)}$$

Apply log on both sides of (5)

$$\log_2 n = \log_2 (3/2)^k$$

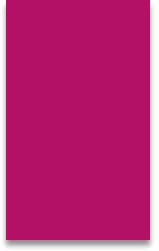
$$\log_2 n = k * \log_2 (3/2)$$

$$k = \log_2 n / \log_2 (3/2)$$

For MAX- HEAPIFY

$$T(n) = kc + T(1)$$

$$\mathbf{T(n) = O(lg\ n)}$$



- ▶ Children's subtrees each have size at most $2n/3$
- ▶ The Running time of Max-Heapify ,
 $T(n) \leq T(2n/3) + \Theta(1)$
- ▶ $T(n) = O(\log n)$
- ▶ The **running time of Max-Heapify** on a node of **height h** as **$O(h)$**

BUILD-MAX-HEAP

BUILD-MAX-HEAP(A)

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

Running time of BUILD-MAX-HEAP

- ▶ Simple upper bound
 - Each call to MAX-HEAPIFY: $O(\log n)$
 - Number of calls to MAX-HEAPIFY: $O(n)$
 - Running time : $O(n \log n)$
- ▶ But this upper bound is not asymptotically tight

Tighter Analysis of BUILD-MAX-HEAP

Based on the following **observations** and the **property** given

Observations:

- Heights of most of the nodes are small
- Running time of MAX-HEAPIFY varies with the height of the node

Property: (Exercises: Prove the following)

- An n -element heap has height $\lfloor \log n \rfloor$
- At most $\lceil n / 2^{h+1} \rceil$ number of nodes of height h in any n -element heap

Prove: At most $\lceil n / 2^{h+1} \rceil$ nodes of height h in any n -element heap

Proof by Induction on height h .

Base case:

$h = 0$

#nodes with height $h=0$ in an n -element heap $\lceil n/2 \rceil$

This is same as the #leaves in an n -element heap.

Hypothesis:

Given statement is true for height $h-1$.

i.e At most $\lceil n / 2^h \rceil$ nodes of height $h-1$ in an n -element heap

Induction Step: Prove the given statement for **height h**.

Let n_h be the number of nodes at **height h** in **n-node tree T**

Consider tree T' by removing the leaves of T

Let n' be the number of nodes in T' .

$$n' = n - n_0, \text{ where } n_0 = \# \text{leaves in } T$$

$$= n - \lceil n/2 \rceil$$

$$= \lfloor n/2 \rfloor \text{ ----- (1)}$$

#nodes at **height h** in T = #nodes at **height h-1** in T'

(T': T in which leaves are removed)

Let n'_{h-1} be the number of nodes at **height h-1** in T'

$$n_h = n'_{h-1}$$

From the induction hypothesis, we can bound n'_{h-1}

$$n_h = n'_{h-1} \leq \lceil n' / 2^h \rceil \text{ ----- (2)}$$

We saw that, $n' = \lfloor n/2 \rfloor$ in equation (1)

Substitute (1) in (2)

$$\leq \lceil \lfloor n/2 \rfloor / 2^h \rceil$$

$$\leq \lceil n / 2^{h+1} \rceil$$

Hence, the given statement is proved.

Tighter Analysis of Build-Max-Heap

- ▶ Time required for MAX-HEAPIFY when called on a node of height h is $O(h)$
- ▶ The total cost of BUILD-MAX-HEAP as being bounded from above by

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

$$\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2}$$

for $|x| < 1$.

- ▶ Summing up the above series by substituting $x = 1/2$

$$\begin{aligned} \sum_{h=0}^{\infty} \frac{h}{2^h} &= \frac{1/2}{(1-1/2)^2} \\ &= 2. \end{aligned}$$

Thus, we can bound the running time of BUILD-MAX-HEAP as

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

Linear time for Building a Max-heap

- ▶ From the tighter analysis, we can see that
 - ▶ Building a Max-heap from an unordered array takes linear time.
 - ▶ Linear in terms of the number of elements in the array

Heap Sort: ALGORITHM

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($A, 1$)

Running time of Heap sort

- ▶ Call to BUILD-MAXHEAP takes time $O(n)$
- ▶ Each of the $n-1$ calls to MAX-HEAPIFY takes time $O(\lg n)$
- ▶ Thus, HEAPSORT takes time $O(n \lg n)$

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( $A, 1$ )
```

Priority Queue (implemented using heap)

Running Time of operations

- ▶ HEAP-MAXIMUM() - $\Theta(1)$
- ▶ HEAP-EXTRACT-MAX() - $O(\lg n)$
- ▶ HEAP-INCREASE-KEY() - $O(\lg n)$
- ▶ MAX-HEAP-INSERT() - $O(\lg n)$

Reference

T H Cormen, C E Leiserson, R L Rivest, C Stein *Introduction to Algorithms*, 3rd ed., PHI, 2010