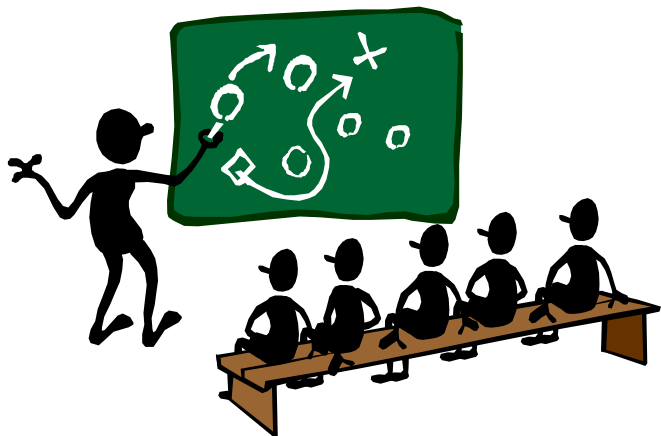


Algorithms – Chapter 6

Heap Sort



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August 2007

Rev. '08, '11, '12, '15, '16, '18, '19, '20, '21

Sorting Algorithms (1/2)

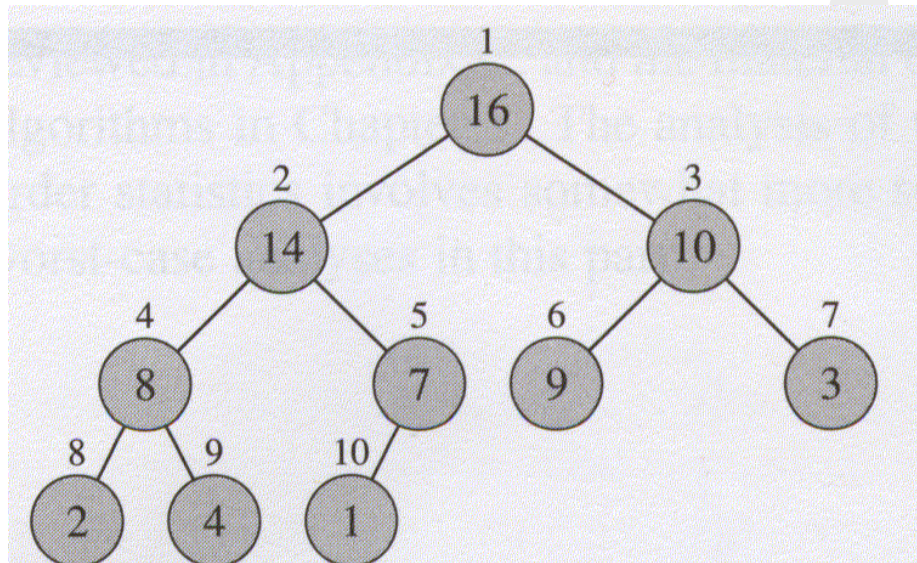
- Insertion sort
 - in place: only a constant number of elements of the input array are sorted outside the array ☺
 - worst case: $O(n^2)$ ☹
- Merge sort
 - worst case: $\Theta(n \lg n)$ ☺
 - not in place ☹
- Heap sort (Chap 6)
 - in place ☺
 - worst case: $O(n \lg n)$ ☺

Sorting Algorithms (2/2)

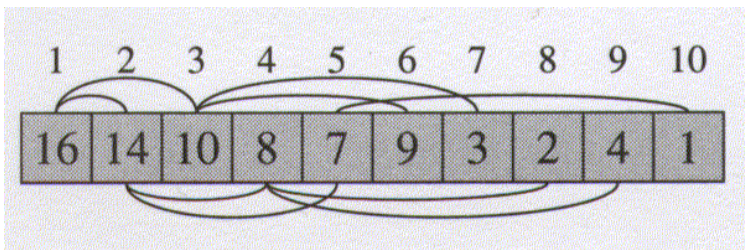
- Quick sort (Chap 7)
 - worst case: $\Theta(n^2)$ ☹️
 - average case: $\Theta(n \lg n)$ 😊
 - in-place 😊
- Worst-case time complexity for all comparison sorts: $\Omega(n \lg n)$
- Counting / Radix / Bucket sort (Chap 8)
 - non-comparison sorts
 - linear-time algorithms
- Order statistics (Chap 9)
 - find i -th smallest element in $O(n)$ time

Heaps

- The binary heap data structure is an array object that can be viewed as a **complete** binary tree



A max heap



Parent(i)
return $\lfloor i/2 \rfloor$
Left(i)
return $2i$
Right(i)
return $2i+1$

Heap Properties

- Max-heap
 - $A[\text{Parent}(i)] \geq A[i]$
- Min-heap
 - $A[\text{Parent}(i)] \leq A[i]$
- The **height** of a node in a heap (tree)
 - the number of edges on the longest simple downward path from the node to a leaf
- The height of a heap (tree)
 - the height of the **root**
- The height of a heap: $\Theta(\lg n)$

Turn an Array into a Max-Heap (1/3)

Max-Heapify (A, i)

```
0    * assume binary trees rooted at Left( $i$ ) & Right( $i$ )  
    are both max-heaps already  
1     $l \leftarrow \text{Left}(i)$   
2     $r \leftarrow \text{Right}(i)$   
3    if  $l \leq \text{heap-size}[A]$  and  $A[l] > A[i]$   
4    *  $A[i]$  is smaller than its left child  
5        then  $\text{largest} \leftarrow l$   
6        else  $\text{largest} \leftarrow i$ 
```

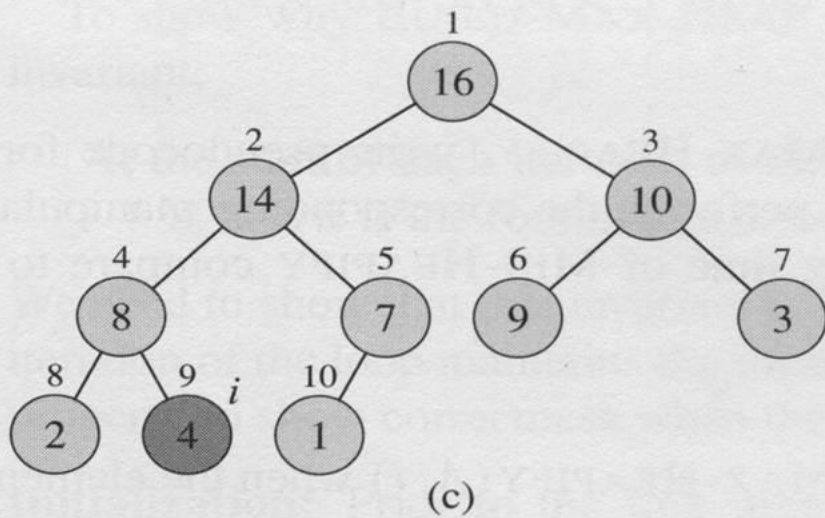
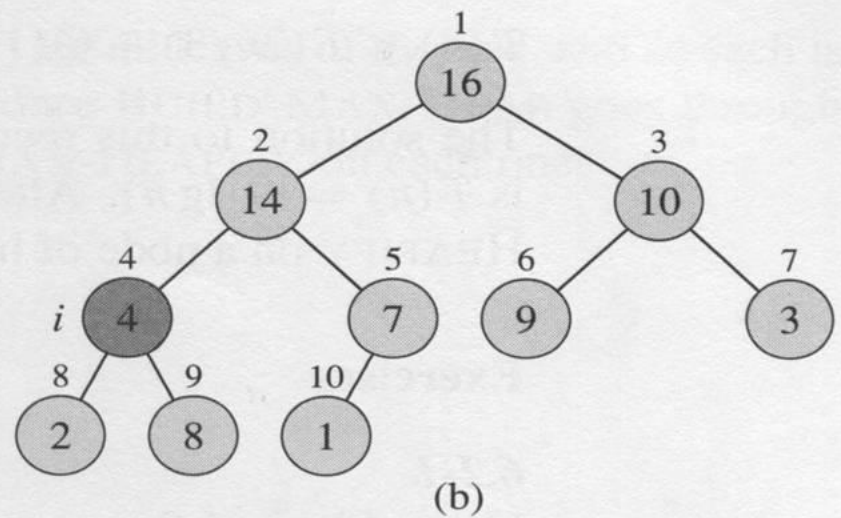
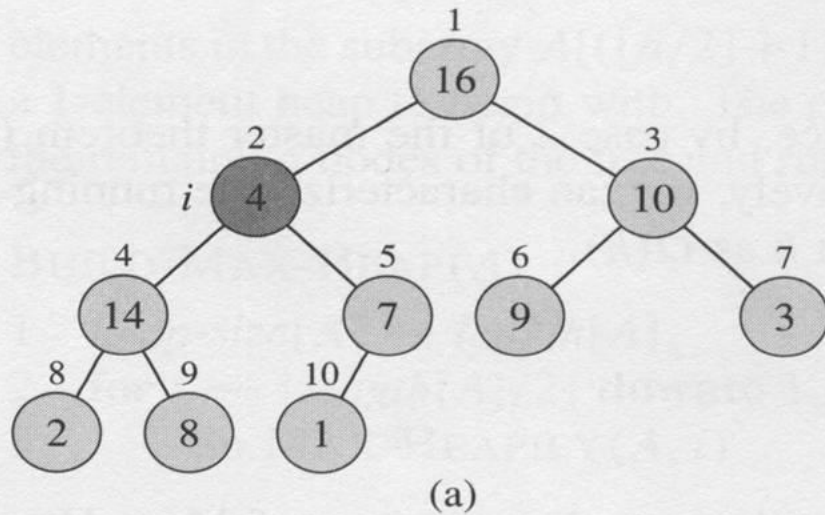
Turn an Array into a Max-Heap (2/3)

```
7   if  $r \leq \text{heap-size}[A]$  and  $A[r] > A[\text{largest}]$ 
8   * right child is the largest among  $l, r, i$ 
9   then  $\text{largest} \leftarrow r$ 
10  if  $\text{largest} \neq i$ 
11  then exchange  $A[i] \leftrightarrow A[\text{largest}]$ 
12      Max-Heapify ( $A, \text{largest}$ )
13      * recursive call
```

Runtime of Max-Heapify on a node of height h is **$O(h)$**

(Note that the maximum height = $\lfloor \lg n \rfloor$)

Example



**Max-Heapify (A, 2) with
heap-size[A] = 10**

Turn an Array into a Max-Heap (3/3)

Build-Max-Heap(A)

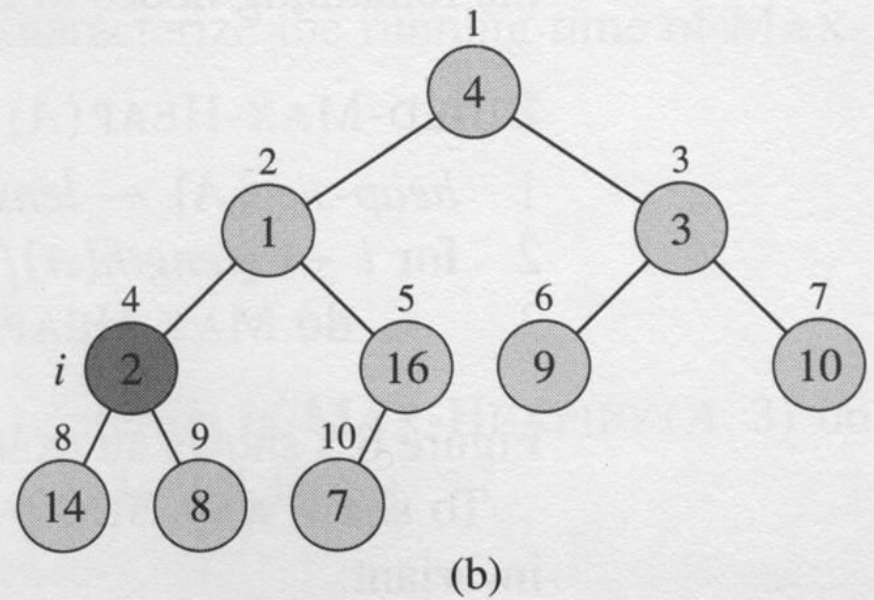
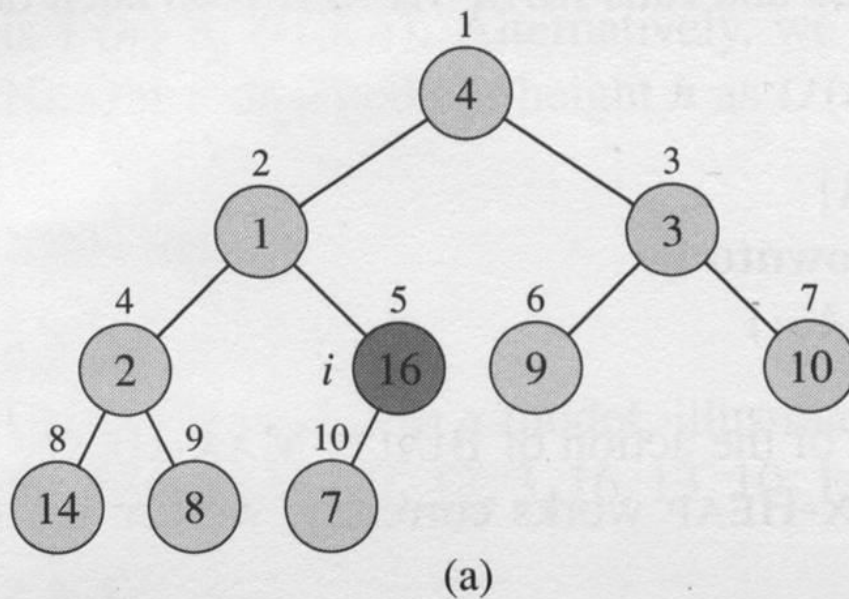
```
1  heap-size[A] ← length[A]
2  for  $i \leftarrow \lfloor \text{length}[A] / 2 \rfloor$  downto 1
3      do Max-Heapify( A,  $i$  )
4  * for  $k > \lfloor \text{length}[A] / 2 \rfloor$ , A[k] is a leaf
```

Example (1/3)

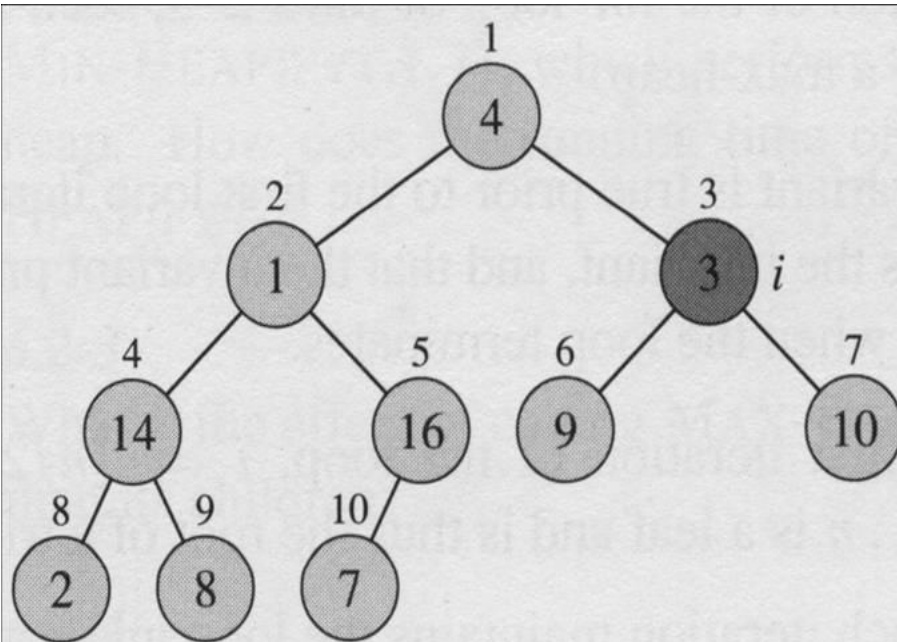
A

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

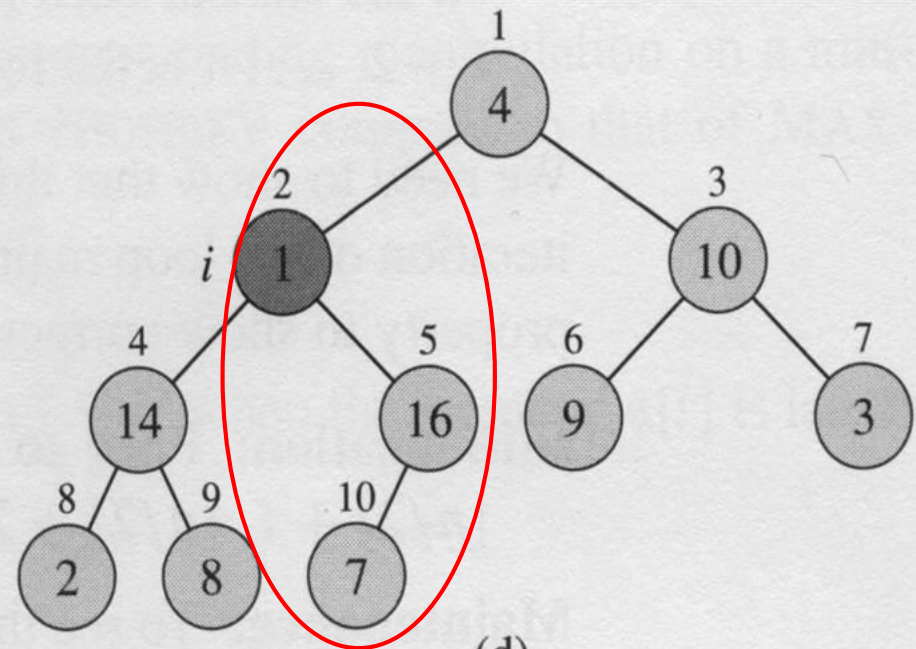
10 elements



Example (2/3)

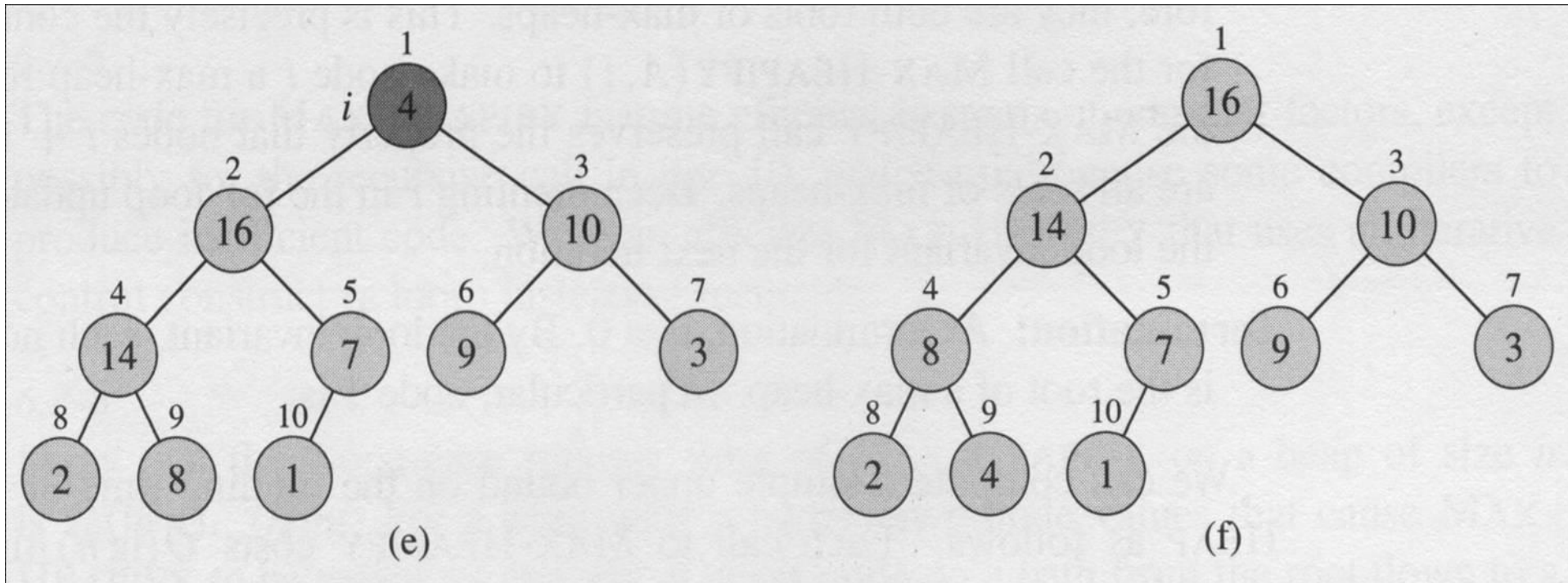


(c)



(d)

Example (3/3)



Time Complexity of Build-Max-Heap (1/2)

- An easy but **non-tight** bound
 - each call to Max-Heapify costs $O(\lg n)$
 - there are $O(n)$ such calls
 - $\rightarrow O(n \lg n)$

Adar

Time Complexity of Build-Max-Heap (2/2)

Asymptotically tight bound

- Corollaries

- an n -element heap has height $\lfloor \lg n \rfloor$
- at most $\lceil n/2^{h+1} \rceil$ nodes of any height h

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

$$\sum_{h=0}^{\infty} \frac{h}{2^h} = 2 \left(\because \sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2} \right) \quad (\text{See Appendix A})$$

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

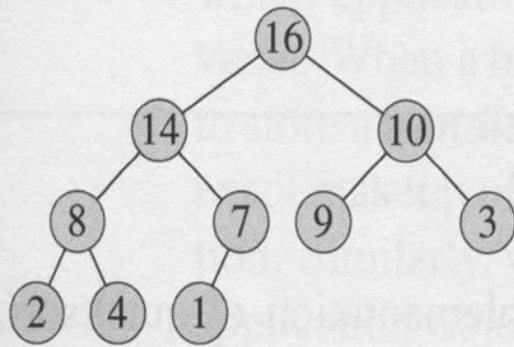
Heap Sort

Heap-Sort(A)

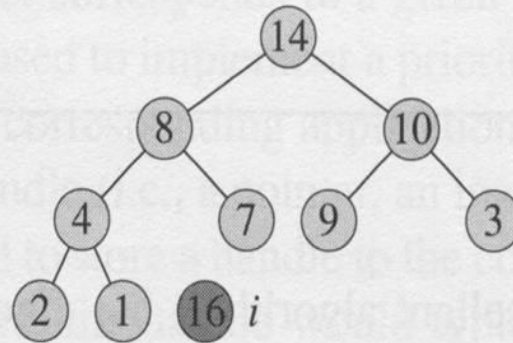
```
1   Build-Max-Heap(A)
2   for  $i \leftarrow \text{length}[A]$  down to 2
3       do exchange  $A[1] \leftrightarrow A[i]$ 
4           heap-size[A]  $\leftarrow$  heap-size[A] - 1
5       Max-Heapify( A, 1 )
```

Time complexity: $O(n \lg n)$

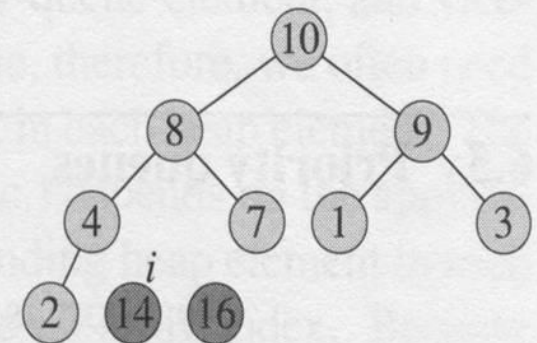
Example (1/2)



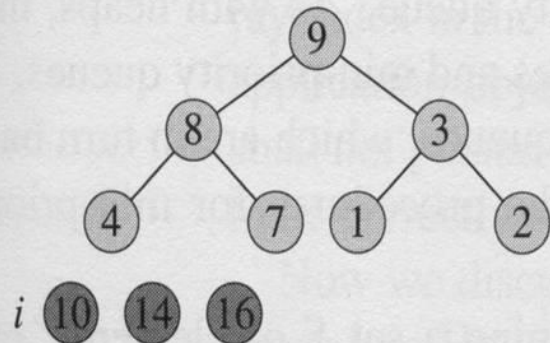
(a)



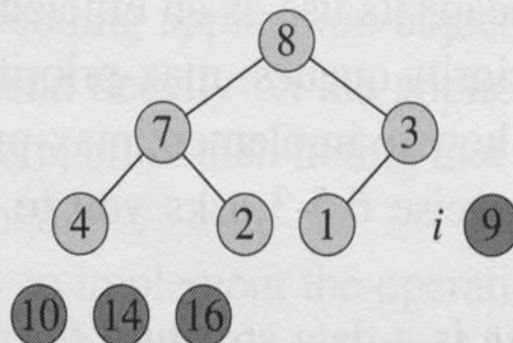
(b)



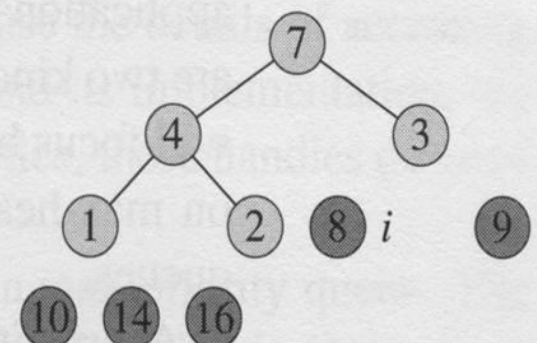
(c)



(d)

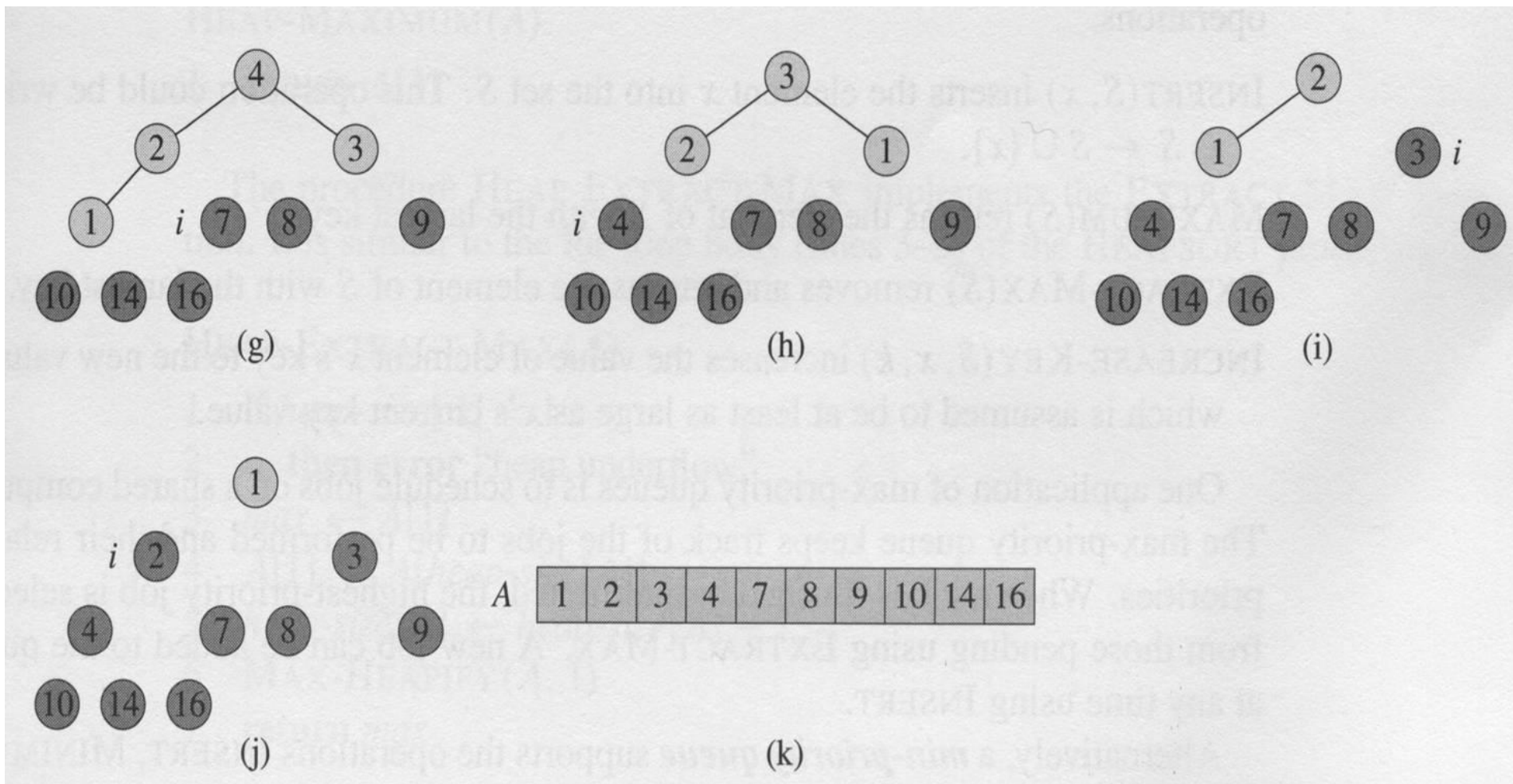


(e)



(f)

Example (2/2)



Priority Queues

- A priority queue is a data structure that maintains a set S of elements, each with an associated value called a key
- A max-priority queue supports the following operations:
 - Insert(S, x) $O(\lg n)$
 - Maximum(S) $\Theta(1)$
 - Extract-Max(S) $O(\lg n)$
 - Increase-Key(S, x, k) $O(\lg n)$
- Of course, there is a min-priority queue
 - the dual of the max-priority queue

Maximum and Extract-Max

Heap-Maximum(A)

$\Theta(1)$

1 **return** A[1]

Heap-Extract-Max(A)

$O(\lg n)$

1 **if** heap-size[A] < 1

2 **then error** “heap underflow”

3 max \leftarrow A[1]

4 A[1] \leftarrow A[heap-size[A]]

5 heap-size[A] \leftarrow heap-size[A] – 1

6 Max-Heapify(A, 1)

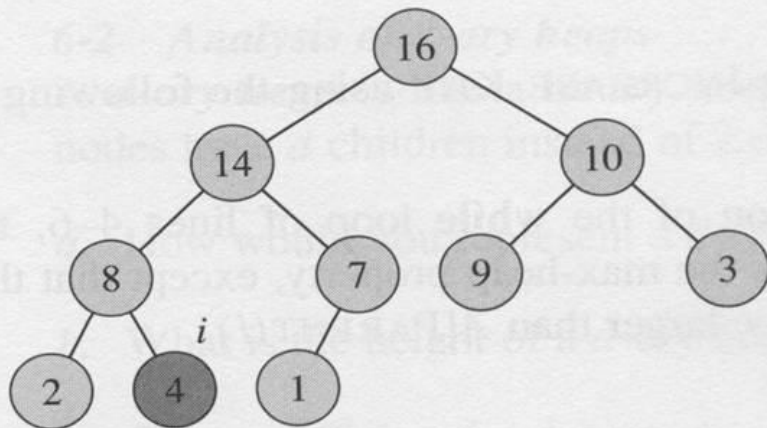
7 **return** max

Heap-Increase-Key

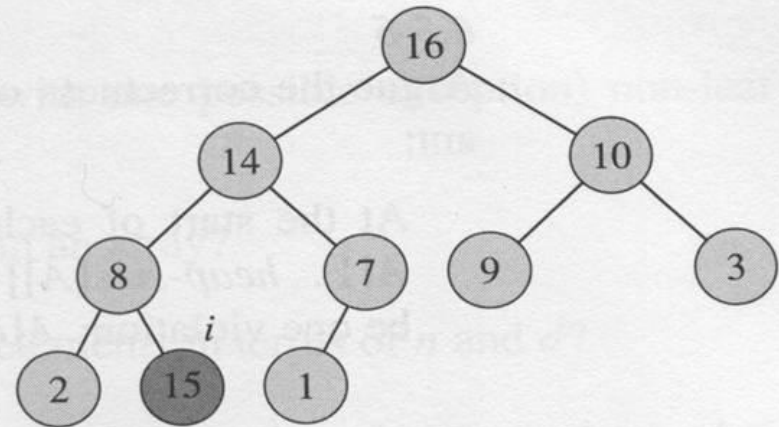
Heap-Increase-Key(A, i, key) **$O(\lg n)$**

```
1   if key < A[ i ]  
2       then error "new key < current key"  
3   A[ i ] ← key  
4   while i > 1 and A[Parent(i)] < A[ i ]  
5       do exchange A[ i ] ↔ A[Parent(i)]  
6       i ← Parent(i)
```

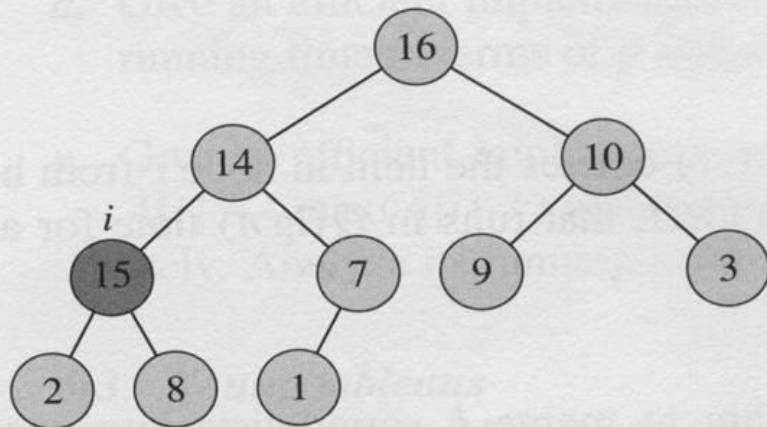

Example



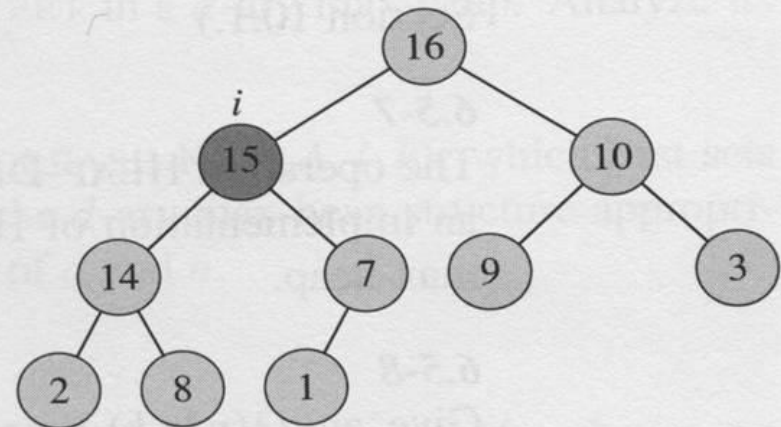
(a)



(b)



(c)



(d)

Max-Heap-Insert

Max-Heap-Insert(A, key) **$O(\lg n)$**

- 1 heap-size[A] \leftarrow heap-size[A] + 1
- 2 A[heap-size[A]] $\leftarrow -\infty$
- 3 Heap-Increase-Key(A, heap-size[A], key)