

Heap Sort

Why sorting

- ☞ Sometimes the need to sort information is inherent in an application.
- ☞ Algorithms often use sorting as a key subroutine.
- ☞ There is a wide variety of sorting algorithms, and they use rich set of techniques.
- ☞ Sorting problem has a nontrivial lower bound.
- ☞ Many engineering issues come to the fore when implementing sorting algorithms.

Heapsort

☞ Heapsort combines the better attributes of sorting algorithms:

- Like merge sort, it's running time is $O(n \lg n)$.
- Like insertion sort, it sorts in place.

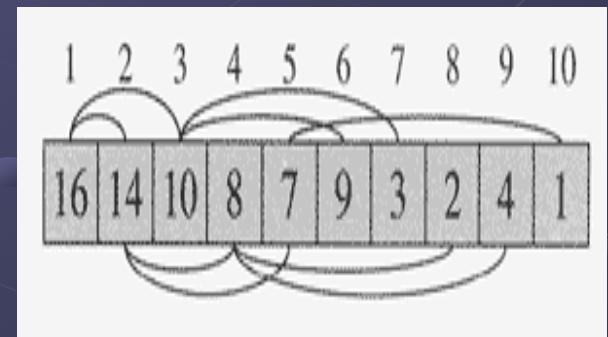
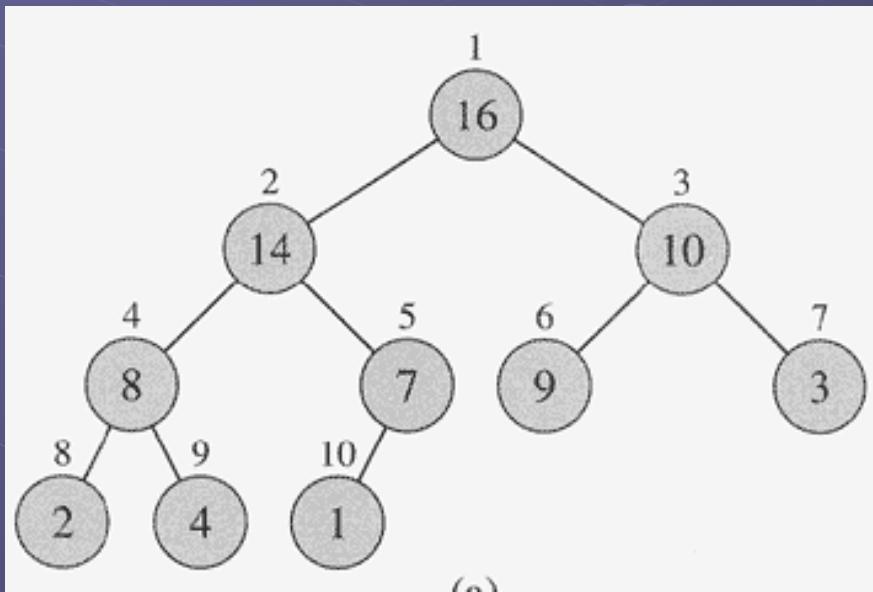
☞ Heapsort introduce another algorithm design technique:

- The use of a data structure to manage information during the execution of the algorithm.

Heaps¹

Heap:

- An array object that can be viewed as a nearly complete binary tree.



Heaps²

☞ Max-heap (Commonly used in heapsort)

- The *max-heap property* is that for every node i other than the root,

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

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

☞ Min-heap

- The *min-heap property* is that for every node i other than the root,

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

- The smallest element in a min-heap is stored at the root.
- Commonly used in priority queues.

Heaps³

☞ Basic procedures:

- MAX-HEAPIFY: Maintaining the max-heap property, runs in $O(\lg n)$ time.
- BUILD-MAX-HEAP: Produce a max-heap from an unordered input array, runs in linear time.
- HEAPSORT: Sorts an array in place, runs in $O(n \lg n)$ time.
- MAX-HEAP-INSERT, HEAP-EXTRACT-MAX, HEAP-INCREASE_KEY, HEAP-MAXIMUM: Allow the heap to be used as a priority queue, run in $O(\lg n)$ time.

Maintaining the Heap Property¹

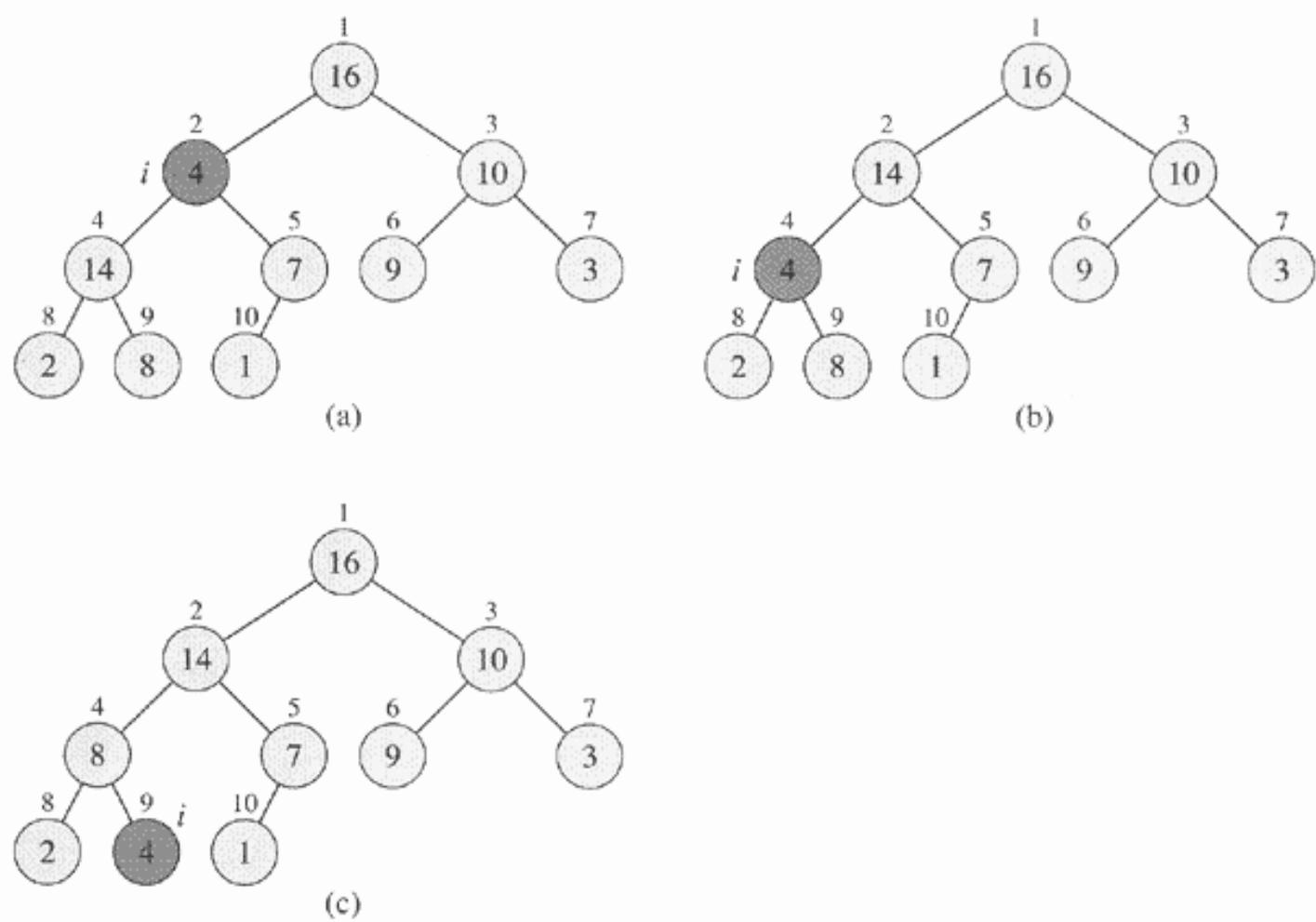
☞ MAX-HEAPIFY

- Let the value at $A[i]$ “float down” in the max-heap so that the subtree rooted at index i become a max-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( $A, \text{largest}$ )
```

Maintaining the Heap Property²



Maintaining the Heap Property³

☞ The running time of MAX-HEAPIFY

- Worst case: The children's subtree each have size at most $2n/3$ when the last row of the tree is exactly half full.

$$T(n) \leq T(2n/3) + \Theta(1) \quad (\text{Case 2 of Master theorem})$$

$$\therefore T(n) = O(\lg n)$$

Building a Heap¹

Algorithm

```
BUILD-MAX-HEAP( $A$ )
```

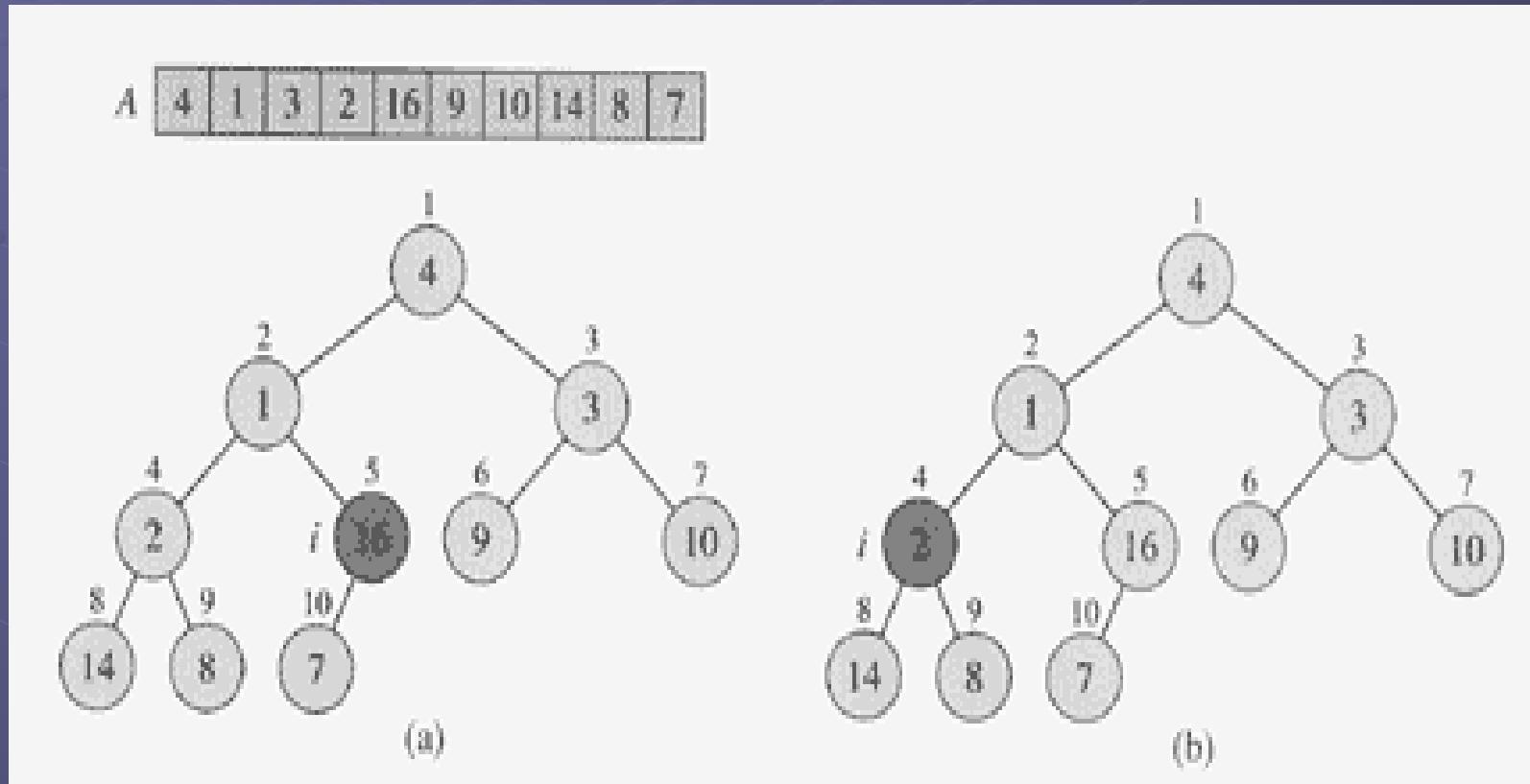
```
1  $A.\text{heap-size} = A.\text{length}$ 
2 for  $i = \lfloor A.\text{length}/2 \rfloor$  downto 1
3   MAX-HEAPIFY( $A, i$ )
```

Correctness

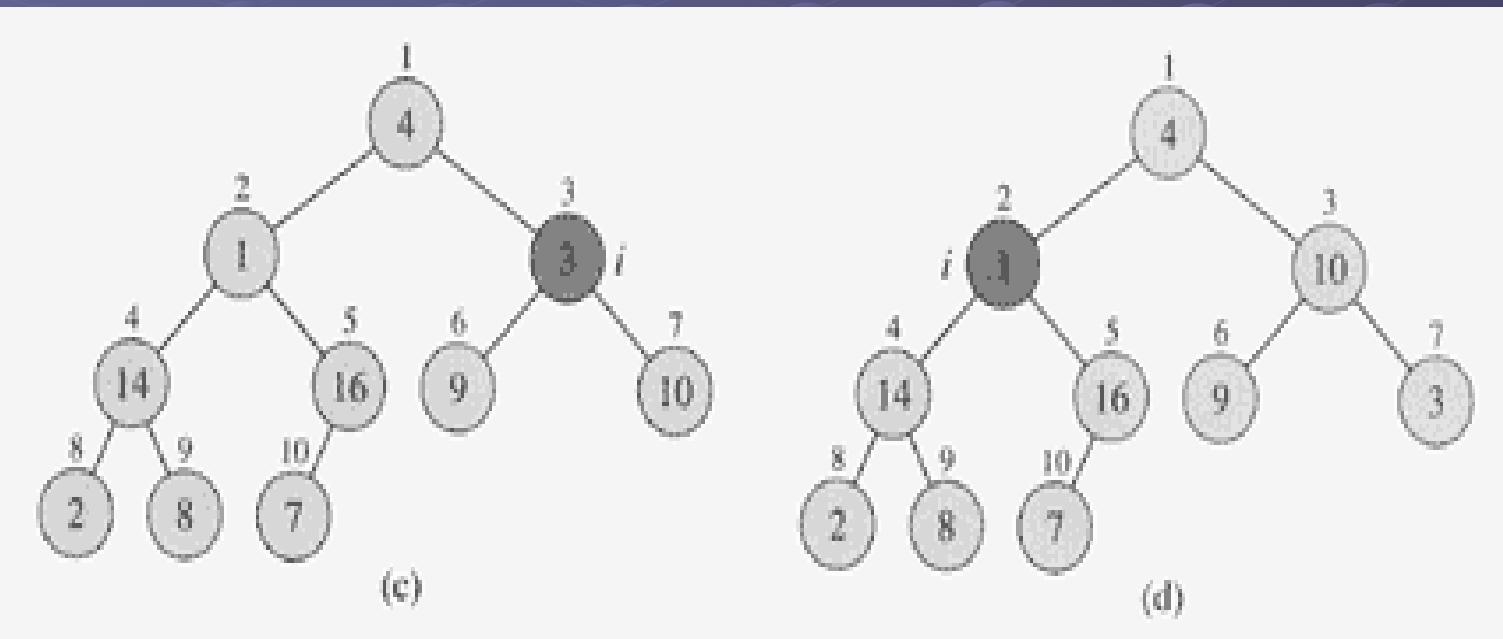
■ Loop invariant

At the start of each iteration of the for loop, each node $i + 1, i + 2, \dots, n$ is the root of a max-heap.

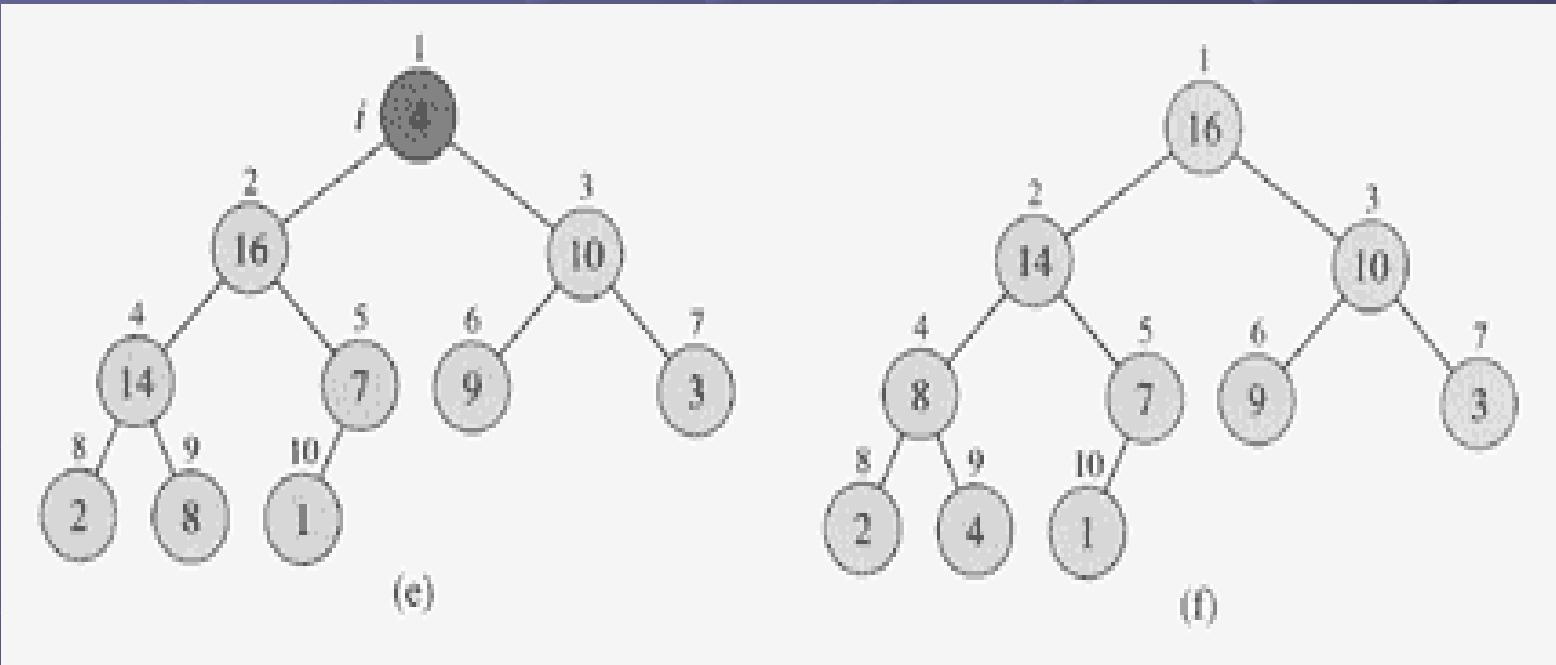
Building a Heap²



Building a Heap³



Building a Heap⁴



Building a Heap⁵

■ Initialization: Prior to the first iteration of the loop, $i = \lfloor n/2 \rfloor$.

■ Maintenance

- The children of node i are both roots of max-heaps.
- MAX-HEAPIFY(A, i) makes node i a max-heap root.

■ Termination: At termination $i = 0$.

☞ Running time: $O(n \lg n)$

- Each call to MAX-HEAPIFY cost $O(\lg n)$ time.
- There are $O(n)$ such call.
- Not asymptotic tight.

*Building a Heap*⁶

☞ Tighter analysis

- An n -element heap has height $\lfloor \lg n \rfloor$ and at most $\lceil n/2^{h+1} \rceil$ nodes of any height h .
- The time required by MAX-HEAPIFY when called on a node of height h is $O(h)$.

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

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

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

Deriving the equation in p.159

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$$

$$\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2} \quad \text{for } |x| < 1$$

Substituting $x = \frac{1}{2}$ into the above equation.

The Heapsort Algorithm^I

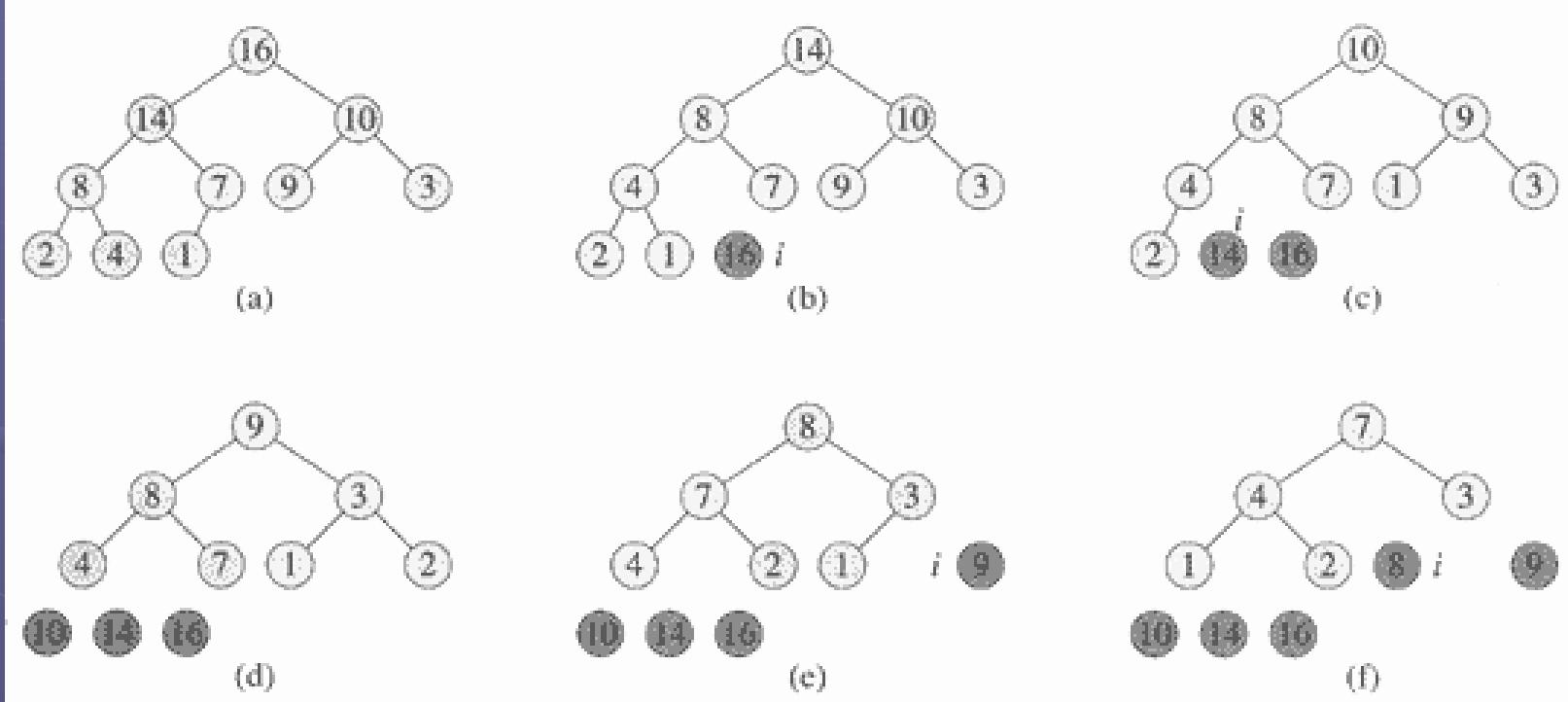
☞ Algorithm

```
HEAPSORT( $A$ )
```

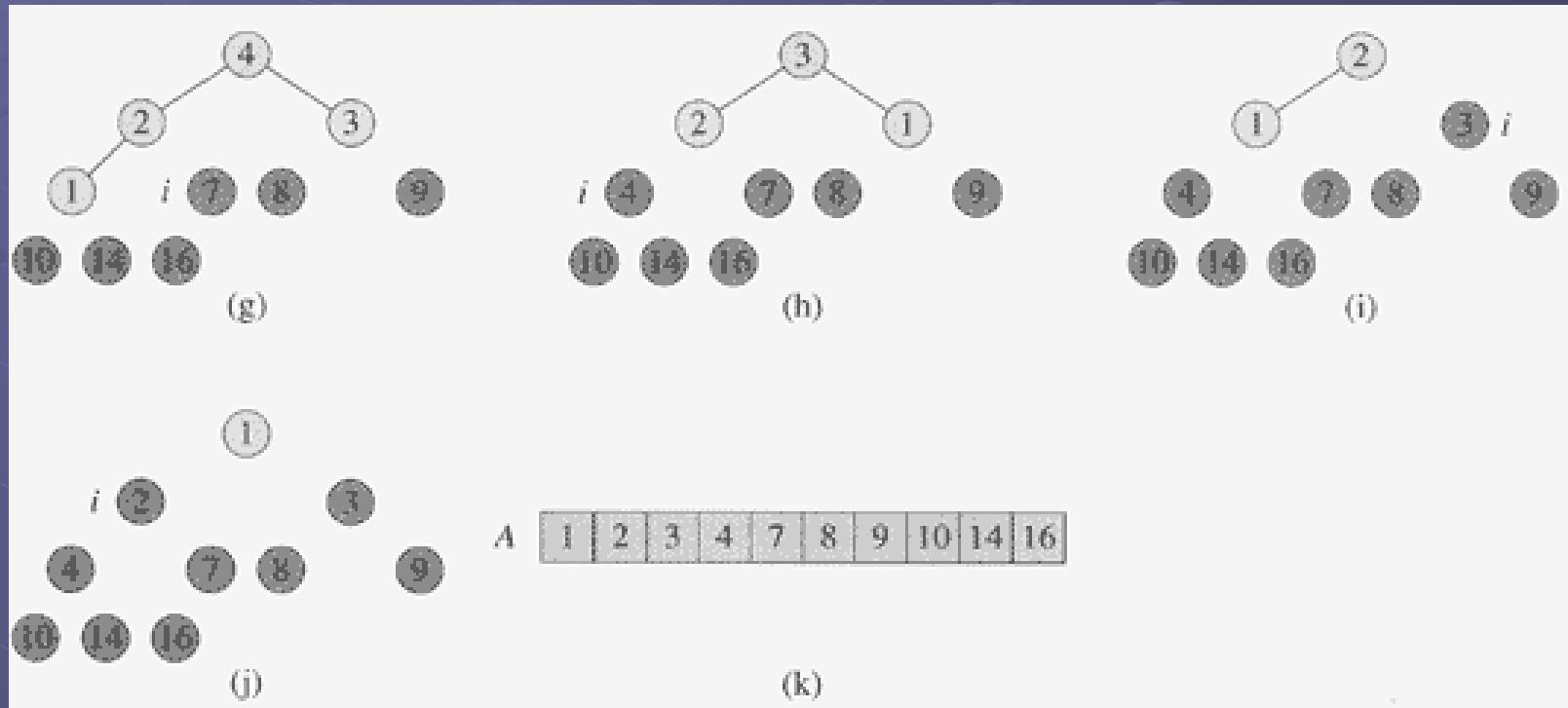
- 1 BUILD-MAX-HEAP(A)
- 2 **for** $i = A.length$ **downto** 2
- 3 exchange $A[1]$ with $A[i]$
- 4 $A.heap-size = A.heap-size - 1$
- 5 MAX-HEAPIFY($A, 1$)

☞ Running time: $O(n \lg n)$.

The Heapsort Algorithm²



The Heapsort Algorithm³



Priority Queues¹

☞ Priority queue:

- A data structure for maintaining a set S of elements, each with an associated value called a *key*.

☞ Max-priority queue

- $\text{INSERT}(S, x)$
- $\text{MAXIMUM}(S)$
- $\text{EXTRACT-MAX}(S)$
- $\text{INCREASE-KEY}(S, x, k)$

☞ Min-priority queue

Priority Queues²

☞ **HEAP-MAXIMUM(A)** $\Theta(1)$

1 **return** $A[1]$

☞ **HEAP-EXTRACT-MAX(A)** $O(\lg n)$

1 **if** $heap\text{-size} < 1$
2 **then error** “heap underflow”
3 $max = A[1]$
4 $A[1] = A[A.heap\text{-size}]$
5 $A.heap\text{-size} = A.heap\text{-size} - 1$
6 MAX-HEAPIFY ($A, 1$)
7 **return** max

Priority Queues³

☞ **HEAP-INCREASE-KEY (A, i, key)** $O(\lg n)$

- 1 **if** $key < A[i]$
- 2 **then 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 $i = \text{PARENT}(i)$

☞ **MAX-HEAP-INSERT(A, key)** $O(\lg n)$

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

Priority Queues⁴

