

# Introduction to Algorithms

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## ❖ Chapter.2 Sorting and Order Statistics

### 1. Heapsort

- 1) Heaps
- 2) Maintaining the heap property
- 3) Building a heap
- 4) The heapsort algorithm
- 5) Priority queues

### 2. Quicksort

- 1) Description of quicksort
- 2) Performance of quicksort
- 3) A randomized version of quicksort

### 3. Sorting in Linear Time

- 1) Lower bounds for sorting
- 2) Counting sort
- 3) Radix sort
- 4) Bucket sort

### 4. Medians and Order Statistics

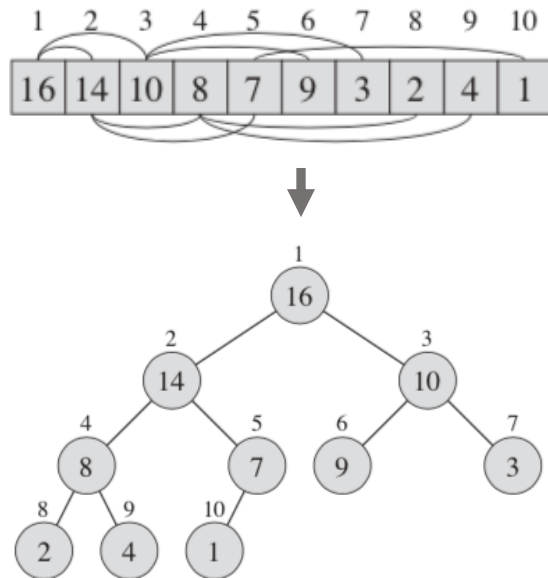
- 1) Minimum and maximum
- 2) Selection in expected linear time
- 3) Selection in worst-case linear time

## ❖ Heap

- Introduction

- Running time of heapsort:  $O(n \lg n)$
- In-place algorithm : only a constant number of array elements are stored outside the input array
- New algorithm design technique : using a data structure (heap)

- Definition



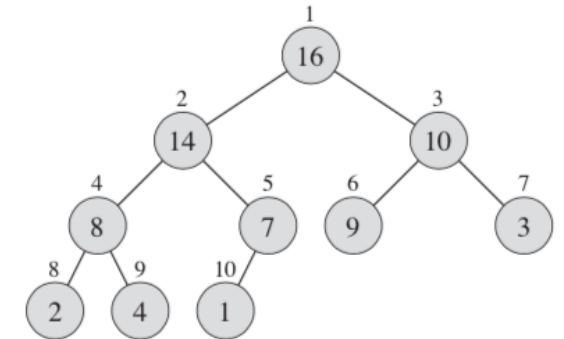
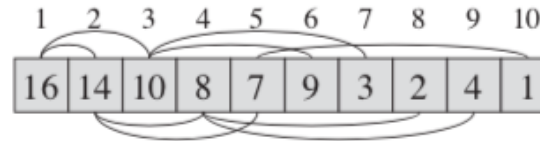
- Array object that we can view as a nearly complete binary tree
- Filled from the left up to a point
- Attributes
  - ✓ *A.length* : gives the number of elements in the array (10)
  - ✓ *Heap – size* : represents how many elements in the heap (10)
  - ✓  $0 \leq A.\text{heap} - \text{size} \leq A.\text{length}$

## ❖ Heap

- Definition

- Index function

- ✓  $i$  : index of the nodes
    - ✓  $A[1]$  : root of the tree
    - ✓  $PARENT(i) = \lfloor i/2 \rfloor$
    - ✓  $LEFT(i) = 2i$
    - ✓  $RIGHT(i) = 2i + 1$



- Kinds of binary heaps

- *Max – heap* :  $A[PARENT(i)] \geq A[i]$ 
    - ✓ Largest value is stored in root node
    - ✓ Heapsort uses max-heap
  - *Min – heap* :  $A[PARENT(i)] \leq A[i]$ 
    - ✓ Smallest value is stored in root node

## ❖ Heap

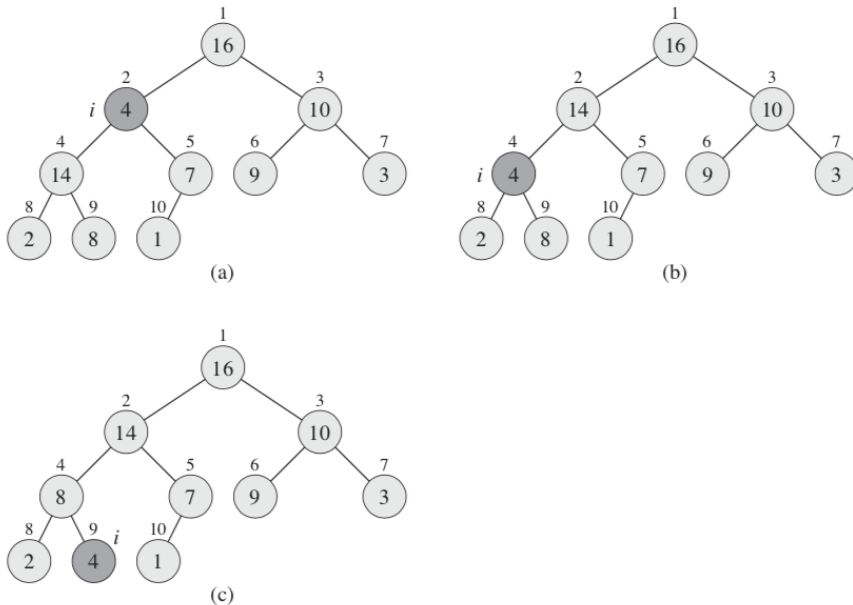
- Operation of heap

- Height of node : the number of edges on the longest simple downward path from node to a leaf
- Height of heap : height of root node
- Height of  $n$  elements-heap :  $\lceil \lg n \rceil$  (complete binary tree)
- The basic operations on heaps run in time proportional to the height of the tree :  $O(\lg n)$ 
  - ✓ *MAX – HEAPIFY* : maintain the max-heap property –  $O(\lg n)$
  - ✓ *BUILD – MAX – HEAP* : produce a max-heap from unordered input array – linear time
  - ✓ *HEAPSORT* : sorts an unordered array in place –  $O(n \lg n)$
  - ✓ *MAX – HEAP – INSERT, HEAP – EXTRACT – MAX, HEAP – INCREASE – KEY, HEAP – MAXIMUM* : to implement a priority queue –  $O(\lg n)$

## ❖ Maintaining the Heap-property

### • Definition

- To maintain the max-heap property, we call *MAX – HEAPIFY*
- Assume that the binary trees rooted at  $LEFT(i)$  and  $RIGHT(i)$  are max-heaps, but  $A[i]$  might be smaller than its children



The process of *MAX – HEAPIFY*

- (a) :  $A[2]$  violates the max-heap property
- (b-1) : find the larger children between  $A[2]$  and  $A[4]$
- (b-2) : exchange  $A[2]$  with  $A[4]$
- (b-3) :  $A[4]$  violates the max-heap property
- (c-1) : find the larger children between  $A[8]$  and  $A[9]$
- (c-2) : exchange  $A[8]$  with  $A[9]$
- (c-3) :  $A[9]$  is fixed up

## ❖ Maintaining the Heap-property

- Pseudocode

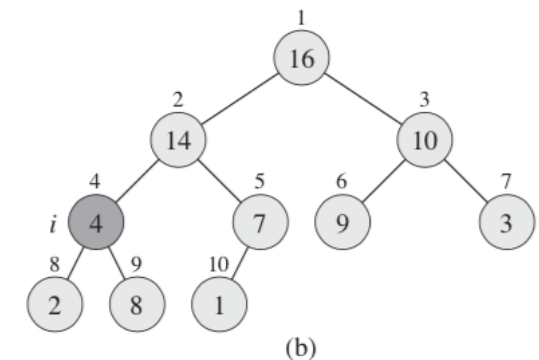
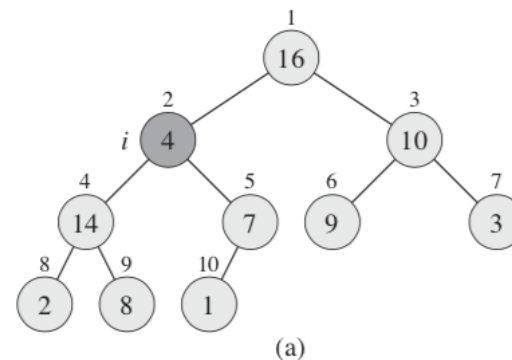
➤ Input : array  $A$ , index  $i$

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

The pseudocode for *MAX-HEAPIFY*

- 1~7 : find the largest values in  $A[i], A[\text{LEFT}(i)], A[\text{RIGHT}(i)]$
- 8~10 : check heap property is violated
  - ✓ If  $A[i]$  is the largest → terminates
  - ✓ Otherwise  $A[i]$  violates the heap property → swap and recursively call *MAX-HEAPIFY*



## ❖ Maintaining the Heap-property

- The running time of **MAX-HEAPIFY**

- Subtree of size  $n$

- Exchange  $A[i]$  with  $A[LEFT(i)]$  or  $A[RIGHT(i)] : \theta(1)$

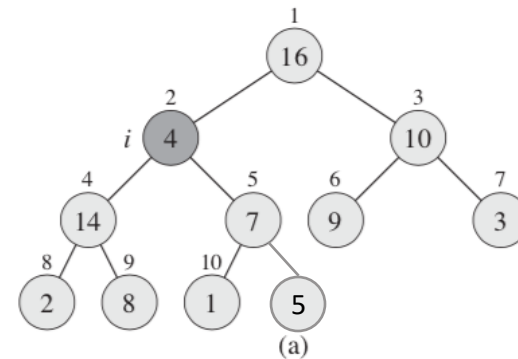
- Run **MAX-HEAPIFY** on a subtree :  $T(2n/3)$

- ✓ Children's subtrees each have size at most  $2n/3$  – the worst case occurs

- ✓ Worst case : exactly half full

- Running time of MAX-HEAPIFY

$$T(n) \leq T\left(\frac{2n}{3}\right) + \theta(1)$$





## ❖ Maintaining the Heap-property

- The running time of **MAX-HEAPIFY**

- Running time of MAX-HEAPIFY

$$T(n) \leq T\left(\frac{2n}{3}\right) + \theta(1)$$

- Solve this recurrence by using case 2 of the master theorem

- ✓ In the form of equation :  $T(n) = aT(n/b) + f(n)$

1. If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ .

2. If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \lg n)$ .

3. If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(n/b) \leq cf(n)$  for some constant  $c < 1$  and all sufficiently large  $n$ , then  $T(n) = \Theta(f(n))$ . ■

- ✓ In this case,  $a = 1$   $b = 3/2$   $f(n) = \theta(1) = \theta(n^{\log_{3/2} 1}) = \theta(1)$

- ✓  $T(n) = \theta(n^{\log_{3/2} 1} \lg n) = \theta(\lg n)$

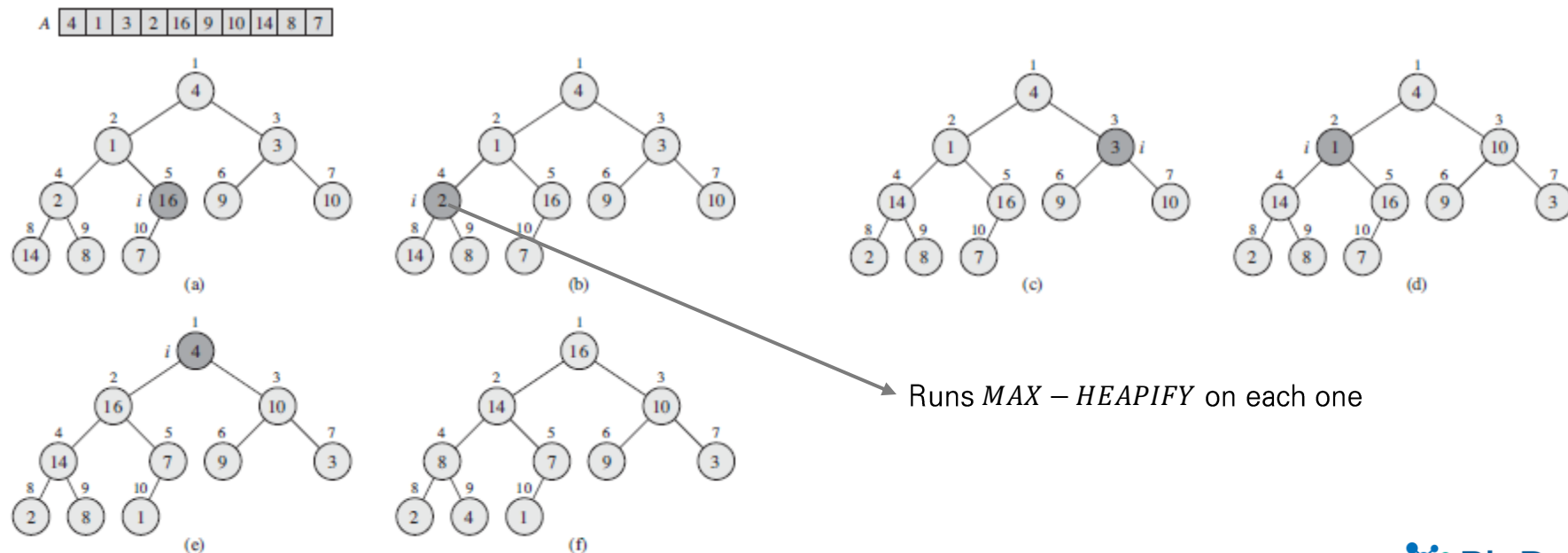
- Height of  $n$  elements-heap( $h$ ) :  $\lg n$

- ✓  $\theta(\lg n) = \theta(h)$

## ❖ Building a heap

- Definition

- Using *MAX\_HEAPIFY* in a bottom-up manner to convert an array  $A[1..n]$  into a max-heap
- The elements in the subarray  $A[(\lfloor n/2 \rfloor + 1)..n]$  are all leaves of the tree
  - ✓ In this case  $N = 10$ ,  $A[6, \dots, 10]$  are all leaves – already satisfy max-heap
- The procedure *BUILD-MAX-HEAP* goes through the remaining nodes of the tree



## ❖ Building a heap

- Pseudocode

- Input : array  $A$

$BUILD-MAX-HEAP(A)$

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

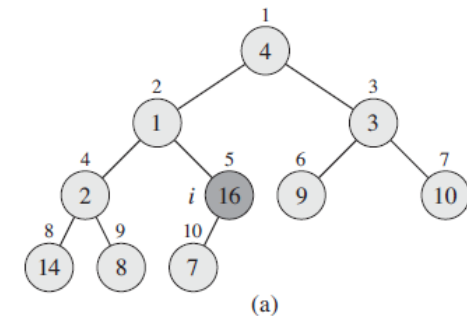
- 2~3 : compute  $MAX-HEAPIFY$  in a bottom-up manner, starting from remaining nodes (not leaves node)

- Loop invariant of  **$BUILD-MAX-HEAP$**

- At the start of each iteration of the for loop of lines 2-3, each node  $i + 1, i + 2, \dots, n$  is the root of a max-heap

- Initialization

- ✓ Prior to the first iteration of the loop,  $i = \lfloor n/2 \rfloor$ , each node  $\lfloor n/2 \rfloor + 1, \lfloor n/2 \rfloor + 2, \dots, n$  is a leaf and is thus the root of a trivial max-heap



Initial state

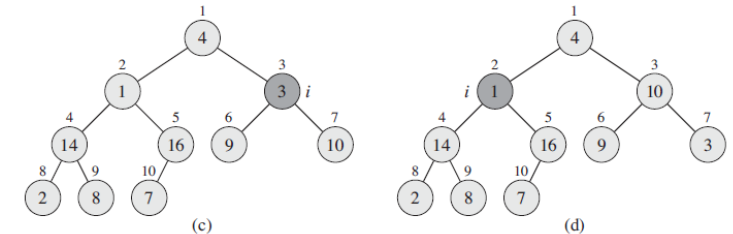
## ❖ Building a heap

- Loop invariant of ***BUILD – MAX – HEAP***

- At the start of each iteration of the for loop of lines 2–3, each node  $i + 1, i + 2, \dots, n$  is the root of a max-heap

- Maintenance

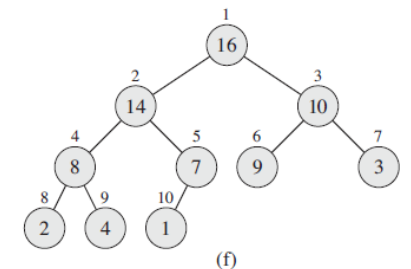
- ✓ Children of node  $i$  are numbered higher than  $i$
- ✓ By the loop invariants, They are both roots of max-heaps
- ✓ ***MAX – HEAPIFY***( $A, i$ ) preserves the property at node  $i$



Maintenance state

- Termination

- ✓ At termination  $i = 0$
- ✓ By the loop invariant, each node  $1, 2, \dots, n$  is the root of a max-heap



Termination state

## ❖ Building a heap

### • Upper bound running time of **BUILD – MAX – HEAP**

- As we computed, *MAX – HEAPIFY* cost  $O(\lg n)$
- *BUILD – MAX – HEAP* makes  $O(n)$ 
  - ✓ Line 2~3 iterate  $n/2$  terms
- The upper bound running time :  $O(n \lg n)$

**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)
    
```

### • Tighter bound running time of **BUILD – MAX – HEAP**

- $n$  elements heap has height  $\lfloor \lg n \rfloor$
- At height  $h$ , there are  $\lfloor n/2^{h+1} \rfloor$  nodes

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

Iterates whole levels

The number of node at level  $h$

Running time of *MAX – HEAPIFY*

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

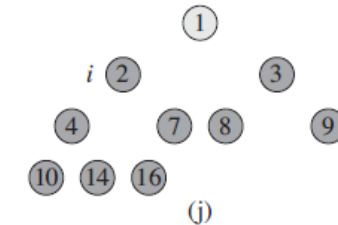
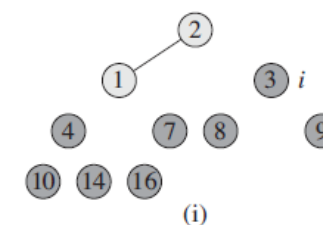
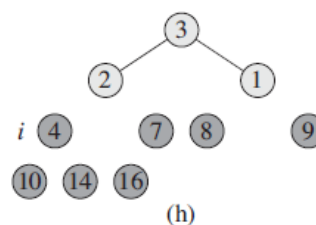
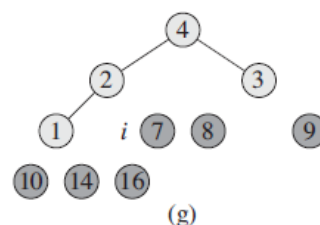
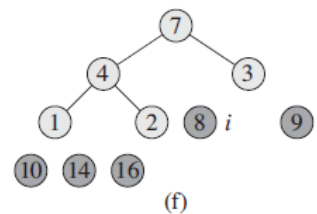
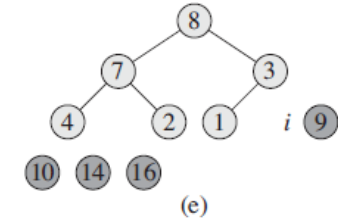
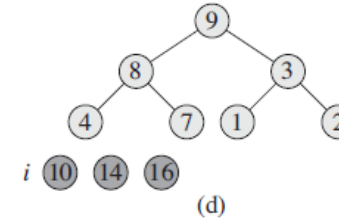
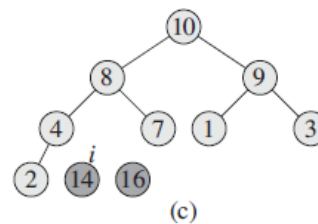
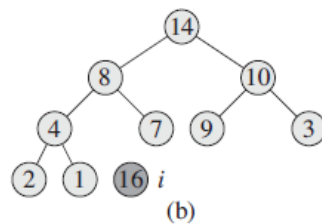
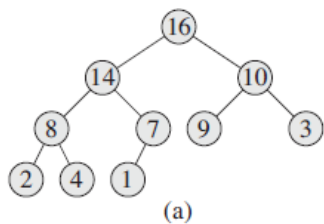
$$O\left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h}\right) = O\left(n \sum_{h=0}^{\infty} \frac{h}{2^h}\right) = O(n).$$

## ❖ The heapsort algorithm

### • Definition

- Start by using *BUILD – MAX – HEAP* to build a max-heap on the unsorted input array  $A[1..n]$
- The maximum element of the array is stored at the root  $A[1]$
- Put the maximum element into final position by exchanging with  $A[n]$
- Discard node  $n \rightarrow (\text{heap\_size} - 1) \rightarrow$  check new root element satisfy max-heap *property* (*MAX – HEAPIFY*)

A [ 1 2 3 4 7 8 9 10 14 16 ]



## ❖ The heapsort algorithm

- Pseudo code

- Input : unordered array  $A$

    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( $A, 1$ )

- 1 : build a max-heap using *BUILD – MAX – HEAP*

- 3~4 : exchange root( $A[1]$ ) with  $A[n] \rightarrow$  resize the heap

- 5 : rebuild the max-heap

- Running time of **HEAPSORT**

- *BUILD – MAX – HEAP* :  $O(n)$

- For loop iterate  $n-1$  terms

- ✓ Call *MAX – HEAPIFY* :  $O(\lg n)$

- Total running time :  $(n - 1) O(\lg n) + O(n) = O(n \lg n)$

## ❖ Priority queues

- Introduction

- Heap data structure itself has many uses
- Focus on the most popular applications of a heap : **priority queue**

- Definition of a priority queue

- Data structure for maintaining a set  $S$  of elements, each with an associated value called a **key**
- A max-priority queue operations
  - ✓ ***INSERT***( $S, x$ ) : insert the element  $x$  into the set  $S$
  - ✓ ***MAXIMUM***( $S$ ) : returns the element of  $S$  with the largest key
  - ✓ ***EXTRACT – MAX***( $S$ ) : removes and return the element of  $S$  with the largest key
  - ✓ ***INCREASE – KEY***( $S, x, k$ ) : increases the value of element  $x$ 's key to the new value  $k$
- Now, implement these operations using heap data structure



## ❖ Priority queues

- ***MAXIMUM(S)***

- The procedure *HEAP – MAXIMUM* implements the *MAXIMUM* operation
- Takes  $\theta(1)$  time

*HEAP-MAXIMUM(A)*

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

- ***EXTRACT – MAX(S)***

- The procedure *HEAP – EXTRACT – MAX* implements the *EXTRACT – MAX* operation
- Similar to *HEAPSORT* loop procedure
- Takes  $O(\lg n)$  time for *MAX – HEAPIFY*

*HEAP-EXTRACT-MAX(A)*

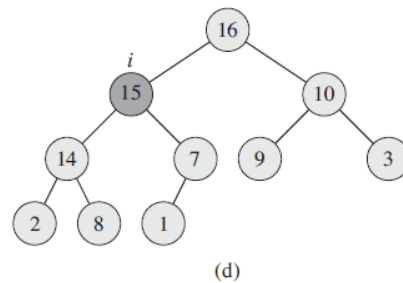
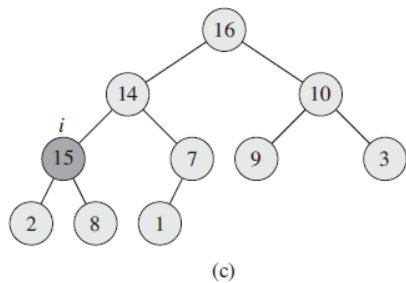
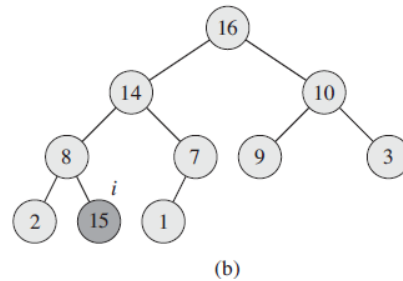
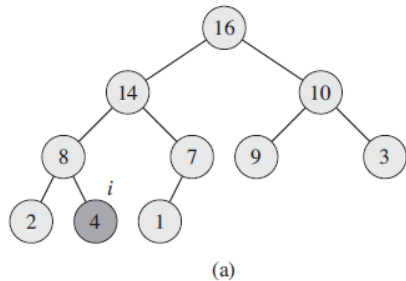
```
1 if  $A.heap-size < 1$ 
2   error "heap underflow"
3  $max = A[1]$ 
4  $A[1] = A[A.heap-size]$ 
5  $A.heap-size = A.heap-size - 1$ 
6 MAX-HEAPIFY( $A, 1$ )
7 return  $max$ 
```

- ✓ 3~4 : save the largest element ( $A[1]$ ) & swapping with  $A[n]$
- ✓ 5~6 : rebuild the max-heap
- ✓ 7 : Return the largest value

## ❖ Priority queues

- **INCREASE – KEY( $S, x, k$ )**

- The procedure *HEAP – INCREASE – KEY* implements the *INCREASE – KEY* operation
- Index  $i$  node want to increase key to a new value → violates max-heap property



- ✓ (a) : update the  $i$  node's key to a new value
- ✓ (b) : new key violates the max-heap property
- ✓ (c) : exchange with parents
- ✓ (d) : iterate (c) process toward the root until finding a proper place for newly increased key

## ❖ Priority queues

- **INCREASE – KEY( $S, x, k$ )**

HEAP-INCREASE-KEY( $A, i, key$ )

```
1  if  $key < A[i]$ 
2      error "new key is smaller than current key"
3   $A[i] = key$ 
4  while  $i > 1$  and  $A[PARENT(i)] < A[i]$ 
5      exchange  $A[i]$  with  $A[PARENT(i)]$ 
6       $i = PARENT(i)$ 
```

- ✓ 3 : change key value of node  $i$
- ✓ 4~6 : exchange with its parent until find proper place

➤ The running time of *HEAP – INCREASE – KEY* :  $O(\lg n)$

- ✓ Find proper place of newly increased key → Traverse from node  $i$  to root
- ✓ The height of the tree :  $\lg n$

## ❖ Priority queues

- ***INSERT – KEY***(*S*, *x*, *k*)

- The procedure *MAX – HEAP – INSERT* implements the *INSERT* operation

*MAX-HEAP-INSERT*(*A*, *key*)

1 *A.heap-size* = *A.heap-size* + 1

2 *A*[*A.heap-size*] =  $-\infty$

3 *HEAP-INCREASE-KEY*(*A*, *A.heap-size*, *key*)

- ✓ 1 : expand the heap-size

- ✓ 2 : add to the tree a new leaf whose key is  $-\infty$

- ✓ 3 : find proper place of new node using *HEAP – INCREASE – KEY*

- The running time of *MAX – HEAP – INSERT* :  $O(\lg n)$

- ✓ *HEAP – INCREASE – KEY*

- **Conclusion**

- A heap can support any priority-queue's operation in  $O(\lg n)$  time

## ❖ Chapter.2 Sorting and Order Statistics

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### 2. Quicksort

- 1) Description of quicksort
- 2) Performance of quicksort
- 3) A randomized version of quicksort

### 3. Sorting in Linear Time

- 1) Lower bounds for sorting
- 2) Counting sort
- 3) Radix sort
- 4) Bucket sort

### 4. Medians and Order Statistics

- 1) Minimum and maximum
- 2) Selection in expected linear time
- 3) Selection in worst-case linear time

## ❖ Description of Quicksort

- Introduction

- Worst-case running time :  $\theta(n^2)$
- Average-case running time :  $\theta(n \lg n)$
- In-place sorting algorithm

- Divide-and-Conquer

- Quicksort applies the divide-and-conquer paradigm

- Divide

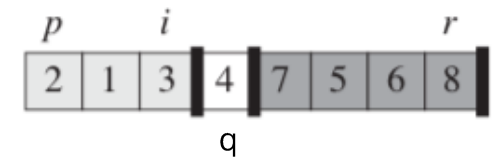
- ✓ Partition the array  $A[p..r]$  into two subarray  $A[p..q-1]$  and  $A[q+1..r]$
- ✓ Each element of  $A[p..q-1]$  is less than or equal to  $A[q]$
- ✓ Compute the index  $q$  as part of partitioning procedure

- Conquer

- ✓ Sort the two subarrays  $A[p..q-1]$  and  $A[q+1..r]$  by recursive calls to quicksort

- Combine

- ✓ Because the Subarrays are already sorted, no work is needed to combine



## ❖ Description of Quicksort

- Procedure of quicksort

QUICKSORT( $A, p, r$ )

1 **if**  $p < r$

2      $q = \text{PARTITION}(A, p, r)$

3     QUICKSORT( $A, p, q - 1$ )

4     QUICKSORT( $A, q + 1, r$ )

✓ 2 : divide the unsorted array  $A$  using *PARTITION*

✓ 3~4 : conquer the subarray by recursive call

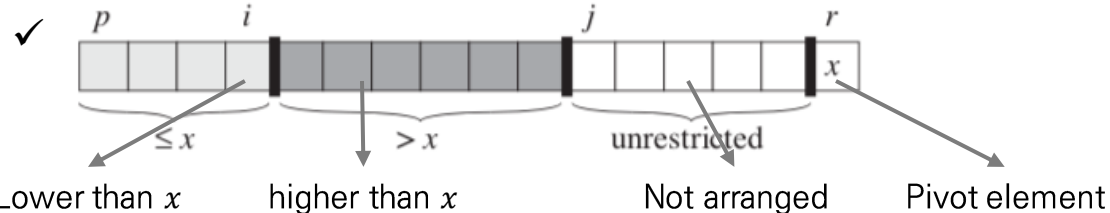
- Partitioning the array

➤ The key procedure of the quicksort algorithm

➤ Rearrange the subarray  $A[p..r]$

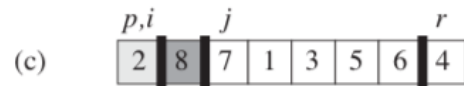
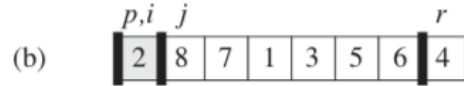
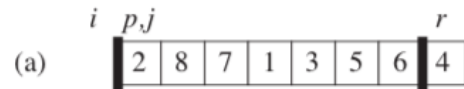
➤ Select an element  $x = A[r]$  as a pivot element which to partition the subarray

✓ Partition the array into four regions



## ❖ Description of Quicksort

- Partitioning the array



➤ (a) : initial state, pivot element is  $A[r]$

➤ (b) : the value 2 is “swapped with itself” and put in the partition of smaller values

➤ (c) (d) : increase index  $j$ , because value 8, 7 is larger than pivot, added to the partition of larger values

➤ (e) : swap  $A[i]$  and  $A[j]$ , value 1 and 8

➤ (f) : swap value 3 and 7

➤ (g) (h) : the larger partition grows to include 5 and 6 and loop terminates

➤ (i) : the pivot element is swapped



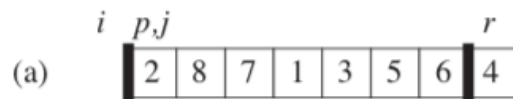
## ❖ Description of Quicksort

- Pseudocode for Partitioning

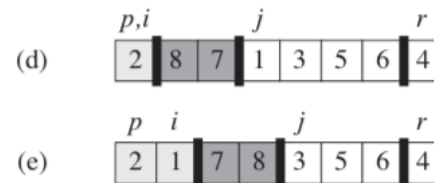
PARTITION( $A, p, r$ )

```
1   $x = A[r]$ 
2   $i = p - 1$ 
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
7  exchange  $A[i + 1]$  with  $A[r]$ 
8  return  $i + 1$ 
```

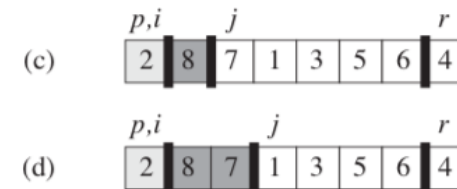
- 1 : initialize the pivot  $x$  and index  $i$
- 3~6 : compare  $A[j]$  with pivot  $x$ , increasing index  $j$ 
  - ✓ If  $A[j] \leq \text{pivot}$  : increase  $i$  and exchange  $A[i]$  with  $A[j]$
  - ✓ Otherwise : only increase index  $j$
- 7~8 : exchange pivot with  $A[i + 1]$



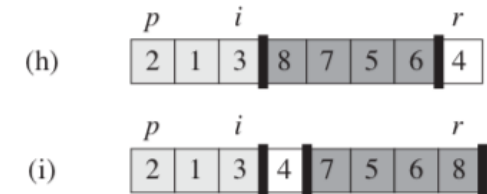
Line 1



Line 3~6  
First case



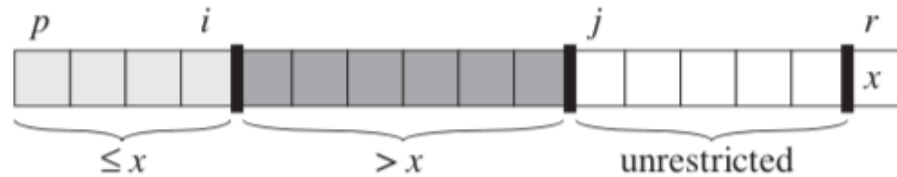
Line 3~6  
Second case



Line 7~8

## ❖ Description of Quicksort

### • Loop invariant of *PARTITION*



```

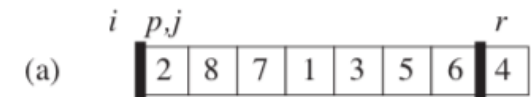
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
    
```

➤ For any array index  $k$ , always satisfy these three property

1. If  $p \leq k \leq i$ , then  $A[k] \leq x$ .  $\longrightarrow$  all elements of  $A[p..i]$  are always smaller than  $x$
2. If  $i + 1 \leq k \leq j - 1$ , then  $A[k] > x$ .  $\longrightarrow$  all elements of  $A[i + 1..j - 1]$  are always larger than  $x$
3. If  $k = r$ , then  $A[k] = x$ .  $\longrightarrow$   $A[r]$  is a pivot element

### ➤ Initialization

- ✓ Prior to the first iteration of the loop,  $i = p - 1$  and  $j = p$
- ✓ No value lie between  $p$  and  $i$  and between  $i + 1$  and  $j - 1$
- ✓ Loop invariant property 1,2 was satisfied
- ✓ Loop invariant property 3 is always true, because of line 1



## ❖ Description of Quicksort

### • Loop invariant of **PARTITION**

1. If  $p \leq k \leq i$ , then  $A[k] \leq x$ .  $\longrightarrow$  all elements of  $A[p..i]$  are always smaller than  $x$
2. If  $i + 1 \leq k \leq j - 1$ , then  $A[k] > x$ .  $\longrightarrow$  all elements of  $A[i + 1..j - 1]$  are always larger than  $x$
3. If  $k = r$ , then  $A[k] = x$ .  $\longrightarrow$   $A[r]$  is a pivot element

### ➤ Maintenance

✓ Consider two cases, depending on outcome of the test in line 4

✓ First case (a) :  $A[j] > x$

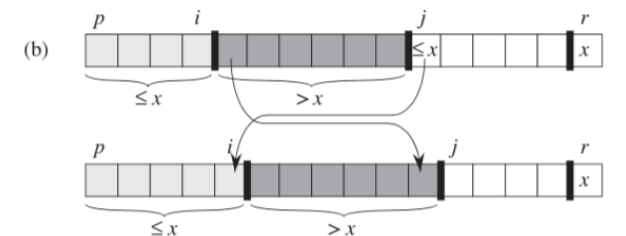
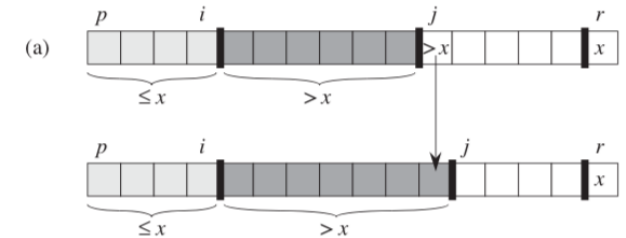
- i. Only action in the loop is to increment  $j$
- ii. After  $j$  incremented, condition 2 holds for  $A[j - 1]$  and all other entries remain unchanged

✓ Second case (b) :  $A[j] \leq x$

- i. The loop increments index  $i$
- ii. Swap  $A[i]$  and  $A[j]$ , and then increments  $j$
- iii.  $A[i] \leq x$  and  $A[j - 1] \geq x$
- iv.  $A[i] \leq x$  Condition 1,2 holds

```

3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
    
```



## ❖ Description of Quicksort

- Loop invariant of ***PARTITION***

- |   |        |  |
|---|--------|--|
| 1. If $p \leq k \leq i$ , then $A[k] \leq x$ .      | —————> | all elements of $A[p..i]$ are always smaller than $x$        |
| 2. If $i + 1 \leq k \leq j - 1$ , then $A[k] > x$ . | —————> | all elements of $A[i + 1..j - 1]$ are always larger than $x$ |
| 3. If $k = r$ , then $A[k] = x$ .                   | —————> | $A[r]$ is a pivot element                                    |

### ➤ Termination

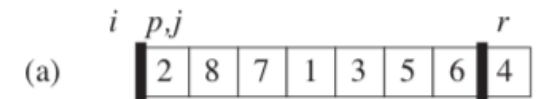
- ✓ At termination  $j = r$ , therefore every entry in the array is in one of the three sets described by the invariant



- The running time of ***PARTITION***

- $\theta(n)$
- Index  $j$  iterate  $n-1$  terms

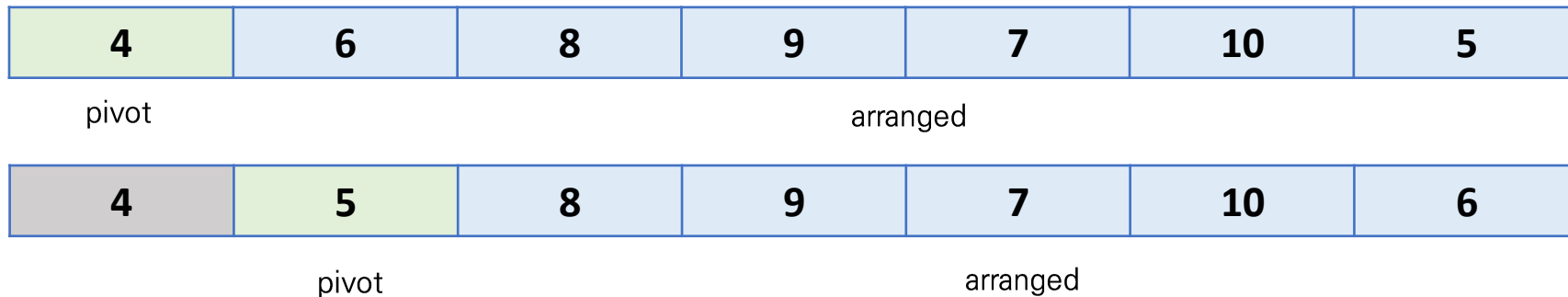
```
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
7  exchange  $A[i + 1]$  with  $A[r]$ 
8  return  $i + 1$ 
```



## ❖ Performance of Quicksort

### • Worst-case Partitioning

- Worst-case behavior for quicksort occurs when the partitioning routine produces one subproblem with  $n - 1$  elements and one with 0 elements
- Assume that this unbalanced partitioning arises in each recursive call



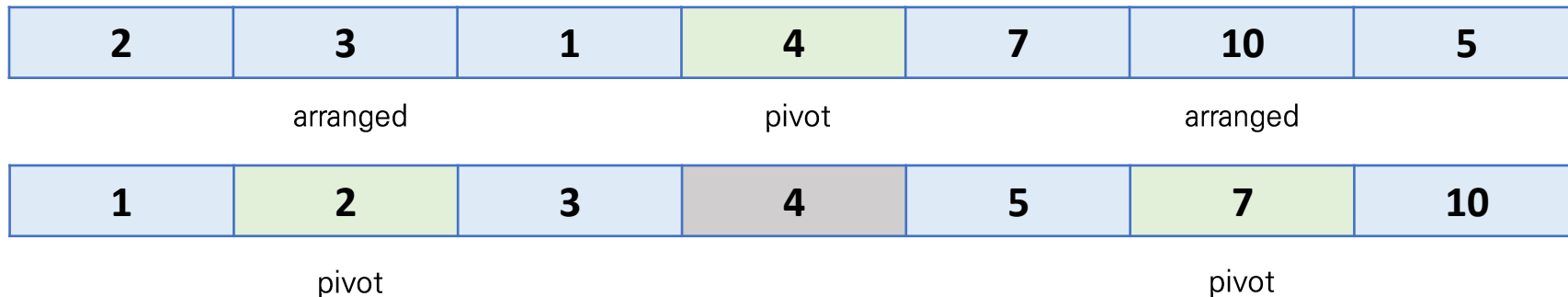
- The running time of *PARTITION* :  $\theta(n)$
- The running time of *PARTITION* in array size 0 (just return) :  $\theta(1)$
- The running time of next iteration *PARTITION* :  $T(n - 1)$

$$\begin{aligned} T(n) &= T(n - 1) + T(0) + \Theta(n) \\ &= T(n - 1) + \Theta(n) . \end{aligned} \quad \longrightarrow \quad \begin{aligned} \sum_{k=1}^n k &= \frac{1}{2}n(n + 1) \\ &= \Theta(n^2) . \end{aligned}$$

## ❖ Performance of Quicksort

- Best-case Partitioning

- Best-case behavior for quicksort occurs when the partitioning routine produces each size  $\lfloor n/2 \rfloor$  and  $\lfloor n/2 - 1 \rfloor$
- Assume that this balanced partitioning arises in each recursive call

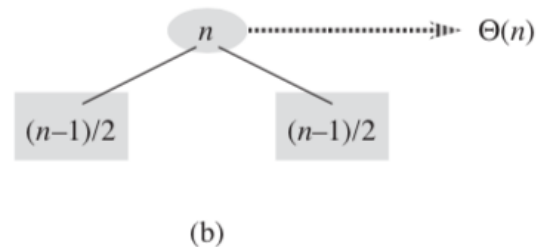
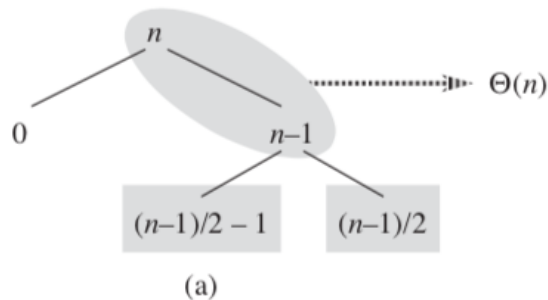


- The running time of *PARTITION* :  $\theta(n)$
- The running time of next iteration *PARTITION* :  $2T(n/2)$
- $T(n) = 2T(n/2) + \Theta(n)$
- Using master theorem,  $T(n) = \theta(n \lg n)$

## ❖ Performance of Quicksort

- average-case Partitioning

- Expect that some of the splits will be reasonably well balanced and that some will be fairly unbalanced
- *PARTITION* produces a mix of “good” and “bad” splits



- (a) : mix of “bad” and “good” split = takes partitioning cost  $\theta(n) + \theta(n-1) = \theta(n)$
  - (b) : only “good” split = takes partitioning cost  $\theta(n)$
  - The cost of (a) and (b) are same!!
- 
- In average case the running time for partitioning is similar with best-case, still have  $O(n \lg n)$  time

# Thank You!