

# Lecture 5: Binary heaps

Sorting algorithms: Heapsort and Quicksort

# Binary heaps

What is a binary heap?

- array  $A$  of objects with 2 special attributes:  $A.length$  and  $A.heap\_size$ .
- it represents a complete binary tree with  $A.heap\_size$  nodes
  - The tree is completely filled on all levels except possibly the lowest, which is filled from left to right
  - $A.length$  represents the maximum number of nodes of the tree. Therefore,  $A.heap\_size \leq A.length$
- The index of the **parent**, **left child**, and **right child** of a node with index  $i$  are computed as follows:

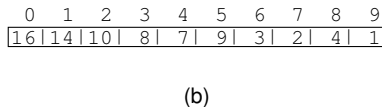
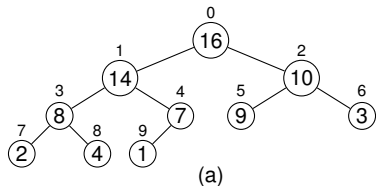
$$parent(i) := \begin{cases} \lfloor (i-1)/2 \rfloor & \text{if } i \neq 0 \\ -1 & \text{if } i = 0 \end{cases}$$

$$left(i) := 2 \cdot i + 1$$

$$right(i) := 2 \cdot i + 2$$

- The **heap property** must hold:  $A[parent(i)] \geq A[i]$  for all  $i \neq 0$ .

# Binary heaps: Example



A heap viewed as **(a) a binary tree** and **(b) an array**. The number within the circle at each node in the tree is the value stored at that node. The number next to a node is the corresponding index in the array.

## AUXILIARY NOTIONS

- **height of a node** in a tree := maximum number of edges from that node to a leaf.
- **height of the tree** := height of the root of the tree.

- The height of a binary heap is  $\Theta(\log_2(n))$  – obvious.
- FIND / INSERT / REMOVE operations in binary heaps take  $O(\log_2(n))$  time – we shall prove this.
- We are interested in the efficient implementation of:
  - 1 HEAPIFY( $A, i$ )
  - 2 BUILDHEAP( $A$ )
  - 3 HEAPSORT( $A$ )
  - 4 EXTRACTMAX( $A$ )
  - 5 INSERT( $A, key$ )

The purpose of these procedures will be explained later.

# HEAPIFY( $A, i$ )

- Takes as input an array  $A$  and an index  $i$ , such that
  - the subtrees rooted at  $left(i)$  and  $right(i)$  are binary heaps.
  - The subtree rooted at  $i$  may not be a binary heap, because  $A[i]$  is smaller than its children.
- Rearranges the elements of  $A$  by letting  $A[i]$  "float down" so that the subtree rooted at index  $i$  becomes a binary heap.

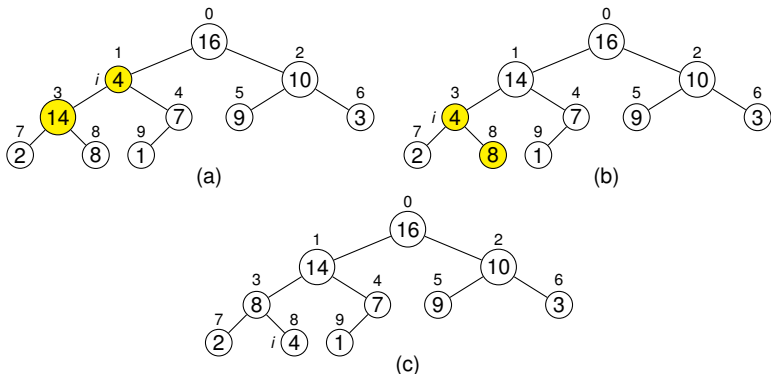
**Thus, the purpose of HEAPIFY is to maintain the heap property of an array of values.**

# HEAPIFY( $A, i$ )

```
HEAPIFY( $A, i$ )
1  $l := \text{left}(i)$ 
2  $r := \text{right}(i)$ 
3 if  $l < A.\text{heap\_size}$  and  $A[l] > A[i]$ 
4    $\text{largest} := l$ 
5 else  $\text{largest} := i$ 
6 if  $r < A.\text{heap\_size}$  and  $A[r] > A[\text{largest}]$ 
7    $\text{largest} := r$ 
8 if  $\text{largest} \neq i$ 
9   exchange  $A[i] \leftrightarrow A[\text{largest}]$ 
10  HEAPIFY( $A, \text{largest}$ )
```

# Example

The action of  $\text{HEAPIFY}(A, 1)$ , where  $A.\text{heap\_size} = 10$ . Configuration **(a)** lacks heap property at index 1. The heap property for index 1 is restored in **(b)** by exchanging  $A[1]$  with  $A[3]$ , which destroys the heap property for index 3. The recursive call  $\text{HEAPIFY}(A, 3)$  sets  $i = 3$ , swaps  $A[3] \leftrightarrow A[8]$  as shown in **(c)**, and the recursive call  $\text{HEAPIFY}(A, 8)$  yields no further change to the data structure.



# Properties of HEAPIFY

- The running time complexity of  $\text{HEAPIFY}(A, i)$  is  $O(h)$ , where  $h$  is the height of node with index  $i$ .
- $\Rightarrow$  In general, the running time of  $\text{HEAPIFY}(A, i)$  is  $O(\log_2(n))$ .
- For a proof, check the references.



# Building a binary heap

BUILDHEAP( $A$ )

- Rearranges the elements of an array  $A$ , to have the binary heap property.
- The rearrangement is achieved by successive runs of HEAPIFY( $A, i$ )

BUILDHEAP( $A$ )

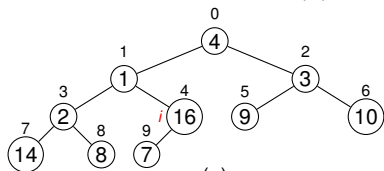
```
1  heap_size( $A$ ) :=  $A.length$ 
2  for  $i := \lfloor (A.length - 1)/2 \rfloor$  downto 0
3      HEAPIFY( $A, i$ )
```

## Remarks

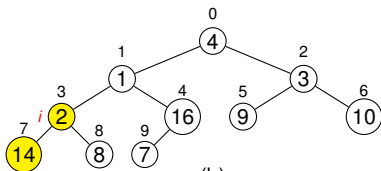
- The order in which the nodes are processed guarantees that the subtrees rooted at children of a node  $i$  are heaps before HEAPIFY is run at that node.
- There are  $O(n)$  calls of HEAPIFY( $A, i$ ), which has time complexity  $O(\log_2 n) \Rightarrow$  time complexity  $O(n \log_2 n)$ .
- Tighter bound of the total runtime of step 3:  $O(n)$  (see refs.)

# Example

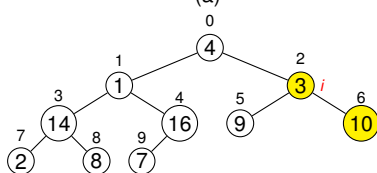
BUILDHEAP(A) for  $A=\{4,1,3,2,16,9,10,14,8,7\}$ .



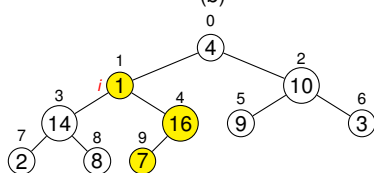
(a)



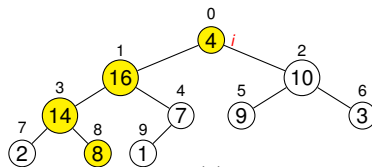
(b)



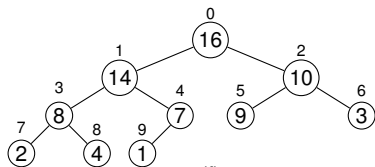
(c)



(d)



(e)



(f)

# The Heapsort algorithm

HEAPSORT( $A$ ) rearranges the elements of an array  $A$  in ascending order, using the following method:

- 1 Call **BUILDHEAP**( $A$ )  $\Rightarrow$  a heap on the elements of the array  $A[0..n-1]$
- 2  $A[0]$  is the maximum element of  $A$ 
  - ▷ exchange  $A[0] \leftrightarrow A[n-1]$ , to place  $A[0]$  into its correct final position.
- 3 Discard  $A[n-1]$  from the heap by decrementing  $A.heap\_size$ . We still have to sort  $A[0..n-2]$ 
  - $A[0..n-2]$  is *almost* a binary heap: 0 is the only index that may violate the heap property.
  - We run **HEAPIFY**( $A, 0$ ) to rearrange  $A[0..n-2]$  into binary heap.
  - The Heapsort algorithm **repeats this process** for the heap of size  $n-1$  down to a heap of size 2.

# Heapsort

HEAPSORT( $A$ )

1 BUILDHEAP( $A$ )

2 **for**  $i := A.length - 1$  **downto** 1

3     exchange  $A[0] \leftrightarrow A[i]$

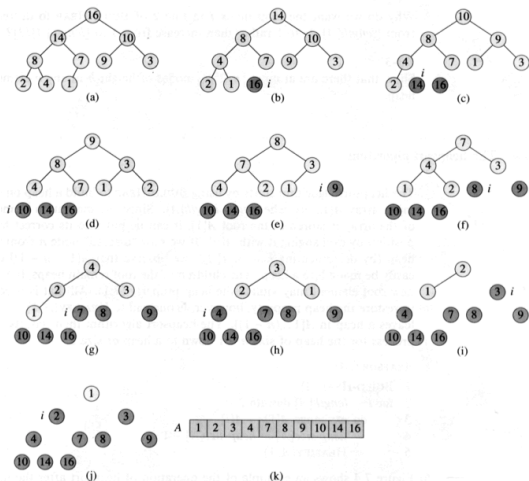
4      $A.heap\_size := A.heap\_size - 1$

5     HEAPIFY( $A, 0$ )

## TIME COMPLEXITY ANALYSIS

- BUILDHEAP( $A$ ) takes  $O(n)$  time.
  - There are  $n - 1$  calls to HEAPIFY( $A, 0$ ), and each one takes  $O(\log_2 n)$  time.
- ⇒ HEAPSORT( $A$ ) takes  $O(n \log_2 n)$  time, where  $n = A.length$ .

# Heapsort – running example



**(a)** The heap data structure just after it has been built by BUILDHEAP. **(b)–(j)** The heap just after each call of HEAPIFY in line 5. The value of  $i$  at that time is shown. Only lightly shaded nodes remain in the heap. **(k)** The resulting sorted array  $A$ .

# Priority queues

A **priority queue** is a data structure for maintaining a set  $S$  of elements, each with an associated value called a **key**. It is intended to support efficient execution of the following operations:

- **INSERT( $S, x$ )**: inserts the element  $x$  into a set  $S$ . We denote this operation by  $S := S \cup \{x\}$ .
- **MAXIMUM( $S$ )**: returns the element of  $S$  with the largest key.
- **EXTRACTMAX( $S$ )**: removes and returns the element of  $S$  with the largest key.

## Applications of priority queues

- Job scheduling on a shared resource
  - The queue keeps track of jobs to be performed, and their relative priorities.
  - When a job is finished or interrupted, the highest-priority job is selected from the queue, using **EXTRACTMAX**
  - New jobs can be added at any time using **INSERT**
- Event-driven simulation: time of event occurrence serves as its key.

# Priority queues

Can be implemented efficiently using binary heaps.

EXTRACTMAX( $A$ )

```
1  if  $A.heap\_size < 1$ 
2    error "heap underflow"
3   $max := A[0]$ 
4   $A[0] := A[A.heap\_size - 1]$ 
5   $A.heap\_size := A.heap\_size - 1$ 
6  HEAPIFY( $A, 0$ )
7  return  $max$ 
```

Running time analysis

- HEAPIFY( $A, 0$ ) takes  $O(\log_2 n)$  time  
⇒ EXTRACTMAX( $A$ ) takes  $O(\log_2 n)$  time.

# Priority queues

INSERT( $A, key$ )

INSERT( $A, key$ ) inserts a node into a binary heap  $A$ :

- First, it expands the heap by adding a new leaf to the tree.
- Then, it traverses a path from this leaf toward the root, to find a proper place for the new element.

INSERT( $A, key$ )

```
1  $A.heap\_size := A.heap\_size + 1$ 
2  $i := A.heap\_size - 1$ 
3 while  $i > 0$  and  $A[parent(i)] < key$ 
4      $A[i] := A[parent(i)]$ 
5      $i := parent(i)$ 
6  $A[i] := key$ 
```

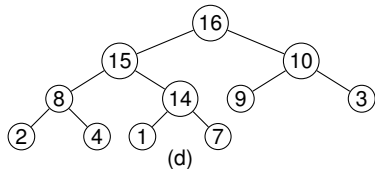
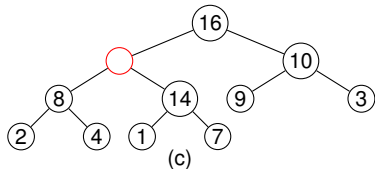
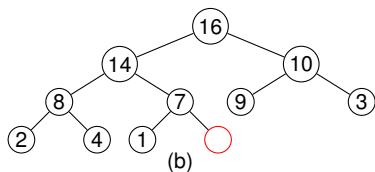
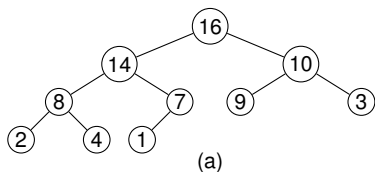
Running time analysis

- The path traced from the new leaf to the root has length  $O(\log_2 n) \Rightarrow$  HEAPINSERT( $A, key$ ) takes  $O(\log_2 n)$  time, where  $n = A.heap\_size$ .



# Priority queues

INSERT( $A, key$ ) illustrated



- (a)** The heap before we insert a node with key 15. **(b)** A new leaf is added to the tree. **(c)** Values on the path from the new leaf to the root are copied down until a place for the key 15 is found. **(d)** Key 15 is inserted into the tree.

- Sorting algorithm with worst-case running time  $\Theta(n^2)$  on an input array of  $n$  numbers.
- Very efficient on average:  $\Theta(n \log n)$
- Often, the best practical choice for sorting

# Quicksort

## Description of the algorithm

3-step divide-and-conquer algorithm for sorting a subarray  $A[p..r]$

**Divide:** The subarray  $A[p..r]$  is partitioned (rearranged) into two nonempty subarrays  $A[p..q]$ ,  $A[q + 1..r]$  such that

- The elements of  $A[p..q]$  are smaller than the elements of  $A[q + 1..r]$

The index  $q$  is computed as part of this partitioning procedure.

**Conquer:** The subarrays  $A[p..q]$  and  $A[q + 1..r]$  are sorted by recursive calls to quicksort.

**Combine:** Since the subarrays are sorted in place, no work is needed to combine them: the entire array  $A[p..r]$  is now sorted.

QUICKSORT( $A, p, r$ )

1. **if**  $p < r$
2.    $q \leftarrow$  PARTITION( $A, p, r$ )
3.   QUICKSORT( $A, p, q$ )
4.   QUICKSORT( $A, q + 1, r$ )

### Partitioning the array

PARTITION( $A, p, r$ )

```
1   $x \leftarrow A[p]$ 
2   $i \leftarrow p - 1$ 
3   $j \leftarrow r + 1$ 
4  while TRUE
5      do repeat  $j \leftarrow j - 1$ 
6          until  $A[j] \leq x$ 
7      repeat  $i \leftarrow i + 1$ 
8          until  $A[i] \geq x$ 
9      if  $i < j$ 
10         then exchange  $A[i] \leftrightarrow A[j]$ 
11         else return  $j$ 
```

# Quicksort

## How does PARTITION work?

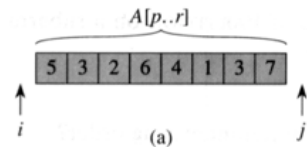
- ▶ Element  $x = A[p]$  from  $A[p..r]$  is selected as **pivot** around which to partition  $A[p..r]$ .
- ▶ The **while** loop grows two regions  $A[p..i]$  and  $A[j..r]$  from the top and bottom of  $A[p..r]$ , respectively, such that
  - Every element in  $A[p..i]$  is less than or equal to  $x$ .
  - Every element in  $A[j..r]$  is greater than or equal to  $x$ .

Initially,  $i = p - 1$  and  $j = r + 1$ , so the two regions are empty.

- ▶ Within the **while** loop, index  $j$  is decremented and index  $i$  is incremented, in lines 5-8, until  $A[i] \geq x \geq A[j]$ .
  - By exchanging  $A[i]$  and  $A[j]$ , the two regions can be extended.
- ▶ The **while** loop repeats until  $i \geq j$ , at which point the entire array  $A[p..r]$  has been partitioned into two subarrays  $A[p..q]$  and  $A[q + 1..r]$  where  $p \leq q < r$ , such that all elements in  $A[p..q]$  are smaller than or equal to any element in  $A[q + 1..r]$ .
- ▶ The value  $q = j$  is returned at the end of the procedure.

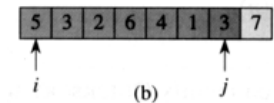
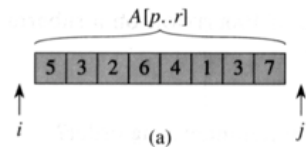
# Quicksort

Example of how PARTITION works



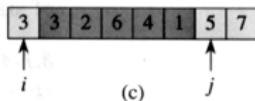
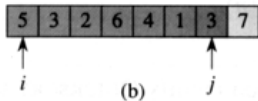
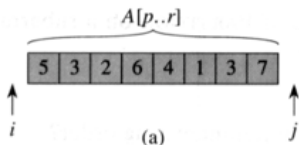
# Quicksort

Example of how PARTITION works



# Quicksort

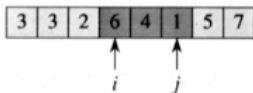
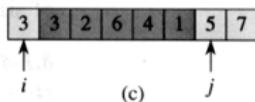
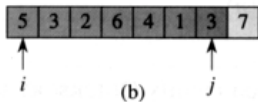
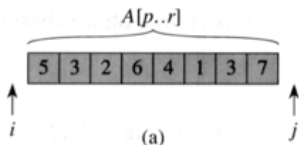
Example of how PARTITION works





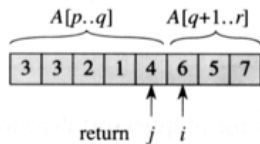
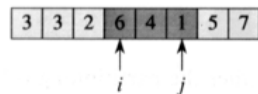
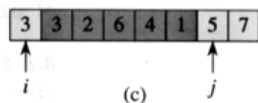
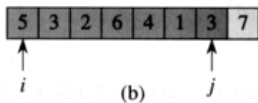
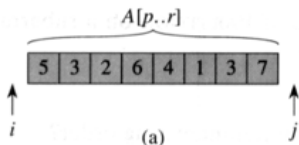
# Quicksort

Example of how PARTITION works



# Quicksort

Example of how PARTITION works



- The running time of PARTITION on an array  $A[p..r]$  is  $\Theta(r - p + 1)$ .
- Worst case behavior happens when the partitioning always produces one partition with 1 element, and the other with all the rest. In this case:
  - Partitioning an array of size  $n$  takes  $\Theta(n)$  time and  $T(1) = \Theta(1)$ .
  - The recurrence relation is  $T(n) = T(n - 1) + \Theta(n - 1) = \dots = \sum_{k=1}^n \Theta(k) = \Theta(\sum_{k=1}^n k) = \Theta(n^2)$ .

$\Rightarrow$  in the worst case, the running time is  $\Theta(n^2)$ .
- Best case is when the partitioning produces regions of equal size  $\Rightarrow$  the recurrence relation  $T(n) = 2 T(n/2) + \Theta(n)$ .
  - $\Rightarrow T(n) = \Theta(n \log n)$   
(Cf. the Master Theorem)

Chapters 7 (Heapsort) and 8 (Quicksort) from the book

- Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest. *Introduction to Algorithms*. McGraw Hill, 2000.