

CH-231-A

**Algorithms and Data Structures**

ADS

**Lecture 13**

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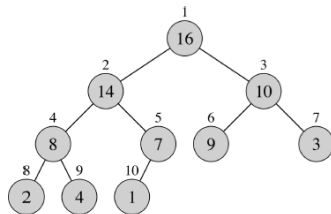
# Heap Sort: Motivation

- ▶ Try to develop an in-situ sorting algorithm with asymptotic runtime  $\Theta(n \lg n)$ .
- ▶ Use a sophisticated data structure to support the computations.

## Heap: Data structure

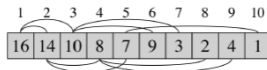
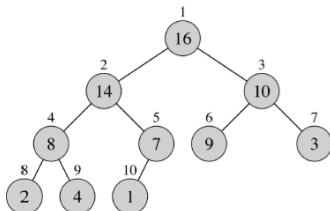
### Defintion:

A (binary) heap data-structure is an array which can be viewed as a nearly complete binary tree: each level is completely full except possibly the last level, which is filled from left to right.



## Heap as an Array (1)

A heap can be stored as an array:



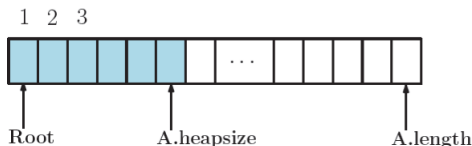
## Heap as an Array (2)

The array  $A$  representing the heap has two attributes:

- ▶  $A.length$
- ▶  $A.heapsize$

such that  $0 \leq A.heapsize \leq A.length$ .

There are only  $A.heapsize$  valid elements of the heap.

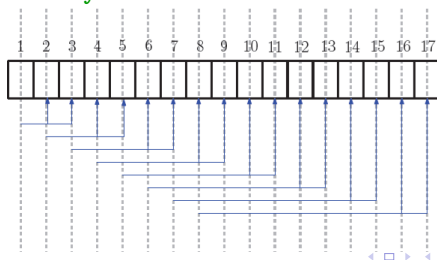


$A[1]$  is the root of the heap (root of the binary tree).

## Heap as an Array (3)

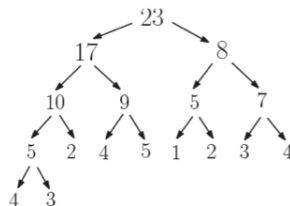
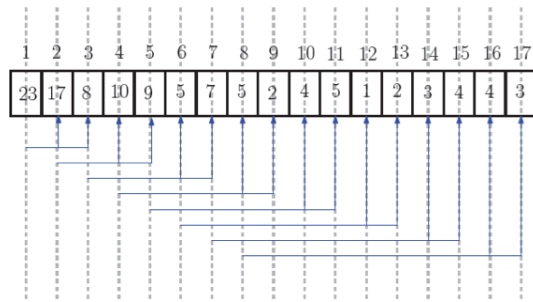
Given the index  $i$  of an element of  $A$ , we can calculate:

- ▶ `Parent(i): return floor(i/2);`  
`// Right shift by 1 bit`
- ▶ `Left(i): return 2i;`  
`// Left shift by 1 bit`
- ▶ `Right(i): return 2i + 1;`  
`// Left shift by 1 bit and set LSB to 1`



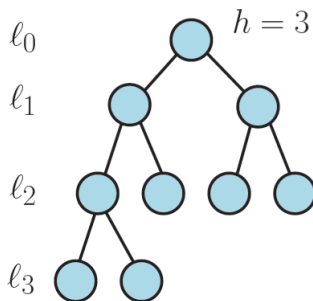
# Max-Heap Property

In a max-heap, for every node  $i$  (other than the root),  
 $A[\text{Parent}(i)] \geq A[i]$ .



## Recall: Height of a Tree

- ▶ The height of a node  $x$  is the length of the longest simple downward path from  $x$  to a leaf.
- ▶ The height of a tree is the height of its root.





# Heap Height (1)

Theorem:

A heap with  $n$  elements has height  $h = \lfloor \lg n \rfloor$ .

## Heap Height (2)

Proof:

Heap height  $h$  implies that there are  $h + 1$  levels (levels 0 to  $h$ ). As a heap is a nearly complete binary tree, the last guaranteed complete level is level  $h - 1$ .

The level  $h$  may be incomplete, but it has at least one element.

The number of elements in complete levels 0 to  $h - 1$  is

$$1 + 2 + 2^2 + \dots + 2^{h-1} = 2^h - 1.$$

So,  $n > 2^h - 1$  or (since it is an integer)  $n \geq 2^h$ .

If all levels 0 to  $h$  were complete, the number of elements would be  $2^{h+1} - 1$ .

So,  $n \leq 2^{h+1} - 1$ .

## Heap Height (3)

Proof (continued):

Combining the two inequalities:

$$2^h \leq n \leq 2^{h+1} - 1$$

As  $2^{h+1} > 2^{h+1} - 1 \geq 2^h$  for  $h \geq 0$ ,

$$h + 1 > \lg(2^{h+1} - 1) \geq h$$

Thus,  $\lg(2^{h+1} - 1) = h + \alpha$  with  $\alpha \in [0, 1)$ , which leads to

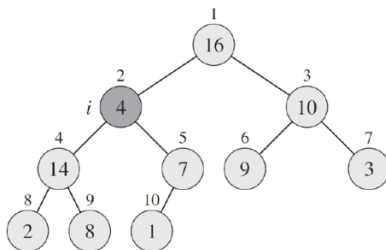
$$h \leq \lg n \leq h + \alpha \text{ with } \alpha \in [0, 1).$$

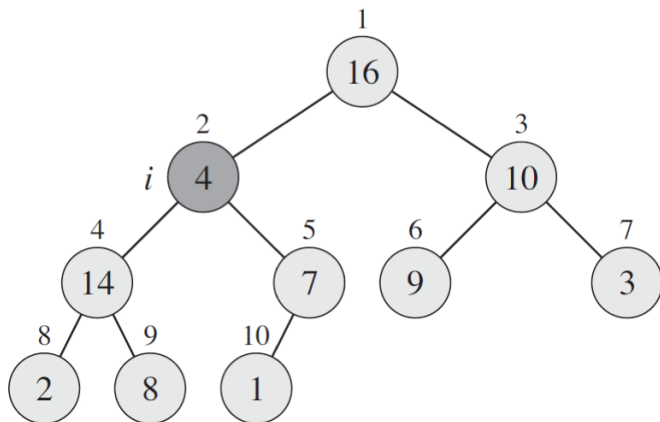
Hence,  $h = \lfloor \lg n \rfloor$ .

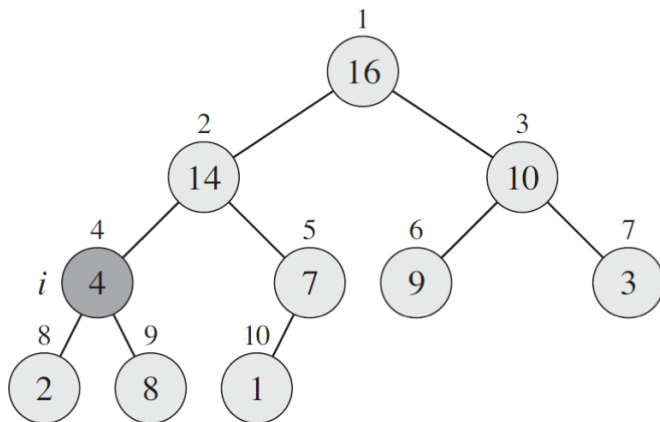
## $Max\text{-Heapify}(A, i)$ (1)

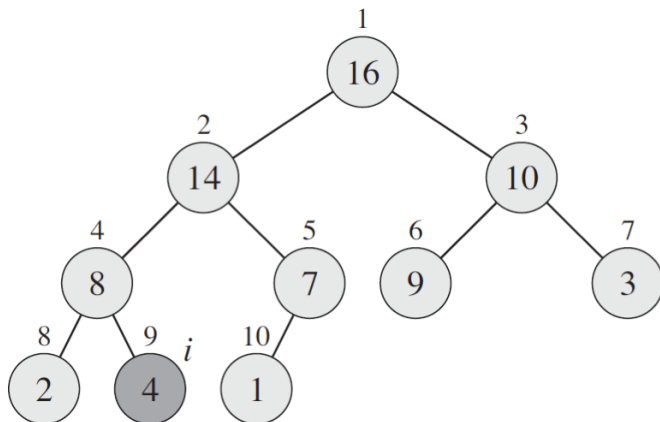
### Precondition:

When  $Max\text{-Heapify}(A, i)$  is called, binary-trees rooted at  $Left(i)$  and  $Right(i)$  are valid max-heaps, but  $A[i]$  may be smaller than its children.



$\text{Max-Heapify}(A, i)$  (2)

$\text{Max-Heapify}(A, i) \ (3)$ 

$\text{Max-Heapify}(A, i)$  (4)

## Max-Heapify( $A, i$ ) (5)

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



## *Max-Heapify*( $A, i$ ) (6)

Time complexity:

$$T(n) = O(h) = O(\lg n),$$

as in the worst case the element from position  $i$  has to go down all the way to the last level.

In the expression above  $h$  is the height of the element from position  $i$ .