# Data Structures and Algorithms: Lecture 6

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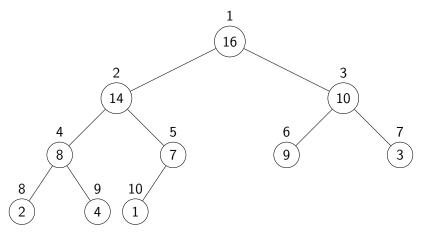
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## Heapsort

- ▶ running time:  $O(n \lg n)$ . Hence like MERGE-SORT, but not like INSERTION-SORT.
- sorts in place. Hence like INSERTION-SORT not like MERGE-SORT

### Heap

- Not a "garbage-collected storage"
- ▶ Definition: (binary) heap is a nearly complete binary tree
- Nearly complete means all levels of the tree are filled except the last one which is filled from the left to the right to some point.



## Stored as an array:

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

#### Stored as an array:

The array has two attributes:

- ► A.length: number of all elements in the array,
- A.heap—size: number of elements of the heap stored in the array

$$A.heap-size \leq A.length$$

The root of the tree is A[1].

#### Height of a node in a heap:

the number of edges on a longest path from the node to the leaf

### Height of the heap:

the height of the root

If the heap has n elements, its height is  $\Theta(\lg(n))$ .

Stored as an array:

Given an index i it is easy to compute the parent and children of i in the tree:

- ▶ Parent(i)
  - 1. return  $\lfloor \frac{i}{n} \rfloor$
- ► Left(*i*)
  - 1. **return** 2*i*

\\ shifting the binary representation of i to the left adding 0 at the end.

- ► RIGHTT(*i*)
  - 1. **return** 2i + 1

 $\backslash \backslash$  shifting the binary representation of i to the left adding 1 at the end.

# Two kinds of heaps

#### There are two kinds of heaps

- max-heaps: satisfy max-heap property
- min-heaps: satisfy min-heap property

#### Max-heap property:

$$\forall (1 \le i \le A.heap-size) \quad A[PARENT(i)] \ge A[i]$$

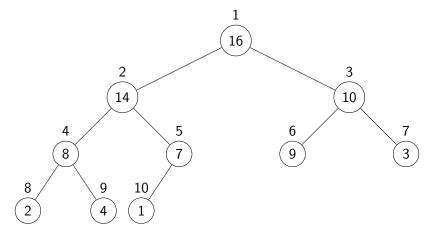
Maximal element is at the root.

#### Min-heap property:

$$\forall (1 \leq i \leq A.heap-size) \quad A[PARENT(i)] \leq A[i]$$

Minimal element is at the root.

We will use max-heap for sorting.



# Maintaining the max-heap property

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Procedure MAX-HEAPIFY(A, i)
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```
1 I = Left(i)
2 r = RIGHT(i)
3 if 1 \le A. heap—size and A[I] > A[i] then
   largest = l
5 else
      largest = i
7 if r \le A.heap—size and A[r] > A[largest] then
      largest = r
9 if largest \neq i then
      exchange values A[i] and A[largest]
10
      MAX-HEAPIFY(A, largest)
11
```

Trace it on: A = <16, 4, 10, 14, 7, 9, 3, 2, 8, 1 >, i = 2.

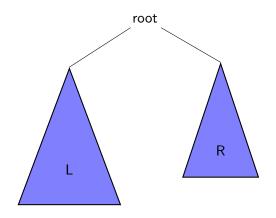
## Running time

- ▶ All lines of the algorithm take constant time, except the last line, which contains the recursive call
- What is the size of the input?
- ▶ The size is the size of the subtree rooted at *i*, call it *n*.
- ► The size of the input in the recursive call is the size of the subtree rooted at *largest*.

$$T(n) = T(\text{size of the subtree}) + \Theta(1)$$

In the worst case the subtree rooted at *largest* is as big as possible.

## Worst case



- Assume the number of nodes in R = k
- Number of nodes in L = k + (k+1)
- n = 1 + k + (2k + 1) = 3k + 2.
- ▶ Size of L is approx  $2/3 \cdot n$

#### Recurrence for MAX-HEAPIFY

The recurrence for Max-heapify is:

$$T(n) \leq T(\frac{2n}{3}) + \Theta(1)$$

#### Solution

$$a = 1, \quad b = 3/2, \quad f(n) = \Theta(1)$$

- $n^{\log_{3/2} 1} = n^0 = 1$
- $f(n) = \Theta(1) = \Theta(n^{\log_{3/2} 1})$

Case 2 of Master Theorem:

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

If h is the height of i, we can write it also O(h).

# Building a max-heap

#### Important observation

In any heap-array, the elements at the indexes:

$$\lfloor \frac{n}{2} \rfloor + 1, \lfloor \frac{n}{2} \rfloor + 2, \ldots, n$$

are leaves! Hence each of them is a heap of the height 0.

#### **Procedure** BUILD-MAX-HEAP(A)

- 1 A.heap-size = A.length
- 2 for  $i = \lfloor \frac{A.length}{2} \rfloor$  down to 1 do
- 3 MAX-HĒAPIFY(A, i)

#### Running time:

- ► Each call to MAX-HEAPIFY costs O(lg n)
- ▶ BUILD-MAX-HEAP makes *O*(*n*) such calls.
- ► Hence  $O(n \lg n)$  not tight!

# Better running time analysis

Notice that:

- ► Heap has height | Ig n |
- ► At each height *h* there are at most  $\lceil \frac{n}{2h+1} \rceil$  nodes.
- ▶ Question: How many times MAX-HEAPIFY is called starting with nodes at height *h* (going up the tree)?
- Answer:  $\sum_{l=0}^{\lfloor \lg n \rfloor} \lceil \frac{n}{2^{h+1}} \rceil$  times.

Hence the running time is:

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

Notice: 
$$\sum_{h=0}^{\infty} \frac{h}{2^h} = \frac{1/2}{(1-1/2)^2} = \frac{1/2}{(1/2)^2} = \frac{4}{2} = 2$$

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