

# Heaps. Heapsort.

(CLRS 6)

## 1 Introduction

- We have discussed several fundamental algorithms (sorting, selection, etc)
- We will now turn to *data structures*; They play an important role in algorithms design.
  - Today we discuss priority queues and next time structures for maintaining ordered sets.

## 2 Priority Queue

- A priority queue supports the following operations on a set  $S$  of  $n$  elements:
  - INSERT: Insert a new element  $e$  in  $S$
  - FINDMIN: Return the minimal element in  $S$
  - DELETEMIN: Delete the minimal element in  $S$
- Sometimes we are also interested in supporting the following operations:
  - CHANGE: Change the key (priority) of an element in  $S$
  - DELETE: Delete an element from  $S$
- Priority queues have many applications, e.g. in discrete event simulation, graph algorithms
- We can obviously sort using a priority queue:
  - Insert all elements using INSERT
  - Delete all elements in order using FINDMIN and DELETEMIN

## 3 Priority Queue implementations

### 3.1 A Priority Queue with an Array or List

- The first implementation that comes to mind is ordered array:

1	3	5	6	7	8	9	11	12	15	17
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- FINDMIN can be performed in  $O(1)$  time

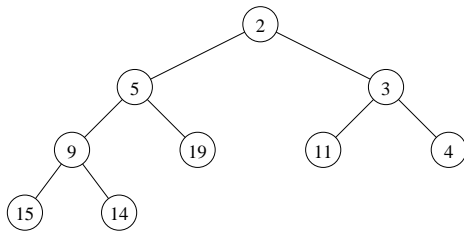
- DELETEMIN and INSERT takes  $O(n)$  time since we need to expand/compress the array after inserting or deleting element.
- If the array is unordered all operations take  $O(n)$  time.
- We could use double linked sorted list instead of array to avoid the  $O(n)$  expansion/compression cost
  - but INSERT can still take  $O(n)$  time.

### 3.2 A Priority Queue with a Heap

- The common way of implementing a priority queue is using a heap
- Heap definition:

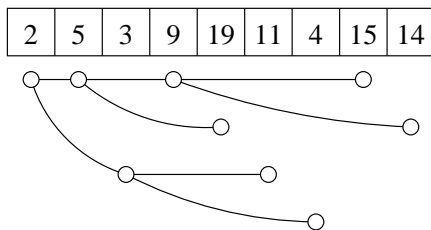
- Perfectly balanced binary tree
  - \* lowest level can be incomplete (but filled from left-to-right)
- For all nodes  $v$  we have  $\text{key}(v) \geq \text{key}(\text{parent}(v))$

- Example:



- Heap can be implemented (stored) in two ways (at least)
  - Using pointers
  - In an array level-by-level, left-to-right

Example:



- \* the left and right children of node in entry  $i$  are in entry  $2i$  and  $2i + 1$ , respectively
- \* the parent of node in entry  $i$  is in entry  $\lfloor \frac{i}{2} \rfloor$

- Properties of heap:
  - Height  $\Theta(\log n)$

– Minimum of  $S$  is stored in root

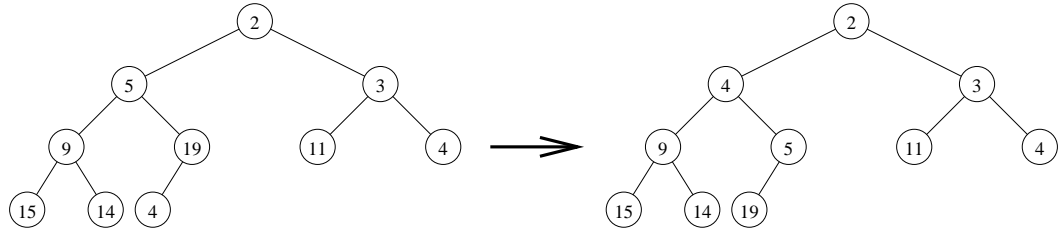
• Operations:

– INSERT

\* Insert element in new leaf in leftmost possible position on lowest level

\* Repeatedly swap element with element in parent node until heap order is reestablished (UP-HEAPIFY)

Example: Insertion of 4



– FINDMIN

\* Return root element

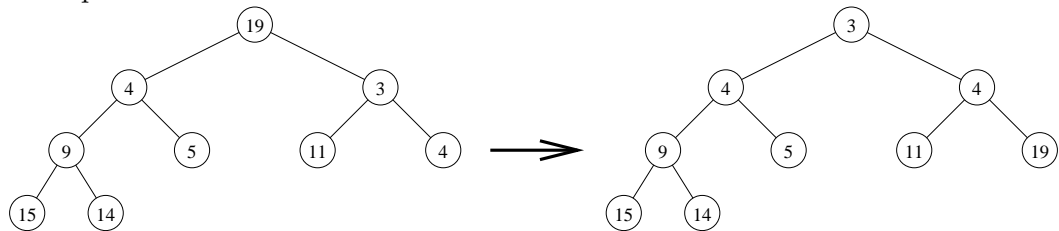
– DELETEMIN

\* Delete element in root

\* Move element from rightmost leaf on lowest level to the root (and delete leaf)

\* Repeatedly swap element with the smaller of the children elements until heap order is reestablished (DOWN-HEAPIFY)

Example:



– CHANGE and DELETE can be handled similarly in  $O(\log n)$  time

\* Note: Assuming that we know the element to be changed/deleted (we cannot search in a heap!!)

• **Correctness:** Exercise.

• **Running time:** All operations traverse at most one root-leaf path  $\Rightarrow O(\log n)$  time.

• Sorting using heap (*HeapSort*) takes  $\Theta(n \log n)$  time.

–  $n \cdot O(\log n)$  time to insert all elements (build the heap)

–  $n \cdot O(\log n)$  time to output sorted elements

• Sometimes we would like to build a heap faster than  $O(n \log n)$

- BUILDHEAP
  - \* Insert elements in any order in perfectly balanced tree
  - \* DOWN-HEAPIFY all nodes level-by-level, bottom-up
- Correctness:
  - \* Induction on height of tree: When doing level  $i$ , all trees rooted at level  $i - 1$  are heaps.
- Analysis:
  - \* The leaves are at height 0, the root is at height  $\log n$
  - \*  $n$  elements  $\Rightarrow \leq \lceil \frac{n}{2} \rceil$  leaves  $\Rightarrow \lceil \frac{n}{2^i} \rceil$  elements at height  $h$
  - \* Cost of DOWN-HEAPIFY on a node at height  $h$  is  $h$
  - \* Total cost:  $\sum_{i=1}^{\log n} h \cdot \lceil \frac{n}{2^i} \rceil = \Theta(n) \cdot \sum_{i=1}^{\log n} \frac{h}{2^i}$
  - \* It can be shown that  $\sum_{i=1}^{\log n} \frac{h}{2^i} = O(1) \implies$  the total buildheap cost is  $\Theta(n)$
  - \* Computing  $\sum_{i=1}^n \frac{h}{2^i}$  and  $\sum_{i=1}^{\infty} \frac{h}{2^i}$ 
    - Differentiate  $\sum_{h=0}^n x^h = \frac{1-x^{n+1}}{1-x}$ , respectively  $\sum_{h=0}^{\infty} x^h = \frac{1}{1-x}$  (assuming  $|x| < 1$ )
    - $\sum_{h=0}^{\infty} hx^{h-1} = \frac{1}{(x-1)^2} \Rightarrow \sum_{h=0}^n hx^h = \frac{x}{(x-1)^2} \Rightarrow \sum_{h=0}^n \frac{h}{2^h} = \frac{1/2}{(1/2-1)^2} = O(1)$