# Heaps. Heapsort.

(CLRS 6)

## 1 Introduction

So far we have discussed tools necessary for analysis of algorithms (growth, summations and recurrences) and we have seen a couple of sorting algorithms as case-studies.

Today we discuss a <u>data structure</u> called <u>priority queue</u>, and its implementation with a heap. The heap will lead us to a different algorithm for sorting, called <u>heapsort</u>.

## 2 Priority Queue

- A priority queue supports the following operations on a set S of n elements:
  - Insert a new element e in S
  - FINDMIN: Return the minimal element in S
  - Delete Temporary Delete the minimal element in S
- Sometimes we are also interested in supporting the following operations:
  - 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:

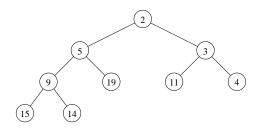


- FINDMIN can be performed in O(1) time

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

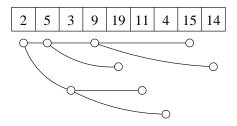
### 3.2 A Priority Queue with a Heap

- 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  $\underline{\text{key}(v)} > \underline{\text{key}(\text{parent}(v))}$
- Note: this is a *min-heap*; a symmetrical definition is possible, giving a *max-heap*.
- Example:



• The beauty of heaps is that <u>although they are trees</u>, <u>they can be implemented as arrays</u>. <u>The elements in the heap are stored level-by-level</u>, <u>left-to-right in the array</u>.

#### Example:



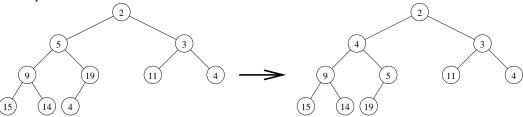
- the left and right children of node in entry  $\underline{i}$  are in entry  $\underline{2i}$  and  $\underline{2i+1}$ , respectively
- the parent of node in entry i is in entry  $\lfloor \frac{i}{2} \rfloor$
- Properties of heap:

- Height  $\Theta(\log n)$
- For a min-heap: Minimum of S is stored in root (for a max-heap, the maximum element is stored in the root).

#### • Operations:

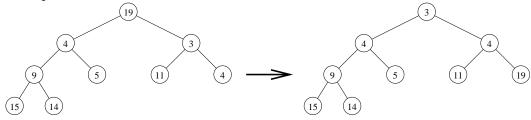
- Insert
  - \* Insert element in new leaf in leftmost possible position on lowest level
  - \* Repeatedly swap element with element in parent node until heap order is reestablished (this is referred to as <u>UP-HEAPIFY</u>.

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 (this is refered to as DOWN-HEAPIFY or sometimes just HEAPIFY).

Example:



- By default Heapify works on the root node (i = 1). Heapify(i) means it's called on node i in the heap. Prior to this call, the left and right children of node i must be heaps. After Heapify (i) is complete, the tree rooted at node i is a heap.
- Changing the priority of a given node or deleting a given node can be handled similarly in  $O(\log n)$  time.
  - \* Note: We can delete or update nodes in a heap if we are given their index in the array. For e.g. we cannot say "delete the node with priority 37" because we cannot search (efficiently) in a heap! But we can say "delete the node at index 5".
- Running time: All operations traverse at most one root-leaf path  $\Rightarrow O(\log n)$  time.

### 3.3 Heapsort

- Sorting using heap 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
- This is not in place. An in-place sorting algorithm with a heap is possible, and is reffered to as heapsort.
  - Build a max-heap
  - Repeatedly, delete the largest element, and put it at the end of the array.

### 3.4 Building a heap in O(n) time

- Sometimes we would like to build a heap faster than  $O(n \log n)$
- By default Heapify works on the root node (i = 1). Heapify(i) means it's called on node i in the heap. Prior to this call, the left and right children of node i must be heaps. After Heapify (i) is complete, the tree rooted at node i is a heap.
  - BUILDHEAP (A)
    - \* DOWN-HEAPIFY all nodes level-by-level, bottom-up (starting at node n/2)
  - 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  $\log n$
    - \* Cost of DOWN-HEAPIFY on a node at height h is h
    - \* n elements  $\Rightarrow \leq \lceil \frac{n}{2} \rceil$  leaves, ...,  $\lceil \frac{n}{2^h} \rceil$  elements at height h
    - \* Total cost:  $\sum_{i=1}^{\log n} h \cdot \lceil \frac{n}{2^h} \rceil = \Theta(n) \cdot \sum_{i=1}^{\log n} \frac{h}{2^h}$
    - \* It can be shown that  $\sum_{i=1}^{\log n} \frac{h}{2^h} = O(1) \Longrightarrow$  the total buildheap cost is  $\Theta(n)$