

# Topic 15: Heaps and Heapsort

(CLRS 6)

CPS 230, Fall 2001

## 1 Introduction

- Data structures play an important role in algorithms design.
  - we will now discuss priority queues and 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  $S$ .
- We can obviously sort using a priority queue:
  - Insert all elements using INSERT.
  - Delete all elements in order using FINDMIN and DELETEMIN.
- Priority queues have many other applications, e.g. in scheduling, discrete event simulation, and graph algorithms.

### 2.1 Array or List implementations

- The first implementation that comes to mind is sorted 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 (i.e., not sorted), then all operations take  $O(n)$  time.
- We could use a doubly-linked sorted list instead of an array to avoid the  $O(n)$  cost of expansion or compression.
  - But INSERT will still take  $O(n)$  time.

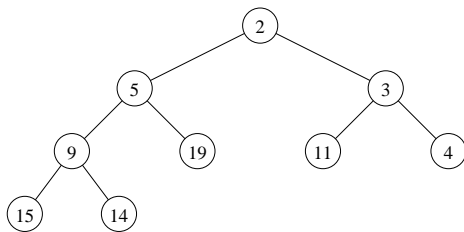
## 2.2 Heap implementation

- One way of implementing a priority queue is using a heap.
  - Note: We use a min-heap; the book describes a max-heap.

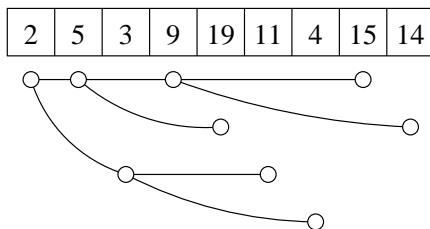
- Heap definition:

- Perfectly balanced binary tree.
  - \* Lowest level can be incomplete (but filled from left-to-right).
- For all non-root nodes  $v$ , we have  $key(v) \geq key(parent(v))$ .

- Example:



- Heap can be implemented (stored) in two ways (at least):
  - Using pointers.
  - In an array level-by-level, left-to-right. The array representation of previous example is pictured below. The left son is pictured on the same vertical level as the parent, and the right son is pictured lower.



- \* Note the nice property that the left and right children of the node stored in entry  $i$  are stored in entries  $2i$  and  $2i + 1$ , respectively.

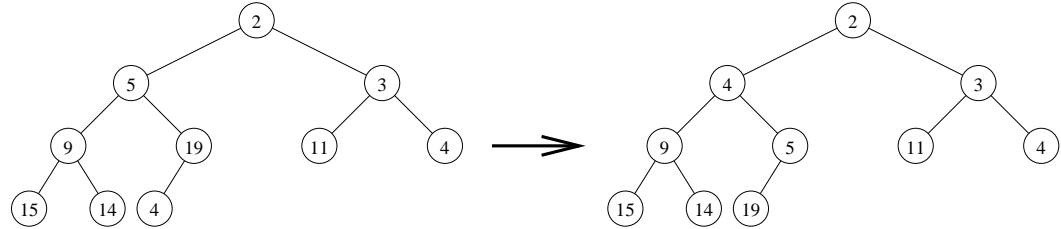
- Properties of heap:
  - Height  $\Theta(\log n)$ .
  - Minimum of  $S$  is stored in root.

- Operations:

- INSERT

- \* Insert the new element in a new leaf at the leftmost possible position on lowest level.
- \* Repeatedly swap the newly inserted element with the element in its parent node until the heap order is reestablished (UP-HEAPIFY).

Example: Insertion of 4



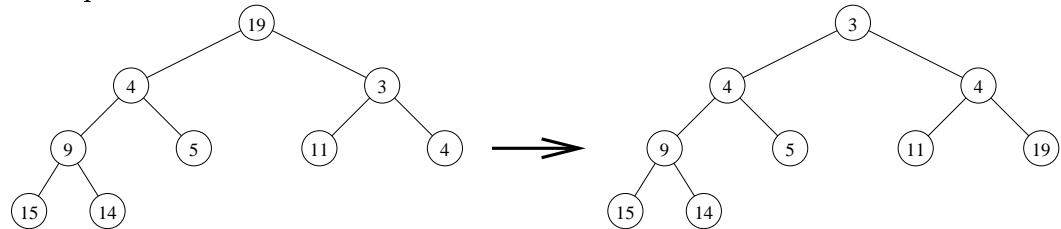
- FINDMIN

- \* Return root element.

- DELETEMIN

- \* Delete element in root.
- \* Move element from rightmost leaf on lowest level to the root (and delete leaf).
- \* Repeatedly swap the element with the smaller of its two children until the heap order is reestablished (DOWN-HEAPIFY).

Example:



- All operations traverse at most one root-leaf path  $\implies O(\log n)$  running time..
- CHANGE and DELETE can be handled similarly in  $O(\log n)$  time.
  - Assuming that we know the element to be changed/deleted.

### 2.3 Heapsort

- Sorting using a heap, which we call *Heapsort*, takes  $\Theta(n \log n)$  time.
  - $n \cdot O(\log n)$  time to insert all elements into the heap.
  - $n \cdot O(\log n)$  time to output the sorted elements by a sequence of DELETEMIN operations.
- Sometimes we would like to build a heap in  $O(n)$  time rather than in  $O(n \log n)$  time. For example, we can get a constant-factor improvement in sorting if we come up with a very efficient way of initially constructing the heap of  $n$  elements. The DELETEMIN operations still take  $O(n \log n)$  time.

- Place elements in any order in a 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 to show  $O(n)$  time for DOWN-HEAPIFY:

- Define leaves to be on level 1 (root on level  $\log n$ ).
- $n$  elements  $\implies \leq \lceil \frac{n}{2} \rceil$  leaves  $\implies \lceil \frac{n}{2^h} \rceil$  elements on level  $h$ .
- Cost of DOWN-HEAPIFY on a node on level  $h$  is  $h$ .
- Total cost:  $\Theta(\sum_{i=1}^{\log n} h \cdot \lceil \frac{n}{2^h} \rceil) = \Theta(n \sum_{i=1}^{\log n} \frac{h}{2^h})$ .
- $\sum_{i=1}^{\log n} \frac{h}{2^h} < 2$ , and thus cost of DOWN-HEAPIFY is  $\Theta(n)$ . Proof:
  - \* Assume  $|x| < 1$  and differentiate  $\sum_{h=0}^{\infty} x^h = \frac{1}{x-1}$ .
  - \* Derivative of LHS is  $\sum_{h=0}^{\infty} h x^{h-1}$ , and derivative of RHS is  $\frac{1}{(x-1)^2}$ .
  - $\implies \sum_{h=0}^{\infty} h x^h = \frac{x}{(x-1)^2} \implies \sum_{h=0}^{\infty} \frac{h}{2^h} = \frac{1/2}{(1/2-1)^2} = 2$ .