

Module 2: Priority Queues

CS 240 - Data Structures and Data Management

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Based on lecture notes by many previous cs240 instructors

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Winter 2015

Abstract Data Types

Abstract Data Type (ADT): A description of *information* and a collection of *operations* on that information.

The information is accessed *only* through the operations.

We can have various *realizations* of an ADT, which specify:

- How the information is stored (*data structure*)
- How the operations are performed (*algorithms*)

Dynamic Arrays

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Dynamic arrays offer a compromise:

$O(1)$ element access, and $O(1)$ insertion/deletion *at the end*.

Two realizations of dynamic arrays:

- Allocate one HUGE array, and only use the first part of it.
- Allocate a small array initially, and double its size as needed.
(*Amortized analysis* is required to justify the $O(1)$ cost for
insertion/deletion at the end — take CS 341/466!)

Stack ADT

Stack: an ADT consisting of a collection of items with operations:

- *push*: inserting an item
- *pop*: removing the most recently inserted item

Items are removed in LIFO (*last-in first-out*) order.

We can have extra operations: *size*, *isEmpty*, and *top*

Applications: Addresses of recently visited sites in a Web browser,
procedure calls

Realizations of Stack ADT

- using arrays
- using linked lists

Queue ADT

Queue: an ADT consisting of a collection of items with operations:

- *enqueue*: inserting an item
- *dequeue*: removing the least recently inserted item

Items are removed in FIFO (*first-in first-out*) order.

Items enter the queue at the *rear* and are removed from the *front*.

We can have extra operations: *size*, *isEmpty*, and *front*

Realizations of Queue ADT

- using (circular) arrays
- using linked lists

Priority Queue ADT

Priority Queue: An ADT consisting of a collection of items (each having a *priority*) with operations

- *insert*: inserting an item tagged with a priority
- *deleteMax*: removing the item of *highest priority*

deleteMax is also called *extractMax*.

Applications: typical “todo” list, simulation systems

The above definition is for a *maximum-oriented* priority queue. A *minimum-oriented* priority queue is defined in the natural way, by replacing the operation *deleteMax* by *deleteMin*.

Using a Priority Queue to Sort

PQ – Sort(A)

1. initialize *PQ* to an empty priority queue
2. **for** $i \leftarrow 0$ **to** $n - 1$ **do**
3. *PQ.insert(A[i], A[i])*
4. **for** $i \leftarrow 0$ **to** $n - 1$ **do**
5. $A[n - 1 - i] \leftarrow PQ.deleteMax()$

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Attempt 2: Use *sorted arrays*

- insert: $O(n)$ *Need to shift rest of array over after binary search*
- deleteMax: $O(1)$

Using sorted linked-lists is identical.

This realization used for sorting yields *insertion sort*.

Insertion Sort

↳ Not too bad

↳ If the list is almost sorted,
then it is much faster

↳ It is an online algorithm

↳ Ez 2 implement & "Natural"

Third Realization: Heaps

A **heap** is a certain type of binary tree.

Recall binary trees:

A binary tree is either

- empty, or
- consists of three parts: a node and two binary trees (left subtree and right subtree).

Terminology: root, leaf, parent, child, level, sibling, ancestor, descendant, etc. .

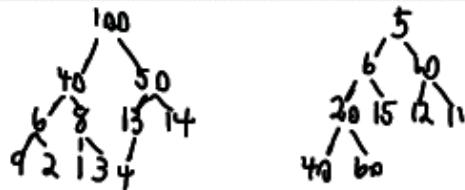
Heaps

A **max-heap** is a binary tree with the following two properties:

- ① **Structural Property:** All the levels of a heap are completely filled, except (possibly) for the last level. The filled items in the last level are *left-justified*.
- ② **Heap-order Property:** For any node i , **key** (priority) of parent of i is larger than or equal to key of i .

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Lemma: Height of a heap with n nodes is $\Theta(\log n)$.

Storing Heaps in Arrays

Let H be a heap (binary tree) of n items and let A be an array of size n . Store root in $A[0]$ and continue with elements *level-by-level* from top to bottom, in each level left-to-right.

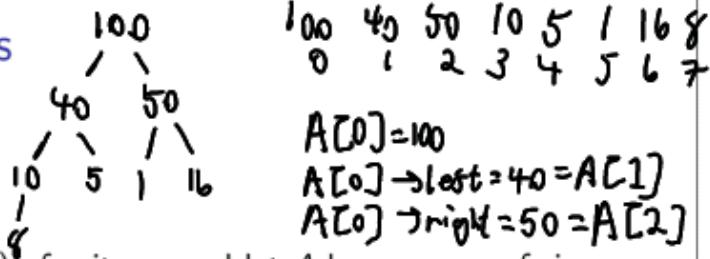
Assume height of a heap is h
 \Rightarrow It has $(h-1)$ full levels

of nodes
in the heap $\rightarrow n > 1 + 2 + \dots + 2^{h-1} = 2^h - 1$

$$n > 2^h - 1$$
$$n+1 > 2^h$$
$$\log(n+1) > h$$
$$h \in O(\log n)$$

$$\boxed{1 + 2 + 4 + 8 = 2^4 - 1 = 15}$$

Storing Heaps in Arrays



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Store root in $A[0]$ and continue with elements **level-by-level** from top to bottom, in each level left-to-right.

It is easy to find parents and children using this array representation:

- the **left child** of $A[i]$ (if it exists) is $A[2i + 1]$,
- the **right child** of $A[i]$ (if it exists) is $A[2i + 2]$,
- the **parent** of $A[i]$ ($i \neq 0$) is $A[\lfloor \frac{i-1}{2} \rfloor]$ ($A[0]$ is the root node).

$$40 = A[1]$$

$$A[1] \rightarrow \text{left} = 10 = A[3]$$

$$A[1] \rightarrow \text{right} = 5 = A[4]$$

$$\left\{ \begin{array}{l} A[i] \rightarrow \text{left} = A[2i+1] \\ A[i] \rightarrow \text{right} = A[2i+2] \end{array} \right.$$

Insertion in Heaps

- Place the new key at the first free leaf
- The heap-order property might be violated: perform a **bubble-up**:

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bubble-up(v)

v : a node of the heap

1. while $\text{parent}(v)$ exists and $\text{key}(\text{parent}(v)) < \text{key}(v)$ do
2. swap v and $\text{parent}(v)$
3. $v \leftarrow \text{parent}(v)$

The new item bubbles up until it reaches its correct place in the heap.

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Time: $O(\text{height of heap}) = O(\log n)$.

deleteMax in Heaps

- The maximum item of a heap is just the root node.
- We replace root by the last leaf (last leaf is taken out).
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```
bubble-down(v)
v: a node of the heap
1.   while v is not a leaf do
2.       u  $\leftarrow$  child of v with largest key
3.       if key(u) > key(v) then
4.           swap v and u
5.           v  $\leftarrow$  u
6.       else
7.           break
```

Time: $O(\text{height of heap}) = O(\log n)$.

Priority Queue Realization Using Heaps

heapInsert(A, x)

A : an array-based heap, x : a new item

1. $\text{size}(A) \leftarrow \text{size}(A) + 1$
2. $A[\text{size}(A) - 1] \leftarrow x$
3. $\text{bubble-up}(A, \text{size}(A) - 1)$

heapDeleteMax(A)

A : an array-based heap

1. $\text{max} \leftarrow A[0]$
2. $\text{swap}(A[0], A[\text{size}(A) - 1])$
3. $\text{size}(A) \leftarrow \text{size}(A) - 1$
4. $\text{bubble-down}(A, 0)$
5. **return** max

Insert and deleteMax: $O(\log n)$

Building Heaps

Problem statement: Given n items (in $A[0 \dots n - 1]$) build a heap containing all of them.

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Solution 1: Start with an empty heap and insert items one at a time:

```
heapify1(A)
A: an array
1. initialize H as an empty heap
2. for i ← 0 to size(A) – 1 do
   3.     heapInsert(H, A[i])
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This corresponds to going from $0 \dots n - 1$ in A and doing *bubble-ups*.
Worst-case running time: $\Theta(n \log n)$.

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heapify( $A$ )
 $A$ : an array
1.    $n \leftarrow \text{size}(A) - 1$ 
2.   for  $i \leftarrow \lfloor n/2 \rfloor$  downto 0 do
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A careful analysis yields a worst-case complexity of $\Theta(n)$.
A heap can be built in linear time.

HeapSort

```
HeapSort( $A$ )
1.   initialize  $H$  to an empty heap
2.   for  $i \leftarrow 0$  to  $n - 1$  do
3.       heapInsert( $H, A[i]$ )
4.   for  $i \leftarrow 0$  to  $n - 1$  do
5.        $A[n - 1 - i] \leftarrow \text{heapDeleteMax}(H)$ 
```

HeapSort

HeapSort(A)

1. initialize H to an empty heap
2. **for** $i \leftarrow 0$ **to** $n - 1$ **do**
3. *heapInsert*($H, A[i]$)
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Running time of HeapSort: $O(n \log n)$

Selection

Problem Statement: The k th-max problem asks to find the *k th largest item* in an array A of n numbers.

Solution 1: Make k passes through the array, deleting the maximum number each time.

Complexity: $\Theta(kn)$.

Solution 2: First sort the numbers. Then return the k th largest number.

Complexity: $\Theta(n \log n)$.

Solution 3: Scan the array and maintain the k largest numbers seen so far in a min-heap

Complexity: $\Theta(n \log k)$.

Solution 4: Make a max-heap by calling *heapify(A)*. Call *deleteMax(A)* k times.

Complexity: $\Theta(n + k \log n)$.

Sol 3.

↪ Form a min heap of size k (first k items) $\rightarrow \Theta(k)$

↪ For any a among $(n-k)$ remaining item :

$\Theta((n-k)\log k)$ ↪ $\text{heap-insert}(a) \rightarrow \log(k) \rightarrow \Theta(\log k)$

$\text{heap-extract-min} \rightarrow \log(k) \rightarrow \Theta(\log k)$

↪ Final heap includes k largest items
root of that heap (min among k largest)

$\therefore \Theta(n \log k)$