CS 240 – Data Structures and Data Management

Module 2: Priority Queues

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References: Sedgewick 9.1-9.4

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Outline

1. Priority Queues
   - Abstract Data Types
   - ADT Priority Queue
   - Binary Heaps
   - Operations in Binary Heaps
   - *PQ-sort* and *Heapsort*
   - Summary

Outline

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**)

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.
Items enter the stack at the *top* and are removed from the *top*.
We can have extra operations: *size*, *isEmpty*, and *top*

Applications: Addresses of recently visited web sites, 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*

Applications: Waiting lines, printer queues

Realizations of Queue ADT

- using (circular) arrays
- using linked lists

Outline

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* or *getmax*.
The priority is also called *key*.

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

Applications: typical "todo" list, simulation systems, sorting

Using a Priority Queue to Sort

PQ-Sort($A[0..n-1]$)
1. initialize PQ to an empty priority queue
2. **for** $k \leftarrow 0$ **to** $n-1$ **do**
3. $\quad PQ.\mathit{insert}(A[k], A[k])$ (priority and item are equal to $A[k]$)
4. **for** $k \leftarrow n-1$ **down to** 0 **do**
5. $\quad A[k] \leftarrow PQ.\mathit{deleteMax}()$

- run-time $O(\sum_{0 \le i < n} \mathit{insert}(i) + \sum_{0 \le i < n} \mathit{deleteMax}(i))$
- depends on how we implement the priority queue

Realizations of Priority Queues

Realization 1: unsorted arrays

- *insert*: $O(1)$
- *deleteMax*: $O(n)$

Note: We assume **dynamic arrays**, i. e., expand by doubling as needed. (Amortized over all insertions this takes $O(1)$ extra time.)

Using unsorted linked lists is identical.

PQ-sort with this realization yields *selection sort*, so runtime is

$$O(\sum_{i<n} i) = O(n^2)$$

Realizations of Priority Queues

Realization 2: sorted arrays

- *insert*: $O(n)$
- *deleteMax*: $O(1)$

Using sorted linked lists is identical.

PQ-sort with this realization yields *insertion sort*, runtime is

$$O(\sum_{i<n} i) = O(n^2)$$

Outline

Realization 3: Heaps

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

You should know:

- 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.
- Any binary tree with n nodes has height at least $\log(n+1) - 1 \in \Omega(\log n)$.

Heaps – Definition

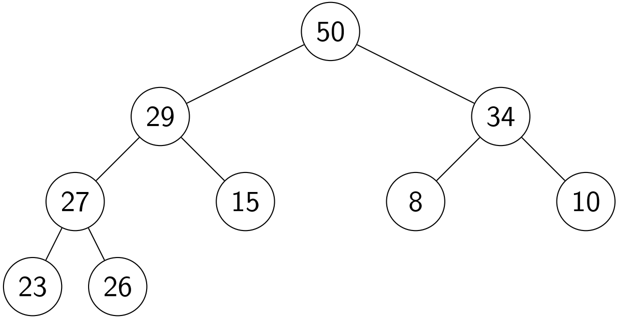
A **heap** is a binary tree with the following two properties:

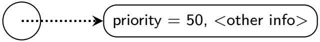
1. **Structural Property:** All the levels of a heap are completely filled, except (possibly) for the last level. The filled items in the last level are *left-justified*.
2. **Heap-order Property:** For any node i, the key of the parent of i is larger than or equal to key of i.

The full name for this is *max-oriented binary heap*.

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

Example Heap



(In our examples we only show the priorities, and we show them directly in the node. A more accurate picture would be 

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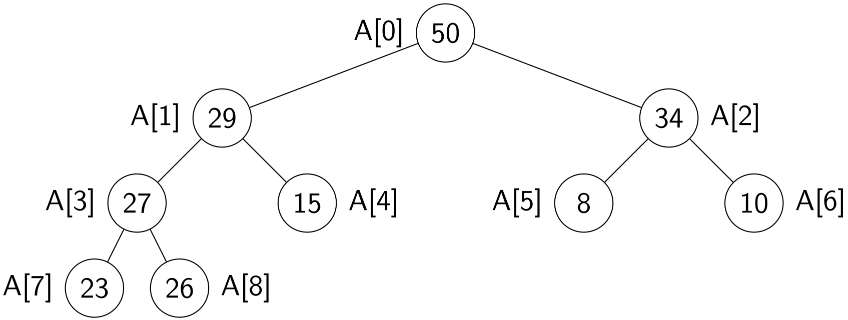
Storing Heaps in Arrays

Heaps should *not* be stored as binary trees!

Let H be a heap 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.



Heaps in Arrays – Navigation

It is easy to navigate the heap using this array representation:

- the *root* node is at index 0
  (We use "node" and "index" interchangeably in this implementation.)
- the *left child* of node i (if it exists) is node $2i + 1$
- the *right child* of node i (if it exists) is node $2i + 2$
- the *parent* of node i (if it exists) is node $\lfloor \frac{i-1}{2} \rfloor$
- the *last* node is $n - 1$

We should hide implementation details using helper-functions!

- functions *root()*, *parent(i)*, *last(n)*, etc.

Outline

Insert in Heaps

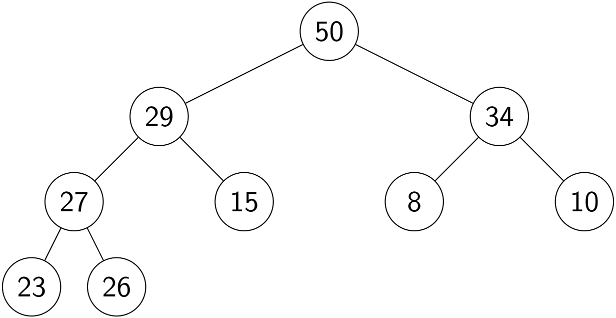
- Place the new key at the first free leaf
- The heap-order property might be violated: perform a *fix-up*:

fix-up(A, k)
k: an index corresponding to a node of the heap
while $parent(k)$ exists **and** $A[parent(k)] < A[k]$ **do**
    swap $A[k]$ and $A[parent(k)]$
    $k \leftarrow parent(k)$

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

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

fix-up example



deleteMax in Heaps

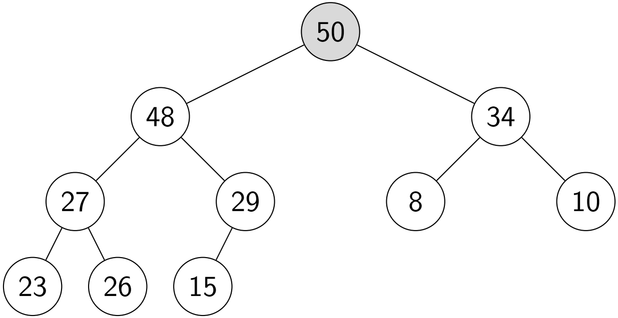
- The maximum item of a heap is just the root node.
- We replace root by the last leaf (last leaf is taken out).
- The heap-order property might be violated: perform a *fix-down*:

fix-down(A, n, k)
A: an array that stores a heap of size n
k: an index corresponding to a node of the heap

while k is not a leaf **do**
    // Find the child with the larger key
    $j \leftarrow$ left child of k
    if (j is not *last*(n) and $A[j+1] > A[j]$)
        $j \leftarrow j+1$
    if $A[k] \geq A[j]$ **break**
    swap $A[j]$ and $A[k]$
    $k \leftarrow j$

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

deleteMax example



Priority Queue Realization Using Heaps

- Store items in array A and globally keep track of *size*

insert(x)
increase *size*
$\ell \leftarrow$ *last*(*size*)
$A[\ell] \leftarrow x$
fix-up(A, ℓ)

deleteMax()
$\ell \leftarrow$ *last*(*size*)
swap A[*root*()] and $A[\ell]$
decrease *size*
fix-down(A, *size*, *root*())
return $A[\ell]$

insert and *deleteMax*: $O(\log n)$

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Sorting using heaps

- Using the binary-heaps implementation of PQs, we obtain:

PQsortWithHeaps(A)	
1.	initialize H to an empty heap
2.	**for** $k \leftarrow 0$ **to** $n-1$ **do**
3.	$\quad H.insert(A[k])$ (we just insert keys, no items)
4.	**for** $k \leftarrow n-1$ **down to** 0 **do**
5.	$\quad A[k] \leftarrow H.deleteMax()$

- Recall: runtime is

$$O\Big(\sum_{0 \le i < n} \mathit{insert}(i) + \sum_{0 \le i < n} \mathit{deleteMax}(i)\Big)$$

- both operations run in $O(\log n)$ time for heaps

$\rightsquigarrow$ *PQ-Ssrt* using heaps takes $O(n \log n)$ time.

- Can improve this with two simple tricks $\rightarrow$ **Heapsort**
  1. Heaps can be built faster if we know all input in advance.
  2. Can use the same array for input and heap. $\rightsquigarrow$ $O(1)$ *auxiliary space!*

Building Heaps with Fix-up

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

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

```
simpleHeapBuilding(A)
A: an array
1.    initialize H as an empty heap
2.    for i ← 0 to size(A) − 1 do
3.        H.insert(A[i])
```

This corresponds to doing *fix-ups*
Worst-case running time: $\Theta(n \log n)$ (we proved $O(\ )$, $\Omega(\ )$ is an exercise)

Building Heaps with Fix-down

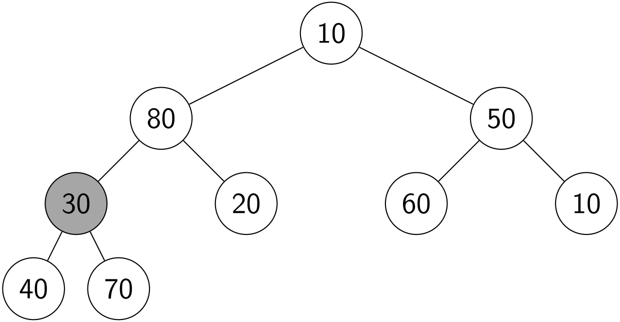
Problem: Given n items all at once (in $A[0 \cdots n-1]$) build a heap containing all of them.

Solution 2: Using *fix-downs* instead:

```
heapify(A)
A: an array
    n ← A.size()
    for i ← parent(last(n)) downto 0 do
        fix-down(A, n, i)
```

A careful analysis yields a worst-case complexity of $\Theta(n)$.
A heap can be built in linear time.

heapify example



HeapSort

- Idea: *PQ-sort* with heaps.
- But: Use same input-array A for storing heap.

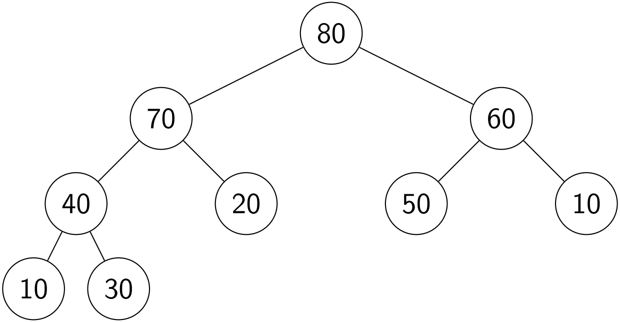
```
HeapSort(A, n)
    // heapify
    n ← A.size()
    for i ← parent(last(n)) downto 0 do
        fix-down(A, n, i)

    // repeatedly find maximum
    while n > 1
        // delete the maximum
        swap items at A[root()] and A[last(n)])
        decrease n
        fix-down(A, n, root())
```

The for-loop takes $\Theta(n)$ time and the while-loop takes $O(n \log n)$ time.

Heapsort example

Continue with the example from heapify:



Outline

Heap summary

- **Binary heap**: A binary tree that satisfies structural property and heap-order property.
- Heaps are one possible realization of ADT PriorityQueue:
  - *insert* takes time $O(\log n)$
  - *deleteMax* takes time $O(\log n)$
  - Also supports *findMax* in time $O(1)$
- A binary heap can be built in linear time.
- *PQ-sort* with binary heaps leads to a sorting algorithm with $O(n \log n)$ worst-case run-time ($\rightsquigarrow$ *HeapSort*)
- We have seen here the *max-oriented version* of heaps (the maximum priority is at the root).
- There exists a symmetric *min-oriented version* that supports *insert* and *deleteMin* with the same run-times.

Finding the smallest item

Problem: Find the *kth smallest item* in an array A of n distinct numbers.

Solution 1: Make k passes through the array, deleting the minimum number each time.
Complexity: $\Theta(kn)$.

Solution 2: Sort A, then return $A[k-1]$.
Complexity: $\Theta(n \log n)$.

Solution 3: Scan the array and maintain the k smallest numbers seen so far in a max-heap
Complexity: $\Theta(n \log k)$.

Solution 4: Create a min-heap with *heapify*(A). Call *deleteMin*(A) k times.
Complexity: $\Theta(n + k \log n)$.