

6 Priority Queues

6.1 Model and simple implementations

Reading: MAW 6.1 and 6.2

- A data structure H that supports $Insert(H, x)$ and $DeleteMin(H)$.
- Implementations:

Implementation	Insert	DeleteMin
array	$O(1)$	$O(n)$
sorted array	$O(n)$	$O(1)$
BST	$O(n)$	$O(n)$
AVL	$O(\log n)$	$O(\log n)$

Question: Can we use an array for data storage to achieve $O(\log n)$ worst-case time for both operations?

6.2 Heap: An implementation of priority queues

Reading: MAW 6.3

- What is a heap?
 - A left-complete binary tree;
 - Each node has a number (key); and
 - Heap-order property (for min heaps): Parent \leq children.

Example: The left tree in Figure 6.5 on page 215.

- How is a heap implemented?

Array: top-down, level-by-level, and left-right.

For example, the heap in our previous example can be represented by array H : 13, 21, 16, 24, 31, 19, 68, 65, 26, 32.

In general, for $H[i]$, its left child is $H[2i]$, its right child is $H[2i + 1]$, and its parent is $H[\lfloor \frac{i}{2} \rfloor]$.

- Height of a heap with n nodes:

Height $h = \lfloor \log n \rfloor + 1 = O(\log n)$.

For example,

n	h
1	$1 = \lfloor \log 1 \rfloor + 1$
2	$2 = \lfloor \log 2 \rfloor + 1$
3	$2 = \lfloor \log 3 \rfloor + 1$
4	$3 = \lfloor \log 4 \rfloor + 1$

- Rebuilding a heap:

Assumptions:

- The left subtree of $H[i]$ is a heap;
- The right subtree of $H[i]$ is a heap; but
- The heap-order property is not satisfied at $H[i]$, i.e., $H[i] > H[2i]$ or $H[i] > H[2i + 1]$.

Question: How can we rebuild the tree rooted at $H[i]$ into a heap?

Example: 10, 5, 6, 7, 8 \rightarrow 5, 7, 6, 10, 8.

```

RebuildHeap(H, i)
  if H[i] is not a leaf
    if H[i] > H[2i] or H[i] > H[2i+1]
      if H[2i] < H[2i+1]
        j = 2i
      else j = 2i+1
      swap H[i] and H[j]
      RebuildHeap(H, j)

```

Worst-case time: $O(\log n)$.

- *Insert(H, x)*:

Example: Insert 14 to 13, 21, 16, 24, 31, 19, 68, 65, 26, 32.

```

Insert(H, x)
  i = size + 1
  H[i] = x
  parent = i / 2
  while parent > 0 and H[i] < H[parent]
    swap H[i] and H[parent]
    i = parent
    parent = i / 2
  size ++

```

Worst-case time: $O(\log n)$.

- *DeleteMin(H)*:

By the heap property of a min heap, the minimum is at the root of the heap, which is $H[1]$.

Example: Delete the minimum from 13, 14, 16, 24, 21, 19, 68, 65, 26, 32, 31.

```

DeleteMin(H)
  min = H[1]
  H[1] = H[size]
  size --
  RebuildHeap(H, 1)
  return min

```

Worst-case time: $O(\log n)$.

- Initialization:

Example: Input array 150, 80, 40, 30, 10, 70, 110, 100, 20, 90, 60, 50, 120, 140, 130.

```

Initialization(H)
  for i = size / 2 to 1
    RebuildHeap(H, i)

```

Worst-case time: $O(n)$. (See page 223 for proof)

Note: Another method for initialization is to insert the keys one by one into an initially empty heap. What is the time complexity?

6.3 Applications

Reading: MAW 6.4

- Operating system: Scheduling jobs on processors to run based on priorities.

When a new job comes, if the processors are all busy, the job is inserted into a waiting queue based on its priority.

When a processor becomes idle, if the waiting queue is not empty, the job with the highest priority is removed from the queue to start execution on the processor.

- The selection problem: Finding the k th largest number among n .

- Algorithm 1: Sort and then select.

Time complexity: $O(n^2)$ or $O(n \log n)$.

- Algorithm 2: Sort $A[1..k]$ into decreasing order. We call the sorted list S . ($S[k]$ is therefore the smallest of the k numbers.) For $i = k + 1..n$, if $A[i] > S[k]$, $S[k]$ is removed and $A[i]$ is inserted into the correct position in S . At the end, return $S[k]$ as the k th largest.

Time complexity: $O(k \log k + (n - k)k) = O(nk)$.

- Algorithm 3: Initialize into a max heap in $O(n)$ time and then perform *DeleteMax* k times.

Time complexity: $O(n + k \log n)$.

- Algorithm 4: Similar idea to Algorithm 2, but maintain S as a min heap instead of a sorted list. $A[1..k]$ is first initialized into a heap S . For $i = k + 1..n$, if $A[i] > S[1]$, do *DeleteMin*(H) and *Insert*($S, A[i]$). At the end, return $S[1]$.

Time complexity: $O(k + (n - k) \log k) = O(n \log k)$.