

ESO211 (Data Structures and Algorithms)

Lecture Notes Set 4

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1 Heap Data Structure

A max-heap is a combination of two structures: (i) It is an almost-complete binary tree and (ii) also a sequence/list such that any node precedes its child-nodes in the sequence. It stores a completely ordered set such that for any node j , $key(parent(j)) \geq key(j)$ for all j . One can similarly define a *min-heap* where the last relation is reversed, i.e., $key(parent(j)) \leq key(j)$ for all j . In this discussion we will only discuss max-heap but the entire discussion also applies to min-heaps.

Observation 1 *The root of a max-heap is a highest key node.*

Usually it is very convenient to implement a heap on an array because the second property is easily achieved. To define an almost complete binary tree structure on an array we use the following relations on the indices: $parent(i) = \lfloor (i-1)/2 \rfloor$, $leftChild(i) = 2i+1$, and $rightChild(i) = 2i+2$. The index of the root is 0.

A few points about the binary tree, especially almost complete binary trees, are in order. The depth of a node is the distance of the node from the root (number of edges on the path). At most 2^i nodes can exist in a binary tree at depth i . An almost complete binary tree of depth d contains 2^i nodes at depth i for all $i < d$ and contains at least 1 vertex in the depth d . Hence such a tree contains at least 2^d vertices and at most $2^{d+1} - 1$ vertices.

One very useful operation on a heap is when the sub-tree rooted at a node i is not a valid heap but those rooted at $leftChild(i)$ and $rightChild(i)$ are valid heaps. So we have, $key(i) < key(leftChild(i))$ or $key(i) < key(rightChild(i))$.

Algorithm 1 shows how to rearrange the elements of the sub-tree rooted at i such that finally the subtree rooted at i also becomes a heap. It may be noted that the sub-tree at $parent(i)$ may not be valid heap. Here A is the array in which the heap is implemented and $HeapSize$ denotes the number of elements in the heap. Observe that the process ripples down to a leaf node.

Verify that the resulting sub-tree rooted at vertex i is a valid heap.

The time complexity of the procedure is $O(depth - depth(i))$ where $depth(x)$ denotes the depth of the node x , the number of edges on the path from the root

```

large := i;
if leftChild(i) ≤ HeapSize − 1 AND A[leftChild(i)] > A[i] then
    | large := leftChild(i);
end
if rightChild(i) ≤ HeapSize − 1 AND A[rightChild(i)] > A[large] then
    | large := rightChild(i);
end
if large ≠ i then
    | Swap(A[large], A[i]);
    | FixDown(A, large, HeapSize);
end

```

Algorithm 1: *FixDown*(*A*, *i*, *HeapSize*)

to *i* in the tree. *depth* is the depth of the tree, i.e., depth of the deepest node in the tree.

To extract the largest element from a max-heap, you need to output the key of the root and then re-fix the heap. This is done by removing the root (top) key, copy the value of *A*[*HeapSize*] to the root, and then perform *FixDown* at the root. See algorithm 2.

```

Output A[0];
A[0] := A[HeapSize − 1];
HeapSize := HeapSize − 1;
FixDown(A, 0, HeapSize);

```

Algorithm 2: *DeleteMax*(*A*, *HeapSize*)

To insert a new element in a heap we store the new element at *A*[*HeapSize*] and then allow the new value to bubble up to its valid position. So we donot use *FixDown* here. See Algorithm 3.

```

A[HeapSize] := x;
HeapSize := HeapSize + 1;
i := HeapSize − 1;
while i > 0 AND key(i) > key(parent(i)) do
    | swap(A[i], A[parent(i)]);
    | i := parent(i);
end

```

Algorithm 3: *Insert*(*A*, *HeapSize*, *x*)

Prove that Algorithm 3 results in a valid heap.

Finally let us make a heap from a set of *m* elements given in an array *A* stored in range 0 : *m* − 1. We will perform this task iteratively in bottom-up order. Note that the elements in the range *m* : *parent*(*m*) + 1 are leaf nodes

hence the sub-trees rooted at these vertices are single nodes and hence these are valid heaps. Starting at $\text{parent}(m)$ down to 0 we will fix the heap using *FixHeap*. See Algorithm 4.

```

for  $i := \text{parent}(m - 1)$  Down to 0 do
  |  $\text{FixDown}(A, i, m)$ ;
end

```

Algorithm 4: BuildHeap(A, m)

Suppose the depth of the tree is d . Also suppose there are r nodes at depth d . So $n = (2^d - 1) + r$ where $1 \leq r \leq 2^d$. The cost of FixDown from a node at depth i is at most $(d - i)$. So the cost of FixDown for the nodes upto depth $d - 2$ is $\sum_{i=0}^{d-2} (d - i) \cdot 2^i$. The cost of $\lceil r/2 \rceil$ nodes at level $d - 1$ (which have children) is $\lceil r/2 \rceil$. So the total cost is

$$\begin{aligned}
\text{Cost} &\leq c \left(\sum_{i=0}^{d-2} (d - i) \cdot 2^i + \lceil r/2 \rceil \right) \\
&= c \left(d \cdot \sum_{i=0}^{d-2} 2^i - \sum_{i=1}^{d-2} i \cdot 2^i + \lceil r/2 \rceil \right) \\
&= c \left(d(2^{d-1} - 1) + \lceil r/2 \rceil - 2d \left(\sum_{i=1}^{d-2} x^i \right) / dx|_{x=2} \right) \\
&= c \left(d(2^{d-1} - 1) + \lceil r/2 \rceil - 2d((x^{d-1} - x)/(x - 1)) / dx|_{x=2} \right) \\
&= c \left(d(2^{d-1} - 1) + \lceil r/2 \rceil - 2((d - 1)2^{d-2} - 1) + 2(2^{d-1} - 2) \right) \\
&= c(3 \cdot 2^{d-1} - 2 - d + \lceil r/2 \rceil) \\
&\leq c \cdot 3n/2
\end{aligned}$$

So the time complexity of *BuildHeap* is $O(n)$.

Exercise:

1. Given a max-Heap in which the key of each node its priority. Does it act a priority queue, i.e., if there are more than one nodes with the same priority, then does *DeleteMax* otuput the oldest of these highest priority node value? Justify your answer.

Explain a suitable modification so the max-Heap acts like priority queue.