

1.**Solution:**

For the delete operation, we need to define some helper methods and follow the below steps:

1. For element at $A[i]$, change its priority to some large value (larger than the current maximum) :takes $O(1)$ time
2. Then shift that $A[i]$ up, to maintain the heap property: takes $O(\log n)$ time
3. Now extract it and remove it using `getMax()`: takes $O(1)$ time
4. Finally, heapify the structure again: takes $O(\log n)$ time

So in total time complexity is $O(1) + O(\log n) + O(\log n) + O(1)$, which is equivalent to $O(\log n)$.

2. Solution:**Pseudocode:**

Let a be the array in which elements of the max heap are stored.

The number of nodes in a binary tree of height $h = 2^{h+1} - 1$

We know that in case of the max heap the k largest elements exist between the root and the maximum height $k-1$ in the binary max heap.

Number of nodes in the binary tree of height $k-1 = 2^{k-1+1} - 1 = 2^k - 1$

So the k largest elements exist among the front $2^k - 1$ elements in the array.

So we can search the array from index 0 to index $2^k - 2$.

The required algorithm is shown below:

`get_k_Biggest_Elements(Array_a)`

integer $m = 2^k - 1$

Create a new array b of size m .

The elements from index 0 to index $2^k - 2$ in array a are copied to array b .

`BUILD-MAX-HEAP(b)`

for $i := 1$ to k

`Heap-Extract-Max(b)`

`MAX-HEAPIFY` //This step results in $O(\log k)$ complexity

end for

Time Complexity Analysis:

Construct a new max-heap using the front $2^k - 1$ elements from the given array (which is a max heap). Note the height of the new max heap is k . The maximum element is present in the root node. We extract the maximum element from the heap and convert it back to the max heap. This step has a complexity $O(\log k)$. Since the previous step is performed k times, hence total time complexity $= k * O(\log k) = O(k \log k)$.

3. Solution:

Denote for a node j having two children its predecessor by p and its successor by s . First, we show by contradiction that the successor of j doesn't have left child. If s has a left child, then the key of s is greater than the key of $\text{left}[s]$. Also the key of s is also

larger than that of j , and since s has one left child, the key of $\text{left}[s]$ is larger than that of j . Thus $\text{key}[s] \geq \text{key}[\text{left}[s]] \geq \text{key}[j]$, which is a contradiction since s is the successor of j . So that successor of j has no left child.

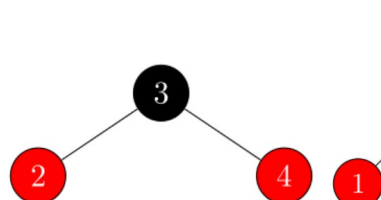
Also we can show by contradiction that the predecessor of j doesn't have right child. Suppose p has a right child. Then we can see that the key of p is less than that of the right child. Also the key of p is less than that of j , and since p has a right child, so the key of $\text{right}[p]$ is less than that of j . Thus $\text{key}[p] \leq \text{key}[\text{right}[p]] \leq \text{key}[j]$, which is a contradiction, since p is the predecessor of j . Hence the predecessor of j has no right child.

4. Solution:

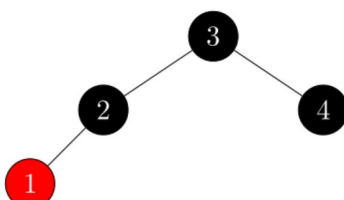
In a red-black tree, the resulting tree after performing insertion and deletion operations may not be the same as the initial red-black tree. Because after insertion or deletion operations, the resulting tree may violate the property of the red-black tree.

For example:

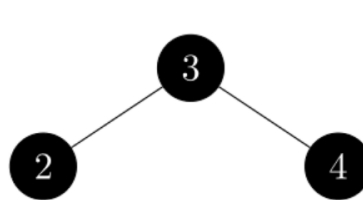
Initial:



insert 1:



delete 1:



5. Solution:

Pseudocode:

initialize $i=1$ and $j=1$ and a sequence C

while $i \leq n$ and $j \leq n$

if $A[i] = A[j]$, do $i++$ and $j++$

else if $A[i] > A[j]$, append $A[j]$ to C and do $j++$

else if $A[j] > A[i]$, append $A[i]$ to C and do $i++$.

if $i == n$ (B is remaining with some elements)

append rest elements of B in C

else if $j == n$ (A is remaining with some elements)

append rest elements of A in C

C will contain the difference of sequence A and B .

Time Complexity :

$O(n)$