

## Question 1

(a) Consider the following algorithm.

---

**Algorithm 1:** Is  $k$ -th smallest key smaller than  $x$

---

**Input:** A min-heap with  $n$  distinct keys in an array  $A[1, \dots, n]$ ,  
a value  $x$  and an integer  $k$  with  $1 \leq k \leq n$ .

**Output:** Yes if the  $k$ -th smallest key in the heap is smaller than  $x$ ; No otherwise.

*/\* Here's my algorithm.*

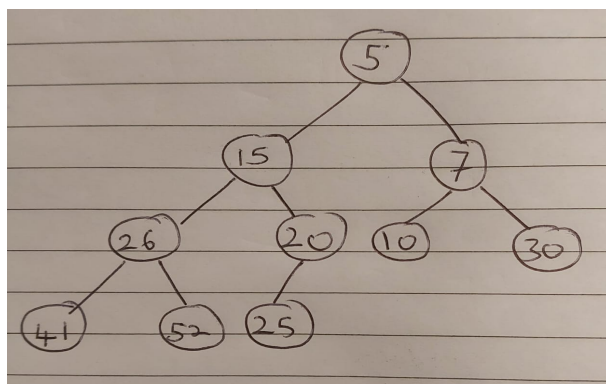
*node: represents current node (root at start)*

*count: counter to keep track (0 at start) \*/*

```

1 checkIfSmaller(node, count, k, x)
2 {
3   flag ← false
4   if node.value ≠ NULL and node.value < x then
5     count++
6     if count ≥ k then
7       flag = true
8       return flag
9   if !flag then
10    checkIfSmaller(node.left, count, k, x)
11    checkIfSmaller(node.right, count, k, x)
12 }
13
14 isSmaller(node, count, k, x) {
15   result = checkIfSmaller(node, count, k, x)
16   if result = true then
17     print Yes
18   else print No
19 }
```

---



Explain the algorithm.

The basic logic as mentioned in the remark is that we just need to know if the node value is lower than  $x$  and don't actually need the value.

This is kind of a depth first search approach. We start at the root, check if the node value

is lesser than  $x$ . If Yes, then we increase the count and move to the child nodes recursively until we have the count  $\geq k$  (indicating that, there are already  $k$  nodes smaller than  $x$ ) or there are no more nodes left to check.

Suppose, Case 1:  $k = 6, x = 30$

We check 5, and since  $5 < 30$  count = 1. We then follow left first traverse acc. to algo.

$15 < 30$ : yes,  $c = 2$

$26 < 30$ : yes,  $c = 3$

$41 < 30$ : no, since it is more than 30, we don't traverse further

$52 < 30$ : no, same as above

$20 < 30$ : yes,  $c = 4$

$25 < 30$ :  $c = 5$

$7 < 30$ :  $c = 6$

Now we have established that there are atleast 6 values less than  $x$ , hence we can say with confidence that  $k$ -th smallest key in the heap is smaller than  $x$ .

Suppose, Case 2:  $k = 6, x = 20$

$5 < 20$ :  $c = 1$

$15 < 20$ :  $c = 2$

$26 < 20$ : no, Don't process further

$20 < 20$ : no, Don't process further

$7 < 20$ :  $c = 3$

$10 < 20$ :  $c = 4$

$30 < 20$ : no, Don't process further

End of Nodes

Hence, by processing just 4 nodes we determined that the answer is no.

(b) Running time analysis.

Since we only Process the nodes with values less than  $x$ , We will require at most  $O(k)$  time (if answer is yes) and even lesser time (if answer is No).

Example: In case 1 above,  $k = 6$  and we processed 6 nodes to determine the answer. In case 2,  $k = 6$  and we processed 4 nodes to determine the answer.

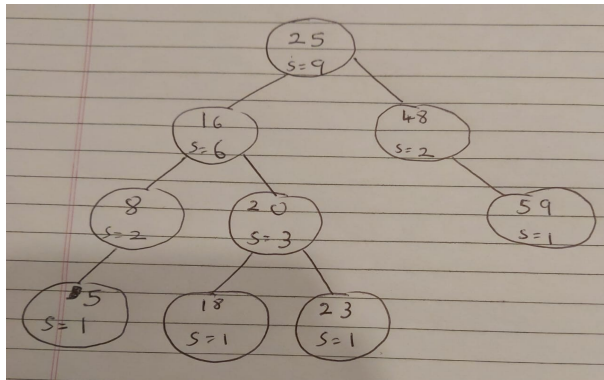
The worst case time complexity will be  $O(k)$ .

## Question 2

- (a) We will augment the tree such that it becomes an AVL tree with each node storing a metadata that contains the size of the node ( $v.size$ ) where  $size = \text{number of childs under the node including the node}$ .

$$v.size = 1 + v.left.size + v.right.size$$

Refer diagram for the augmented tree.



- (b) Consider the following algorithm.

---

**Algorithm 2:** Determine the range( $x$ ) in a BST.

---

**Input:** an augmented balanced binary search tree (AVL)  $T$  of  $n$  nodes (each node  $v$  has  $v.left$ ,  $v.right$ , and  $v.key$  and stores the size metadata), a value  $x$  (to determine rank), rank initialized as 1.

**Output:** Rank of  $x$ . (How does  $X$  rank in an ordered list)

*/\* Here's my algorithm.*

*Rank is initialized as 1 \*/*

---

```

GetRank ( T, x, rank ) {
  if ( T.v = NULL ) then
    return rank
  end

  if ( T.v.value <= x ) then
    if ( T.v.value == x ) then
      rank--
    end
    if ( T.v.left = NULL ) then
      rank <- GetRank ( T.v.right, x, rank ) + 1
    else
      rank <- GetRank ( T.v.right, x, rank ) + T.v.left.size + 1
    end
  else
    rank <- GetRank ( T.v.left, x, rank )
  end
end
}
  
```

Explanation with Example: case 1: Rank (21)

Current Rank = 1 (initialized value)

25  $\leq$  21 : No, Go to left subtree

16  $\leq$  21: Yes, and since there is a left subtree, rank = current rank + 2 (left tree size) + 1 = 4 (goto right subtree).

20  $\leq$  21: Yes, and since there is a left subtree, rank = curr. rank + 1 (left tree size) + 1 = 6 (goto right subtree).

23  $\leq$  21: No, Go to left subtree. Since left subtree is empty. return rank = 6.

case 2: Rank (48)

Current Rank = 1

25  $\leq$  48 : Yes, and since there is a left subtree, rank = current rank (1) + 6 (left tree size) + 1 = 8 (goto right subtree).

48  $\leq$  48: Yes, and since node value = x, we decrement rank once as the node is present in the tree and we don't have to count it twice current rank = 8 - 1 = 7. There is no left subtree, rank = current rank + 1 = 8 (goto right subtree).

59  $\leq$  48: No, Go to left subtree. Since left subtree is empty. return rank = 8.

Running time analysis. This method just traverses a single path in the tree to find the rank. Worst case, it traverses the entire height of the tree making its time complexity as  $O(h)$ . Since height of a balanced tree is  $\log(n)$ . The running time complexity of the ALgo is  $\log(n)$ .

- (c) According to the Main Theorem: Let  $A$  be an attribute that augments an AVL tree of  $n$  nodes, and suppose that the value of  $A$  for each node  $v$  only depends on the info in the nodes  $v$ ,  $v.left$  and  $v.right$ . Then we can maintain the values of  $A$  in all nodes of the tree during search, insertion and deletion without affecting the  $O(\log(n))$  performance of the operations. Hence, Here when we augment the tree with the size of its child nodes, performance of the other operations of the tree remains unaffected as the tree balances itself based on the size indicated by the the theorem above.

### Question 3

- (a) Consider the following algorithm.

---

**Algorithm 3:** List all keys in order for the given range  $[Xl, Xr]$

---

**Input:** a balanced binary search tree  $T$  of  $n$  nodes (each node  $v$  has  $v.left$ ,  $v.right$ , and  $v.key$ , a range  $[Xl, Xr]$  such that  $Xl \leq Xr$

**Output:** A list of keys in order that lies within the range

*/\* Here's my algorithm. \*/*

---

```

GetKeysInRange( T, Q, Xl, Xr) {
    if (T.v == NULL)
        return
    end

    if (Xl < T.v.key)
        GetKeysInRange( T.v.left, Q, Xl, Xr)
    end

    if (Xl <= T.v.key and Xr >= T.v.key)
        Q.enqueue(T.v.key)
    end

    if (Xr > T.v.key)
        GetKeysInRange( T.v.right, Q, Xl, Xr)
    end
}

```

Explain the algorithm.

We kind of do an in-order traversal over here. We first check all the nodes in left subtree recursively that fall in the given range. move on to the parent and then the right sub tree similarly. This would give us the list of keys in order that lie in the given range. We only traverse those nodes that lie in the range.

Example: Consider the tree in previous example.

Suppose we need to find all keys between 16-48

$Xl = 16, Xr = 48$ .

From root:

$16 < 25$ : Yes, Goto left subtree

$16 < 16$ : No, Add in queue and go to right subtree(as  $48 > 16$ ).

$16 < 20$ : Yes, Goto left subtree.

$16 < 18$ : Yes, Goto left subtree.

Since left subtree is null, we return to last recursive call. Hence we print 18 (lies in range) and goto right subtree ( $48 > 18$ ).

Since right subtree is empty, goto prev call, Print 20 and goto right subtree.

$16 < 23$ : Yes, Goto left subtree.

Since left subtree is null, we return to last recursive call. Hence we print 23 and goto right subtree ( $48 > 23$ ). Since right subtree is empty, goto prev call, Print 25 and goto right sub-

tree.

$16 < 48$ : Yes, Goto left subtree.

Since left subtree is null, we return to last recursive call. Hence we print 48 and won't goto right subtree as  $(48 > 48)$  is not true.

- (b) Running time analysis. Worst time it can take to calculate keys in range is Traverse the height of the tree twice (left and right subtree ends).  $2 * \log(n)$ . and it takes  $O(k)$  to report the values obtained. In total it takes  $O(k + \log(n))$  time to run the algo in worst case complexity.

## Question 4

- (a) We will Augment the tree by adding a metadata 'sum', which basically stores the sum of the keys of all child nodes including the node.  $v.sum = v.key + v.left.sum + v.right.sum$
- (b) Consider the following algorithm.

---

**Algorithm 4:** Calculate sum of all keys that lie in the range in Log n time

---

**Input:** an augmented balanced binary search tree (AVL) T of n nodes (each node v has v.left, v.right, and v.key and stores the sum metadata - v.sum), a value Xl (start range), Xr (end range) such that  $Xl \leq Xr$ .

**Output:** the sum of the keys in T that are in the range [Xl, Xr]

*/\* Here's my algorithm. \*/*

---

```

range-sum(Xl, Xr){
  u = LCA (T, Xl, Xr)
  sum = u.key
  v = u.left;
  while v != NULL{
    if (Xl < v.key)
      sum += v.key + v.right.sum
      v = v.left
    end

    if (Xl > v.key)
      v = v.right
    end
    if (Xl = v.key)
      sum += v.key + v.right.sum
      break
    end
  }
  v = u.right;
  while v != NULL{
    if (Xr > v.key)
      sum += v.key + v.left.sum
      v = v.right
    end

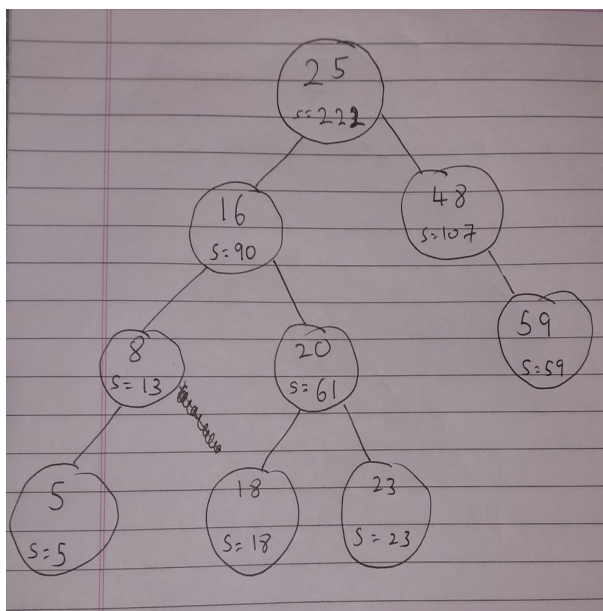
    if (Xr < v.key)
      v = v.left
    end
    if (Xr = v.key)
      sum += v.key + v.left.sum
      break
    end
  }
  return sum;
}

```

```

LCA(T, Xl, Xr){
  v = T.root
  while (v != NULL and v.key not in [ Xl, Xr ] ){
    if (v.key > Xr)
      v = v.left
    else
      v = v.right
    end
  }
  return v
}

```



Example:

Given:  $X_l = 7$ ,  $X_r = 49$

On running LCA we get root node as 25.

sum = 25 (initial for LCA node)

For Left subtree (Start node = 16):

Since  $X_l < 16$ , sum =  $25 + 16 + 61$  (V.right.sum) = 102

Go left

Since  $X_l < 8$ , sum =  $102 + 8 + 0$  (V.right.sum) = 110

Go left

Since  $X_l > 5$ , Go right, and since right is null, we end this.

For Right subtree (Start node = 48):

Since  $X_r > 48$ , sum =  $110 + 48 + 0$  (V.left.sum) = 158

Go right

Since  $X_r < 59$ , Go left, Since node = null, return sum

Final sum = 158

Running time analysis. The worst case time complexity is  $O(\log n)$ . We first find the LCA and check where does the node splits and we begin our operation range-sum. We first traverse to the left, worst case we reach the leaf and add all the right subtree sums of the Left Path. Similarly on right subtree of our LCA node. Hence augmenting the tree allows to get the range-sum in  $O(\log n)$

- (c) According to the Main Theorem: Let  $A$  be an attribute that augments an AVL tree of  $n$  nodes, and suppose that the value of  $A$  for each node  $v$  only depends on the info in the nodes  $v$ ,  $v$ .left and  $v$ .right. Then we can maintain the values of  $A$  in all nodes of the tree during search, insertion and deletion without affecting the  $O(\log(n))$  performance of the operations. Hence, Here when we augment the tree with the sum of child nodes, the performance of the other operations of the tree remains unaffected by balancing due to augmentation indicated by the theorem above.