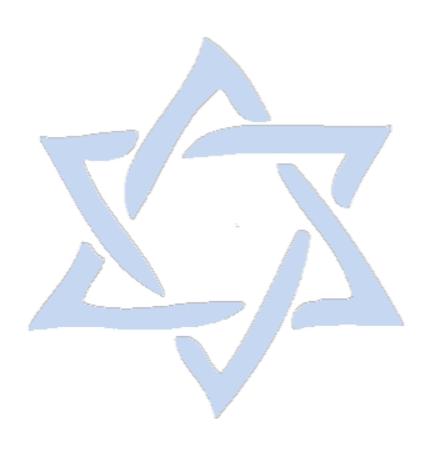
<u>HEAP</u> <u>Lesson-2</u>



Print all nodes less than a value x in a Min Heap

Approach

The idea is to do a preorder traversal of the given Binary heap. While doing preorder traversal, if the value of a node is greater than the given value x, we return to the previous recursive call. Because all children nodes in a min heap are greater than the parent node. Otherwise we print current nodes and recur for its children.

```
void printSmallerThan(int x, int pos)
1.
             /* Make sure item exists */
2.
             if (pos >= heap size)
3.
4.
                 return;
5.
             if (harr[pos] >= x) {
6.
                 /* Skip this node and its descendants,
7.
                  as they are all >= x \cdot */
8.
                 return;
9.
10.
11.
             System.out.print(harr[pos] + " ");
12.
13.
             printSmallerThan(x, left(pos));
14.
             printSmallerThan(x, right(pos));
15.
16.
```

K largest(or smallest) elements in an array | added Min Heap method

Approach

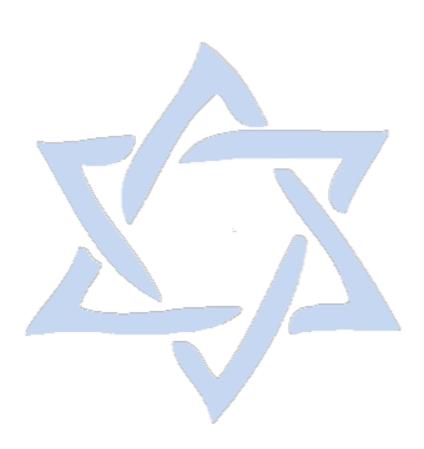
- 1) Build a Min Heap MH of the first k elements (arr[0] to arr[k-1]) of the given array. O(k)
- 2) For each element, after the kth element (arr[k] to arr[n-1]), compare it with the root of MH.
 - (a) If the element is greater than the root then make it root and call heapify for MH
 - (b) Else ignore it.

// The step 2 is O((n-k)*logk)

3) Finally, MH has k largest elements and root of MH is the kth largest element.

```
void FirstKelements(int arr[],int size,int k){
1.
          // Creating Min Heap for given
2.
          // array with only k elements
3.
4.
         MinHeap* m = new MinHeap(k, arr);
5.
          // Loop For each element in array
6.
7.
          // after the kth element
          for (int i = k; i < size; i++) {
8.
9.
              // if current element is smaller
10.
11.
              // than minimum element, do nothing
              // and continue to next element
12.
              if (arr[0] > arr[i])
13.
                  continue;
14.
15.
16.
              // Otherwise Change minimum element to
17.
              // current element, and call heapify to
18.
              // restore the heap property
19.
              else {
                  arr[0] = arr[i];
20.
                  m->heapify(0);
21.
22.
23.
24.
          // Now min heap contains k maximum
          // elements, Iterate and print
25.
          for (int i = 0; i < k; i++) {
26.
              cout << arr[i] << " ";
27.
28.
```

O(k + (n-k)Logk) without sorted output. If sorted output is needed then O(k + (n-k)Logk + kLogk)



Sort a nearly sorted (or K sorted) array

Approach

- 1) Create a Min Heap of size k+1 with first k+1 elements. This will take O(k) time
- 2) One by one remove min element from heap, put it in result array, and add a new element to heap from remaining elements.

```
int sortK(int arr[], int n, int k)
```

```
1.
        // Insert first k+1 items in a priority queue (or min
2.
        // heap)
3.
        //(A O(k) \text{ operation}). We assume, k < n.
4.
        priority queue<int, vector<int>, greater<int> > pq(arr, arr + k + 1);
5.
6.
        // i is index for remaining elements in arr[] and index
7.
        // is target index of for current minimum element in
8.
        // Min Heap 'pq'.
9.
        int index = 0;
10.
        for (int i = k + 1; i < n; i++) {
11.
             arr[index++] = pq.top();
12.
13.
             pq.pop();/
             pq.push(arr[i]);
14.
15.
16.
        while (pq.empty() == false) {
17.
18.
             arr[index++] = pq.top();
19.
             pq.pop();
20.
         }
21.
```

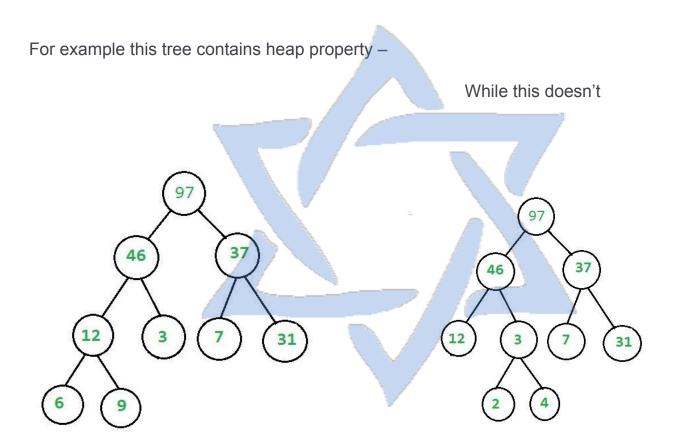
Time Complexity:

Removing an element and adding a new element to min heap will take log k time. So overall complexity will be O(k) + O((n-k) * log(k)).

Check if a given Binary Tree is Heap

Conditions for an Heap:

- 1. It should be a complete tree (i.e. all levels except last should be full).
- 2. Every node's value should be greater than or equal to its child node (considering max-heap).



Approach

- 1. Every Node can have 2 children, 0 child (last level nodes) or 1 child (there can be at most one such node).
- 2. If Node has No child then it's a leaf node and returns true (Base case)
- 3. If Node has one child (it must be left child because it is a complete tree) then we need to compare this node with its single child only.

4. If the Node has both child then check heap property at Node at recur for both subtrees .

```
unsigned int countNodes(struct Node* root)
2.
3.
        if (root == NULL)
4.
            return (0);
5.
       return (1 + countNodes(root->left)
6.
                + countNodes(root->right));
7.
    }
8.
9.
     /* This function checks if the
10.
    binary tree is complete or not */
     bool isCompleteUtil (struct Node* root,
11.
12.
                          unsigned int index,
13.
                          unsigned int number nodes)
14.
15.
        // An empty tree is complete
16.
       if (root == NULL)
17.
            return (true);
18.
19.
       // If index assigned to
20.
       // current node is more than
       // number of nodes in tree,
21.
       // then tree is not complete
22.
23.
        if (index >= number nodes)
            return (false);
24.
25.
       // Recur for left and right subtrees
26.
       return (isCompleteUtil(root->left, 2*index + 1,
27.
28.
                                number nodes) &&
29.
                 isCompleteUtil(root->right, 2*index + 2,
30.
                                number nodes));
31.
    }
32.
33.
     // This Function checks the
34.
    // heap property in the tree.
     bool isHeapUtil(struct Node* root)
35.
36.
        // Base case : single
37.
38.
        // node satisfies property
       if (root->left == NULL && root->right == NULL)
39.
40.
            return (true);
41.
       // node will be in
42.
```

```
43.
        // second last level
44.
        if (root->right == NULL)
45.
        {
46.
            // check heap property at Node
47.
            // No recursive call ,
48.
            // because no need to check last level
49.
            return (root->key >= root->left->key);
50.
        }
51.
        else
52.
        {
53.
            // Check heap property at Node and
            // Recursive check heap
54.
            // property at left and right subtree
55.
56.
            if (root->key >= root->left->key &&
                root->key >= root->right->key)
57.
                return ((isHeapUtil(root->left)) &&
58.
                         (isHeapUtil(root->right)));
59.
60.
            else
61.
                return (false);
62.
     }
63.
64.
     // Function to check binary
     // tree is a Heap or Not.
65.
66.
     bool isHeap(struct Node* root)
67.
68.
        // These two are used
        // in isCompleteUtil()
69.
70.
       unsigned int node count = countNodes(root);
       unsigned int index = 0;
71.
72.
       if (isCompleteUtil(root, index,
73.
                            node count)
74.
            && isHeapUtil(root))
75.
            return true;
76.
        return false;
77.
```

How to check if a given array represents a Binary Heap?

Approach:

A **Simple Solution** is to first check root if it's greater than all of its descendants. Then check for children of the root. Time complexity of this solution is $O(n_2)$

An **Efficient Solution** is to compare root only with its children (not all descendants), if root is greater than its children and the same is true for all nodes, then tree is max-heap (This conclusion is based on transitive property of > operator, i.e., if x > y and y > z, then x > z). The last internal node is present at index (n-2)/2 assuming that indexing begins with 0.

```
1.
    bool isHeap(int arr[], int i, int n)
2.
3.
        // If a leaf node
        if (i >= (n - 2) / 2)
4.
5.
            return true;
6.
7.
        // If an internal node and is
        // greater than its children,
8.
9.
        // and same is recursively
10.
        // true for the children
11.
        if (arr[i] >= arr[2 * i + 1] &&
12.
            arr[i] >= arr[2 * i + 2]
            && isHeap(arr, 2 * i + 1, n)
13.
14.
            && isHeap(arr, 2 * i + 2, n))
15.
            return true;
16.
17.
        return false;
18.
    }
```

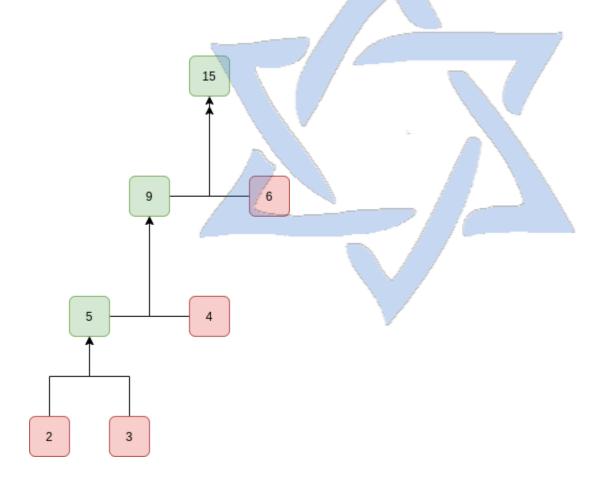
Time Complexity

Time complexity of this solution is O(n).

Connect n ropes with minimum cost

Approach

- 1. Create a min-heap and insert all lengths into the min-heap.
- 2. Do the following while the number of elements in min-heap is not one.
- 1. Extract the minimum and second minimum from min-heap
- 2. Add the above two extracted values and insert the added value to the min-heap.
- 3. Maintain a variable for total cost and keep incrementing it by the sum of extracted values.
- 3. Return the value of this total cost.



```
1.
        int cost = 0; // Initialize result
2.
3.
        // Create a min heap of capacity
4.
        // equal to n and put all ropes in it
5.
        struct MinHeap* minHeap = createAndBuildMinHeap(len, n);
6.
7.
        // Iterate while size of heap doesn't become 1
8.
9.
        while (!isSizeOne(minHeap)) {
            // Extract two minimum length
10.
            // ropes from min heap
11.
            int min = extractMin(minHeap);
12.
            int sec min = extractMin(minHeap);
13.
14.
            cost += (min + sec min); // Update total cost
15.
16.
17.
            // Insert a new rope in min heap
18.
            // with length equal to sum
19.
            // of two extracted minimum lengths
            insertMinHeap(minHeap, min + sec min);
20.
21.
        }
22.
        // Finally return total minimum
23.
        // cost for connecting all ropes
24.
25.
        return cost;
26.
```

Complexity Analysis:

- **Time Complexity:** O(nLogn), assuming that we use a O(nLogn) sorting algorithm. Note that heap operations like insert and extract take O(Logn) time.
- Auxiliary Complexity:O(n), The space required to store the values in min heap

Merge k sorted arrays

Approach:

- 1. Create a min Heap and insert the first element of all k arrays.
- 2. Run a loop until the size of MinHeap is greater than zero.
- 3. Remove the top element of the MinHeap and print the element.
- 4. Now insert the next element from the same array in which the removed element belonged.
- 5. If the array doesn't have any more elements, then replace root with infinite.After replacing the root, heapify the tree.

```
nt *mergeKArrays(int arr[][n], int k)
2.
     {
3.
4.
     // To store output array
5.
        int *output = new int[n*k];
6.
        // Create a min heap with k heap nodes.
7.
        // Every heap node has first element of an array
8.
        MinHeapNode *harr = new MinHeapNode[k];
9.
        for (int i = 0; i < k; i++)
10.
11.
12.
13.
     // Store the first element
14.
            harr[i].element = arr[i][0];
15.
16.
     // index of array
17.
            harr[i].i = i;
18.
19.
     // Index of next element to be stored from the array
            harr[i].j = 1;
20.
21.
22.
23.
    // Create the heap
24.
        MinHeap hp(harr, k);
```

```
25.
26.
        // Now one by one get the minimum element from min
27.
        // heap and replace it with next element of its array
        for (int count = 0; count < n*k; count++)</pre>
28.
29.
30.
             // Get the minimum element and store it in output
            MinHeapNode root = hp.getMin();
31.
            output[count] = root.element;
32.
33.
            // Find the next elelement that will replace current
34.
35.
            // root of heap. The next element belongs to same
36.
            // array as the current root.
37.
            if (root.j < n)
38.
39.
                 root.element = arr[root.i][root.j];
40.
                 root.j += 1;
41.
             // If root was the last element of its array
42.
     // INT MAX is for infinite
43.
     else root.element = INT MAX;
44.
45.
             // Replace root with next element of array
46.
            hp.replaceMin(root);
47.
48.
        }
49.
50.
        return output;
51.
```

Complexity Analysis:

- Time Complexity: O(n * k * log k), Insertion and deletion in a Min Heap requires log k time. So the Overall time complexity is O(n * k * log k)
- **Space Complexity**: O(k), If Output is not stored then the only space required is the Min-Heap of k elements. So space Complexity is O(k).