

## **Experiment 5.1**

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**Branch:** BE-CSE **Section/Group:** IoT\_641(A) **Date of Performance:** 28/2/25

**Subject Name:** Advanced Programming-2 **Subject Code:** 22CSP-351

**1. Aim:** Find the k most frequent elements in an array. Use a hash map to count frequencies, then use a min-heap or bucket sort to extract the top k elements. The heap approach runs in O(n log k) time. Edge cases include k = 1 or all elements being unique.

- **2. Objective:** Implement an algorithm to identify and return the k most frequently occurring elements within a given array.
- 3. Implementation/Code:

```
class Solution {
public:
    vector<int> topKFrequent(vector<int>& nums, int k) {
        unordered_map<int,int> mp;
        int n=nums.size();
        for(int i=0;i<n;i++){
            mp[nums[i]]++;
        }
        vector<pair<int,int>> freq(mp.begin(),mp.end());
        sort(freq.begin(), freq.end(), [](auto &a, auto &b) {
            return a.second > b.second;
        });
        vector<int> ans;
        for(int i=0;i<k;i++){
            ans.push_back(freq[i].first);
        }
        return ans;
    }
};</pre>
```

## 4. Output

```
Accepted Runtime: 0 ms

• Case 1
• Case 2

Input

nums =
[1,1,1,2,2,3]

k =
2

Output
[1,2]
```

### **5.** Learning Outcomes

- Understand the concept of hash maps and min-heaps.
- Learnt how to implement the algorithm in a programming language.
- Learnt how to evaluate and compare the min-heap and bucket sort approaches for top-k extraction.
- Understand the importance of code efficiency, and how to create code that runs within the desired time constraints.

## **Experiment 5.2**

- **1. Aim**: Find the kth largest element in an array using a min-heap or Quickselect. The heap approach runs in  $O(n \log k)$  time, while Quickselect runs in O(n) on average. Edge cases include k = 1 (max element) and k = n (min element).
- **2. Objective**: Implement algorithms to efficiently find the kth largest element within a given array.

## 3. Code:



```
priority_queue<int>max;
int a=1;
for(auto p:nums){
    max.push(p);
}
for(int i=0;i<nums.size();i++){
    if(a<k){
     max.pop();
        a++;
    }
}
return max.top();
}</pre>
```

#### 4. Output



# 5. Learning Outcomes:

- Understand the steps involved in both the min-heap and Quickselect algorithms for finding the kth largest element.
- Learn how to analyze and compare the time complexity of the min-heap (O(n log k)) and Quickselect (O(n) average) approaches.
- Implement both the min-heap and Quickselect algorithms in a programming language.

# **Experiment 5.3**

- **1. Aim**: Find the kth smallest element in a row- and column-sorted matrix. Use a min-heap  $(O(k \log n))$  or binary search on values  $(O(n \log \max \min))$ . Edge cases include k = 1 (smallest element) and  $k = n^2$  (largest element).
- **2. Objective**: Implement an algorithm to find the kth smallest element using binary search on values.

#### 3. Code:

```
class Solution
public:
    int kthSmallest(vector<vector<int>>& matrix, int k)
        int n = matrix.size();
        int le = matrix[0][0], ri = matrix[n - 1][n - 1];
        int mid = 0;
        while (le < ri)
            mid = le + (ri-le)/2;
            int num = 0;
            for (int i = 0; i < n; i++)
                int pos = upper_bound(matrix[i].begin(), matrix[i].end(), mid) -
matrix[i].begin();
                num += pos;
            if (num < k)
                le = mid + 1;
            else
                ri = mid;
        return le;
};
```

## 4. Output

```
Accepted Runtime: 0 ms

• Case 1
• Case 2

Input

matrix =

[[1,5,9],[10,11,13],[12,13,15]]

k =

8

Output

13
```

#### 5. Learning Outcomes:

- Understand the steps involved in both the min-heap and binary search algorithms.
- Learn how to analyze and compare the time complexity.
- Implement appropriate algorithm based on the problem's constraints.