

Top K Frequent Elements

Given an integer array `nums` and an integer `k`, return the `k` most frequent elements.
You may return the answer in any order.

Ex1 :

Input: `nums = [1,1,1,2,2,3]`, `k = 2`
Output: `[1,2]`

Ex2 :

Input: `nums = [1]`, `k = 1`
Output: `[1]`

Constraints:

$1 \leq \text{nums.length} \leq 10^5$

`k` is in the range `[1, the number of unique elements in the array]`.

It is guaranteed that the answer is unique.

Follow up: Your algorithm's time complexity must be better than $O(n \log n)$, where `n` is the array's size.

Ex1 :

Input: nums = [1,1,1,2,2,3], k = 2
Output: [1,2]

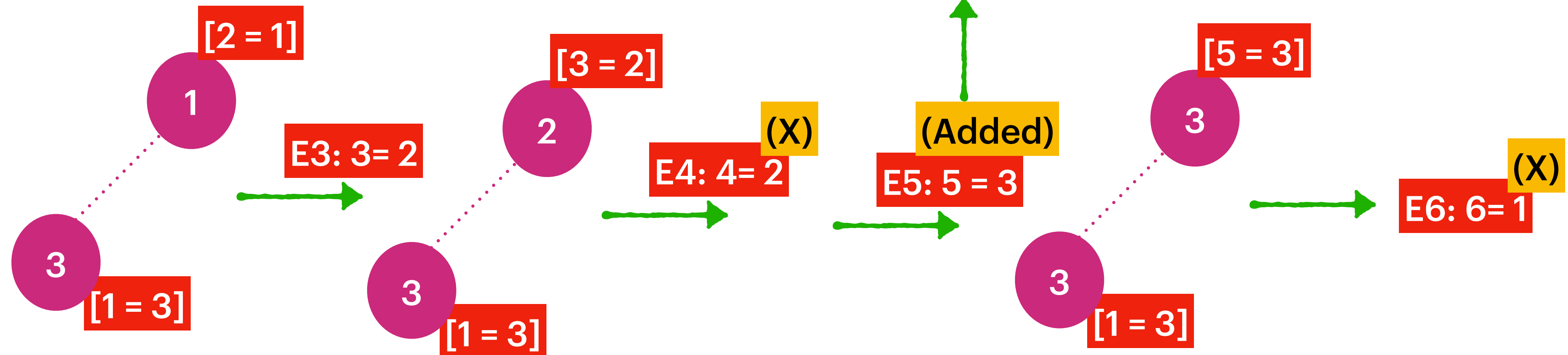
Ex2 :

Input: nums = [1,1,1,2,3,3,4,5,6,4,5,5],
k = 2
Output: [1,5]

[1,1,1,2,3,3,4,5,6,4,5,5], k = 2

counterMap :

E1: 1 = 3
E2: 2 = 1
E3: 3 = 2
E4: 4 = 2
E5: 5 = 3
E6: 6 = 1



Heap has [5=3] , [1=3] these elements return keys = {5,1}

Constructing Map + Add the elements int to minHeap [Sort by Counter]
 $O(n) + O(n \log k) = O(n \log k)$

Kth Largest Element in a Stream

Design a class to find the kth largest element in a stream. Note that it is the kth largest element in the sorted order, not the kth distinct element.

Implement KthLargest class:

KthLargest(int k, int[] nums) Initializes the object with the integer k and the stream of integers nums.
int add(int val) Appends the integer val to the stream and returns the element representing the kth largest element in the stream.

Example 1:

Input

```
["KthLargest", "add", "add", "add", "add", "add"]  
[[3, [4, 5, 8, 2]], [3], [5], [10], [9], [4]]
```

Output

```
[null, 4, 5, 5, 8, 8]
```

Explanation

```
KthLargest kthLargest = new KthLargest(3, [4, 5, 8, 2]);  
kthLargest.add(3); // return 4  
kthLargest.add(5); // return 5  
kthLargest.add(10); // return 5  
kthLargest.add(9); // return 8  
kthLargest.add(4); // return 8
```

Constraints:

$1 \leq k \leq 104$

$0 \leq \text{nums.length} \leq 104$

$-104 \leq \text{nums}[i] \leq 104$

$-104 \leq \text{val} \leq 104$

At most 104 calls will be made to add.

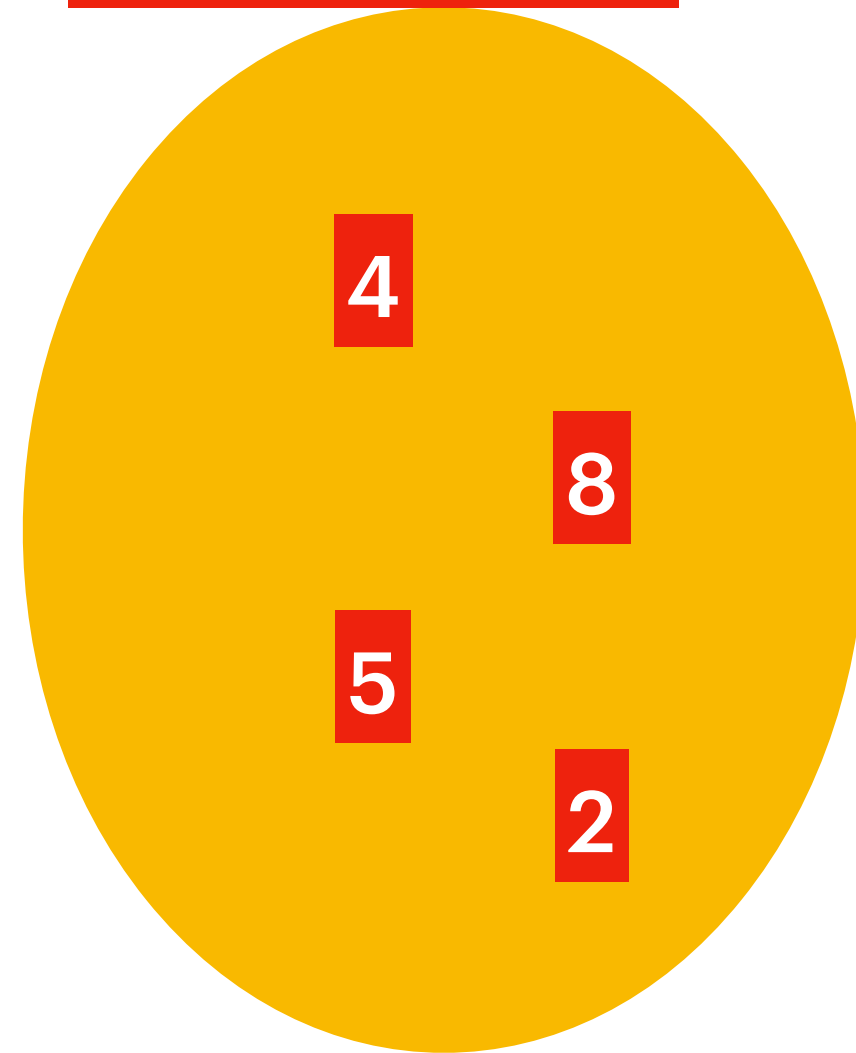
It is guaranteed that there will be at least k elements
in the array
when you search for the kth element.

Kth Largest Element in a Stream

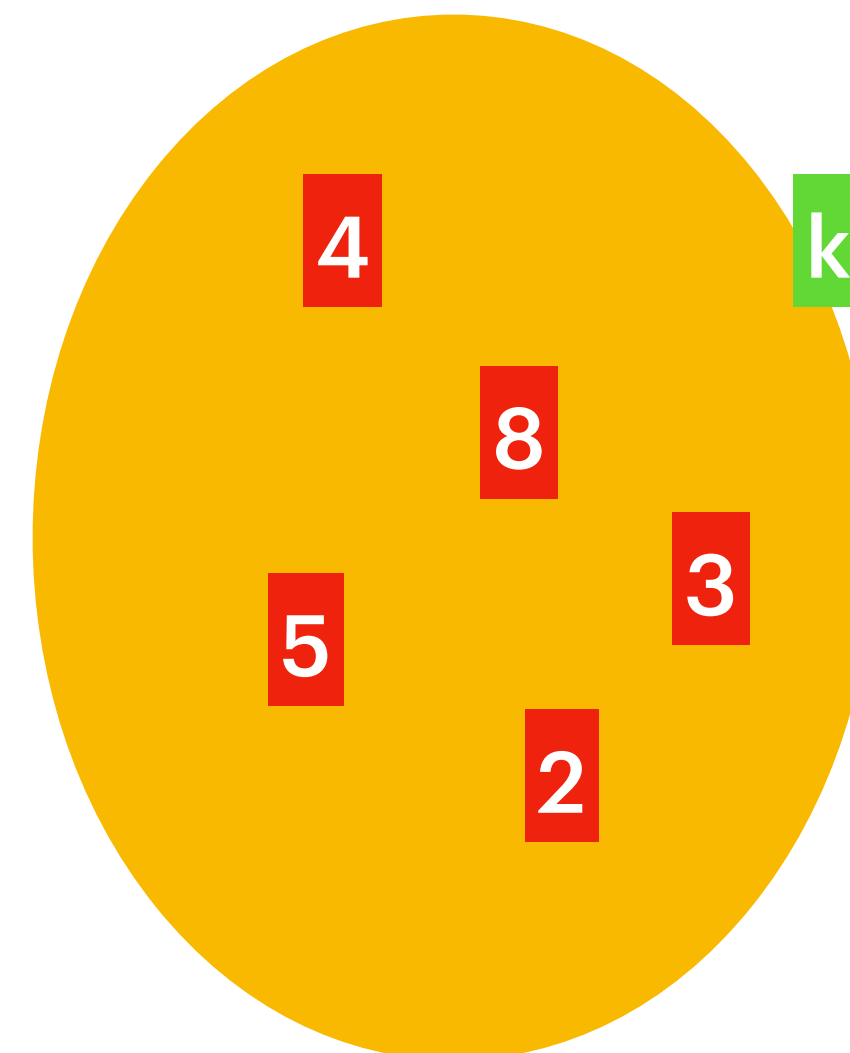
[[k=3, construct[4, 5, 8, 2]], add[3], add[5], add[10], add[9], add[4]]

Output : [null, 4, 5, 5, 8, 8]

Initial Stream :

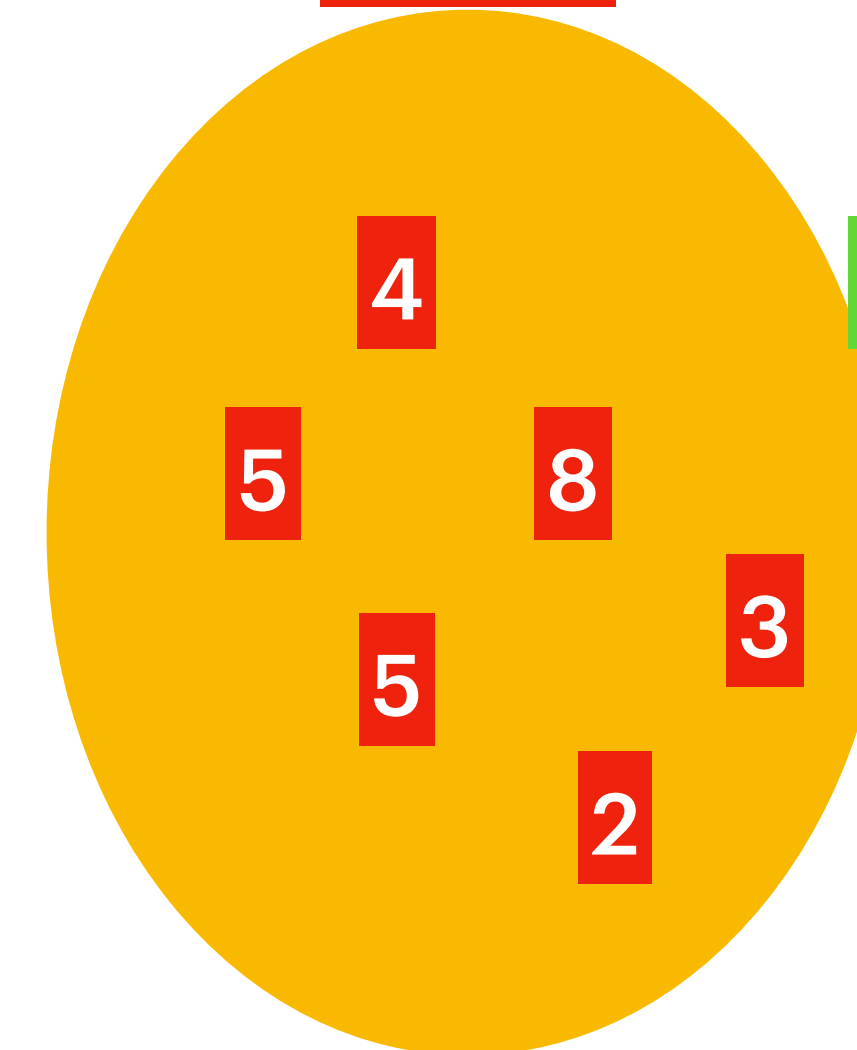


add[3]



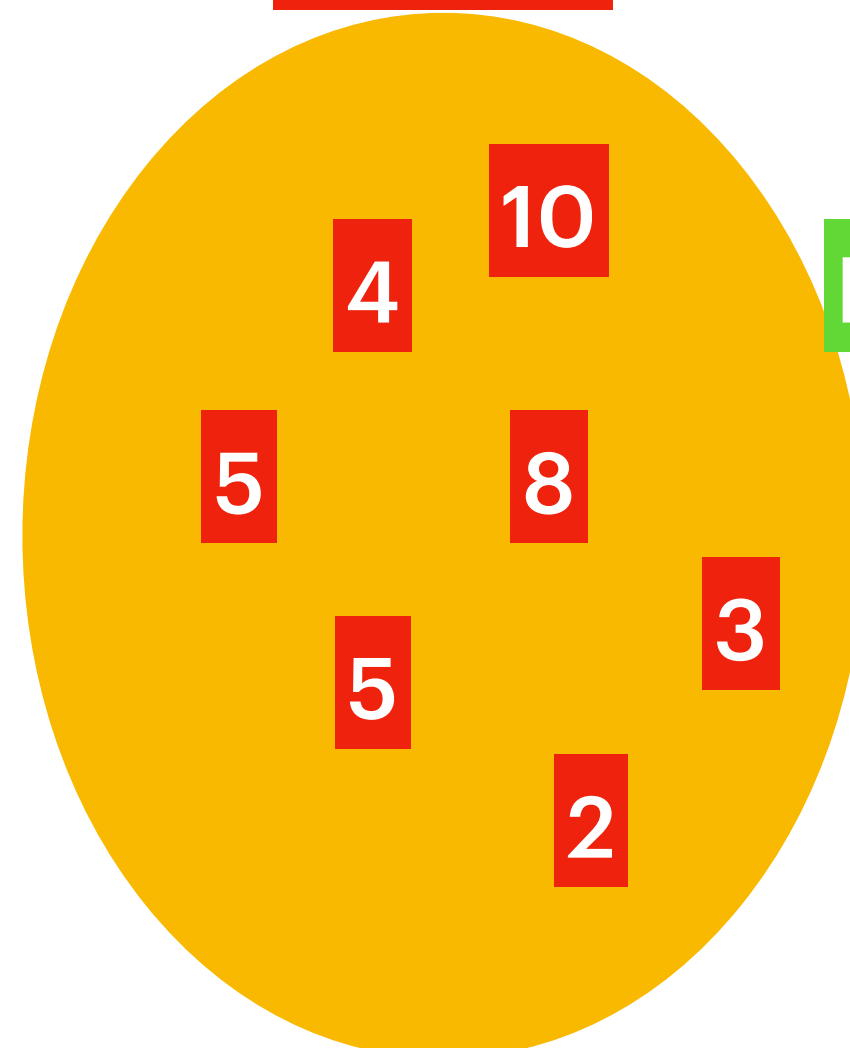
kth Largest 4

add[5]



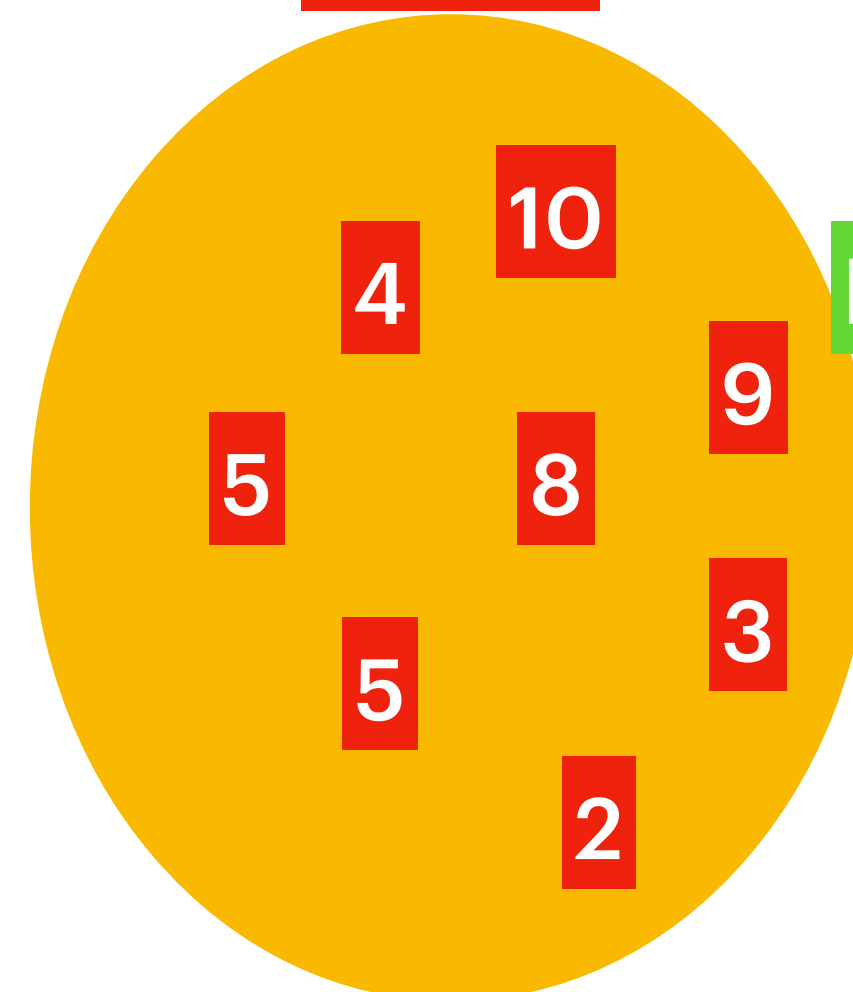
kth Largest 5

add[10]



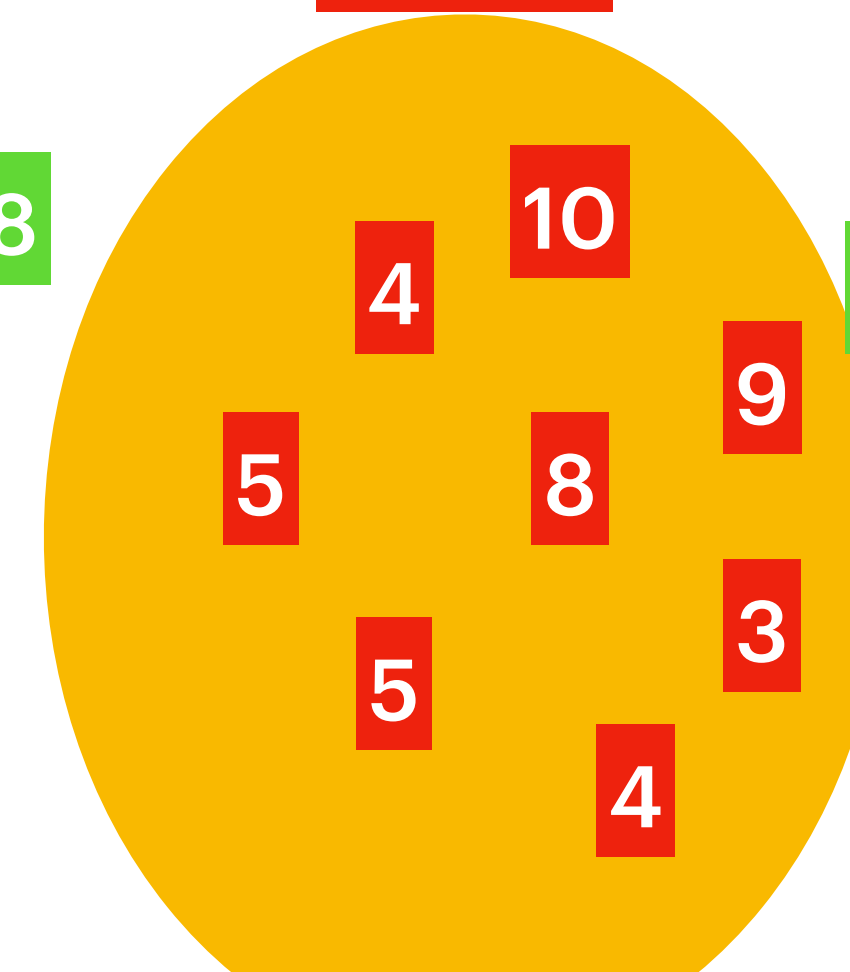
kth Largest 5

add[9]



kth Largest 8

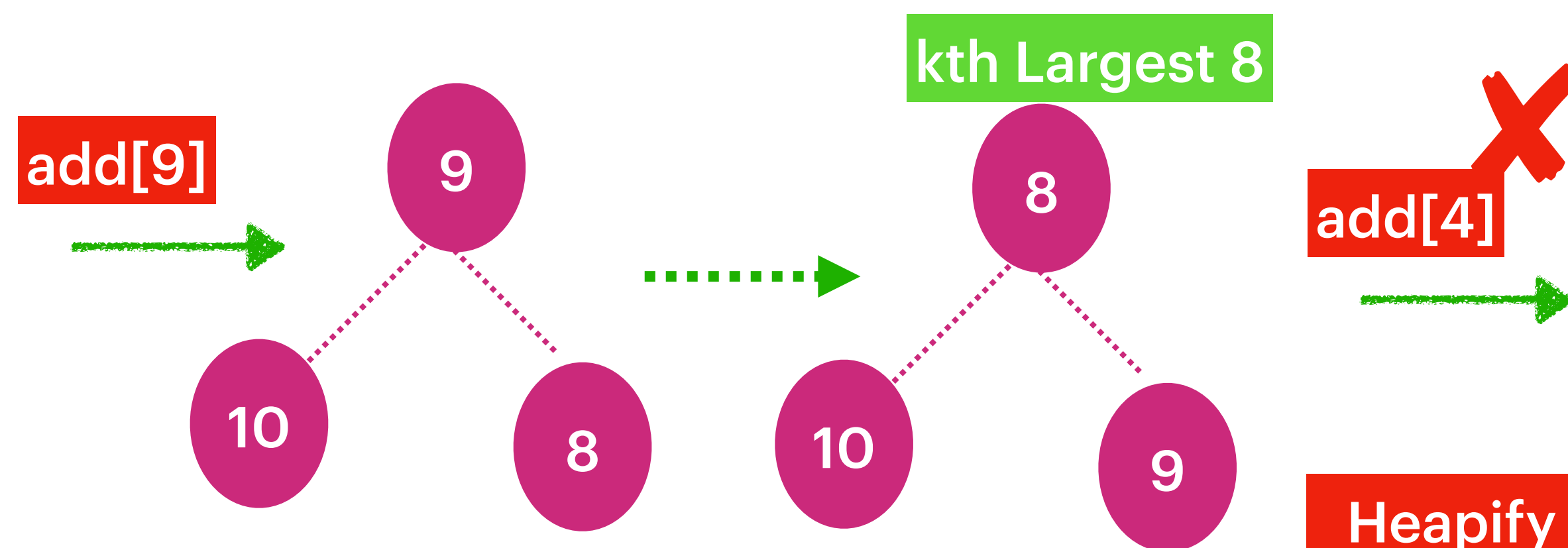
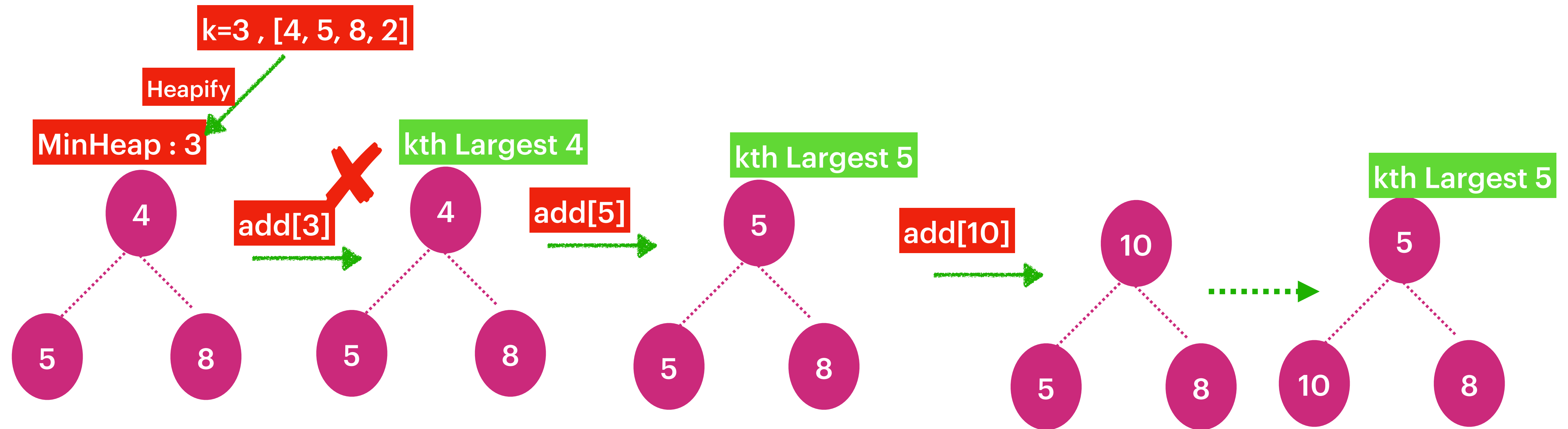
add[4]



kth Largest 8

[4,5,5,8,8]

[[k=3, construct[4, 5, 8, 2]], add[3], add[5], add[10], add[9], add[4]]



Algorithm :

Construct a MinHeap with K Size :

If the current element is greater than peek element
Then remove the peek and add current to the Heap.

Heapify + add/remove element to the Heap
Time Complexity : $O(n \log k) + O(n \log k) = n \log k$
Space Complexity : $O(k)$

Meeting Rooms II

Given an array of meeting time intervals intervals where $\text{intervals}[i] = [\text{start}_i, \text{end}_i]$, return the minimum number of conference rooms required.

Example 1:

Input: intervals = [[0,30],[5,10],
[15,20]]
Output: 2

Example 2:

Input: intervals = [[7,10],[2,4]]
Output: 1

Constraints:

$1 \leq \text{intervals.length} \leq 10^4$
 $0 \leq \text{start}_i < \text{end}_i$

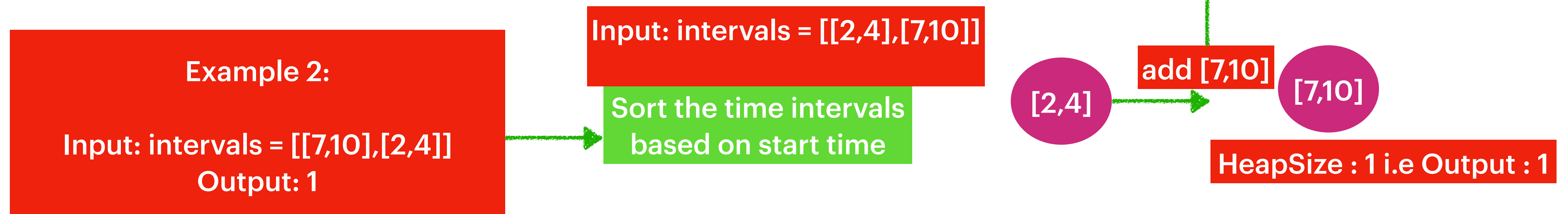
Example 1:
Input: intervals = $[[0,30],[5,10],[15,20]]$
Output: 2

By the time meeting3 starts
Meeting2 would over in Room2 so that we can use Room2



Output : Total Minimal No.of Rooms required = 2

Current meeting startTime \geq peek().endTime
So we can use same room.
Remove the peek & add the current one.



[[7,10],[45,60], [2,4],[3,7],[41,44],[9,15],[15,30]]

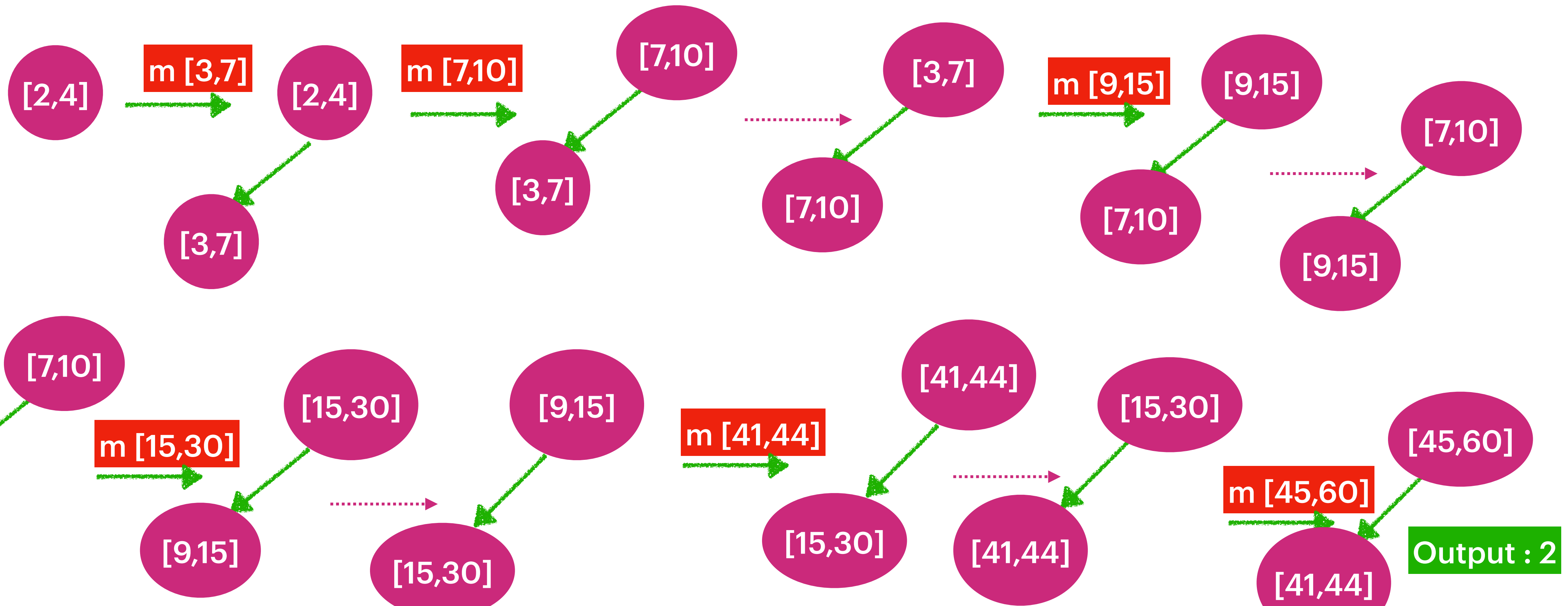


Sort the intervals based startTime

[[2,4],[3,7],[7,10],[9,15],[15,30] ,[41,44],[45,60]]

MinHeap based on endTime

[2,4]



Algorithm :

Sort the intervals based on startTime.

Construct minHeap based on endTime,
While adding interval to the heap, if the current interval
startTime >= peek() endTime then remove the
peek() & add the currentInterval.

Output would be the heapSize:

Time Complexity : sorting + adding/removing element to Heap
 $O(n \log n) + O(n \log k)$ where $k \leq n = O(n \log n)$