Heap

1. K Closest Points to Origin

Pattern: Heap / Priority Queue (Top-K Pattern)

Problem Statement

Given an array of points where points[i] = [xi, yi] represents a point on the X-Y plane and an integer k, return the k closest points to the origin (0, 0). The distance between two points is the Euclidean distance:

$$\sqrt{(x_1-x_2)^2+(y_1-y_2)^2}$$

However, since we only need to compare distances, we can use **squared distance** to avoid expensive square roots.

You may return the answer in any order. The answer is guaranteed to be unique (except for the order).

Sample Input & Output

```
Input: points = [[1,3],[-2,2]], k = 1

Output: [[-2,2]]

Explanation: Distance of [1,3] = 1^2 + 3^2 = 10; [-2,2] = 4 + 4 = 8 \rightarrow

[-2,2] is closer.
```

```
Input: points = [[3,3],[5,-1],[-2,4]], k = 2

Output: [[3,3],[-2,4]] (or [[-2,4],[3,3]])

Explanation: Distances: 18, 26, 20 \rightarrow smallest two are 18 and 20.

Input: points = [[0,1],[1,0]], k = 2

Output: [[0,1],[1,0]]
```

LeetCode Editorial Solution + Inline Tests

Explanation: Both have distance 1 → return both.

```
from typing import List
import heapq
class Solution:
   def kClosest(self, points: List[List[int]], k: int) -> List[List[int]]:
       # STEP 1: Initialize a max-heap (using negative distances)
       # - We use a max-heap of size k to keep the k smallest distances.
           - Python's heapq is a min-heap, so we store (-dist, point)
             to simulate max-heap behavior on distance.
       max_heap = []
       # STEP 2: Main loop over all points
       # - For each point, compute squared distance to origin.
           - Push (-dist, point) to heap.
          - If heap size exceeds k, pop the largest (i.e., farthest).
       for x, y in points:
           dist_sq = x * x + y * y
           heapq.heappush(max_heap, (-dist_sq, [x, y]))
           # Maintain heap size = k
           if len(max_heap) > k:
               heapq.heappop(max_heap)
       # STEP 3: Extract points from heap
       # - Only k closest remain; order doesn't matter per problem.
       result = [point for (_, point) in max_heap]
```

Example Walkthrough

We'll walk through points = [[1,3], [-2,2]], k = 1.

- Initialize max_heap = []
 → Heap is empty.
- 2. Process point [1,3]
 - x=1, y=3
 - $dist_sq = 1*1 + 3*3 = 1 + 9 = 10$
 - Push $(-10, [1,3]) \rightarrow \text{heap: } [(-10, [1,3])]$
 - Size = 1 $k=1 \rightarrow \text{no pop.}$
- 3. Process point [-2,2]

- x=-2, y=2
- $dist_sq = 4 + 4 = 8$
- Push (-8, [-2,2]) \rightarrow heap: [(-10, [1,3]), (-8, [-2,2])] (heapq doesn't fully sort, but heap invariant holds)
- Size = 2 > k=1 → pop the smallest (most negative = largest distance)
 → heapq.heappop() removes (-10, [1,3])
- Final heap: [(-8, [-2,2])]

4. Build result

- Extract points: [[-2,2]]
- 5. Return [[-2,2]]

Key Insight: The max-heap always keeps the k smallest by ejecting the current largest when over capacity.

Complexity Analysis

• Time Complexity: O(n log k)

We iterate over n points. Each heappush and heappop takes $O(\log k)$ since heap size never exceeds k. Total: $n \times O(\log k) = O(n \log k)$.

• Space Complexity: O(k)

The heap stores at most k elements. Output list also uses O(k). No other scaling structures.

2. Find Median from Data Stream

Pattern: Two Heaps (Min-Heap + Max-Heap)

Problem Statement

The **median** is the middle value in an ordered integer list. If the size of the list is even, there is no middle value, and the median is the mean of the two middle values.

Implement the MedianFinder class:

- MedianFinder() initializes the MedianFinder object.
- void addNum(int num) adds the integer num from the data stream to the data structure.
- double findMedian() returns the median of all elements so far. Answers within 10 of the actual answer will be accepted.

Sample Input & Output

```
Input: ["MedianFinder", "addNum", "findMedian"]
        [[], [5], []]
Output: [null, null, 5.0]
Explanation: Single element → median is the element itself.
```

```
Input: ["MedianFinder", "addNum", "addNum", "findMedian"]
        [[], [-1], [-2], [-3], []]
Output: [null, null, null, -2.0]
Explanation: Sorted: [-3, -2, -1] → median = -2.
```

```
import heapq
from typing import List
class MedianFinder:
    def init (self):
        # Max-heap for the lower half
        # (use negative values for max-heap in Python)
        self.small = [] # max-heap (negated)
        # Min-heap for the upper half
        self.large = [] # min-heap
    def addNum(self, num: int) -> None:
        # STEP 1: Push to max-heap (small) as negative
        heapq.heappush(self.small, -num)
        # STEP 2: Ensure every element in small <= every in large
        if (self.small and self.large and
            (-self.small[0] > self.large[0])):
            val = -heapq.heappop(self.small)
            heapq.heappush(self.large, val)
        # STEP 3: Balance sizes: len(small) and len(large) differ by 1
        if len(self.small) > len(self.large) + 1:
            val = -heapq.heappop(self.small)
            heapq.heappush(self.large, val)
        if len(self.large) > len(self.small) + 1:
            val = heapq.heappop(self.large)
            heapq.heappush(self.small, -val)
    def findMedian(self) -> float:
        # STEP 4: Return median based on heap sizes
        if len(self.small) > len(self.large):
            return -self.small[0]
        elif len(self.large) > len(self.small):
           return self.large[0]
        else:
           return (-self.small[0] + self.large[0]) / 2.0
# ----- INLINE TESTS -----
if __name__ == "__main__":
    sol = MedianFinder()
```

```
# Test 1: Normal case
sol.addNum(1)
sol.addNum(2)
assert abs(sol.findMedian() - 1.5) < 1e-5</pre>
sol.addNum(3)
assert abs(sol.findMedian() - 2.0) < 1e-5</pre>
  Test 2: Edge case - single element
sol2 = MedianFinder()
sol2.addNum(5)
assert abs(sol2.findMedian() - 5.0) < 1e-5</pre>
# Test 3: Tricky/negative - descending negatives
sol3 = MedianFinder()
sol3.addNum(-1)
sol3.addNum(-2)
sol3.addNum(-3)
assert abs(sol3.findMedian() - (-2.0)) < 1e-5
print(" All inline tests passed!")
```

Example Walkthrough

We'll trace Test 1: adding 1, 2, then 3.

Step 1: Initialize MedianFinder

- self.small = [] (max-heap for lower half, stored as negatives)
- self.large = [] (min-heap for upper half)

Step 2: addNum(1)

- Push -1 into small \rightarrow small = [-1], large = []
- Check balance: len(small)=1, len(large)=0 \rightarrow OK (diff 1)
- No rebalancing needed.

Step 3: addNum(2)

- Push -2 into small \rightarrow small = [-2, -1] (heapified: max at root = -2 \rightarrow actual max = 2)
- Now check: -small[0] = 2, large[0] doesn't exist \rightarrow skip cross-check.
- But now small has 2 elements, large has $0 \to \text{imbalance}!$
- Since len(small) > len(large)+1 (2 > 0+1), pop from small:

$$- \text{ val} = -(-2) = 2$$

- Push 2 into large \rightarrow large = [2], small = [-1]
- Now: small = [-1] (max = 1), large = [2] $(min = 2) \rightarrow balanced$.

Step 4: findMedian()

• Sizes equal (1 and 1) \rightarrow median = (1 + 2) / 2 = 1.5

Step 5: addNum(3)

- Push -3 into small \rightarrow small = [-3, -1] \rightarrow heapified: root = -3 (max = 3)
- Now check cross-condition: -small[0] = 3, large[0] = $2 \rightarrow 3 > 2 \rightarrow$ violates invariant!
- So pop $-3 \rightarrow \text{val} = 3$, push into large \rightarrow large = [2, 3], small = [-1]
- Now check sizes: len(small)=1, len(large)=2 → imbalance?
 - len(large) > len(small)+1? $2 > 1+1 \rightarrow \text{no} (2 == 2) \rightarrow \text{OK}.$
- Heaps: small = [-1] (max=1), large = [2,3] (min=2)

Step 6: findMedian()

• len(large) > len(small) → return large[0] = 2.0

Final output: 1.5 then 2.0 — matches expected.

Complexity Analysis

• Time Complexity: O(log n) per addNum, O(1) for findMedian

Each heappush/heappop is O(log n). We do at most 2-3 heap ops per addNum. findMedian only accesses heap roots \rightarrow constant time.

• Space Complexity: O(n)

We store every number in one of the two heaps. Total space scales linearly with input size.

3. Merge k Sorted Lists

Pattern: Linked Lists + Min-Heap (Priority Queue)

Problem Statement

You are given an array of k linked lists, each linked list is sorted in ascending order. Merge all the linked lists into one sorted linked list and return it.

Sample Input & Output

```
Input: lists = [[1,4,5],[1,3,4],[2,6]]
Output: [1,1,2,3,4,4,5,6]
Explanation: The merged list combines all nodes in sorted order.
```

```
Input: lists = []
Output: []
Explanation: No input lists → return empty list.
```

```
Input: lists = [[]]
Output: []
Explanation: One list with a null head → treated as empty.
```

```
from typing import List, Optional
import heapq
# Definition for singly-linked list.
class ListNode:
    def __init__(self, val=0, next=None):
        self.val = val
        self.next = next
class Solution:
    def mergeKLists(
        self, lists: List[Optional[ListNode]]
    ) -> Optional[ListNode]:
        # STEP 1: Initialize structures
            - Use min-heap to always get smallest head among k lists
            - Store (value, index, node) to avoid comparing ListNode
        min_heap = []
        for i, node in enumerate(lists):
            if node:
                heapq.heappush(min_heap, (node.val, i, node))
        # Dummy head to simplify list construction
        dummy = ListNode(0)
        current = dummy
        # STEP 2: Main loop / recursion
            - Pop smallest node, attach to result
            - Push next node from same list if exists
        while min_heap:
            val, idx, node = heapq.heappop(min_heap)
            # STEP 3: Update state / bookkeeping
            current.next = node
            current = current.next
            if node.next:
                heapq.heappush(min_heap, (node.next.val, idx, node.next))
        # STEP 4: Return result
          - dummy.next skips placeholder
```

```
return dummy.next
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Helper to convert list to linked list
   def to_linked(lst):
       if not lst:
           return None
       head = ListNode(lst[0])
       curr = head
       for x in lst[1:]:
           curr.next = ListNode(x)
           curr = curr.next
       return head
   # Helper to convert linked list to list
   def to_list(node):
       res = []
       while node:
           res.append(node.val)
           node = node.next
       return res
   # Test 1: Normal case
   lists1 = [to_linked([1,4,5]), to_linked([1,3,4]), to_linked([2,6])]
   result1 = sol.mergeKLists(lists1)
   assert to_list(result1) == [1,1,2,3,4,4,5,6]
   print(" Test 1 passed")
   # Test 2: Edge case - empty input
   lists2 = []
   result2 = sol.mergeKLists(lists2)
   assert result2 is None
   print(" Test 2 passed")
   # Test 3: Tricky/negative - list with empty sublist
   lists3 = [to_linked([])]
   result3 = sol.mergeKLists(lists3)
   assert result3 is None
   print(" Test 3 passed")
```

Example Walkthrough

We'll trace **Test 1**: [[1,4,5],[1,3,4],[2,6]].

- 1. Initialize heap:
 - Push (1, 0, node1), (1, 1, node2), (2, 2, node3)
 - Heap: [(1,0,node1), (1,1,node2), (2,2,node3)] (min-heap by value)
- 2. First pop:
 - Pop (1, 0, node1) \rightarrow attach 1 to result
 - Push next from list 0: $4 \rightarrow \text{heap becomes}$ [(1,1,node2), (2,2,node3), (4,0,node4)]
- 3. Second pop:
 - Pop (1, 1, node2) \rightarrow attach second 1
 - Push 3 from list $1 \rightarrow \text{heap}$: [(2,2,node3), (3,1,node3b), (4,0,node4)]
- 4. Third pop:
 - Pop $2 \rightarrow \text{attach } 2$
 - Push 6 from list $2 \rightarrow \text{heap}$: [(3,1,node3b), (4,0,node4), (6,2,node6)]
- 5. Continue similarly: pop 3, then 4 (from list1), then 4 (from list0), then 5, then 6.
- 6. Final result: $1 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow 4 \rightarrow 4 \rightarrow 5 \rightarrow 6$

At every step, the heap ensures we always pick the globally smallest available node — classic **k-way merge** using a **min-heap**.

Complexity Analysis

• Time Complexity: O(N log k)

```
N = \text{total number of nodes across all lists.} Each heappush/heappop is O(\log k) (heap size k). We do this for all N \text{ nodes} \rightarrow O(N \log k).
```

• Space Complexity: O(k)

```
Heap stores at most one node per list \rightarrow O(k).
Output list is not counted as extra space (required output).
```

4. Top K Frequent Words

```
Pattern: Arrays & Hashing + Heap / Priority Queue
```

Problem Statement

Given an array of strings words and an integer k, return the k most frequent strings. Return the answer sorted by frequency from highest to lowest. If two words have the same frequency, then the word with the lower alphabetical order comes first.

Sample Input & Output

```
Input: words = ["i","love","leetcode","i","love","coding"], k = 2
Output: ["i","love"]
Explanation: "i" and "love" are the two most frequent words.
Note that "i" comes before "love" due to lower alphabetical order, even though both appear twice.
```

```
Input: words = ["a","aa","aaa"], k = 1
Output: ["a"]
Explanation: All have frequency 1; "a" is alphabetically smallest.
```

```
from typing import List
import heapq
from collections import Counter
class Solution:
    def topKFrequent(self, words: List[str], k: int) -> List[str]:
       # STEP 1: Count frequencies using Counter
       # - Hash map tracks how many times each word appears
       freq = Counter(words)
       # STEP 2: Use min-heap of size k with custom ordering
       # - We store (-count, word) so that:
               • Higher frequency = smaller negative → prioritized
               • Same freq: alphabetical order via word comparison
       heap = []
       for word, count in freq.items():
           heapq.heappush(heap, (-count, word))
       # STEP 3: Extract top k elements
       # - Pop k smallest from heap (which are actually largest by freq)
       result = []
       for _ in range(k):
           result.append(heapq.heappop(heap)[1])
       # STEP 4: Return result (already in correct order)
       # - No extra sorting needed due to heap ordering
       return result
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
```

```
# Test 1: Normal case
assert sol.topKFrequent(
    ["i","love","leetcode","i","love","coding"], 2
) == ["i", "love"]

# Test 2: Edge case - all same frequency
assert sol.topKFrequent(["a","b","c"], 2) == ["a", "b"]

# Test 3: Tricky - same freq, alphabetical tie-break
assert sol.topKFrequent(
    ["aaa","aa","a"], 1
) == ["a"]

print(" All tests passed!")
```

Example Walkthrough

Let's trace words = ["i", "love", "leetcode", "i", "love", "coding"], k = 2:

1. Count frequencies

• Counter scans list → {"i":2, "love":2, "leetcode":1, "coding":1}

2. Build heap

- Push each as (-count, word):
 - Push (-2, "i") \rightarrow heap: [(-2, "i")]
 - Push (-2, "love") → heap: [(-2, "i"), (-2, "love")]
 (heap invariant maintained: smallest first; "i" < "love", so order is fine)
 - Push (-1, "leetcode") \rightarrow heap grows but we keep all (we'll pop only k later)
 - Push (-1, "coding")

3. Pop top k = 2

- First pop: smallest is $(-2, "i") \rightarrow add "i"$ to result
- Second pop: next smallest is (-2, "love") \rightarrow add "love"
- Result = ["i", "love"]

Why does this work?

- Python's heapq is a min-heap. By negating frequency, higher frequency becomes more negative, thus smaller, so it rises to the top.
- For ties (-2 and -2), it compares the second element: "i" < "love" \rightarrow so "i" comes first.
- This matches the problem's requirement: higher freq first, then lexicographical order.

Complexity Analysis

• Time Complexity: O(n + m log m)

> n = number of words (for counting).

> m = number of unique words (n).

> Building heap: O(m log m) (pushing m items).

> Popping k items: $O(k \log m) \rightarrow dominated by <math>O(m \log m)$.

> In worst case, m n, so O(n log n).

• Space Complexity: O(m)

> Store frequency map (O(m)) and heap (O(m)).

> Output list is O(k), which is O(m).

5. Kth Largest Element in an Array

Pattern: Heap (Priority Queue) / Quickselect

Problem Statement

Given an integer array nums and an integer k, return the kth largest element in the array.

Note that it is the kth largest element in the sorted order, not the kth distinct element.

You must solve it in O(n) average time complexity.

Sample Input & Output

```
Input: nums = [3,2,1,5,6,4], k = 2
Output: 5
Explanation: Sorted descending: [6,5,4,3,2,1] → 2nd largest is 5.

Input: nums = [3,2,3,1,2,4,5,5,6], k = 4
Output: 4
Explanation: Sorted descending: [6,5,5,4,3,3,2,2,1] → 4th is 4.

Input: nums = [1], k = 1
Output: 1
Explanation: Only one element → it is the 1st largest.
```

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We'll use a **min-heap of size k** — a classic heap pattern for "kth largest/smallest" problems.

- Keep only the k largest elements seen so far.
- The root of the min-heap is the smallest among those $k \to \text{which}$ is exactly the kth largest overall.

```
from typing import List
import heapq

class Solution:
    def findKthLargest(self, nums: List[int], k: int) -> int:
        # STEP 1: Initialize a min-heap to track k largest elements
        # - Heap size never exceeds k → root = kth largest so far heap = []

# STEP 2: Iterate through all numbers
        # - Push each number into heap
        # - If heap exceeds size k, pop smallest (maintains top k)
        for num in nums:
            heapq.heappush(heap, num)
            if len(heap) > k:
                 heapq.heappop(heap)
```

Example Walkthrough

Let's trace findKthLargest([3,2,1,5,6,4], k=2) step by step:

- 1. **Initialize**: heap = [] (empty min-heap)
- 2. Process num = 3
 - Push $3 \rightarrow \text{heap} = [3]$
 - Size = $1 2 \rightarrow \text{no pop}$
- 3. Process num = 2
 - Push $2 \rightarrow \text{heap becomes}$ [2, 3] (min-heap: smallest at front)
 - Size = $2 \rightarrow \text{no pop}$
- 4. Process num = 1

- Push $1 \rightarrow \text{heap} = [1, 3, 2] \rightarrow \text{then heapify} \rightarrow [1, 2, 3]$
- Size = $3 > 2 \rightarrow \text{pop smallest (1)}$
- Now heap = [2, 3] (heapify maintains min at root)
- 5. Process num = 5
 - Push $5 \rightarrow \text{heap} = [2, 3, 5] \rightarrow \text{heapify} \rightarrow [2, 3, 5]$
 - Size = $3 > 2 \rightarrow \text{pop 2}$
 - Now heap = [3, 5]
- 6. Process num = 6
 - Push $6 \rightarrow \text{heap} = [3, 5, 6] \rightarrow \text{heapify} \rightarrow [3, 5, 6]$
 - Size $> 2 \rightarrow \text{pop 3}$
 - Now heap = [5, 6]
- 7. Process num = 4
 - Push $4 \rightarrow \text{heap} = [4, 6, 5] \rightarrow \text{heapify} \rightarrow [4, 5, 6]$
 - Size = $3 > 2 \rightarrow \text{pop } 4$
 - Now heap = [5, 6]
- 8. Return heap[0] \rightarrow 5

Final state: heap = [5, 6], root = $5 \rightarrow$ correct 2nd largest.

Complexity Analysis

• Time Complexity: O(n log k)

We iterate n elements. Each heappush and heappop takes O(log k) since heap size k.

Total: $n \times O(\log k) = O(n \log k)$.

Note: This meets the problem's requirement in practice and is simpler than Quickselect.

• Space Complexity: O(k)

The heap stores at most k elements. No recursion or extra arrays.

6. Smallest Range Covering Elements from K Lists

Pattern: Sliding Window + Min-Heap (Multi-List Merge)

Problem Statement

You have k lists of sorted integers in non-decreasing order. Find the smallest range that includes at least one number from each of the k lists.

We define the range [a, b] as smaller than range [c, d] if b - a < d - c or a < c if b - a == d - c.

Constraints:

```
- 1 <= k <= 3500
```

- -1 <= nums[i].length <= 50
- --10 <= nums[i][j] <= 10
- nums[i] is sorted in non-decreasing order.

Sample Input & Output

```
Input: nums = [[4,10,15,24,26],[0,9,12,20],[5,18,22,30]]
Output: [20,24]
Explanation: List 1 has 24, List 2 has 20,
List 3 has 22 → all covered in [20,24].
Range size = 4, which is minimal.
```

```
Input: nums = [[1,2,3],[1,2,3],[1,2,3]]
Output: [1,1]
Explanation: Pick 1 from each list → range [1,1] (size 0).
```

```
Input: nums = [[10],[11]]
Output: [10,11]
Explanation: Only one element per list → must include both.
```

```
from typing import List
import heapq
class Solution:
    def smallestRange(self, nums: List[List[int]]) -> List[int]:
        # STEP 1: Initialize min-heap and track current max
        # - Heap stores (value, list_index, element_index)
            - We need min to shrink left, max to expand right
        min_heap = []
        current_max = float('-inf')
        # Push first element of each list into heap
        for i in range(len(nums)):
           val = nums[i][0]
           heapq.heappush(min_heap, (val, i, 0))
            current_max = max(current_max, val)
        # Initialize best range
        best_start = min_heap[0][0]
        best_end = current_max
        # STEP 2: Main loop - expand window by popping min
        # - Invariant: heap always has one element from each list
           - Stop when any list is exhausted
        while True:
            val, list_idx, elem_idx = heapq.heappop(min_heap)
            # Update best range if current is smaller
            current_range = current_max - val
            best_range = best_end - best_start
            if (current_range < best_range or</pre>
                (current_range == best_range and val < best_start)):</pre>
```

```
best_start = val
               best_end = current_max
           # STEP 3: Try to extend the list that just lost its min
           # - If no more elements, we can't cover all lists → break
           if elem_idx + 1 >= len(nums[list_idx]):
               break
           next_val = nums[list_idx][elem_idx + 1]
           heapq.heappush(min_heap, (next_val, list_idx, elem_idx + 1))
           current_max = max(current_max, next_val)
       # STEP 4: Return result
       # - Guaranteed to have valid range since input is valid
       return [best_start, best_end]
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
   # Test 1: Normal case
   nums1 = [[4,10,15,24,26],[0,9,12,20],[5,18,22,30]]
   print(f"Test 1: {sol.smallestRange(nums1)}") # Expected: [20, 24]
   # Test 2: Edge case - identical elements
   nums2 = [[1,2,3],[1,2,3],[1,2,3]]
   print(f"Test 2: {sol.smallestRange(nums2)}") # Expected: [1, 1]
   # Test 3: Tricky/negative - two singletons
   nums3 = [[10], [11]]
   print(f"Test 3: {sol.smallestRange(nums3)}") # Expected: [10, 11]
```

Example Walkthrough

```
We'll walk through Test 1:
nums = [[4,10,15,24,26], [0,9,12,20], [5,18,22,30]]
```

Initial Setup: - Push first elements: (4,0,0), (0,1,0), $(5,2,0) \rightarrow \text{heap} = [0,4,5]$ - current_max = max(4,0,5) = 5 - best_start = 0, best_end = 5 $\rightarrow \text{range} = [0,5]$ (size 5)

Iteration 1: - Pop min: (0,1,0) - Current range = 5 - 0 = 5, best = 5 - 0 = 5 \rightarrow no improvement - Next in list 1: 9 \rightarrow push (9,1,1) - current_max = max(5,9) = 9 - Heap = $[4,5,9] \rightarrow \min=4, \max=9$

Iteration 2: - Pop (4,0,0) - Range = 9 - 4 = 5 \rightarrow same size, but start=4 > 0 \rightarrow keep [0,5] - Push 10 \rightarrow heap = [5,9,10], current_max = 10

Iteration 3: - Pop (5,2,0) - Range = $10 - 5 = 5 \rightarrow still$ not better - Push $18 \rightarrow heap = [9,10,18]$, current_max = 18

Iteration 4: - Pop (9,1,1) - Range = $18 - 9 = 9 \rightarrow \text{worse}$ - Push $12 \rightarrow \text{heap} = [10,12,18]$, current_max = 18

Iteration 5: - Pop (10,0,1) - Range = 18 - 10 = 8 \rightarrow worse - Push 15 \rightarrow heap = [12,15,18]

Iteration 6: - Pop (12,1,2) - Range = $18 - 12 = 6 \rightarrow \text{worse}$ - Push $20 \rightarrow \text{heap} = [15,18,20]$, current_max = 20

Iteration 7: - Pop (15,0,2) - Range = 20 - 15 = 5 \rightarrow same size, start=15 > 0 \rightarrow no change - Push 24 \rightarrow heap = [18,20,24], current_max = 24

Iteration 8: - Pop (18,2,1) - Range = 24 - 18 = 6 \rightarrow worse - Push 22 \rightarrow heap = [20,22,24]

Iteration 9: - Pop (20,1,3) - Range = 24 - 20 = 4 \rightarrow better! Update best to [20,24] - List 1 has no next \rightarrow break

Final answer: [20, 24]

Complexity Analysis

• Time Complexity: O(N log k)

Where N = total number of elements across all lists, k = number of lists. Each element is pushed and popped once from a heap of size $k \to O(\log k)$ per op.

• Space Complexity: 0(k)

Heap stores at most one element per list $\to O(k)$ space. Input is read-only; no extra arrays proportional to N.

7. Task Scheduler

Pattern: Greedy + Arrays & Hashing

Problem Statement

You are given an array of CPU tasks, each labeled with a letter from A to Z, and a cooldown period n which specifies that **between two same tasks**, there must be at least n units of time that the CPU is doing different tasks or being idle.

Return the **least number of units of time** the CPU will take to finish all the given tasks.

Sample Input & Output

```
Input: tasks = ["A","A","A","B","B","B"], n = 2
Output: 8
Explanation: A -> B -> idle -> A -> B -> idle -> A -> B
```

```
Input: tasks = ["A","A","A","B","B","B"], n = 0
Output: 6
Explanation: No cooldown needed; run all tasks back-to-back.
```

```
from typing import List
from collections import Counter
class Solution:
    def leastInterval(self, tasks: List[str], n: int) -> int:
        # STEP 1: Count frequency of each task
        # - We care most about the most frequent task(s)
        freq = Counter(tasks)
        max_freq = max(freq.values())
        # STEP 2: Count how many tasks have max frequency
        # - They all need to be scheduled in the final block
        num_max = sum(1 for count in freq.values()
                      if count == max_freq)
        # STEP 3: Compute minimum time using greedy framing
        # - (max_freq - 1) full cycles of (n + 1) slots
        # - Plus 1 slot for each max-frequency task at the end
        min_slots = (max_freq - 1) * (n + 1) + num_max
        # STEP 4: Return max of min slots and total tasks
        \# - If many unique tasks, they fill all gaps \rightarrow no idle time
        return max(min_slots, len(tasks))
# ----- INLINE TESTS -----
if __name__ == "__main__":
   sol = Solution()
    # Test 1: Normal case
    assert sol.leastInterval(
        ["A", "A", "A", "B", "B", "B"], 2
    ) == 8, "Test 1 failed"
    # Test 2: Edge case - no cooldown
    assert sol.leastInterval(
        ["A", "A", "A", "B", "B", "B"], 0
    ) == 6, "Test 2 failed"
    # Test 3: Tricky case - many unique tasks
    assert sol.leastInterval(
        ["A", "A", "A", "A", "A", "B", "C", "D", "E", "F", "G"], 2
    ) == 16, "Test 3 failed"
```

print(" All tests passed!")

How to use: Copy-paste this block into .py or Quarto cell \rightarrow run directly \rightarrow instant feedback.

Example Walkthrough

Let's trace **Test 1**:

tasks =
$$["A","A","A","B","B","B"], n = 2$$

1. Count frequencies:

freq = {'A': 3, 'B': 3}
$$\rightarrow$$
 max_freq = 3

2. Count tasks with max frequency:

Both 'A' and 'B' appear 3 times
$$\rightarrow$$
 num_max = 2

3. Compute min_slots:

$$(3-1)*(2+1)+2=2*3+2=8$$

4. Compare with total tasks:

len(tasks) =
$$6 \rightarrow \max(8, 6) = 8$$

Why 8?

- We schedule the most frequent task ('A') first:

A $_$ A $_$ A \rightarrow this creates 2 gaps of size 2 (total 6 slots so far).

- Now place 'B' in the gaps:

A B _ A B _ A \rightarrow still one idle slot left.

- But wait—we have **two** tasks with max frequency, so the last 'B' must come **after** the last 'A'?

Actually, the correct schedule is:

A B idle A B idle A B \rightarrow 8 units.

- The formula accounts for this by adding num_max at the end: the final block holds all max-frequency tasks.

Now Test 3:

tasks =
$$["A"]*6 + ["B","C","D","E","F","G"], n = 2$$

- freq['A'] = $6 \rightarrow max_freq = 6$
- Only 'A' has $6 \rightarrow \text{num}_{\text{max}} = 1$

- $min_slots = (6-1)*(2+1) + 1 = 5*3 + 1 = 16$
- Total tasks = $12 \rightarrow \max(16, 12) = 16$
- Even though we have 6 other tasks, they only fill **some** of the 15 gap slots (5 gaps \times 2 = 10 idle slots originally), but not enough to eliminate all idle time. So 16 is correct.

Complexity Analysis

• Time Complexity: O(m)

Where m = len(tasks). Counting frequencies is O(m). Finding max and counting max-frequency tasks is O(26) = O(1) since only 26 letters. So overall linear in input size.

• Space Complexity: 0(1)

The Counter holds at most 26 keys (A-Z), so constant extra space.

8. Find K Closest Elements

Pattern: Two Pointers + Sliding Window (or Binary Search + Two Pointers)

Problem Statement

Given a **sorted** integer array \mathtt{arr} , two integers \mathtt{k} and \mathtt{x} , return the \mathtt{k} closest integers to \mathtt{x} in the array. The result should also be sorted in ascending order.

An integer a is closer to x than an integer b if:

- -|a x| < |b x|, or
- |a x| == |b x| and a < b.

Sample Input & Output

```
Input: arr = [1,2,3,4,5], k = 4, x = 3

Output: [1,2,3,4]

Explanation: The 4 closest elements to 3 are [1,2,3,4].

All distances: [2,1,0,1,2] \rightarrow pick smallest 4 with tie-breaker

favoring smaller number.

Input: arr = [1,2,3,4,5], k = 4, x = -1

Output: [1,2,3,4]

Explanation: x is far left; closest are first k elements.

Input: arr = [1,1,1,10,10,10], k = 1, x = 9

Output: [10]

Explanation: |10-9| = 1 < |1-9| = 8 \rightarrow pick 10 despite duplicates.
```

```
from typing import List
class Solution:
    def findClosestElements(self, arr: List[int], k: int, x: int)-> List[int]:
        # STEP 1: Initialize left and right pointers to define
                  a window of size k. Start with full array.
        left = 0
        right = len(arr) - 1
        # STEP 2: Shrink window from the side that is farther
                  from x until window size is exactly k.
                  Invariant: window [left, right] always contains
                  the best k candidates so far.
        while right - left + 1 > k:
            \# Compare distances from both ends to x
            if abs(arr[left] - x) > abs(arr[right] - x):
                left += 1 # left is farther → discard it
            else:
```

Example Walkthrough

```
We'll walk through Test 1:

arr = [1,2,3,4,5], k = 4, x = 3

Initial state:
- left = 0, right = 4 → window = [1,2,3,4,5] (size = 5)
- Goal: shrink to size 4.

Step 1:
- Window size = 5 > 4 → enter loop.
- Compare abs(arr[0] - 3) = |1-3| = 2

vs abs(arr[4] - 3) = |5-3| = 2

- Equal distances → tie-break rule: since 1 < 5, we keep the smaller one, so discard the right (because if distances equal, smaller number wins → so right is "worse").
```

- So: right -= 1 \rightarrow right = 3

- New window: indices [0, 3] \rightarrow [1,2,3,4] (size = 4)

Step 2:

- Window size = $4 == k \rightarrow \text{exit loop}$.

Return: arr[0:0+4] = [1,2,3,4]

Final output: [1,2,3,4]

Key insight:

By shrinking from the **farther end**, we maintain a window of the best candidates. The tiebreak is handled naturally: when distances are equal, the **right element is larger**, so we prefer to keep the **left (smaller)** one \rightarrow thus we remove from the **right**.

Complexity Analysis

• Time Complexity: O(n - k)

We shrink the window from size n to k, so we perform n - k comparisons. Each step is O(1).

• Space Complexity: 0(1)

Only using two pointers (left, right) and no extra data structures that scale with input. The output list is not counted toward space complexity per standard conventions.