



# Top K Elements: Introduction

Let's go over the Top K Elements pattern, its real-world applications, and some problems we can solve with it.

## We'll cover the following



- Overview
- Examples
- Does my problem match this pattern?
- Real-world problems
- Strategy time!

## Overview

The **top K elements** pattern helps find some specific  $k$  number of elements from the given data with optimum time complexity.

Many problems ask us to find the top, the smallest, or the most/least frequent  $k$  elements in an unsorted list of elements. To solve such problems, sorting the list takes  $O(n \log(n))$  time, then finding the  $k$  elements takes  $O(k)$  time. However, the top  $k$  elements pattern can allow us to solve the problem using  $O(n \cdot \log k)$  time without sorting the list first.

Which data structure can we use to solve such problems? The best data structure to keep track of the smallest or largest  $k$  elements is heap. With this pattern, we either use a max-heap or a min-heap to find the smallest or largest  $k$  elements, respectively.

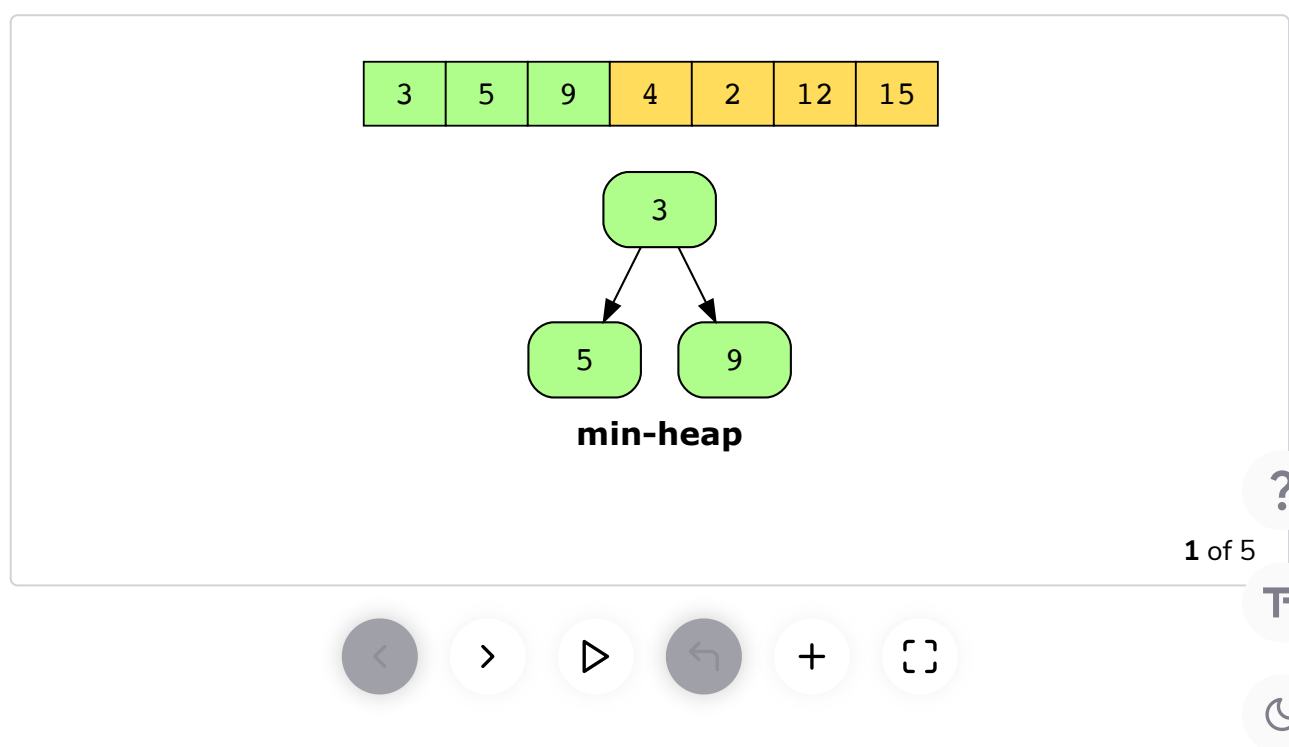


For example, let's look at how this pattern takes steps to solve the problem of finding the top  $k$  largest elements (using min-heap) or top  $k$  smallest elements (using max-heap):

1. Insert the first  $k$  elements from the given set of elements to the min-heap or max-heap.
2. Iterate through the rest of the elements.
  - I. For min-heap, if you find the larger element, remove the top (smallest number) of the min-heap and insert the new larger element.
  - II. For max-heap, if you find the smaller element, remove the top (largest number) of the max-heap and insert the new smaller element.

Iterating the complete list takes  $O(n)$  time, and the heap takes  $O(\log k)$  time for insertion. However, we get the  $O(1)$  access to the  $k$  elements using the heap.

Let's look at the following illustration to understand how we can use min-heap to find the top  $k$  elements.

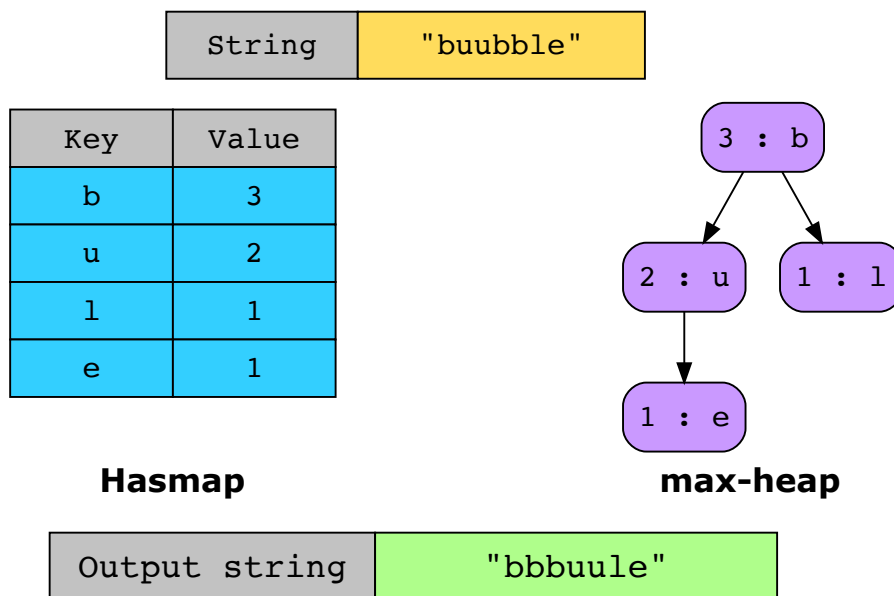


# Examples

The following examples illustrate some problems that can be solved with this approach:

## Sort characters by frequency

We'll calculate frequencies of each character and store them in the hash map. We'll use the frequencies to maintain the max-heap and sort characters by frequency.



1 of 2



## Does my problem match this pattern?

- Yes, if both of these conditions are fulfilled:
  - We need to find the largest, smallest, most frequent, or least frequent subset of elements in an unsorted list.
  - This may be the requirement of the final solution, or it may be necessary as an intermediate step toward the final solution.
- No, if any of these conditions is fulfilled:
  - The input data structure does not support random access.



- The input data is already sorted according to the criteria relevant to solving the problem.
- If only 1 extreme value is required, that is,  $k = 1$ , as that problem can be solved in  $O(n)$  with a simple scan through the input array.

## Real-world problems

Many problems in the real world use the top K elements pattern. Let's look at some examples.

- **Uber:** Select at least the  $n$  nearest drivers within the user's vicinity, avoiding the drivers that are too far away.
- **Stocks:** Given the set of IDs of brokers, determine the top K broker's



### Strategy time!

Match the problems that can be solved using the top K elements pattern.

**Note:** Select a problem in the left-hand column by clicking it, and then click one of the two options in the right-hand column.

#### Match The Answer

ⓘ Select an option from the left-hand side

Rearrange the given string so that no two identical characters are adjacent to each other.

Top K Elements



Detect a cycle in a linked list.

Some other pattern

Sort the characters of the given string by frequency.

Implement a stack in which the pop operation removes the most frequent element.

Reset

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Median of Two Sorted...

Kth Largest Element i...



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