

Find the Kth Largest Array Element [LeetCode](#)

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.

Example: [3,2,1,5,6,4], *k* = 3

Output: The 3rd largest element: 4

Approach 1: Function to find the kth largest element in an array using a brute force approach

- **Function Purpose:** To find the kth largest element in an array using a brute force approach.
- **Explanation:**
 - Sort the array in ascending order and return the kth element from the end.
- **Time Complexity:** $O(N * \log(N))$, where *N* is the number of elements in the array.
- **Space Complexity:** $O(1)$ since it operates in-place.

Approach 2: Function to find the kth largest element in an array using a max heap

- **Function Purpose:** To find the kth largest element in an array using a max heap.
- **Explanation:**
 - Use a max heap to store elements.
 - Remove the *k* - 1 largest elements, leaving the kth largest.
- **Time Complexity:** $O(N * \log(N))$, where *N* is the number of elements in the array.
- **Space Complexity:** $O(N)$ due to the max heap.

Approach 3: Function to find the kth largest element in an array using a min heap

- **Function Purpose:** To find the kth largest element in an array using a min heap.
- **Explanation:**
 - Use a min heap to store elements.
 - Remove elements until the heap size is *k*, leaving the kth largest.
- **Time Complexity:** $O(N * \log(N))$, where *N* is the number of elements in the array.

- **Space Complexity:** $O(N)$ due to the min heap.

Approach 4: Function to find the kth largest element in an array using a min heap (Optimized Approach)

- **Function Purpose:** To find the kth largest element in an array using an optimized min heap approach.
- **Explanation:**
 - Create a min heap to maintain the k largest elements.
 - Insert the first k elements into the min heap.
 - For the remaining elements, if an element is greater than the current minimum in the heap, replace the minimum with the larger element.
 - The top element of the min heap is the kth largest element.
- **Time Complexity:** $O(N * \log(K))$, where K is the value of k.
- **Space Complexity:** $O(K)$ for the min heap.

Conclusion:

- All four approaches yield the same result for finding the 3rd largest element, which is 4.
- The Brute Force approach is straightforward but has a time complexity of $O(N * \log(N))$.
- The Max Heap and Min Heap approaches also have a time complexity of $O(N * \log(N))$ but require $O(N)$ space for the heap.
- **The Optimized Min Heap approach has a better time complexity of $O(N * \log(K))$ and requires $O(K)$ space for the heap.**