

# Sorting before Two Pointer

The **total time complexity** is derived by analyzing each step of the algorithm and summing their individual complexities. Here's a detailed breakdown:

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## Steps of the Algorithm

1. **Sorting the Array ( `arr.sort()` ):**
    - The `sort()` function sorts the array in ascending order.
    - Sorting an array of size `n` takes  **$O(n \log n)$**  using efficient sorting algorithms like Timsort (used in Python).
  2. **Two-Pointer Traversal (the `while` loop):**
    - After sorting, the two-pointer technique is applied.
    - The two pointers (`left` and `right`) traverse the array at most once. In each iteration:
      - Either `left` is incremented, or `right` is decremented.
      - This ensures that the loop runs for at most  **$O(n)$**  iterations.
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## Combining the Steps

- Sorting takes  **$O(n \log n)$** .
  - The two-pointer traversal takes  **$O(n)$** .
  - Since these steps are performed sequentially (not nested), their complexities are **added**:  
$$\text{Total Time Complexity} = O(n \log n) + O(n)$$
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## Simplifying the Complexity

- In Big O notation, only the **dominant term** matters as `n` grows large.
    - **$O(n \log n)$**  dominates  **$O(n)$**  because logarithmic growth adds a significant factor to linear growth.
  - Therefore, the total complexity simplifies to:  $O(n \log n)$
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## Why Approximation?

The symbol  $\approx$  in  $O(n \log n) + O(n) \approx O(n \log n)$  indicates that:

- The  $O(n)$  term is negligible compared to  $O(n \log n)$  for large  $n$ .
  - So the effective time complexity is considered  $O(n \log n)$ .
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## Practical Example

Let's assume  $n = 1,000,000$ :

- **Sorting:**  $n \log n = 1,000,000 \cdot \log_2 1,000,000 \approx 20,000,000$  operations.
  - **Traversal:**  $O(n) = 1,000,000$  operations.
  - Clearly, the sorting step dominates.
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