

Sorting Algorithms

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Abstract

Sorting algorithms are fundamental to computer science, providing essential tools for organizing data efficiently. This paper explores and compares five widely used sorting algorithms: Insertion Sort, Merge Sort, Bubble Sort, Heap Sort, and Quick Sort. We analyze each algorithm's time complexity, space complexity, and performance in different scenarios. Through empirical analysis and theoretical discussion, this study aims to provide a comprehensive understanding of these algorithms' strengths and weaknesses, helping practitioners select the most appropriate sorting method for specific applications.

Keywords: Sorting algorithms, Insertion Sort, Merge Sort, Bubble Sort, Heap Sort, Quick Sort, Time complexity, Space Complexity, Algorithm Performance

Introduction

The history of sorting algorithms dates back to the early days of computer science and remains a critical area of study. Sorting is the process of arranging data in a specific order, typically ascending or descending. The efficiency of sorting algorithms directly impacts the performance of numerous applications, including database management, search algorithms, and data analysis.

This paper focuses on five prominent sorting algorithms: Insertion Sort, Merge Sort, Bubble Sort, Heap Sort, and Quick Sort. These algorithms have been extensively studied and applied in various domains due to their unique characteristics and performance profiles. Insertion Sort, known for its simplicity, performs well on small or nearly sorted datasets. Merge Sort, a divide-and-conquer algorithm, is stable and effective for large datasets. Bubble Sort, though less efficient for large datasets, is easy to implement and understand. Heap Sort, utilizing a binary heap data structure, offers reliable

performance without requiring additional space. Quick Sort, often the fastest in practice, uses a partitioning approach but can degrade in performance under certain conditions.

The primary goal of this study is to compare these sorting algorithms based on their time and space complexities, stability, and practical performance across different types of datasets. By providing a detailed analysis, we aim to assist software developers and computer scientists in selecting the most suitable sorting algorithm for their specific needs.

Our contributions include:

- A comprehensive review of the theoretical foundations of each sorting algorithm.
- An empirical performance comparison using various datasets.
- A discussion on the practical considerations for choosing an appropriate sorting algorithm based on specific application requirements.

Through this work, we hope to contribute to the ongoing research and development of efficient sorting techniques, ensuring that the best practices are employed in both academic and industrial settings.

Preliminaries

Sorting algorithms are critical to the efficiency of numerous computational tasks, making their study a cornerstone of computer science. This section provides the necessary background and preliminaries required to understand the sorting algorithms discussed in this paper.

0.1 Sorting Algorithms

A sorting algorithm is a method for rearranging a list or array of elements in a specific order, typically in ascending or descending numerical or lexicographical order.

0.2 Input Array

An array A of n elements, where $A = [a_1, a_2, \dots, a_n]$, is provided as the input to the sorting algorithm. The goal is to transform this array into a sorted array A' where $a'_1 \leq a'_2 \leq \dots \leq a'_n$.

0.3 Comparison-Based Sorting

Most traditional sorting algorithms operate by comparing elements to determine their order. The efficiency of these algorithms often depends on the number of comparisons made.

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0.4 Inversions

An inversion in an array is a pair of elements where the preceding element is greater than the following one. Inversions are used to measure how far an array is from being sorted. The number of inversions in an array can give insights into the efficiency of sorting algorithms, especially those that are comparison-based.

Formally, given an array A of n elements, an inversion is a pair of indices (i, j) such that $i < j$ and $A[i] > A[j]$.

0.5 Random Access

The input array is stored in a structure that allows random access, meaning any element a_i can be accessed in constant time $O(1)$.

0.6 Big O Notation

Big O notation, denoted as $O(f(n))$, is used to describe the upper bound of an algorithm's running time. It provides an asymptotic analysis of the algorithm's performance, allowing us to understand how the running time grows relative to the input size n . Formally, we say that a function $T(n)$ is $O(f(n))$ if there exist positive constants c and n_0 such that for all $n \geq n_0$, the following inequality holds:

$$T(n) \leq c \cdot f(n)$$

0.7 Time Complexity

Time complexity is a measure of the amount of time an algorithm takes to complete as a function of the size of the input. It helps in comparing the efficiency of different algorithms. Time complexities are typically classified into three categories: average case, best case, and worst case.

- **Best Case:** The scenario where the algorithm performs the minimum number of operations. For sorting algorithms, this often occurs when the input array is already sorted.
- **Average Case:** The expected time over all possible inputs of size n . This provides a realistic measure of the algorithm's performance in everyday use.
- **Worst Case:** The scenario where the algorithm performs the maximum number of operations. This case is crucial as it provides an upper bound on the running time, ensuring that the algorithm will not exceed this time for any input.

0.8 Space Complexity

Space complexity refers to the amount of memory an algorithm uses relative to the input size. It is crucial for understanding the overall resource requirements of an algorithm. Space complexity is often categorized into:

- **In-place:** Algorithms that use a constant amount of extra space, usually $O(1)$.
- **Not in-place:** Algorithms that require additional space proportional to the input size, typically $O(n)$ or more.

1 Insertion Sort

Insertion Sort is a straightforward and efficient comparison-based sorting algorithm. It is widely used for sorting small datasets or nearly sorted arrays due to its simplicity and ease of implementation. The algorithm operates similarly to the way humans sort playing cards, making it an intuitive method for understanding basic sorting principles.

Insertion Sort constructs the final sorted array one element at a time. It starts by assuming that the first element is already sorted. It then picks the next element and inserts it into the correct position relative to the already sorted elements. This process is repeated for each element until the entire array is sorted.

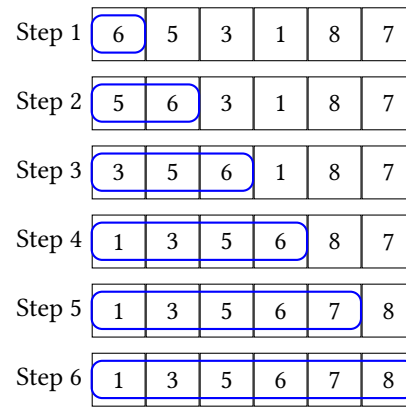


Figure 1. The operation of INSERTION-SORT on the array $A = \langle 6, 5, 3, 1, 8, 7 \rangle$.

1.1 Pseudocode

[H]

Insertion Sort	
1:	InsertionSort (A):
2:	for $j = 2$ to $\text{length}(A)$ do
3:	$\text{key} = A[j]$ \triangleright Insert $A[j]$ into the sorted sequence $A[1 \dots j-1]$
4:	$i = j - 1$
5:	while $i \geq 1$ and $A[i] > \text{key}$ do
6:	$A[i + 1] = A[i]$
7:	$i = i - 1$
8:	end while
9:	$A[i + 1] = \text{key}$
10:	end for

1.2 Time and Space Complexity

Time Complexity:

- **Best Case:** $O(n)$ - This occurs when the array is already sorted. The inner loop (line 5) is never executed.

- **Average Case:** $O(n^2)$ - On average, each insertion takes $O(n)$ time.
- **Worst Case:** $O(n^2)$ - This occurs when the array is sorted in reverse order. Each insertion requires shifting all previously sorted elements.

Space Complexity:

- $O(1)$ - Insertion Sort is an in-place sorting algorithm, meaning it requires only a constant amount of additional memory space beyond the input array.

1.3 Practical Applications

Insertion Sort is particularly useful when:

- The array is already mostly sorted, as it can operate in nearly linear time.
- The dataset is small, making the simplicity and low overhead of Insertion Sort advantageous over more complex algorithms.
- A stable sort is required, as Insertion Sort maintains the relative order of equal elements.

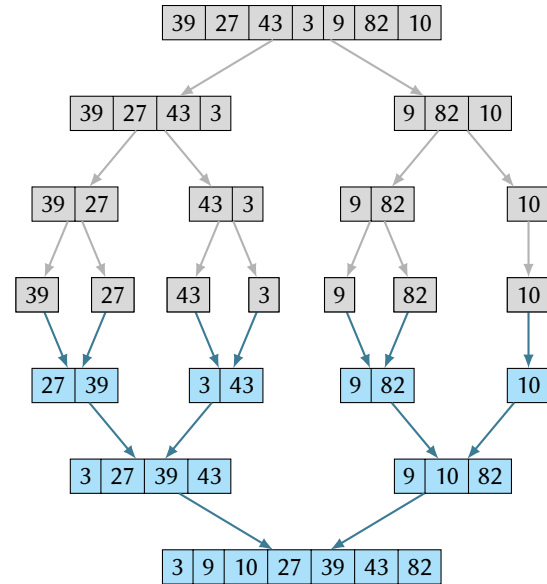
2 Merge Sort

Merge sort is a widely-used and highly efficient sorting algorithm that follows the divide-and-conquer paradigm. It was developed by John von Neumann in 1945 and is considered one of the most important and influential algorithms in computer science.

The basic idea behind merge sort is to recursively divide the input array into smaller subarrays until they are small enough to sort, and then merge these sorted subarrays back together to produce the final sorted array.

The algorithm works as follows:

- **1. Base Case:** If the input array has 0 or 1 elements, it is already sorted, so return it.
- **2. Divide:** Divide the input array into two halves, left and right.
- **3. Conquer:** Recursively sort the left and right halves using merge sort.
- **4. Merge:** Merge the two sorted halves into a single sorted array.



2.1 Pseudocode

Merge Sort

```

1: function MERGE SORT(array)
2:   if length(array) ≤ 1 then return array
3:   end if
4:   middleIndex ← ⌊length(array)/2⌋
5:   left ← array[0 : middleIndex - 1]
6:   right ← array[middleIndex : end]
7:   left ← MERGE SORT(left)
8:   right ← MERGE SORT(right)
9:   MERGE(left, right)
10: end function
10: function MERGE(left, right)
11:   result ← {}
12:   while left ≠ {} and right ≠ {} do
13:     if first(left) < first(right) then
14:       append(result, first(left))
15:       remove(left, first(left))
16:     else
17:       append(result, first(right))
18:       remove(right, first(right))
19:     end if
20:   end while
21:   while left ≠ {} do
22:     append(result, first(left))
23:     remove(left, first(left))
24:   end while
25:   while right ≠ {} do
26:     append(result, first(right))
27:     remove(right, first(right))
28:   end while
29:   return result

```

2.2 Time and Space Complexity

Time Complexity:

- The time complexity of merge sort in the best, average, and worst cases is $O(n \log n)$. This means that the time it takes to sort an array of size n using merge sort is proportional to n multiplied by the logarithm of n .

Space Complexity:

- The space complexity of merge sort is $O(n)$ in the worst case. This is because merge sort requires additional space to store the temporary arrays used during the merging phase. This extra space is proportional to the size of the input array. However, in practice, this additional space requirement makes merge sort less memory-efficient than some other sorting algorithms.

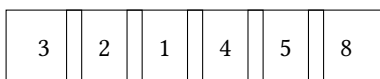
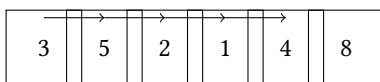
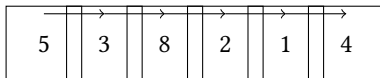
2.3 Practical Applications

Some of the practical applications of merge sort include:

- **External sorting:** Merge sort is often used for sorting large data sets that cannot fit entirely in memory. It is commonly used in external sorting algorithms, where data is sorted on external storage such as hard drives or SSDs.
- **Parallel processing:** Merge sort can be easily parallelized, making it suitable for parallel processing environments. Each subarray can be sorted independently, and then the sorted subarrays can be merged in parallel, improving overall performance.
- **Stable sorting:** Merge sort is a stable sorting algorithm, meaning that it preserves the relative order of equal elements. This property makes it suitable for applications where maintaining the original order of equal elements is important, such as in financial data processing and record-keeping systems.

3 Bubble Sort

Bubble sort is a simple comparison-based sorting algorithm that repeatedly steps through the list to be sorted, compares each pair of adjacent items, and swaps them if they are in the wrong order. The pass through the list is repeated until no swaps are needed, indicating that the list is sorted.



3.1 Pseudocode

Bubble Sort

```

1: procedure BUBBLESORT( $A$ )
2:    $n \leftarrow$  length of  $A$ 
3:   for  $i \leftarrow 0$  to  $n - 1$  do
4:     for  $j \leftarrow 0$  to  $n - i - 1$  do
5:       if  $A[j] > A[j + 1]$  then
6:         swap  $A[j]$  and  $A[j + 1]$ 
7:       end if
8:     end for
9:   end for
10: end procedure

```

3.2 Time and Space Complexity

Time Complexity:

- **Best Case:** $O(n)$ - when the input list is already sorted, and no swaps are needed.
- **Average Case:** $O(n^2)$ - on average, bubble sort requires $n^2/2$ comparisons and swaps.
- **Worst Case:** $O(n^2)$ - when the input list is sorted in reverse order, and each pair of elements needs to be swapped.

Space Complexity:

- $O(1)$ - Bubble sort has a space complexity of $O(1)$ as it only requires a constant amount of extra space for temporary variables during the swapping process.

3.3 Practical Applications

Bubble sort can be used in situations where simplicity and ease of implementation are more important than efficiency:

- It can be used for sorting small arrays or lists where the number of elements is relatively small.
- Bubble sort is a stable sorting algorithm, meaning that it preserves the relative order of equal elements. If stability is a requirement, bubble sort can be used in such cases.

4 Heap Sort

Heap sort is a comparison-based sorting algorithm. It divides its input into a sorted and an unsorted region, and it iteratively shrinks the unsorted region by extracting the largest element from it and inserting it into the sorted region. Heapsort does not waste time with a linear-time scan of the unsorted region; rather, heap sort maintains the unsorted region in a heap data structure to efficiently find the largest element in each step.

The heap Sort algorithm begins by rearranging the array into a binary max-heap. The algorithm then repeatedly swaps the root of the heap (the greatest element remaining in the heap) with its last element, which is then declared to be part of the sorted suffix. Then the heap, which was

damaged by replacing the root, is repaired so that the greatest element is again at the root. This repeats until only one value remains in the heap.

4.1 Pseudocode

Heap Sort

```

function HEAPSORT(arr, n)
  BUILDHEAP(arr, n)
  for i ← n to 2 do SWAP(arr[1], arr[i])
    n ← n − 1 HEAPIFY(arr, 1, n)
  end for
end function
function BUILDHEAP(arr, n)
  for i ← n/2 downto 1 do HEAPIFY(arr, i, n)
  end for
end function
function HEAPIFY(arr, i, n)
  largest ← i
  l ← 2i
  r ← 2i + 1
  if l ≤ n and arr[l] > arr[largest] then
    largest ← l
  end if
  if r ≤ n and arr[r] > arr[largest] then
    largest ← r
  end if
  if largest ≠ i then SWAP(arr[i], arr[largest])
    HEAPIFY(arr, largest, n)
  end if
end function
function SWAP(a, b)
  temp ← a
  a ← b
  b ← temp
end function

```

4.2 Time and Space Complexity

Time Complexity:

- Heap sort has a time complexity of $O(n \log n)$ for the best, average and worst case scenarios. This is because the heapify operation takes $O(\log n)$ time and it needs to be performed for each element in the array, resulting in a total time complexity of $O(n \log n)$.

Space Complexity:

- The space complexity of heap sort is $O(1)$ as it operates in place and does not require any additional space other than the input array

4.3 Practical Applications

Applications of Heap Sort include:

- Sorting large datasets: Heap sort is efficient for sorting large datasets due to its $O(n \log n)$ time complexity.

- External sorting: Heap sort can be used for external sorting where the data is too large to fit into memory and needs to be sorted using disk-based algorithms.

5 Quick Sort

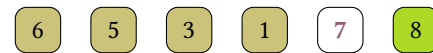
Quicksort is an efficient, general-purpose sorting algorithm. Quicksort is a divide-and-conquer algorithm. It works by selecting a 'pivot' element from the array and partitioning the other elements into two sub-arrays, according to whether they are less than or greater than the pivot. For this reason, it is sometimes called partition-exchange sort. The sub-arrays are then sorted recursively. This can be done in-place, requiring small additional amounts of memory to perform the sorting.



Step 1: Choose a pivot



Step 2: Lesser values go to the left, equal or greater values go to the right



Step 3: Repeat step 1 with the two sub lists



Step 4: Repeat step 2 with the sub lists:



Step 5: and again and again!

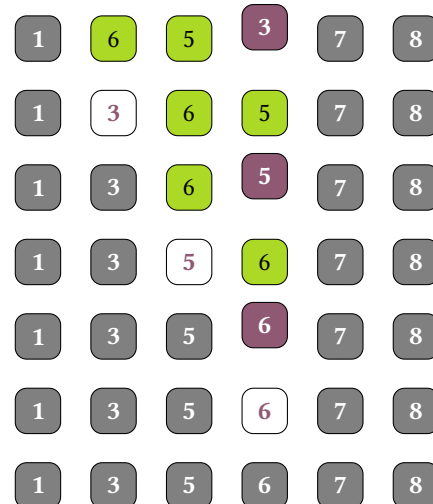


Figure 2. The operation of QUICK-SORT on the array $A = \langle 6, 5, 3, 1, 8, 7 \rangle$.

5.1 Pseudocode

Quick Sort	
1:	QuickSort(A):
2:	if $A.length < 1$ then
3:	choose a pivot
4:	while (there are items left in the array) do
5:	if $item < pivot$ then
6:	put item in subarray1
7:	else
8:	put item in subarray2
9:	end if
10:	end while
11:	QuickSort(subarray1)
12:	QuickSort(subarray2)
13:	end if

5.2 Time and Space Complexity

Time Complexity:

- **Best Case:** $O(n \log n)$ - The scenario occurs when the pivot divides the array into two nearly equal halves, leading to well-balanced partitioning.
- **Average Case:** $O(n \log n)$ - This scenario considers all possible cases and calculates the average time complexity. Moreover, it's assumed that all permutations of array element orders are equally likely.
- **Worst Case:** $O(n^2)$ - The worst-case scenario occurs when the smallest or largest element is chosen as a pivot. In this case, the partitioning is heavily unbalanced, leading to significantly higher time complexity.

Space Complexity:

- $O(\log n)$ - Quicksort has a space complexity of $O(\log n)$ in the average case. This arises from the recursive function calls and the partitioning process. It can be $O(n)$ due to an unbalanced partitioning leading to a deep recursion stack in the worst case.

5.3 Practical Applications

Quick Sort is particularly useful when:

- It is used everywhere where a stable sort is not needed.
- The sorting algorithm is used for information searching and as Quicksort is the fastest algorithm so it is widely used as a better way of searching.

6 Related Works

Sorting algorithms have been extensively studied, leading to various methods optimized for different data and computing environments. This section reviews key developments in Mergesort, Heapsort, Quicksort, Insertionsort, and Bubblesort, focusing on their theoretical foundations and practical implementations.

- **Insertionsort:** Insertionsort is a simple algorithm with $(O(n^2))$ average and worst-case time complexity but performs well with $(O(n))$ complexity for nearly sorted arrays. It is stable and efficient for small datasets and is often used as a subroutine in more complex algorithms (Knuth, 1997; Cormen et al., 2009).
- **Mergesort:** Introduced by John von Neumann in 1945, Mergesort is a divide-and-conquer algorithm with a time complexity of $(O(n \log n))$. It is stable and efficient for large datasets but has an $(O(n))$ space complexity, which can be a drawback (Cormen et al., 2009; Knuth, 1997).
- **Bubblesort:** Bubblesort, known for its simplicity, has an $(O(n^2))$ time complexity, making it impractical for large datasets. It is primarily used for educational purposes to illustrate basic sorting concepts (Knuth, 1997; Sedgewick, 2002).
- **Heapsort:** Developed by J. W. J. Williams in 1964, Heapsort transforms the array into a binary heap, operating in $(O(n \log n))$ time and $(O(1))$ space. It is efficient for memory-constrained environments but is less stable and often slower than Quicksort in practice (Williams, 1964; Sedgewick, 2002).
- **Quicksort:** Tony Hoare's Quicksort (1960) is renowned for its average-case $(O(n \log n))$ time complexity and high practical performance. Despite its worst-case $(O(n^2))$ complexity, optimizations like randomized partitioning enhance its efficiency (Hoare, 1962; Musser, 1997).
- **Comparative Analyses:** Studies comparing these algorithms, such as those by Bentley and McIlroy (1993) and Sedgewick (2002), provide insights into their performance and optimizations. Recent work by LaMarca and Ladner (1999) examines cache performance, highlighting the practical implications of algorithmic choices in modern computing.

7 Conclusion

Algorithms	Time Complexity	Space Complexity
Insertion sort	$O(n^2)$	$O(1)$
Merge sort	$O(n \log n)$	$O(n)$
Bubble sort	$O(n^2)$	$O(1)$
Heap sort	$O(n \log n)$	$O(1)$
Quick sort	$O(n \log n)$	$O(n)$

In conclusion, sorting algorithms play a fundamental role in computer science, serving as critical tools for organizing data efficiently. Each of the discussed algorithms (mergesort, heapsort, quicksort, insertionsort, and bubblesort) offers unique advantages and limitations that make them suitable for different scenarios and data sets.

Mergesort, with its stable and predictable $O(n \log n)$ performance, is ideal for large datasets and is well-suited for external sorting due to its consistent splitting and merging approach. Heapsort, also with $O(n \log n)$ complexity, is notable for its in-place sorting and good worst-case performance, making it valuable in memory-constrained environments.

Quicksort is often preferred in practice due to its average-case efficiency and cache-friendly nature, despite its $O(n^2)$ worst-case scenario, which can be mitigated with good pivot selection strategies like median-of-three. Insertionsort, while inefficient for large datasets with its $O(n^2)$ complexity, excels with small or nearly sorted arrays due to its simplicity and low overhead. Bubblesort, although largely impractical for serious applications due to its $O(n^2)$ complexity, serves as an educational tool for understanding basic algorithmic principles.

In practical applications, the choice of sorting algorithm is heavily influenced by the specific requirements of the task

at hand, including the size of the dataset, memory constraints, and the need for stability. Understanding the underlying mechanics and performance implications of each algorithm enables informed decisions that optimize both efficiency and resource utilization in software development.

By comprehensively evaluating these sorting algorithms, we appreciate the diverse strategies employed in data organization and the continuous evolution of algorithmic design, highlighting the significance of both theoretical knowledge and practical application in computer science.

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