

# Better Sorting Algorithms

Josh Wilcox (jw14g24@soton.ac.uk)

March 8, 2025

## Contents

<b>1</b>	<b>Correctness of Sorting Algorithms</b>	2
1.1	Proving correctness of insertion sort . . . . .	2
<b>2</b>	<b>Comparison Based sorting algorithms</b>	2
<b>3</b>	<b>Decision Trees</b>	2
3.1	Time complexity with decision trees . . . . .	2
<b>4</b>	<b>Lower bound on comparison based sorting</b>	2
4.1	Calculating the lower bound . . . . .	2
<b>5</b>	<b>Merge Sort</b>	3
5.1	Recursion Algorithm . . . . .	3
5.2	Merging Sub Arrays . . . . .	3
5.3	Properties of Merge Sort . . . . .	3
5.4	Improving Merge sort with Insertion Sort . . . . .	3
<b>6</b>	<b>Quick Sort</b>	4
6.1	High level idea . . . . .	4
6.2	Selecting a Pivot . . . . .	4
6.3	Time complexity . . . . .	4

# 1 Correctness of Sorting Algorithms

- A sorting algorithm is correct when:
  - Its output is in non-decreasing order
  - The items in the output are a permutation of the items in the input

## 1.1 Proving correctness of insertion sort

At the start of each iteration of the **for** loop, the sub-array consists of all items originally in the list but in sorted order. The for loop ends when the list is fully sorted

# 2 Comparison Based sorting algorithms

- A comparison based sorting algorithm can **only** gain information about the items in the input sequence by performing *pairwise comparisons*
  - **Pairwise Comparison** - A single query asking if  $a_i < a_j$  (between exactly two elements)

# 3 Decision Trees

- Decision trees are a way to visualise **many algorithms**
- Shows a series of decisions made during an algorithm
- For sorting algorithms:
  - Shows what the algorithm does at every comparison

## 3.1 Time complexity with decision trees

- The time taken to complete a task is the *depth* of the decision tree
- **Worst Case** - depth of deepest leaf
- **Best case** - depth of shallowest leaf
- **Average Case** - average depth of leaves

# 4 Lower bound on comparison based sorting

- Any sorting algorithm with *pairwise comparisons* must have a leaf in the decision tree **for every permutation of sorting the list**
  - Each leaf gives a different possible ordering of elements
  - There are  $n!$  possible permutations
- Remember the decision tree is binary (yes or no):

$$\text{Height of Decision Tree for Sorting} \geq \log_2(n!)$$

## 4.1 Calculating the lower bound

$$\begin{aligned}\log_2(n!) &= \log_2\left(1 \cdot 2 \cdot \dots \cdot \frac{n}{2} \cdot \dots \cdot n\right) \\ &= \log_2(1) + \dots + \log_2\left(\frac{n}{2}\right) + \dots + \log_2(n) \\ &\geq \log_2\left(\frac{n}{2}\right) + \dots + \log_2(n) \\ &\geq \frac{n}{2}\log_2\left(\frac{n}{2}\right) = \frac{n}{2}\log_2(n) - \frac{n}{2}\end{aligned}$$

$$\log_2(n!) = \Omega(n \log_2(n))$$

## 5 Merge Sort

### 5.1 Recursion Algorithm

```
1 public void MergeSort(a, start, end) {  
2     if (start < end){  
3         mid = floor((start + end) / 2) // Middle element index  
4         MergeSort(a, start, mid) // Left half  
5         MergeSort(a, mid+1, end) // Right half  
6         Merge(a, start, mid, end) // Merge the split up arrays  
7     }  
8 }
```

### 5.2 Merging Sub Arrays

- The merge operation takes two sorted sub-arrays and combines them into a single sorted array.
- Assume the two sub-arrays are `left` and `right`.
- Initialize three pointers: `i` for the `left` sub-array, `j` for the `right` sub-array, and `k` for the position in the merged array.
- Compare the elements at `left[i]` and `right[j]`:
  - If `left[i]` is smaller, place `left[i]` in the merged array at position `k` and increment `i` and `k`.
  - Otherwise, place `right[j]` in the merged array at position `k` and increment `j` and `k`.
- Repeat the comparison until one of the sub-arrays is exhausted.
- Copy any remaining elements from the non-exhausted sub-array into the merged array.

### 5.3 Properties of Merge Sort

- It is **stable**
- It is **not in-place**
- Merging is quick
  - At most  $n - 1$  comparisons to merge two subarrays
- Recurrence Relation:
$$T(n) = 2T\left(\frac{n}{2}\right) + \mathcal{O}(n)$$
  - Therefore **worst case** complexity:  $\mathcal{O}(n \log(n))$

### 5.4 Improving Merge sort with Insertion Sort

- Insertion sort is very efficient for short arrays
- Therefore, if a sub-array size falls below a certain threshold, it is better to switch to insertion sort to sort the array instead of splitting even more
- These can then be merged in the same ray

## 6 Quick Sort

### 6.1 High level idea

- Separate the array into two parts depending on whether the elements are smaller or greater than some **pivot**
- Recurse on both parts until the array is sorted

### 6.2 Selecting a Pivot

- Pivot is the best when the partitions it creates are of the same length
- Possible options:
  - Choosing first element in array
  - Choose the median of first, middle, and last element of array
    - \* Pivot is somewhat more likely to be the median of the whole array
    - \* Therefore, split in two parts more often
  - Choose the pivot randomly

### 6.3 Time complexity

- Partitioning takes  $\Theta(n)$  operations
- When the pivot element is the smallest in each partition, we would need  $n - 1$  partitioning rounds
  - This means partitioning is  $\mathcal{O}(n)$
- **Worst-Case complexity** -  $\mathcal{O}(n^2)$
- If the pivot is the median value, the array is split in half each time therefore we have  $\Omega(\log(n))$  partitions
  - Therefore on **average**, Quicksort is  $\mathcal{O}(n \log(n))$