

# Lazy sort analysis

## 1. Implementation Analysis:

### Merge sort:

I implemented threading by assigning each half of the array at every step a new thread. Pros and cons of this approach are:

Pros:

- a. Efficient utilization of resources
- b. Scalability: The algorithm scales well as the input size increases because more threads can be created to handle increasingly smaller sections in parallel.

Cons:

- a. High overhead for small data sets. The overhead for handling threads introduces more overhead than benefit from parallelization.
- b. Increased memory usage.
- c. Thread limitations. Some systems have a limit on the number of threads, which might cause problems for large data sets.

### Count sort:

I implemented my distributed count sort by taking inspiration from this site.

<https://stackoverflow.com/questions/39903181/can-you-do-a-parallel-counting-sort-in-on-p-time>

Pros and cons of this approach are:

Pros:

- a. Parallel efficiency. For large datasets, dividing the array into  $p$  parts significantly reduces runtime.
- b. Scalability. This approach works well with large datasets, because of the decrease of time by a factor  $p$ .

Cons:

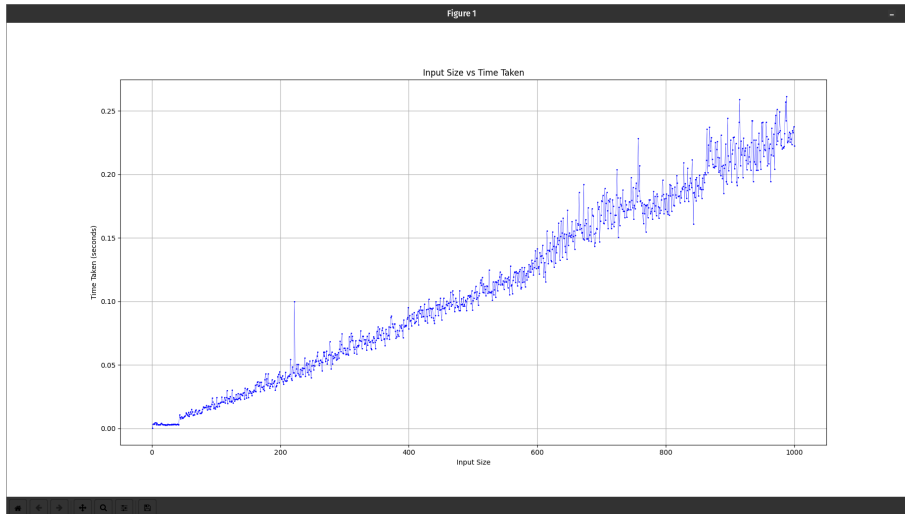
- a. High memory overhead.
- b. Limited benefit for small datasets.
- c. Assumes uniform distribution of work among  $p$  parts of array.

### Assumption:

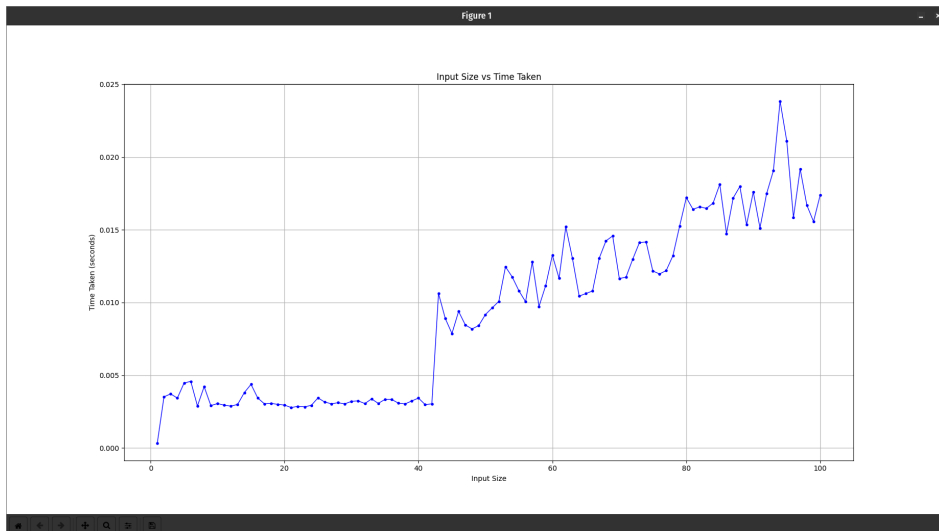
In count sort for names, I am taking only the first four characters for hashing. Because taking more than that makes the hash value too big.

# Lazy sort analysis

## 2. Time analysis:



Large dataset (n = 1 to 1000)



Small dataset (n=1 to 100)

Observation: From  $n = 42$  to  $n = 43$  there is a steep increase in the time because count sort is much more efficient than merge sort.

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