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# Review paper on Merge Sort

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**—This paper aims at introducing a new sorting al- gorithm which sorts the elements of an array In Place. This algorithm has ( ) best case Time Complexity and ( log ) average and worst case Time Complexity. We achieve our goal using Recursive Partitioning combined with In Place merging to sort a given array. A comparison is made between this particular idea and other popular implementations. We finally draw out a conclusion and observe the cases where this outperforms other sorting algorithms. We also look at its shortcomings and list the scope for future improvements that could be made.**

#### Keywords-Time In Place, **Complexity, Recursive Partitioning**

#### **INTRODUCTION**

Jyrki Katajainen, Tomi Pasanen and Jukka Teuhola[4], in the year 1996 explained the uses and performance analysis of an In Place Merge Sort algorithm. Initially, a straightforward variant was applied with  $O(n \log 2n) + O(n)$  comparisons and 3(*n*)  $log 2n$ ) +  $O(n)$  moves. Later, a more advanced variant was introduced which required at most (n  $log 2n$ ) +  $O(n)$  comparisons and (*n* log 2*n*) moves, for any fixed array of size 'n'.

for their execution.

In this paper, we introduce a new algorithm which uses the concept of divide and conquer to sort the array recursively using bottom up approach. Instead of using an external array to merge the two sorted sub arrays, we use multiple pivots to keep track of the minimum element of both the sub arrays and sort it In Place.

In mathematics and computer science, the process of ar- ranging similar elements in a definite order is known as sorting. Sorting is not a new term in computing. It finds its significance in various day to day applications and forms the backbone of computational problem solving. From complex search engine algorithms to stock markets, sorting has an impeccable presence in this modern day era of information technology. Efficient sorting also leads in optimization of many other complex problems. Algorithms related to sorting have always attracted a great deal of Computer Scientists and Mathematicians. Due to the simplicity of the problem and the need for solving it more systematically, more and more sorting algorithms are being devised to suit the purpose. There are many factors on which the performance of a sorting algorithm depends, varying from code complexity to effective memory usage. No single algorithm covers all aspects of efficiency at once. Hence, we use different algorithms under different constraints. When we look on developing a new algorithm, it is impor- tant for us to understand how long might the algorithm take to run. It is known that the time for any algorithm to execute depends on the size of the input data. In order to analyze the efficiency of an algorithm, we try to find a relationship on it's time dependence with the amount of data given. Another factor to take into consideration is the space used up by the code with respect to the input. Algorithms that need constant minimum extra space are called In Place. They are generally preferred over algorithms that take extra memory space ISBN No.978-1-4673-9545-8

Rest of the paper is organized as follows. Section II discusses the various references used in making this paper. Section III describes the basic working idea behind this algorithm. Section IV contains the pseudo code required for the implementation of this algorithm. In Section V, we do a Case Study on the merging process over an array. In Section VI, we derive the time and space complexities of our code. In Section VII, we do an experimental analysis of this algorithm on arrays of varying sizes. In Section VIII ,Advantages of merge sort In section IX , Disadvantages of merge sort .

> In section X ,Disawe draw out asymptotic conclusion based on Section VI and VII. We finally list out the scope for future improvements and conclude the paper in Section XI.

#### II. LITERATURE SURVEY

You Ying, Ping You and Yan Gan[2], in the year 2011 made a comparison between the 5 major types of sorting algorithms. They came to a conclusion that Insertion or Selection sort performs well for small range of elements. It was also noted that Bubble or Insertion sort should be preferred for ordered set of elements. Finally, for large random input parameters, Quick or Merge sort outperforms other sorting algorithms.

Wang Xiang[7], in the year 2011 presented a brief analysis of the performance measure of Quick

Sort algorithm this paper discusses about the TimeComplexity of Quick Sort algorithm and makes a comparison between the improved Bubble Sort and Quick Sort through analysing the first order derivative of the function that is found to co-relate Quick Sort with other sorting algorithms.

Rohit Yadav, Kratika Varshney and Nitin Verma[20], in the year 2013 discussed the run time complexities of the recursive and non recursive approach of the merge sort algorithm using a simple unit cost model. New implementations the for two way and four way bottom-up merge sort were given, the worst case complexities of which were shown to be bounded by 5.5n log  $2n + O(n)$ and 3.25n log  $2n + O(n)$ , respectively.

Shrinu Kushagra, Alejandro Lopez-Ortiz and J. Ian Munro [8], in 2013, presented a new approach which consisted of multiple pivots in order to sort elements. They performed an experimental study and also provided analysis on cache behavior of these algorithms. Here, they proposed a 3 pivot mechanism for sorting and improved the performance by 7-8%.

### III. METHODOLOGY

#### \*ar = pointer to the array  $i =$  starting point of the first sub array  $j =$  ending point of the second sub array

We use 4 pivots 'a', 'x', 'y' and 'b' in the code to , , accomplish our task. 'a' and 'b' initially mark starting points of the two sorted sub arrays respectively. As a result, 'a' is initialized to 'i' and 'b' is obtained by dividing the sum of 'i' and 'j' by two and incrementing it. 'x' is the point below which our final array is sorted and is initialized to  $i'$ . 'y' is an intermediate pivot, which marks the bound for pivot 'a' and is initialized to 'b'. All in all, our function is targeted at sorting the main array 'ar' from position 'i' to 'j' given that the elements from 'i' to 'b-1' and from 'b' to 'j' are already sorted.

#### 3.1 DIVIDE AND CONQUER

In this particular section, we lay emphasis on the idea behind the working of this algorithm. The proposed algorithm solves our problem in two steps, the strategies behind which are stated below.

We use the Divide and Conquer strategy to split the given array into individual elements. Starting from individual elements, we sort the array using Bottom Up Approach, keeping track of the minimum and maximum value of the sub arrays at all times. The technique used for splitting the array is similar to that of a standard Merge Sort, where we recursively partition the array from start to mid, and from mid to last after which, we call the sort function to sort that particular sub array.

**3.2 PIVOT BASED MERGING** 

The variable 'a' is used for keeping track of the minimum value in the first sub array that has not yet been accessed (for most of the time, barring a few passes). Similarly, 'b' is used for keeping track of the minimum value in the second sub array that has not yet been accessed (again, for most of the time, barring a few passes). As mentioned earlier,  $x'$  is the point before which our final array is sorted. So at any point, the array from  $i'$  to 'x-1' is sorted. Finally, we have another variable called 'ctr', which is initialized and always kept equal to 'b' till the second sub array (from 'b' to 'j') is sorted. If not, we keep on incrementing 'ctr' and swap it with its next value until the element at 'ctr' gets placed in its correct position and the second sub array becomes sorted once again. We then make 'ctr' equal to 'b'.



Fig. 1. A recursive algorithm to split and sort array elements

Our logic revolves around comparing the current minimum values in the two sorted sub arrays (values at ' $a'$  and ' $b'$ ), and swapping the smaller

This is the part from which this algorithm differs from a standard Merge Sort. Instead of merging the two sorted sub arrays in a different array, we use multiple pivots to sort them In Place and save the extra space consumed. Our function prototype to sort the array looks somewhat like this:

Procedure sort (int  $*ar$ , int *i*, int *j*)



In this Case Study, we take a look into the merging process of the two sorted sub arrays. Let us consider an array of size 18 elements for the sake of this example. The two sorted sub arrays are from '*i'* to '*b*-1' and from '*b'* to '*j'*.



 $a r[b] < a r[a]$  then



# 59: end while

#### V. CASE STUDY

#### INITIAL ARRAY:



This is the first pass inside the 'while' loop of our' sort' procedure. As mentioned, we compare the current minimum values of the two sub arrays (value at 'a'  $(-4)$  and 'b'  $(-3)$ ). The value at 'a' is less than that at 'b'. Since 'a' is equal to 'x', we don't need to swap the values at 'a' and 'x'. Instead, we increment 'a' and 'x'. The first element  $(-4)$  is now in its correct position. 'a' holds the minimum value of first sub array that has not yet been accessed (-1) and the array before  $x'$  is sorted.

if  $a = x$  and  $b = y$  and  $ar[b] \geq ar[a]$  then  $x \leftarrow x + 1$  and  $a \leftarrow a + 1$  end if ctr  $-3$  $-2$  $-2$  $5<sup>5</sup>$  $\bullet$ 8  $16$ 24  $32<sub>2</sub>$  $5<sub>5</sub>$  $10$  12 14  $a, x$  $_{\rm b,y}$ 

In this pass, the value at 'b' is less than that at 'a'. So we swap the value at 'b' with the value at 'x'. 'x' is incremented accordingly. The second element (-3) is now in its correct sorted position. We reassign 'a' as 'b' and increment 'b'. 'b' now contains the current minimum value of the second sub array (-2) and 'a' keeps track of its previously pointed value (-1).

if  $a = x$  and  $b = y$  and  $ar[b] < ar[a]$  then swap ( $ar[x]$ ,  $ar[b]$ )  $a \leftarrow b$ ,  $b \leftarrow b + 1$ , x  $\leftarrow$  x + 1 if ctr = b - 1 then  $ctr \leftarrow ctr + 1$ end if end if



#### PASS 2:

Our motto behind each pass is to assign 'a' and '  $b'$  such that they contain the current minimum values of the two sub arrays (This condition is true for all but few passes (discussed in Pass 7)).

```
if b = a + 1 and a \cdot [b] < a \cdot [a] then
swap ( ar[b], ar[x]) swap ( ar[a],
ar[b]) b \leftarrow b+1, x \leftarrow x+1, a \leftarrowa + 1 if ctr = b - 1 then ctr \leftarrow ctr +1 end if end if
```


```
if b = a + 1 and ar[b] < ar[a] then swap (
ar[b], ar[x]) swap (ar[a], ar[b]) b \leftarrow b + 1,
x \leftarrow x + 1, a \leftarrow a + 1 if ctr = b - 1 then ctr
\leftarrow ctr + 1
end if
end if
```


if  $b = a + 1$  and  $a \cdot [b] \ge a \cdot [a]$  then swap (  $ar[x]$  ,  $ar[a]$  )  $a \leftarrow y$  ,  $x \leftarrow x +$ 1 end if



Till now, we had assumed that the value of the variable ' $ctr'$  to be equal to 'b'. This was only because  $ar[ctr]$  was less than or equal to  $ar[ctr+1]$  i. e. the array starting from 'b' was sorted. However, to preserve the two conditions stated in Pass 6, we make a swap that costs us the order of the two sub arrays. We solve this dilemma by swapping  $ar[ctr]$  with  $ar[ctr+1]$  and increment ' $ctr'$ . We keep on doing this until  $ar[ctr]$  becomes less than or equal to ar[ctr+1]. After this, 'ctr' once again is made equal to ' $b'$  (PASS 8).

#### PASS 3:

if  $a > y$  and  $b > a + 1$  and  $ar[b] < ar[a]$  and  $ctr = b$  then swap (  $ar[a]$ ,  $ar[b]$ ) swap ( $ar[a]$ ,  $ar[x]$ )  $x \leftarrow x+1$ ,  $a \leftarrow a+1$ if  $a$  $[ctr]$  >  $a$  $[ctr + 1]$  then swap ( $a$  $f$  $[ctr]$ ,  $a$  $[ctr + 1]$ )  $ctr \leftarrow ctr + 1$  end if end if

```
if b = a + 1 and ar[b] < ar[a] then swap (ar[b],
ar[x]) swap (ar[a], ar[b]) b \leftarrow b + 1, x \leftarrow x + 1, a
 \leftarrow a + 1 if ctr = b - 1 then ctr \leftarrow ctr + 1
end if end if
```


#### PASS 4:

if  $b = a + 1$  and  $ar[b] < ar[a]$  then swap ( $ar[b]$ ,  $ar[x]$ ) swap ( $ar[a]$ ,  $ar[b]$ )  $b \leftarrow b + 1$ ,  $x \leftarrow x + 1$ , a  $\leftarrow$  a + 1 if ctr = b - 1 then ctr  $\leftarrow$  ctr + 1 end if end if PASS 11:

if  $x = y$  then  $y \leftarrow a$ end if



#### PASS 5:

#### PASS 6:

if  $b > a + 1$  and  $ar[b] \ge ar[a]$  then swap ( ar[x], ar[a] )  $x \leftarrow x + 1$  , a  $\leftarrow$  $a + 1$  end if



#### FEW NOTES:

It is noticeable up till now that our aim has been to keep elements from 'a' to ' $b - 1'$  and elements from 'y' to 'a - 1' sorted (for  $a \ge y$ ).

2. Another thing worth observing is that elements from 'a' to ' $b-1$ ' are less than the elements from 'y' to 'a-1', (provided  $a \ge y$ ). This means that the first sub array can be accessed in sorted order from 'a' to 'b-1' and then from 'y' to 'a-1'. PASS 7:



PASS 9:

#### PASS 10:

#### PASS 12:

if  $b = a + 1$  and  $ar[b] \ge ar[a]$  then swap (  $ar[x]$ ,  $ar[a]$ )  $a \leftarrow y$ ,  $x \leftarrow x + 1$  end if



#### PASS 13:

if  $b = a + 1$  and  $ar[b] \ge ar[a]$  then swap (  $ar[x]$ ,  $ar[a]$ )  $a \leftarrow y$ ,  $x \leftarrow x + 1$  end if



PASS 14:



Since the value of 'b' is greater than 'j', it has gone out of bounds. Hence, we re- initialize the values of our pivots accordingly.

if  $b > j$  and  $a > x$  and  $b = a + 1$  and  $a > y$  and  $ctr = b$  then

 $b \leftarrow a$ , ctr  $\leftarrow b$ , a  $\leftarrow$  y end

if 8 8 10 12 14 24 32 16  $5 \mid 5 \mid 6 \mid 7$  $-3$   $-2$   $-2$   $-1$  $3<sup>1</sup>$ x, y b PASS 20: if  $a = x$  and  $x = y$  and  $ctr = b$  then  $\overline{b}$  $y \leftarrow$ end if  $ctr$ 

#### $5 \mid 6 \mid 7$ 8 8 10 12  $\sqrt{3}$  $-2$   $-1$  $5$  $3<sup>1</sup>$  $14$ 24  $32<sup>2</sup>$ 16 y,b

end if  $3 \mid 5 \mid 5 \mid 6 \mid 7 \mid 8 \mid 8 \mid 10 \mid 12 \mid 32$  $3$   $-2$   $-2$   $-1$  $16<sup>1</sup>$ x y a b PASS 18: if  $b = a + 1$  and  $ar[b] < ar[a]$  then swap  $($  ar[b], ar[x]  $)$  swap  $($  ar[a], ar[b]  $)$   $b \leftarrow b$  $+ 1$ ,  $x \leftarrow x + 1$ ,  $a \leftarrow a + 1$  if ctr = b - 1 then  $ctr \leftarrow ctr + 1$ end if end if

ctr

ctr

16

8 8 10 12 14 24 32  $6 \mid 7$  $-3$  $-2$   $-2$  $5|5|$  $\mathbf{3}$ x, y

PASS 23: if  $a = x$  and  $b = y$  and  $ar[b] < ar[a]$ then swap (  $ar[x]$  ,  $ar[b]$  )  $a \leftarrow b$ ,  $b \leftarrow b$  $+ 1$ ,  $x \leftarrow x + 1$ if  $ctr = b - 1$  then  $ctr \leftarrow 1$ end if end if  $-4$  $-3$  $-2$  $\mathbf{2}$ 5 5  $6\phantom{a}$  $\overline{7}$  $\mathbf{8}$  $\bf{8}$ 10 12 14 16 24 32  $\mathbf{3}$ 

PASS 19:

PASS 21:



#### PASS 22:

In the previous pass, 'b' and 'ctr' have again gone out of bounds. This condition is similar to that of Pass 19 but with different side condition. if  $b > j$  and  $a > x$  and  $ctr = b$  and  $a = y$  then  $b \leftarrow y$ , ctr  $\leftarrow$  b, a  $\leftarrow$  x end if



It took us about 23 passes to do an In Place merge on 18 elements. Although in code, it would have taken less iterations since multiple conditions can be evaluated at the same time. This more or less covers our sorting logic.

### VI. COMPLEXITY ANALYSIS

In this section, we analyze the time and space complexity for this algorithm's best and worst case scenarios.

#### **6.1 TIME COMPLEXITY**

### 6.1.1 WORST CASE

We saw in previous sections that our code structure was similar to the following:

- 1. **Procedure**  $split(int * ar, int i, int j)$ : 2. **if**  $j = i + 1$  or  $j = i$  then **if**  $ar[i] > ar[j]$  then 3.  $swap ( ar[j], ar[i])$ 4. **end if** 5. return 6. **else**
- 7.  $mid = (i + j)/2$ split (ar, i, mid) split (ar, mid+1, j)
- 8. **if**  $ar$ [ $mid$  + 1] <  $ar$ [ $mid$ ] **then**
- 9. sort (ar, i, j)
- 10. **end if end if**
- 1.  $\therefore$  **Procedure** sort (int  $*$  ar, int *i*, int *j*):
- 2.  $while x < b$  do
- 3. // Sorting logic (multiple if-else statements)
- 4. **end while**

Our algorithm starts its execution in the 'split' procedure. Let 'C1' be the constant time taken to execute the 'if' condition in this procedure and 'C2' be the constant time taken to execute the 'else' condition. Inside the 'else' condition, we recursively call the same function twice for size  $n/2'$ . We then call the 'sort' procedure.

Our 'sort' procedure comprises of a 'while' loop that has the logic for merging the two sorted arrays into a single sorted array. Let 'C3' be the constant time taken to execute this procedure. Our overall equation for time complexity becomes:

We use Recurrence Relation to find out the time complexity for the code. For a large input size " $n$ ", , the above equation for calculating the Time Complexity  $T(n)$  can be simplified as:

$$
T(n) = \begin{cases} C_1, & \text{if } n \leq 2, \\ 2T(n/2) + n + C_2 + C_3 & \text{otherwise} \end{cases}
$$

 $T(n) = 2T(n/2) + n + C_2 + C_3$ 

For **k** iterations the equation for T(n) becomes:  $T(n) = 2kT(n/2k) + kn + (2k - 1)C_2 + (2k - 1)C_3$ 

$$
T(n) = 2[2T(n/4) + n/2 + C_2 + C_3] + n + C_2 + C_3T(n) =
$$

 $4T(n/4) + 2n + 3C_2 + 3C_3$ 

 $T(n) = 4[2T(n/8) + n/4 + C<sub>2</sub>+ C<sub>3</sub>] + 2n + 3C<sub>2</sub>+ 3C<sub>3</sub> T(n) = 8T$  $(n/8) + 3n + 7C<sub>2</sub>+7C<sub>3</sub>$ 

Similar to the previous condition, the base condition for recursion occurs when " $n$ " is equal to 2 i.e. " $T(2)$ ". This implies:  $n/2k = 2 k$  $=$ log<sub>2</sub>  $n - 1$ 

The base condition for recursion in our algorithm

 $T(n) = 2kT(n/2k) + (2k - 1)C2$ 

This also means that the recursion runs up to log2  $n -1$  times before reaching its base condition. Substituting this value of " $k$ " in the above equation, we get:

 $T(n) = 2\log 2 n - 1T (n/(2\log 2 n-1)) + (\log 2 n-1)n$ 

 $+(2log2 n-1-1)C2+(2log2 n-1-1)C3$ 

We know,  $n/(2\log 2 n - 1) = 2$  Substituting

this and the value of  $T(2)$ , we get:

 $T(n) = (n/2)C1 + n(\log 2 n) - n + (n/2 - 1)C2 + (n/2 - 1)C3$ 

#### 6.1.2 BEST CASE

The efficiency of this sorting algorithm is directly proportional to the orderliness in the given array. As a result, the best case of this algorithm occurs when the array is already sorted (or even almost sorted). Let us consider the sorted array given as an example to find out the time taken by this algorithm in its best case:

 $T(n) = 4[2T(n/8) + C_2] + 3C_2$ 

 $T(n) = 8T(n/8) + 7C_2$ 

For " $k$ " iterations the equation for  $T(n)$  becomes:

Again, this means that the recursion runs up to log<sup>2</sup>  $n$  -1 times before reaching its base condition. Substituting this value of " $k$ " in the above equation, we get:

 $T(n) = 2\log_2 n - 1 T(n/(2\log_2 n - 1)) + (2\log_2 n - 1 - 1)C_2$  $T(n) = (n/2)C_1 + (n/2 - 1)C_2$ 

**6.2 SPACE COMPLEXITY** 

This is an In Place sorting algorithm and takes constant amount of memory for sorting a particular array. This property is quite important for any algorithm since it results in almost nonexistent computational space in the memory. In some cases, this is even considered more important than an algorithm"s Time Complexity.

### **6.3 STABILITY**

Instability is a major drawback in this sorting algorithm. Due to this, similar elements are not evaluated as distinct and lose their order as a result.

This issue can be sorted out by increasing the number of pivots and treating similar elements as distinct, but the implementation becomes way too complicated and is beyond the scope of this paper.

For a sorted array however, the stability is maintained since no swaps are being made.

occurs when " $n$ " is equal to 2 i.e.  $T(2)$ . This implies:  $n/2k = 2$  k=  $log_2 n -1$ 

#### VIII EXPERIMENTAL ANALYSIS

We evaluated the performance of the code for array inputs up to 32,000 elements by denoting the time taken to sort the elements.

7.1 WORST CASE

The worst case scenario in this algorithm occurs The Best Case scenario happens when the when each element in the input array is distinct and array is completely sorted or has similar elements. Since we know that the Time there is no order in the array whatsoever. We noticed that for large values of input elements, this Complexity for this condition is  $O(n)$ , we only algorithm performs slower than the Standard compare this algorithm with those having the Merge Sort and Quick Sort. However, for array elements up to 1000, this algorithm is faster than If the array is already sorted, this would imply that the starting element of the second sub array would be greater than the ending element of first sub array i.e. m Hence the program would not  $m$ even enter the s procedure. This means that our time complexity for merging the two sorted sub arrays is constant. Hence, the time taken to split the array is the only factor affecting the total Time Complexity. As a result, our overall equation becomes:  $T(n) = 2T(n/2) + C$ 

We consider the average case to be an array with partial order in it.

both Merge and Quick Sort even in its worst case.

# 7.1 BEST CASE

similar Time Complexities (Bubble and Insertion sort). Also, since the time difference between them was very small and non-comparable, we compared the 3 algorithms with respect to the number of iterations taken to sort the array. We noticed that this algorithm performs better than Bubble Sort but is slightly slower than Insertion Sort.





### VIII ASYMPTOTIC ANALYSIS

In this section, based on experimental analysis and previously stated proof, we draw out an asymptotic analysis of our algorithm.



# **7.1 AVERAGE CASE**

=

- IX Advantages[17]
	- Stability. o
	- **Consistent performance.** o
	- Parallelizability. o
	- **Efficient for linked list.** o
	- **External sorting.** o
	- Minimal comparisons. o
	- Divide and conquer approach. o
- X Disadvantages[18]
	- o Space complexity.
	- Not in-place. o
	- Complexity in iterative o *implementation.*
	- Slower on small datasets. o
	- Not adaptive. o

# XI CONCLUSION

This idea, like most standard algorithms has room for improvement. During our implementation phase, we noticed that the code slows down for very large values of  $'n'$  The instability of the algorithm is also a cause of concern.

Future improvements can be made to enhance the performance over larger number of input array. Since we have the minimum and maximum value of the sub array at any time, instead of starting from the beginning, we can combine the current logic with an end first search to reduce the number of iterations. Regarding its stability, as mentioned earlier, this algorithm can be made stable by increasing the number of pivots but this would lead to other complications. Any improvement though, however trivial, would be highly appreciated.

- [1] Dr. D. E. Knuth. "Sorting and Searching", The Art of **Computer** Programming, 3rd volume, second edition
- [2] You Yang, Ping Yu and Yan Gan. "Experimental Study on
- the Five Sort Algorithms", International Conference on Mechanic Automation and Control Engineering (MACE), 2011
- [3] W. A. Martin. "Sorting", ACM Comp Survey., 3(4):147-174, 1971
- [4] Jyrki Katajainen, Tomi Pasanen and Jukka Teuhola. " Practical in-place mergesort", Nordic Journal of Computing Archive Volume 3 Issue 1, 1996
- [5] R. Cole. "Parallel Merge Sort," Proc. 27th IEEE Symp. FOCS, pp. 511516, 1988
- [6] Wang Xiang. " Analysis of the Time Complexity of Quick Sort Algo- rithm", Information Management, Innovation

Management and Indus- trial Engineering (ICIII), International Conference, 2011

- [7] Shrinu Kushagra, Alejandro Lopez, J. Ian Munro and Aurick Qiao "Multi-Pivot Quicksort: Theory and Experiments", Proceedings of the 16th Meeting on AlgorithmEngineering and Experiments (ALENEX), pp. 47- 60, 2014.
- [8] L. T. Pardo. "Stable sorting and merging with optimal space and time bounds", SIAM Journal on Computing, 6(2):351–372, 1 9 7 7 .
- [9] Jeffrey Ullman, John Hopcroft and Alfred Aho. " The Design and Analysisof Computer Algorithms", 1974.
- [10] E.Horowitz and S.Sahni. "Fundamentals of Data Structures", Computer Science Press, Rockville, 1976
- [11] A.Symvonis. "Optimal stable merging", Computer Journal, 38:681–690, 1995
- [12] F. K. Hwang and S. Lin. "A Simple algorithm for merging two disjoint linear ordered sets", SIAM Journal on Computing, 1972
- [13] Hovarth, E. C. "Stable sorting in asymptotically optimal time and extra space", Journal of the ACM 177-199, 1978
- [14] S.Dvorak and B.Durian. "Stable linear time sub linear space merging", The Computer Journal 30 372-375, 1987
- [15] J.Chen. "Optimizing stable in-place merging", Theoretical Computer Science, 302(1/3):191–210, 2003.
- [16] Rohit Yadav, Kratika Varshney and Nitin Verma. "Analysis of Recursive and Non-Recursive Merge Sort Algorithm", International Journal of Advanced Research in Computer Science and

Software Engineering Volume 3, Issue 11, November 2013

[17] advantages of merge sort, chatgpt.

[18] disadvantages of merge sort, chatgpt.

#### **REFERENCES**