Ministry of Education, Culture and Research of the Republic of Moldova

Technical University of Moldova

Department of Software and Automation Engineering

REPORT

Laboratory work No. 2
Discipline: Algorithms' Analysis
Topic: Study and empirical analysis of sorting algorithms

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Chişinău – 2023

Algorithm analysis

Objective:

Study and analyze different sorting algorithms

Tasks:

- 1. Implement 4 sorting algorithms
- 2. Decide properties of input format that will be used for algorithm analysis;
- 3. Decide the comparison metric for the algorithms;
- 4. Analyze empirically the algorithms;
- 5. Present the results of the obtained data;
- 6. Deduce conclusions of the laboratory.

Theoretical Notes:

An alternative to mathematical analysis of complexity is empirical analysis.

This may be useful for: obtaining preliminary information on the complexity class of an algorithm; comparing the efficiency of two (or more) algorithms for solving the same problems; comparing the efficiency of several implementations of the same algorithm; obtaining information on the efficiency of implementing an algorithm on a particular computer.

Introduction:

Sorting refers to arranging data in a particular format. Sorting algorithm specifies the way to arrange data in a particular order. Most common orders are in numerical or lexicographical order.

Comparison metric:

The comparison metric for this laboratory work will be considered the time of execution of each algorithm (T(n))

Input format:

As input, each algorithm will receive a random array of 100000 numbers

IMPLEMENTATION

All four algorithms will be implemented in python an analyzed empirically based on the time required for their completion. The particular efficiency in rapport with input will vary depending on memory of the device used.

Quick Sort

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.

Quicksort is a comparison sort, meaning that it can sort items of any type for which a "less-than" relation (formally, a total order) is defined. Most implementations of quicksort are not stable, meaning that the relative order of equal sort items is not preserved.

Mathematical analysis of quicksort shows that, on average, the algorithm takes $O(n \log_n)$ comparisons to sort n items. In the worst case, it makes $O(n^2)$ comparisons.

Implementation:

```
def partition(array, low, high):
   pivot = array[high]
   for j in range(low, high):
       if array[j] <= pivot:</pre>
            (array[i], array[j]) = (array[j], array[i])
   (array[i + 1], array[high]) = (array[high], array[i + 1])
def quickSort(array, low, high):
       pi = partition(array, low, high)
```

Figure 1. Quick sort implementation

Results:

	array length	execution 1	time	(milliseconds)
Θ	10000			158.717632
1	20000			131.485462
2	40000			338.089228
3	80000			1161.364555
4	160000			1941.892147
5	320000			4552.301645
6	640000			14566.983938
7	1000000			30666.592121

Figure 2. Quick sort result table

The plot:

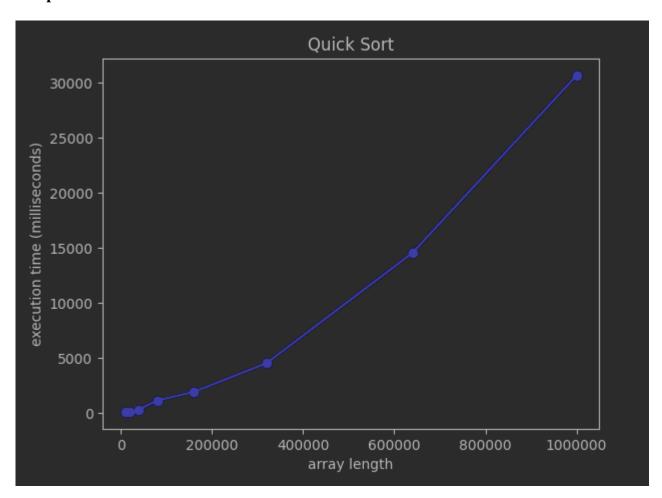


Figure 3. Quick sort plot result

Space complexity: O(1)

Best case: O (n*logn)

Worst case: O (n**2)

Stable: No

Merge Sort

Merge sort is a sorting algorithm that works by dividing an array into smaller subarrays, sorting each subarray, and then merging the sorted subarrays back together to form the final sorted array.

Conceptually, a merge sort works as follows:

- 1. Divide the unsorted list into n sublists, each containing one element (a list of one element is considered sorted).
- 2. Repeatedly merge sublists to produce new sorted sublists until there is only one sublist remaining. This will be the sorted list.

Imp	lement	tation:
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```
def merge(arr, l, m, r):
   L = [0] * (n1)
   R = [0] * (n2)
   for i in range(0, n1):
       L[i] = arr[l + i]
   for j in range(0, n2):
       R[j] = arr[m + 1 + j]
        if L[i] <= R[j]:</pre>
            arr[k] = L[i]
            arr[k] = R[j]
        arr[k] = L[i]
       arr[k] = R[j]
```

Figure 4. Merge sort implementation 1

```
# Same as (l+r)//2, but avoids overflow for large l and h
m = l+(r-l)//2

# Sort first and second halves
mergeSort(arr, l, m)
mergeSort(arr, m+1, r)
merge(arr, l, m, r)
```

Figure 5. Merge sort implementation 2

Results:

array length	execution	time	(milliseconds)
10000			155.855417
20000			170.241833
40000			239.865303
80000			739.741325
160000			1385.490179
320000			2778.556585
640000			4769.016981
1000000			7761.406898
	10000 20000 40000 80000 160000 320000 640000	10000 20000 40000 80000 160000 320000 640000	10000 20000 40000 80000 160000 320000 640000

Figure 6. Merge sort result table

The plot:

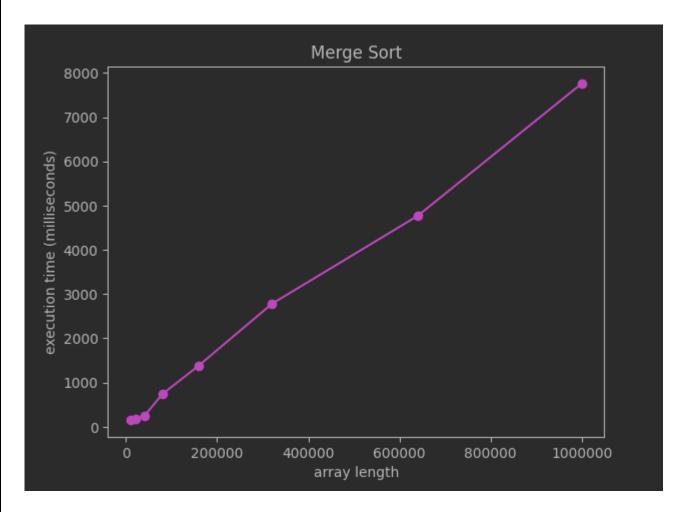


Figure 7. Merge sort plot result

Space complexity: O(n)

Best case: O (n*logn)

Worst case: O (n*logn)

Stable: Yes

Heap Sort

Heapsort 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. Unlike selection sort, 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 more quickly find the largest element in each step.

Implementation:

```
def heapify(arr, n, i):
    largest = i # Initialize largest as root
    if l < n and arr[i] < arr[l]:</pre>
        largest = l
    if r < n and arr[largest] < arr[r]:</pre>
        largest = r
    if largest != i:
        (arr[i], arr[largest]) = (arr[largest], arr[i]) # swap
        heapify(arr, n, largest)
def heapSort(arr):
    n = len(arr)
        heapify(arr, n, i)
        (arr[i], arr[0]) = (arr[0], arr[i]) # swap
        heapify(arr, i, 0)
```

Figure 8. Heap sort implementation

Results:

			***	(:33:d-)	
	array Length	execution	time	(milliseconds)	
Θ	10000			140.676498	
1	20000			246.214867	
2	40000			461.436272	
3	80000			1058.922291	
4	160000			2124.501228	
5	320000			4748.028994	
6	640000			11152.838230	
7	1000000			16665.899038	

Figure 9. Heap sort table result

The plot:

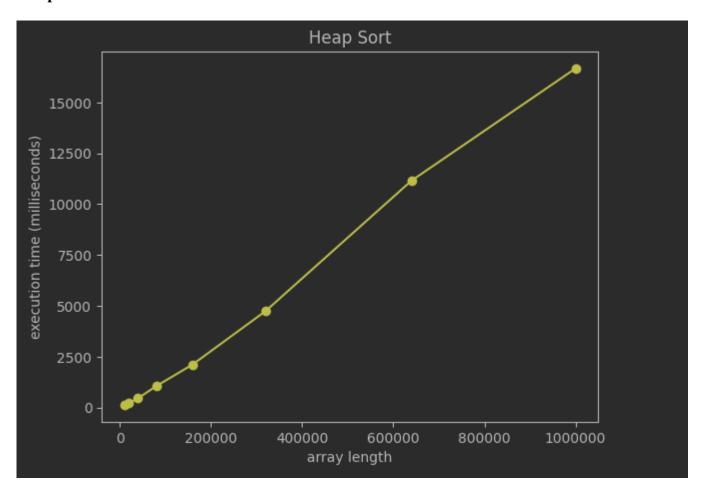


Figure 10. Heap sort plot result

Space complexity: O(n)

Best case: O (n)

Worst case: O (n*logn)

Stable: Yes

Counting Sort

Counting sort is an algorithm for sorting a collection of objects according to keys that are small positive integers; that is, it is an integer sorting algorithm. It operates by counting the number of objects that possess distinct key values, and applying prefix sum on those counts to determine the positions of each key value in the output sequence. Its running time is linear in the number of items and the difference between the maximum key value and the minimum key value, so it is only suitable for direct use in situations where the variation in keys is not significantly greater than the number of items.

Implementation:

```
def countingSort(arr):
    size = len(arr)
    output = [0] * size

# count array initialization
    count = [0] * size

# storing the count of each element
    for m in range(0, size):
        count[arr[m]] += 1

# storing the cumulative count
    for m in range(1, 10):
        count[m] += count[m - 1]

# place the elements in output array after finding the index of each element of original array in count array
    m = size - 1

while m >= 0:
    output[count[arr[m]] - 1] = arr[m]
    count[arr[m]] -= 1
    m -= 1

for m in range(0, size):
    arr[m] = output[m]
```

Figure 11. Counting sort implementation

Results:

	array length	execution	time	(milliseconds)
Θ	10000			12.528419
1	20000			24.437904
2	40000			45.604944
3	80000			75.562954
4	160000			167.082787
5	320000			445.294380
6	640000			463.932276
7	1000000			893.347740

Figure 12. Counting sort table result

The plot:

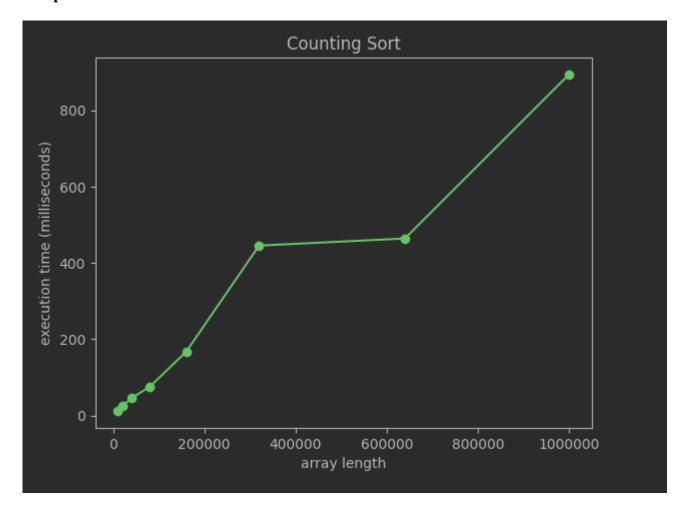


Figure 13. Counting sort plot result

Space complexity: O(k)

Best case: O (n+k)

Worst case: O (n+k)

Stable: Yes

n is the nr of elements in the array, k is the range of input

Conclusion:

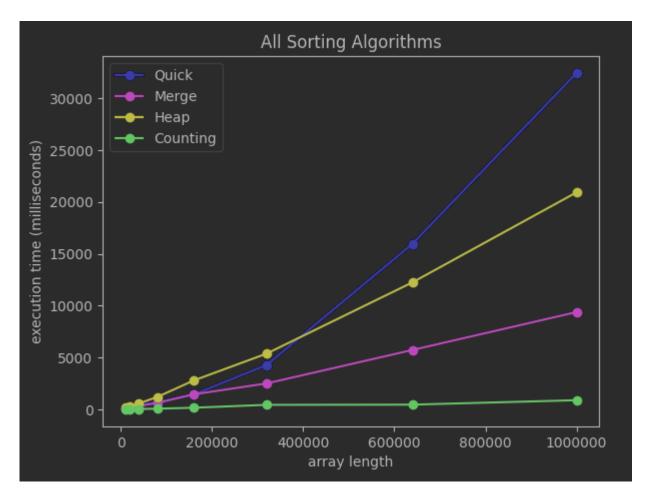


Figure 14. All sorting algorithms

The slowest algorithm, ironically, turned out to be Quick sort. The fastest is Counting Sort.

Link to GitHub: https://github.com/CristinaT21/APA_LABS