

Homework III - Analysis of Algorithms

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Abstract

In this homework, we analyze the Tim Sort algorithm on Python and compare it with the three linear sorts (Counting, Radix and Bucket sorts).

PROBLEM 1

Implement the three linear sortings using C

- (a) Counting Sort
- (b) Radix Sort
 - i. Queue version
 - ii. Give its new complexity in big O and prove it
- (c) Bucketsort

Answer

Attached to this document are the three linear sortings in C.

The complexity for Radix sort using queues is

$$T(n, d) = O(1) + d * [O(n) + O(n + 10)] \quad (1)$$

Simplifying we have

$$T(n, d) = O(1) + d * [O(2n + 10)] \quad (2)$$

$$T(n, d) = O(1) + O(2dn + 10d) \quad (3)$$

$$T(n) = O(d * n) \quad (4)$$

where d is the number of digits on the array. Also, there is an implementation on the code to confirm this, and it give us the following result.

Comprobando que $T(n, d)$ es proporcional a $d * n$:				
n	d	Tiempo(s)	d * n	T/(d*n)
1000	2	0.001000	2000	0.0000005000
1000	4	0.001000	4000	0.0000002500
1000	6	0.001000	6000	0.0000001667
5000	2	0.002000	10000	0.0000002000
5000	4	0.004000	20000	0.0000002000
5000	6	0.008000	30000	0.0000002667
10000	2	0.004000	20000	0.0000002000
10000	4	0.013000	40000	0.0000003250
10000	6	0.013000	60000	0.0000002167

Fig. 1: Probe of $O(d*n)$

Here we can see that when d increases from 2 to 6 (having n=1000), the execution time remains with the same value, and when n increases, the time value also increases d times. So, what does this means?, it means that when we have a constant value of d, the behavior is $O(n)$, and when we have a non constant value of d, we have $O(d)$. The last column show us that proportion of time against the behavior $O(d*n)$

PROBLEM 2

Compare them against Tim Sort the one in Python

- (a) Please provide with plots
- (b) Explain why the differences
 - i. Use the papers
 - A. Auger, Nicolas; Nicaud, Cyril; Pivoteau, Carine (2015). "Merge Strategies: from Merge Sort to TimSort"
 - B. Auger, Jugé, Nicaud, Pivoteau (2018). "On the Worst-Case Complexity of TimSort"

Answer

(a) Here are the plots of execution time, taking an average time of each sorting algorithm after running 5 times each one.

```
===COUNTING SORT===  
Your array is: 9 1 5 2 4 7 3  
Ordered array  
Your array is: 1 2 3 4 5 7 9  
execution time: 4217 microseconds
```

Fig. 2: Counting sort

```
===RADIX SORT===  
Your array is: 170 45 75 90 2 802 24 66  
Ordered array  
Your array is: 2 24 45 66 75 90 170 802  
execution time: 1328 microseconds
```

Fig. 3: Radix sort using queue

```
Original: Array: 0.780 0.170 0.390 0.260 0.720 0.940 0.210 0.120 0.230 0.680  
Ordenado: Array: 0.120 0.170 0.210 0.230 0.260 0.390 0.680 0.720 0.780 0.940  
execution time: 1080 microseconds
```

Fig. 4: Bucket sort

```
Given Array is  
[-2, 7, 15, -14, 0, 15, 0, 7, -7, -4, -13, 5, 8, -14, 12]  
After Sorting Array is  
[-14, -14, -13, -7, -4, -2, 0, 0, 5, 7, 7, 8, 12, 15, 15]  
Exec time is: 0.00016780001169536263
```

Fig. 5: Tim sort (167 microseconds)

Also here is their asymptotic behavior

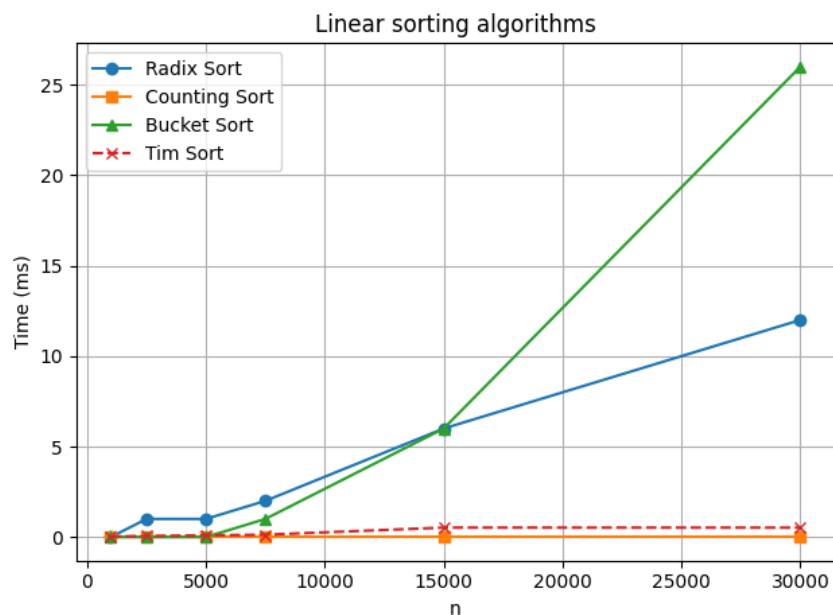


Fig. 6: Asymptotic behavior

(b) We can see that the execution time of Tim sort is less than the linear ones. This is due Tim sort is implemented as a hybrid of Merge sort and Insertion Sort, using a fusion of its recent runs on the system cache. This hybrid has a complexity of $O(n \log n)$ on the worst case and $O(n)$ on the best case. Unfortunately, to reach the best case, we need to apply a pre-sorting on

the input array in order to help the Tim sort algorithm; in fact, Tim sort uses stacks to drive the runs ensuring that the length of the runs grow at least as fast as Fibonacci numbers, keeping the stack height within a logarithmic limit.

Couting Sort and radix sort work more efficiently for integers with a smaller range. Bucket Sort assumes that the data is already uniformly distributed.

One point to keep in mind is that Tim Sort is a more general approach; it works really good with partially sorted data. This means that there will be some cases that even if it's pretty good, another algorithms could be a better approach for specific data types as you can see in the following image.

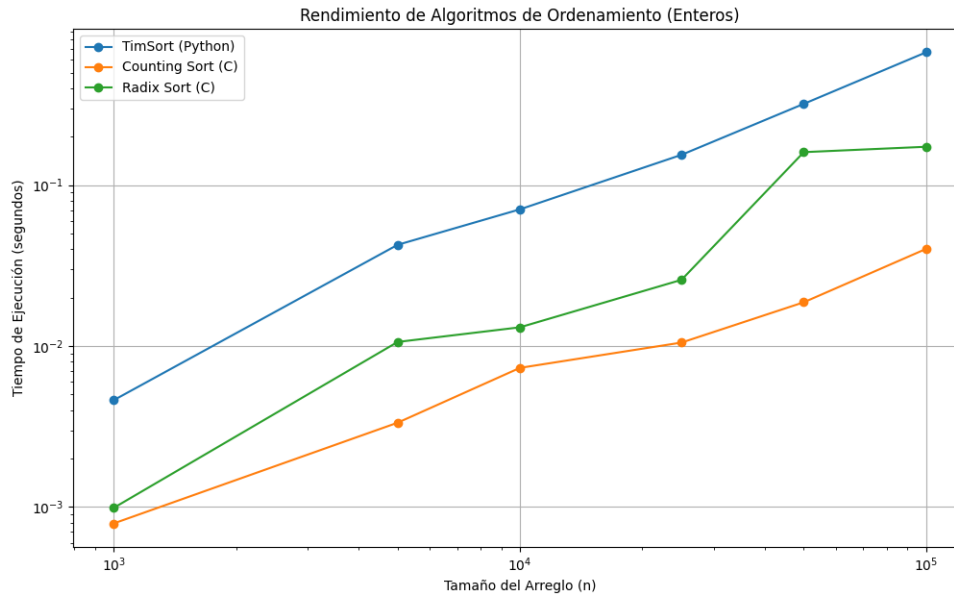


Fig. 7: specific cases

Complexity between algorithms are on the following table.

Algorithm	Worst Case	Average Case
Counting Sort	$O(n + k)$	$O(n)$
Radix Sort	$O(d * k)$	$O(n)$
Bucket Sort	$O(n^2)$	$O(n)$
Tim Sort	$O(n \log n)$	$O(n)$

TABLE I: Algorithms Complexity