

## DATA STRUCTURES AND ALGORITHMS

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ECE RUTGERS SPRING 2018

HOMEWORK – 1

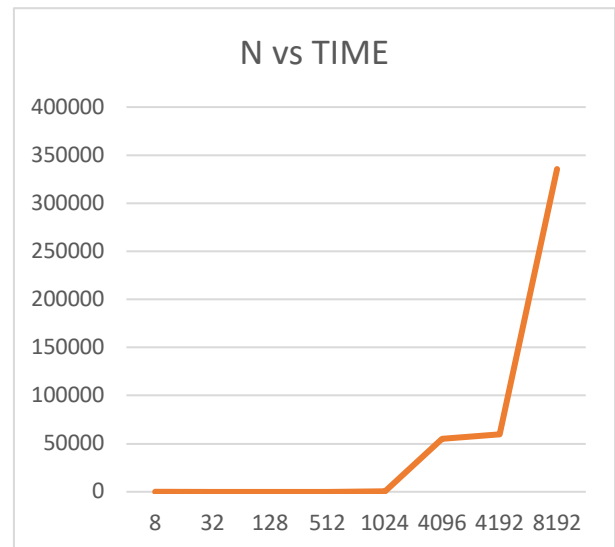
### SOLUTION AND IMPORTANT POINTS

#### Q1: 3-sum problem

Naive implementation of 3-sum problem ( $O(N^3)$ )

This implementation also known as the brute-force algorithm was implemented and the time-cost is captured and logged and plotted. The order of this algorithm is  $O(N^3)$  because the sum is calculated inside the 3-nested loop.

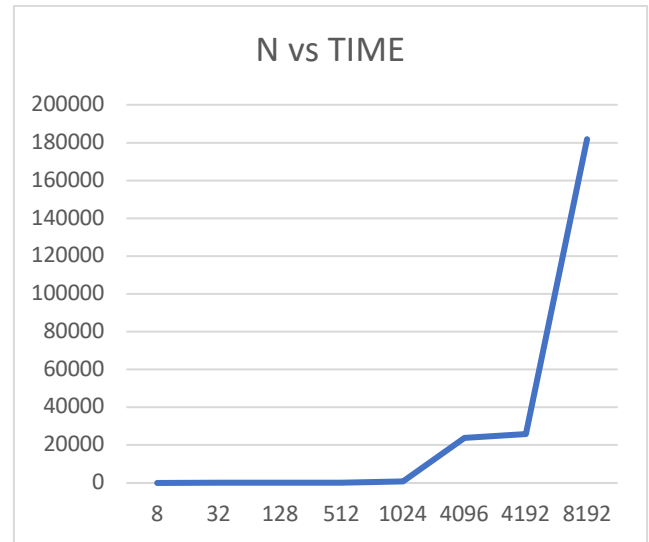
N	TIME(Naive) (in ms)
8	0.4
32	2.368
128	27.508
512	59.271
1024	334.343
4096	55197.534
4192	59705.585
8192	335568.364



Sophisticated implementation of 3-sum problem ( $O(N^2 \log N)$ )

This implementation is also called binary-search based linearithmic algorithm was implemented and the time-cost is captured and logged and plotted. The order of this algorithm is  $O(N^2 \log N)$  because the sum is calculated inside the 2-nested loop with the third element searched using binary search whose summation gives the value 0.

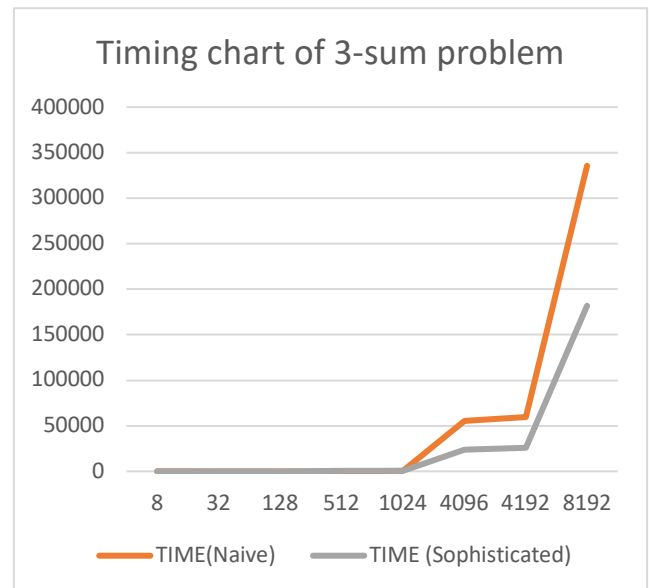
N	TIME (Sophisticated) (in ms)
8	1.269
32	6.659
128	37.541
512	232.054
1024	763.733
4096	23837.512
4192	26025.181
8192	181814.616



### Comparison of the Naïve and Sophisticated algorithms

We see that for lower value of N, the time taken by the naïve algorithm to solve the 3-sum problem is the least. But as the order of N increases, the sophisticated algorithm shows better performance by consuming lesser time because it has lower growth rate than the naïve algorithm.

N	TIME (Naive) (in ms)	TIME (Sophisticated) (in ms)
8	0.4	1.269
32	2.368	6.659
128	27.508	37.541
512	59.271	232.054
1024	334.343	763.733
4096	55197.534	23837.512
4192	59705.585	26025.181
8192	335568.364	181814.616



### Important points of Q1:

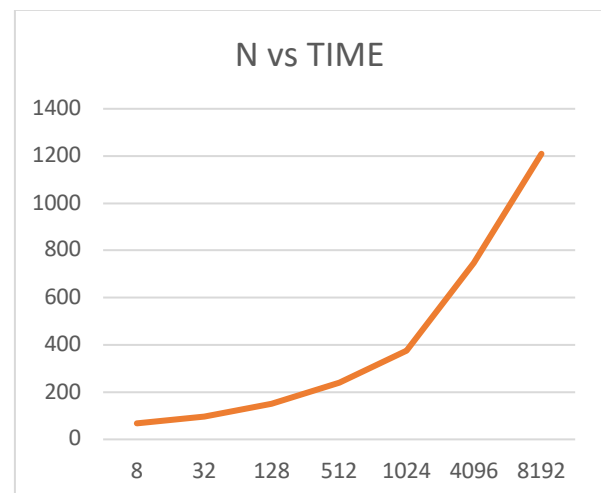
1. Graphs, Tables and Comparison between the two implementations
2. Inclusion of time complexity in the analysis. So,  $O(N^3)$  for naïve based approach and  $O(N^2 \log N)$  for sophisticated approach.

### Q2: Union Find algorithms

#### (i) Quick Find

The quick find implementation is the slowest with respect to the other two implementations. In quick find, the find operation is  $O(1)$ , a constant growth rate. But the union operation is expensive. For each union operation, we must go through  $n$  array elements to ensure successful implementation i.e. is  $O(N)$ . Thus, for  $M$  union functions, the growth rate would be  $O(MN)$ , which is a square function and expensive, with a worst case of  $O(N^2)$

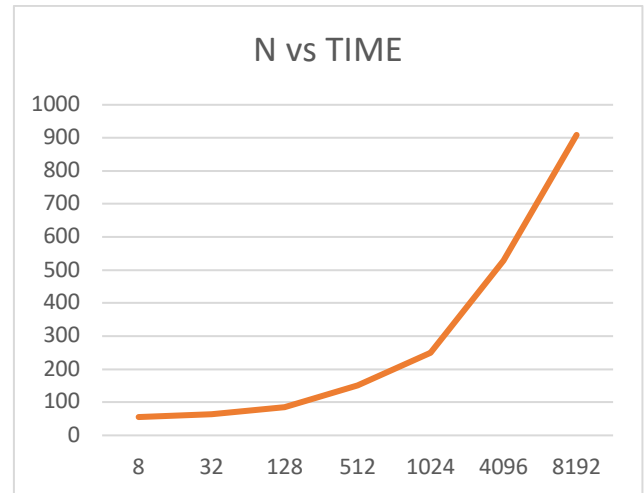
N	TIME (in ms)
8	67.82
32	96.062
128	149.858
512	241.287
1024	375.192
4096	747.392
8192	1209.469



#### (ii) Quick Union

The quick union is a better implementation of the union function. The find function is now expensive, which in the worst case is  $O(N)$ . On the other hand, the union is faster for an average case. The number of find operations the root operation in an average case is less than  $N$ . Thus, for the average case, the  $m$  union operation is less than  $O(MN)$ . But for the worst case, it is  $O(MN)$ , specifically  $O(N^2)$  for  $n$  union operations

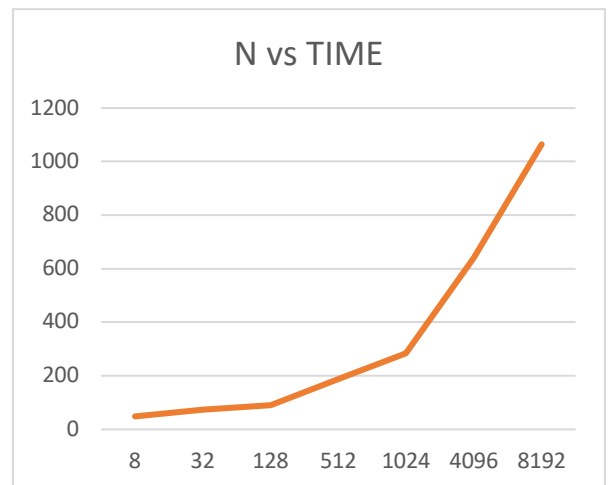
N	TIME (in ms)
8	55.42
32	63.938
128	84.45
512	150.142
1024	250.506
4096	528.749
8192	908.933



### (iii) Weighted Quick Union

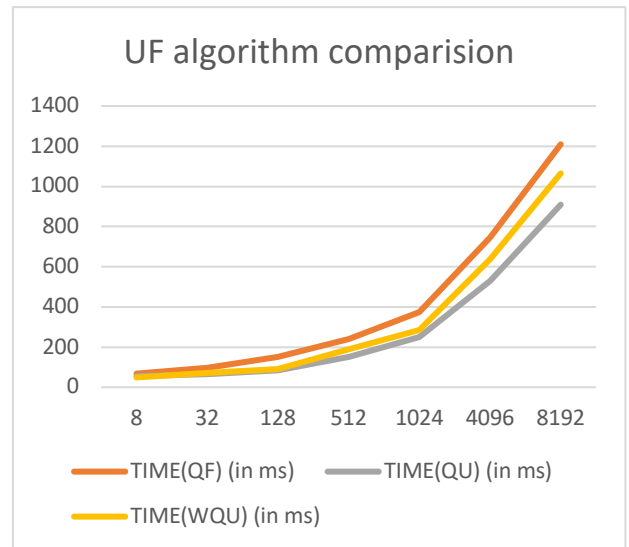
The weighted quick union is the fastest of the three. It decreases the average distance to finding a root. The union operation and the find operation on an average lead to  $O(\log N)$ .

N	TIME(in ms)
8	48.598
32	72.692
128	91.681
512	188.78
1024	283.895
4096	640.106
8192	1064.627



### Analysis of all the three algorithms collectively:

N	TIME (QF) (in ms)	TIME (QU) (in ms)	TIME (WQU) (in ms)
8	67.82	55.42	48.598
32	96.062	63.938	72.692
128	149.858	84.45	91.681
512	241.287	150.142	188.78
1024	375.192	250.506	283.895
4096	747.392	528.749	640.106
8192	1209.47	908.933	1064.63



#### Important points of Q2:

1. Graphs, Tables and Comparison between the two implementations
2. Inclusion of time complexity in the analysis in terms of big Oh notations as done above.

#### Q4: Farthest Pair (1 Dimension)

The algorithm goes through the 1-D array once using a for loop to find the minimum and maximum value. Thus, it is a linear function with growth rate  $O(N)$ , where  $n$  is the size of input data.

#### Important points of Q4:

1. Correct and runnable code.
2. Discussion of time complexity in terms of Big Oh notation.
3. I have attached the code in the assignment folder. Have a look at it if there is any confusion.

### Q3: "Big Oh"

I used my TI-83 calculator to list the data points for the naive implementation and then used the cubic regression function to determine the best fit line for the data. I received the equation  $0.167x^3 - 0.499x^2 + 0.333x + 1.71 \times 10^{-5}$ . This means that we can find the values of  $c$  and  $N_c$  by making all of the exponents of the  $x$  variables be equal to 3 and disregarding all of the negative coefficients, so  $0.167x^3 - 0.499x^2 + 0.333x + 1.71 \times 10^{-5} \leq 0.167x^3 + 0.333x^3$ , and from that,  $0.167x^3 - 0.499x^2 + 0.333x + 1.71 \times 10^{-5} \leq 0.5x^3$ , so the constant  $c$  in  $f(n) \leq c(g(n))$  is equal to 0.5. Further, the value of  $N_c$  is then 1, as can be shown by substituting  $x$  with 1 in both of the equations:  $0.167(3^3) - 0.499(3^2) + 0.333(3) + 1.71 \times 10^{-5} \leq 0.999(3^3) \Rightarrow 0.0010171 \leq 0.5$ . For the sophisticated implementation, I also used the same calculation method, but using quadratic regression instead to obtain an equation of  $7.989x^2 - 4585.769x + 816521.094$ . Using this equation and knowing that the sophisticated implementation is  $O(N^2 \log(N))$ , I used a guess and check method to try to modify the equation to obtain  $0.63 \log_2(x)x^2 - 1585.799x + 816521.094$  as the best fit line while still being  $O(N^2 \log(N))$ . Now, we get the equation  $0.63 \log_2(x)x^2 - 1585.799x + 816521.094 \leq 0.63 \log_2(x)x^2 + 816521.094x^2$ , and therefore  $0.63 \log_2(x)x^2 - 1585.799x + 816521.094 \leq 816521.724x^2 \log_2(x)$  where  $c$  is 816521.724 and  $N_c$  is 2 because if  $N_c$  were 1,  $\log(1)$  is 0.

Using the same approach for Problem 2, for quick find, the equation using linear regression is  $7.679x + 0.007$ , for quick union, the equation is  $8.668x - 0.005$ , for weighted quick find, the equation is  $7.608x + 0.032$ , and finally for weighted quick union the equation is  $8.757x - 0.015$ . From this information, the values of  $c$  and  $N_c$  can be calculated as follows: for quick find,  $7.679x + 0.007 \leq 7.679x + 0.007x$  so  $7.679x + 0.007 \leq 7.686x$  and the value of  $c$  is 7.686 and  $N_c$  is 1. For quick union,  $8.668x - 0.005 \leq 8.668x$ , so the value of  $c$  is 8.668 and  $N_c$  is 1. For weighted quick find,  $7.608x + 0.032 \leq 7.608x + 0.032x$  so  $7.608x + 0.032 \leq 7.64x$ , and the value of  $c$  is 7.64 and  $N_c$  is 1. Finally, for weighted quick union,  $8.757x - 0.015 \leq 8.757x$ , so the value of  $c$  is 8.757 and  $N_c$  is 1.

**Important points on Q3:**

1. Analysis on how to get  $c$ ,  $N_c$ .
2. Get values of  $c$ ,  $N_c$  based on the analysis.
3. Some people have not used curve fitting to get  $c$ ,  $N_c$  and have evaluated the values using the equations. Those answers are fine as well.

**Q5: Faster-est-ist 3-sum**

For the two pair implementation, I iterated through the sorted number  $n$  in one pass, with two iterators at the beginning and at the end. I observed if sum is less than 0, thus we have needed a bigger value. Hence you move the iterator in the beginning. Similarly, if sum is more than 0, thus we have needed a smaller value and you move the iterator at the end. This results in an  $O(N)$  growth rate. Using that same thinking in the three-sum implementation, we again fix one value of the element and fix two iterators at the beginning and at the end. I observed if sum is less than 0, thus we have needed a bigger value. Hence you move the iterator in the beginning. Similarly, if sum is more than 0, thus we have needed a smaller value and you move the iterator at the end. Thus, this leads to an  $O(N^2)$  growth rate.

**Important points:**

1. Correct and runnable code.
2. Discussion of time complexity in terms of Big Oh notation.