

## Written Assignment

Q1. What is the big-Oh (O) time complexity for the following algorithm (shown in pseudocode) in terms of input size  $n$ ? Show all necessary steps:

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(a) For the time complexity, (note that  $C_i$  where  $i$  is positive integer is constant and  $n$  is input size)

1.  $\text{count} = 0 \rightarrow C_1$
2.  $\text{for } i = 0 \text{ to } n-1 \rightarrow C_2n$  (For loop take  $n$  times, constant time will be omitted on  $C_1$ )
3.  $\text{sum} = 0 \rightarrow C_3n$  (because it is under for loop)
4.  $\text{for } j = 0 \text{ to } n-1 \rightarrow C_4n^2$  (Under the for loop, there is another for loop so it will run  $n^2$  times.  $n$  times by constant will be omitted)
5.  $\text{sum} = \text{sum} + A[0] \rightarrow C_5n^2$  (Calling  $A[0]$  and assigning new sum value, it is constant time \*  $n^2$  times)
6.  $\text{for } k = 1 \text{ to } j \rightarrow C_6n^3$  (This for loop is increasing  $k(1+2+3+\dots+n-1) = k*(n-1)*n/2$  and the first for loop also multiplied, then the result will be  $C_6*n^3$ )
7.  $\text{sum} = \text{sum} + A[k] \rightarrow C_7*n^3$  (Since it is under 3<sup>rd</sup> for loop)
8.  $\text{if } B[i] == \text{sum} \text{ then } \text{count} = \text{count} + 1 \rightarrow C_8*n$  (it is under the first loop)
9.  $\text{return count} \rightarrow C_9$  (Return only once (constant time))

The sum of all the steps is  $C'_1 + C'_2*n + C'_3*n^2 + C'_4*n^3$  and it satisfies  $C'_1 + C'_2*n + C'_3*n^2 + C'_4*n^3 < c*n^3$  where any  $c > C'_1 + C'_2 + C'_3 + C'_4$  or  $c = C'_1 + C'_2 + C'_3 + C'_4 + 1$  and  $n_0 = 1$  ( $n \geq 1$ )

Therefore, the answer is time complexity is  **$O(n^3)$**

(b) Document a hand-run on MyAlgorithm for input arrays  $A = [1\ 2\ 5\ 9]$  and  $B = [2\ 29\ 40\ 57]$  and show the final output.

1.  $\text{count} = 0$
2. input size is the same as size of Array which is 4 (A and B are the same input size given pseudo code)
3. Note that when ever for loop is executed, sum will be reset as  $i$  value increments. Therefore, we only can consider  $j$  for loop case. When  $j = 0$ ,  $k$  loop does not execute because  $k = 1$  and  $j = 0$  so sum is the same as  $A[0]$  which is 1.
4. Next,  $j = 1$ , then  $k$  loop executes 1 time so sum is  $1 + A[0] + A[1] = 4$ .
5. Next,  $j = 2$ , then  $k$  loop executes 2 times so sum is  $4 + A[0] + A[1] + A[2] = 12$
6. Lastly,  $j = 3$ , then  $k$  loop executes 3 times so sum is  $12 + A[0] + A[1] + A[2] = 29$
7.  $j$  loop is done, and if  $B[i] == \text{sum}$  will be executed. However when  $i = 0$ ,  $\text{sum} = 29$  and  $B[0] = 2$  so count will not increment.

8. when  $i = 1$ , do the same procedures 3 to 6 and sum is exactly the same value as 29 (since sum value becomes zero when  $i$  is incremented.). Therefore, count will be incremented.

9. when  $i = 2$  and 3, do the same procedures 3 to 6 and since  $B[2]$  and  $B[3]$  are not the same as 29, count will not be incremented.

10. return count which is 1 and **output will be 1.**

Q2. Consider the following code fragments (a), (b) and (c) where  $n$  is the variable specifying data size and  $C$  is a constant. What is the big-Oh time complexity in terms of  $n$  in each case? Show all necessary steps.

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Note that  $k_i$  is arbitrary constant for question 2.

(a) for (int  $i = 0$ ;  $i < n$ ;  $i = i + C$ )

for (int  $j = 0$ ;  $j < 10$ ;  $j++$ )

Sum[ $i$ ] +=  $j * \text{Sum}[i]$ ;

Since  $C$  is constant, for the first for loop,  $i = 0$ , one operation,  $i = i + C$  is running  $n/C$  times because of  $i < n$ .  $i < n$  is  $n/C + 1$  operation. Total operation is  $k_1n + k_2$  operations.

For the second for loop, it is constant time of loop, so we can simply conclude that it will be multiplication of  $k_3$  times on first loop. Therefore, it is  $k_4n + k_5$  operations (note that  $k_1 * k_3 = k_4$  and  $k_2 * k_3 = k_5$ ).

Sum[ $i$ ] +=  $j * \text{Sum}[i]$  is actually Sum[ $i$ ] = Sum[ $i$ ] +  $j * \text{Sum}[i]$ . Therefore, sum[ $i$ ] is calling constant times. For the simplicity, we denote another constant time  $k_6$  times will be multiplied. Therefore, it is  $k_7 * n + k_8$  operations.

In total, we can simply denote  $k'_1 + k'_2n$  and it satisfies  $k'_1 + k'_2n < c * n$  where any  $c > k'_1 + k'_2 + 1$  and  $n_0 = 1$  ( $n \geq 1$ ).

The big-Oh is thus  **$O(n)$** .

(b) for (int  $i = 1$ ;  $i < n$ ;  $i = i * C$ )

for (int  $j = 0$ ;  $j < i$ ;  $j++$ )

Sum[ $i$ ] +=  $j * \text{Sum}[i]$ ;

$i = 1$  is 1 operation. For  $i = i * C$ , it will accumulate until the value of  $1 * C * C * \dots * C$  is reached to  $n$  because of  $i < n$ , so it means that when  $\log_c n$  is reached, the loop will be stopped (Note that  $C^x = n$ , then  $x = \log_c n$ ). Therefore, we can let it  $k_1 * \log_c n$  operations.

For the second loop,  $j = 0$  is 1 operation and since  $j < i$  and  $j$  increment by 1, we can say that it is  $1 + C + C * C$  (which is  $C^2$ ) +  $C * C * C$  (which is  $C^3$ ) + ... +  $C^{(\log_c n)}$  times. Note that it is combining from previous loop. The sum of this calculation is  $(1 - C^{(\log_c n + 1)}) / (1 - C) = 1 - C^{n+1} / (1 - C)$  so we can denote  $k_2n$  operations.

In total, it is  $k_1 \cdot \log_c n + k_2 n$  and  $k_1 \cdot \log_c n + k_2 n < c \cdot n$  where  $c > k_1 + k_2$  and  $n_0 = 1$  ( $n \geq 1$ )

Therefore, the big-Oh is  **$O(n)$** .

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(c) for (int i = 1 ; i < n; i = i * 2)
    for (int j = 0; j < n; j = j + 2)
        Sum[i] += j * sum[i];
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Similar to question (b), first for loop operates  $k_1 \cdot \log(n)$  times because of similar reason to question (b) (just it is  $\log(n)$  instead of  $\log_c n$ ).

However, for the second loop, the loop will be done  $n/2$  times. The second for loop will run  $n/2$  times because  $j$  is incremented by 2. Therefore,  $n/2$  times can be written  $k_2 \cdot n$  times. Note that regardless of previous loop it will run the same amount of time. Therefore, combining with previous loop, it will be  $k_3 \cdot n \log(n)$  times.

The last operation is Same as (a), (b), simple constant time, so combining with loop, it will be  $k_4 \cdot n \log(n)$ .

In total, total operations can be denoted as  $k'_1 \cdot \log(n) + k'_2 \cdot n \log(n)$  and it satisfies  $k'_1 \cdot \log(n) + k'_2 \cdot n \log(n) < c \cdot n \log(n)$  where  $c > k'_1 + k'_2$  or  $c = k'_1 + k'_2 + 1$  and  $n_0 = 1$  ( $n \geq 1$ ).

Therefore, the big-Oh is  **$O(n \cdot \log(n))$** .

Q3. The number of operations executed by algorithms A and B are  $12n^3 + 40n \log n$  and  $5n^4 - 100n^2$  respectively. Determine an  $n_0$  such that B is greater than A for  $n \geq n_0$ .

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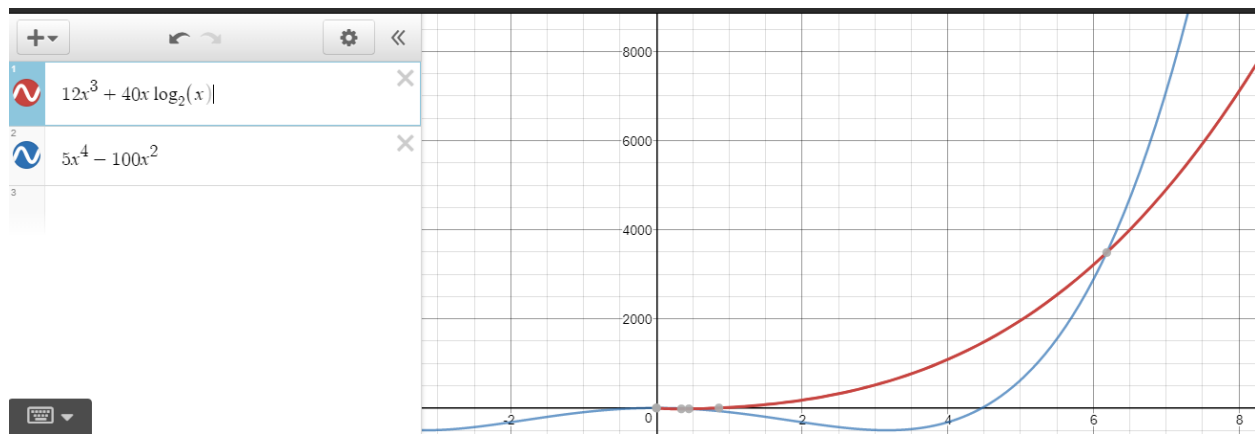
B-A should be greater than 0.  $5n^4 - 100n^2 - 12n^3 - 40n \log n > 0$ . We know that  $n \log(n) \geq n$  when  $n \geq 2$ . Please note that in algorithm  $\log(n) = \log_2 n$ . Assume  $5n^4 - 100n^2 - 12n^3 - 40n > 0$  then  $n > 5.97$ , base on this, we test 6 or 7 (or more if it's applicable)

When  $n = 6$ , B-A = -332.39 and when  $n = 7$ , B-A = 2202.94

Therefore, B is greater than A for  $n \geq n_0$  where  **$n_0 = 7$**

Also, we can simply prove by the graph. Check the below image. In this case, it is also  $n_0 = 7$

You can also verify with below image.



Please note that if by mistake, if we denote  $\log(n)$  as  $\log_{10}(n)$  (normal mathematical intuition), Then the answer will be different or wrong, However, in algorithm calculation,  $\log(n)$  should be  $\log_2(n)$

Q4. Answer the following questions:

a) Show that if  $d(n)$  is  $O(f(n))$  and  $e(n)$  is  $O(g(n))$ , then  $d(n) + e(n)$  is  $O(f(n) + g(n))$ .

By definition,  $d(n) \leq C_1 * f(n)$ ,  $n \geq n_1$  and  $e(n) \leq C_2 * g(n)$ ,  $n \geq n_2$ . Adding them, then,

$d(n) + e(n) \leq C_1 * f(n) + C_2 * g(n)$ . Actually,  $O(f(n) + g(n))$  is  $C_3(f(n) + g(n))$ .

If we denote  $C_3 = C_1 + C_2$ . We get

$d(n) + e(n) \leq C_1 * f(n) + C_2 * g(n) \leq (C_1 + C_2) * (f(n) + g(n))$  where  $n \geq \max(n_1, n_2)$  (Note that if max is not allowed, we can simply determine  $n \geq n_1 + n_2$ )

b) Show that if  $d(n)$  is  $O(f(n))$  and  $e(n)$  is  $O(g(n))$ , then  $d(n) - e(n)$  is not necessarily  $O(f(n) - g(n))$ .

We can prove by counter example.

Assume that  $d(n) = 3x$ , then  $f(n) = n$ .  $e(n) = 2x$ , then  $g(n) = n$  as well.  $d(n) - e(n) = x$ ; however,  $f(n) - g(n) = 0$  which we denote in big-Oh as  $O(1)$ . It shows that  $d(n) - e(n) = x$  so its big-Oh should be  $O(x)$ , but  $f(n) - g(n)$  can be 0 in some cases so given statement is true.

c) Show that  $2^{n+1} + n^2$  is  $O(2^n)$

$2^{n+1} + n^2 \leq g(n)$  where  $g(n) = c * 2^n$  where  $c = 5$  and  $n \geq 1$ .

Therefore  $g(n) = 2^n$  so its big-Oh is  $O(2^n)$ .

d) Show that  $f(n) = \sum_{i=1}^n i^2$  is  $O(n^3)$

$\sum_{i=1}^n i^2 = 1 + 4 + 9 + \dots + n^2$ . It can be shown as  $n(n+1)(2n+1)/6 = (2n^3+3n^2+n)/6$

$(2n^3+3n^2+n)/6 \leq C \cdot n^3$  where  $C = 1$  and  $n \geq 1$  so big-Oh is  $O(n^3)$