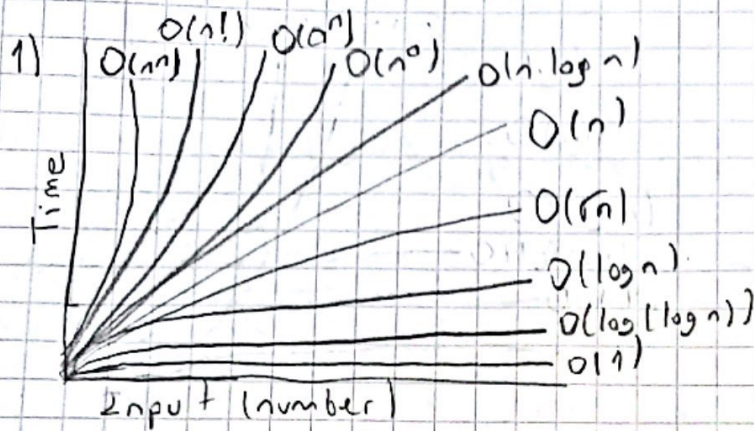


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CSE 321 - HOMEWORK 1



$$T_1(n) = 3 \log n + 3 \rightarrow O(\log \log n) < O(3 \log n + 3) < O(n) \quad +$$

$$T_2(n) = 4 \log(\log n) \rightarrow O(1) < O(4 \log(\log n)) < O(\log n) \quad +$$

$$T_3(n) = n^5 + 8n^4 \rightarrow O(n \log n) < O(n^5 + 8n^4) < O(a^n) \quad +$$

$$T_4(n) = 2000n + 1 \rightarrow O(n) < O(2000n + 1) < O(n \log n) \quad +$$

$$T_5(n) = \left(\frac{n}{6}\right)^2 \rightarrow O(n \log n) < O\left(\frac{n}{6}\right)^2 < O(a^n) \quad +$$

$$T_6(n) = 3^n + n^2 \rightarrow O(n^2) < O(3^n + n^2) < O(n!)$$

$$T_7(n) = n^n + 1000n \rightarrow O(n!) < n^n + 1000n$$

$$T_8(n) = 2^n + n^3 \rightarrow O(n^2) < O(2^n + n^3) < O(n!)$$

$$T_2(n) < T_1(n) < T_4(n) < T_5(n) < T_3(n) < T_8(n) < T_6(n) < T_7(n)$$

2) a) $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{99n}{n} = 99 \rightarrow \text{constant}$ so $f(n) \in \Theta(g(n))$
L'Hospital rule

b) $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{2n^4 + n^2}{(\log n)^6}$ → We must apply L'Hospital 6 times.

$\frac{8n^3 + 2n}{6(\frac{1}{n} \log n)^5}$ → it's not important to take again, $\frac{32n^3}{36(\frac{1}{n} \log n)^4}$... $\frac{2^3 \cdot 2^{12} \cdot n^3}{6!} = \infty$ so $f(n) \in \omega(g(n))$

c) $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \lim_{n \rightarrow \infty} \frac{n^2 + n}{2(n \log n)}$ → It's not important because small degree
 $\sum_{x=1}^n x = \frac{n \cdot (n+1)}{2}$

$= \lim_{n \rightarrow \infty} \frac{n(n+1)}{8n} = \infty$ so $f(n) \in \omega(g(n))$

d) $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \frac{3^n}{5^n} = \left(\frac{3}{5} \right)^n = 0$ so $f(n) \in o(g(n))$

3) a) This algorithm is finding the occurrences of numbers in an array. If the occurrence of the number is bigger than the array's half size, then returns that number. If it can not found, returns -1.

Lets consider that array (1, 2, 3, 1, 1, 1). Algorithm counts occurrences of 1. It finds 4. It's bigger than half size of array so it returns 1.

b) In best case, Like the example above, our number is in the first element, and the algorithm works n time. So its $\Theta(n)$.

In worst case, our array will be like (2, 3, 4, 1, 1, 1, 1) and it works with two for loop. $\Theta(n^2)$. $\Theta(\frac{n}{2})$

4) a) This algorithm finds max number and allocates a pointer space with $\text{max}+1$. Then every space in pointer represents a different number's count. For example if array is (1, 2, 3, 1, 1, 1) $\rightarrow \text{map}[1] \rightarrow$ represents count of 1, so its 4, $\text{map}[2]$ represents count of 2, and it's 1. Then it returns the number which's count is bigger than half count of array. If It can not found, returns -1.

b) This algorithm work on $\Theta(n)$ time for both best case and worst case because one for loop always works for n times.

5) For time complexity, in best case both algorithm works same, but in worst case second algorithm is better. Second algorithm takes more space than first algorithm because it creates a pointer too. And in some cases, pointers some elements never use. For example, (1, 2, 101, 1, 1) case, 1, 2, 3 elements usable, but the rest of the elements don't use and requires more space. There is 3 different number, but we use 3 number, rest of them is junk.

96)

```

a) for i=0 to n do {
    for j=0 to m do {
        multiply a[i].b[j]
        find and store max a[i].b[j]
    }
}

```

Both best case and worst case take $\Theta(n^2)$ time because this is nested loop

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b)
    for j=0 to j=n do {
        copy all elements of a[j] array to new_array[i]
        increase i
    }

    for j=0 to j=m do {
        copy all elements of b[j] array to new_array[i]
        increase i
    }

for i=0 to n+m do {
    for j=i to n+m do {
        control if new_array[i] smaller than new_array[j] {
            swap new_array[i] and new_array[j]
        }
    }
}

```

This algorithm takes $\Theta(n^2)$ time for both best and worst cases. Because it's insertion sort

c) if adding element to last {

$a[n+1] = \text{new-element}$.

if adding element to middle {

copy $a[n/2]$ to temp

copy new-element to $a[n/2]$

slide after $a[n/2]$ of array to right

}

if adding first {

copy $a[0]$ to temp

copy new-element to $a[0]$

slide after $a[0]$ of array to right

}

best case is $O(1)$ if adding to last element.
worst case is $O(n)$ if adding to first element.

d) Same as in c, best case is delete from last
and it takes $O(1)$ time

• delete $a[n]$

if its in middle;

delete $a[n/2]$

slide array to left

if its first element;

delete $a[0]$

slide array to left.

Worst case is deleting from first element and
it takes $O(n)$ time.