

Sunday 8/22/2020

Saturday, August 22, 2020

12:12 PM

Study Notes

PreReq Asymptotic Complexity:

O notation expresses an asymptotic upper bound

Ω notation is for asymptotic lower bound

Θ notation is for asymptotic tight bound

1. $3n^2 = O(n^2)$
2. $3n^2 = \Omega(n^2)$
3. $3n^2 = \Theta(n^2)$
4. $3n^2 \neq O(n \lg n)$
5. $3n^2 = O(n^3)$
6. $3n^2 \neq \Omega(n^3)$
7. $\lg n = O(n^{0.1})$

Intro to Algorithms Book

Page 22: Sorting Examples problem

Chapter 3: Asymptotic Notation, Functions and Running Times, pg 44

Worst case running time of insertion sort is $T(n) = \Theta(n^2)$

For any two functions $f(n)$ and $g(n)$, we have $f(n) = \Theta(g(n))$ if and only if $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$.

Example, a search function


```
For x in len(array):  
    If x == targetvalue:  
        Return Print("Found it")
```

The size of the array is n . The max number of times the for loop can run is n , and the worst case is the target value being in the array.

Each time the for-loop iterates, it has to do several things:

- compare guess with array.length
- compare array[guess] with targetValue
- possibly return the value of guess
- increment guess.

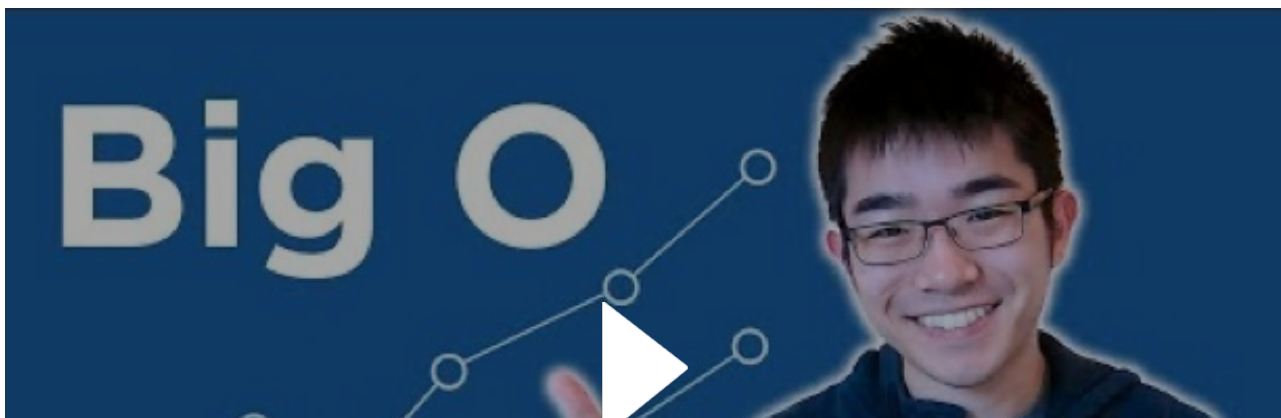
Each of these little computations takes a constant amount of time each time it executes. If the time for one iteration is c_1 , then the time for all n iterations is $c_1 \cdot n$ where c_1 is the sum of the times for the computations. Now, we cannot say here what the value of c_1 is, because it depends on the speed of the computer, the programming language used, the compiler or interpreter that translates the source program into runnable code.

This code has a little bit of extra overhead, for setting up the for-loop (including initializing guess and returning -1 at the end. Let's call the time for this overhead c_2 , which is also a constant. Therefore, the time for linear search in the worst case is $c_1 \cdot n + c_2$.

We suddenly change our minds about what to call the constants, c , to calling them k instead of c .

Intro to Big O Notation and Time Complexity

[Introduction to Big O Notation and Time Complexity \(Data Structures & Algorithms #7\)](#)



e scenario is the value not

for-loop iterates n times,
s in one loop iteration.
outer, the programming
ode, and other factors.

ss to 0) and possibly
ore, the total time for

hen we refer to Θ



To find the Big O for a problem,
If your problem is Linear, $O(n)$
If constant, $O(1)$
If quadratic, $O(n^2)$

When you're finding the time complexity, you want to take the highest variable in your problem, coefficient.

For example,

$$5n^2 + 3n + 1$$

$$T = 5n^2 = n^2 = O(n^2)$$

If you have no variables, for example, if your problem is only a constant, c , then you can look at

$$c = 0.115$$

$$T = c = 0.115 = 0.115 * 1 = O(1)$$

If you're finding the time complexity of a function, it's important to take into account the time step of the function. For example,

```
def function(given_array):  
    total = 0  
    return total
```

The first step, $total = 0$, does not rely on any variables, since it will be the same every time. The second step, $return total$, performs the same action every time without relying on any changes, therefore, $O(1)$

, and remove the

t it like this

it takes to perform each

erefore, $O(1)$
anging variables as well,

The time complexity will be

$$T = O(1) + O(1) = c_1 + c_2$$

You can call $c_1 + c_2$ a new variable of itself, c_3

$$c_1 + c_2 = c_3 = c_3 * 1 = O(1)$$

This proves that any function in which each step is $O(1)$ will ultimately have a time complexity of $O(1)$

New example

```
def find_sum(given_array):  
    total=0  
    for i in given_array:  
        total += i  
    return total
```

If you assume that `given_array` will always be a reasonable length, then each step here is also $O(1)$

*I don't understand why you would assume that, cuz it seems like this relies on a variable, but

*Never mind, he went back and said it's actually $n * O(1)$

*Ok ok I get it now. He's saying the addition of a number to the variable `total` takes the same amount of time repeated n times

All of em are $O(1)$ except for the line `total += i`, because it relies on the length of the `given_array`
 $n * O(1)$

$$T = O(1) + n * O(1) + O(1) = c_1 + n * c_2 + c_3 = c_4 + n * c_5 = O(n)$$

New example, say you're finding the time complexity for a for loop that goes into a 2d array

```
def find_sum_2d(array_2d):  
    total = 0  
    for row in array_2d:  
        for i in row:  
            total += i  
    return total
```

This is similar to the previous problem in that each line takes $O(1)$ amount of time, but not both

of $O(1)$

$O(1)$
whatev

amount of time. It's just

ay, so it will be denoted as

different to this is a constant

This is similar to the previous problem in that each line takes $O(1)$ amount of time, but what's the time for the for loop.

Since we know that each for loop is depending on a variable, `array_2d` and `row`, and we don't know the size of `array_2d`, it will be n .

$$T = O(1) + n^2 * O(1) + O(1)$$

$$T = c_3 + n^2 * c_4$$

$$T = n^2$$

$$T = O(n^2)$$

different is this is a nested

know what those are, they