

Analysis of Assignment 5 part 1 Algorithms

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1 largest34

For this problem I wrote the following code:

```
1 def largest_34(a):
2     """(list of numbers) -> number
3         Returns the sum of the 3rd and 4th largest number
4         Preconditions: len(a) >= 4
5     """
6     first = float("-inf")#lowest possible numbers so will get replaced with larger
7     second = float("-inf")
8     third = float("-inf")
9     fourth = float("-inf")
10    for num in a:
11        if num >= first:
12            fourth = third
13            third = second
14            second = first
15            first = num
16        elif num >= second:
17            fourth = third
18            third = second
19            second = num
20        elif num >= third:
21            fourth = third
22            third = num
23        elif num >= fourth:
24            fourth = num
25    return third + fourth
```

Note Line 6 and onward are the only bits of actual code running.

Consider them as Lines 1, 2, ... and 20 in the program to keep things simple
(we will ignore non-code lines in all examples following this too).

The number of operations could be then described with the following equation:

$$T_n = T_1 + T_2 + T_3 + T_4 \dots T_{20}$$

Where T_n is the total number of the operations, and $T_1, T_2, \dots T_{20}$.

The first 4 lines are assignment operators which all take constant time, therefore $T_1 = 1, T_2 = 1, T_3 = 1, T_4 = 1$.

Line 5's loop check will run about n times where n denotes the length of the list, so $T_5 = n$.

Line 6- 19 execute all execute less than or equal to n times since they are contained within the for loop.

$$T_6 \leq n, T_7 \leq n, T_8 \leq n, \dots T_{19} \leq n.$$

Note adding each of the ns to the total number of operations is not exactly correct as not all of them will run n times but doesn't matter really, since the big O notation gets rid of constants and such. Finally, the last line is constant so $T_{20} = 1$.

Giving us the following equation for the number of operations.

$$T_n \leq 1 + 1 + 1 + 1 + n + n + n + n + n + n + n + n + n + n + n + n + n + n + n + n + 1$$

We can write this in big O notation as the following: $\mathcal{O}(n)$

So, the time complexity of 1 a) $\mathcal{O}(n)$. Meaning it runs in linear time.

2 largest_third

For this problem I wrote the following code:

```

1 def largest_third(a):
2     """(list of numbers) -> number
3         Returns the sum of the largest third of numbers
4         Preconditions: len(a) >= 3
5     """
6     tmp = sorted(a, reverse=True)
7     tmp = tmp[:len(a)//3]
8     summation = 0
9     for i in tmp:
10         summation += i
11     return summation

```

This algorithm's number of operations can be described as:

$$T_n = T_1 + T_2 + T_3 + T_4 + T_5 + T_6$$

Where T_n is the total number of the operations, and $T_1, T_2, T_3, T_4, T_5, T_6$ are operations on their respective lines.

Now we know Python's inbuilt sort does approximately $n \log n$ operations, so we can say $T_1 \approx n \log n$. Slicing in python is $\mathcal{O}(k)$ but we are doing a slice of one third of a list of length n $T_2 = \frac{n}{3}$.

Initializing a variable is constant so $T_3 = 1$

The for loop test line executes n times so $T_4 = n$

The summation line also executes n times since it is in the for loop so $T_5 = n$

Returning is constant so $T_6 = 1$

We now have $T_n \approx n \log n + \frac{n}{3} + 1 + n + n + 1$. We can write this in big O notation as the following: $\mathcal{O}(n \log n)$

So, the time complexity of 1 b) is $\mathcal{O}(n \log n)$.

3 third_at_least

For this problem I wrote the following code:

```

1 def third_at_least(a):
2     """(list of numbers) -> number or None
3         Returns the value that occurs len(a)//3+1 times, if none
4             exist returns None. If more than one exists returns smaller
5                 value
6         Preconditions: len(a) >= 4
7     """

```

```

8     tmp = sorted(a)
9     third_of_a = len(a)//3
10    third = None
11    for i in range(len(tmp) - third_of_a):
12        if tmp[i] == tmp[i+third_of_a]:
13            if third == None:
14                third = tmp[i]
15            elif tmp[i] < third:
16                third = tmp[i]
17
18    return third

```

This algorithm's number of operations can be described as:

$$T_n = T_1 + T_2 + T_3 + T_4 + T_5 + T_6 + T_7 + T_8 + T_9 + T_{10}$$

Where T_n is the total number of the operations, and $T_1, T_2, T_3, T_4, T_5, T_6, T_7, T_8, T_9, T_{10}$ are operations on their respective lines.

Now we know Python's inbuilt sort does approximately $n \log n$ operations, so we can say $T_1 \approx n \log n$. Line 2 and Line 3 are constant due to `len(a)` being executed in constant time in python so $T_2 = 1, T_3 = 1$

Line 4 executes $\frac{2n}{3}$ times so $T_4 = \frac{2n}{3}$

All the Lines inside the for loop are constant, but are executed $\frac{2n}{3}$ times due to being inside of the for loop.

So, $T_5 = \frac{2n}{3}, T_6 = \frac{2n}{3}, T_7 = \frac{2n}{3}, T_8 = \frac{2n}{3}, T_9 = \frac{2n}{3}$

Returning is constant so $T_{10} = 1$

We now have $T_n \approx n \log n + 1 + 1 + \frac{2n}{3} + \frac{2n}{3} + \frac{2n}{3} + \frac{2n}{3} + \frac{2n}{3} + \frac{2n}{3} + 1$. We can write this in big O notation as the following: $\mathcal{O}(n \log n)$

So, the time complexity of 1 c) is $\mathcal{O}(n \log n)$.

4 sum_tri

For this problem I wrote the following code:

```

1 def sum_tri(a, x):
2     """(list of numbers, number) -> bool
3         Returns True if a value a[i] + a[j] + a[k] = x exists.
4         Otherwise returns False.
5     """
6     tmp = sorted(a)
7     for i in range(len(tmp)):
8         j = i
9         k = len(tmp) - 1
10        while j <= k:
11            test_sum = tmp[i]+tmp[j]+tmp[k]
12            if test_sum == x:
13                return True
14            elif test_sum > x:
15                k -= 1
16            else:
17                j += 1
18    return False

```

Finally, this algorithm's number of operations can be described as:

$$T_n = T_1 + T_2 + T_3 + T_4 + T_5 + T_6 + T_7 + T_8 + T_9 + T_{10} + T_{11} + T_{12} + T_{13}$$

Where T_n is the total number of the operations, and $T_1, T_2, T_3, T_4, T_5, T_6, T_7, T_8, T_9, T_{10}, T_{11}, T_{12}, T_{13}$ are operations on their respective lines.

Now we know Python's inbuilt sort does approximately $n \log n$ operations, so we can say $T_1 \approx n \log n$.

Line 2 executes n times, where $n = \text{len}(a)$ so $T_2 = n$

Line 3 and line 4 execute $\text{constant} * n$ times so $T_3 = n, T_4 = n$

Now here's where things get expensive.

Line 5 will execute at most $n * n$ times i.e if the sum isn't possible as k will approach j over n elements n times.

Therefore, $T_5 = n^2$

Line 6, 7, 9, 10 and 11, 12 will run n^2 times at most so $T_6 = n^2, T_7 = n^2, T_9 = n^2, T_{10} = n^2, T_{11} = n^2, T_{12} = n^2$. Note this is not exactly correct since we know that there is no way that they will all execute n^2 times, but it will not affect our analysis, since we will turn it into big O anyways.

Line 8 will execute at most once so $T_8 = 1$

Returning is constant so $T_{13} \leq 1$

Either T_{13} or T_8 will execute not both, so I'll exclude T_8 from the calculation considering the worst possible case.

This is allowed since we get rid of constants anyways in big O.

We now have $T_n < \approx n \log n + n + n + n + n^2 + 1$. We can write this in big O notation as the following: $\mathcal{O}(n^2)$

So, the time complexity of 1 d) is $\mathcal{O}(n^2)$.