# Gebze Technical University Computer Engineering

**CSE 222 - 2018 Spring** 

**HOMEWORK 4 REPORT** 

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### 1 PROBLEM-1

Given a single linked list of integers, we want to find the maximum length sorted sublist in this list. For example for the list  $L=\{1, 9, 2, 7, 20, 13\}$  the returned list should be  $S=\{2, 7, 20\}$ .

1.1 Write an iterative function which performs this task. Analyze its complexity.

Code:

```
static LinkedList<Integer> maxList(LinkedList<Integer> list)
while((now = list.pollFirst()) != null)
        if(counter1 > counter2)
            list1.add(now);
```

```
// Keep the previous element in temp.
    temp = now;
}
// Returns the list with a large number of elements.
if(counter1 > counter2) return list1;
else return list2;
}
```

Analyze Complexity: T(n): Time Complexity, S(n): Space Complexity I will do the calculations by dividing the code into certain parts.

n => size of list. m => amount of sublist.

1.) T1 =  $\Theta$  (1), S1 =  $\Theta$  (1), Because All processes take constant time.

```
LinkedList<Integer> list1 = new LinkedList<Integer>();
LinkedList<Integer> list2 = null;

Integer temp = 0, now = 0;
int counter1 = 0, counter2 = 0;
```

2.) T2 =  $\Theta$  (n), S2 =  $\Theta$  (1), Because loop turn n times, and condition take constant time.

```
while((now = list.pollFirst()) != null) // Loop turn n times.
```

3.) T3 =  $\Theta$  (1), S3 =  $\Theta$  (m), Because there is no loop or recursive call, it take constant time.

4.) T4 =  $\Theta$  (1), S4 =  $\Theta$  (1), Just return list take 1 complexity.

```
if(counter1 > counter2) return list1;
else return list2;
```

```
Conclusion:
```

```
T(n) = T1 + T2^*(T3+T4)
= T(n) = T1 + T2^*(max(T3,T4)) = T1 + T2^*(max(\Theta(1),\Theta(1))) = T1 + T2^*(\Theta(1))
= T(n) = max(T1,T2^*(\Theta(1))) = max(\Theta(1),\Theta(n)^*(\Theta(1))) = max(\Theta(1),\Theta(n))
Result Time complexity: T(n) = \Theta(n)
S(n) = \Theta(m)
m is amount of sublists, so it not directly connected n. It can be change for situation. Best case is m=1 So, S(n) = \Theta(1) Worst case is m=n So, S(n) = \Theta(n) Result Space complexity: S(n) = O(n), S(n) = O(1)
```

# 1.2 Write a recursive function for the same purpose. Analyze its complexity by using both the Master theorem and induction.

#### Code:

```
{
    list2 = list1;
    counter2 = counter1;
    list1 = new LinkedList<Integer>();
    counter1 = 0;
}

// Otherwise, reset the list 1.
    else
    {
        list1 = new LinkedList<Integer>();
        counter1 = 0;
    }
}
// Recursive call method. Then, repeat same process.
return maxList_recursive(list, list1, list2, counter1, counter2);
}
```

Analyze Complexity: T(n): Time Complexity, S(n): Space Complexity

1.) T1 =  $\Theta$  (1), S1 =  $\Theta$  (1), Because All processes take constant time.

```
Integer now = list.pollFirst(); //poolFirst and peekFirst 's time Integer next = list.peekFirst(); //complexity is \Theta (1)
```

2.) T2 =  $\Theta$  (1), S2 =  $\Theta$  (1), Because All processes take constant time.

3.) T3 =  $\Theta$  (1), S3 =  $\Theta$  (1), Processes take constant time but each sublist need new space.

4.) T4 =  $\Theta$  (T(n)-1), S4 =  $\Theta$  (S(n)-1), Recursive calling.

return maxList recursive(list, list1, list2, counter1, counter2);

### Conclusion:

$$T(n) = T1 + T2 + T3 + T4$$

$$T(n) = max(T1,T2,T3) + T4$$

$$\mathsf{T}(\mathsf{n}) = \mathsf{max}(\ \Theta\ (\mathsf{1}),\ \Theta\ (\mathsf{1}),\ \Theta\ (\mathsf{1})\ ) + \mathsf{T4}$$

$$T(n) = T4 + \Theta(1)$$

$$T(n) = \Theta (T(n-1)) + \Theta (1)$$

$$a = 1, b = 1, k = 0$$

#### **Induction Method:**

$$T(n) = \Theta (T(n-1)) + \Theta (1)$$

$$T(n) = \Theta(T(n-1)) + c$$

$$T(n-1) = \Theta (T(n-2)) + c$$

$$T(n) = (\Theta (T(n-2)) + c) + c$$

$$T(1) = \Theta(T(0)) + c => T(1) = c$$

$$\mathsf{T}(1) = \Theta \; \mathsf{T}((\mathsf{n}\text{-}(\mathsf{n}\text{-}1))$$

So, 
$$T(n) = \Theta(T(0)) + n*c => T(0) = 0 => T(n) = \Theta(n*c)$$
 (c is constant)

Result :  $T(n) = \Theta(n)$ 

#### Master Theorem:

$$T(n)=\{ O(n^k), \text{ if } a<1, \}$$

$$O(n^{(k+1)})$$
, if a=1,

$$O(n^k a^{\frac{n}{b}})$$
, if a >1 }

So a = 1, result is : 
$$O(n^{(0+1)}) = O(n)$$

Result : T(n) = O(n) Also,  $T(n) = \theta(n)$ 

#### 2 PROBLEM-2

## 2.1 Describe and analyze a $\Theta$ (n) time algorithm that given a sorted array searches two numbers in the array whose sum is exactly x.

- 1.) Let's create a function that takes sorted array where the numbers are kept, the size of the array, and the sum to be searched as a parameter. We accept that the array that is a parameter is ordered from small to big.
- 2.) Create variables that hold left and right indexes. Left index start with 0, the right index start with (size 1). In this way, the left index starts from the left of the array, the right index starts from the right of the array.
- 3.) Until the left and right indexs show the same item, we will compare the sum of the values shown by the left and right indexes and the sum we are looking for.
- 4.) If the sum of the values shown by the left and right indexes and the sum we are looking for are equal, It will print values shown by the left and right indexes. We have found a pair that provides the sum, we must increase the index to 1 to find other pairs.
- 5.) If the sum of the values shown by the left and right indexes smaller than the sum we are looking for. It moves from the left to a big element to increase the sum of the values shown by the indexes. To makes this process, it will increase 1 the left index. This will not cause trouble because the array is sequential.
- 6.) If the sum of the values shown by the left and right indexes bigger than the sum we are looking for. It moves from the right to a small element to reduce the sum of the values shown by the indexes. To makes this process, it will reduce 1 the right index.
- 7.) The program will continue until the indexes show the same element, ie there are no pairs that can provide the sum we are searching for.

#### TIME COMPLEXITY:

- 1.) The distances of the left index and the right index are equal to the size of the array.
- 2.) In each cycle, either the left index 1 is increasing or the right index 1 is decreasing.
- 3.) This means that the distance between them in each round is decreasing 1.
- 4.) We know the loop will end when the distance between them is 0.
- 5.) As a result we know that the loop will repeat as much as the size of the array.
- 6.) Assume that n = size of aray and Time complexity is T(n).
- 7.) Result :  $T(n) = \Theta(n)$

### 3 PROBLEM-3

### 3.1 Calculate the running time of the code snippet below:

```
for (i=2*n; i>=1; i=i-1)
    for (j=1; j<=i; j=j+1)
        for (k=1; k<=j; k=k*3)
            print("hello")</pre>
```

```
print("hello")
// Time complexity of print is @(1)
```

T1(n) = 
$$\sum_{i=1}^{2n} \sum_{j=1}^{i} T3(n)$$

$$T2(n) = \sum_{i=1}^{i} T3(n)$$

 $T3(n) = 1 * log_3 j = log_3 j$  Because k \*= 3 so, Increase amount of logarithmic

$$\mathsf{T2(n)} = \sum_{j=1}^{i} T3(n) = \frac{T3(i)*T3(i+1)}{2} = \frac{(\log_3 i)*(\log_3 i+1)}{2} \cong \frac{(\log_3 i)*(\log_3 i)}{2} \cong \frac{(\log_3 i)^2}{2} \cong \log_3 i$$

$$\begin{aligned} &\mathsf{T2}(\mathsf{n}) = \log_3 i \\ &\mathsf{T1}(\mathsf{n}) = \sum_{i=1}^{2n} \sum_{j=1}^i T3(n) = \sum_{i=1}^{2n} \log_3 i = \frac{(2n)*(2n+1)}{2} * \frac{(\log_3 n)*(\log_3 n+1)}{2} \\ &\mathsf{T1}(\mathsf{n}) \cong \frac{(2n)*(2n+1)}{2} * \frac{(\log_3 n)*(\log_3 n)}{2} = \frac{4n^2+2n}{2} * \frac{(\log_3 n)^2}{2} = (2n^2+n) * \log_3 n \\ &\mathsf{T1}(\mathsf{n}) = (2n^2+n) * \log_3 n = \theta(n^2 \log n) \\ &\mathsf{T1}(\mathsf{n}) = \theta(n^2 \log n) \end{aligned}$$

RESULT:  $T(n) = \theta(n^2 log n)$ 

#### 4 PROBLEM-4

4.1 Write a recurrence relation for the following function and analyze its time complexity T(n).

- 1.) First loop turns (n/2) times and second loop turns (n/2). Each round it will do 7 processes that are taken constant time. So,  $f(n) = \theta\left(\frac{n}{2}\right) * \theta\left(\frac{n}{2}\right) * \theta(1) = \theta\left(\frac{n^2}{4}\right) = \theta(n^2)$
- 2.) 4 times the function calls itself again. Each call invokes the size as n / 2. So, Recursive complexity is  $4T(\frac{n}{2})$

3.) 
$$T(n) = 4*T(\frac{n}{2}) + f(n) = 4*T(\frac{n}{2}) + \theta(n^2)$$
  
4.)  $a = 4$ ,  $b = 2$ ,  $d = 2$   $a = b^d$   $4 = 4$   
5.)  $T(n) = \theta(n^d \log n) = \theta(n^2 \log n)$