## GTU Department of Computer Engineering CSE 222/505 - Spring 2022

## HOMEWORK 2 Atacan BAŞARAN(200104004008)

1)

By definition c.f(n) 
$$\geq T(n)$$
 for  $n \geq n_0$ 
 $c=1$ 
 $n_0=2$ 

1.  $n \geq \log_2 n^2 + 1$  for all  $n \geq 2$ 

TRUE

b)  $\Lambda$  (f(n))

By definition c.f(n)  $\leq T(n)$  for  $n \geq n_0$ 
 $c=1$ 
 $n_0=1$ 

1.  $n \leq \sqrt{n_0(n+1)}$  for all  $n \geq 1$ 

TRUE

c)  $O(n^2)$ 

By definition if  $n^{n-1} = O(n^2)$  and  $n^{n-1} = \Lambda(n^2)$ 

Then  $O(n^2)$  is true

 $O(n^2)$  (c.f(n)  $\geq T(n)$ )

 $O(n^2)$  (c.f(n)  $\geq T(n)$ )

For this equation

Chas to be in terms of  $n$ .

But  $n \geq 1$ 

So  $n \geq 1$ 

S

2) 
$$\lim_{n\to\infty} \frac{f(n)}{g(n)} = 0 \Rightarrow f(n) = o(g(n))$$
,  $g(n)$  has bigger growth rate  $1 \Rightarrow g(n) = 0$   $\Rightarrow f(n) = o(g(n))$ , they have some growth rate  $1 \Rightarrow g(n) = o(f(n))$ ,  $f(n)$  has bigger growth rate

lim 
$$\frac{\log n}{\sqrt{n}} = 0$$
 growth rate  $\log n \angle \sqrt{n}$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} = 0$  growth rate  $\sqrt{n} \angle \sqrt{n}^2$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} = 0$  growth rate  $\sqrt{n}^2 \angle \sqrt{n}^2 \log n$ 

lim  $\frac{n^2}{\sqrt{n} \log n} = 0$  growth rate  $\sqrt{n}^2 \log n \angle \sqrt{n}^2$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} \log n = 0$  growth rate  $\sqrt{n}^2 = 8^{\log n} \angle \sqrt{n}^2$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} = 0$  growth rate  $\sqrt{n}^2 = 8^{\log n} \angle \sqrt{n}$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} = 0$  growth rate  $\sqrt{n}^2 = 2^{\log n} \angle \sqrt{n}$ 

lim  $\frac{\sqrt{n}}{\sqrt{n}} = 0$  growth rate  $\sqrt{n}^2 \angle \sqrt{n}$ 

lim  $\sqrt{n} = 0$  growth rate  $\sqrt{n}^2 \angle \sqrt{n}$ 

logn  $\angle \sqrt{n} \angle \sqrt{n} \log n \angle \sqrt{n} = 8^{\log n} \angle \sqrt{n} \angle \sqrt{n}$ 

```
3)
    a)
                                                        T(n) = O(logn)
   int p_1 (int my_array[]){
             for(int i=2; i<=n; i++){ O(logn)
                     if(i\%2==0){\theta(1)}
                             count++; θ(1)
                     } else{
                              i=(i-1)i; \theta(1)
                     }
            }
    }
     b)
     int p_2 (int my_array[]){
                                                                       T(n) = \theta(n)
             first_element = my_array[0]; \theta(1)
             second_element = my_array[0]; \theta(1)
              for(int i=0; i<sizeofArray; i++){ θ(n)
                      if(my_array[i]<first_element){ θ(1)
                              second element=first element; 0(1)
                              first element=my array[i];
                                                               0(1)
                      }else if(my_array[i]<second_element){ θ(1
                              if(my_array[i]!= first_element){ θ(1)
                                       second_element= my_array[i]; θ(1)
```

}

}

}

```
c)
 int p_3 (int array[]) {
                                                              T(n) = \theta(1)
             return array[0] * array[2]; θ(1)
}
d)
                                                                      T(n) = \theta(n)
int p_4(int array[], int n) {
           Int sum = 0 \theta(1)
           for (int i = 0; i < n; i=i+5) \theta(n)
                       sum += array[i] * array[i]; θ(1)
            return sum; \theta(1)
}
 e)
                                                                                        T(n) = \theta(n*logn)
 void p_5 (int array[], int n){
            for (int i = 0; i < n; i++) \theta(n)
                        for (int j = 1; j < i; j=j*2) \theta(logn)
                                   printf("%d", array[i] * array[j]); 0(1)
 }
f)
                                                                                     Best case : \theta(n)
int p_6(int array[], int n) {
                                                                                     Worst case: θ(n*logn)
          If (p_4(array, n)) > 1000 \theta(n) + \theta(1)(Comparison) = \theta(n)
                                                                                     Average case: O(n*logn),Ω(n)
                  p_5(array, n) \theta(n*logn)
         else printf("%d", p_3(array) * p_4(array, n)) \theta(1) (p_3)
                                                             + \Theta(n) (p 4)
}
                                                             + \theta(1) (printf) = \theta(n)
```

```
g)
                                                                        T(n) = \theta(\log n) * \theta(n) = \theta(\log n * n)
int p_7(int n){
          int i = n; \theta(1)
          while (i > 0) { \theta(logn)
                    for (int j = 0; j < n; j++) \theta(n)
                              System.out.println("*"); \theta(1)
                     i = i/2; \theta(1)
           }
}
h)
int p_8( int n ){
                                                                       T(n) = O(\log n) * \theta(\log n) = O(\log^2 n)
         while (n > 0) { O(logn)
                   for (int j = 0; j < n; j++) \theta(logn)
                              System.out.println("*"); θ(1)
                    n = n / 2; \theta(1)
          }
}
i)
int p_9(n){
                                                                       Best case: θ(1)
                                                                       Worst case: θ(n)
              if (n = 0) \theta(1)
                                                                       Average case: O(n)
                             return 1 θ(1)
                                                                       Program runs for n times,
              else
                                                                       T(n) = \theta(n)
                             return n * p_9(n-1) \theta(1)
}
j)
                                                                         Best case: \theta(1)
int p_10 (int A[], int n) {
                                                                         Worst case: θ(n^2)
            if (n == 1) \theta(1)
                                                                         T(n) = O(n^2)
                       return; θ(1)
             p_10(A, n-1); \theta(1)
            j=n-1; \quad \theta(1)
            while (j > 0 \text{ and } A[j] < A[j-1]) \{ \theta(n) \}
                       SWAP(A[j], A[j-1]); \theta(1)
                       j=j-1; \quad \theta(1)
            }
 }
```

a) Big O notation is used for upper bound so it con't be used for at least " (lower bound) judgment. Also a and no values can charge. Therefore, we can not reach a clear judgment.

b) 
$$T$$
.)  $2^{n+1} = O(2^n)$   
By the definition if  $2^{n+1} = O(2^n)$  and  $2^{n+1} = \Omega(2^n)$   
Then  $O(2^n)$  is true
$$O(2^n) \Rightarrow (c.f(n) \geq T(n))$$

$$O(2^n) \Rightarrow (c.f(n) \geq T(n))$$

$$\Rightarrow c=3 \qquad 3.2^{n} \geq 2^{n+1}$$

$$por all \quad n \geq 1$$

II) 
$$2^{2n} = O(2^n)$$
By same asymptotic definition as above.

c=1 1.2°  $\leq 2^{n+1}$ for all  $n \geq 1$ 

TRUE V

 $O(n^2)$  notation can represent linear function (f(n)=n)  $O(n^2)$  notation represent quadratic function( $g(n)=n^2$ ) f(n) = g(n) can be  $n = n^2 = n^3$  and  $n^3$  can not be represented as  $O(n^4)$  because  $O(n^4)$  represent fourth degree function.

a) 
$$T(n) = 2T(n/2) + n$$
  
 $T(n) = 2^2T(n/2) + n + n$   
 $T(n) = 2^3T(n/2) + 3n$   
 $T(n) = 2^kT(n/2) + k.n$   
 $T(n) = n.1 + n.logn$   
Ly  $O(nlogn)$ 

b) 
$$T(n) = 2T(n-1)+1$$
  
 $T(n) = 2[2T(n-2)+1]+1$   
 $T(n) = 2^3T(n-3)+2^2+2+1$   
 $T(n) = 2^kT(n-k)+2^{k-1}+2^{k-2}+....+2^l+1$   
 $T(n) = 2^n.T(0)+2^n-1$   $T(n-k) = T(0) = 0$   
 $T(n) = 2^n-1$   $k = n$   
 $T(n) = 2^n-1$   $k = n$ 

 $T(n) = \theta(n)*\theta(n) = \theta(n^2)$ 

Running time in 1000 size array(millisecond): 3

Running time in 10 000 size array(millisecond): 27

Running time in 100 000 size array(millisecond): 2151

Running time in 1 000 000 size array(millisecond): 220 140

Size 1k to 10k time increase 9x Size 10k to 100k time increase 80x Size 100k to 1million time increase 102x

Theoretically, time should increase 100x when i increase size 10x because of time complexity  $\theta(n^2)$ 

Theoretical and practical values almost same in big sizes

```
public static int recursiveFindPair(int arr[], int sum, int index, int index2) {
    if (index >= arr.length) θ(1)
        return 0; θ(1)

    if (index2 >= arr.length) { θ(1)
            return recursiveFindPair(arr, sum, index + 1, index + 2); θ(n)
        }
    else if (arr[index] + arr[index2] == sum) θ(1)
        return 1 + recursiveFindPair(arr, sum, index, index2 + 1); θ(n)
    return 0 + recursiveFindPair(arr, sum, index, index2 + 1); θ(n)
}
```

Running time in 5 size array(microsecond): 3

Running time in 10 size array(microsecond): 6

Running time in 50 size array(microsecond): 213

Running time in 100 size array(microsecond): 592

Size 5 to 10 time increase 2x Size 10 to 50 time increase 35x Size 50 to 100 time increase 2.77x

Theoretically, time should increase 4x when i increase size 2x because of time complexity  $\theta(n^2)$  Theoretical and practical values almost same