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1) a) Base case  $n=1$  First: if part  $\Rightarrow \Theta(1)$   
 else part  $\Rightarrow T(n-1) \Rightarrow \text{tmp} = \log_2(LCA(n-2))$   
 if part  $\Rightarrow \Theta(1)$   
 else part  $= \Theta(1)$

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

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

$$T(n) = T(1) + \underbrace{\Theta(1) + \Theta(1) + \dots + \Theta(1)}_{n \text{ times}}$$

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

$$T(1) = \Theta(1)$$

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

By recurrence relation rule

$$x(n+1) = ax(n) + b \quad x(n) = a^n$$

characteristic equation is

$$x^2 = ax + b \quad x^2 = ax$$

$$b=0 \Rightarrow x^2 = ax$$

$$a=1$$

$$x = 1 \text{ or } x = -1$$

it can't be negative

roots are same

$$T(n) = c_1 x^n + c_2 n x^n$$

$$T(1) = 1 \Rightarrow 1 = c_1 + c_2$$

$$T(2) = 2 \Rightarrow 2 = c_1 + 2c_2 \quad c_2 = 1, c_1 = 0$$

$$T(n) = n \in \Theta(n)$$

$$\text{so } T(n) = n$$

$$T(n) \in \Theta(n)$$

6) if part  $\Rightarrow$  constant time

else part  $\Rightarrow$  floor  $\Rightarrow$  constant time

$$\text{tmp1} = \log_2(1) \Rightarrow T(n/2)$$

$$\text{tmp2} = \log_2(1) \Rightarrow T(n/2)$$

if part  $=$  constant time

else part  $=$  constant time

$$T(n) = 2T(n/2) + \Theta(1)$$

By master theorem  $T(n) = aT(n/b) + f(n)$

$$T(n) = \begin{cases} \Theta(nd) & \text{if } a < b^d \\ \Theta(n \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

$$f(n) \in \Theta(nd)$$

$$a=2 \quad b=2$$

$$\Theta(1) \Rightarrow nd=1 \quad d=0$$

$$2 > 2^0 \Rightarrow 2 > 1$$

$$\text{So 3. case } \Rightarrow T(n) \in \Theta(n^{\log_2 2}) = \Theta(n)$$

Both a and b time complexity is same, but in algorithm a we are using  $n$  additional memory. In algorithm b we are using  $n^3$  additional memory. So I would prefer algorithm a

$$a \Rightarrow \text{tmp}$$

$$b \Rightarrow \text{if, tmp1, tmp2}$$



$$2) p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

algo polynomial( $x_0, a[0 \dots n]$ ) // let  $a_0$  is at 0th index  
 $sum = 0$   
 for  $i = 0$  to  $n$   
 $sum = sum + pow(x_0, i) * a[i]$   
 end  
 return sum

$T(n) \in \Theta(n)$  because we are looking every element of the array. It is not possible to find better algorithm  
 Worst case = Best case = Average case

3) algo count\_str( $s$ -letter,  $e$ -letter, string  $[0 \dots n]$ )

counter = 0  
 for  $i = 0$  to  $n$   
 if string  $[i] == s$ -letter  
 for  $j = i+1$  to  $n$   
 if string  $[j] == e$ -letter  
 counter++  
 end if  
 end for  
 end for  
 return counter

$$A(n) = \Theta(n)$$

Average case: outer loop executed  $n$  times. Inner loop will be executed less than  $n$  times. If we remove lower order terms it will be  $\Theta(n)$

Best case: Best case occurs if start letter is not in the string.

Outer for loop will be executed  $n$  times complexity =  $\Theta(n)$

Worst case: Occurs if all elements of the string is same with start letter and end letter. Outer loop will be executed  $n$  times. Inner loop will be executed  $n-1 + n-2 + n-3 + \dots + 1$

$$\frac{n(n-1)}{2} = \frac{n^2 - n}{2} \in \Theta(n^2)$$

Best case:  $\sum_{i=0}^n 1 = n \in \Theta(n)$       Worst case:  $\sum_{i=0}^n \sum_{j=i+1}^n 1 = \sum_{i=0}^n (n-i-1+1) = \frac{n(n+1)}{2} \in \Theta(n^2)$

4) There are  $n$  points,  $k$  dimensions

$$n \rightarrow 0$$

Euclidean distance function:  $d = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2 + \dots}$

algo min\_distance(points  $[0 \dots n][0 \dots k]$ ,  $k$ )

mindist = MAX-VALUE, temp = 0

for  $i = 0$  to  $n$   
 for  $j = i+1$  to  $n$   
 for  $z = 0$  to  $k$   
 $temp = temp + pow(points[i][z] - points[j][z], 2)$   
 $temp = sqrt(temp)$   
 if  $(temp < min\_distance)$   
 $min\_distance = temp$   
 $temp = 0$



Best case = Worst case = Average case (because we are looking every elements of the points and every dimensions)

Time complexity depends on  $n$  and  $k$   $n = \text{size}$   
 $k = \text{dimensions}$

$$T(n, k) = \Theta(n^2 \times k)$$

$$\sum_{i=0}^n \sum_{j=i+1}^n \sum_{k=0}^k 1 = \sum_{i=0}^n \sum_{j=i+1}^n k = k \sum_{i=0}^n \sum_{j=i+1}^n 1 = k \sum_{i=0}^n n-i-1 = k \left( \frac{n^2 - n}{2} \right) \in \Theta(n^2 k)$$

When calculating

$x-x_0$   $y-y_0$   $z-z_0$  and so forth, I used two dimensional array because in first dimension I thought that there are points, and in second dimension of the array, there are dimensions of the points.

5) algo most\_profitable(stations[0...n])

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a)
left_index = 0 right_index = 0
profit_sum = MIN-VALUE
current_max = 0
for i = 0 to n
    current_max = stations[i]
    for j = i+1 to n
        current_max = stations[j]
        if profit_sum < current_max
            profit_sum = current_max
            left = i
            right = j
    current_max = 0

```

return subarray of stations, start index left, end index right

Worst-Average-Best  $\sum_{i=0}^n \sum_{j=i+1}^n 1 \in \Theta(n^2)$   $\sum_{i=0}^n n-i-1 = \frac{n(n+1)}{2} \in \Theta(n^2)$   
 $n = 10$

6) algo max-profit(cluster[0...n], index)

if (len(cluster) == 1)  
 return cluster[0]

cur\_index = int(len(cluster)/2)

left = max-profit(cluster[0:cur\_index])

right = max-profit(cluster[cur\_index:len(cluster)])

return left + right

$T(n) = aT(n/b) + f(n)$  we are dividing problem into 2 and solving in 2 steps

$a=2$   $b=2$   $T(n) = 2T(n/2) + \log n$   $f(n) \log n = n^{1/2} \log n$

$\epsilon = 1/2$   $n^{1-\epsilon} = n^{1/2}$   $\log n \in \Theta(n^{1/2})$  so case 1 in master theorem  
 $T(n) = \Theta(n)$