

Question 1

a) Algorithm $\text{alg1}(L[0 \dots n-1]) \rightarrow T(n)$

```

    if ( $n == 1$ )
        return  $L[0]$  }  $\Theta(1)$ 
    else
        tmp =  $\text{alg1}(L[0 \dots n-2])$  }  $T(n-1)$ 
        if ( $\text{tmp} \leq L[n-1]$ )
            return tmp
        else
            return  $L[n-1]$  }  $\Theta(1)$ 

```

Recurrence Relation

$$T(n) = T(n-1) + 1, \quad T(1) = 0$$

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

$$\rightarrow T(n) = (T(n-3) + 1) + 2$$

$$\rightarrow T(n) = \dots$$

$$\rightarrow T(n) = \underbrace{(T(1) + 1)}_0 + (n-2) \Rightarrow T(n) = n-1 \Rightarrow T(n) \in \Theta(n)$$

b) Algorithm $\text{alg2}(X[l \dots r]) \rightarrow T(n)$

```

    if ( $l == r$ )
        return  $X[l]$  }  $\Theta(1)$ 

```

else

$$\text{flr} = \text{floor}((l+r)/2) \rightarrow \Theta(1)$$

$$\text{tmp1} = \text{alg2}(X[l \dots \text{flr}]) \rightarrow T(n/2)$$

$$\text{tmp2} = \text{alg2}(X[\text{flr}+1 \dots r]) \rightarrow T(n/2)$$

```

    if ( $\text{tmp1} \leq \text{tmp2}$ )
        return tmp1
    else
        return tmp2
    }  $\Theta(1)$ 

```

Recurrence Relation

$$T(n) = 2T\left(\frac{n}{2}\right) + 1, \quad T(1) = 0$$

Master's Theorem

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

$$a : 2 \quad * \quad n^{\log_b a} = n^{\log_2 2} = n$$

$$b : 2$$

$$f(n) : 1 \quad * \quad n > f(n), \text{ thus we use case 1 in theorem}$$

$$\text{Case 1: } T(n) = \Theta(n^{\log_b a}) \Rightarrow T(n) \in \Theta(n)$$

Which one to choose?

* I would choose alg1 because solving a problem with recursion is costly. It is costly because the function is called over and over again. Since alg2 divides the problem into more subproblems it costs more, in the end they both make same amount of comparison.

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

Algorithm $\text{alg}(a[0..n], x)$

result = $a[0]$ } $\Theta(1)$

currX = 1

for $i=1$ to n do $\rightarrow \Theta(n)$

currX = currX * x

result = result + (currX * $a[i]$) } $\Theta(1)$

return result $\rightarrow \Theta(1)$

} $\Theta(n)$

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

Can we have a better time complexity?

\rightarrow No, we can't. Because in order to calculate a polynomial function we have to know every x value from x^n to x^0 and we have to multiply them with their coefficients. Thus, this problem can't be solved lower than $\Omega(n)$ complexity.

Question 3

Algorithm $\text{alg}(\text{string}) \rightarrow T(n)$
 $\left. \begin{array}{l} \text{str} = \text{string} \\ \text{size} = \text{length of the string} \\ \text{count} = 0 \end{array} \right\} \Theta(1)$
 $\Theta(n^2) \left\{ \begin{array}{l} \text{for } i=0 \text{ to } (\text{size}-1) \text{ do} \\ \quad \Theta(n) \left\{ \begin{array}{l} \text{for } j=i+1 \text{ to } \text{size} \text{ do} \\ \quad \text{if } ((\text{str}[i] == "X") \text{ and } (\text{str}[j] == "Z")) \end{array} \right. \left. \begin{array}{l} \text{count} = \text{count} + 1 \end{array} \right\} \Theta(1) \end{array} \right\}$
return count

$$* T(n) \in \Theta(n^2)$$

Question 4

Algorithm $\text{alg}(\text{Array}[k][n], k) \rightarrow T(n)$
 $d = \infty \rightarrow \Theta(1)$
 $\Theta(n^2 k) \left\{ \begin{array}{l} \text{for } i=0 \text{ to } n-1 \text{ do} \\ \quad \Theta(n \cdot k) \left\{ \begin{array}{l} \text{for } j=i+1 \text{ to } n \text{ do} \\ \quad \Theta(k) \left\{ \begin{array}{l} \text{sum} = 0 \rightarrow \Theta(1) \\ \text{for } m=0 \text{ to } k \text{ do} \\ \quad \text{sum} = \text{sum} + (\text{array}[m][i] - \text{array}[m][j])^2 \rightarrow \Theta(1) \end{array} \right. \\ \text{d} = \min(d, \text{sqr}(\text{sum})) \rightarrow \Theta(1) \end{array} \right. \end{array} \right\}$
return d

$$* T(n) \in \Theta(k \cdot n^2)$$

Question 5

a) Algorithm $\text{alg}(\text{stations}, \text{profit_rates}) \rightarrow T(n)$
size = number of stations $\rightarrow \Theta(1)$
start = 0, end = 0, curr_profit = 0, max_profit = 0 $\rightarrow \Theta(1)$
 $\Theta(n^2) \left\{ \begin{array}{l} \text{for } i=0 \text{ to } \text{size}-1 \text{ do} \\ \quad \text{curr_profit} = \text{profit_rates}[i] \\ \quad \Theta(n) \left\{ \begin{array}{l} \text{for } j=i+1 \text{ to } \text{size} \text{ do} \\ \quad \text{if } (\text{curr_profit} > \text{max_profit}) \text{ do} \\ \quad \quad \text{start} = i \\ \quad \quad \text{end} = j+1 \\ \quad \quad \text{max_profit} = \text{curr_profit} \end{array} \right. \end{array} \right\} \Theta(1)$
return (stations[start:end]) $\rightarrow \Theta(1)$

$$T(n) \in \Theta(n^2)$$

b) Algorithm alg (profit_rates, low, high) $\rightarrow T(n)$
 if (high \leq low) $\left. \begin{array}{l} \text{return profit_rates[low]} \end{array} \right\} \Theta(1)$

$$\text{mid} = (\text{low} + \text{high}) / 2 \rightarrow \Theta(1)$$

$$\text{left_max_profit} = -\infty \quad \left. \begin{array}{l} \text{sum} = 0 \end{array} \right\} \Theta(1)$$

for i = mid to low, decreasing by 1, do $\left. \begin{array}{l} \text{sum} += \text{profit_rates}[i] \\ \text{if} (\text{sum} > \text{left_max_profit}) \\ \quad \text{left_max_profit} = \text{sum} \end{array} \right\} \Theta(n)$

$$\text{right_max_profit} = -\infty \quad \left. \begin{array}{l} \text{sum} = 0 \end{array} \right\} \Theta(1)$$

for i = mid + 1 to high do $\left. \begin{array}{l} \text{sum} += \text{profit_rates}[i] \\ \text{if} (\text{sum} > \text{right_max_profit}) \\ \quad \text{right_max_profit} = \text{sum} \end{array} \right\} \Theta(n)$

$$\text{left} = \text{alg}(\text{profit_rates}, \text{low}, \text{mid}) \rightarrow T(n/2)$$

$$\text{right} = \text{alg}(\text{profit_rates}, \text{mid} + 1, \text{high}) \rightarrow T(n/2)$$

$$\text{return } \max((\text{left} + \text{right}), (\text{left_max_profit} + \text{right_max_profit})) \rightarrow \Theta(1)$$

Recurrence Relation

$$T(n) = 2T(n/2) + n$$

Master's theorem

$$a : 2 \quad * \quad n^{\log_b a} = n^{\log_2 2} = n$$

$$b : 2$$

$$f(n) : n \quad * \quad n \Leftrightarrow f(n) \rightarrow n == n, \text{ thus we will use case 2}$$

$$\text{Case 2 : } T(n) = n^{\log_b a} \cdot \log n \rightarrow T(n) \in \Theta(n \cdot \log n)$$