

Project Report

Design & Analysis of Algorithms

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Problem 1 (Strings):

Pseudocode:

FUNCTION rabinkarp(array, data, q, d):

 FOR find IN array:

 p = 0

 t = 0

 n = length(data)

 m = length(find)

 IF $n < m$:

 RETURN false

 result = 1

 FOR i = 0 to (m-1) - 1:

 result = result * d

 h = result % q

FOR i = 0 to m-1 :

 p = p + find[i]

 t = t + data[i]

matched = false

i = 0

FOR i = 0 to n - m:

 IF p == t:

 matched = true

 BREAK

 IF i < n - m:

 t = (t - data[i]) + data[i + m]

IF NOT matched:

 RETURN false

ne = ""

FOR j = 0 to n - 1:

 IF j < i:

 ne = ne + data[j]

 ELSE IF j > (i + m + 1):

 ne = ne + data[j]

data = ne

RETURN true

Time Complexity Analysis:

- The outer loop runs $O(k)$ times, where k represents the number of strings in array.
- Initializing p , t , n and m take $O(1)$ time.
- The inner power loop runs $O(m)$ times, where m is the length of string in array.
- P and t are computed in $O(m)$ time.
- The next inner loop runs $O(n-m+1)$ times performing constant operations i.e. comparing p and t and updating the value of t if p and t don't match.
- Next loop for constructing a new string runs $O(n)$ times.

The overall time complexity is $O(k(m+m+(n-m+1)+n)) = O(k(2n)) = O(kn)$.

Space Complexity Analysis:

- Variables ' p ' and ' t ' have a space complexity of $O(m)$ where m is the length of the string to be found.
 - The string variable ' ne ' has a maximum space complexity of $O(n)$. That is the worst case where the previous version of data remains unchanged.
 - The remaining variables use $O(1)$ space.
- Therefore the total space complexity is $O(m+n) = O(n)$.

Problem 2 (Graphs):

Pseudocode:

FUNCTION NearestNeighbour(vertices, size, matrix, vertexCost, totalCost):

verticesVisited = 1

dynamically declare array vertices of size 'size' initialized to false

visited[0]=true

declare vector indexes

indexes.push_back(0)

index=0

```

while (verticesVisited != size)
    min=INT_MAX
    for i from 0 to size-1
        if (visited[i] != true && matrix[index][i] != 0 && matrix[index][i] <
min)
            min = i;
    index=min
    visited[index]=true
    indexes.push_back(index)
    verticesVisited++
indexes.push_back(0)
twoOpt(indexes, vertices, matrix, size)

```

FUNCTION twoOpt(indexes, vertices, matrix, size):

```

improve=true
while(improve)
    improve=false
    for i from 0 to size-2
        for j from i+2 to size
            first=0
            second=0
            total=0
            for i from 0 to size-1
                first=indexes[i]
                second=indexes[i+1]
                total=total+matrix[first][second]
            oldDistance=total

```

```

while(i+1<j-1)
    int temp=indexes[i]
    indexes[i]=indexes[j]
    indexes[j]=temp
    (i+1)+1
    (j-1)-1
first=0
second=0
total=0
for i from 0 to size-1
    first=indexes[i]
    second=indexes[i+1]
    total=total+matrix[first][second]
newDistance=total
if( newDistance < oldDistance)
    improve=true
    break
while(i+1<j-1)
    int temp=indexes[i]
    indexes[i]=indexes[j]
    indexes[j]=temp
    (i+1)+1
    (j-1)-1
if(improve)
    break

```

Time Complexity Analysis:

- The function NearestNeighbour runs the main while loop for the number of vertices V where V is equals to n and the inner loop runs V times to find the next vertex whose edge has the minimum possible distance. Hence the code runs $O(V^2)$ or $O(n^2)$ times.
- Initializing all the variables take $O(1)$ time.
- The function NearestNeighbour then calls the function twoOpt after it has executed its codes means the complexities for both the code will be added.
- The function twoOpt executes and the two outer for loop run for $n-2$ and $n+2$ times respectfully in nested format which means their complexity is $O(n^2)$ and there's an inner loop that calculates the distance each time the loops run and the while loop inside the nested loop run for $n/2$ times hence the total time complexity for this code becomes $O(n^3)$.
- The main outer while loop is used to execute the code till there are improvements being made however if an improvement is made the code automatically breaks the execution of the double for loop hence the complexity of the function does not change from $O(n^3)$
- The overall time complexity is
 - $O(n^2 + n^2(n + n/2)) = O(n^2 + n^3 + n^3/2) = O(n^2 + 2n^3) = O(n^3)$

Space Complexity Analysis:

- The array matrix has a space complexity of n^2 as it forms a 2D matrix for the number of vertices to store the value of each edge.
- There's a 1D array of size n to save the battery drained by the robot after visiting each vertex and another 1D Boolean array to check if vertices have been visited or not.
- There's 2 vectors or size $n+1$ to save the indexes of the path and the names of the vertices.
- In total $O(n^2 + n + n + n + n) = O(n^2 + 4n) = O(n^2)$

Websites used to learn about the algorithms used in Q2:

[Nearest neighbour algorithm - Wikipedia](#)

[2-opt - Wikipedia](#)

[Traveling Salesman Problem With the 2-opt Algorithm | by Adam Davis | Medium](#)

Problem 3 (Dynamic Programming):

PART A:

Pseudocode:

ComputeTotalNumberOfWays(n):

Declare a dynamic array dp of size n+1

dp[0]=1

dp[1]=1

dp[2]=2

for i from 3 to n

 dp[i]=dp[i-1]+dp[i-2]

totalWays=dp[n]

Deallocate memory for dp array

RETURN totalWays

Time Complexity Analysis:

- The loop runs for time complexity of $O(n)$.
- The remaining statements take $O(1)$ time complexity.

The overall time complexity is $O(n)$.

Space Complexity Analysis:

- The array declared takes space complexity of $O(n)$.

The overall space complexity is $O(n)$.

| "n" e-mails | Number of ways |
|-------------|------------------------|
| 3 | 3 |
| 8 | 34 |
| 75 | 3416454622906707 |
| 1225 | 7.40×10^{255} |

PART B:

Pseudocode:

FUNCTION compute_optimal_cost_and_path(dp[], n, c[][MAX_N] , parent[]):

dp[0]=0

for j from 1 to n-1

dp[j]=INF

for i from 0 to j-1

if (dp[i] + c[i][j] < dp[j])

dp[j] = dp[i] + c[i][j];

parent[j] = i;

creating a stack called path of type int

curr = n-1

total_cost = dp[n-1]

while curr is not equal to 0

push curr+1 onto path

curr = parent [curr]

push 1 onto path

output “The optimal path is”

while path is not empty

output path.top()

path.pop()

if path is not empty

output “->”

output newline

output “The total optimal cost is” , total_cost

Time Complexity Analysis:

- The first outer loop runs $O(n-1)$ times and the inner loop runs $O(n-2)$ times which leads to time complexity of $O(n^2)$.
- The while loop pushes path into stack which at worst time complexity will be $O(n)$.
- Next while pops and prints the path which can be at max $O(n)$.
- The remaining statements take constant time $O(1)$.

The overall time complexity is $O(n^2)$.

Space Complexity Analysis:

- The space complexity of 'c' is $O(n^2)$.
- The space complexity of 'dp' and 'parent' is $O(n)$.
- The space complexity of stack 'path' can be up to $O(n)$.
- The remaining variables take constant space complexity $O(1)$.

The overall space complexity is $O(n^2)$.