

# CSC265 Homework 5

Note: This assignment was rushed in a few hours and poorly written

## Question 1 (Mark: 6.5/8)

In this question, we will take use of the, combined union-by-rank and path-compression heuristic disjoint set, which is proved to runs in  $O(m \log^*(n))$  time for  $m$  disjoint-set operations on  $n$  elements.

In the following code, we use the array  $T$  to record which vertices added into the traversal. For the sake of brevity, after each MAKE-SET( $u$ ), we treat  $u$  as both a vertex and an object in the disjoint set.  $\mathcal{S}$  will be maintained and represented by the set of all FIND-SET( $u$ ) for which  $u$  that we MAKE-SET.

```
1 Initialize Array D[1...n]
2 Initialize Array T[1...n] where each entry is False
3 D[1] = 0
4 DISTANCE(1)
5 TRAVERSE(1, NULL)
6
7 DISTANCE(u):
8     for v in A[u]:
9         D[v] = D[u] + w(u,v)
10        DISTANCE(v)
11
12 TRAVERSE(u, p):
13     MAKE-SET(u)
14     for v in A[u]:
15         TRAVERSE(v, u)
16     T[u] = True
17     if p != NULL:
18         UNION(u,p) and exchange u and p to make p the root if necessary
19     for v in B[u]:
20         if T[v] is True:
21             w = FIND-SET(v)
22              $\delta(u,v) = D[u] + D[v] - 2D[w]$ 
```

**We first prove the correctness:**

For the DISTANCE function, we conducted a pre-order traversal, resulting in  $D[u]$  for each  $u \in T$  is defined. The correctness is established through induction. Initially, we set  $D[1] = 0$ , as it represents the distance from the root to itself. Assuming  $D[u]$  is correct, we then set  $D[v]$  for each child  $v$  of  $u$  to be  $D[u] + w(u, v)$ , where  $w(u, v)$  is the weight associated with the edge between  $u$  and  $v$ . Since  $T$  is a rooted tree, the distance from a node to the root equals the sum of the distance between the node and its parent and the distance between the parent and the root. Therefore, by induction,  $D[1...n]$  is correct after the execution of line 4.

According to the definition of the TRAVERSE function,  $p$  always represents the parent of  $u$ , and MAKE-SET( $p$ ) is executed if  $p$  is not NULL. We establish the maintenance of  $\mathcal{S}$  through induction. Lines 12-16 demonstrate that TRAVERSE follows a post-order traversal. When the first node is incorporated into the traversal, it has no children. While traversing down, each node on the path undergoes MAKE-SET, resulting in all nodes being singletons. Therefore, the property of  $\mathcal{S}$  is initially fulfilled.

As this node is singleton, line 18 makes  $p$  as the root of the new set. This sets the stage for the next node to be added to the traversal. Since  $p$  will be on its path to the root, and  $u$  is a descendant of  $p$  that has been previously processed and is not included in the descendants of any other child of  $p$ , the algorithm may traverse down from  $p$  before the next node is added, adding more singleton sets  $S_i$  to  $\mathcal{S}$ . Alternatively, the next node added might be  $p$ . In either case, at the moment the new node is included in the traversal, the property of  $\mathcal{S}$  is maintained.

Assuming that when  $u$  is added to the traversal, the property of  $\mathcal{S}$  is maintained, following the identical rationale as presented earlier, the algorithm readies itself for the next node, ensuring that at the moment of its addition, the

property of  $\mathcal{S}$  holds.

The traversal terminates when the root is added to the traversal, with  $p$  being NULL. Consequently, in this iteration, no operation is performed on  $\mathcal{S}$ . Therefore, by induction, the property of  $\mathcal{S}$  consistently holds immediately after a node is added to the traversal.

**(-1.5 Marks, flawed proof)** Considering each edge  $(u, v) \in P$ , one of them must be added later than the other. Without loss of generality, let's assume that  $u$  is added first. Then, when  $v$  is added, the algorithm iterates through  $u$  in line 17, and  $T[v]$  is true. According to the invariant of set  $\mathcal{S}$ ,  $v$  belongs to  $S_i$  for some  $i$ . Following line 18 in our algorithm, the root of  $S_i$  must always be  $u_i$  on the path from  $u$  to  $u_0 = 1$ . Consequently, both  $u$  and  $v$  are descendants of  $w$ . To demonstrate that  $w$  is the deepest common ancestor of  $u$  and  $v$ , we assume the existence of  $u_j$  on the path that is deeper than  $w$ . Then,  $u$  is a descendant of  $u_j$ . However, the path  $u - u_j - w - u$  forms a cycle, contradicting the fact that  $T$  is a tree. Therefore, the path  $u - w - v$  is simple. Since  $T$  is a tree, the path from  $w$  to the root 1 is unique, as well as the paths  $u$  to  $w$  and  $w$  to  $v$ . Consequently, the distance between  $u$  and  $v$  is correctly calculated in line 22.

### We now analyze the time complexity:

Initializing two arrays takes  $O(n)$ . Calling DISTANCE(1) preforms a pre-order traverse through  $T$ , with constant number of array access and calculation, which contributes to a complexity of  $O(n)$ .

Similarly, TRAVERSE(1, NULL) preforms a post-order traverse through  $T$ , a total of  $n$  nodes (line 14-15). We first analyze line 13 and 16-18. For each node, other than constant number of array access and comparison, one MAKE-SET and one UNION is preformed, contributing to a total of  $2n$  operations along with  $O(n)$  for other operations. For line 19-22, it traverse through all edges twice in  $B$ , however, for each edge, only once the if condition of line 20 is true. Thus  $m$  FIND-SET are executed and along with  $O(m)$  for operations in line 22.

Therefore we preformed  $2n + m$  operations on this disjoint set of  $m$  elements, which has a complexity of  $O((2n + m) \log^*(n)) = O((n + m) \log^*(n))$ . Summing all complexity mentioned above, we have  $O(n) + O(n) + O(n) + O(m) + O((n + m) \log^*(n)) = O((n + m) \log^*(n))$  as desired.

## Question 2 (Mark: 9/12)

### Part a. (Mark: 2/2) Bad implementation of BFS, have typos

```

1 Initialize Matrix D[1...n][1...n] such that each entry is  $\infty$ 
2
3 For i = 1 to n:
4     Initialize a linkedlist L
5     For j = 1 to n skipping i:
6         If A[i][j] = 1:
7             L.append(j)
8             D[i][j] = 1
9     For j = 2 to n-1:
10        Initialize a linkedlist L'
11        For k in L:
12            For z = 1 to n skipping i:
13                If M[k][z] = 1 AND D[i][z] =  $\infty$ :
14                    L'.append(z)
15                    D[i,z] = j
16        L = L'

```

**We first prove the correctness:**

Consider an arbitrary  $i$  from 1 to  $n$ . In lines 4-8, vertices  $j \neq i$  that are adjacent to  $i$  are identified, and their distance is set to 1, stored in  $L$ . Thus,  $L$  now contains vertices at a distance of 1 from  $i$ .

Let's establish an invariant through induction. Assume that for  $j \in 2, \dots, n-1$ ,  $L$  holds vertices at exactly a distance of  $j-1$  from  $i$ . Line 11 iterates through such vertex  $k$ , and line 12 identifies vertices  $z$  not reached by  $i$  at a distance  $\leq j-1$ , setting their distance to  $k$  to 1. These vertices  $z$  are stored in  $L'$  and have a distance of  $j$  to  $i$ , making line 15 correct. Setting  $L$  to  $L'$  concludes the correctness of the invariant.

As there are only  $n$  vertices and the maximum distance is  $n-1$ , the for-loop in line 9 indeed finds all distances for  $i$  and stores them in  $D$ . Since  $i$  is chosen arbitrarily, we conclude that the algorithm correctly generates the distance matrix  $D$ .

**Now we analyze the time complexity:**

Let's analyze lines 9-16. For each  $j$ ,  $L$  contains vertices at exactly distance  $j$  from  $i$ . Iterating from  $j = 2$  to  $n-1$ , line 11 is executed at most  $n-1$  times since  $i$  cannot be in  $L$ . Line 12 takes  $n-1$  iterations, each with a constant number of matrix accesses, comparisons, and linked list appends, all taking constant time. Thus, lines 11-15 take  $O(n^2)$  during the entire for-loop in line 9. Since lines 10 and 16 each take constant time, we conclude that lines 9-16 take a total of  $O(n^2)$ .

Line 4 takes constant time, and the for-loop in line 5 has  $n-1$  iterations, each with a constant number of matrix accesses, comparisons, and linked list appends, contributing to  $O(n)$ . Summing all together, lines 4-16 take  $O(n^2)$ .

As initializing  $D$  takes  $O(n^2)$  and line 3 has  $n$  iterations, the overall time complexity is  $O(n^2) + O(n \times n^2) = O(n^3)$  as desired.

**Part b. (Mark: 3/4) -1 Marks, lots of typos and wrong sign at line 5**

```
1 COMPUTE_D( $D_{sq}$ ,  $P$ )
2   Initialize Matrix  $D[1\dots n][1\dots n]$  such that each entry is 0
3   For  $i = 1$  to  $n$ :
4       For  $j = 1$  to  $n$ :
5            $D[i][j] = 2S_{sq}[i][j] + P[i][j]$ 
6   return  $D$ 
```

The initialization step takes  $O(n^2)$ , and the nested for-loop iterates  $n^2$  times, each with a constant number of matrix accesses and calculations. Overall, this contributes to a time complexity of  $O(n^2)$  as desired.

To establish correctness, we derive a formula for  $D_{sq}$  in terms of  $D$  and  $P$ . Let the distance between  $u$  and  $v$  be  $d$ ; then the shortest path from  $u$  to  $v$  has  $d$  edges.

As the squared graph introduces a new edge for vertices at distance 2, we can pair each two adjacent edges in the path into one. Therefore, if  $d$  is even, the length of the new path is  $d/2$ , and if  $d$  is odd, the new path length is  $(d+1)/2$ . We assert that this new path is the shortest; if there exists a path of length  $\leq (d+p)/2 - 1$ , where  $p$  is the parity, it implies that the distance between every adjacent vertex on the path is  $\leq 2$  by the definition of a squared graph. However, the path obtained by joining the shortest paths between those vertices in  $D$  has a length of  $d+p-2 \leq d$ , contradicting the fact that  $d$  is the distance between  $u$  and  $v$ .

Thus, we have proven that the distance between  $u$  and  $v$  in the squared graph is  $(d+p)/2$ , precisely corresponding to line 6 in the algorithm. Hence, our algorithm correctly generates the matrix  $D$ .

**Part c. (Mark: 3/4) -1 Marks, includes typo and flawed proof**

```
1 COMPUTE_P( $D_{sq}$ ,  $T$ )
2   Initialize Matrix  $P[1..n][1..n]$  such that each entry is 0
3   For  $i = 1$  to  $n$ :
4       For  $j = 1$  to  $n$  skipping  $i$ :
5           If  $T[i,j] < D_{sq}[i][j]$ :
6                $P[i][j] = 1$ 
7   return  $P$ 
```

The initialization step takes  $O(n^2)$ , and the nested for-loop iterates  $n^2$  times, each with a constant number of matrix accesses and comparisons. Overall, this contributes to a time complexity of  $O(n^2)$  as desired.

Consider  $u \neq v$ , let  $w$  be adjacent to  $v$ . (Existence of  $w$  can be shown by choosing the second vertex on the path from  $v$  to  $w$ ). Let the distance between  $u$  and  $v$  be  $d$  in  $G$ . By definition of distance, the distance between  $u$  and  $w$  is either  $d - 1$  or  $d + 1$ . The existence of  $w$  shows that at least one vertex adjacent to  $v$  that  $\delta(u, w) = d - 1$ .

If  $d$  is odd, then for  $w$  such that  $\delta(u, w) = d - 1$ ,  $D_{sq}[u][w] = D_{sq}[u][v] - 1$ . And for  $w$  such that  $\delta(u, w) = d + 1$ ,  $D_{sq}[u][w] = D_{sq}[u][v]$ . By definition,  $T[u, v]$  is the average of  $D_{sq}[u][w]$  for all such  $w$ , and at least one  $w$  that  $\delta(u, w) < D_{sq}[u, v]$ . Therefore  $T[u, v] < D_{sq}[u, v]$ .

If  $d$  is even, then for  $w$  such that  $\delta(u, w) = d - 1$ ,  $D_{sq}[u][w] = D_{sq}[u][v]$ . And for  $w$  such that  $\delta(u, w) = d + 1$ ,  $D_{sq}[u][w] = D_{sq}[u][v] + 1$ . By definition,  $T[u, v]$  is the average of  $D_{sq}[u][w]$  for all such  $w$ , and at least one  $w$  that  $\delta(u, w) = D_{sq}[u, v]$ . Therefore  $T[u, v] \geq D_{sq}[u, v]$ .

Thus,  $P[i][j] = 1 \iff d \text{ is odd} \iff T[i, j] < D_{sq}[i][j]$ . This implies that our algorithm is correct.

**Part d. (Mark: 3/4)**

```
1 D = D_MATRIX(A, 1)
2
3 D_MATRIX(A, order)
4   If order  $\geq n$ :
5       Initialize  $D[1..n][1..n]$  such that all entries are 1 except 0 on diagonal
6       return  $D$ 
7   else:
8        $A^2 = \text{SQUARE}(A)$ 
9        $D_{sq} = \text{D\_MATRIX}(A^2, 2 \times \text{order})$ 
10       $T = \text{COMPUTE\_T}(D_{sq}, A)$ 
11       $P = \text{COMPUTE\_P}(D_{sq}, T)$ 
12       $D = \text{COMPUTE\_D}(D_{sq}, P)$ 
13      return  $D$ 
```

**(-1 Marks, insufficient justification)** Let  $\text{order} = 2^k$  for some  $k \in \mathbb{N}$ , distinct vertices  $u$  and  $v$  are adjacent in  $G^{\text{order}}$  if  $\delta(u, v) \leq \text{order}$  in  $G$ , which can be shown by induction on  $k$ . In our algorithm,  $\text{order}$  is always a power of 2.

We first prove the correctness by induction. The variable  $\text{order}$  tracks that the algorithm is calculating distance matrix for  $G^{\text{order}}$ . For the base case where  $\text{order} \geq n$ , every pair  $u \neq v$  satisfies  $\delta(u, v)$  in  $G$  since it's a connected graph. Thus every pair  $u \neq v$  is adjacent, which implies that our  $D$  in line 5 is correct.

Assume DMATRIX is correct for  $\text{order}' = 2^k$  and now let  $\text{order} = 2^{k-1}$  then  $D$  is correctly generated since every function in line 8-13 is assumed to be correct. Thus by induction, DMATRIX is correct for  $\text{order} = 2^k$  for all  $k \in \mathbb{N}$ . Then line 1 is correct, and our algorithm correctly generates  $D$  as desired.

Since each recursion we multiply  $\text{order}$  by 2 until  $\text{order} \geq n$ , the depth of the recursion is  $O(\log n)$ . For the last iteration, initializing  $D$  takes  $O(n^2)$ . And for each other iteration, we add the complexity of line 8, 10-12 and get a total of  $O(n^\omega) + O(n^\omega) + O(n^2) + O(n^2) = O(n^\omega)$ . Then each iteration, including the last iteration, have time complexity of  $O(n^\omega)$ .

Finally, we conclude that the algorithm takes  $O(\log n) \times O(n^\omega) = O(n^\omega \log n)$  as desired.  $\square$