Algorithms and Datastructures

Assignment 2

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1

The number of the item in the list indicated the iteration of BFS. In the discovered list, we denote (p, i) where p is the predecessor and i is the iteration. Initialization:

Q = [2], Discovered = [2]

- 1. Q = [3, 4], Discovered = [2, 3(2, 1), 4(2, 1)], Done = [2]
- 2. Q = [4], Discovered = [2, 3(2, 1), 4(2, 1)], Done = [2, 3]
- 3. Q = [5], Discovered = [2, 3(2, 1), 4(2, 1), 5(4, 3)], Done = [2, 3, 4]
- 4. Q = [1], Discovered = [2, 3(2, 1), 4(2, 1), 5(4, 3), 1(5, 4)], Done = [2, 3, 4, 5]
- 5. Q = [], Discovered = [2, 3(2, 1), 4(2, 1), 5(4, 3), 1(5, 4)], Done = [2, 3, 4, 5, 1]

Algorithm 1 Push

```
1: Data: Two queues q1, q2 with functions: size, enqueue, dequeue
 2: procedure Push(n)
        size \leftarrow q1.size()
 3:
        if size == 0 then
 4:
            q1.enqueue(n)
 5:
        else
 6:
            i \leftarrow 0
 7:
            while i < \text{size do}
 8:
 9:
                q2.enqueue(q1.dequeue())
10:
                i \leftarrow i + 1
            end while
11:
            q1.enqueue(n)
12:
            i \leftarrow 0
13:
            while i < \text{size do}
14:
                q1.enqueue(q2.dequeue())
15:
16:
                i \leftarrow i + 1
            end while
17:
        end if
18:
        return n
19:
20: end procedure
```

Algorithm 2 Pop

```
1: Data: Two queues q1, q2 with functions: size, enqueue, dequeue
2: procedure Pop
3: return q1.dequeue()
4: end procedure
```

Push: O(n), we have 2n because we go through a loop n times twice. **Pop:** O(1), we only dequeue a single item which is by itself only 1 action.

Algorithm 3 Find universal sink - Adjacency Matrix

```
a) 1: Require: adj - The adjacency matrix
    2: i \leftarrow j \leftarrow 0
    3: while i < |V| and j < |V| do
           if adj[i][j] = 1 then
               i \leftarrow i + 1
    5:
    6:
           else
    7:
               j \leftarrow j + 1
           end if
    8:
    9: end while
   10: if i > |V| then
           return False
   12: end if
   13: for k..|V| do
   14:
           if adj[i][k] = 1 then
               return False
   15:
           end if
   16:
           if adj[k][i] = 0 and k \neq i then
   17:
               return False
   18:
           end if
   19:
   20: end for
   21: return True
```

Algorithm 4 Find universal sink - Adjacency List

```
b) 1: in \leftarrow out \leftarrow [0 \cdot |V|]
                                                                           \triangleright lists of length |V| filled with zeros
     2: for i..|V| do
              \operatorname{out}[i] = |\operatorname{adj}[i]|
     3:
     4:
              for each j \in \text{adj}[i] do
                  in \leftarrow in + 1
     5:
              end for
     6:
     7: end for
     8: for i..|in| do
              if in[i] = |V| - 1 and out[i] = 0 then
                  return True
    10:
              end if
    11:
    12: end for
    13: return False
```

 $\mathcal{O}(|V|)$ does not exist because we can only access the adjacencies by going through all the adjacency lists. This makes it $\mathcal{O}(|V|+|E|)$.

4

No time left, I did find an algorithm real quickly online that does it using DFS. But that is not my solution, so I will not just quickly paste it here: https://www.baeldung.com/cs/detecting-cycles-in-directed-graph

DISCLAIMER: I realized after I made it that my assumptions were wrong, but I didn't have the time to fix it.

Just to give some assumptions that I made, based on reading the text:

- Inter-species friendships: This means that only A-H or H-A friendships exist, because inter-species means "between" species, ergo between a human and an android. If this assumption is wrong, I see no way to partition based on human/android.
- Complete list of the r pairs: This means that all the humanlike entities are connected through pairs to each other.
- r: R is is a list of friendships that contain indexes the n

Algorithm 5 Create Adjacency Matrix

```
1: Data:
      r - The friendships consisting of {first, second} which are both indexes to n,
      n - group of human-like species
 4: procedure CreateAdJMatrix(n, r)
        a_{ij} \leftarrow [|n|][|n|]
        for i \leftarrow 0 to |r| do
 6:
            a_{ij}[r.first][r.second] = 1
 7:
            a_{ij}[r.second][r.first] = 1
 8:
        end for
 9:
10:
        return \leftarrow a_{ii}
11: end procedure
```

Algorithm 6 Partition Human-like species

```
1: Require:
       r - The friendships consisting of {first, second} which are both indexes to n
       n - group of human-like species
 4: Q_1, Q_2 \leftarrow \text{Queue}
 5: list1, list2 \leftarrow []
 6: adj \leftarrow CreateAdjMatrix(n, r)
 7: Enqueue(Q_1, 0)
                                                                           \triangleright We select that we start at 0
 8: while Q_1 \neq \emptyset or Q_2 \neq \emptyset do
         while Q_1 \neq \emptyset do
9:
10:
             u \leftarrow \text{Dequeue}(Q_1)
             for each v \in \operatorname{adj}[u] do
11:
                  if v == 1 and v \notin \text{list2 then}
12:
                      Enqueue(Q_2, v)
13:
                      ListAdd(list2, v)
14:
                  end if
15:
             end for
16:
         end while
17:
         while Q_2 \neq \emptyset do
18:
             u \leftarrow \text{Dequeue}(Q_2)
19:
             for each v \in \text{adj}[u] do
20:
21:
                  if v == 1 and v \notin \text{list1 then}
22:
                      Enqueue(Q_1, v)
                      ListAdd(list1, v)
23:
                  end if
24:
             end for
25:
         end while
26:
27: end while
```

At the end of this algorithm, we have 2 separate lists, 1 for the humans and 1 for the androids. We do however not know which list contains the humans and which list contains the androids. If we cannot assume that all humanlike entities are connected, as I assumed at the start, then we can simply check if they are all connected. We can simply check at the end, by looping through |n| and check if the entry is in either list1 or list 2. If the entry is in neither, we know that we did not partition it correctly.

This algorithm is functionally correct, because we do a modified BFS, in which we have an adjacency matrix that contains all the edges. We can partition everything correctly, because we are adding every edge to the opposite list and then rinse and repeat, like in BFS.

This algorithm is $\mathcal{O}(n+r)$ because, just like in BFS, we go through every vertex once (n) and the creation of the adjacency matrix is a loop through |r|, thus it is $\mathcal{O}(n+r)$.