

## Problem 1

- (a) Consider the following algorithm with input being an adjacency list representation  $L$  of the given graph:

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**Algorithm 1** NeighborhoodDeg( $L$ )
 

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1: Initialize degree to be an array of size  $n$ 
2: Initialize nbdegree to be an array of size  $n$ 
3: for each vertex  $i$  in the graph do
4:   degree $[i] \leftarrow 0$ 
5:   for each neighbor  $j$  of vertex  $i$  do
6:     degree $[i] \leftarrow \mathbf{degree}[i] + 1$ 
7:   end for
8: end for
9: for each vertex  $i$  in the graph do
10:  nbdegree $[i] \leftarrow \mathbf{degree}[i]$ 
11:  for each neighbor  $j$  of vertex  $i$  do
12:    nbdegree $[i] \leftarrow \mathbf{nbdegree}[i] + \mathbf{degree}[j]$ 
13:  end for
14: end for
15: return nbdegree

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- (b) For the algorithm above, the overall running time can be calculated as follows. The first loop (Lines 3-8) computes the degree of each vertex in the graph and stores it in the **degree** array. Each vertex  $i$  is considered in the outer for-loop (Line 3) and then each neighbor  $j$  of vertex  $i$  is considered. Thus, for a vertex  $i$ , the inner loop runs in time  $\Theta(1) + \Theta(\text{degree}(i))$ . Summing up over all vertices  $i$ , we see that the total running time of the first loop is  $\Theta(m + n)$ . The second loop (Lines 9-14) sums up the degrees of all the neighbors of each vertex (along with the degree of that vertex) and stores this in an array **nbdegree**. The structure of this for-loop is essentially the same as the structure of the first for-loop and therefore the running time of this loop is also  $\Theta(m + n)$ . Thus the total running time of this algorithm is  $\Theta(m + n)$ .
- (c) Consider a new implementation of the algorithm with input being an adjacency matrix  $A$  instead of an adjacency list  $L$ .

The major differences between the two implementations is that in an adjacency matrix representation, examining all neighbors of any vertex  $i$  takes  $\Theta(n)$  time independent of the number of neighbors that vertex  $i$  has. This means that the first loop (Line 3-7) executes in  $\Theta(n^2)$  time and similarly, the second loop (Lines 8-15) executes in  $\Theta(n^2)$  time. Thus the total running time of this new implementation is  $\Theta(n^2)$ .

## Problem 2

- (a) Initially, all vertices are white. After each iteration of the **while** loop, the results are as below:
- 1) white:  $E, F, G$ ; grey:  $B, C, D$ ; black:  $A$
  - 2) white:  $G$ ; grey:  $C, D, E, F$ ; black:  $A, B$
  - 3) white: None; grey:  $C, D, E, F, G$ ; black:  $A, B, D$
- The final dominating set is  $A, B, D$ .

**Algorithm 2** NewNeighborhoodDeg(A)

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1: Initialize degree to be an array of size  $n$ 
2: Initialize nbdegree to be an array of size  $n$ 
3: for each vertex  $i$  in the graph do
4:   for each vertex  $j$  in the graph do
5:     degree[ $i$ ]  $\leftarrow$  degree[ $i$ ] +  $A_{ij}$ 
6:   end for
7: end for
8: for for each vertex  $i$  in  $L$  do
9:   nbdegree[ $i$ ]  $\leftarrow$  degree[ $i$ ]
10:  for each vertex  $j$  in the graph do
11:    if  $A_{ij} == 1$  then
12:      nbdegree[ $i$ ]  $\leftarrow$  nbdegree[ $i$ ] + degree[ $j$ ]
13:    end if
14:  end for
15: end for
16: return nbdegree

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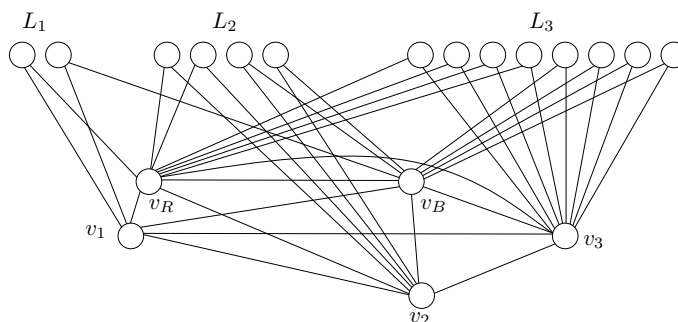


Figure 1: A bad example for greedy algorithm for Minimum Dominating Set.

- (b) The execution of the greedy algorithm will repeatedly pick the vertex with the maximum number of white neighbors. In the beginning,  $v_3$  has 8 white neighbors from  $L_3$  plus  $v_1, v_2, v_R, v_B$ , which are white as well. Thus  $v_3$  has a white neighborhood size of 13 (including itself). The remaining vertices have the following white neighborhood sizes:  $v_2$ : 9,  $v_1$ : 7,  $v_R$ : 12,  $v_B$ : 12, and every vertex  $x \in L_1 \cup L_2 \cup L_3$  has white neighborhood size equal to 3. Thus  $v_3$  will be picked first and colored black. Once  $v_3$  is colored black, the white neighborhood sizes become:  $v_2$ : 4,  $v_1$ : 2,  $v_R$ : 3,  $v_B$ : 3 and every vertex  $x \in L_1 \cup L_2$  has white neighborhood size equal to 1. Thus  $v_2$  will be picked next. Now the white neighborhood sizes are  $v_1$ : 2,  $v_R$ : 1,  $v_B$ : 1 and every vertex  $x \in L_1$  has white neighborhood size equal to 1. So  $v_1$  is picked in the last iteration. In this way, the dominating set created by the algorithm is the set  $\{v_1, v_2, v_3\}$ , but the minimum dominating set is easily seen as  $\{v_R, v_B\}$ .
- (c) The minimum dominating set in  $G_n$  has size 2. The greedy algorithm returns a dominating set  $\{v_1, v_2, \dots, v_n\}$  of size  $n$  in  $G_n$ .
- (d) The graph  $G_{21}$  serves as a counterexample to the claim that the greedy algorithm is a 10-approximation. This is because the greedy algorithm produces a solution of size 21 which is *strictly more* than 10 times the size of a minimum dominating set.

### Problem 3

- (a) The greedy algorithm in Problem 2 with input adjacency list can be implemented in the following way:

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**Algorithm 3** Dominate(L)
 

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1: Set nonblack be an empty object to host non-black vertices
2: Let ds be an empty set for hosting the dominating set
3: Let color be a length- $n$  array, all of whose slots are initialized to white
4: for each vertex  $i$  in the graph do
5:   nonblack.insert( $i$ ,  $L[i].length+1$ )
6: end for
7: ( $v$ , whiteDeg)  $\leftarrow$  nonblack.getMax()
8: while whiteDeg  $>$  0 do
9:   Save  $v$  to ds
10:  if color[ $v$ ] == white then
11:    for each neighbor  $j$  of vertex  $v$  do
12:      nonblack.decreaseValue( $j$ , 1)
13:    end for
14:  end if
15:  for each neighbor  $j$  of vertex  $v$  do
16:    if color[ $j$ ] == white then
17:      for each neighbor  $k$  of vertex  $j$  do
18:        nonblack.decreaseValue( $k$ , 1)
19:      end for
20:      color[ $j$ ]  $\leftarrow$  gray
21:    end if
22:  end for
23:  color[ $v$ ]  $\leftarrow$  black
24:  ( $v$ , whiteDeg)  $\leftarrow$  nonblack.getMax()
25: end while
26: return ds

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- (b) Given the running time of the 3 methods, `getMax`, `insert`, and `decreaseValue`, we can analyze the algorithm's running time complexity as follows. The for-loop (Lines 4-6) executes `insert( $k$ ,  $v$ )`  $n$  times, taking  $O(\log n)$  time for each insertion. Thus, this for-loop will run in  $O(n \log n)$  time. The while-loop is executed  $n$  times because with each execution, one vertex is deleted from **nonblack**. Each execution of `getMax`, takes  $O(\log n)$  time and therefore extracting vertices with largest white neighborhood from **nonblack** take  $O(n \log n)$  time. After a vertex  $v$  is extracted from **nonblack** and added to **ds**, we have to update white neighborhood sizes associated with vertices in **nonblack**. Now note that for each vertex that changes from white to gray or black, we update its neighbors' values in **nonblack**. A vertex changes from white to some other color only once and therefore for each edge we perform this update at most twice. Updates of these values (via `decreaseValue`) take  $O(\log n)$  time. Thus the total time to update sizes of white neighborhood sizes is  $O(m \log n)$ . Thus the total running time of this algorithm is  $O((m + n) \log n)$ .

- (c) The data structure that can fulfill the runtime specifications is a max-heap.

## Problem 4

This problem is a graph modeling problem. We can convert the number maze  $M$  into graph and use BFS on this graph to find a solution. First, we make each number in the number maze a vertex. Thus there are a total of  $n^2$  vertices corresponding to an  $n \times n$  maze. Then for each vertex  $M_{i,j}$ , we read its value  $k$  in the number maze and connect  $M_{i,j}$  to  $M_{i+k,j}$ ,  $M_{i-k,j}$ ,  $M_{i,j+k}$ ,  $M_{i,j-k}$  (provided these vertices exist). Once the graph is completely constructed, we can use BFS to find a shortest path from the vertex corresponding to the top-left square to the vertex corresponding to the bottom-right square.