

1) a) Maybe-mst( $G$ ):

while there is a cycle  $C$  in  $G$ :

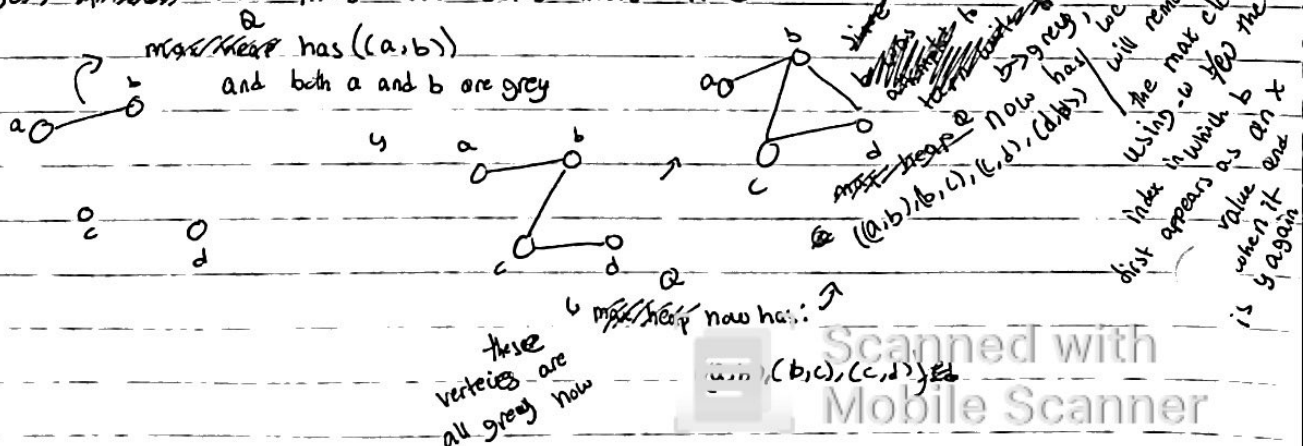
Remove the maximum weight edge in  $C$ .

Because of the fact that this algorithm only removes the maximum weighted edge in a cycle, the graph still stays connected after the edge removal, therefore if we eliminate the maximum weighted edges from all the cycles, we will always end up with a minimum spanning tree as all the remaining edges are crucial for ensuring that the graph is connected.

b) ~~Maybe-MST is similar to MST-KRUSKAL, only MST-Kruskal builds a MST by not including edges that try to connect two vertices that are a part of the same set (hence avoiding cycles) where as we are given an already connected graph consisting of cycles and are told to remove the maximum-weighted edge within the cycles, hence removing cycles.~~

~~For this algorithm, I will use disjoint sets and constants, not as a unionize vertices. If I end up encountering a cycle, I will still unionize the vertices, but then I will remove the maximum weighted edge from the cycle.~~

→ I will assume that my input is a connected and undirected graph  $G=(V,E)$  in which  $E$  is a vector of pairs of every possible vertex-vertex connections,  $(u,v)$  and each vertex-to-vertex connection has a weight  $propto(u,v).w$ . I will use DFS and DFS-visit to visit every single vertex and I will use a data structure ~~(like a heap for vertex weights)~~ to keep track of the neighboring edges. ~~approx~~ This may look some thing like:



DFS\_MAYBE\_MST (G)

for each vertex  $u \in G.V$

$u.color = WHITE \rightarrow O(V)$

for each vertex  $u \in G.V \rightarrow O(V)$

$Q = \{\}$

if  $u.color == WHITE$

DFS-VISIT (G, u, Q)

DFS-VISIT (G, u, Q)

$u.color = GRAY$

for each  $v \in G.adjs[u]$

$Q = Q \cup ((u, v))$

if  $v.color == WHITE$

DFS-VISIT (G, v, Q)

if  $v.color == GRAY$

Find the maximum  $((a, b), w)$  value in Q between when v first appears as (a) in (a, b) in Q and now when it appears as (b) in (a, b) and remove that  $((a, b))$  combo from G.E.

this step deletes the cycle and removes

the max edge in a cycle.

If a vertex is grey that means it was being visited and it been entered into Q as (v, something) and it has now cycled back and instead of has been entered as (something, v) into

Q. All the edges between (v, something) and (something, v) form a cycle. So by removing the (something, something), w with the maximum (w) from E, we have removed the maximum weighted edge in a cycle.

$u.color = BLACK$

$Q = Q - \text{all } (a, b) \text{ combos that appear after } u \text{ appears as } (a)$

this step just empties the Q data structure as we go. If we reach end of DFS-visit recursion calls, Q ends up being empty.

this step looks something like:

$Q = ((a, b), (u, \text{something}), (\text{something}, \text{something}), \dots)$

→ We need Q to be a data structure that allows us to find max value b/w certain indices.

overall runtime:  $O(V \cdot E)$

+ whatever time is required to probe through Q and remove max elements b/w indices, + time to remove pairs from G.E.

If Q is max heap and there are C nodes

Runtime:  $O(V(E) + C \cdot \lg E)$

assuming Q has E edges in the worst case

if its a max-heap it will take  $\lg N$  being size of index to index to extract max.

2) MST-prim ( $G, w, r$ )

for each  $u \in G.V$

$u.key = \infty$

$u.\pi = NIL$

$r.key = 0$

$Q = G.V$

while  $Q \neq \emptyset$

$u = \text{Extract-min}(Q)$

for each  $v \in G.Adj[u]$

if  $v \in Q$  and  $w(u, v) < v.key$

$v.\pi = u$

$v.key = w(u, v)$

done  
in constant  
time by keeping  
a bit for each  
vertex for whether  
or not it is in  
 $Q$ .

↳ this implies that we  
are making an edge  
and inserting to another  
linked list keeping track  
of keys of vertices of  
the growing minimum  
spanning tree and  
this happens in constant  
time for linked lists.

$(V-Q)$   
\*  $Q$  is all the keys  
not in the MST and  
 $V-Q$  is all the keys  
inside MST

total runtime:

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

~~$O(V^2 + E)$~~

\* This explanation in the book was not very clear,  
although the book example used a min heap,  
they had the same for loop inside a while loop,  
yet they put their final runtime for prims  
as  $O(V \lg V + E \lg V)$  instead of  
 $O(V \lg V + E \lg V)$ . Since we have to  
use a linked list the  $\lg V$  for extract min

changes to  $V$  and  $\lg V$  for decrease key simply  
changes to  $O(1)$  for insertion but it is  
still unclear why our answer is not  $O(V(V+E))$

RUNTIME:  $\Theta(V(V+E))$

For trees  
 $E = V - 1$   
for connected  
graph  $E = \Omega(V-1)$

this is  
true for simple  
graphs  
→  $E = O(V^2)$   
and that the for loop  
runs for each while loop

total run time:  
 $O(V \cdot V + V \cdot E)$   
leaving my  
answer in terms  
of  $E$

3) Bellman-Ford ( $G, w, s$ )  $\rightarrow$  original

Initialize-single-source ( $G, s$ )

for  $i = 1$  to  $|G.V| - 1$

for each edge  $(u, v) \in G.E$

Relax ( $u, v, w$ )

for each edge  $(u, v) \in G.E$

if  $v.d > u.d + w(u, v)$

return false

Return true

Initialize-single-source ( $G, s$ )

for each vertex  $v \in G.V$

$v.d = \infty$

$v.\pi = \text{NIL}$

$s.d = 0$

original

Relax ( $u, v, w$ )

if  $v.d > u.d + w(u, v)$

$v.d = u.d + w(u, v)$

$v.\pi = u$

~~Relax-New~~

New Relax will take in count\_vector as an

input argument and update its values based on

if statement. ~~Count\_vector~~ Count\_vector ( $v$ ) holds

how many shortest paths from  $s$  to  $v$  have

been encountered

Relax-New ( $u, v, w, \text{count\_vector}$ )

if  $v.d > u.d + w(u, v)$

$v.d = u.d + w(u, v)$

$v.\pi = u$

Count\_vector ( $v$ ) = 1

elseif  $v.d == u.d + w(u, v)$

Count\_vector ( $v$ ) += 1

Our new Belmont-Ford Algorithm is assumed to

always return true and the second outer for loop

is ~~not needed~~ ~~can be removed~~ ~~new one remove time the~~

same,

Bellman-Ford-New ( $G, w, s$ )

Initialize-single-source ( $G, s$ )

Initialize countvector  $\forall v$  as indices

for  $i = 1$  to  $|G.V| - 1$

for each edge  $(u, v) \in G.E$

Relax ( $u, v, w, \text{count\_vector}$ )

return Count\_vector

~~Relax-New~~ ~~(u, v, w, count\_vector)~~

Relax-New ( $u, v, w, \text{count\_vector}$ )

\* I honestly don't understand the Boolean component of this algorithm and it does not seem necessary for the prompt so I am just excluding it.\*

the count\_vector

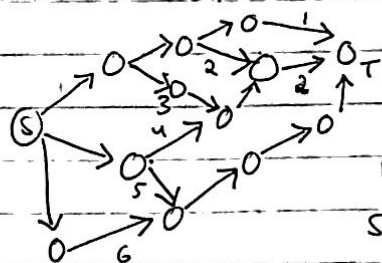
holds how many

shortest path exist

from the source to

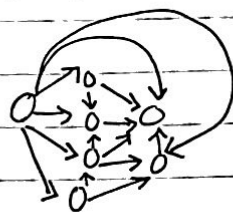
each vertex  $v$ .

- 4) Since it is a directed graph you can have multiple ways to reach  $t$  from  $s$  without forming a cycle.



6 Paths to reach  $t$  in this case

In the worst case every vertex could have many different ways of reaching  $t$  from it and  $s$  can be connected to every vertex:



Every vertex can be connected to every vertex in front of it

BFS( $G, s, t$ )

For each vertex  $u \in G.V - \{s\}$

$u.color = WHITE$

$u.d = \infty$

$u.\pi = NIL$

$s.color = GRAY$

$s.d = 0$

$s.\pi = NIL$

Initialize ~~vector~~ vector to hold paths called PATHS

Enqueue( $Q, s$ )

while  $Q \neq \emptyset$

$u = Dequeue(Q)$

for each  $v \in G.adj[u]$

if  $v == t$

PATHS.append( $s \xrightarrow{P} u \xrightarrow{P} v$ )

else if  $v.color == white$

$v.color = GRAY$

$v.d = u.d + 1$

$v.\pi = u$

Enqueue( $Q, v$ )

$u.color = BLACK$

Return PATHS

only works if  $s$  is connected to any vertex that is connected to  $t$ .

$$O(|V| + |E|) = O(b^d)$$

where

$b$  = branching factor of graph or average out factor

and  $d$  = distance from start node

$d = n \rightarrow$  can be  $n$  nodes away and assuming each ~~branch~~ node has 2 children on average

run time is:

$$O(2^n)$$



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→ Necessary supplemental algorithms:

 $u = 0.1$ 

return X.P

to run it  
n times

else  $x.p = y$

Union  $(x, y)$

$$O(m \alpha(n)) = O(m \lg n)$$

if  $x.\text{rank} == y.\text{rank}$

Link (Find-Set (x), Find-Set (y))  $\rightarrow$  ~~set~~  
 $\hookrightarrow O(\log n)$

$$y.\text{rank} = y.\text{rank} + 1$$

~~and these all of them take time or constant time~~

algorithm to delete an edge  $\rightarrow$

```
deleteEdge(vector<int> adj[i], int u, int v) {  
    for all neighbors of u
```

↳ Our Algorithm will create a disjoint set by creating a ~~set~~ new set for each

if  $v$  is a neighbor edge the connection

vertex. It will then loop through all the

for all neighbors of  $v$

edges of the given graph (tree  $T$ ) and

if  $u$  is a neighbor, erase the connection

for each edge it will connect the two sets

3  $\rightarrow O(v) \rightarrow$  linear time

if they are not already a part of the same

Set. If an edge attempts to connect two ~~sets~~ vertices that are already a part of the same set, we remove that edge because the two ~~sets~~ vertices are already connected as a part of a tree.

```
void Remove_Extra_Edge (Graph G(V,E))
```

$$v \in \text{for all } v \{ \text{make-set}(v) \} \rightarrow v \cdot O(1) = O(v)$$

$e \in \text{free}$  each edge  $(u, v)$  in  $E \rightarrow E \cdot \log(V)$

if  $\text{find-set}(u) \neq \text{find-set}(v) \rightarrow \text{log}(u)$

$$\text{unid}_n(u, v)$$

else

$$\text{delete-edge}(G, u, v) \rightarrow \alpha(v)$$

overall runtime:  $O(E \lg V)$

can beat a runtime of  $O(n^2)$

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- 6) Social Distancing : cell  $(i, j) = 1$  if some one is standing there  
 $\rightarrow$   $sd$  has to be <sup>smaller</sup> than manhattan distance b/w any 2 points w/ 1.

$\rightarrow$  An Adjacency matrix can be used to hold the grid.

$\rightarrow$  // vector < vector < int >> adj-matrix  $(n+1, \text{vector} < \text{int} > (n+1, 0))$ ;

If All the positions are given in a vector of pairs, loop through this vector and set the corresponding coordinates in the matrix to one.

/\* People-standing =  $[(1, 2), (3, 4), (7, 6), \dots]$

for (int i = 0; i < People-standing.size(); i++) {

adj-matrix [People-standing[i].first][people-standing[i].second] = 1;

} \*/

Given The above statements our algorithm would take the size of the matrix  $(n)$ , the (people-standing) vector, and the social distancing requirements ( $sd$ ), as its 3 argument inputs.

$\rightarrow$  initialize a vector of pairs to hold "sd violators" and an adjacency matrix } comments above

~~loop through the rows and columns of the adjacency matrix and if you encounter a [row][column] combo containing a 1, start a new set of for loops that run from row-sd to row+sd and column-sd to column+sd.~~

~~for (int row = 0; row < n; row++) {~~

~~for (int column = 0; column < n; column++) {~~

~~if (adj-matrix[row][column] == 1) {~~

// have a way to make sure

// negative indices aren't used

~~for (int row2 = row-sd; row2 <= row+sd; row2++) {~~

~~for (int column2 = column-sd; column2 <= column+sd; column2++) {~~

~~if (adj-matrix[row2][column2] == 1) {~~

~~sd-violators.push\_back((row, column), (row2, column2));~~

~~// if two people are too close record their coordinates~~

~~}~~

~~}~~

~~}~~

~~Runtime  $(n \times n \times sd \times sd) = O(n^2 \cdot sd^2)$~~

~~Space complexity  $(n \times n)$~~

$\rightarrow$  The sd-violator vector will hold any coordinates that a manhattan

loop through each row and for each row loop through the columns. If a 1 is encountered, inside the loops, loop through all the rows and columns again. Each time another 1 is encountered, calculate the manhattan distance b/w the second set of rows and columns and the first set. If sd happens to be bigger than these calculated values add both set of rows and columns to sd-violators. at the end, output the sd-violators vector of pairs as the coordinates of those points that violate the social distancing requirement.

```

for (row = 0 to n) {
  for (column = 0 to n) {
    if (adj-matrix[row][column] == 1) {
      for (row2 = 0 to n) {
        for (column2 = 0 to n) {
          if (adj-matrix[row2][column2] == 1) {
            if (sd > (abs(row2-row) + abs(column2-column))) {
              if ((row, column) not in sd-violators)
                sd-violators.push-back(row, column)
              if ((row2, column2) not in sd-violators)
                sd-violators.push-back(row2, column2)
            }
          }
        }
      }
    }
  }
}

```

RUNTIME:  $O(n \cdot n \cdot n \cdot n) = O(n^4)$

SPACE COMPLEXITY:  $O(n \times n)$