



\* Shortest Path in an Undirected graph

Simple BFS and updating distances would work here

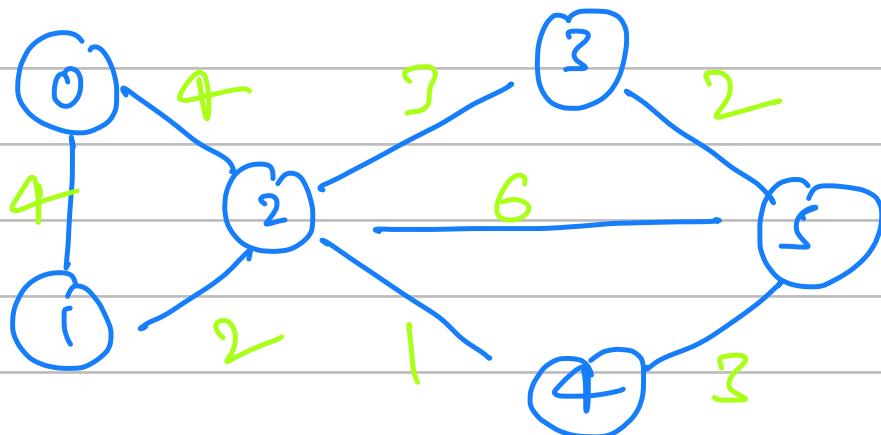
↳ This would not work for a DAG

→ you will need topo-order

This is a perfect()

\* Dijkstra Algorithm

→ Undirected graph



Path from 0 — 1 st

0 → 2 → 1      cost = 4 + 2 = 6

0 → 1      cost = 4



$$\text{Cost} = 4 + 3 + 2 + 6 + 2$$

### ① Statement

Given a source node find the min distance to every node

3 methods to do that

- ↳ Queue (BFS)
- ↳ Priority Queue
- ↳ Set

### ② Priority Queue approach

#### Initial Configuration

↳ Create a min-heap data structure

that stores  $\equiv \{\text{node}, \underline{\text{distance}}\}$

$\overline{T}$   
Criterion for  
min heap

↳ distance array  $\equiv []$

↳ Initialize the  $\text{dist}[\text{src}] = 0$

↳ put that into min-heap  $= [[\text{src}, 0]]$

→ Iteration

① takes out the node from priority queue  
node, dist =  $pq.pop()$

for every neighbor of node:

if  $dist[node] + \text{edge\_weight}$  is better:

$dist[neighbor] = dist[node] + \text{edge\_weight}.$  neighbor  
 $pq.add([neighbor, dist[neighbor]])$

② Return the distance array

```
from heapq import heapify, heappush, heappop  
class Solution:
```

```
# Function to find the shortest distance of all the vertices  
# from the source vertex S.  
def dijkstra(self, V, adj, S):  
    # Initial configuration  
    pq = []  
    distance = [float('inf')] * V  
    distance[S] = 0  
    heappush(pq, (0, S))  
  
    # Iteration  
    while len(pq) > 0:  
        d, u = heappop(pq)  
        for neighbor in adj[u]:  
            v, edge_weight = neighbor  
            if distance[v] > distance[u] + edge_weight:  
                distance[v] = distance[u] + edge_weight  
                heappush(pq, (distance[v], v))  
    return distance
```

Note → This would not work for negative weights on cycle that results in negative

Example



0	∞
0	1

when going from  $0 \rightarrow 1$   $\text{dist} = -$

0	-2
0	1

when going from  $1 \rightarrow 0$   $\text{dist} = -2 - 2 = -4$

-4	-2
0	1

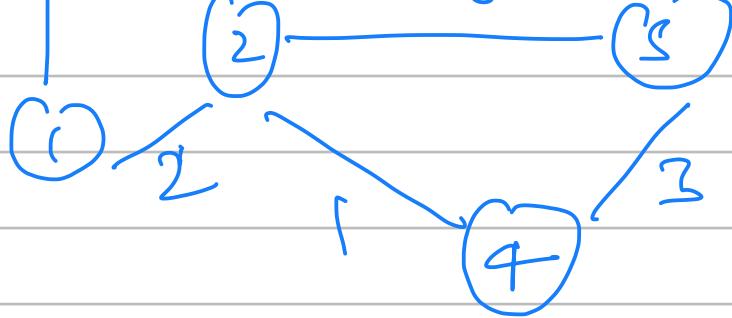
See the logic, Algorithm stuck in a loop

\* Dijkstra using Set data structure

→ Everything is same as priority queue  
except

Consider this snap shot





Distance  $\equiv$ 

0	4	4	7	5	10
---	---	---	---	---	----

0 1 2 3 4 5

Set = (10, 5) (7, 3) (5, 4)

A diagram illustrating a search or connection attempt between two nodes. A green curved arrow originates from the top left and points towards a blue circle labeled '4'. From node 4, a blue wavy line extends downwards and to the right, ending at another blue circle labeled '5'. To the right of this connection, the text 'checks for neighbour' is written in blue. Below node 4, the word 'node' is written in blue, and below node 5, the word 'current' is written in blue.

it says that someone has already reached  $s$   
with a distance of 10 but it has better

value of  $(S+3) = 8 \leq 10$

76mous the (1015) entry from the set

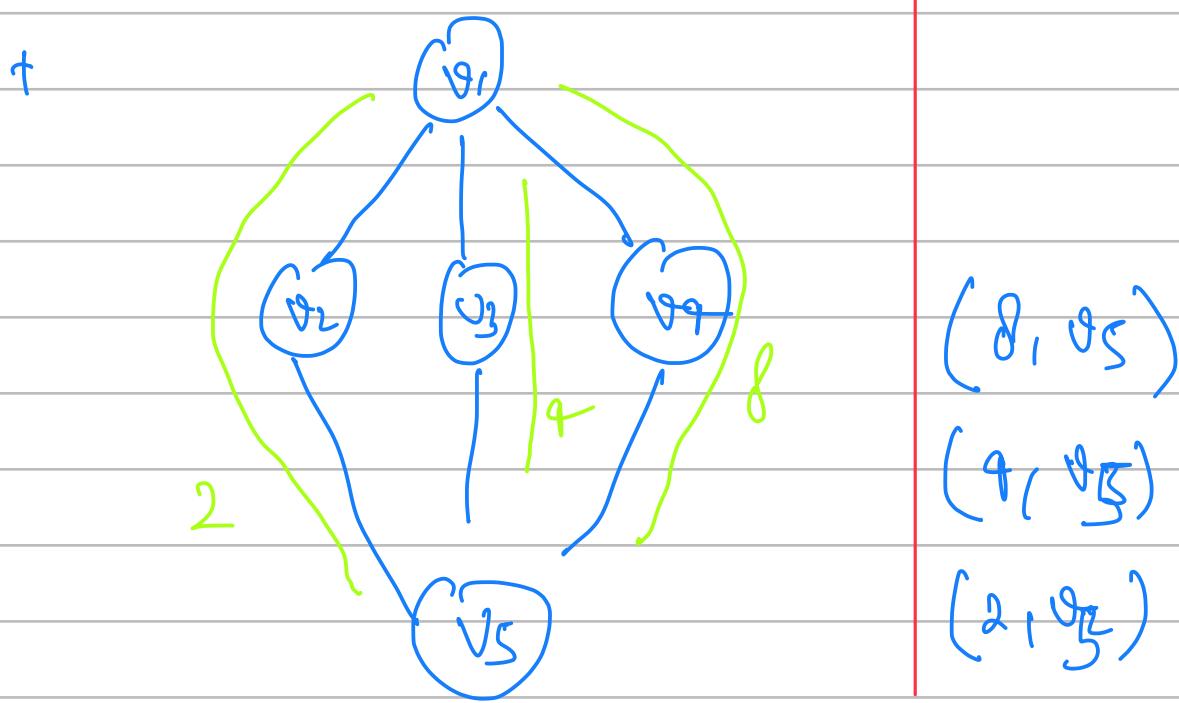
## On& Confined



# Intuition

- Why a Priority Queue instead of a Queue
- Queues will also work

\* Simple using DQOS and doing a BFS and updating distance as you encounter is just about same in which you consider all the paths



To reach  $v_5$

$(v_5)$  can be reached in three ways and with simple queue all the paths will be explored

= To eliminate this Dijkstra said why not use priority queues and consider the path with min distance first

in case  $(2, v_1)$  first which will

1.0 consider (a -> )

calculates all the paths with min time and  
will nullify future entries

= Using set idea  $\rightarrow$  On set if you found  
a better path removes the entries from set

Time complexity =  $E \cdot \underline{\log V}$

worst case scenario

\* Tell the shortest path b/w src and target

→ apply dijkstra but memoize where you come  
from

→ maintain a parent array

```

class Solution:
    def shortestPath(self, n: int, m: int, edges: List[List[int]]) -> List[int]:
        adj = []
        for i in range(n+1):
            adj.append([])
            # construct the adjacency list
        for egde in edges:
            u, v, w = egde
            adj[u].append([v, w])
            adj[v].append([u, w])
        # Initial configuraton
        distance = [float('inf')] * (n+1)
        parent = [-1] * (n+1)
        queue = []

        distance[1] = 0
        heappush(queue, (0, 1))

        # Iterate
        while len(queue) > 0:
            d, u = heappop(queue)
            for neighbor in adj[u]:
                v, w = neighbor
                if distance[v] > distance[u] + w:
                    distance[v] = distance[u] + w
                    heappush(queue, (distance[v], v))
                    parent[v] = u
        if distance[n] == float('inf'):
            return [-1]
        result = []
        p = n
        while p != -1:
            result.append(p)
            p = parent[p]
        result.append(distance[n])
        result.reverse()
        return result

```

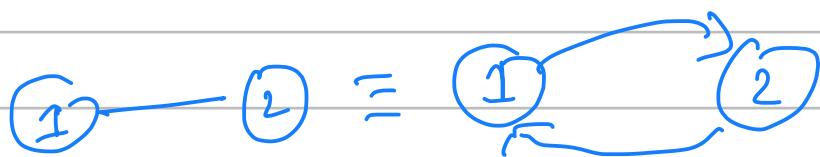
## \* Bellman Ford Algorithm

\* Shortest Path from src to all nodes

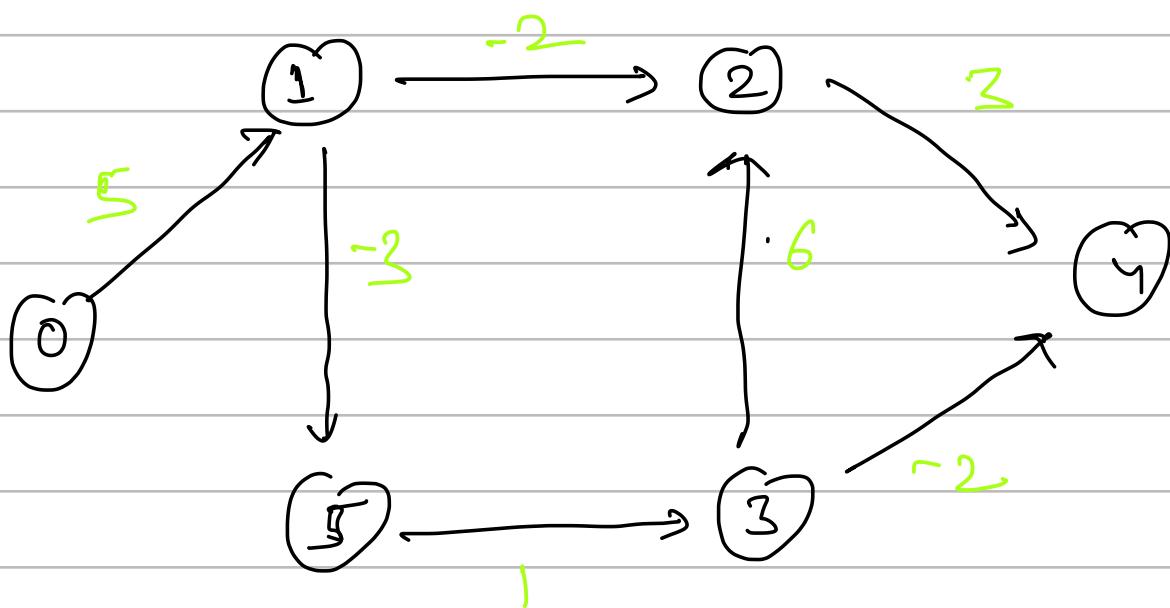
\* Dijkstra fails with negative edges and negative cycles

\* Applicable only on Directed graphs

(  
↳ to make it work on Undirected convert it to  
Directed)



\* Single source shortest path algo



① Relax all edges  $n-1$  times sequentially

Relax  $\equiv$   $v \xrightarrow{wt} u$

if  $\text{distance}[v] + wt < \text{distance}[u]$ :

// got a better path

$\text{distance}[v] = \text{distance}[u] + wt$

② initial configuration

distance =  $[\infty]$  at

distance [src] = 0

- (3) So there are  $(n^2)$  iterations or  
 $(n^2)$  iteration needs to be done

In each iteration you have to go over  
all the edges each time

- \* If you haven't reached a node in an  
iteration ignore it

### A) Intuition

We know the thought process but the question is  
why  $(n^2)$  iterations = will it update all  
the nodes's distance

- \* Edges can be given in any order

Consider this



$$\text{dist} = [0 \infty \infty \infty \infty]$$

1st iteration

Edges

3	4	1
2	3	1

1	2	1
---	---	---

$\text{dist}[3] + 1 < \text{dist}[4]$  — nothing  
 $\text{dist}[2] + 1 < \text{dist}[3]$  — nothing  
 $\text{dist}[1] + 1 < \text{dist}[2]$  — nothing

0 1 1

$$\text{dist}[0] + 1 < \text{dist}[1] \equiv [0 \ 1 \ \infty \ \infty \ \infty]$$

2nd iteration

$$\text{dist}[3] + 1 < \text{dist}[4] \times$$

$$\text{dist}[2] + 1 < \text{dist}[3] \times$$

$$\text{dist}[1] + 1 < \text{dist}[2] \equiv [0 \ 1 \ 2 \ \infty \ \infty]$$

$$\text{dist}[0] + 1 < \text{dist}[1] \times$$

⋮

→ and so on

If the order of edges has been given in below order

0 1 1  
1 2 1  
2 3 1  
3 4 1

On a single iteration all the min dist would have been calculated

This is the intuition of running iterations (n-1) times

(\*) How to detect a negative cycle

We know  $(n-1)$  iteration will update the distances completely

If on  $n^{\text{th}}$  iteration distance array updates for any node then there is a negative cycle.

class Solution:

```
# Function to construct and return cost of MST for a graph  
# represented using adjacency matrix representation
```

```
'''
```

```
V: nodes in graph
```

```
edges: adjacency list for the graph
```

```
S: Source
```

```
'''
```

```
def bellman_ford(self, V, edges, S):
```

```
    inf = 10**8
```

```
    # Initial configuration
```

```
    distance = [inf] * V
```

```
    distance[S] = 0
```

```
    # Perform V-1 iterations
```

```
    for _ in range(V-1):
```

```
        # Go Through each edge
```

```
        for edge in edges:
```

```
            u, v, wt = edge
```

```
            if distance[u] != inf and distance[u] + wt < distance[v]:
```

```
                distance[v] = distance[u] + wt
```

```
# check for negative cycle by running Vth iteration
```

```
for edge in edges:
```

```
    u, v, wt = edge
```

```
    if distance[u] != inf and distance[u] + wt < distance[v]:
```

```
        return [-1]
```

```
return distance
```

(\*) Floyd Warshall Algorithm

Bellman Ford and Dijkstra were single

Source shortest path algo

→ This is different

- multi source shortest path algo
- helps detect negative cycles well

Thought process →

Go via buoyy node & calculate

- use adjacency matrix to represent the graph
- initialize a cost matrix of the same size

\*Initial configuration

- ① store buoyy edges into cost matrix
- ②  $(i, i) = \text{store } 0$  as cost of itself  
will always be 0

- ③ iterate through every vertex  $0 \dots n-1$   
and do

$\text{via} = 0$  to  $\text{via} = n-1$ :

for  $i=0$  to  $n-1$ :

for  $j=0$  to  $n-1$ :

loop through  
cost matrix

$$\text{cost}[i][j] = \min(\text{cost}[i][j],$$

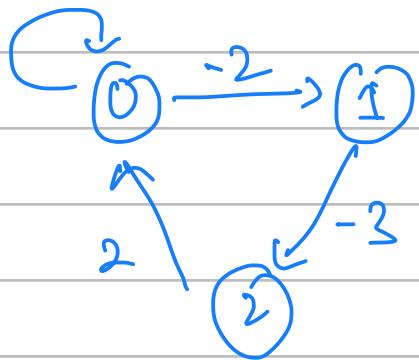
$$\text{cost}[i][\text{via}] + \text{cost}[\text{via}][j])$$

④ The resultant cost matrix will give shortest path from every vertex to another

The algorithm is not much intuitive as the other ones. It is more of a brute force, where all combination of paths have been tried to get the shortest paths.

Nothing to be panic much on the intuition, it is a simple brute on all paths. Focus on the three for loops.

④ How to detect a negative cycle



Cost of each vertex to itself should always be 0 but for negative cycles the cost to itself will be calculated as  $\infty$

for  $i=0$  to  $n$ :

if  $\text{cost}[i][i] < 0$ :

cyclic detected

```

class Solution:
    def shortest_distance(self, matrix):
        n = len(matrix)
        # Initial configuration
        # Initialize a cost matrix
        cost = []
        for _ in range(n):
            temp = [float('inf')]*n
            cost.append(temp)
        # u->u should be 0 and direct edges cost u->v in adj matrix
        for i in range(n):
            for j in range(n):
                if matrix[i][j] == -1: # no edge
                    cost[i][j] = float('inf')
                else:
                    cost[i][j] = matrix[i][j]

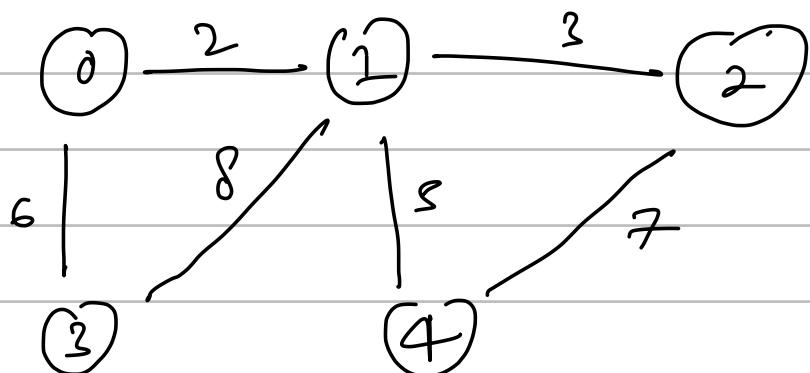
        # Iterate through every vertex 'via' and calculate the costs
        for via in range(n):
            # loop through the cost matrix and update
            for i in range(n):
                for j in range(n):
                    cost[i][j] = min(cost[i][j], cost[i][via] + cost[via][j])
        # update -1 for v which are not reachable from u
        for i in range(n):
            for j in range(n):
                if cost[i][j] == float('inf'):
                    cost[i][j] = -1

        return cost

```

\* Minimum Spanning Tree (MST)

undirected graph

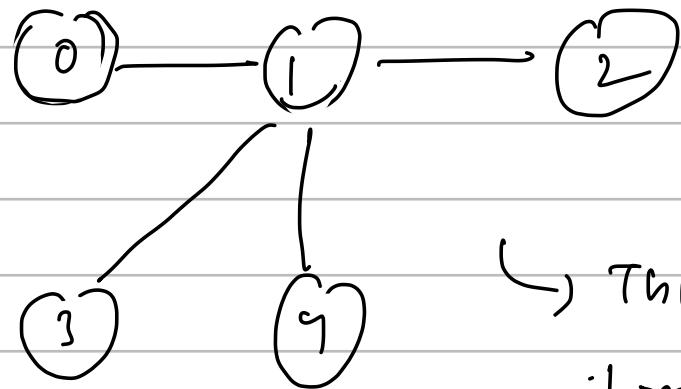


A tree in which we have  $n$  nodes and  $n-1$

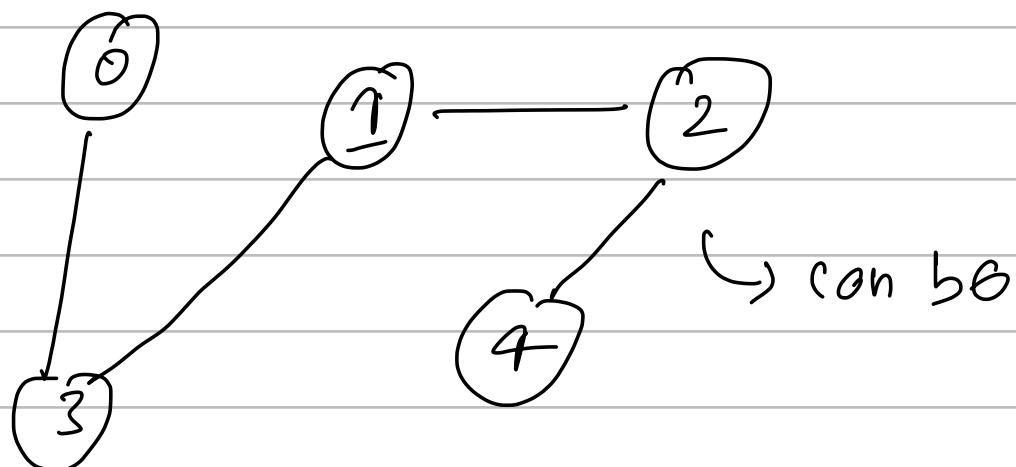
edges = no cycles

all nodes are reachable

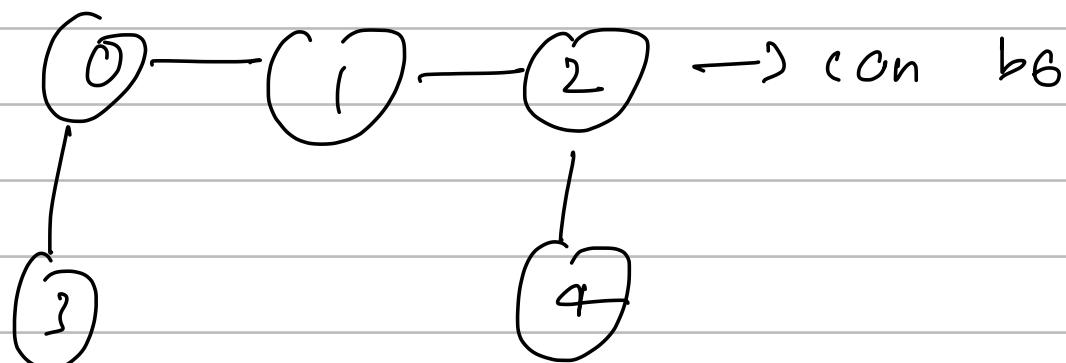
If a graph contains  $n$  vertices and  $n-1$  edges and is connected it cannot have any cycles



↳ This can be a spanning tree

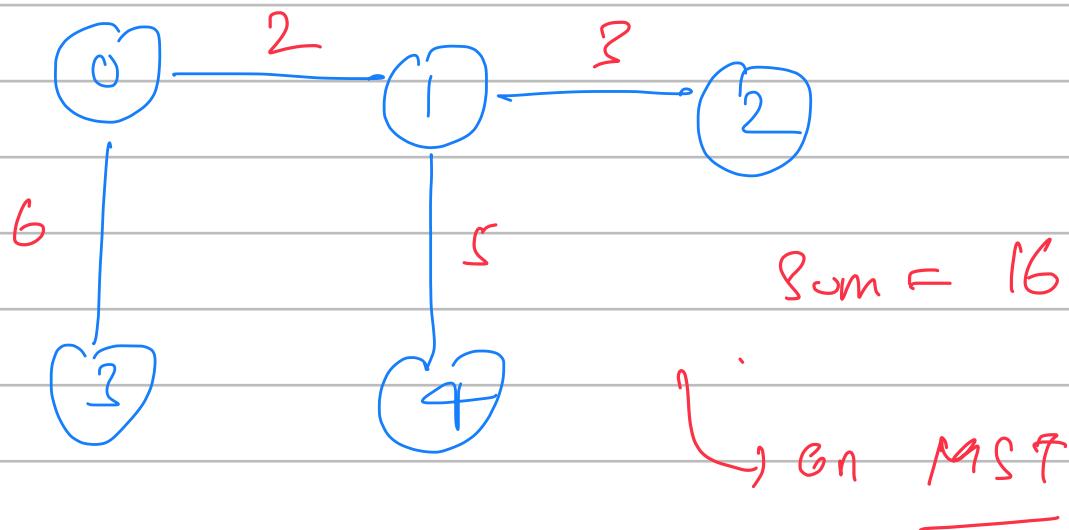


↳ can be



These are all spanning trees

\* The spanning tree which has the least or the minimum sum will be MST of a given graph



There can also be multiple MSTs for a graph

\* Prim's Algorithm

helps to find MST for a graph

Required 2 things

{ Priority queue  
Visited array }  
MST list = []      { Initial Configuration }

\* Initial Configuration



Priority queue  $\equiv$  (wt, node, parent)

visited  $\equiv [ ]$

MST  $\equiv [ ]$

- ① Can start with any node

### Iteration

- ① Take the element out of PQ.

$$(\text{wt}, \text{v}, \vartheta) \equiv \text{PQ.pop()}$$

- ② if  $v$  is already visited continue

- ③ if  $v$  is not visited :

- a) mark  $v$  visited

- b) put  $(v - \vartheta)$  edge in MST

- c) Update the sum = sum + wt

- d) iterate through children of  $v$

- \* if neighbour is not visited

- \* push them to the PQ

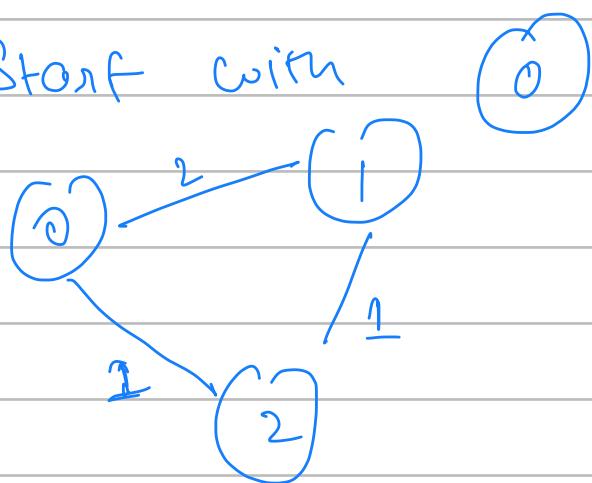
```
from heapq import heapify, heappush, heappop
class Solution:
```

```
#Function to find sum of weights of edges of the Minimum Spanning Tree.
def spanningTree(self, V, adj):
    # Initial configuration
    visited = [False] * V
    MST = []
    pque = [] # type (wt, u, parent)
    heappush(pque, (0, 0, -1))
    sum = 0

    # Iteration
    while len(pque) > 0:
        wt, u, parent = heappop(pque)
        if visited[u] is True:
            continue
        visited[u] = True
        sum += wt
        MST.append((u, parent))
        for neighbor in adj[u]:
            v, wt = neighbor
            if visited[v] is False:
                heappush(pque, (wt, v, u))
    return sum
```

## \* Intuition

Start with



(0) pushed (1) and (2)  
in pque and in mst

iteration it took min of

the 2 edges which is (0) — (2)

now as soon as you take out (2) from pq



2

it pushed  $\textcircled{2} \xrightarrow{\perp} \textcircled{1}$  which resulted in not  
considering  $\textcircled{0} \longrightarrow \textcircled{1}$

(\*) Greedy is the intuition here

(\*) TC  $\rightarrow O(\Sigma \log \Sigma)$

(\*) Disjoint Set Data Structure

→ Traversing through DFS or BFS costs  
 $O(V+E)$

→ for dynamic graphs use DS

Two methods

→ find Parent → Path compression  
→ union → Rank  
→ size

Initially

Every node is alone

(1) (2) (3) (4) (5) (6) (7)

Union will connect two vertices

\* Union by rank

rank  $\in [ ]$

parent  $\in [ ]$

UNION ( $v, w$ ):

① find parent of  $v$  and  $w \in P_1$  and  $P_2$

② find rank of  $P_1$  and  $P_2 \in r_1$  and  $r_2$

③ Attach the smaller rank to larger rank

FIND-PARENT ( $v$ ):

if parent [ $v$ ] ==  $v$ :

return  $v$

parent [ $v$ ] = FIND-PARENT (parent [ $v$ ])

return parent [ $v$ ])

$T \in O(N \alpha)$

Path compression technique

\* Union by size technique

$\text{Size} = [1, 1, 1, 1, \dots]$

$\text{parent} = [0, 1, 2, 3, \dots]$

Consider edges

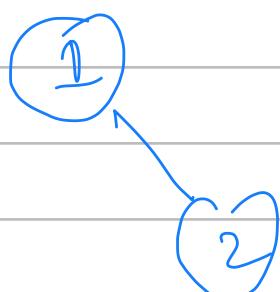
(1, 2)

(2, 3)

(4, 5)



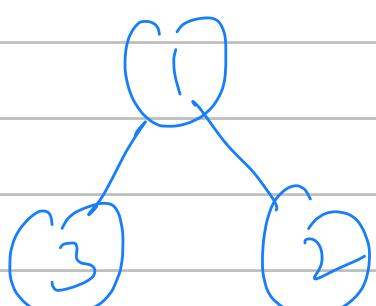
Union (1, 2)



$\text{Size} = [2, 1, 1, 1, \dots]$

$\text{parent} = [1, 1, 3, 4, 5, \dots]$

Union (2, 3)

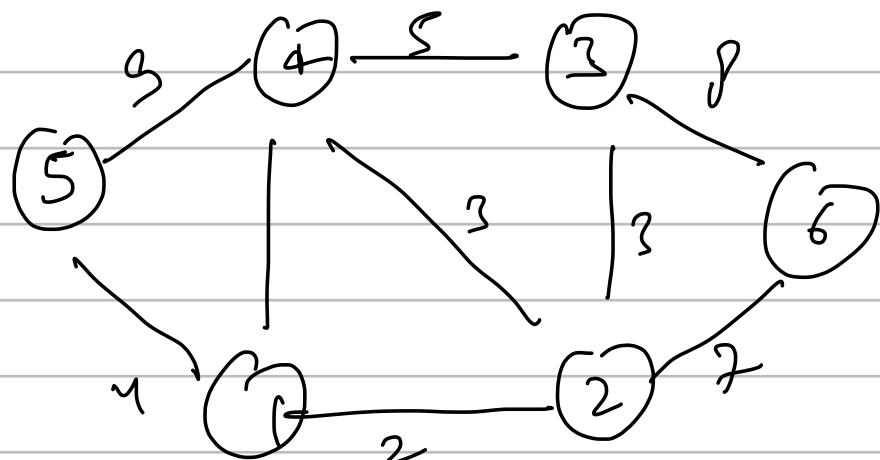


$\text{Size} = [3, 1, 1, 1, \dots]$

$\text{parent} = [1, 1, 1, 4, 8, \dots]$

\* Kruskal's algorithm (Disjoint Set)

- ① Sort all the edges according to weights
- ② Initialize a disjoint set data structure with the given nodes



$\text{DS} = \textcircled{1} \textcircled{2} \textcircled{3} \textcircled{4} \textcircled{5} \textcircled{6}$   
 (Edges sorted)

wt	U	V
1	1	4
2	1	2
3	2	3
3	2	4
4	1	3
5	3	4
7	2	6
8	3	6
9	4	5

- ③ Pick the edge from by one and add it to

the Disjoint Set and calculate the total weight

↳ if  $\text{① } \text{②}$  already connected then ignore the edge as this will only create a cycle.

## \* Disjoint Set implementation

```
class DisjointSet:  
    def __init__(self, n):  
        self.parent = []  
        self.rank = []  
        for i in range(n+1):  
            self.parent.append(i)  
            self.rank.append(0)  
  
    def getParent(self, u):  
        if self.parent[u] == u:  
            return u  
        self.parent[u] = self.getParent(self.parent[u])  
        return self.parent[u]  
    def union(self, u, v):  
        p_u = self.getParent(u)  
        p_v = self.getParent(v)  
  
        if p_u == p_v:  
            return  
        if self.rank[p_u] > self.rank[p_v]:  
            self.parent[p_v] = p_u  
        elif self.rank[p_u] < self.rank[p_v]:  
            self.parent[p_u] = p_v  
        else:  
            self.parent[p_v] = p_u  
            self.rank[p_u] += 1  
    def connected(self, u, v):  
        if self.getParent(u) == self.getParent(v):  
            return True  
        return False
```

## \* Kruskal's implementation

```
class Solution:
```

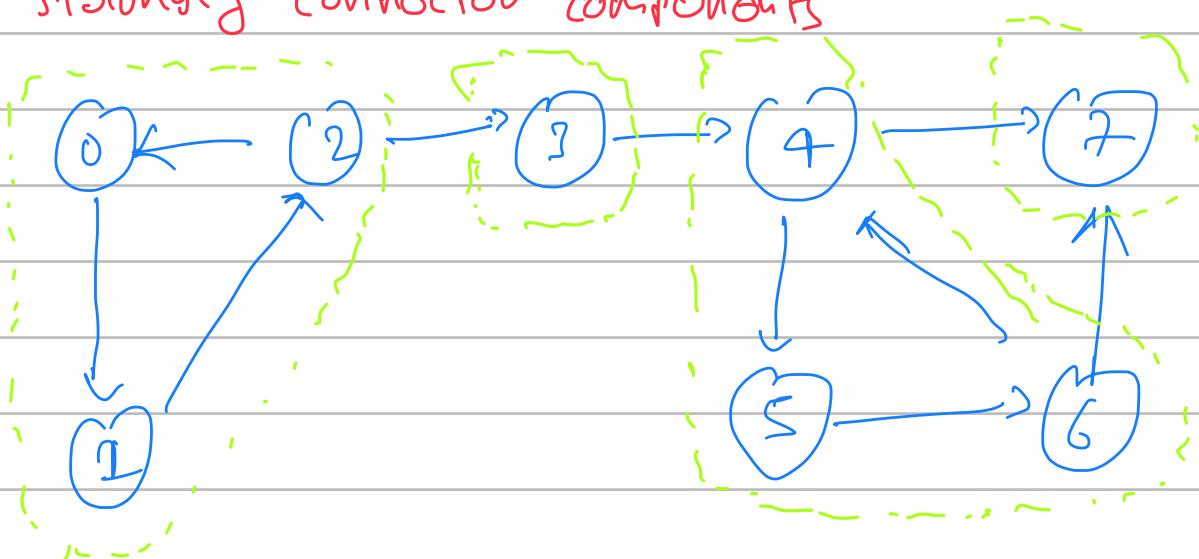
```
    # Function to find sum of weights of edges of the Minimum Spanning Tree.
    def spanningTree(self, V, adj):
        # Initial configuration
        # 1. Get the edges
        edges = []
        for i in range(len(adj)):
            for neighbor in adj[i]:
                u = i
                v, wt = neighbor
                edges.append([wt, u, v])
        # 2. Sort the edges
        sorted_edges = sorted(edges, key=lambda x: x[0])
        # 3. Create a disjoint set data structure
        ds = DisjointSet(V)

        weight = 0
        # Go through every edge
        for edge in sorted_edges:
            wt, u, v = edge
            if ds.connected(u, v) == False:
                weight += wt
                ds.union(u, v)
        return weight
```

## \* Kosaraju's Algorithm

Note → Strongly Connected Comps are only valid  
for Directed Graphs

→ Strongly Connected Components



Consider a  $\textcircled{v}$  and  $\textcircled{w}$  if in a graph

you can reach from  $v \rightarrow w$  and you can also reach from  $w \rightarrow v$  then that set of vertices are known as strongly connected

Component

$\rightarrow 0 \ 1 \ 2$

$\rightarrow 3$

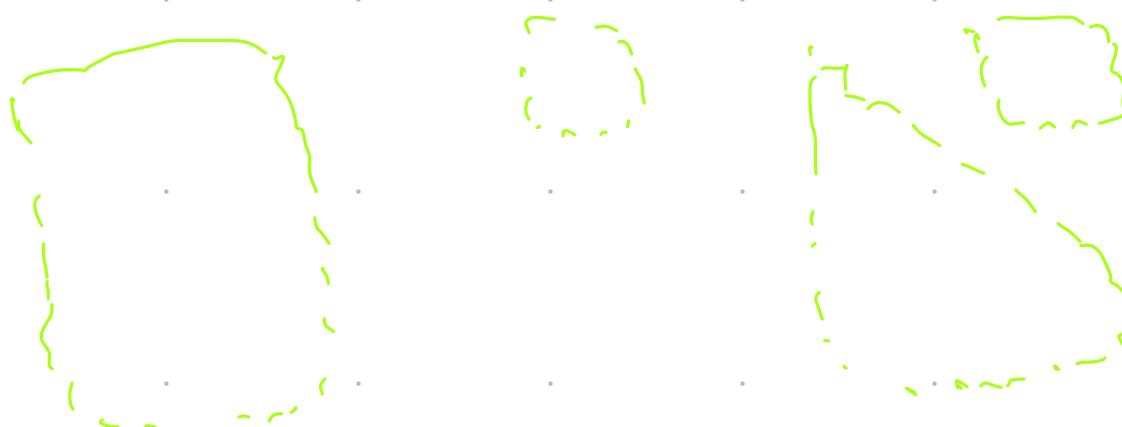
$\rightarrow 4 \ 5 \ 6$

$\rightarrow 7$

## \* Basic intuition

if you apply DFS it will visit all nodes and that we don't require.

We can divide the graph as follows



SCC 1  $\longrightarrow$  SCC 2  $\longrightarrow$  SCC 3  $\longrightarrow$  SCC 4

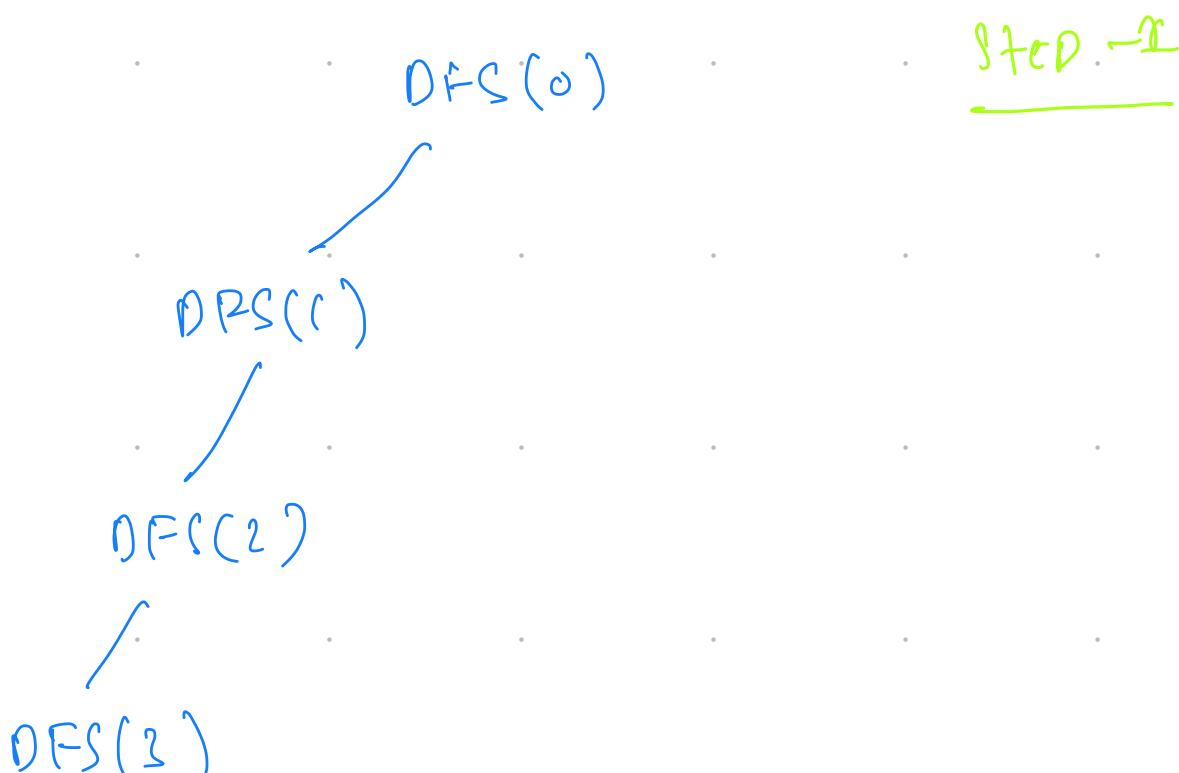
\* Now if you removes every edge SCC won't change you will still be able to travel it but you won't be able to go to SCC 2 from SCC 1

\* Kosaraju steps

1  $\rightarrow$  Sort all the edges according to finishing time

2, Reverse the graph

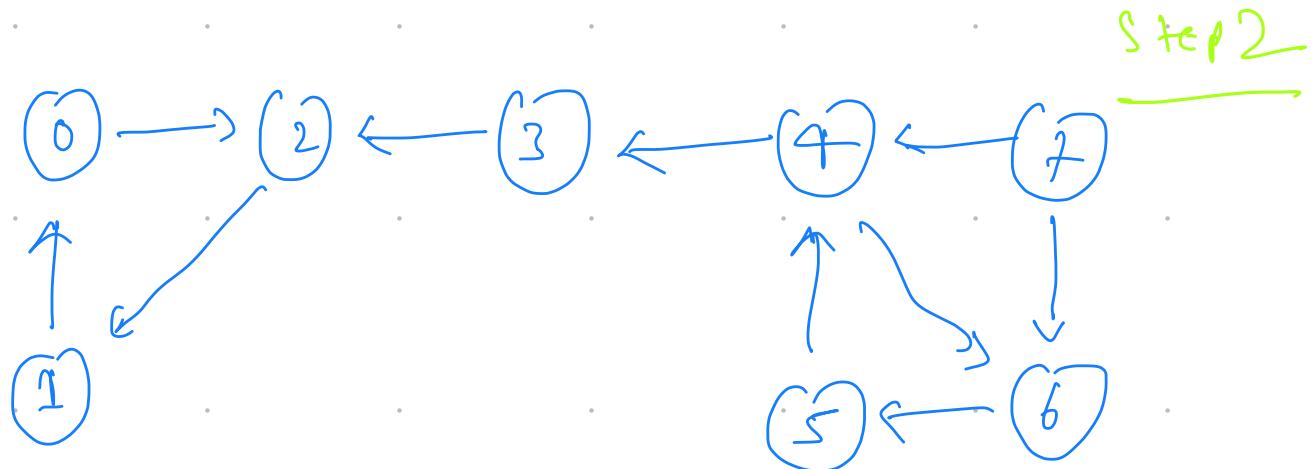
3  $\rightarrow$  Do a DFS





{ 7 is the first guy and  
 is the last portion in the  
 SCC }

Store the vertices when you back track



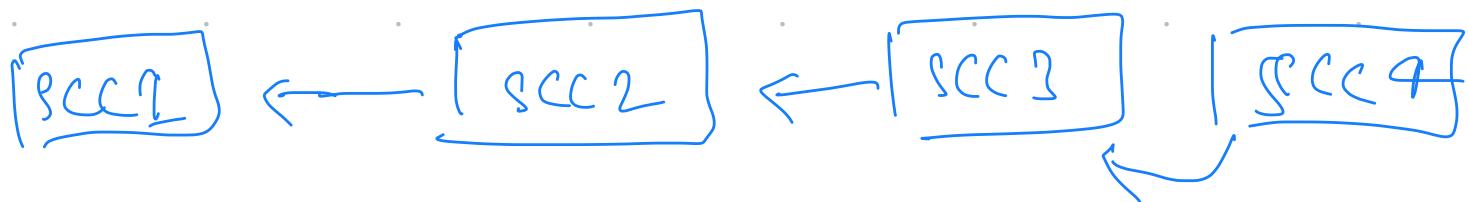
Step -3

Initially if you did a DFS on the original

graph you will do something



Now since we have traversed all the edges



we want to do a DFS of SCC1 first

as doing a DFS of any other SCC would  
reach the previous one

\* this order we can get from the finishing  
time stack that we got in Step 1

```
def DFS_1(self, u, V, adj, visited, order):
    visited[u] = True
    for v in adj[u]:
        if visited[v] == False:
            self.DFS_1(v, V, adj, visited, order)
    order.append(u)

def DFS_3(self, u, adj, visited, scc):
    visited[u] = True
    scc.append(u)
    for neighbor in adj[u]:
        if visited[neighbor] == False:
            self.DFS_3(neighbor, adj, visited, scc)
```

class Solution:

```
#Function to find number of strongly connected components in the graph.
def kosaraju(self, V, adj):
    visited = [False] * V

    # Step-1 sort all the vertices w.r.t finishing order
    order = []
    for i in range(V):
        if visited[i] == False:
            self.DFS_1(i, V, adj, visited, order)

    # Step-2 Reverse all the edges
    adjT = []
    for _ in range(V):
        adjT.append([])
    for u in range(V):
        visited[u] = False
        for v in adj[u]:
            adjT[v].append(u)

    # Step-3 Perform DFS on finishing order
    scc_s = [] # strongly connected comps
    while len(order) > 0:
        u = order.pop()
        if visited[u] == False:
            scc = []
            self.DFS_3(u, adjT, visited, scc)
            scc_s.append(scc)
    return len(scc_s)
```

\* Bridges in graph

Any edge whose removal breaks down the graph into two components or more components is a Bridge.

