

# Graph Traversal Algorithms

## → Breadth-First Search (BFS)

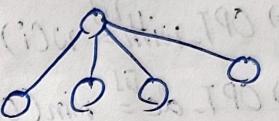
Given a graph  $G(V, E)$  & source vertex  $s$ , BFS "discovers" every vertex reachable from  $s$

BFS colours nodes using 3 colours

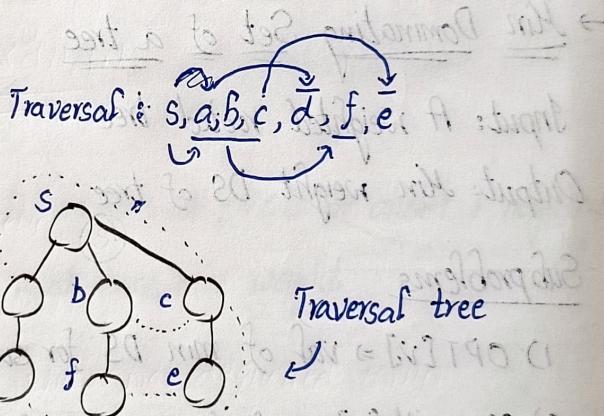
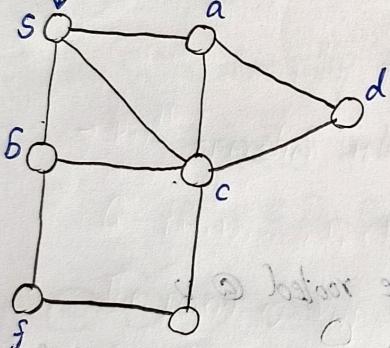
↳ White: undiscovered node

↳ Grey: Discovered node with white neighbours

↳ Black: Discovered node with neighbours either black or grey



Ex: Graph



\* Edges in a traversal tree are tree edges

\* Non-tree edges can connect nodes who are in the same level or one level apart

↳ If the edge connects nodes 2 levels apart, the bottom node must have been discovered ~~before~~ by the top node itself making the level difference as 1.

## Algorithm

BFS( $G, s$ ) {

for each vertex  $u$  in  $v(G) - \{s\}$  &:

$\text{colour}(u) = \text{white} \rightarrow \text{colour of vertex}$

$d[u] = \infty \rightarrow \text{distance from } s$

$\pi[u] = \text{NIL} \rightarrow \text{parent node in traversal tree}$

} preprocessing

$\text{colour}(s) = \text{grey}$

$d[s] = 0$

$\pi[s] = \text{NIL}$

$O(n)$

$Q = \emptyset \rightarrow \text{queue of grey nodes.}$

Enqueue( $Q, s$ )

while  $Q \neq \emptyset$ :

$u = \text{dequeue}(Q)$

for  $v$  in  $\text{adjlist}[u]$ :

if  $\text{colour}[v] = \text{white}$ :

$\text{colour}[v] = \text{grey}$

Enqueue  $d[v] = d[u] + 1$

$\pi[v] = u$

Enqueue( $Q, v$ )

$\text{colour}[u] = \text{black}$

\* Every vertex is enqueued & dequeued once  $\Rightarrow O(n)$

\* For every vertex, we traverse through all edges connected to it

↳ Total time taken  $\Rightarrow O(m)$

Overall:  $O(n+m)$  [ $n \Rightarrow \text{nodes}$ ,  $m \Rightarrow \text{edges}$ ]

Properties of BFS

① Shortest path

$d[v]$  holds the value of shortest path from  $s$  to  $v$

↳ To prove this, we need property 2.

② Suppose that during execution of BFS on graph  $(V, E)$ , the queue  $Q$  contains  $\{v_1, v_2, \dots, v_r\}$  where  $v_1$  is head of  $Q$  &  $v_r$  is tail, then  
 $\Rightarrow d[v_i] \leq d[v_{i+1}] \quad \Rightarrow d[v_1] + 1 \geq d[v_r]$   
 $1 \leq i \leq r-1$

#### ↳ Proof by induction on queue operations

Base case: 1 operation  $\Rightarrow$  Enqueue  $\Rightarrow Q = \{s\}$

↳ Trivially true

Hypothesis: True for  $k$  operations

$$Q = \{v_1, v_2, \dots, v_r\}$$

Inductive step:  $k+1^{\text{th}}$  operation

$\rightarrow k+1^{\text{th}}$  operation is dequeue  $\Rightarrow v_i$  is removed

We know that  $d[v_r] \leq d[v_i] + 1 \leq d[v_2] + 1$  ( $d[v_i] \leq d[v_2]$ )

$$\underline{\underline{d[v_r] \leq d[v_2] + 1}}$$

Proved

$\rightarrow k+1^{\text{th}}$  operation is enqueue

$$Q = \{v_1, v_2, \dots, v_r, v_{r+1}\}$$

The parent of  $v_{r+1}$  is  $u$  which was dequeued earlier

↳  $d[u] \leq d[v_i]$  &  $d[v_{r+1}] = d[u] + 1$

$$\underline{\underline{d[v_{r+1}] \leq d[v_i] + 1}}$$

Now we need to show  $d[v_r] \leq d[v_{r+1}]$

As  $d[v_r] \leq d[u] + 1$  (Before  $u$  was dequeued)

$$d[v_r] \leq d[v_{r+1}] \quad (d[u] + 1 = d[v_{r+1}])$$

Hence proved

## Proof of property ①

$\delta(u, v) = \text{shortest path from } u \text{ to } v$ . We need to show  $d[v] = \delta(s, v)$

Let  $u, v \in V[G]$ ,  $\& (u, v) \in E[G]$ ,  $\delta(s, v) \leq \delta(s, u) + 1$

Clearly,  $d[v] \geq \delta(s, v)$

↳ If  $d[v] < \delta(s, v)$ , there exists a shorter path from  $s$  to  $v$  than  $\delta(s, v)$

↳ Contradiction

Claim:  $d[v] = \delta(s, v)$

Let  $v$  be the first vertex from  $s$  where  $d[v] > \delta(s, v)$  [All prior vertices follow  $d[u] = \delta(s, u)$ ]

Say  $(u, v)$  is the last edge in shortest path from  $s$  to  $v$

$$d[v] > \delta(s, v) = \delta(s, u) + 1 \Rightarrow d[u] + 1$$

↳  $d[v] > d[u] + 1 \Rightarrow$  Contradiction

When  $u$  is dequeued,  $v$  can't be black as  
 ~~$u$  is neighbour  $\Rightarrow v$  can't be black~~

if  $v$  was white,  $d[v] = d[u] + 1$

if  $v$  was grey,  $d[v] = d[u] + 1 \leq d[u] + 1$

Where  $w$  is the node that discovers  $v$  before  $u$ .

↳ By prop ②,  $d[w] \leq d[u]$  as

$w$  comes before  $u$

## Depth-First Search

$\pi[v]$   $\Rightarrow$  predecessor of  $v$

$d[v]$   $\Rightarrow$  time when  $v$  is discovered (when  $v$  turns grey)

$f[v]$   $\Rightarrow$  time when all neighbours of  $v$  are visited (when  $v$  turns black)

$\text{DFS}(G)$ :

for each vertex  $u \in V[G]$ :

$\text{colour}[u] = \text{white}$

$\pi[u] = \text{NIL}$

$\text{time} = 0$

for each vertex  $u \in V[G]$ :

if  $\text{colour}[u] = \text{white}$ :

$\text{DFS\_visit}(u)$

$\text{DFS\_visit}(u)$ :

$\text{colour}[u] = \text{grey}$

$\text{time} = \text{time} + 1$

$d[u] = \text{time}$

for each vertex  $v \in \text{adjList}[u]$ :

if  $\text{colour}[v] = \text{white}$ :

$\pi[v] = u$

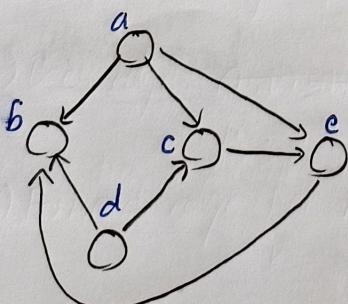
$\text{DFS\_visit}(v)$

$\text{colour}[u] = \text{black}$

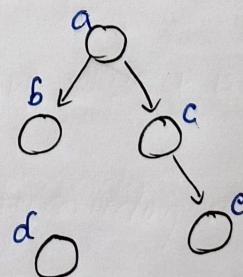
~~$\text{time} = \text{time} + 1$~~   $f[u] = \text{time}$

$f[u] = \text{time}$   $\text{time} = \text{time} + 1$

Eg:



Traversal tree :



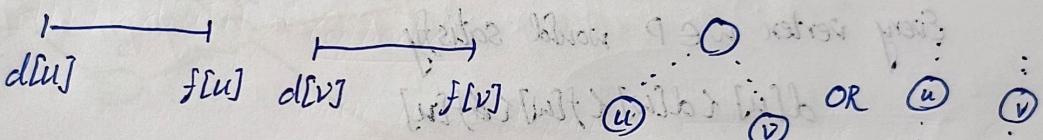
time: 1 2 3 4 5 6 7 8 9 10  
 node: a b-b c e-e c a d-d  
 $d[a] d[b] f[b] d[e] d[e] f[e] f[e] f[a] d[d] f[d]$

### Properties of DFS

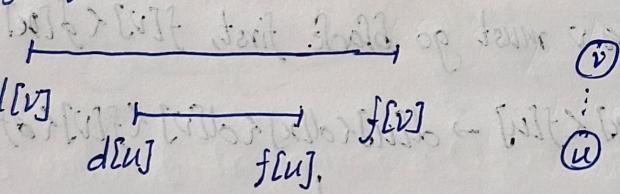
- ① We get multiple trees in DFS. They form a forest.
- ②  $\pi[v] = u$  if & only if  $\text{DFS.visit}(v)$  was called during a search of  $u$ 's adjList.
- ③ Parenthesis Form

In any DFS of a directed or undirected graph  $G = (V, E)$ , for any 2 vertices  $u \in V$ , exactly 1 of the 3 conditions hold:

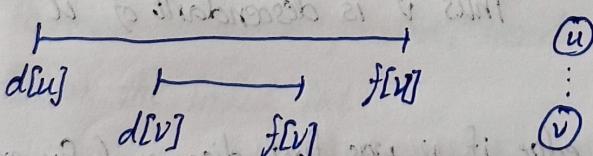
- 1) Intervals  $[d[u], f[u]] \cap [d[v], f[v]]$  are completely disjoint. Neither  $u$  nor  $v$  is a descendant of the other in the DFS forest.



- 2)  $[d[u], f[u]]$  is entirely contained in  $[d[v], f[v]]$  &  $u$  is descendant of  $v$ .  
 Same tree. Diff trees



- 3)  $[d[v], f[v]]$  is entirely contained in  $[d[u], f[u]]$  &  $v$  is descendant of  $u$ .



Provable by simple proof by contradiction

- 4) If  $v$  is a proper descendant of  $u$  if & only if  $d[u] < d[v] < f[v] < f[u]$

## ⑤ White Path Thm

$v$  is descendant of  $u$  iff at the time  $d[u]$ ,  $v$  can be reached from  $u$  along a path consisting entirely of white vertices.

### Proof

Assume  $v$  is descendant of  $u$

↳ This means there is a path from  $u$  to  $v$ , in traversal tree. This path would be white (undiscovered) when we first discover  $u$  (at time  $d[u]$ )

Assume there exists a white path from  $u$  to  $v$  in  $G$  at  $d[u]$  time

↳ Let path be  $P$

Assume  $v$  is not descendant of  $u$

However every vertex in  $P$  would be a descendant of  $u$

Every vertex  $w \in P$  would satisfy

$$d[u] < d[w] < f[w] < f[u]$$

Say  $x$  is the prev vertex of  $v$  in the path  $P$ .

As  $x$  would be discovered first,  $d[x] < d[v]$

For  $x$  to go black,  $v$  must go black first,  $f[v] < f[x]$

$$\text{As } d[u] < d[x] < f[x] < f[u] \rightarrow d[u] < d[x] < d[v] < f[v] < f[x] < f[u]$$

$$d[u] < d[v] < f[v] < f[u]$$

Thus  $v$  is descendant of  $u$

## ⑥ Classification of Edges

1) Tree edges :  $(u, v)$  is a tree edge if  $v$  was first discovered by exploring edge  $(u, v)$

2) Back edge :  $(u, v)$  where Non-tree edge  $(u, v)$  where  $v$  is ancestor of  $u$

3) Forward edge: Non-tree edge  $(u, v)$  connecting  $u$  to descendant  $v$

4) Cross Edge: Edge Non-tree edge b/w diff DFS trees or edges b/w vertices in the same tree, when they aren't descendant / descendant of one-another.

Basically all other edges apart from 1, 2, & 3

~~Ex~~

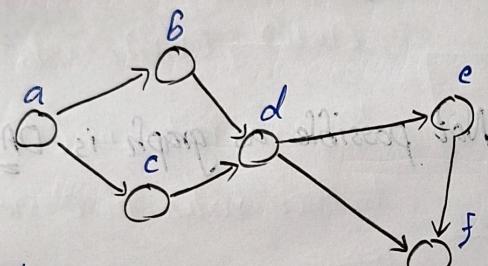
## → Applications of BFS & DFS

### ① Topological Sort

directed graph (directed acyclic graph)

A topological sort of a DAG  $G(V, E)$  is a linear ordering of vertices of  $G$  s.t. if  $G$  contains an edge  $(u, v)$ , then  $u$  appears before  $v$  in the ordering.

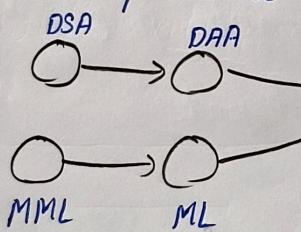
Eg:



a, b, c, d, e, f

\* Used to show precedence among events

Eg:



DSA, MML, ML, DAA, NLP

Logic

- ⇒ Call DFS( $G$ ) to compute  $f[v]$  for each  $v$
- ⇒ As each vertex is finished, insert it to the front of a linked list
- ⇒ Return the linked list

\* A directed graph  $G$  is acyclic iff. a DFS of  $G$  yields no back edges.

↳ Proof

Assume  $(u, v)$  is a back edge

↳ Path from  $v$  to  $u$  in DFS tree exists. With  $(u, v)$  we get a cycle. Contradiction as  $G$  is acyclic.

Assume cycle  $C(v_1, v_2, \dots, v_k, v_1)$  exists

Let  $v_1$  be the first vertex in  $C$  to be discovered.

At  $d[v_1]$ ,  $v_2 - v_3 - \dots - v_k$  are all white.

There is a white path  $v_1 \rightarrow v_k \rightarrow v_1$  is descendant of  $v_1$ .

$(v_k, v_1)$  is a back edge

### Proof of correctness of Topological Sort logic

If  $(u, v) \in E$ , then  $f(u) > f(v)$

When the edge  $(u, v)$  is explored

Case 1)  $v$  is white

$v$  is descendant of  $u \Rightarrow f(u) > f(v)$

Case 2)  $v$  is grey

$(u, v)$  is back edge  $\Rightarrow$  Not possible as graph is DAG

Case 3)  $v$  is black

$f(v) < f(u)$

### ② Strongly Connected Components

This makes sense in directed graphs only

A SCC of a directed graph  $G(V, E)$  is maximal set of vertices  $C \subseteq V$  st for every pair of vertices  $u \notin C$ ,  $v \in C$ ,  $v$  is reachable from  $u$  &  $u$  is reachable from  $v$ .

Algorithm

Logic used

1) Call DFS( $G$ ) & compute  $f[u]$  for every vertex  $u$

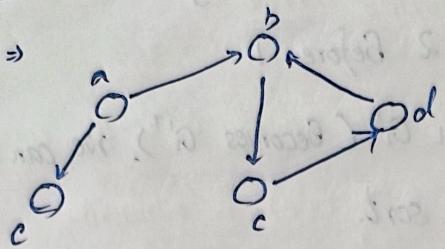
2) Compute  $G^T \Rightarrow G^T$  is the same graph with the edges reversed.

3) Call DFS( $G^T$ ), consider the vertices in order of decreasing order of  $f[u]$

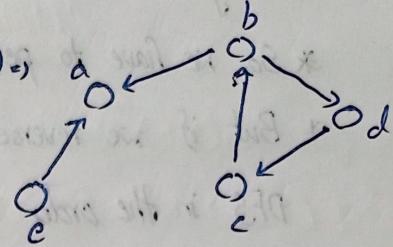
$\hookrightarrow$  Each tree in the DFS forest of  $\text{DFS}(G^T)$  is a separate SCC.

Dry run

$$G(V, E) \Rightarrow$$



$$G^T(V, E) \Rightarrow$$



Step 1: DFS tree of  $G(V, E) \Rightarrow a \rightarrow b \rightarrow c \rightarrow d$   
 $\downarrow$   
 $\rightarrow e$

$f[u]$  ordered in decreasing order:  $d \ a \ c \ b \ c \ d$

Step 2: Done above

Step 3:  $\text{DFS}(G^T)$  on  $a \Rightarrow$  ①

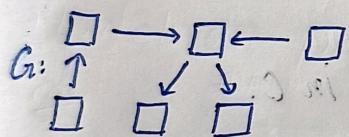
$\text{DFS}(G^T)$  on  $c \Rightarrow$  ②

$\text{DFS}(G^T)$  on  $b \Rightarrow$  ③ → ② → ①

} These are the 3  
SCCs in the graph  $G$

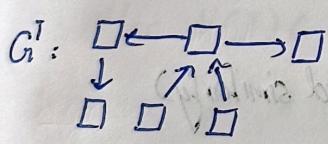
Logic behind the algorithm

Say we have the following graph (We compiled all the SCCs together)



Here,  $\boxed{\quad}$  is a SCC (It has nodes inside)

In this graph, if there is an edge b/w 2SCCs,  
one node in the first SCC has an edge to a node  
in the second SCC



If we reverse the edges, the ~~old~~ SCCs won't be  
affected. The edges would be reversed.

Aim: We need to develop an algorithm that outputs each SCC (each  $\boxed{\quad}$ ) separately.

↳ For achieving this we need will perform DFS.

↳ The challenge here is there might be a chance to get  $\boxed{\quad} \rightarrow \boxed{\quad}$  which is wrong.

Graph:



Solution to this challenge:

( $G^T$ ) Call DFS on SCC 2 before SCC 1.

This way we get  $\boxed{\quad} \notin \boxed{\quad}$  separately.

\* If you notice, topological sort of  $G'$  is  $\{1, 2\}$

\* So, we have to perform DFS on 2 before 1.

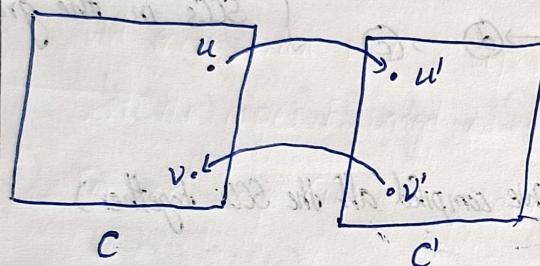
\* But if we reverse the edges in  $G'$  (Becomes  $G'^T$ ), we can perform DFS in the order of topological sort.

This is essentially what we are doing.

### Proof of correctness

Claim: B/w 2<sup>n</sup> SCCs  $C \& C'$ , if there is a path from  $C$  to  $C'$ , there can't exist a path from  $C'$  to  $C$ .

↳ Prove claim using proof by contradiction



$(u, u') \& (v, v')$  exists

Now each vertex in  $C$  is connected to each vertex in  $C'$  via  $(u, u')$

↳ Each ver

Each vertex in  $C'$  is connected to each vertex in  $C$ .

This means  $C \& C'$  aren't separate & disjoint.

↳ Contradiction

So, there are 2 cycles in the graph of SCCs (can be proved similarly)

Claim: If  $(u, v) \in E$  st.  $u \in C \& v \in C'$ , then  $f(C) > f(C')$

↳ Proof

$$f(C) = \max_{v \in C} f(v) : d(C) = \min_{v \in C} d(v)$$

Case I:  $d(C) < d(C')$

Say  $d(C) = d[x]$  ( $x$  is first node in  $C$  to be discovered)

Every vertex in  $C \& C'$  (except  $x$ ) is white @  $d[x]$

There is a white path from  $x$  to each vertex in  $C \& C'$

For  $C$  to turn black,  $C'$  must turn black as  $C \rightarrow C' \Rightarrow f(C) > f(C')$

Case II:  $d(C) > d(C')$

$$d(C') = d[y]$$

$\exists$  white path from  $y$  to every node in  $C'$

$$f(C') < d(C) < f(C)$$

Claim: Suppose  $(u, v) \in E^T$  where  $u \in C$ ,  $v \in C'$ , then  $f(C) < f(C')$

Now proof by induction of on DFS trees  $k$

↳ Base case:  $k=0$ , trivially true

↳ Inductive hypothesis:  $k$  trees returned corresponding to SCC.  
↳ Constructed in  $\downarrow$  order of  $f(u)$

↳ Inductive Step

Let  $u$  be root of  $(k+1)^{\text{th}}$  DFS tree.

Let  $S$  be the SCC which contains  $u$ .

The  $(k+1)^{\text{th}}$  tree must be  $S$

$$V(S) \subseteq V(T) \quad (V(T) \subseteq V(S))$$

As all vertices in  $S$  are connected to  $u$  via a white path. So  
 $V(S) \subseteq V(T)$

Now for  $V(T) \subseteq V(S)$

↳ Assume not,  $\exists$   $x \in T$  s.t.  $x \notin S$

Say  $x \in S'$ , there is an edge from  $S$  to  $S'$

But  $f(S') < f(S)$  as  $S \rightarrow S'$

This means  $S'$  would be one of the  $k$  trees before  $S$ . We won't visit it again

$$\text{Hence, } V(T) \subseteq V(S) \Rightarrow \underline{\underline{S = T}}$$

Hence Proved

# Minimum Spanning Tree

Given  $G(V, E)$  graph & the weights of the edges  $c: E \rightarrow \mathbb{N}$ , construct tree  $T (T \subseteq E)$  s.t.  $G(V, T)$  is connected &  $\sum_{e \in T} c(e)$  is minimum

This can be computed with a greedy strategy.

There are 2 strategies:

## 1) Kruskal's Algorithm

Insert edges in increasing order of cost so that no cycles are formed.

Kruskal ( $V, E$ )

$$A = \emptyset$$

Add each vertex  $v$  to a separate component of  $A$

Sort the edges  $E$  based on their weights

For each edge  $(u, v)$  in order:

if  $u \notin v$  are not in same component:

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

## Runtime Analysis

Sorting the edges  $\rightarrow O(m \log m)$

Checking whether 2 elements are of same component & merge  $\rightarrow O(n \log n)$

↳ Done  $m$  times  $\rightarrow O(m \log n)$

Adding vertices to  $A \rightarrow O(n)$

So,  $O(n + m \log n + n \log n)$

## 2) Prim's Algorithm

Start @ a root node  $s$  & try to greedily grow a tree from  $s$  outwards

At each step, add the node that can be attached as cheaply as possible to the partial tree we already have

Prims ( $V, E$ ):

For each  $u \in V$ :

$\text{key}[u] = \infty$  (Min cost to add  $u$  to the soln)

$\pi[u] = \text{NIL}$  (The neighbour of  $u$  that realised the key)

$\text{key}[r] = 0$

$Q = V$  (Ordered by key; Priority Queue with heap implementation)

While  $Q \neq \emptyset$ :

$u = \text{Extract Min}(Q)$

Add edge  $(u, \pi(u))$  to solution tree

For each edge  $v \in \text{AdjList}[u]$ :

if  $v \in Q$  &  $w(u, v) < \text{key}[v]$ :

$\pi[v] = u$

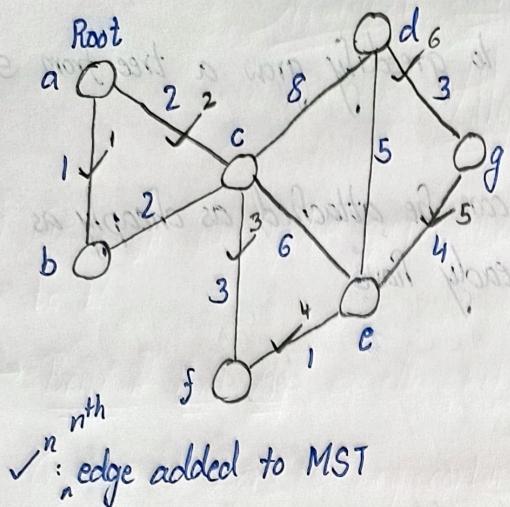
$\text{key}[v] = w(u, v)$

### Runtime Analysis

Finding Min Element from  $Q \rightarrow n$  times  $\xrightarrow{\text{Binary heap}} O(n \log n)$   $\xrightarrow{\text{Fibonacci heap}} O(n \log n)$

Updating the value of key  $\rightarrow m$  times  $\xrightarrow{\text{Binary heap}} O(m \log n)$   $\xrightarrow{\text{Fibonacci heap}} O(m)$

# \* Dry Run of Prim's Algorithm:



Vertex	Key	$\pi$
a	0 ✓	NIL
b	$\infty \rightarrow 1$ ✓	a
c	$\infty \rightarrow 2$ ✓	a
d	$\infty \rightarrow 8 \rightarrow 5 \rightarrow 3$	g
e	$\infty \rightarrow 6 \rightarrow 1$ ✓	f
f	$\infty \rightarrow 3$ ✓	c
g	$\infty \rightarrow 4$ ✓	e

## → Proof of Correctness of Krushkal's & Prim's Algorithm

\* Property: Some edges are always safe to be added to a MST

### \* Cut Property

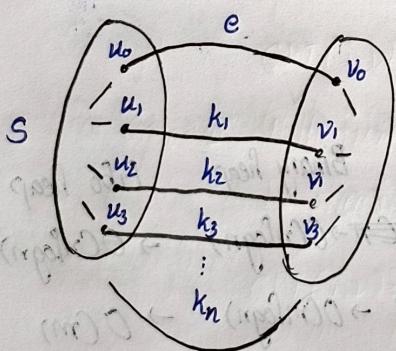
Assume all edge costs are distinct,  $S \subset V$  &  $S \neq \emptyset$

We have 2 disjoint set of vertices:  $S$  &  $V \setminus S$

Let  $e$  be the edge of min. weight, connecting  $S$  &  $V \setminus S$

Claim:  $e$  is a safe edge, aka, every MST contains  $e$

### ↳ Proof by contradiction



Say there exists an MST with the connecting edges b/w  $S$  &  $V \setminus S$  be  $\{k_1, k_2, \dots, k_n\}$  s.t no  $k_i = e$

Let us add  $e$  & remove any  $k_i$  (w.l.o.g  $k_1$ )

↳ This set of edges would cost less than the original MST

↳ We have to show this set of edges is a MST

1) There must be no loops ✓

↳ Original MST had no loops, adding  $e$  formed a loop, removing  $k_1$  broke that loop

2) Every 2 vertices are connected ✓

↳ Say 2 vertices were connected via  $k_1$  in original MST

These 2 vertices now would be connected via  $c$  or  $k_2 k_3 \dots k_n$   
Bridge

↳ The other vertices pairs remain undisturbed.

So, the given set of edges is a ST with less weight than original MST

↳ Contradiction

Hence Proved.

### Optimality of Prims

if we take  $S$  as a part then boundary edges will be  
It would be enough to show that every edge added is a safe edge.

Apply Cut Property :  $S \Rightarrow$  Partial Solution

Every edge added is edge b/w  $S$  &  $V \setminus S$  which has least weight

Hence Proved

### Optimality of Krushals

Similar logic

Every edge added is min weight edge connecting 2 disjoint components  
(they don't form a cycle)

Hence Proved

# Shortest Path

## Algorithms

If  $G$  is an unweighted graph, how do we find a the shortest path from  $s$  to  $t$ ?

→ Bellman-Ford Algo. Algorithm and time is a ratio to be very slow

Condition to work:  $G$  may contain -ve edges, BUT NO -ve cycles.

- Now that  $G$  has no -ve cycles, the algorithm won't loop in the -ve cycle forever
- So, the ~~shortest~~ shortest path from  $s$  to  $t$  would simple & have almost  $n-1$  edges.

Let us solve this using DP.

Step 1: Define the subproblem & final answer

$\text{OPT}(i, v) \Rightarrow \min \text{ cost to reach } v \text{ from } t \text{ with atmost } i \text{ edges.}$

$\text{OPT}(n-1, s) \Rightarrow \text{final answer}$

Step 2: Write the recurrence

Case I: Path from  $t$  to  $v$  has atmost  $i-1$  edges

$$\text{OPT}(i, v) = \text{OPT}(i-1, v)$$

Case II: Path from  $t$  to  $v$  has  $i$  edges exactly.

We check paths from neighbours of  $v$

$$\text{OPT}(i, v) = \min_{w \in \text{AdjList}(v)} [\text{weight}(v, w) + \text{OPT}(i-1, w)]$$

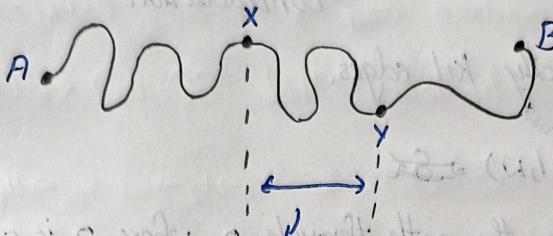
$$\text{So, } \text{OPT}(i, v) = \min \left\{ \text{OPT}(i-1, v), \min_{w \in N(v)} [\text{weight}(v, w) + \text{OPT}(i-1, w)] \right\}$$

### Step 3: Proof of correctness

Let us apply induction upon  $i$  & use optimal substructure property.

What is optimal substructure property?

Say the shortest path b/w nodes A & B is as follows:



This path is the shortest path b/w X & Y

Optimal substructure guarantees that the sub-path b/w X & Y in the shortest path from A to B is the shortest path b/w X & Y.

This is trivial & can be proved by contradiction. If there exists a shorter path b/w X & Y, shortest path b/w A & B would take it.

Proof of Bellman-Ford Algo using induction on  $i$ :

Base Case:  $i=0$

$$OPT(0, t) = 0, OPT(0, w) = \infty \text{ for any } w \neq t$$

True

Inductive hypothesis:  $i = k$

We have  $OPT(k, w)$  for all  $w \in V$

This means we have shortest path b/w  $t$  &  $w$  of at most  $k$  edges.

Inductive step:  $i = k+1$

We compute  $OPT(k+1, w)$  using recurrence.

Proof by contradiction. Say  $OPT(k+1, w)$  is not optimal cast shortest path of atmost  $k+1$  edges b/w  $w$  &  $t$ . Say  $S$  is the shortest cast path as such.

This means  $|S| < OPT(k+1, w)$

Case I:  $S$  has atmost  $k$  edges

Then  $\text{OPT}(\text{ctrl}, w)$  would take the value of  $\text{OPT}(k, w)$

Then  $\text{OPT}(\text{ctrl}, w) = S$

Contradiction.

Case II  $S < \text{OPT}(k+1, w) \Rightarrow S < \text{OPT}(k, w)$

Contradiction

Case II:  $S$  has exactly  $k+1$  edges.

$S < \text{OPT}(k+1, w)$

Say  $S$  takes the path through  $p$ , where  $p$  is a neighbour of  $w$ .

Then,  $S < \text{OPT}(k+1, w)$

$$S_{t-p} + c(p, w) < \text{OPT}(k, p) + c(p, w)$$

Cost of path  
from  $t$  to  $p$

$$S_{t-p} < \text{OPT}(k, p)$$

Contradiction

Thus  $\text{OPT}(k+1, w)$  is correct

Hence Proved

### Runtime Analysis

$O(n, m)$

For a fixed  $(i, v)$ , the value of  $\text{OPT}(i, v)$  is used for the computation of  $\text{OPT}(\text{ctrl}, w)$  for degree( $v$ ) number of times.

$\sum \text{degree}(v) = m$

We are computing it,  $n$ , times (for diff  $i$ )  $\Rightarrow \underline{\underline{O(n, m)}}$

## Dijkstra's Algorithm

Given a graph  $G$  & ~~some~~ source  $s$ , find shortest path from  $s$  to all vertices. (All edge weights must be positive)

Let us solve this by a greedy strategy:

Maintain a set  $S$  of vertices  $u$  for which we have determined a shortest path distance  $d(u)$  from  $s$ , (or from explored)

Dijkstra( $G$ ):

$$S = \{s\}$$

$$d(s) = 0$$

While  $S \neq V$ :

Select a node  $v \notin S$  s.t.  $d'(v) = \min_{e=(u,v), u \in S} d(u) + l_e$  is min

$$S = S \cup \{v\}$$

$$d(v) = d'(v)$$

Basically,

Greedy choice: Always pick the next vertex with the smallest current distance.

Optimal Substructure: Shortest paths can be built by extending shortest paths.

Intuition behind the Alg

The alg picks the vertex  $u$  with smallest tentative distance  $d(u)$  & fixes it as shortest path from  $u$  to  $s$ .

We have to prove that this path is indeed the shortest path to  $u$ .

↳ Proof by induction on size of set  $S$ .

Base case:  $|S| = 1$

$$S = \{s\}, d(s) = 0$$

Trivially true.

Inductive hypothesis:  $|S| = k$

$$S = \{s, u_1, u_2, \dots, u_{k-1}\}$$

$$d(u_i) = \delta(s, u_i)$$

Here  $\delta(s, u_i)$  is the actual shortest path from  $s$  to  $u_i$ .

Inductive step:  $|S| = k+1$

We are adding  ~~$u_{k+1}$~~   $u_k$  to  $S$  &  $d(u_k) = \delta(s, u_k)$

Assume (contradiction) that  $\delta(s, u_k) < d(u_k)$

Say path  $\delta(s, u_k)$  is  $s \rightarrow v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_m \rightarrow u_k$

In this path say  $v_j$  is the first vertex yet to be extracted from the queue. In the sense,  $v_{j-1}$  has been extracted.

$$\text{So, } \delta(s, v_{j-1}) = d(v_{j-1})$$

$$\text{As } w(v_{j-1}, v_j) \geq 0, d(v_j) \leq d(v_{j-1}) + w(v_{j-1}, v_j) = \delta(s, v_j)$$

By optimal substructure property.

$$\text{This means } d(v_j) \leq \delta(s, v_j)$$

$$d(v_j) \leq \delta(s, u_k) < d(u_k)$$

$$d(v_j) < d(u_k)$$

This is a contradiction as  $u_k$  is extracted from queue before  $v_j$  ( $v_j$  is still in queue) but  $d(v_j) < d(u_k)$

Thus, assumption is wrong.

$$d(u_k) = \delta(s, u_k)$$

Hence Proved

Runtime Analysis

\*  $n$  Extraction operations  $\rightarrow n \log n$

$\rightarrow n \log n$

\*  $m$  Updates

$\rightarrow m \log n$

$\rightarrow m \log n$

Binary heaps

Fibonacci heaps

Binary  $\Rightarrow O((m+n)\log n)$

Fibonacci  $\Rightarrow O(n \log n + m) //$