

## Algorithms on Graphs: Part I

### Graphs

**Graphs.** A *graph* is a pair  $G = \langle V, E \rangle$ , where

- $V = \{v_1, \dots, v_n\}$  is a set of *vertices*, and
- $E = \{(v_i, v_j)\}$  is the set of *edges*.

**Directed and Undirected Graphs.** In *directed graphs* edges have *start* and *end*: if  $(v, v') \in E$  is an edge in a directed graph  $G = \langle V, E \rangle$ , then  $v$  is its start and  $v'$  is its end, and the *direction* of  $(v, v')$  is from  $v$  to  $v'$ .

In *undirected graphs*, edges do not have directions:  $(v, v') = (v', v)$  for any edge  $(v, v') \in E$ .

**Weighted Graphs.** A *vertex weighted graph* is a graph  $G = \langle V, E, w \rangle$  where  $w : V \rightarrow \mathcal{R}$ . Here,  $w$  is the *vertex weight function*.

An *edge weighted graph* is a graph  $G = \langle V, E, w \rangle$ , where  $w : E \rightarrow \mathcal{R}$ . Here,  $w$  is the *edge weight function*.

### Graph Representation

Graphs have two typical representations as data structures: *adjacency matrices* and *adjacency lists*.

**Adjacency matrix representation.** A graph  $G = \langle V, E \rangle$  where  $V = \{v_1, \dots, v_n\}$  is represented as a two-dimensional array  $G[1..n, 1..n]$ .  $G[i, j] = 1$  if  $(v_i, v_j) \in G$ .  $G[i, j] = 0$  otherwise.

A *vertex weighted graph*  $G = \langle V, E, w \rangle$  is represented as a two-dimensional array  $G[1..N][0..N]$ , where  $G[i][0] = w(v_i)$  and  $G[i][j] = 1$  if  $(v_i, v_j) \in G$  and  $G[i][j] = 0$  otherwise for  $j > 0$ .

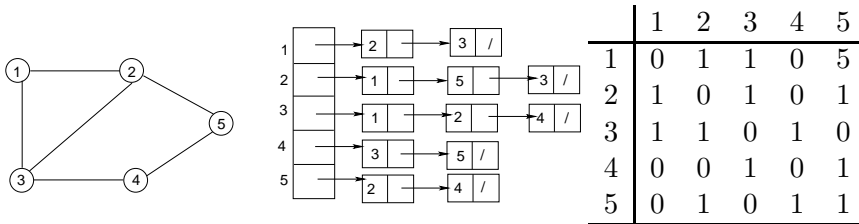


Figure 1: An undirected graph and its adjacency list and adjacency matrix representations.

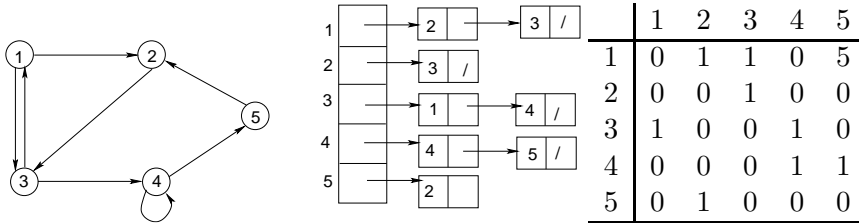


Figure 2: An undirected graph and its adjacency list and adjacency matrix representations.

An *edge weighted graph*  $G = \langle V, E, w \rangle$  is represented as a two-dimensional array  $G[1..N][1..N]$ , where  $g[i][j] = w(v_i, v_j)$ .

Adjacency matrices for *undirected graphs* are symmetric.

**Adjacency list representation.** A graph  $G = \langle V, E \rangle$  is represented as an array  $Adj[1..n]$  of *lists*.  $Adj[v_i]$  contains all  $v_j$ , such that  $(v_i, v_j) \in E$ . If  $G$  is *vertex weighted*, an additional array  $w[1..n]$  is used to store vertex weights. If  $G$  is *edge weighted*, then  $Adj[v_i]$  stores pairs  $(v_j, w(v_i, v_j))$ .

Figures 1 and 2 show the adjacency list and adjacency matrix representations of undirected and directed graphs (respectively).

## Algorithms: Graph Traversal

**Graph Traversal Problem.** Given a graph  $G = \langle V, E \rangle$  and a node  $s \in V$  (referred to as the *source node*, visit all nodes of the graph starting with the *source node*.

### Breadth-First Search

**Breadth-First Search** is a traversal technique which visits all yet unvisited neighbors of a node  $v$  right after visiting  $v$ .

**Node coloring.** In the breadth-first search algorithm we use *colors* of nodes as node labels to represent current state of a node (visited, enqueued, unvisited). This is done for aesthetic reasons, numeric labels can be used instead.

**Data Structures.** Breadth-first search algorithm (BFS algorithm) maintains a *queue* of vertices.

```

ALGORITHM BFS(V,Adj,s)
begin
  foreach  $v \in V - \{s\}$  do //initialization
     $v.color \leftarrow WHITE$ ; //color
     $v.d \leftarrow \infty$ ; //distance from the source
     $v.\pi \leftarrow NIL$ ; // "parent" (in the traversal)
  endfor  $s.color \leftarrow GRAY$ ;
   $s.d \leftarrow 0$ ;
   $s.\pi \leftarrow NIL$ ;
   $Q \leftarrow \emptyset$ ; //initialization of the queue
  Enqueue( $Q$ );
  //Main loop while  $Q \neq \emptyset$  do
   $u \leftarrow Dequeue(Q)$ ;
  foreach  $v \in Adj[u]$  do
    if  $v.color = WHITE$  then  $v.color \leftarrow GRAY$ ;
       $v.d \leftarrow u.d + 1$ ;
       $v.\pi \leftarrow u$ ;
      Enqueue( $Q, v$ );
    endif
  endfor
   $u.color \leftarrow BLACK$ ;
endwhile
end

```

### BFS: Analysis

**Runnin time.**  $T(BFS) = O(|V| + |E|)$ .

*Notes:* Initialization costs  $O(V)$ . For each node  $v$  visited, its adjacency list  $Adj[v]$  will be scanned once, leading to the overall  $O(\sum_{v \in V} |Adj[v]|) = O(|E|)$ .

**Path.** A *path* in a graph  $G = \langle V, E \rangle$  is a sequence  $e_1, e_2, \dots, e_k$  of edges  $e_i = (v_i, u_i) \in E$ , such that  $u_1 = v_2, u_2 = v_3, \dots, u_{k-1} = v_k$ .

**Shortest Path.** A **shortest path distance** between two nodes  $s$  and  $v$  in a graph  $G$ , denoted  $\delta(s, v)$  is the minimum number  $k$  of edges on a path from  $s$  to  $v$ .

A shortest path between  $s$  and  $v$  is any path whose length is equal to the *shortest path distance* between  $s$  and  $v$ .

$v$  is reachable from  $s$  if there exists at least one path in  $G$  from  $s$  to  $v$ .

**Lemma 1.** Let  $G = (V, E)$  be a graph,  $s \in V$  and  $(u, v) \in E$ . Then

$$\delta(s, v) \leq \delta(s, u) + 1.$$

**Proof.** If  $u$  is reachable from  $s$ , then, of course  $v$  is reachable as well.

*Case 1.*  $u$  is on the shortest path from  $s$  to  $v$ . Then  $\delta(s, v) = \delta(s, u) + 1$ .

*Case 2.*  $u$  is NOT on the shortest path from  $s$  to  $v$ . Then  $\delta(s, v) < \delta(s, u) + 1$ .

**Lemma 2.** For each node  $n \in V$ , in algorithm BSF  $v.d \geq \delta(s, d)$ .

**Proof.** By induction.

**Lemma 3.** Consider some state of the queue  $Q = (u_1, \dots, u_r)$  in the BFS algorithm. Then  $u_1.d \leq u_2.d \leq \dots \leq u_r.d$ .

**Proof.** Induction on the number of Enqueue operations.

**Theorem 1.** Let  $G = \langle V, E \rangle$  be a graph and  $s \in V$ . Then:

1. Algorithm BFS discovers all nodes in  $V$  *reachable* from  $s$ .
2. At the end of the algorithm, for each node  $v \in V$ ,  $v.d = \delta(s, d)$ .
3. For each node  $v \neq s$ , one of the *shortest paths* from  $s$  to  $v$  goes through the node  $v.\pi$  (and edge  $(v.\pi, v)$ ).

**Proof.** By contradiction.

## Depth-First Search

**Depth-First Search** traversal is a graph traversal technique that visits the neighbors of most recently visited node on each step.

**DFS node colors.** The Depth-First search (DFS) algorithm colors nodes as follows. Unvisited nodes are **white**; discovered nodes are **gray** and visited nodes are **black**.

**DFS timestamps.** Each node  $v \in V$  receives two *timestamps* during the DFS algorithm. The first timestamp,  $v.d$ , records the step on which  $v$  was discovered (became **gray**). The second timestamp,  $v.f$  records the step on which  $v$  was visited (became **black**).

```
ALGORITHM DFS(V,Adj)
begin
  foreach  $v \in V$  do //initialization
     $v.color \leftarrow$  WHITE; //vertex color: unvisited
     $v.\pi \leftarrow$  NIL; // "parent" (in the traversal)
  endfor  $time \leftarrow 0$  //behaves as global variable
//Main loop
  foreach  $v \in V$  do
    if  $v.color =$  WHITE then DFS_VISIT(V,Adj,v);
  endfor
end
```

```

ALGORITHM DFS_VISIT(V,Adj,u)
begin
  time ← time + 1;
  u.d ← time;
  u.color ← GRAY;    //mark vertex as discovered
  foreach v ∈ Adj[u] do    //visit neighbors
    if v.color =WHITE then
      v.π ←u;
      DFS_VISIT(V,Adj,v);
    endif
  endfor   u.color ← BLACK;    //mark vertex as visited
  time ← time + 1;
  u.f ← time;
end

```

### DFS: Analysis

**Running time.**  $T(DFS) = \Theta(|V| + |E|)$ .

*Note.* Initialization takes  $\Theta(|V|)$  steps. On each call of DFS\_VISIT, with node  $v$  as input, at most  $|Adj[v]|$  of recursive calls will be made. So, the total number of recursive calls of DFS\_VISIT is  $\theta(|E|)$

**Predecessor subgraph.** Given a graph  $G$ , its predecessor subgraph  $G_\pi = \langle V, E_\pi \rangle$  contains only the edges  $(v.\pi, v)$  for each  $v \in V$ .

**Predecessor subgraph in DFS.** A forest of trees.

**Parenthesis theorem.** In any DFS order of the traversal of a graph  $G = \langle V, E \rangle$ , for any two nodes  $u, v \in V$ , one of the following three conditions holds:

1.  $[u.d, u.f] \subset [v.d, v.f]$ ;  $u$  is a descendant of  $v$ .
2.  $[v.d, v.f] \subset [u.d, u.f]$ ;  $v$  is a descendant of  $u$ .
3.  $[u.d, u.f] \cap [v.d, v.f] = \emptyset$ ;  $u, v$  are not in ancestor-descendant relationship.

**Proof.** Consider two cases:  $u.d < v.d$  and  $u.d > v.d$ . For each subcase, establish two possible outcomes.

**White-path Theorem.** In a depth-first forest of a graph  $G = \langle V, E \rangle$ , vertex  $v$  is a descendant of vertex  $u$  **iff** at the time  $u.d$ , there is a path from  $u$  to  $v$  which only encounters white nodes.

**Proof.** Prove in both directions. For the  $\Rightarrow$  direction, the white path is constructed. For the  $\Leftarrow$  direction, prove by contradiction.

**Classification of edges.** DFS algorithm splits edges in  $G$  into the following categories:

- **Tree edges.** Edges in the  $G_\pi$  depth-first forest of  $G$ .
- **Back edges.** Edges  $(u, v)$ , where  $u$  is a descendant of  $v$ .
- **Forward edges.** Edges  $(u, v)$  not in  $G_\pi$ , where  $v$  is a descendant of  $u$ .
- **Cross edges.** All other edges.

## Algorithms: Topological Sort on DAGs

**Directed Acyclic Graphs (DAGs).** A directed graph  $G = \langle V, E \rangle$  is *acyclic* if for any node  $v \in V$ , there is **no path** from  $v$  back to  $v$ .

Note, DAGs do not have *back edges*.

**Topological Sort Problem.** Given a directed acyclic graph  $G = \langle V, E \rangle$  a topological sort of  $G$  is a **linear order**  $<$  on the vertices from  $V$ , such that:

if there is an edge  $(u, v) \in E$ , then  $u < v$ .

The problem is to find a(ny) topological sort given a DAG  $G = \langle V, E \rangle$

**Algorithm.** The algorithm for topological sort uses DFS:

```
run DFS(G), compute all  $v.f$ 
sort  $V$  in ascending order by  $v.f$ 
return sorted list of nodes
```

## Minimal Spanning Trees

**Spanning Tree.** Let  $G = \langle V, E, w \rangle$  be a (connected) edge-weighted undirected graph. A *spanning tree* of  $G$  is a subset  $T \subseteq E$  of edges, such that  $G_T = \langle V, T \rangle$  is *connected* and *acyclic*.

The weight of a spanning tree:  $w(T) = \sum_{(u,v) \in T} w(u, v)$ .

**Minimal Spanning Tree Problem.** Given an undirected edge-weighted graph, find a spanning tree with minimum weight.

**Greedy approach.** (if we can make it work).

```
ALGORITHM GENERIC_MST( $V, \text{Adj}, w$ )
```

```
begin
```

```
   $A \leftarrow \emptyset;$ 
```

```
  while  $A$  is not a spanning tree do
```

```
    find a safe edge  $(u, v) \in E$  to include in  $A$ 
```

```
     $A = A \cup \{(u, v)\}$ 
```

```
  endwhile
```

```
  return  $A;$ 
```

```
end
```

*How to find a safe edge?*

**Cuts.** A cut  $S, V - S$  in an undirected graph  $G = \langle V, E \rangle$  is a partition of  $V$  into two sets.

An edge  $(u, v)$  **crosses** the cut if  $u \in S$  and  $v \in V - S$  or vice versa.

A cut **respects** a set  $A$  of edges if no edge in  $A$  crosses the cut.

An edge  $(u, v)$  is a **light edge** crossing the cut if its weight is the minimum among all edges crossing the cut.

**Theorem 3.** Let  $G = \langle E, V, w \rangle$  be a connected undirected edge-weighted graph. Let  $A \subset E$  be in some *minimal spanning tree*. and let  $(S, V - S)$  be some cut of  $G$ .

If  $(S, V - S)$  respects  $A$  and  $(u, v)$  is a light edge crossing  $A$ , then  $(u, v)$  is safe for  $A$ .