Algorithms 09 CS201

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Representation of Graphs

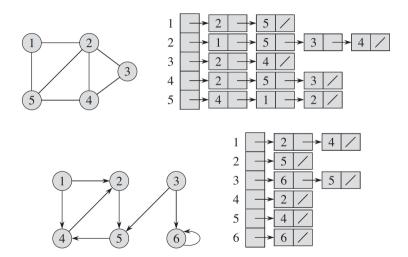
Two Ways of Representing a Graph G = (V, E)

- Adjacency-list: Provides a compact way to represent sparse graphs—those for which E is much less than $|V|^2$.
- ▶ **Adjacency-matrix:** Preferred to represent dense graphs—those for which E is close to $|V|^2$ when we need to be able to tell quickly if there is an edge connecting two given vertices.

Adjacency-List

- ► The adjacency-list representation of a graph G = (V, E) consists of an array Adj of |V| lists, one for each vertex in V.
- For each $u \in V$, the adjacency list Adj[u] contains all the vertices v such that there is an edge $(u,v) \in E$.
- ▶ Adj[u] consists of all the vertices adjacent to $u \in G$. (Alternatively, it may contain pointers to these vertices.)
- Since the adjacency lists represent the edges of a graph, in pseudocode we treat the array Adj as an attribute of the graph, just as we treat the edge set E.
- ▶ If G is a directed graph, the sum of the lengths of all the adjacency lists is |E|. If G is an undirected graph, the sum of the lengths of all the adjacency lists is 2|E|.
- For both directed and undirected graphs, the adjacency-list representation has the memory requirement of $\theta(V+E)$.

Adjacency-List



Adjacency-List

- For both directed and undirected graphs, the adjacency-list representation has the memory requirement of $\theta(V+E)$.
- We can readily adapt adjacency lists to represent weighted graphs, that is, graphs for which each edge has an associated weight, typically given by a weight function $w: E \to \mathbb{R}$.
- Provides no quicker way to determine whether a given edge (u, v) present in the graph than to search for v in the adjacency list Adj[u].

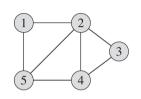
Adjacency-Matrix

We assume that the vertices are numbered 1, 2, ..., |V| in some arbitrary manner. Then the adjacency-matrix representation of a graph G consists of a $|V| \times |V|$ matrix $A = (a_{ij})$ such that

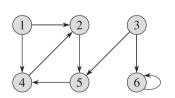
$$a_{ij} = \begin{cases} 1 \text{ if } (i,j) \in E, \\ 0 \text{ otherwise.} \end{cases}$$

- ▶ The adjacency matrix of a graph requires $|V|^2$ memory, independent of the number of edges in the graph.
- Since in an undirected graph, (u, v) and (v, u) represent the same edge, the adjacency matrix A of an undirected graph is its own transpose: $A = A^T$.
- ▶ We may store above the diagonal of the adjacency matrix, thereby cutting the memory needed to store the graph almost in half.

Adjacency-Matrix



	1	2	3	4	5
1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	0 1 1 0 1	0

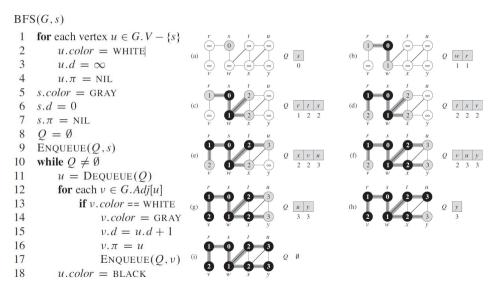


	1	2	3 0 0 0 0 0 0	4	5	6
1	0	1	0	1	0	0
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

Adjacency-Matrix

- ► An adjacency matrix can represent a weighted graph.
- ▶ We can store the weight w(u, v) of the edge $(u, v) \in E$ as the entry in row u and column v of the adjacency matrix. If an edge does not exist, we can store a NIL, 0, or ∞ .

- ▶ Breadth-first search is one of the simplest algorithms for searching a graph and the archetype for many important graph algorithms.
- ▶ It produces a "breadth-first tree" with root s that contains all reachable vertices. For any vertex v reachable from source vertex s, the simple path in the breadth-first tree from s to v corresponds to a "shortest path" from s to v in G, that is, a path containing the smallest number of edges.
- ▶ Breadth-first search is so named because it expands the frontier between discovered and undiscovered vertices uniformly across the breadth of the frontier. The algorithm discovers all vertices at distance k from s before discovering any vertices at distance k+1.



Printing Path

```
PRINT-PATH(G, s, \nu)

1 if \nu == s

2 print s

3 elseif \nu.\pi == \text{NIL}

4 print "no path from" s "to" \nu "exists"

5 else PRINT-PATH(G, s, \nu.\pi)

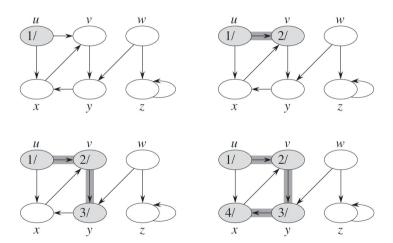
6 print \nu
```

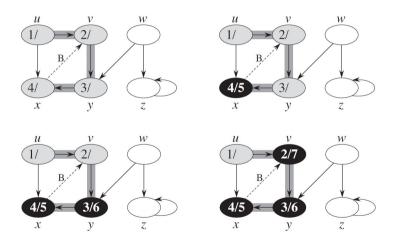
Analysis

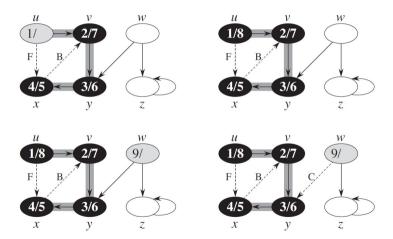
- ▶ The overhead for initialization is O(|V|).
- ▶ After initialization, breadth-first search never whitens a vertex, and thus the test in line 13 ensures that each vertex is enqueued at most once, and hence dequeued at most once.
- The operations of enqueuing and dequeuing take O(1) time, and so the total time devoted to queue operations is O(|V|).
- ▶ Because the procedure scans the adjacency list of each vertex only when the vertex is dequeued, it scans each adjacency list at most once.
- Since the sum of the lengths of all the adjacency lists is $\Theta(|E|)$, the total time spent in scanning adjacency lists is O(|E|).
- ▶ Thus the total running time of the BFS procedure is O(|V|+|E|).

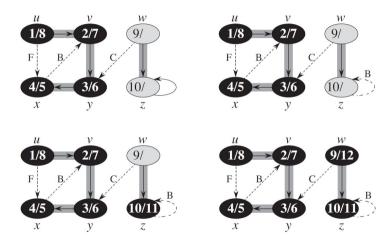
- ► The strategy followed by depth-first search (DFS) searches "deeper" in the graph whenever possible.
- ▶ DFS explores edges out of the most recently discovered vertex *v* that still has unexplored edges leaving it.
- \triangleright Once all of v's edges have been explored, the search "backtracks" to explore edges leaving the vertex from which v was discovered.
- ► This process continues until we have discovered all the vertices that are reachable from the original source vertex.
- ▶ If any undiscovered vertices remain, then depth-first search selects one of them as a new source, and it repeats the search from that source. The algorithm repeats this entire process until it has discovered every vertex.

```
DFS-VISIT(G, u)
                                    time = time + 1
                                  2 u.d = time
DFS(G)
                                  3 u.color = GRAY
  for each vertex u \in G.V
                                  4 for each v \in G.Adj[u]
      u.color = WHITE
                                         if v.color == WHITE
      u.\pi = NIL
                                             \nu.\pi = u
  time = 0
                                             DFS-VISIT(G, \nu)
  for each vertex u \in G.V
                                  8 u.color = BLACK
       if u.color == WHITE
                                  9 time = time + 1
           DFS-VISIT(G, u)
                                 10 u.f = time
```









Analysis

- ► The loops on lines 1–3 and lines 5–7 of DFS take time $\Theta(|V|)$, exclusive of the time to execute the calls to DFS-VISIT.
- The procedure DFS-VISIT is called exactly once for each vertex $v \in V$, since the vertex u on which DFS-VISIT is invoked must be white and the first thing DFS-VISIT does is paint vertex u gray.
- ▶ During an execution of DFS-VISIT (G, v), the loop on lines 4–7 executes |Adj[v]| times.
- Since

$$\sum_{v \in V} |Adj[v]| = \Theta(|E|)$$

the total cost of executing lines 4–7 of DFS-VISIT is $\Theta(|E|)$.

▶ The running time of DFS is therefore $\Theta(|V|+|E|)$.



BFS: Lemmas

Lemma 1

Let G = (V, E) be a directed or undirected graph, and let $s \in V$ be an arbitrary vertex. Then, for any edge $(u, v) \in E$,

$$\delta(s, v) \le \delta(s, u) + 1$$

where $\delta(s, v)$ is the shortest-path distance from s to v defined as the minimum number of edges in any path from vertex s to vertex v; if there is no path from s to v, then $\delta(s, v) = \infty$.

BFS: Lemmas

Lemma 2

Let G=(V,E) be a directed or undirected graph, and suppose that BFS is run on G from a given source vertex $s \in V$. Then upon termination, for each vertex $v \in V$, the value v.d computed by BFS satisfies $v.d \geq \delta(s,v)$.

Lemma 3

Suppose that during the execution of BFS on a graph G = (V, E), the queue Q contains the vertices $\langle v_1, v_2, \dots, v_r \rangle$, where v_1 is the head of Q and v_r is the tail. Then, $v_r \cdot d \leq v_1 \cdot d + 1$ and $v_i \cdot d \leq v_{i+1} \cdot d$ for $i = 1, 2, \dots, r-1$.

Lemma 4

Suppose that vertices v_i and v_j are enqueued during the execution of BFS, and that v_i is enqueued before v_j . Then $v_i d \le v_j d$ at the time that v_j is enqueued.

Breadth-first Trees

For a graph G = (V, E) with source s, we define the predecessor subgraph of G as $G_{\pi} = (V_{\pi}, E_{\pi})$, where

$$V_{\pi} = \{ v \in V : v.\pi \neq \text{NIL} \} \cup \{ s \}$$

and

$$E_{\pi} = \{ (v.\pi, v) : v \in V_{\pi} - \{s\} \}$$

The predecessor subgraph G_{π} is a breadth-first tree if V_{π} consists of the vertices reachable from s and, for all $v \in V_{\pi}$, the subgraph G_{π} contains a unique simple path from s to v that is also a shortest path from s to v in G. A breadth-first tree is in fact a tree, since it is connected and $|E_{\pi}| = |V_{\pi}| - 1$. We call the edges in E_{π} tree edges.

Predecessor Subgraph of a Depth-first Search

For a graph G = (V, E), predecessor subgraph of a depth-first search can be defined as $G_{\pi} = (V, E_{\pi})$ where

$$E_{\pi} = \{(v.\pi, v) : v \in V \text{ and } v.\pi \neq \text{NIL}\}\}$$

The predecessor subgraph of a depth-first search forms a depth-first forest comprising several depth-first trees. The edges in E_{π} are tree edges.

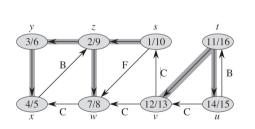
DFS: Theorems

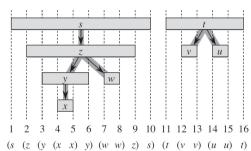
Parenthesis Theorem

In any depth-first search of a (directed or undirected) graph G = (V, E), for any two vertices u and v, exactly one of the following three conditions holds:

- \blacktriangleright the intervals [u.d, u.f] and [v.d, v.f] are entirely disjoint, and neither u nor v is a descendant of the other in the depth-first forest,
- ▶ the interval [u.d, u.f] is contained entirely within the interval [v.d, v.f], and u is a descendant of v in a depth-first tree, or
- ▶ the interval [v.d, v.f] is contained entirely within the interval [u.d, u.f], and v is a descendant of u in a depth-first tree.

DFS: Theorems





DFS: Theorems and Corolaries

Corollary: Nesting of Descendants' Intervals

Vertex v is a proper descendant of vertex u in the depth-first forest for a (directed or undirected) graph G if and only if u.d < v.d < v.f < u.f.

Theorem: White-path Theorem

In a depth-first forest of a (directed or undirected) graph G = (V, E), vertex v is a descendant of vertex u if and only if at the time u.d that the search discovers u, there is a path from u to v consisting entirely of white vertices.

Classification of Edges

Tree edges

Tree edges are edges in the depth-first forest G_{π} . Edge (u, v) is a tree edge if v was first discovered by exploring edge (u, v).

Back edges

Back edges are those edges (u, v) connecting a vertex u to an ancestor v in a depth-first tree. We consider self-loops, which may occur in directed graphs, to be back edges.

Forward edges

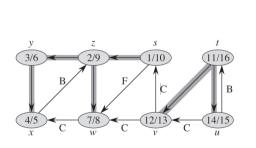
Forward edges are those nontree edges (u, v) connecting a vertex u to a descendant v in a depth-first tree.

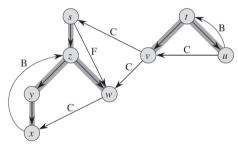
Cross edges

Cross edges are all other edges. They can go between vertices in the same depth-first tree, as long as one vertex is not an ancestor of the other, or they can go between vertices in different depth-first trees.



Classification of Edges

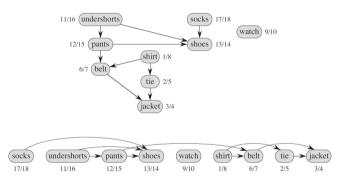




Topological Sort

- A topological sort of a dag G = (V, E) is a linear ordering of all its vertices such that if G contains an edge (u, v), then u appears before v in the ordering.
- ► We can view a topological sort of a graph as an ordering of its vertices along a horizontal line so that all directed edges go from left to right.
- ▶ If the graph contains a cycle, then topological sorting is not possible.
- ▶ Many applications use directed acyclic graphs to indicate precedences among events.

Topological Sort



Topological Sort

Algorithm

TOPOLOGICAL-SORT(G)

- 1 call DFS(G) to compute finishing times ν . f for each vertex ν
- 2 as each vertex is finished, insert it onto the front of a linked list
- 3 **return** the linked list of vertices

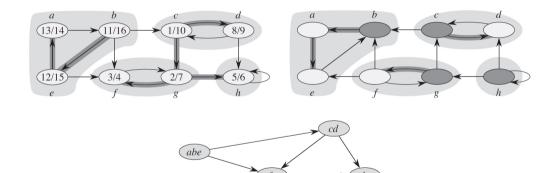
Analysis

- We can perform a topological sort in time $\Theta(|V|+|E|)$, since depth-first search takes $\Theta(|V|+|E|)$ time and it takes O(1) time to insert each of the |V| vertices onto the front of the linked list.
- ▶ A directed graph *G* is acyclic if and only if a depth-first search of *G* yields no back edges.

Strongly Connected Component

- ▶ A strongly connected component of a directed graph G = (V, E) is a maximal set of vertices $C \subseteq V$ such that for every pair of vertices u and v in C, we have both $u \leadsto v$ and $v \leadsto u$, that is, vertices u and v are reachable from each other.
- ▶ The transpose of a graph G = (V, E) is $G^T = (V, E^T)$, where $E^T = \{(u, v) : (v, u) \in E\}$.
- ▶ G and G^T have exactly the same strongly connected components: u and v are reachable from each other in G if and only if they are reachable from each other in G^T .

Strongly Connected Component



Strongly Connected Component

Algorithm

STRONGLY-CONNECTED-COMPONENTS (G)

- 1 call DFS(G) to compute finishing times u.f for each vertex u
- 2 compute G^{T}
- call DFS(G^{T}), but in the main loop of DFS, consider the vertices in order of decreasing u.f (as computed in line 1)
- 4 output the vertices of each tree in the depth-first forest formed in line 3 as a separate strongly connected component

White Board

White Board