

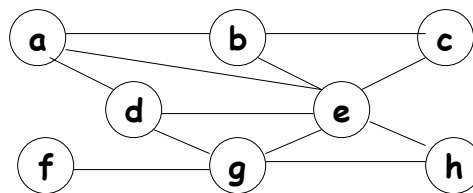
## Graph Algorithms - Outline of Topics

- **Elementary Graph Algorithms** - Chapter 22
  - graph representation
  - breadth-first-search, depth-first-search, topological sort
- **Minimum Spanning Trees** - Chapter 23
  - Kruskal's and Prim's algorithms (greedy algorithms)
- **Single-Source Shortest Paths** - Chapter 24
  - Dijkstra's algorithm (greedy algorithm)

## Undirected Graphs

An **undirected graph**  $G = (V, E)$  consists of

- A set  $V = |V|$  of nodes (vertices) , and
- A set  $E = |E|$  of undirected edges represented by node pairs



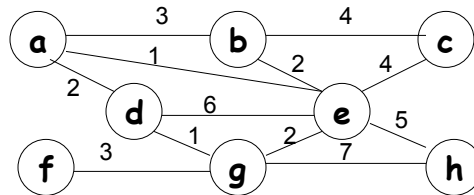
In this graph, both (b,c) and (c,b) are edges.

## Graph Terminology

**sparse graph:**  $|E|$  is  $o(|V|^2)$ .

**dense graph:**  $|E|$  is  $\Theta(|V|^2)$ .

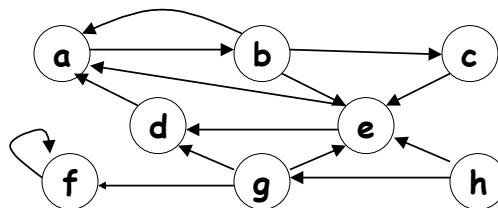
**weighted graph:** each edge has a number, called a *weight* attached to it. The weight is usually a positive number and may represent distance, cost, etc.



## Graph Terminology

A **directed graph (digraph)**  $G = (V, E)$  consists of

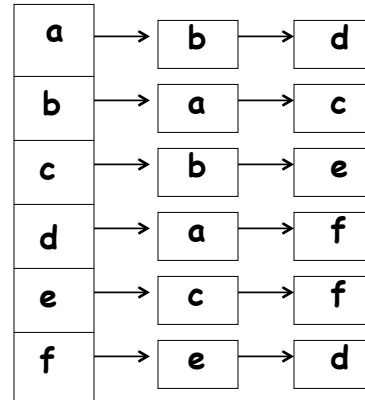
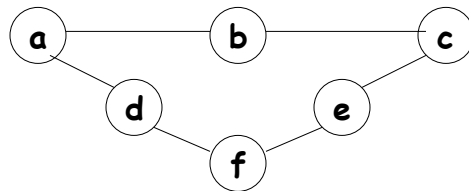
- A set  $V$  of nodes (vertices) , and
- A set  $E$  of *unidirectional* edges (represented by arrows)
- Self-loops are possible (as shown on node f)



Note: In this graph, (b,c) is an edge, but (c,b) is not an edge.

## Adjacency List Graph Representation

**Adjacency list:** An array  $A[1, |V|]$  of lists, one for each node  $v \in V$  (vertex set). Each node  $v$ 's list contains pointers to all nodes adjacent to  $v$  in  $G$ . Each edge repeated twice.



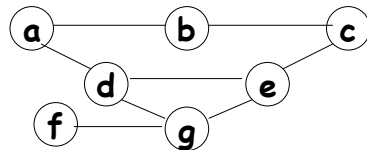
### Complexity issues

- advantage - storage is  $O(V + E)$   
(good for sparse graphs)
- drawback - list traversal to find edge

## Representing Undirected Graphs with Adjacency Matrices

**Adjacency matrix:** An array  $A[V, V]$  such that

$A[i, j] = 1$  if  $(i, j) \in E$  and  
0 otherwise



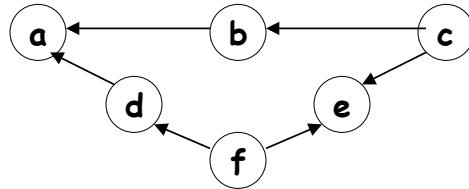
	a	b	c	d	e	f	g
a	0	1	0	1	0	0	0
b	1	0	1	0	0	0	0
c	0	1	0	0	1	0	0
d	1	0	0	0	1	0	1
e	0	0	1	1	0	0	1
f	0	0	0	0	0	0	1
g	0	0	0	1	1	1	0

### Complexity issues

- advantage -  $O(1)$  time to check for edge
- drawback - storage is  $O(V^2)$  (practical only for dense graphs)

In undirected graph, only the entries above the upper left to lower right diagonal need to be stored.

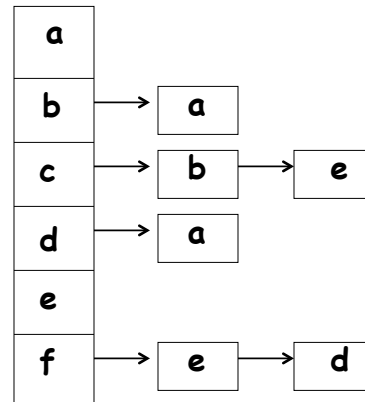
## Representing Digraphs with Adjacency Lists



### Complexity issues

- advantage - storage is  $O(V + E)$   
Good for sparse graphs, and most graphs we will use are sparse
- drawback - list traversal to find edge

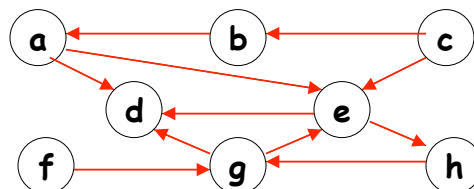
Only store outgoing edges in adj. lists



If  $(u,v)$  is an edge, then it is **incident** on both  $u$  and  $v$  and we say vertex  $v$  is **adjacent** to vertex  $u$ . The same terms hold for undirected graphs. Adjacent vertices are called **neighbors**

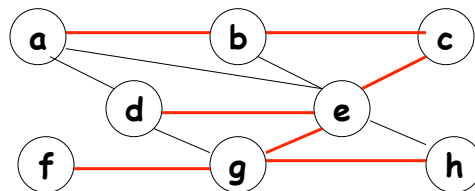
The **degree** of a node in an undirected graph is the number of edges incident on it.

The **in-degree** of a node in a digraph is the number of edges entering it and its **out-degree** is the number of edges leaving it.



A **path** of length  $k$  from a node  $u$  to a node  $u'$  is a sequence  $(v_0, v_1, \dots, v_k)$  of nodes such that  $u = v_0$ ,  $u' = v_k$  and there is an edge between each  $v_i$ ,  $i = 0, 1, 2, \dots, k$ . In a digraph, a path exists between nodes  $a$  and  $b$  only if there is a sequence of *outgoing* edges from  $a$  to  $b$

If there is a path  $p$  between vertices  $u$  and  $v$ , we say  $v$  is **reachable** from  $u$  via  $p$ .



A **simple** path has all distinct vertices.

The red edges in this graph trace a simple path between each pair of nodes.

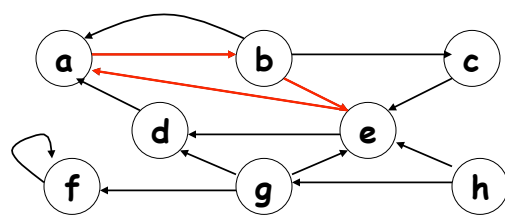
An undirected graph is **connected** if there is a simple path between every pair of nodes. A graph may have several **connected components** that are disjoint subsets of nodes.

A **completely connected** graph is an undirected graph in which every pair of nodes is adjacent.

In a digraph, a path  $(v_0, v_1, \dots, v_k)$  forms a **cycle** if  $v_0 = v_k$  and the path contains at least one edge

The cycle is **simple** if, in addition,  $v_1, v_2, \dots, v_k$  are distinct.

A digraph with no cycles is called a *directed acyclic graph*, abbreviated DAG



Not a DAG

## Breadth-First Search

**Breadth-First Search** finds the shortest-path distance (number of edges) between a source node and every other node in  $G$ .

Called breadth-first because it discovers all vertices at distance  $k$  from a *source node*  $s$  before it discovers any vertices at distance  $k+1$  from  $s$ , spanning the breadth before the depth of  $G$ .

BFS finds all vertices  $v$  that are reachable from  $s$  by building a breadth-first tree, where the path in the tree from  $s$  to  $v$  has the fewest number of edges of all paths from  $s$  to  $v$ .

## Breadth-First Search

**Breadth-First Search** has time complexity of  $O(V + E)$  and is often used as a building block of other algorithms.

BFS is particularly useful in finding shortest paths on un-weighted graphs.

BFS starts at a node  $s$  in a graph and explores all its neighbor nodes before moving to the next level (neighbors of neighbors).

Explores nodes in "layers".

Maintains a queue of nodes to keep track of which node it should visit next.

## Breadth-First Search Implementation

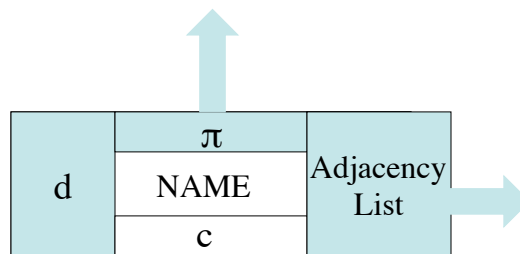
The algorithm from our book maintains a FIFO queue,  $Q$ , to manage the set of nodes and starts by enqueueing  $s$ , the source node

BFS algorithm maintains the following information for each vertex  $u$ :

- $u.c$ : white, gray, or black to indicate status
  - white = not discovered yet; initially, all nodes except  $s$  are undiscovered.
  - gray = discovered, but not finished; initially only  $s$ .
  - black = finished; initially none are finished.
- $u.d$  : distance from  $s$  to  $u$ ; initially  $\infty$  for all but  $s.d=0$
- $u.\pi$  : predecessor of  $u$  in BF tree; initially NIL for all ( $s.\pi = \text{NIL}$  and remains NIL)

## BFS node

Each node has fields for predecessor ( $\pi$ ), distance from source, and color. Each node also has an associated adjacency list with pointers to neighboring nodes.



## Breadth-First Search

BFS ( $G, s$ ):

0.  $s.c = \text{gray}$ ;  $s.\pi = \text{NIL}$
1.  $Q.\text{enqueue}(s)$  //  $Q$  is a FIFO ds
2. **while**  $Q \neq \emptyset$
3.      $u = Q.\text{dequeue}()$
4.     **for each**  $v$  adjacent to  $u$
5.         **if**  $v.c == \text{white}$
6.              $v.c = \text{gray}$
7.              $v.d = u.d + 1$
8.              $v.\pi = u$
9.              $Q.\text{enqueue}(v)$
10.      $u.c = \text{black}$

$Q.\text{enqueue}(s)$  adds  $s$  to the rear of  $Q$

$Q.\text{dequeue}()$  removes and returns the item at the head of  $Q$

**Note:** If  $G$  is not connected, then BFS will not visit the entire graph (without some extra provisions in the algorithm)



## Breadth-First Search

BFS ( $G, s$ ):

0.  $s.c = \text{gray}; s.\pi = \text{NIL}$
1.  $Q.\text{enqueue}(s)$  //  $Q$  is a FIFO ds
2. **while**  $Q \neq \emptyset$
3.      $u = Q.\text{dequeue}()$
4.     **for each**  $v$  adjacent to  $u$
5.         **if**  $v.c == \text{white}$
6.              $v.c = \text{gray}$
7.              $v.d = u.d + 1$
8.              $v.\pi = u$
9.              $Q.\text{enqueue}(v)$
10.     $u.c = \text{black}$

### Complexity (Adjacency List)

- each node enqueued and dequeued once =  $O(V)$  time
- each edge considered once (in each direction on undirected  $G$ ) =  $O(E)$  time
- total =  $O(V + E)$   
=  $O(V^2)$  (w-c)

## Analysis of Breadth-First Search

**Shortest-path distance  $\delta(s, v)$**  : minimum number of edges in any path from vertex  $s$  to  $v$ . If no path exists from  $s$  to  $v$ , then  $\delta(s, v) = \infty$ .

The ultimate goal of the proof of correctness is to show that  $v.d = \delta(s, v)$  when the algorithm is done and that a path is found from  $s$  to all reachable vertices.

- L. 22.1 : children of a node  $u$  are given a higher  $d$  value than  $u$ .
- L. 22.2 : for every edge  $(u, v)$ , the shortest path from  $s$  to  $v$  can be no longer than the (shortest path from  $s$  to  $u$ ) + 1.
- L. 22.3 : at any time,  $Q$  holds at most 2 distinct  $d$  values. I.e., the range of values in  $Q$  is at most 2. Why?
- C. 22.4: the  $d$  values are monotonically increasing over time as the algorithm runs.

**Theorem 22.5: (Correctness of BFS)**

Let  $G = (V, E)$  be a directed or undirected graph, and suppose that BFS is run from a given source vertex  $s \in V$ . Then, during execution, BFS discovers every vertex  $v \neq s$  that is reachable from the source  $s$ , and upon termination,  $v.d = \delta(s, v)$  for every *reachable* or *unreachable* vertex  $v$ .

Proof by contradiction.

Assume that for some vertex  $v$  that  $v.d \neq \delta(s, v)$  after running BFS. Also, assume that  $v$  is the vertex with *minimum*  $\delta(s, v)$  that receives an incorrect  $d$  value. By Lemma 22.2, it must be that  $v.d > \delta(s, v)$ .

Case 1:  $v$  is not reachable from  $s$ . This is a contradiction to the assumption that  $v$  is reachable, and the Thm holds.

Case 2:  $v$  is reachable from  $s$ . Let  $u$  be the vertex immediately preceding  $v$  on a shortest path from  $s$  to  $v$ , so that  $\delta(s, v) = \delta(s, u) + 1$ . Because  $\delta(s, u) < \delta(s, v)$  and because  $v$  is the vertex with the *minimum*  $\delta(s, v)$  that receives an incorrect  $d$  value,  $u.d = \delta(s, u)$ .

So we have  $v.d > \delta(s, v) = \delta(s, u) + 1 = u.d + 1$ .

Consider the time  $t$  when  $u$  is dequeued. At time  $t$ ,  $v$  is either white, gray, or black. We can derive a contradiction in each of these cases.

Case 1:  $v$  is white. Then in line 12,  $v.d = u.d + 1$ .

Case 2:  $v$  is black. Then  $v$  was already dequeued, and therefore  $v.d \leq u.d$  (by L. 22.3).

Case 3:  $v$  is gray. Then  $v$  turned gray when it was visited from some vertex  $w$ , which was dequeued before  $u$ . Then  $v.d = w.d + 1$ . Since  $w.d \leq u.d$  (by L. 22.3),  $v.d \leq u.d + 1$ .

Each of these cases is a contradiction to  $v.d > \delta(s, v)$ , so we conclude that  $v.d = \delta(s, v)$ . ■

## Breadth-First Trees

BFS builds a breadth-first tree that can be identified by using the  $\pi$  values at each node.

The edges defined by each  $v.\pi$  are called tree edges.

```
Print-Path (G, s, v) // finds the tree edges between s and v,
1. if v == s           // starting at v
2.   print s
3. else
4.   if v. $\pi$  == NIL
5.     print "no path from " s " to " v " exists"
6.   else
7.     Print-Path(G, s, v. $\pi$ )
8.     print v
```

## Breadth-First Search v2

BFS (G, s):

0. let *marked* be a boolean array of size  $|V|$  // init all false
1. let *edgeTo* be an array of  $|V|$  integers
2. Q.enqueue (s)
3. marked[s] = true
4. **while** Q  $\neq \emptyset$
5. u = Q.dequeue()
6. **for each** v adjacent to u
7. **if** marked[v] == false
8. edgeTo[v] = u
9. marked[v] = true
10. Q.enqueue(v)

## Enumerating shortest path, $s \rightarrow v$

pathTo(v):

1. if (!marked[v]) return false
2. Stack<Integer> path = new Stack<Integer>()
3. for (int x = v; x != s; x = edgeTo[x])
4.     path.push(x)
5. path.push(s)
6. return path

When pathTo finishes, the path will contain the path from s to v and they can be popped off the stack in order.

## Analysis of Breadth-First Search

**Proposition A1:** For any vertex v reachable from s, BFS computes a shortest path from s to v such that no path from s to v has fewer edges.

*Informal proof:*

It is easy to prove by induction that Q always consists of zero or more vertices of distance k from s, followed by zero or more vertices of distance k+1 from s, for some integer k, starting with k = 0. This property implies, in particular, that vertices enter and leave Q in order of their increasing distance from s. When a vertex v enters Q, no shorter path to v will be found before v comes off Q, and no path to v that is discovered after v comes off Q can be shorter than the path length s to v.

## Analysis of Breadth-First Search

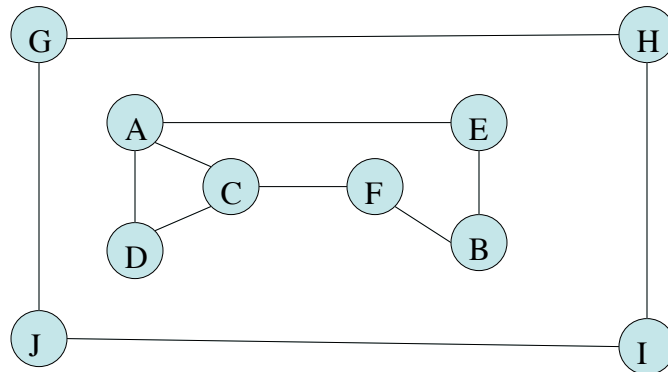
**Proposition A2: BFS takes time proportional to  $V + E$  in the worst case.**

*Informal proof:*

BFS marks all the vertices connected to  $s$  in time proportional to the degree of  $s$ . If the graph is connected, this sum equals the sum of the degrees of all the vertices, or  $2E$ .

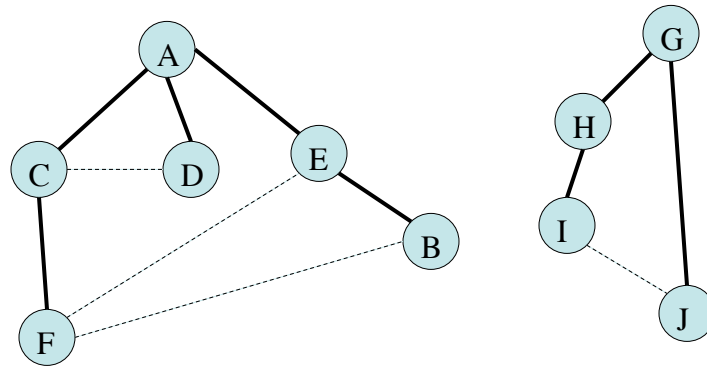
Initializing the `marked[ ]` and `edgeTo[ ]` arrays takes time proportional to  $V$ .

### Example BFS Traversal



Order of visiting:	$a_1$	$c_2$	$d_3$	$e_4$	$f_5$	$b_6$	$g_7$	$h_8$	$i_9$	$j_{10}$
Distance of vertex :	0	1	1	1	2	2	$\infty$	$\infty$	$\infty$	$\infty$

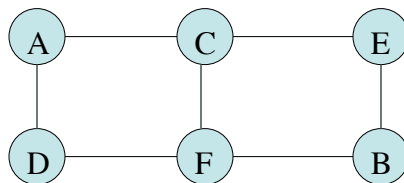
## Breadth-first Search Forest



Tree edges are solid lines and dashed lines are cross edges.

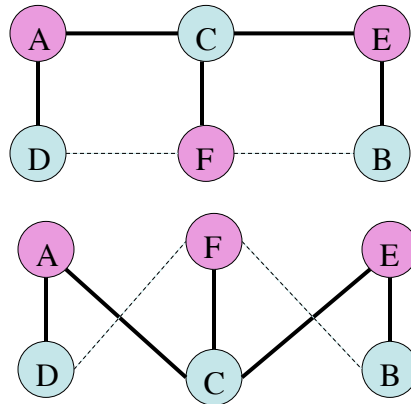
## Bipartite Graphs

A graph is bipartite if all its vertices can be partitioned into two disjoint subsets  $X$  and  $Y$  so that every edge connects a vertex in  $X$  with a vertex in  $Y$ , i.e., if its vertices can be colored in 2 colors so that every edge has its end points colored in different colors.



## Bipartite Graphs

Explain how BFS could be used to detect a bipartite graph.

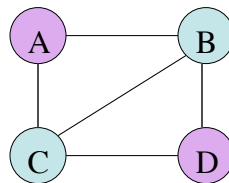


Mark the source, A, with color1, mark the nodes at level 1 with color2, and so on.

Every node on an even numbered level will be color1 and on every odd level color2

## Bipartite Graphs

Is this graph bipartite?



## Applications of BFS

Based upon the BFS, there are  $O(V + E)$ -time algorithms for the following problems:

- Testing whether graph is connected.
- Computing a spanning forest of graph.
- Computing, for every vertex in graph, a path with the minimum number of edges between start vertex and current vertex or reporting that no such path exists.
- Computing a cycle in graph or reporting that no such cycle exists.