

Applied Algorithms

CSCI-B505 / INFO-I500

Lecture 17.

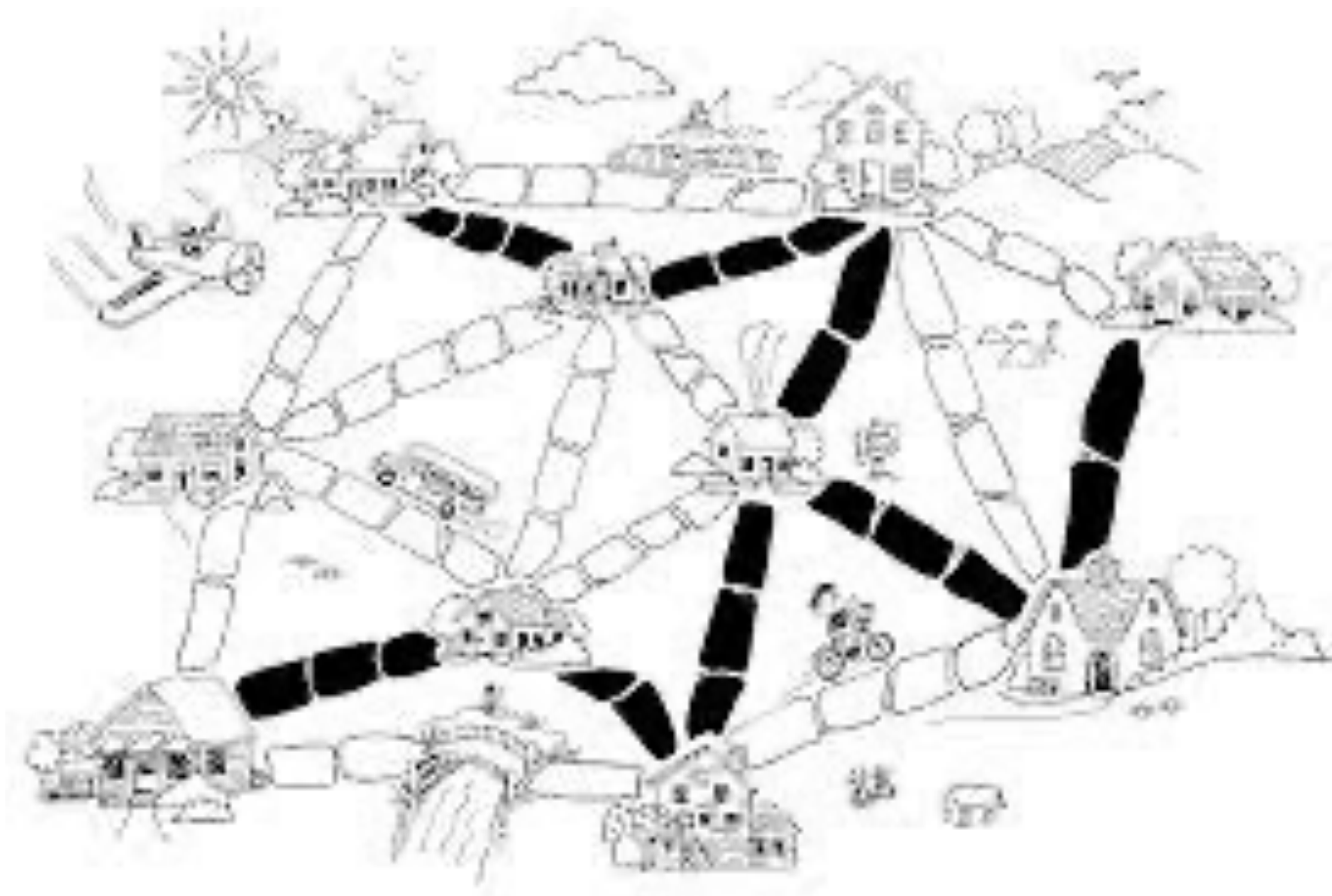
Graph Data Structures and Algorithms - I

M. Oguzhan Kulekci

- Graphs
- How to represent graphs
- Graph Traversals

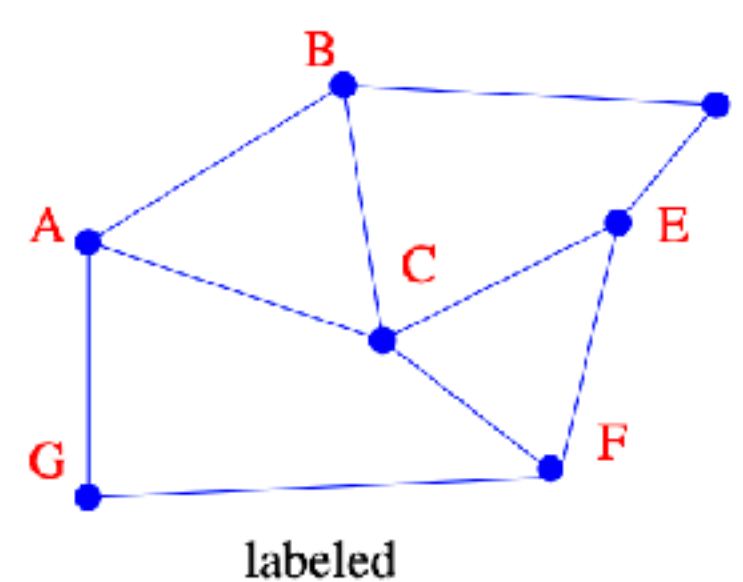
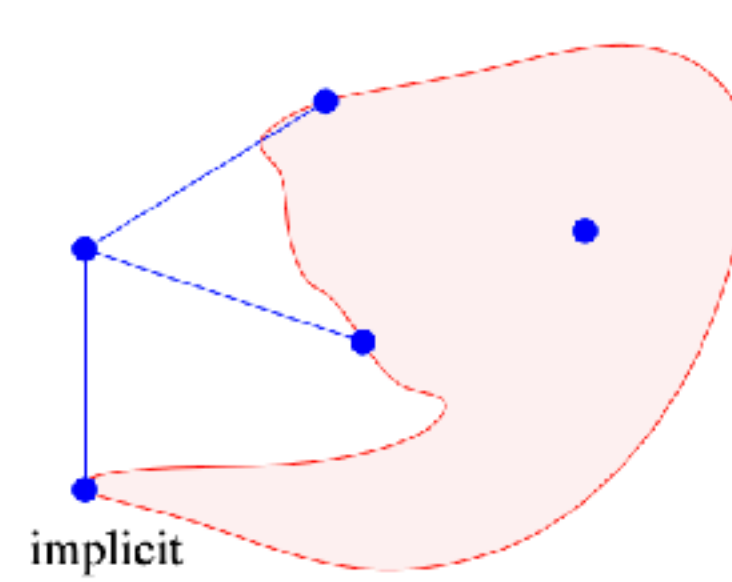
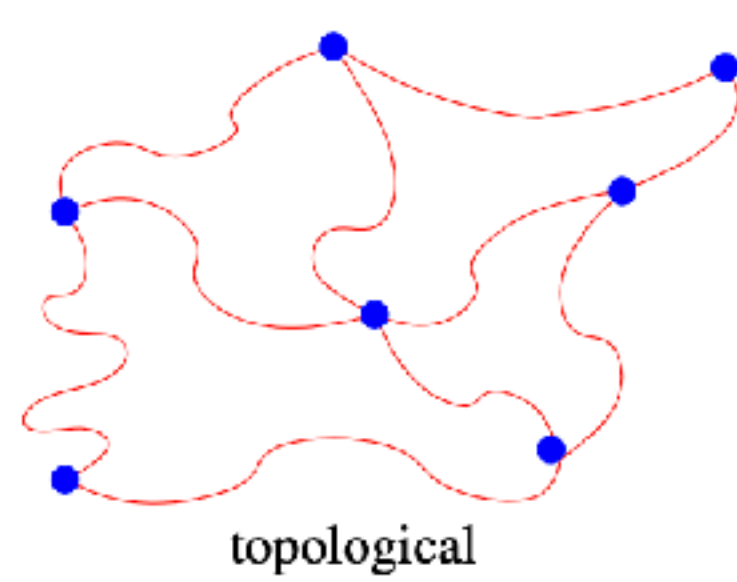
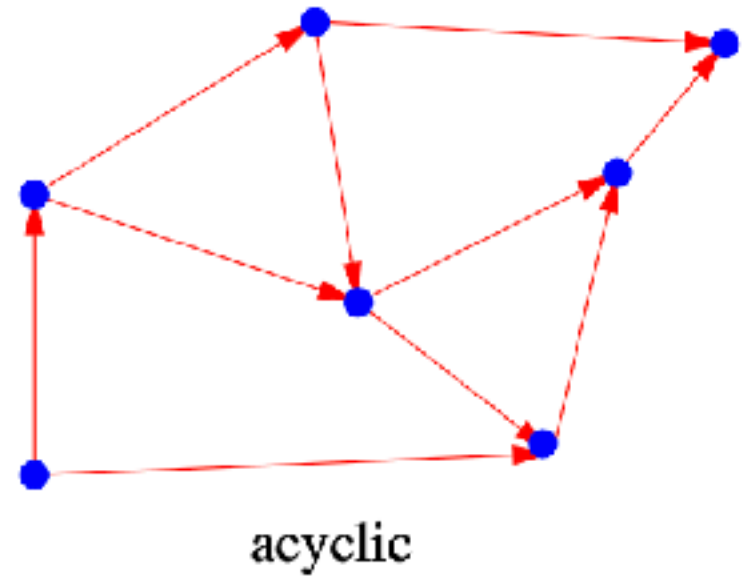
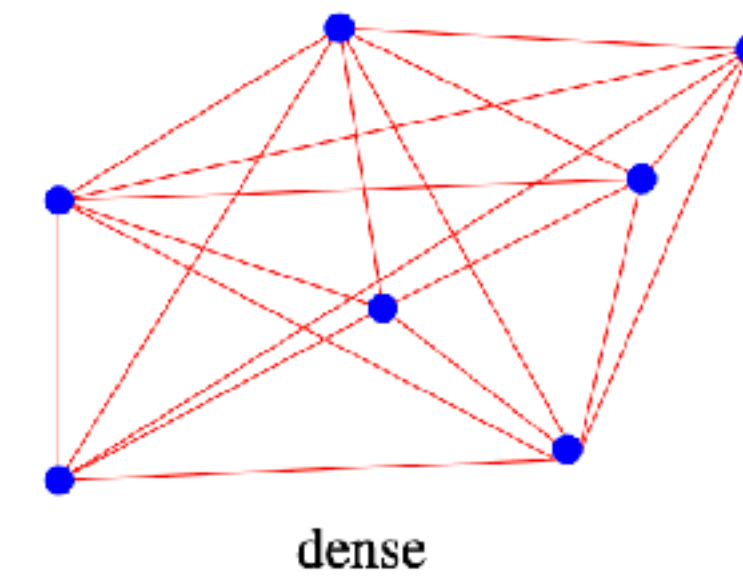
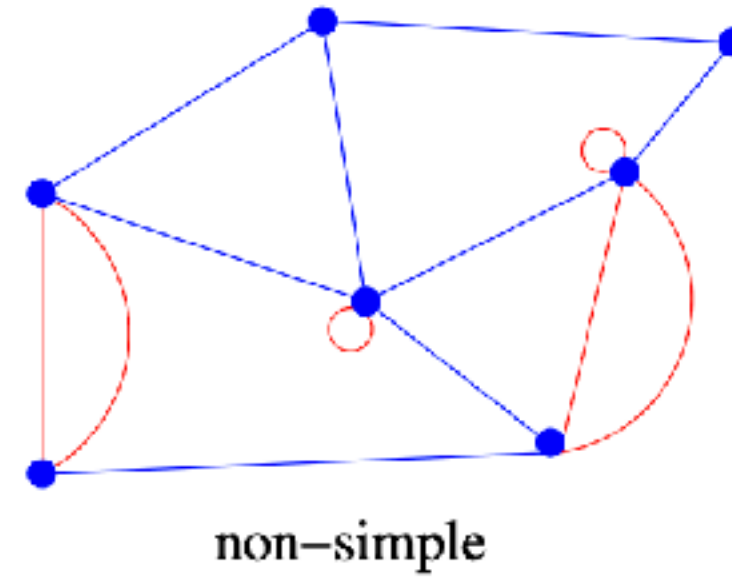
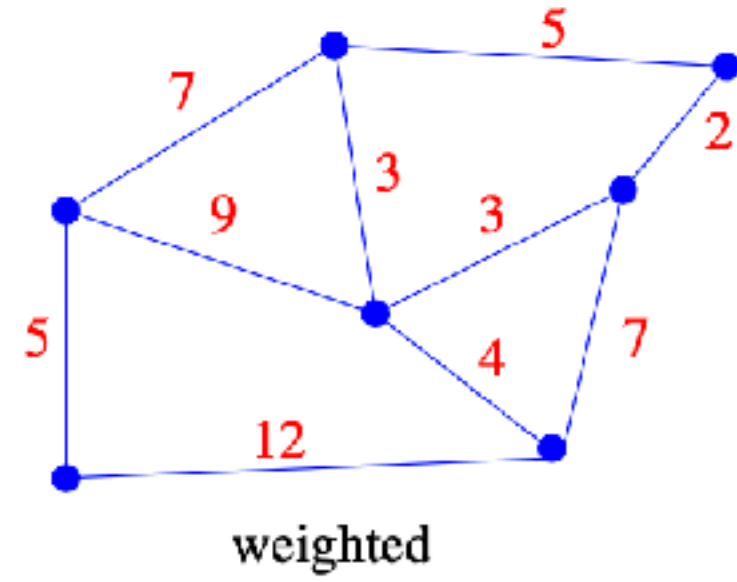
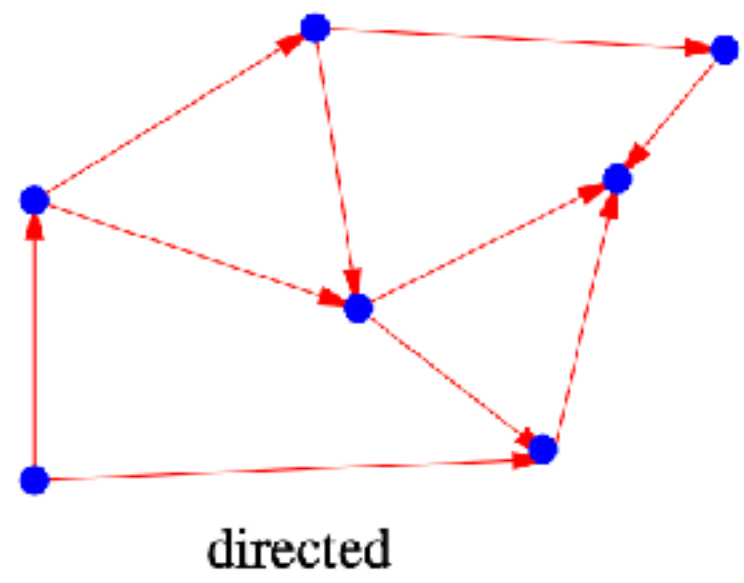
Graphs

- Graphs encode many practical real life scenarios
- Major usage: Modeling our problem with a graph representation and investigating which graph algorithms can help us to solve it.
- Notice that it is very hard to come up with a novel graph algorithm



Some terminology

- $G(V,E)$ represents a graph.



Graph Algorithms - General View

Polynomial Time

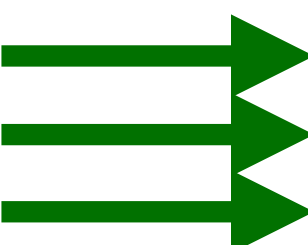
- Connected components
- Minimum spanning tree
- Shortest path
- Eulerian cycle
- Edge/vertex connectivity
- Transitive closure
- Network flow
- Planarity testing
- Graph drawing
-

NP-Hard

- Clique
- Independent set/vertex cover
- Traveling salesman
- Hamiltonian cycle
- Vertex/edge coloring
- Graph partitioning
- Graph isomorphism
-

Representing Graphs

- Different data structures
- Important notice: The complexity of graph algorithms strongly depend on the data structure used to represent the graph.

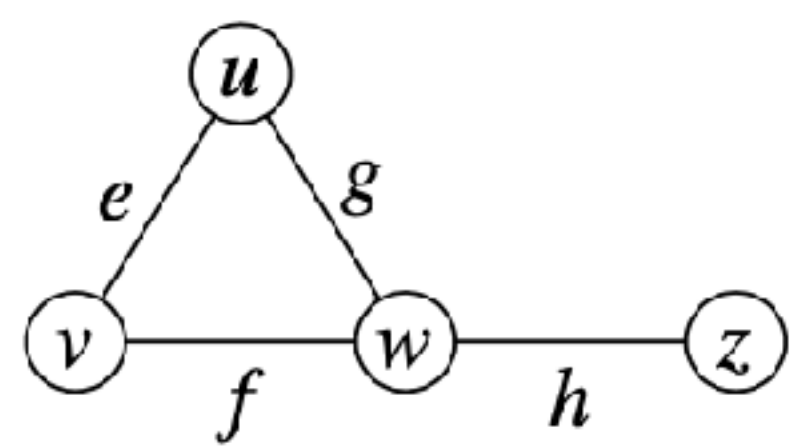


Operation	Edge List	Adj. List	Adj. Map	Adj. Matrix
vertex_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
edge_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
vertices()	$O(n)$	$O(n)$	$O(n)$	$O(n)$
edges()	$O(m)$	$O(m)$	$O(m)$	$O(m)$
get_edge(u,v)	$O(m)$	$O(\min(d_u, d_v))$	$O(1)$ exp.	$O(1)$
degree(v)	$O(m)$	$O(1)$	$O(1)$	$O(n)$
incident_edges(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n)$
insert_vertex(x)	$O(1)$	$O(1)$	$O(1)$	$O(n^2)$
remove_vertex(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n^2)$
insert_edge(u,v,x)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$
remove_edge(e)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$

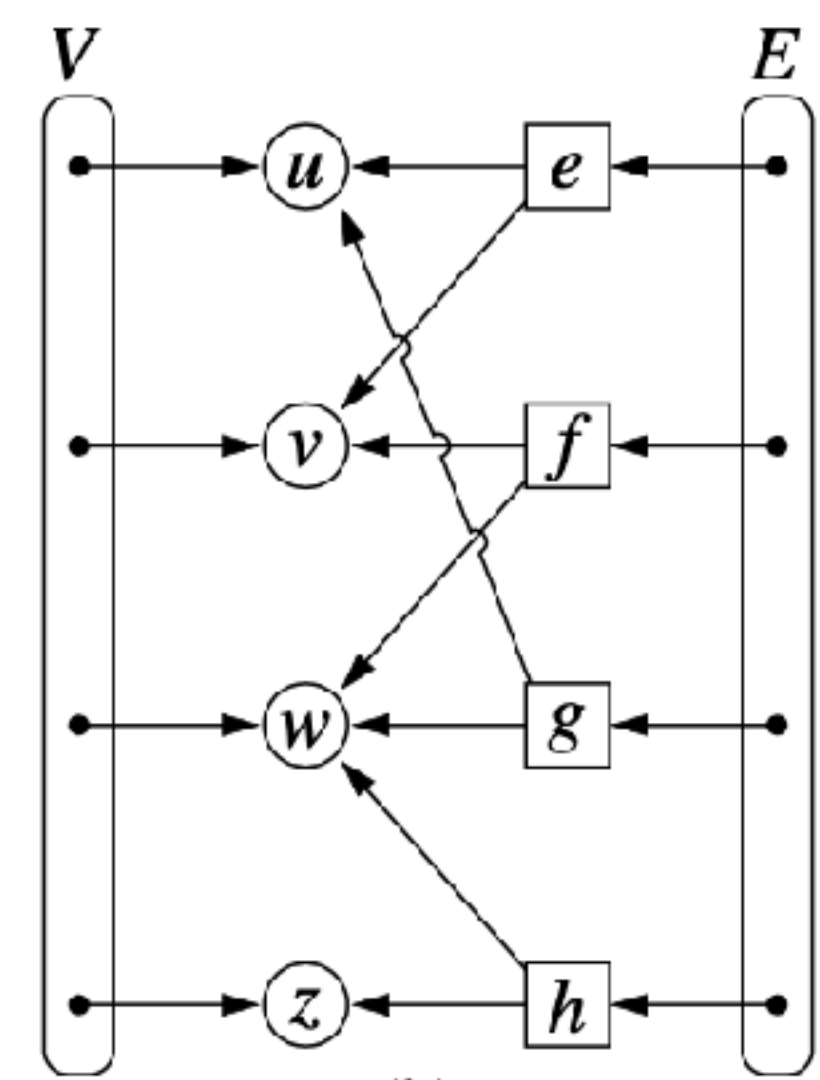
Table 14.1: A summary of the running times for the methods of the graph ADT, using the graph representations discussed in this section. We let n denote the number of vertices, m the number of edges, and d_v the degree of vertex v . Note that the adjacency matrix uses $O(n^2)$ space, while all other structures use $O(n + m)$ space.

- Edge list
- Adjacency list/map
- Adjacency matrix

Edge List



(a)



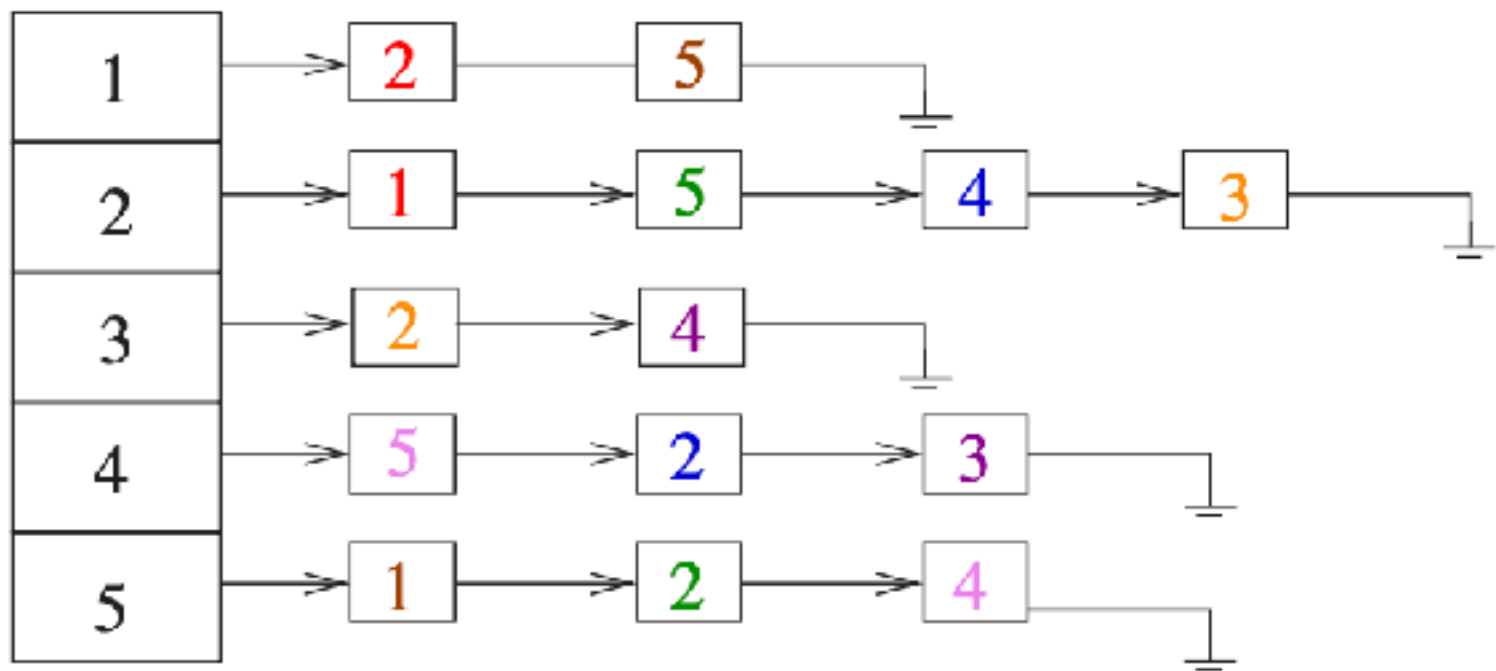
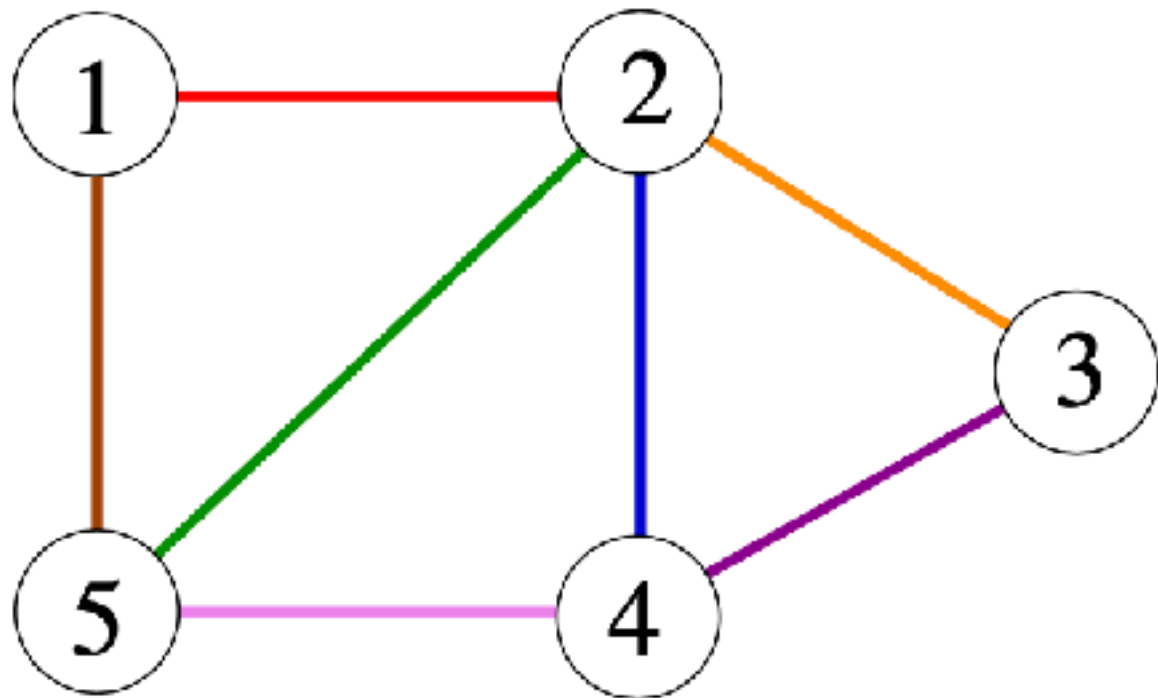
(b)

Operation	Running Time
vertex_count(), edge_count()	$O(1)$
vertices()	$O(n)$
edges()	$O(m)$
get_edge(u,v), degree(v), incident_edges(v)	$O(m)$
insert_vertex(x), insert_edge(u,v,x), remove_edge(e)	$O(1)$
remove_vertex(v)	$O(m)$

Adjacency Matrix / List

Comparison	Winner
Faster to test if (x, y) is in graph?	adjacency matrices
Faster to find the degree of a vertex?	adjacency lists
Less memory on sparse graphs?	adjacency lists $(m + n)$ vs. (n^2)
Less memory on dense graphs?	adjacency matrices (a small win)
Edge insertion or deletion?	adjacency matrices $O(1)$ vs. $O(d)$
Faster to traverse the graph?	adjacency lists $\Theta(m + n)$ vs. $\Theta(n^2)$
Better for most problems?	adjacency lists

	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	1	0



Adjacency Matrix

Operation	Edge List	Adj. List	Adj. Map	Adj. Matrix
vertex_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
edge_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
vertices()	$O(n)$	$O(n)$	$O(n)$	$O(n)$
edges()	$O(m)$	$O(m)$	$O(m)$	$O(m)$
get_edge(u,v)	$O(m)$	$O(\min(d_u, d_v))$	$O(1)$ exp.	$O(1)$
degree(v)	$O(m)$	$O(1)$	$O(1)$	$O(n)$
incident_edges(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n)$
insert_vertex(x)	$O(1)$	$O(1)$	$O(1)$	$O(n^2)$
remove_vertex(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n^2)$
insert_edge(u,v,x)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$
remove_edge(e)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$

← R/S Dictionaries, $O(1)$

Adjacency List

Operation	Running Time
vertex_count(), edge_count()	$O(1)$
vertices()	$O(n)$
edges()	$O(m)$
get_edge(u,v)	$O(\min(\deg(u), \deg(v)))$
degree(v)	$O(1)$
incident_edges(v)	$O(\deg(v))$
insert_vertex(x), insert edge(u,v,x)	$O(1)$
remove_edge(e)	$O(1)$
remove_vertex(v)	$O(\deg(v))$

Graph Traversal

- Visit every vertex in a graph
- A key operation in many graph algorithms, e.g., connected components, reachability, shortest-path, spanning-tree, cycle-detection etc...
- We need a systematic way to avoid again and again visiting the same vertex or get stuck somewhere in the graph
- So, keep track of visited vertices
- Breadth-first or depth-first

Breadth-First Graph Traversal

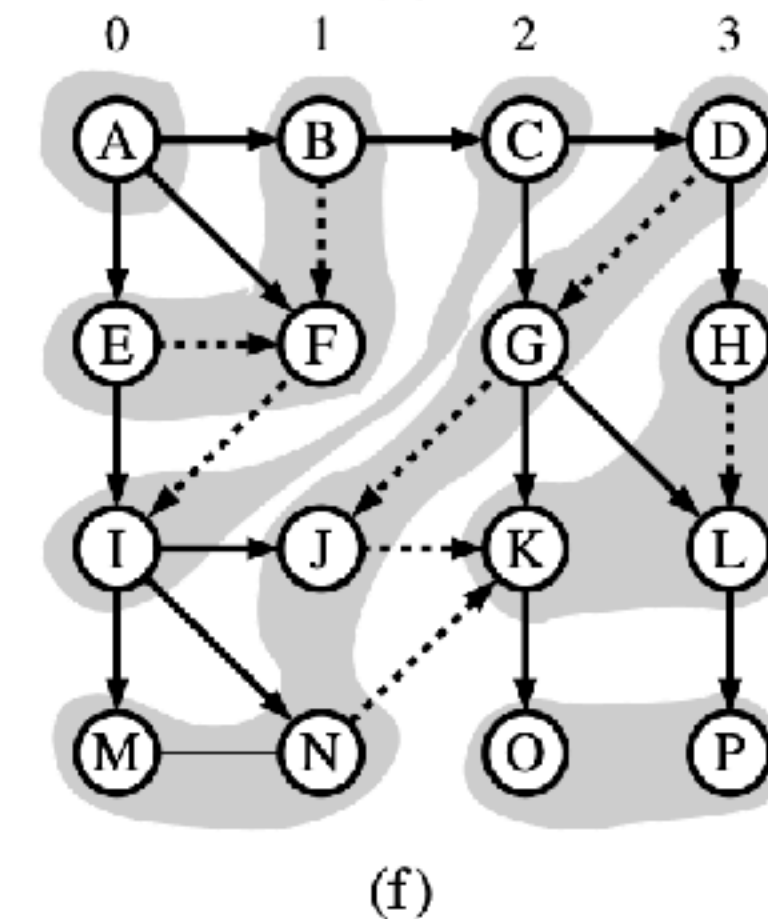
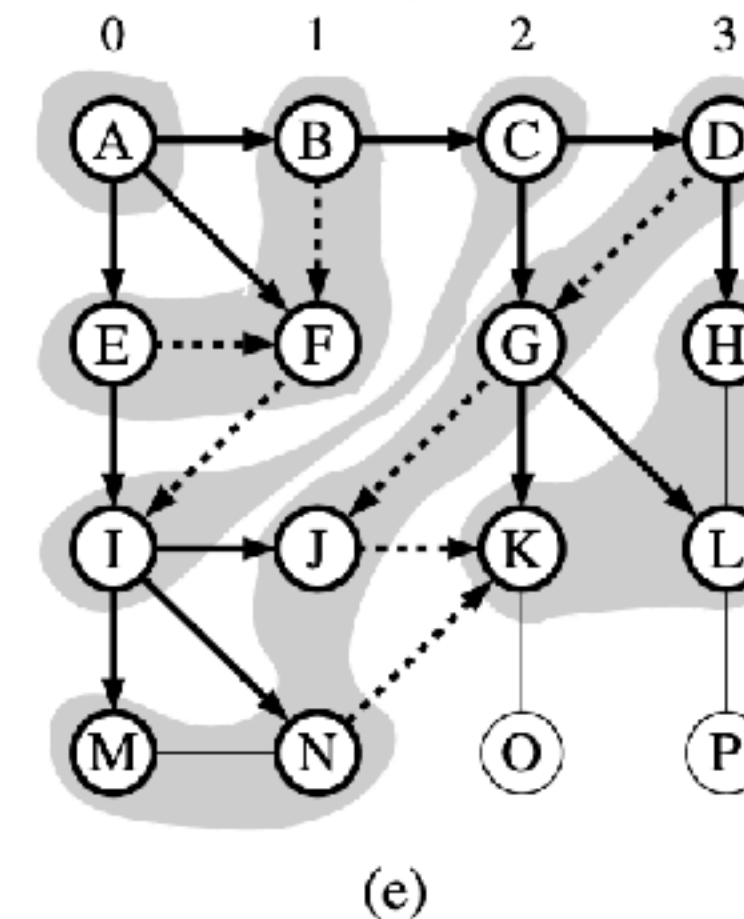
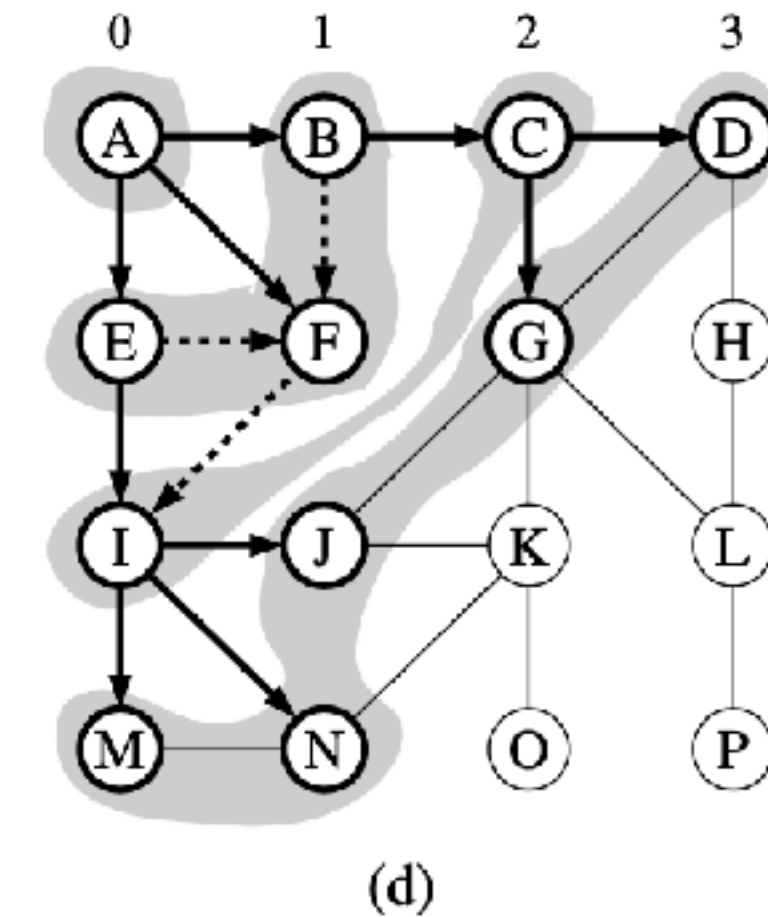
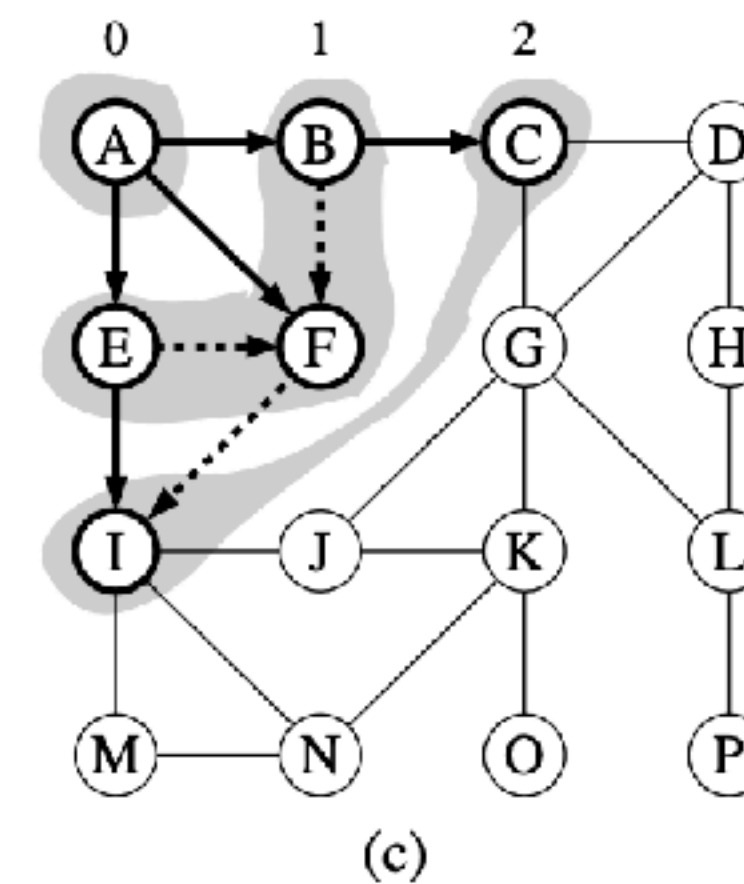
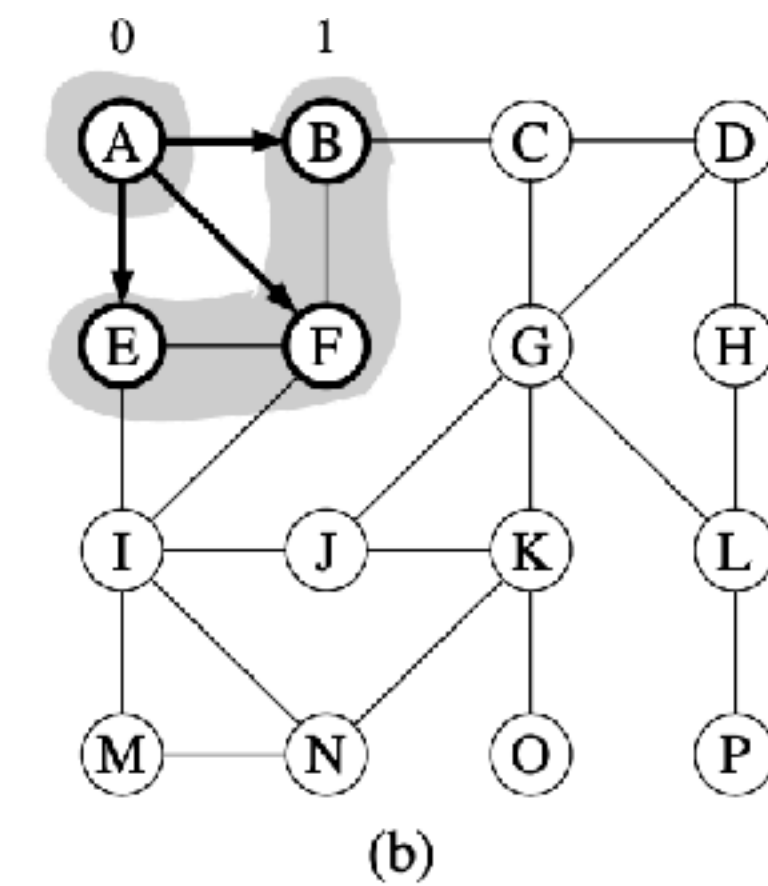
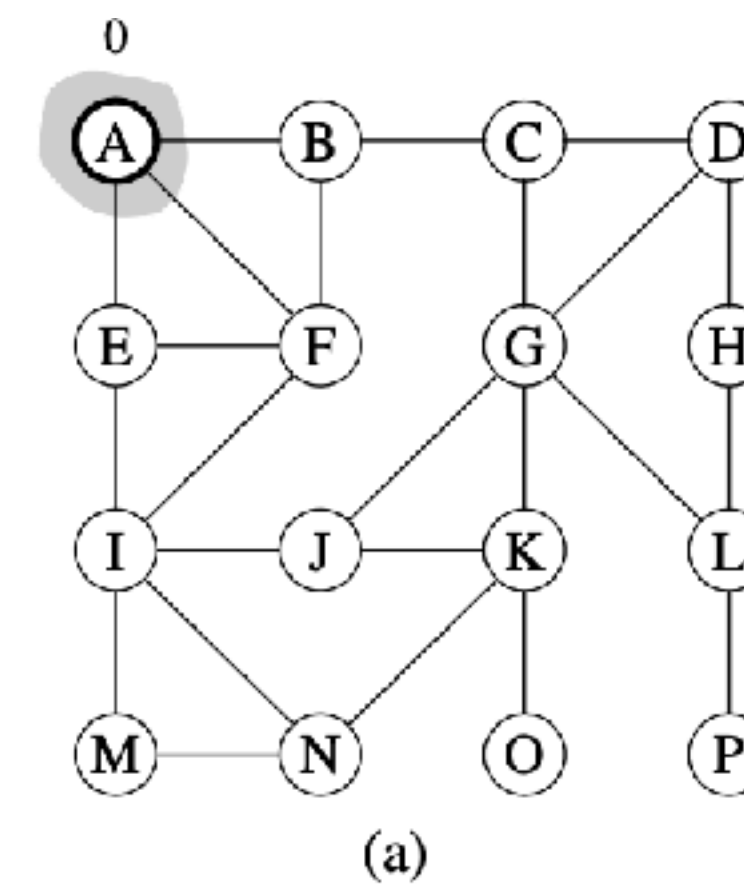
```

1 def BFS(g, s, discovered):
2     """Perform BFS of the undiscovered portion of Graph g starting at Vertex s.
3
4     discovered is a dictionary mapping each vertex to the edge that was used to
5     discover it during the BFS (s should be mapped to None prior to the call).
6     Newly discovered vertices will be added to the dictionary as a result.
7     """
8     level = [s]                # first level includes only s
9     while len(level) > 0:
10        next_level = []        # prepare to gather newly found vertices
11        for u in level:
12            for e in g.incident_edges(u): # for every outgoing edge from u
13                v = e.opposite(u)
14                if v not in discovered: # v is an unvisited vertex
15                    discovered[v] = e  # e is the tree edge that discovered v
16                    next_level.append(v) # v will be further considered in next pass
17        level = next_level      # relabel 'next' level to become current

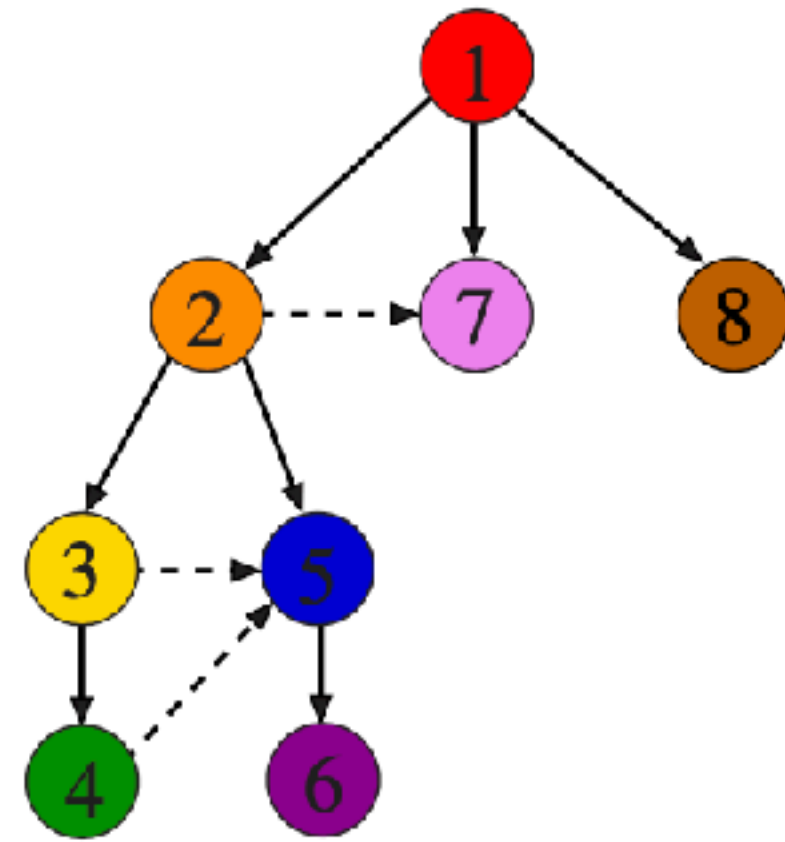
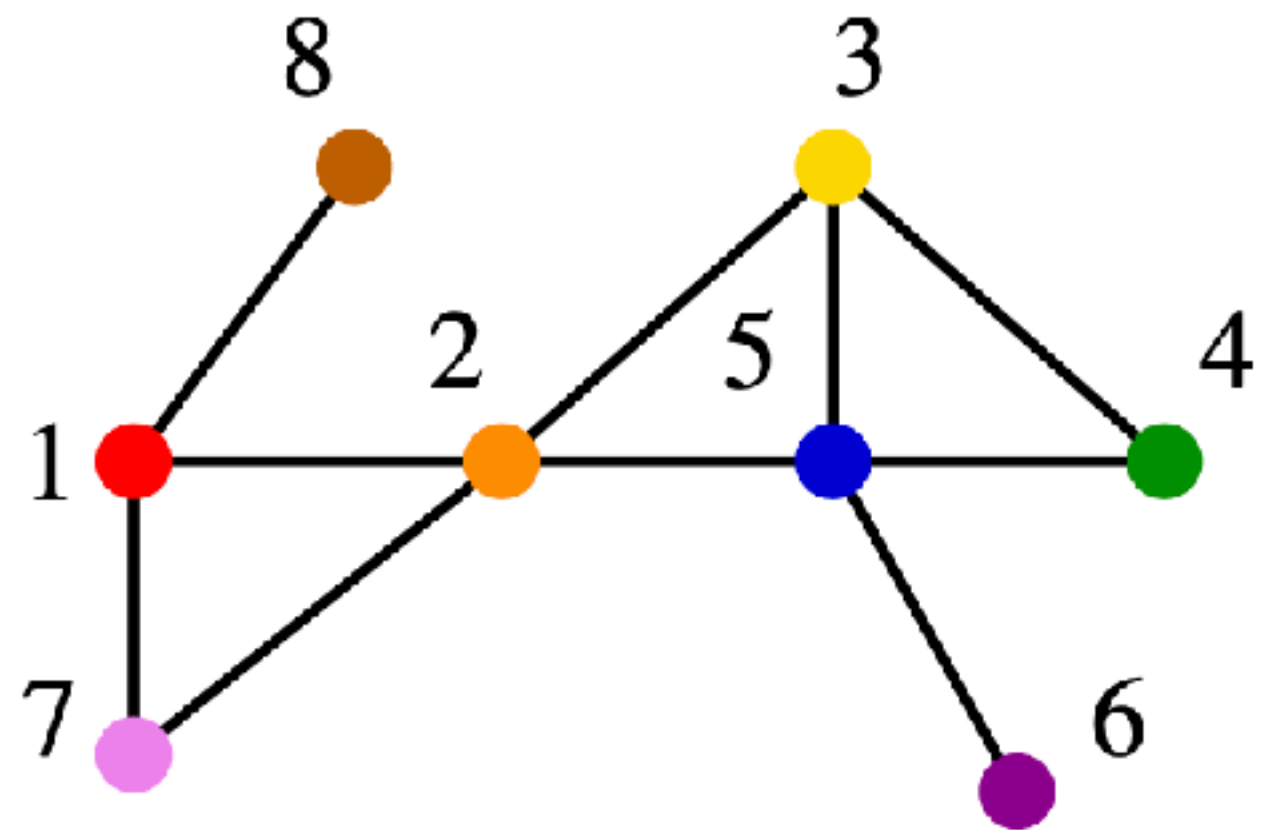
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$O(n + m)$ -time with adjacency list data structure

Breadth-first search procedure has some interesting properties....



Breadth-First Graph Traversal



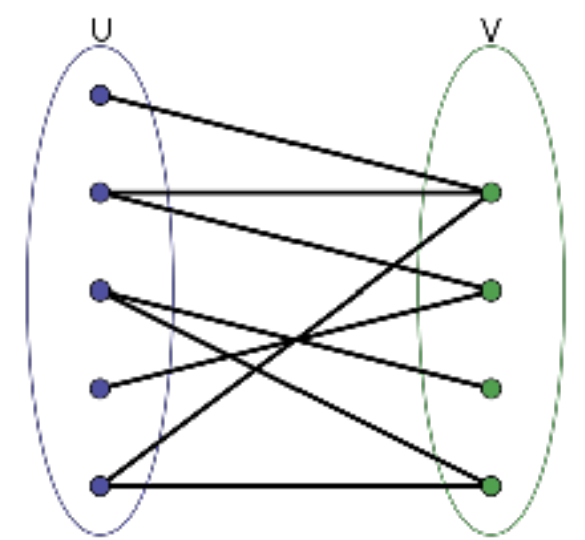
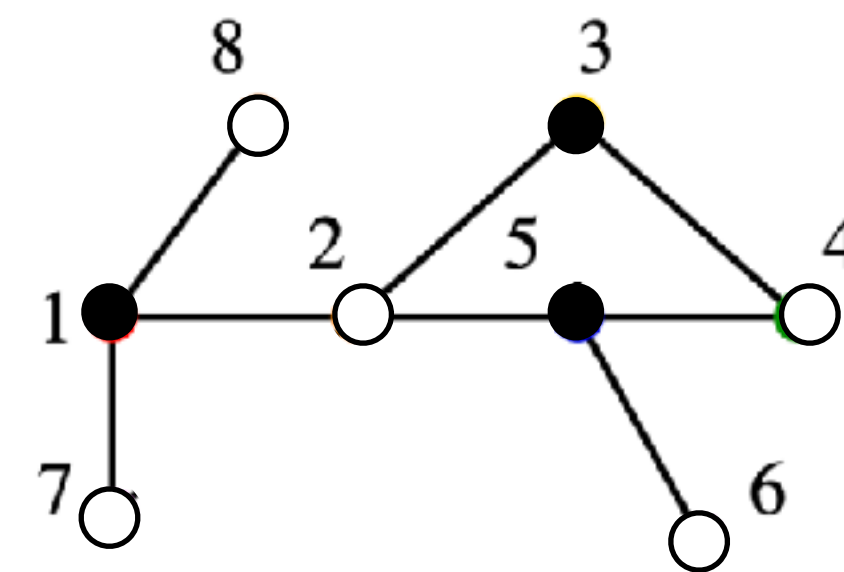
Proposition 14.16: Let G be an undirected or directed graph on which a BFS traversal starting at vertex s has been performed. Then

- The traversal visits all vertices of G that are reachable from s .
- For each vertex v at level i , the path of the BFS tree T between s and v has i edges, and any other path of G from s to v has at least i edges.
- If (u, v) is an edge that is not in the BFS tree, then the level number of v can be at most 1 greater than the level number of u .

“a path in a breadth- first search tree rooted at vertex s to any other vertex v is guaranteed to be the **shortest** such path from s to v in terms of the number of edges “

BFS can also be used in

- detecting the connected components of a graph
- two-coloring verification of the graph (is it a bipartite graph)



Depth-First Graph Traversal

Algorithm DFS(G,u): {We assume u has already been marked as visited}

Input: A graph G and a vertex u of G

Output: A collection of vertices reachable from u , with their discovery edges

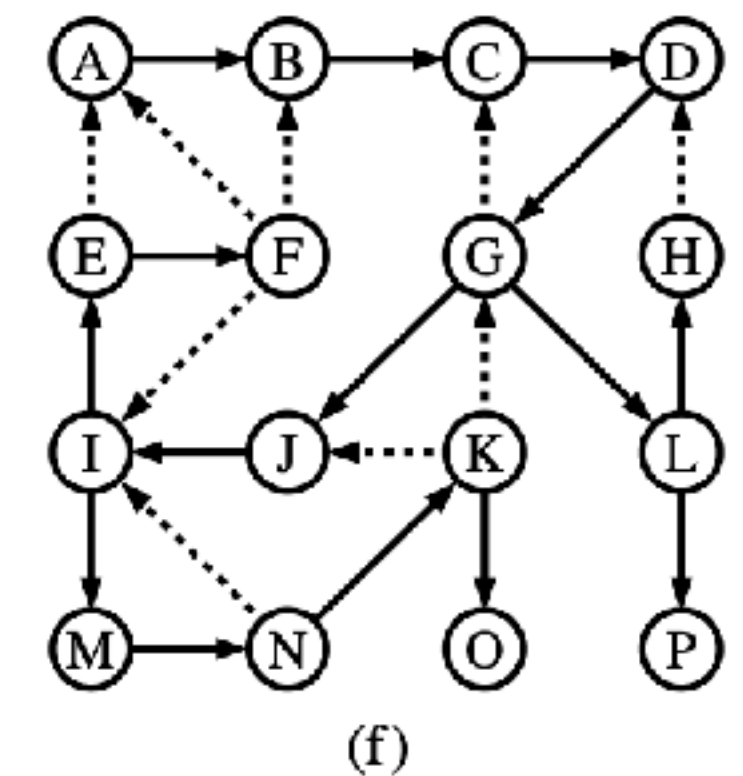
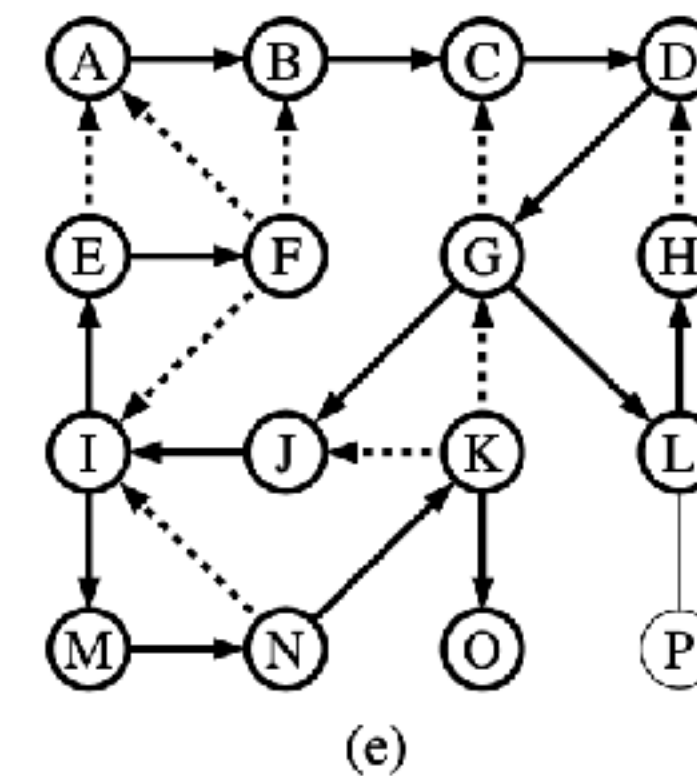
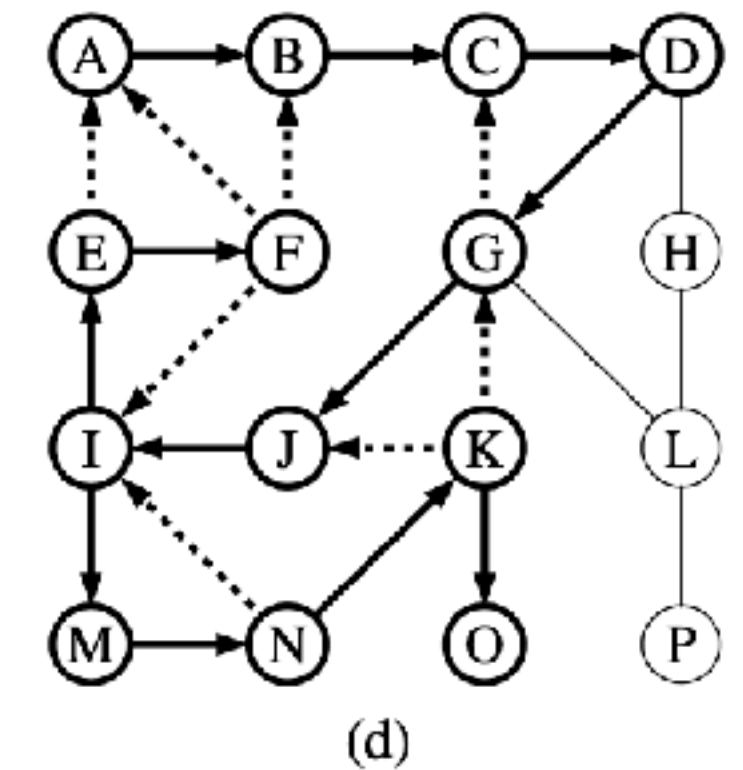
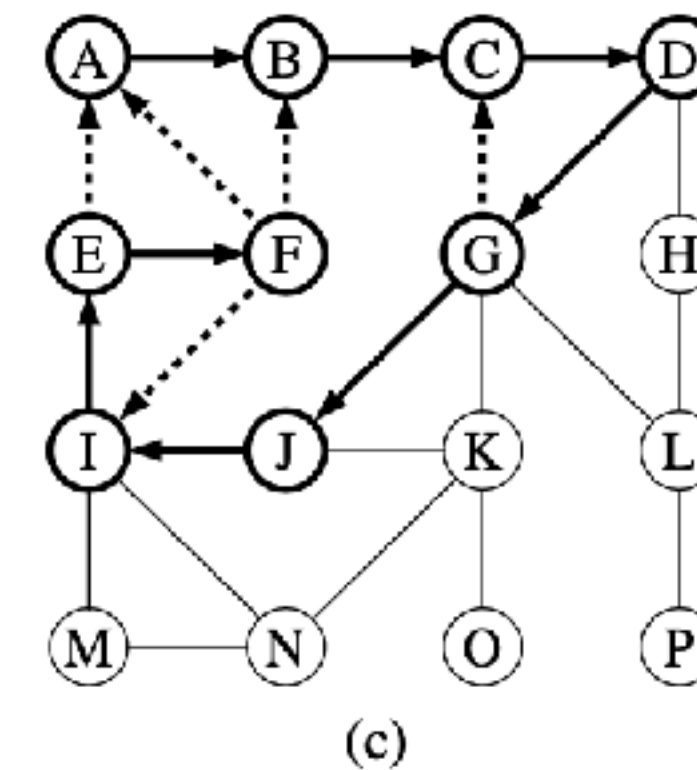
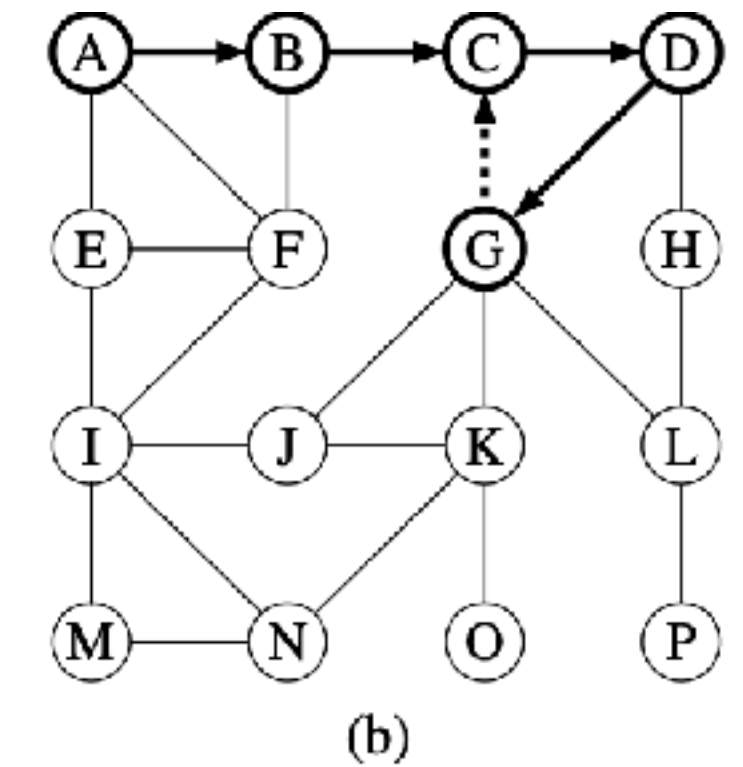
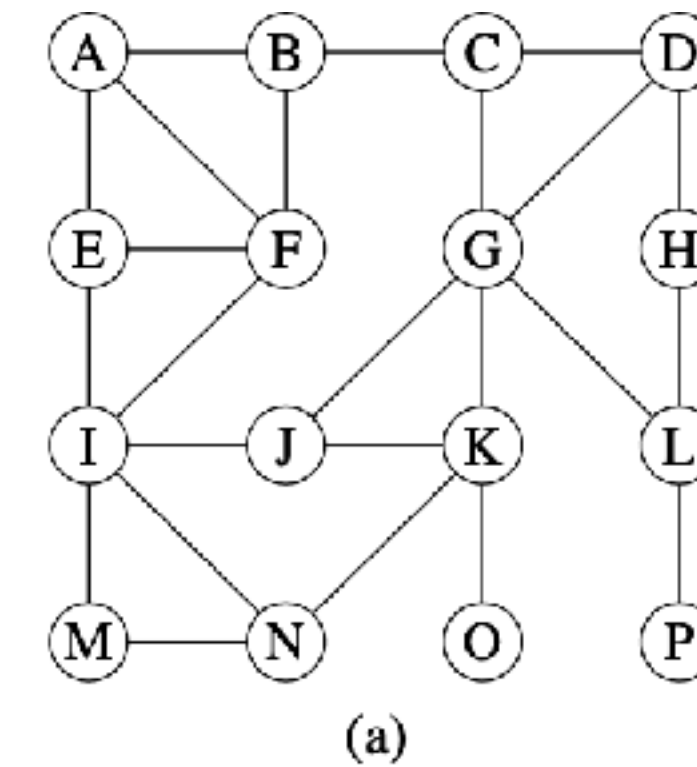
for each outgoing edge $e = (u, v)$ of u do

if vertex v has not been visited **then**

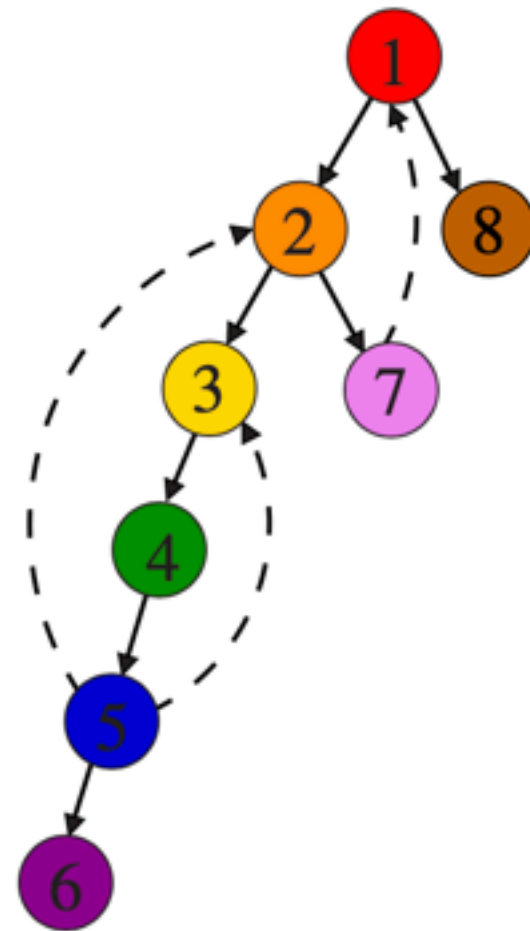
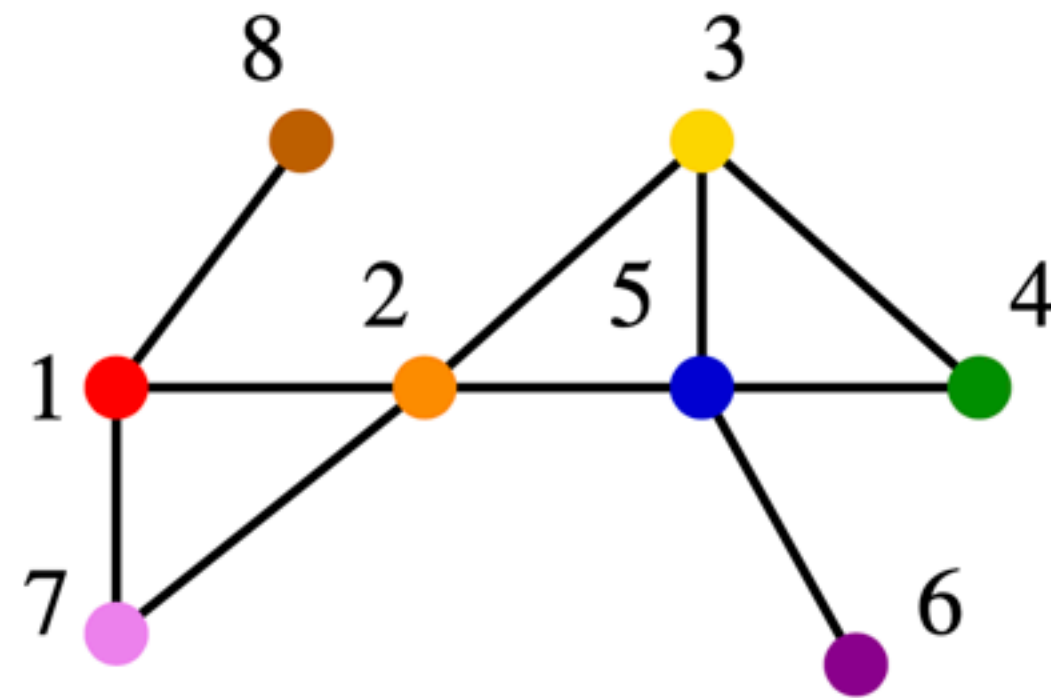
Mark vertex v as visited (via edge e).

Recursively call DFS(G, v).

$O(n + m)$ -time with adjacency list data structure



Depth-First Graph Traversal



Proposition 14.14: Let G be an undirected graph with n vertices and m edges. A DFS traversal of G can be performed in $O(n + m)$ time, and can be used to solve the following problems in $O(n + m)$ time:

- Computing a path between two given vertices of G , if one exists.
- Testing whether G is connected.
- Computing a spanning tree of G , if G is connected.
- Computing the connected components of G .
- Computing a cycle in G , or reporting that G has no cycles.

Proposition 14.15: Let \vec{G} be a directed graph with n vertices and m edges. A DFS traversal of \vec{G} can be performed in $O(n + m)$ time, and can be used to solve the following problems in $O(n + m)$ time:

- Computing a directed path between two given vertices of \vec{G} , if one exists.
- Computing the set of vertices of \vec{G} that are reachable from a given vertex s .
- Testing whether \vec{G} is strongly connected.
- Computing a directed cycle in \vec{G} , or reporting that \vec{G} is acyclic.
- Computing the **transitive closure** of \vec{G} (see Section 14.4).

DFS can also be used in

- Testing for connectivity
 - Undirected graphs: Start from a random vertex and see all vertices are reached or not
 - Directed graphs: Run DFS from a random point, reverse all directions of the graph and run DFS again. If all are visited on both cases, it is strongly connected.
- Computing all connected components: Run DFS from a random point. If unvisited vertices are left at the end, repeat until all are visited.
- Detecting cycles: Through the creation of the DFS tree...

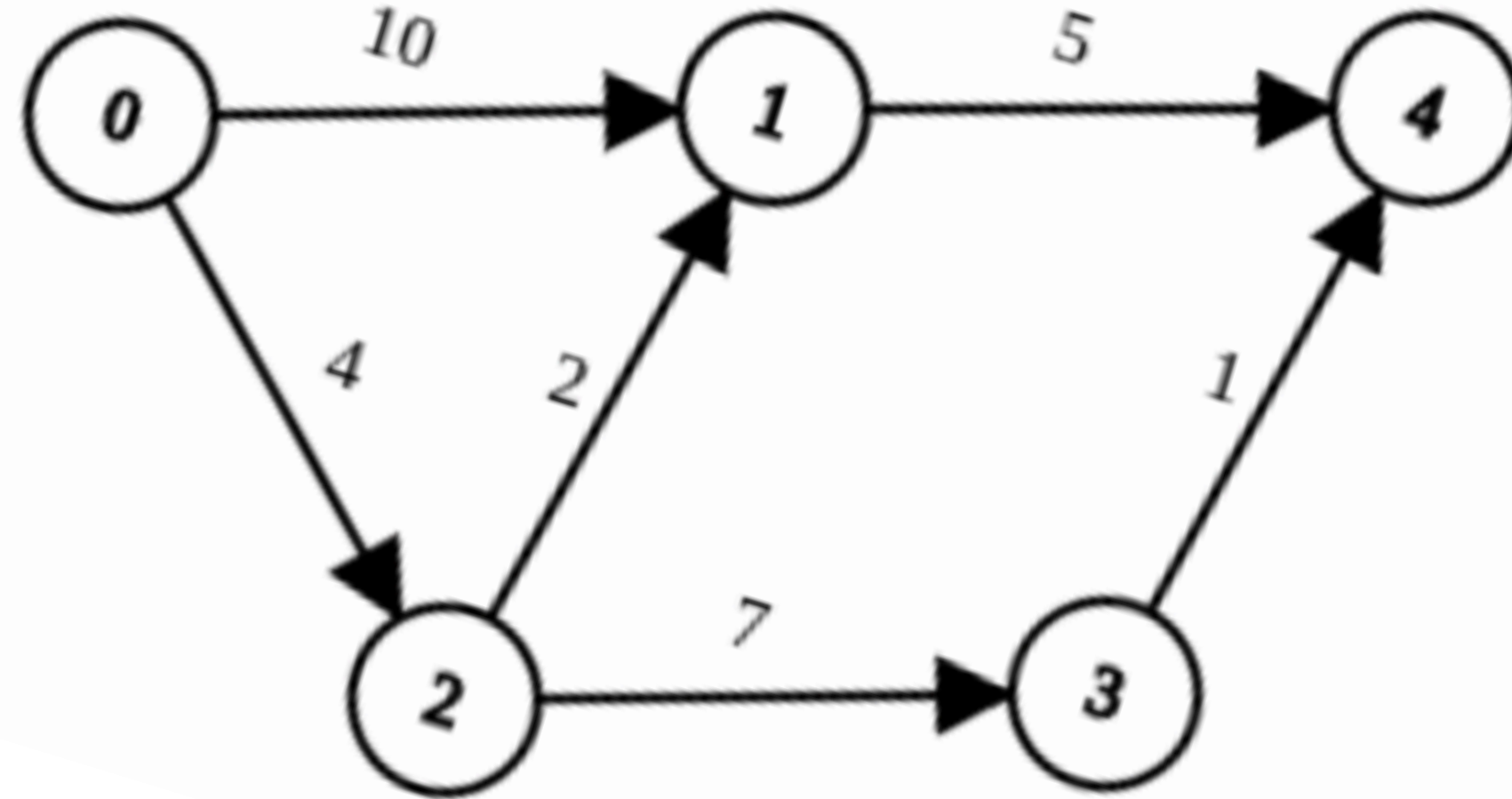
Reading assignment

- Skiena chapter 7.
- Goodrich et al. 14.1, 14.2, 14.3

$$D_0 = 0$$

$$D_1 = \infty$$

$$D_4 = \infty$$



$$D_2 = \infty$$

$$D_3 = \infty$$