



# Elementary Graph Algorithms

## Graphs



- Motivation and Terminology
- Representations
- Traversals

# Graphs

- A graph  $G = (V, E)$ 
  - $V$  = set of vertices
  - $E$  = set of edges = subset of  $V \times V$
  - Thus  $|E| = O(|V|^2)$

## Graph Variations

- Variations:
  - A *connected graph* has a path from every vertex to every other
  - In an *undirected graph*:
    - Edge  $(u,v)$  = edge  $(v,u)$
    - No self-loops
  - In a *directed graph*:
    - Edge  $(u,v)$  goes from vertex  $u$  to vertex  $v$ .
    - It is denoted  $u \rightarrow v$

## Graph Variations

- More variations:
  - A *weighted graph* associates weights with either the edges or the vertices
    - E.g., a road map: edges might be weighted as distance
  - A *multigraph* allows multiple edges between the same vertices
    - E.g., the call graph in a program (a function can get called from multiple points in another function)

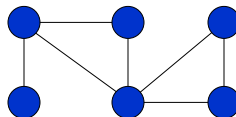
## Graphs

A graph  $G$  consists of a set of *vertices*  $V$  together with a set  $E$  of vertex pairs or *edges*.

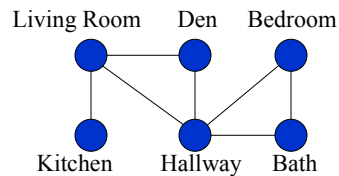
$G = (V, E)$  [in some texts they use  $G(V, E)$ ].

We also use  $V$  and  $E$  to represent # of nodes, edges

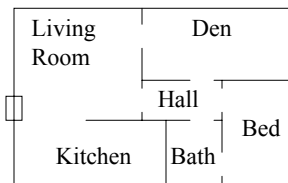
Graphs are important because any binary relation is a graph, so graphs can be used to represent essentially *any* relationship.



## Graph Interpretations



The vertices could represent rooms in a house, and the edges could indicate which of those rooms are connected to each other.



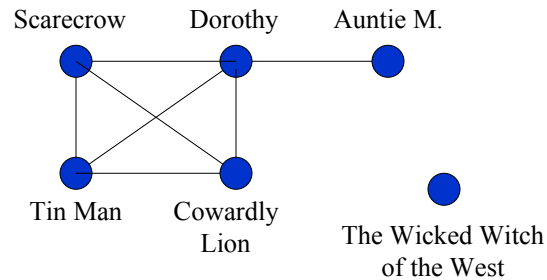
Apartment Blueprint

Sometimes using a graph will be an easy simplification for a problem.

## More interpretations

- Vertices are cities and edges are the roads connecting them.
- Edges are the components in a circuit and vertices are junctions where they connect.
- Vertices are software packages and edges indicate those that can interact.
- Edges are phone conversations and vertices are the households being connected.

## Friendship Graphs



Each vertex represents a person, and each edge indicates that the two people are friends.

## Graph Terminology

### Directed and undirected graphs

A graph is said to be **undirected** if edge  $(x, y)$  always implies  $(y, x)$ . Otherwise it is said to be **directed**. Often called an **arc**.

### Loops, multiedges, and simple graphs

An edge of the form  $(x, x)$  is said to be a **loop**. If  $x$  was  $y$ 's friend several times over, we can model this relationship using **multiedges**. A graph is said to be **simple** if it contains no loops or multiedges.

### Weighted edges

A graph is said to be **weighted** if each edge has an associated numerical attribute. In an **unweighted** graph, all edges are assumed to be of equal weight.

## Graph Terminology - contd.

### Paths

A **path** is a any sequence of edges that connect two vertices. A **simple path** never goes through any vertex more than once. The **shortest path** is the minimum number edges needed to connect two vertices.

### Connectivity

The “six degrees of separation” theory argues that there is always a short path between any two people in the world. A graph is **connected** if there is there is a path between any two vertices. A directed graph is **strongly connected** if there is always a directed path between vertices. Any subgraph that is connected can be referred to as a **connected component**.

## Graph Terminology - contd.

### Degree and graph types

The **degree** of a vertex is the number of edges connected to it. The most popular person will have a vertex of the highest degree. Remote hermits may have degree-zero vertices. In **dense** graphs, most vertices have high degree. In **sparse** graphs, most vertices have low degree. In a **regular graph**, all vertices have exactly the same degree.

### Clique

A graph is called **complete** if every pair of vertices is connected by an edge. A **clique** is a sub-graph that is complete.

## Graph Terminology - contd.

### Cycles and Dags

A **cycle** is a path where the last vertex is adjacent to the first. A cycle in which no vertex is repeated is said to be a **simple cycle**. The shortest cycle in a graph determines the graph's **girth**. A simple cycle that passes through every vertex is said to be a **Hamiltonian cycle**. An undirected graph with no cycles is a **tree** if it is connected, or a **forest** if it is not. A directed graph with no directed cycles is said to be a **directed acyclic graph** (or a DAG)

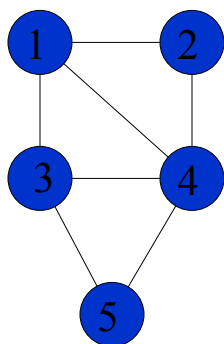
## Graphs

- We will typically express running times in terms of  $|E|$  and  $|V|$  (often dropping the  $|$ 's)
  - If  $|E| \approx |V|^2$  the graph is **dense**
  - If  $|E| \approx |V|$  the graph is **sparse**
- If you know you are dealing with dense or sparse graphs, different data structures may make sense

# Graphs

- Motivation and Terminology
- Representations
- Traversals

## Adjacency Matrix



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

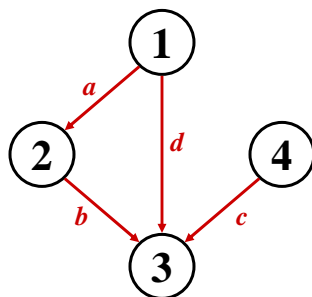


## Representing Graphs

- Assume  $V = \{1, 2, \dots, n\}$
- An *adjacency matrix* represents the graph as a  $n \times n$  matrix  $A$ :
  - $A[i, j] = 1$  if edge  $(i, j) \in E$   
 $= 0$  if edge  $(i, j) \notin E$

## Graphs: Adjacency Matrix

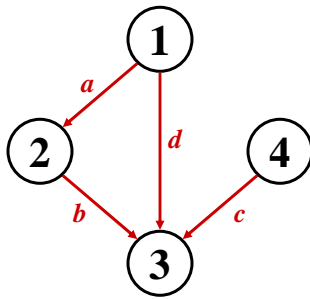
- Example:



A	1	2	3	4
1				
2				
3			??	
4				

## Graphs: Adjacency Matrix

- Example:



A	1	2	3	4
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## Graphs: Adjacency Matrix

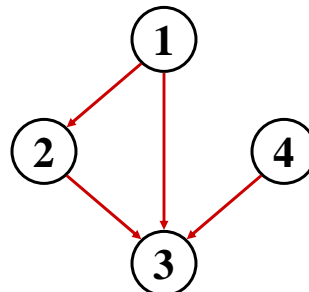
- *How much storage does the adjacency matrix require?*
- A:  $O(V^2)$
- *What is the minimum amount of storage needed by an adjacency matrix representation of an undirected graph with 4 vertices?*
- A: ? bits
  - Undirected graph → matrix is symmetric
  - No self-loops → don't need diagonal

## Graphs: Adjacency Matrix

- The adjacency matrix is a dense representation
  - Usually too much storage for large graphs
  - But can be very efficient for small graphs
- Most large interesting graphs are sparse
  - E.g., planar graphs, in which no edges cross, have  $|E| = O(|V|)$  by Euler's formula
  - For this reason the *adjacency list* is often a more appropriate representation

## Graphs: Adjacency List

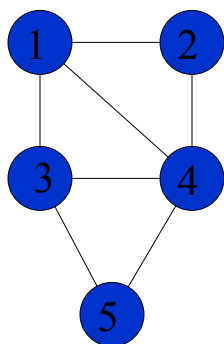
- Adjacency list: for each vertex  $v \in V$ , store a list of vertices adjacent to  $v$
- Example:
  - $\text{Adj}[1] = \{2,3\}$
  - $\text{Adj}[2] = \{3\}$
  - $\text{Adj}[3] = \{\}$
  - $\text{Adj}[4] = \{3\}$
- Variation: can also keep a list of edges coming *into* vertex



## Graphs: Adjacency List

- How much storage is required?
  - The *degree* of a vertex  $v$  = # incident edges
    - Directed graphs have in-degree, out-degree
  - For directed graphs, # of items in adjacency lists is  $\sum \text{out-degree}(v) = |E|$   
takes  $\Theta(V + E)$  storage (*Why?*)
  - For undirected graphs, # items in adj lists is  $\sum \text{degree}(v) = 2|E|$  (*handshaking lemma*)  
also  $\Theta(V + E)$  storage
- So: Adjacency lists take  $O(V+E)$  storage

## Adjacency List



1	→ 2 → 3 → 4
2	→ 1 → 4
3	→ 1 → 4 → 5
4	→ 1 → 2 → 3 → 5
5	→ 3 → 4

## Tradeoffs Between Adjacency Lists and Adjacency Matrices

### Comparison

Faster to test if  $(x, y)$  exists?

Faster to find vertex degree?

Less memory on sparse graphs?

Less memory on dense graphs?

Edge insertion or deletion?

Faster to traverse the graph?

Better for most problems?

### Winner (for worst case)

matrices:  $\Theta(1)$  vs.  $\Theta(V)$

lists:  $\Theta(1)$  vs.  $\Theta(V)$

lists:  $\Theta(V+E)$  vs.  $\Theta(V^2)$

matrices: (small win)

matrices:  $\Theta(1)$  vs.  $\Theta(V)$

lists:  $\Theta(E+V)$  vs.  $\Theta(V^2)$

**lists**