



# CS201 DISCRETE MATHEMATICS FOR COMPUTER SCIENCE

Dr. QI WANG

Department of Computer Science and Engineering

Office: Room413, CoE South Tower

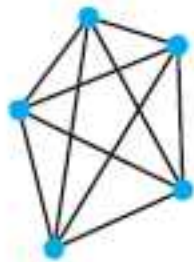
Email: [wangqi@sustech.edu.cn](mailto:wangqi@sustech.edu.cn)

# Definition of a Graph

- **Definition.** A *graph*  $G = (V, E)$  consists of a nonempty set  $V$  of *vertices* (or *nodes*) and a set  $E$  of *edges*. Each edge has either one or two vertices associated with it, called its *endpoints*. An edge is said to be *incident to* (or *connect* its endpoints).



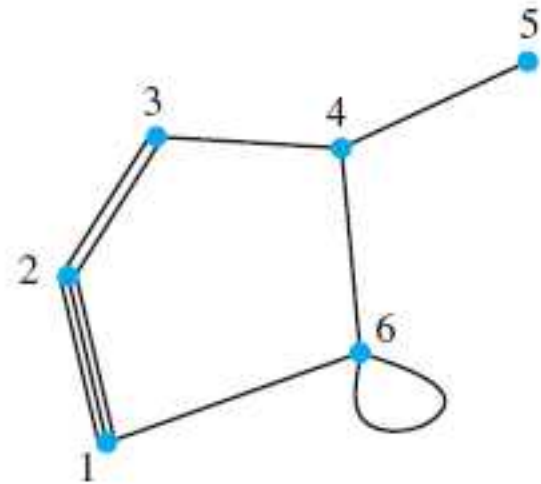
a



b



c



d

# Complete Graphs

- A *complete graph* on  $n$  vertices, denoted by  $K_n$ , is the simple graph that contains exactly one edge between **each pair** of distinct vertices.

$K_1$

$K_2$

$K_3$

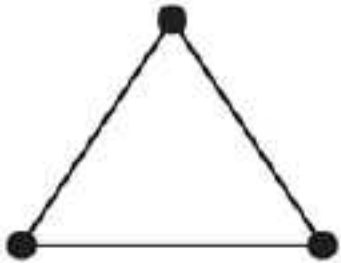
$K_4$

$K_5$

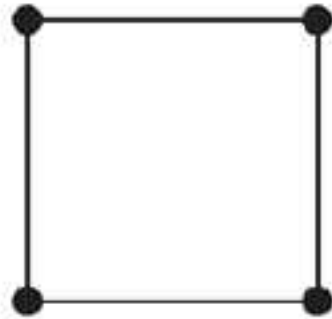
$K_6$

# Cycles

- A *cycle*  $C_n$  for  $n \geq 3$  consists of  $n$  vertices  $v_1, v_2, \dots, v_n$ , and edges  $\{v_1, v_2\}, \{v_2, v_3\}, \dots, \{v_{n-1}, v_n\}, \{v_n, v_1\}$ .



$C_3$



$C_4$



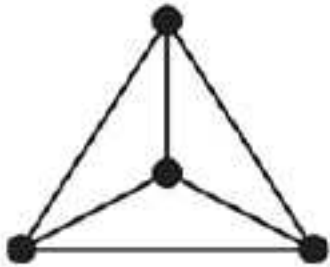
$C_5$



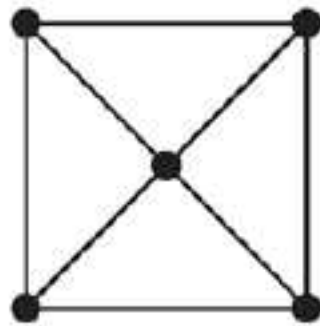
$C_6$

# Wheels

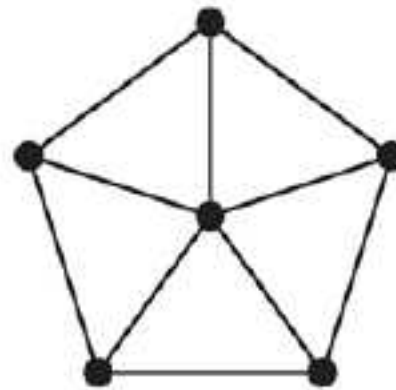
- A *wheel*  $W_n$  is obtained by adding an additional vertex to a cycle  $C_n$ .



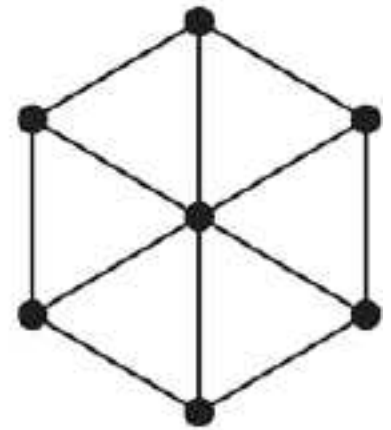
$W_3$



$W_4$



$W_5$



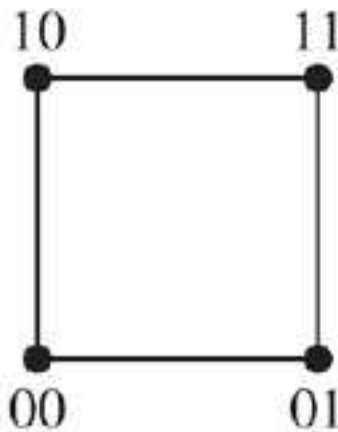
$W_6$

# $N$ -dimensional Hypercube

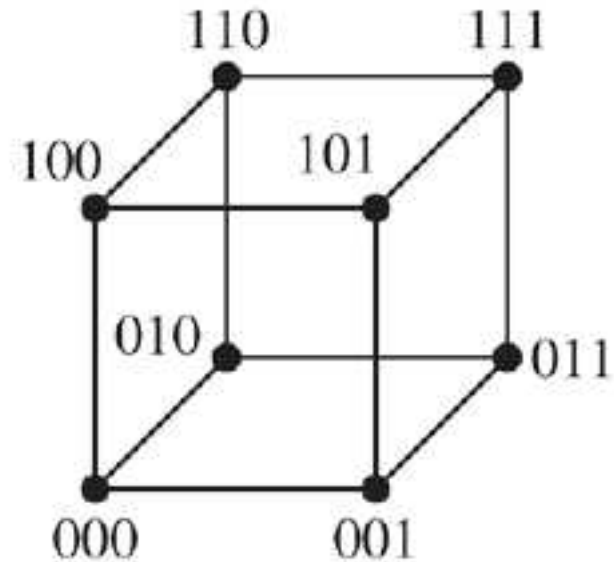
- An  $n$ -dimensional hypercube, or  $n$ -cube,  $Q_n$  is a graph with  $2^n$  vertices representing all bit strings of length  $n$ , where there is an edge between two vertices that differ in exactly one bit position.



$Q_1$



$Q_2$



$Q_3$

How many vertices? How many edges?



# Bipartite Graphs

- **Definition** A simple graph  $G$  is *bipartite* if  $V$  can be partitioned into two disjoint subsets  $V_1$  and  $V_2$  such that every edge connects a vertex in  $V_1$  and a vertex in  $V_2$ .



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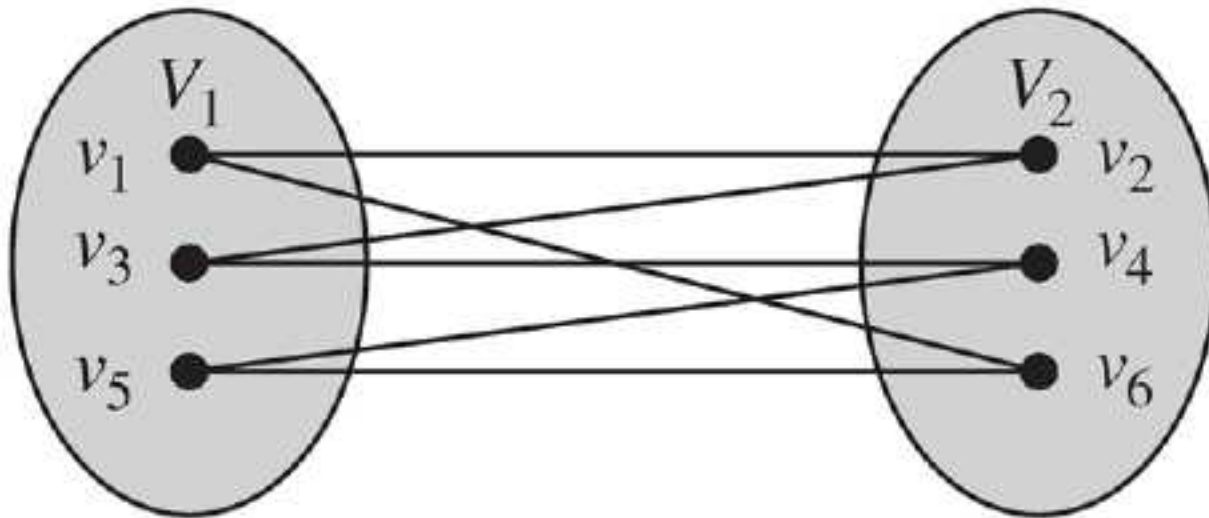




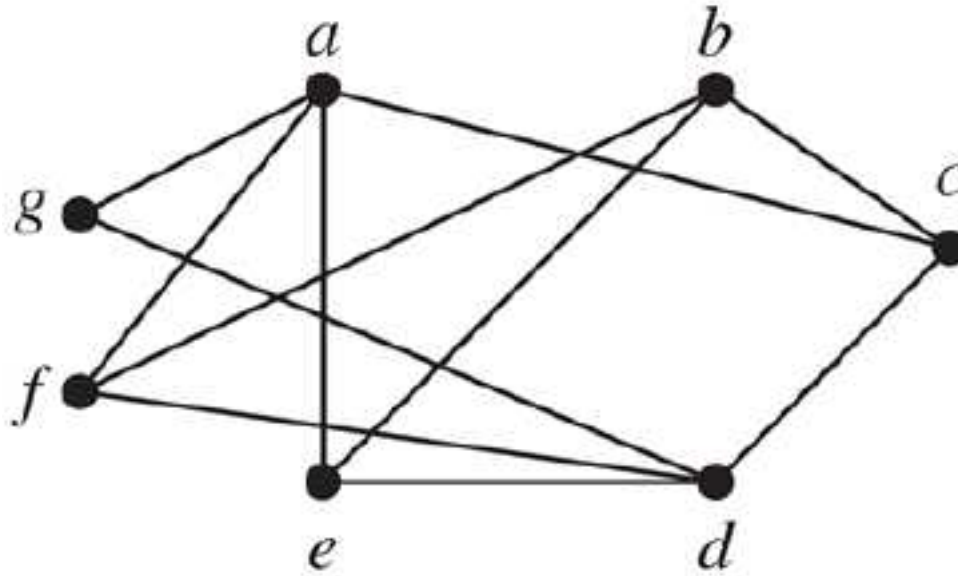
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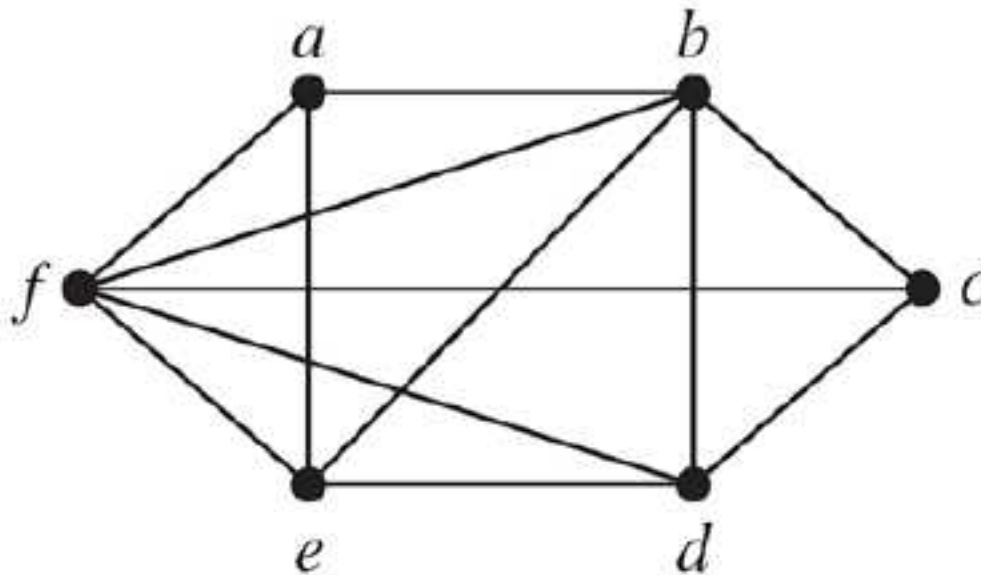
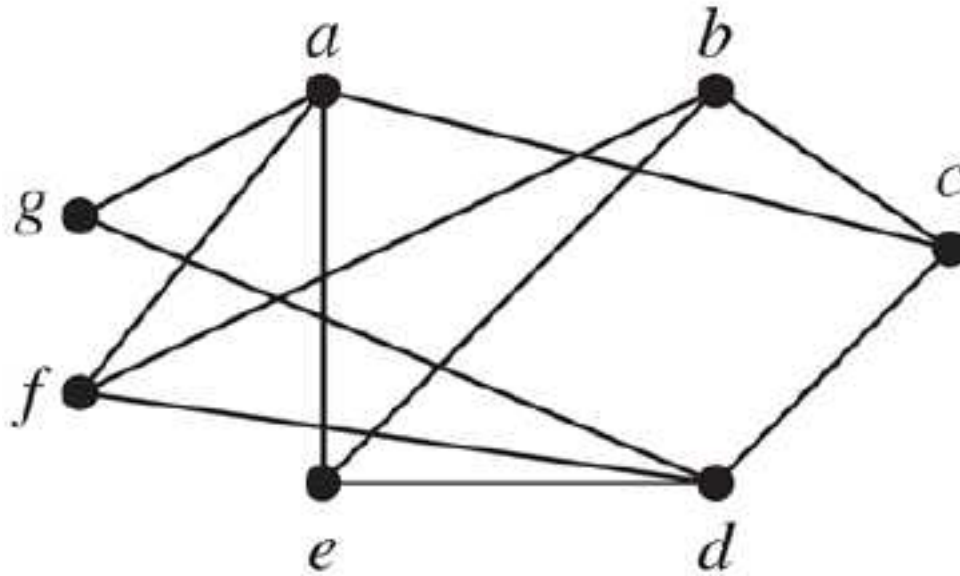
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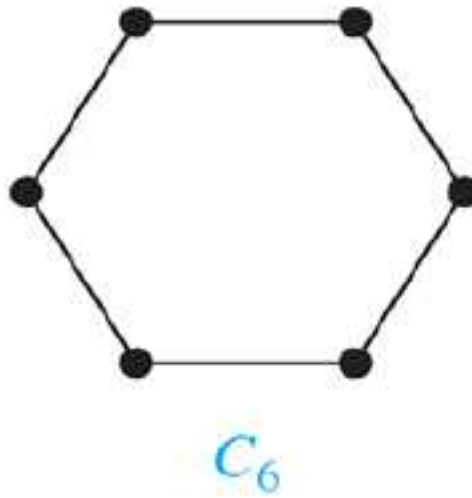


# Bipartite Graphs



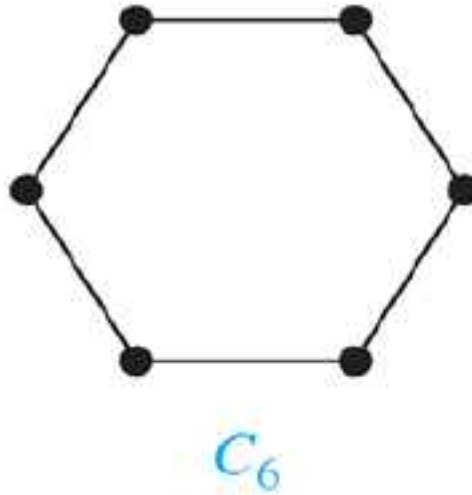
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- **Example** Show that  $C_6$  is bipartite.

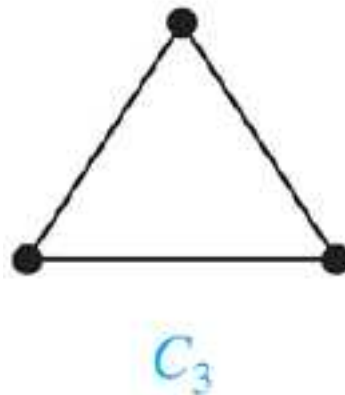


# Bipartite Graphs

- **Example** Show that  $C_6$  is bipartite.



**Example** Show that  $C_3$  is not bipartite.



# Complete Bipartite Graphs

- **Definition** A *complete bipartite graph*  $K_{m,n}$  is a graph that has its vertex set partitioned into two subsets  $V_1$  of size  $m$  and  $V_2$  of size  $n$  such that there is an edge from every vertex in  $V_1$  to every vertex in  $V_2$ .

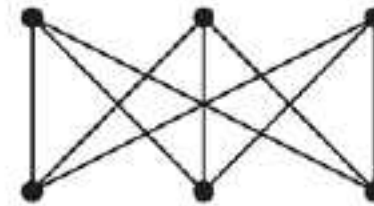


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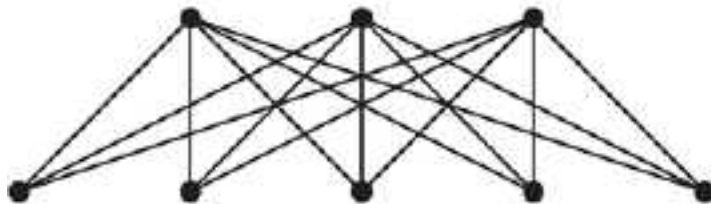
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$K_{2,3}$



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$K_{3,5}$

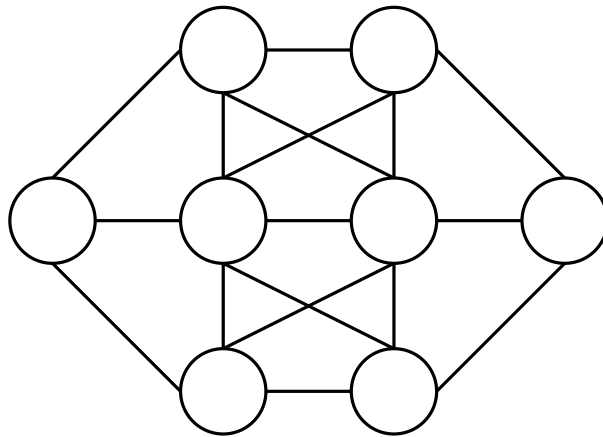


$K_{2,6}$



# Puzzles using Graphs

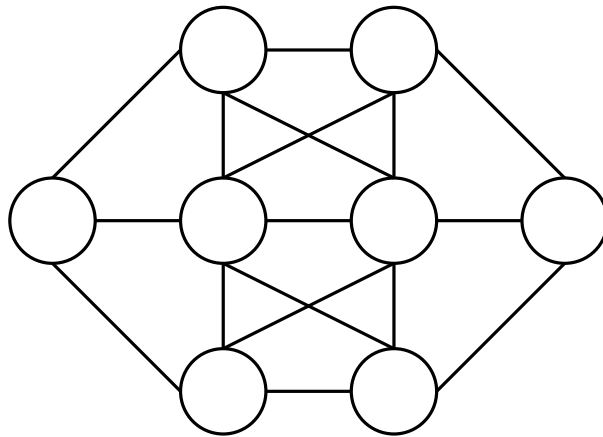
- **The eight-circles problem** Place the letters A, B, C, D, E, F, G, H into the eight circles in the figure, in such a way that **no** letter is adjacent to a letter that is next to it in the alphabet.





# Puzzles using Graphs

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- **Six people at a party** Show that, in any gathering of six people, there are either three people who all know each other, or three people none of which knows either of the other two.

# Bipartite Graphs and Matchings

- *Matching* the elements of one set to elements in another. A *matching* is a subset of  $E$  s.t. **no two edges are incident with the same vertex.**



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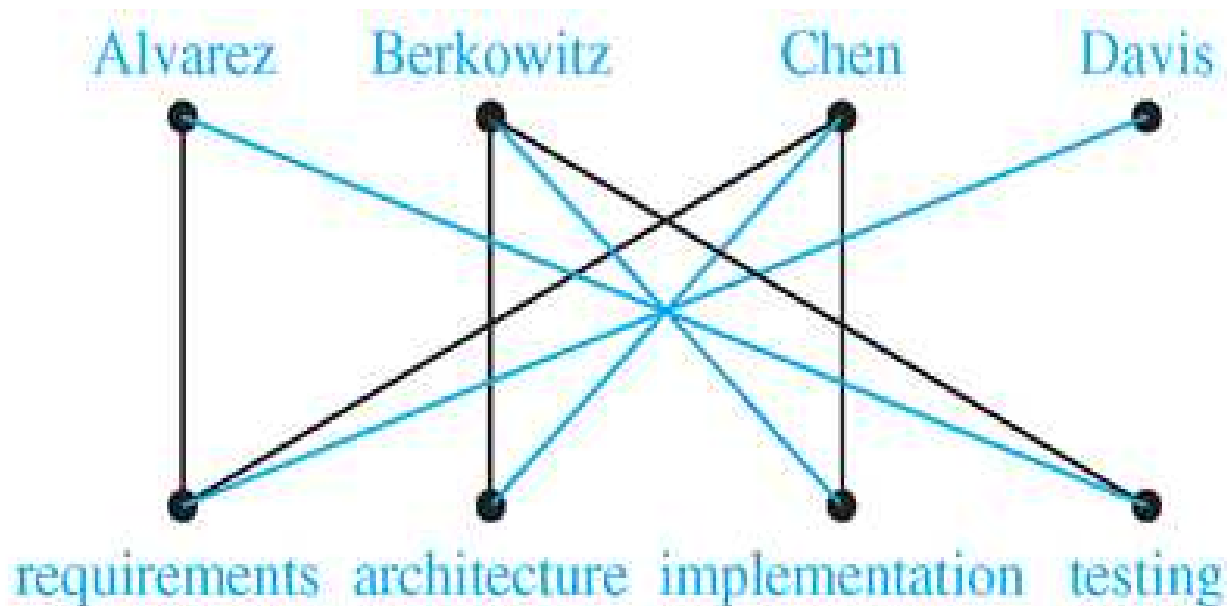
*Job assignments*: vertices represent the jobs and the employees, edges link employees with those jobs they have been trained to do. A **common goal** is to **match jobs to employees** so that the most jobs are done.



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A matching  $M$  in a bipartite graph  $G = (V, E)$  with bipartition  $(V_1, V_2)$  is a *complete matching from  $V_1$  to  $V_2$*  if **every vertex in  $V_1$  is the endpoint of an edge in the matching**, or equivalently, if  $|M| = |V_1|$ .





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**Theorem** (Hall's Marriage Theorem) The bipartite graph  $G = (V, E)$  with bipartition  $(V_1, V_2)$  has a complete matching from  $V_1$  to  $V_2$  if and only if  $|N(A)| \geq |A|$  for all subsets  $A$  of  $V_1$ .



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Hence,  $|N(A)| \geq |A|$ .



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Case (i): For all integers  $j$  with  $1 \leq j \leq k$ , the vertices in every set of  $j$  elements from  $W_1$  are adjacent to at least  $j + 1$  elements of  $W_2$

Case (ii): For some integer  $j$  with  $1 \leq j \leq k$ , there is a subset  $W'_1$  of  $j$  vertices such that there are exactly  $j$  neighbors of these vertices in  $W_2$



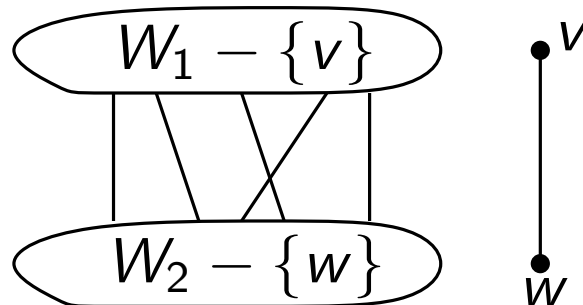
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If not, there is a subset  $B$  of  $t$  vertices with  $1 \leq t \leq k + 1 - j$  s.t.  $|N(B)| < t$ .



# Subgraphs

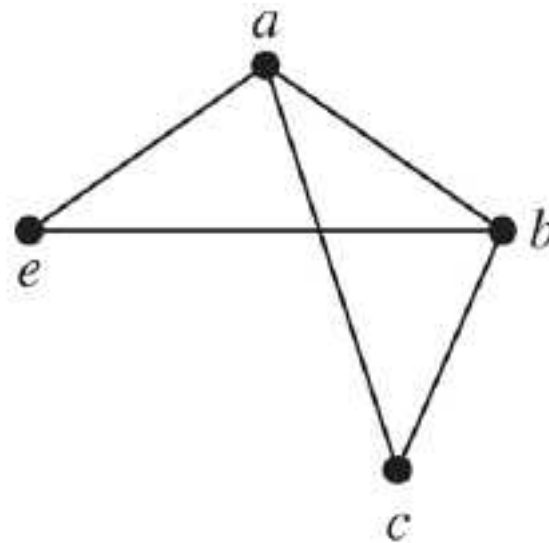
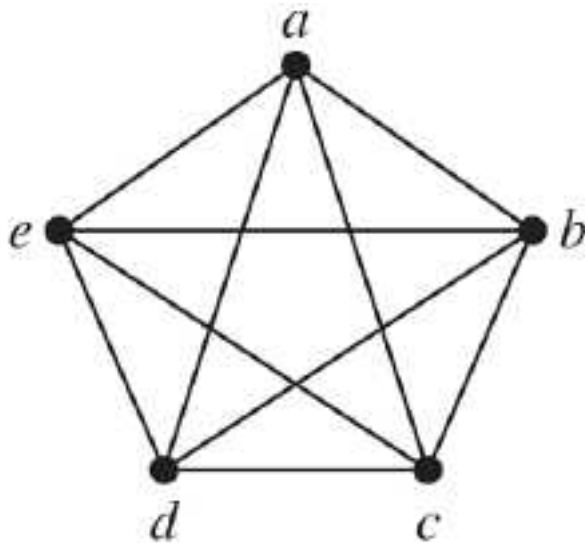
- **Definition** A *subgraph of a graph*  $G = (V, E)$  is a graph  $(W, F)$ , where  $W \subseteq V$  and  $F \subseteq E$ . A subgraph  $H$  of  $G$  is a *proper subgraph* of  $G$  if  $H \neq G$ .





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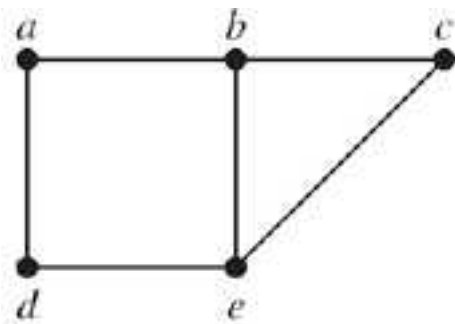
# Union of Graphs

- **Definition** The *union of two simple graphs*  $G_1 = (V_1, E_1)$  and  $G_2 = (V_2, E_2)$  is the simple graph with vertex set  $V_1 \cup V_2$  and edge set  $E_1 \cup E_2$ , denoted by  $G_1 \cup G_2$ .

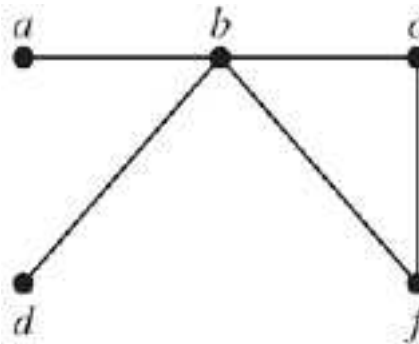


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- **Definition** The *union of two simple graphs*  $G_1 = (V_1, E_1)$  and  $G_2 = (V_2, E_2)$  is the simple graph with vertex set  $V_1 \cup V_2$  and edge set  $E_1 \cup E_2$ , denoted by  $G_1 \cup G_2$ .



$G_1$

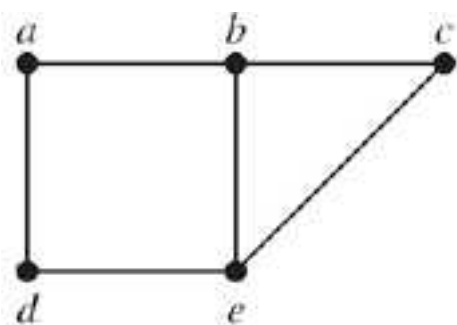


$G_2$

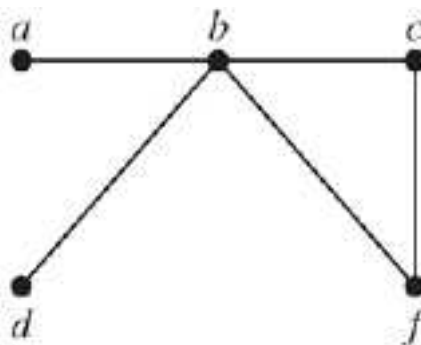


# Union of Graphs

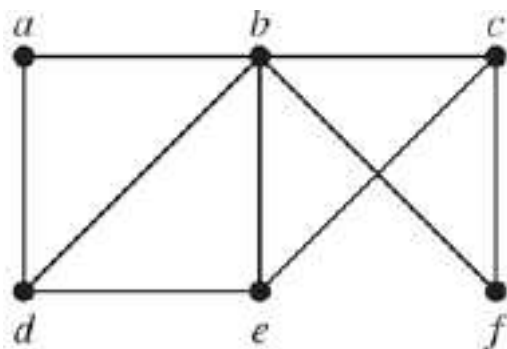
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$G_2$



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# Representation of Graphs

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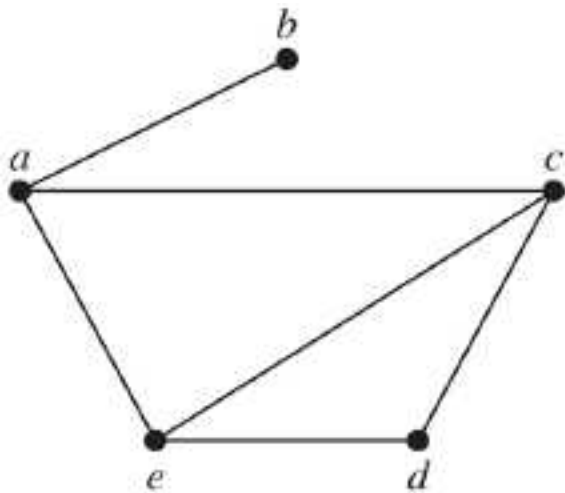
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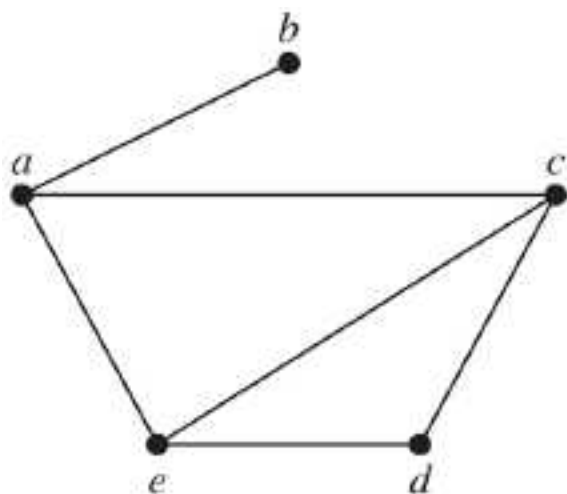
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**TABLE 1** An Adjacency List for a Simple Graph.

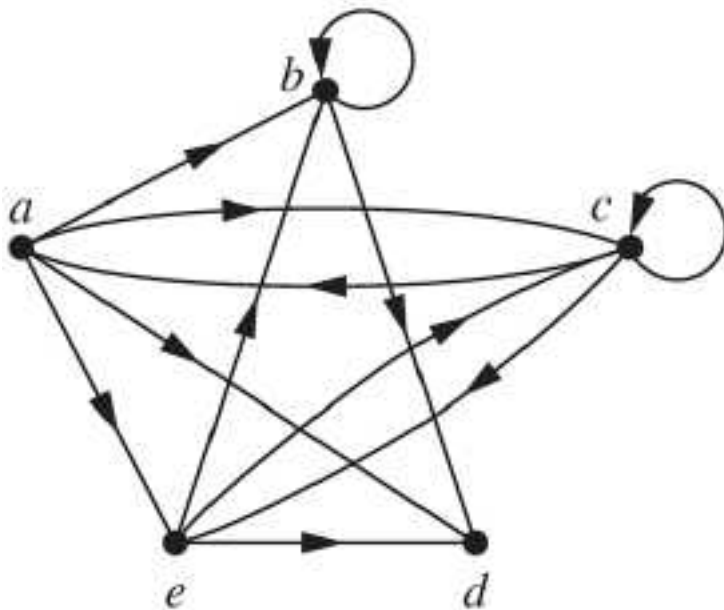
Vertex	Adjacent Vertices
a	b, c, e
b	a
c	a, d, e
d	c, e
e	a, c, d





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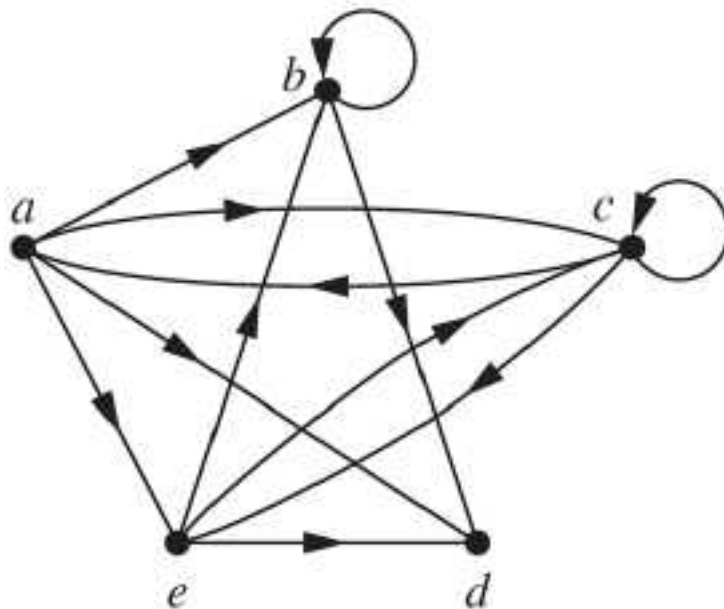


TABLE 2 An Adjacency List for a Directed Graph.	
Initial Vertex	Terminal Vertices
a	b, c, d, e
b	b, d
c	a, c, e
d	
e	b, c, d



# Adjacency Matrices

- **Definition** Suppose that  $G = (V, E)$  is a simple graph with  $|V| = n$ . Arbitrarily list the vertices of  $G$  as  $v_1, v_2, \dots, v_n$ . The *adjacency matrix*  $\mathbf{A}_G$  of  $G$ , is the  $n \times n$  zero-one matrix with 1 as its  $(i, j)$ -th entry when  $v_i$  and  $v_j$  are adjacent, and 0 as its  $(i, j)$ -th entry when they are not adjacent.



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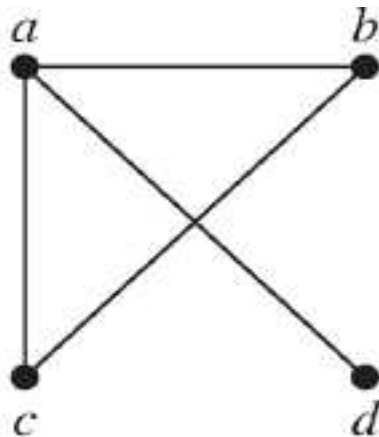


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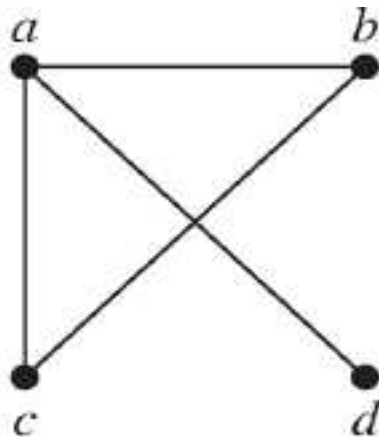


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$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}$$



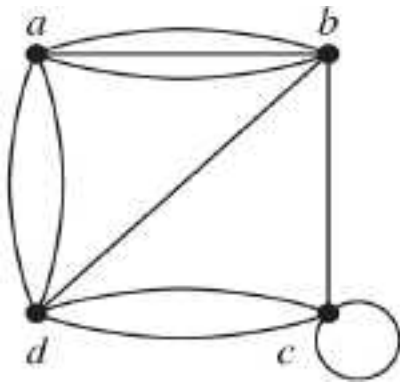
# Adjacency Matrices

- Adjacency matrices can also be used to represent graphs with loops and multiple edges. The matrix is no longer a zero-one matrix.



# Adjacency Matrices

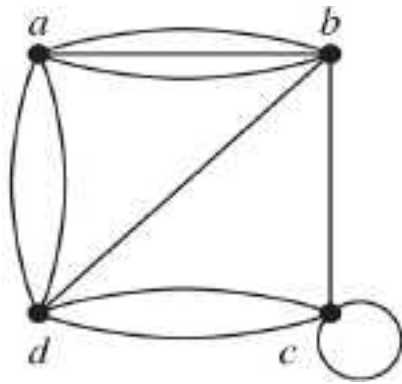
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$$\begin{bmatrix} 0 & 3 & 0 & 2 \\ 3 & 0 & 1 & 1 \\ 0 & 1 & 1 & 2 \\ 2 & 1 & 2 & 0 \end{bmatrix}$$



# Incidence Matrices

- **Definition** Let  $G = (V, E)$  be an undirected graph with vertices  $v_1, v_2, \dots, v_n$  and edges  $e_1, e_2, \dots, e_m$ . The *incidence matrix* with respect to the ordering of  $V$  and  $E$  is the  $n \times m$  matrix  $\mathbf{M} = [m_{ij}]_{n \times m}$ , where

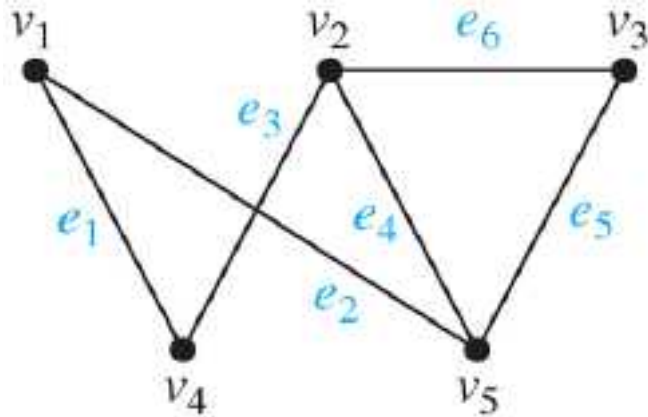
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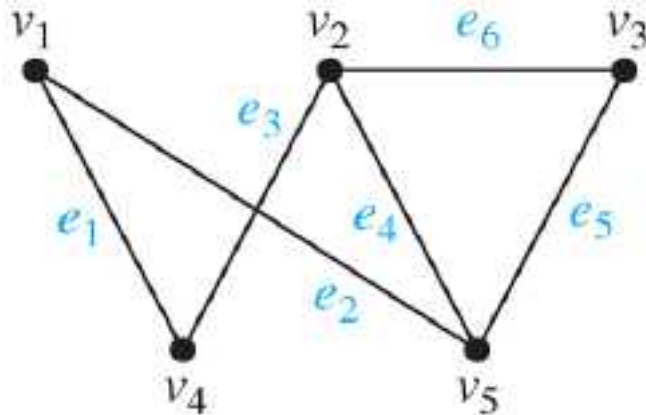
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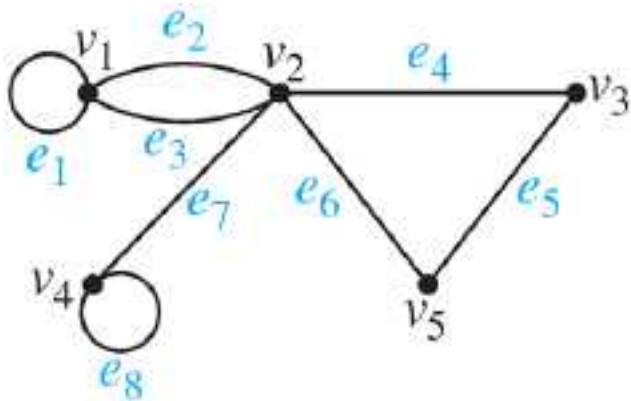
$$\begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 & 0 \end{bmatrix}$$



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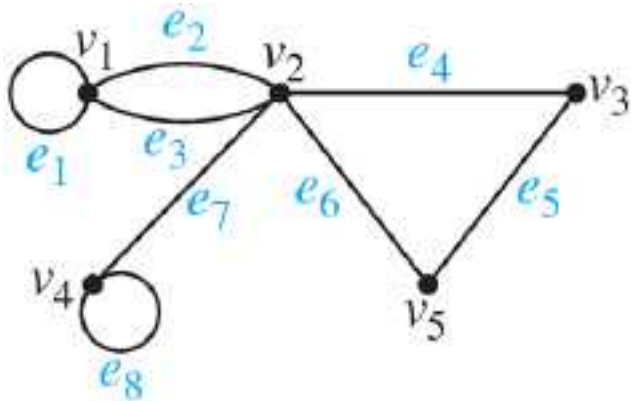
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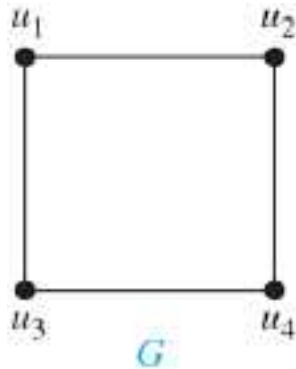
# Isomorphism of Graphs

- **Definition** The simple graphs  $G_1 = (V_1, E_1)$  and  $G_2 = (V_2, E_2)$  are *isomorphic* if there is a *one-to-one* and *onto* function from  $V_1$  to  $V_2$  with the property that  *$a$  and  $b$  are adjacent in  $G_1$  if and only if  $f(a)$  and  $f(b)$  are adjacent in  $G_2$* , for all  $a$  and  $b$  in  $V_1$ . Such a function is called an *isomorphism*.

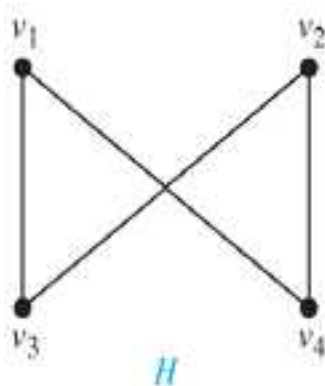


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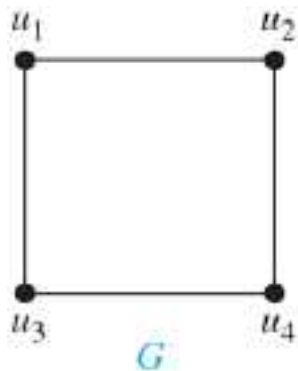
Are the two graphs **isomorphic**?





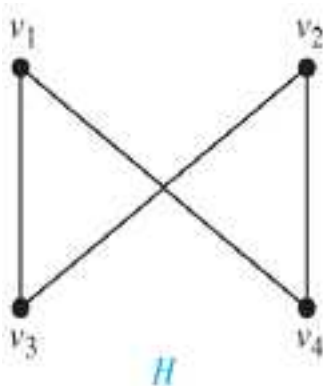
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Are the two graphs **isomorphic**?

Define a **one-to-one correspondence**:  
 $f(u_1) = v_1, f(u_2) = v_4, f(u_3) = v_3,$  and  
 $f(u_4) = v_2$



# Isomorphism of Graphs

- It is usually **difficult** to determine whether two simple graphs are isomorphic **using brute force** since there are  $n!$  possible **one-to-one correspondences**.



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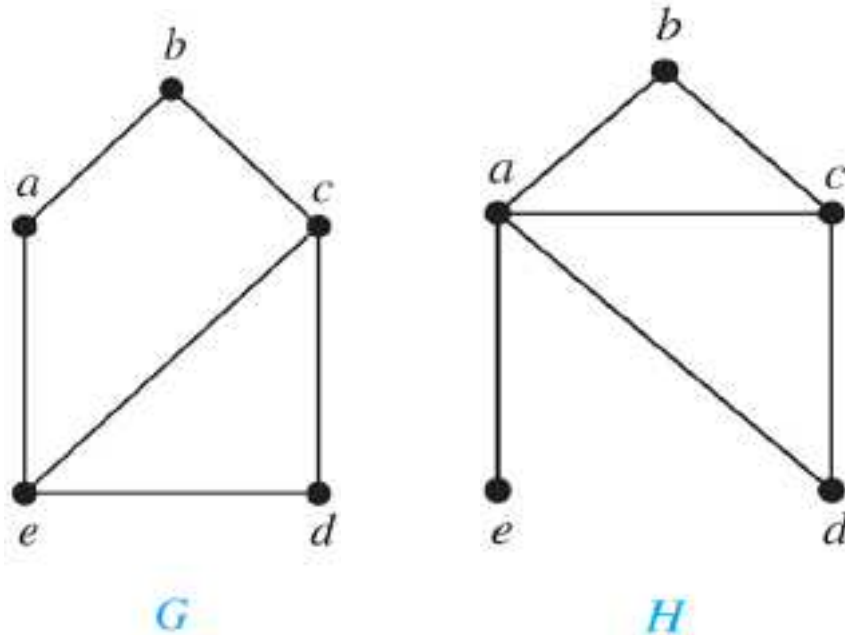
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- Useful **graph invariants** include **the number of vertices**, **number of edges**, **degree sequence**, etc.



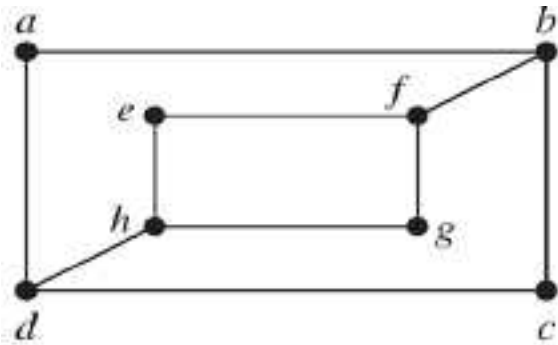
# Isomorphism of Graphs

- **Example** Determine whether these two graphs are **isomorphic**.

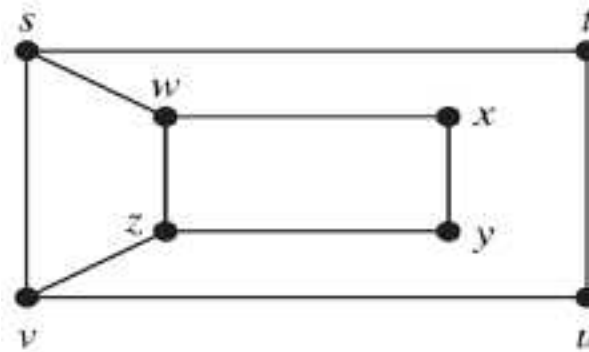


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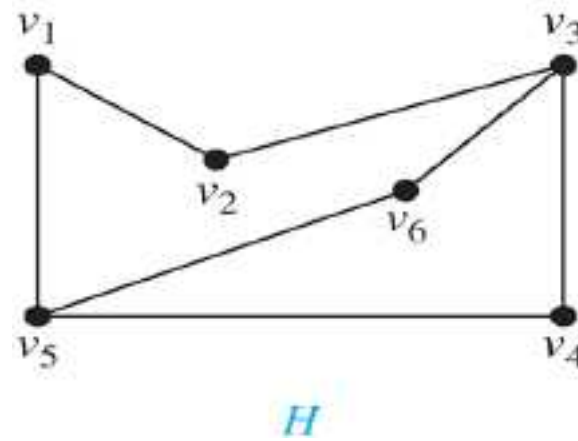
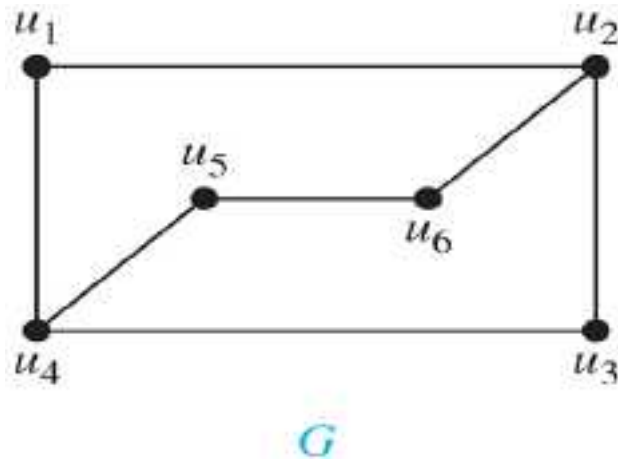
$G$



$H$

# Isomorphism of Graphs

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# Path

- **Definition** Let  $n$  be a nonnegative integer and  $G$  an undirected graph. A *path of length  $n$*  from  $u$  to  $v$  in  $G$  is a sequence of  *$n$  edges*  $e_1, e_2, \dots, e_n$  of  $G$  for which there exists a sequence  $x_0 = u, x_1, \dots, x_{n-1}, x_n = v$  of vertices such that  $e_i$  has the endpoints  $x_{i-1}$  and  $x_i$  for  $i = 1, \dots, n$ . The path is a *circuit* if it *begins and ends at the same vertex*, i.e., if  $u = v$  and has length greater than zero. A path or circuit is *simple* if it *does not* contain *repeating vertices*.





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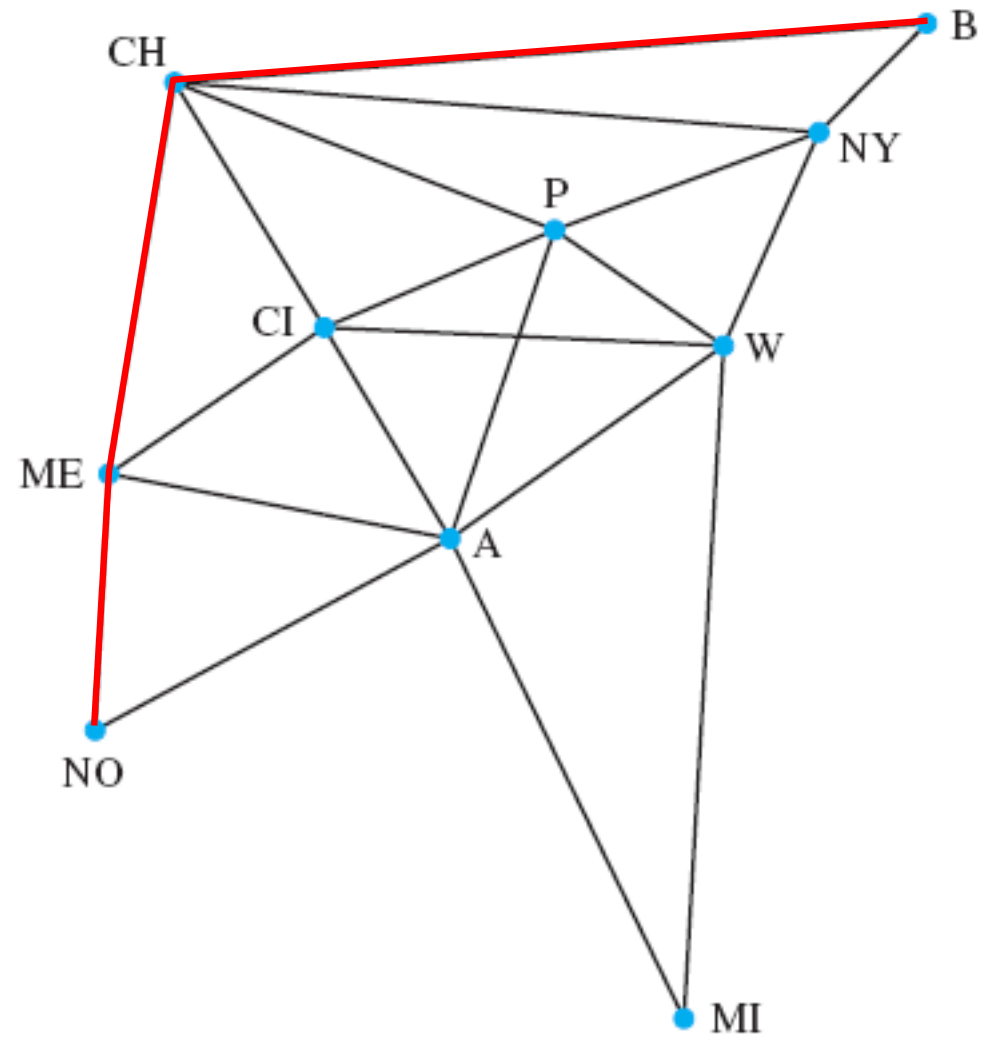
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Length of a path = # of edges on path

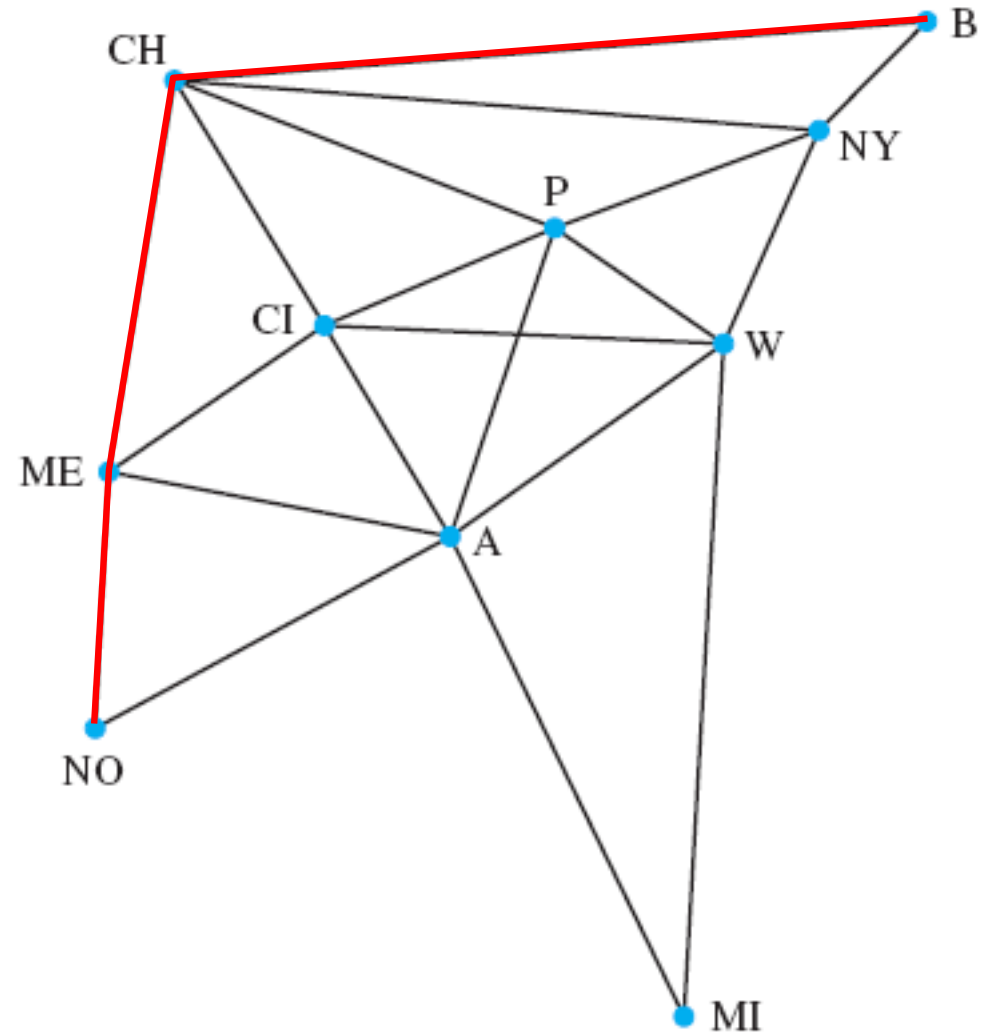


# Path



# Path

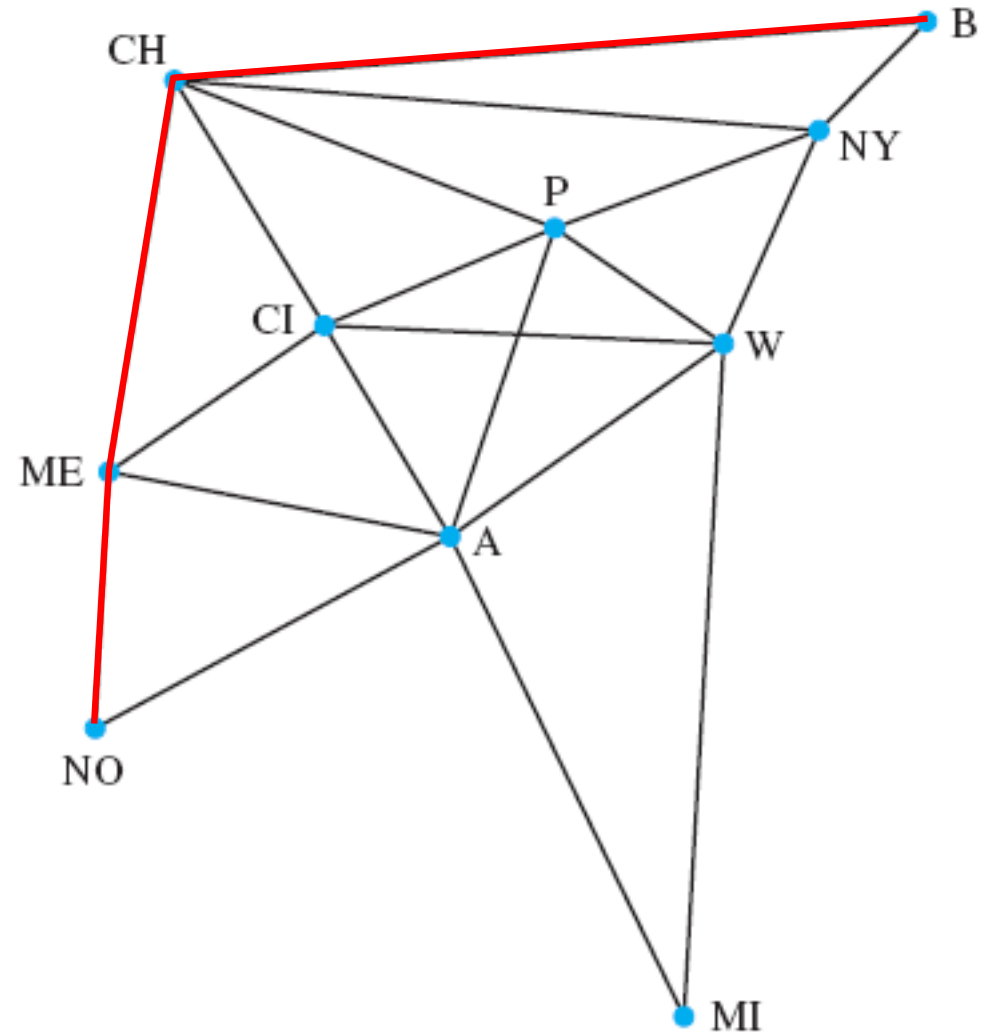
Path from Boston to New Orleans is B, CH, ME, NO



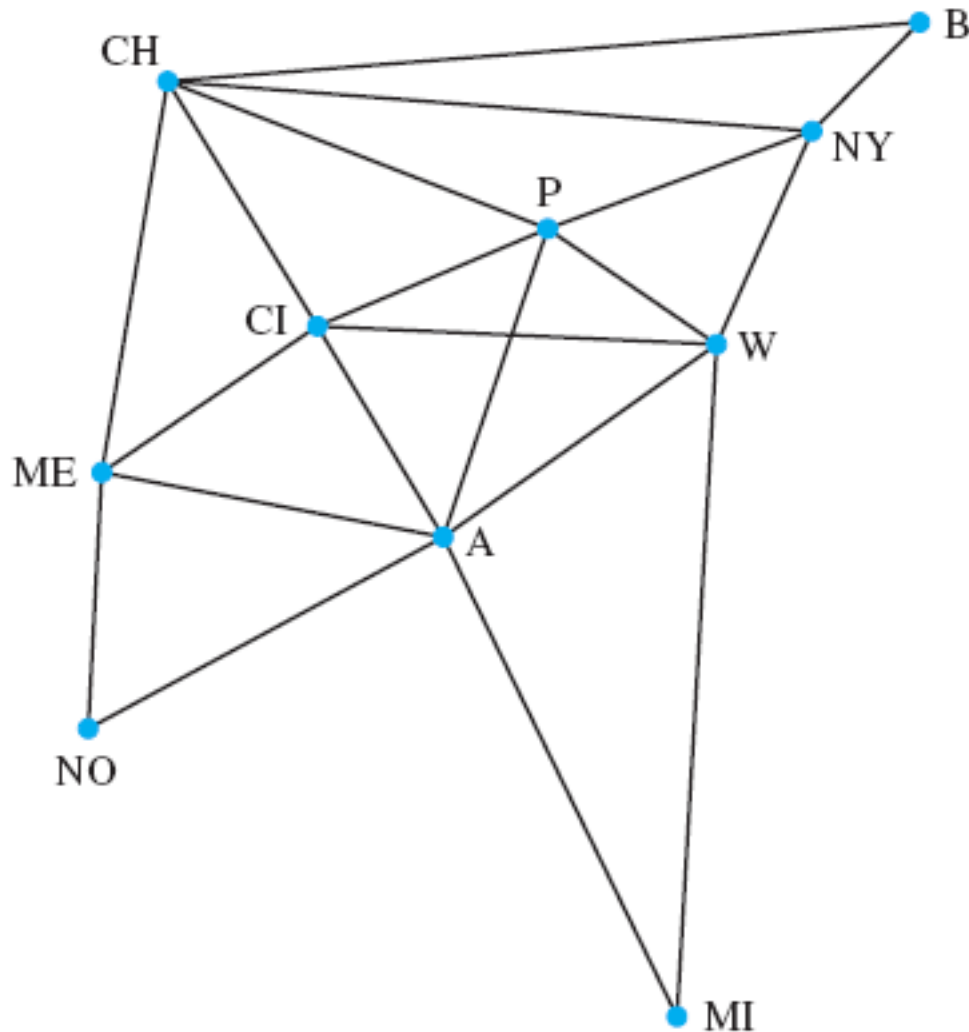
# Path

Path from Boston to New Orleans is B, CH, ME, NO

This path has length 3.



# Connectivity



Company decides to lease only **minimum number** of communication lines it needs to be able to send a message from any city to any other city by using any number of intermediate cities.

What is the **minimum** number of lines it needs to lease?



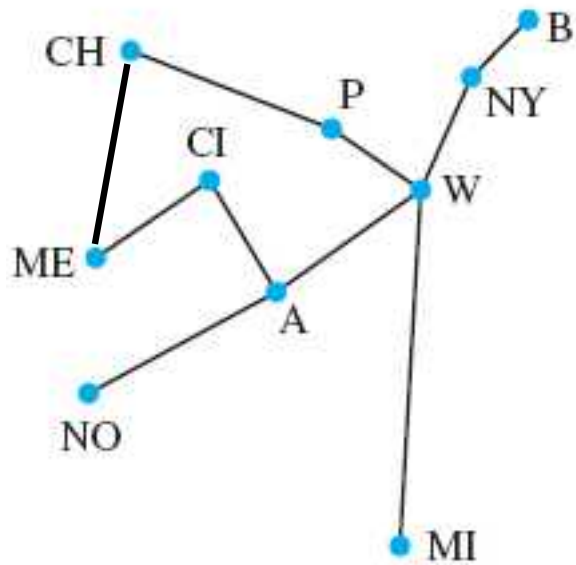
# Connectivity

- Choosing 10 edges?



# Connectivity

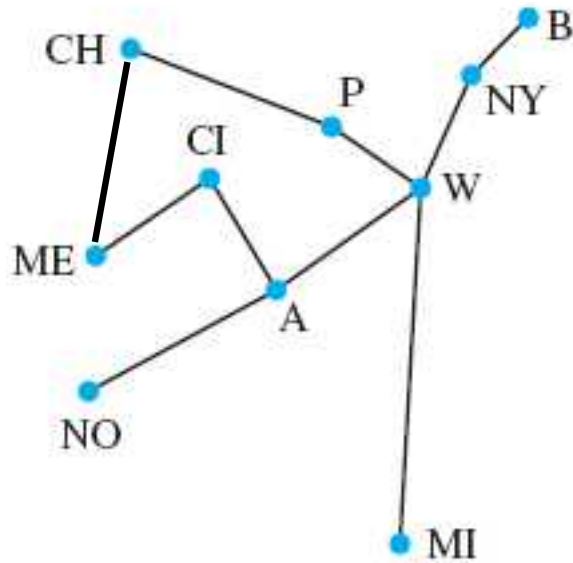
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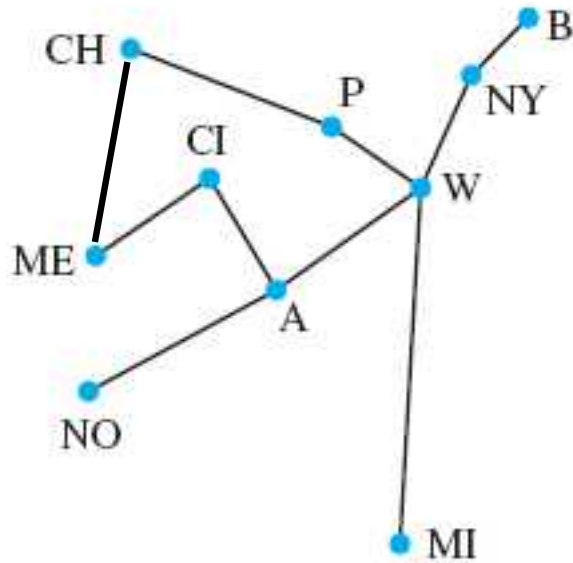
Too many.

Could throw away edge **CI**, **A**, and still have a solution.



# Connectivity

- Choosing 10 edges?



Too many.

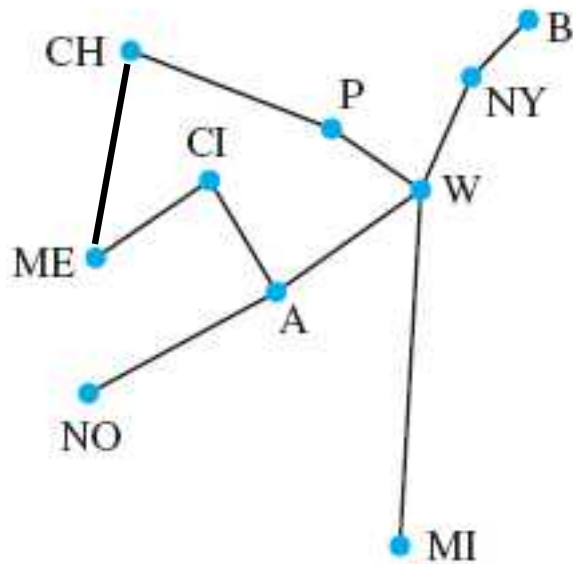
Could throw away edge **CI**, **A**, and still have a solution.

Choosing 8 edges?



# Connectivity

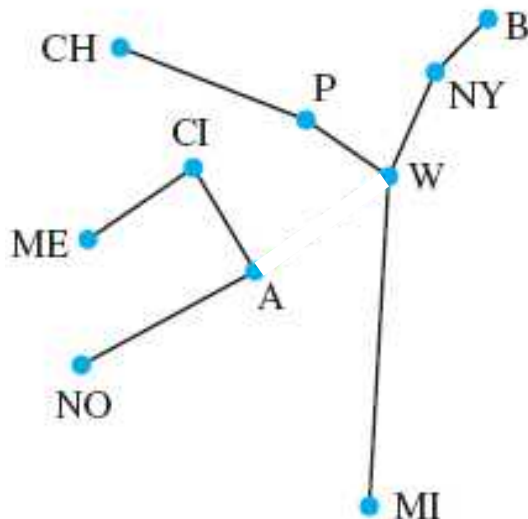
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Too many.

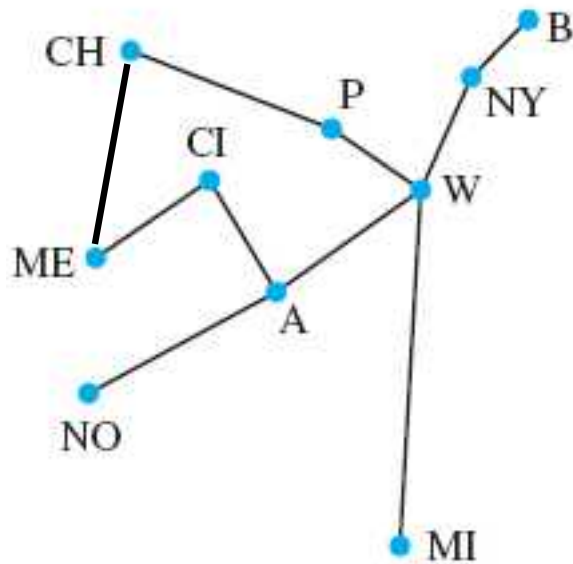
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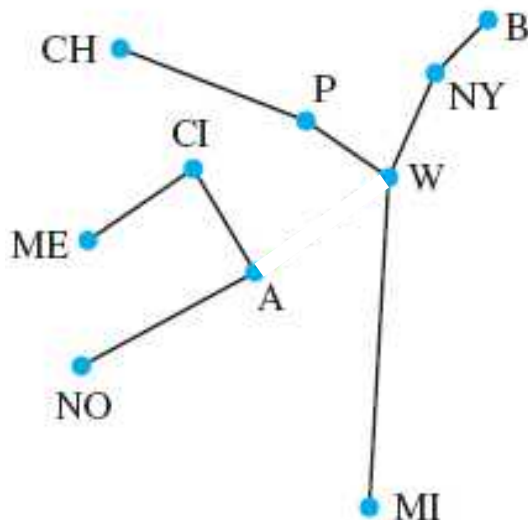
- Choosing 10 edges?



Too many.

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Choosing 8 edges?



Not enough.

There is **no path** from, e.g., **NO** to **B**.

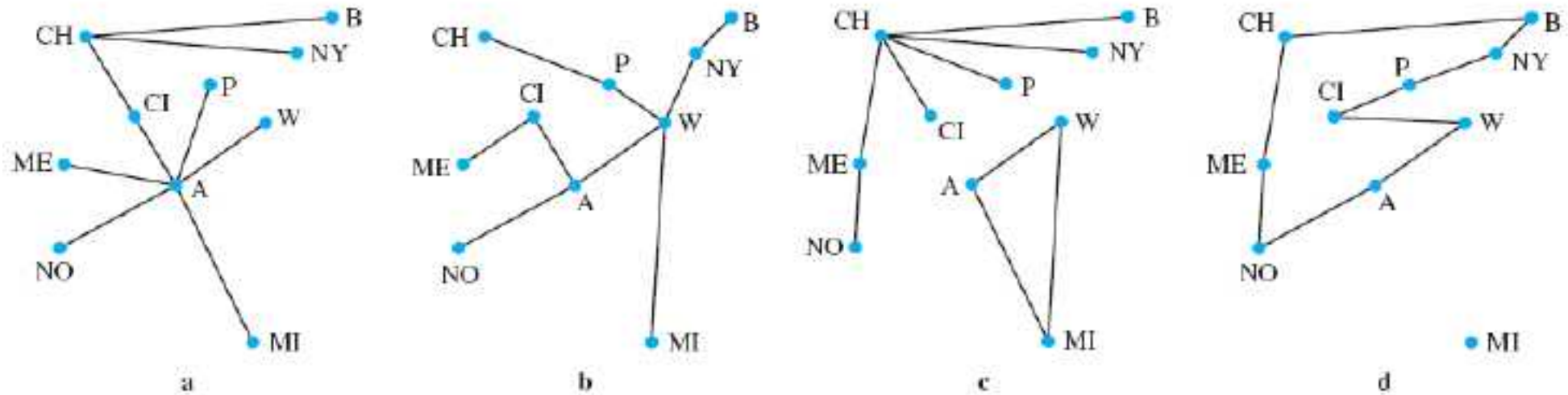


# Connectivity

- Choosing 9 edges:

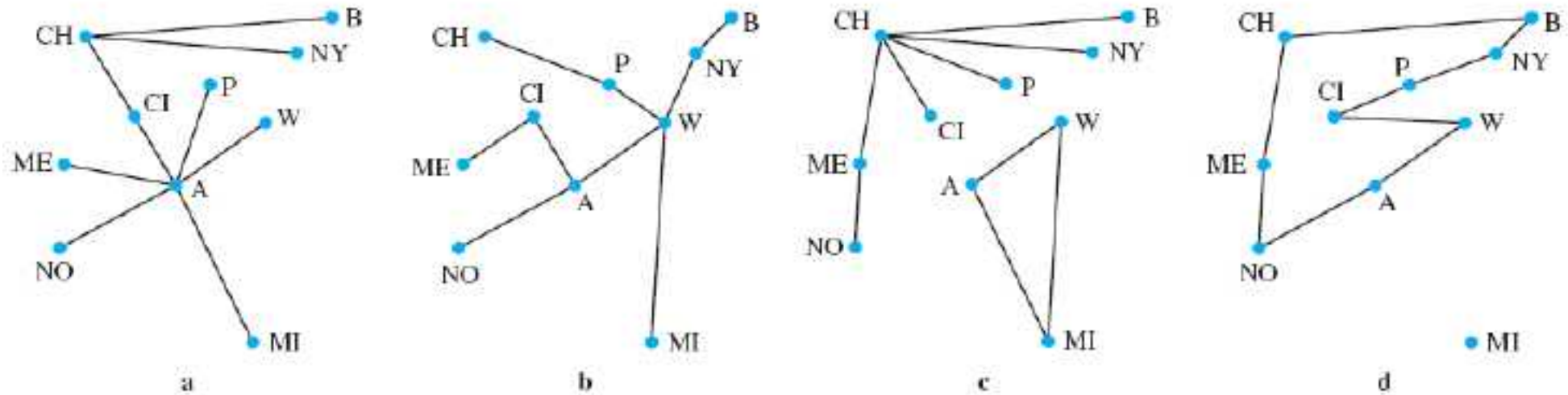
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## ■ Choosing 9 edges:



# Connectivity

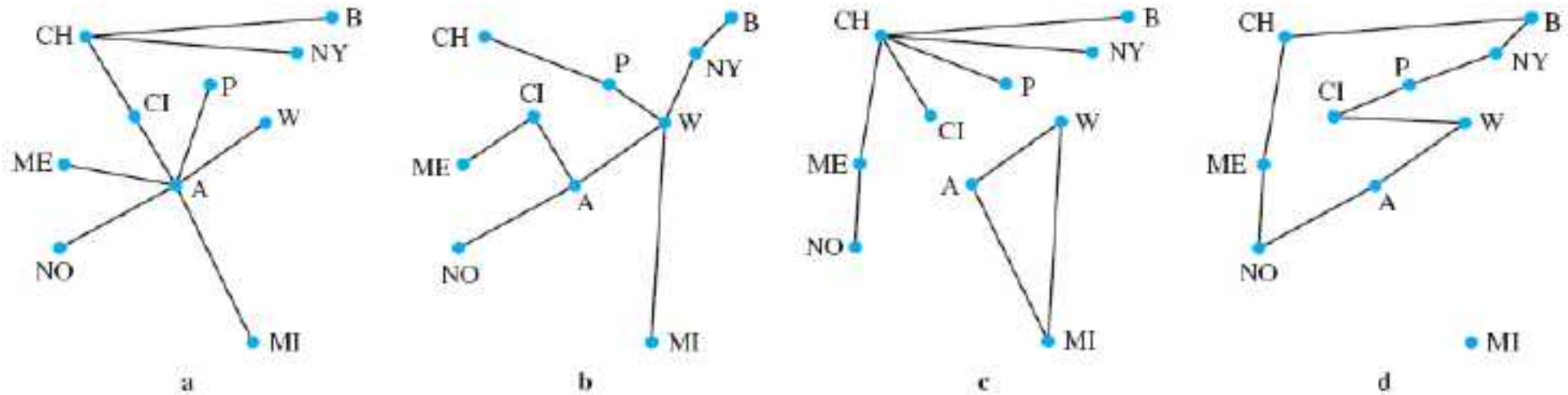
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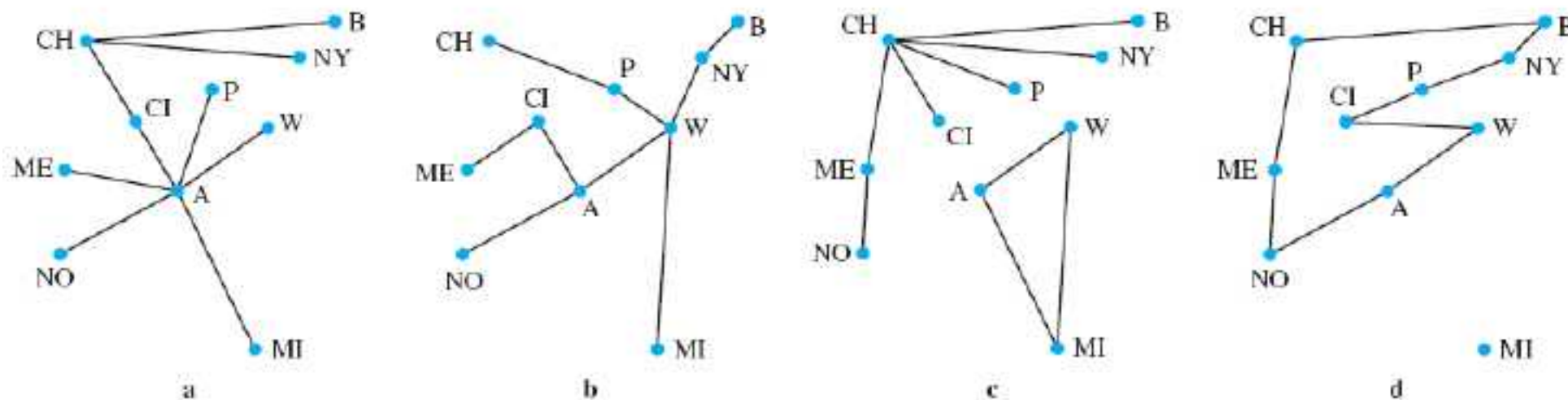
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**Definition** An undirected graph is called *connected* if there is a path between every pair of distinct vertices of the graph.

**Example:** (a) and (b) are *connected*, (c) and (d) are *disconnected*.

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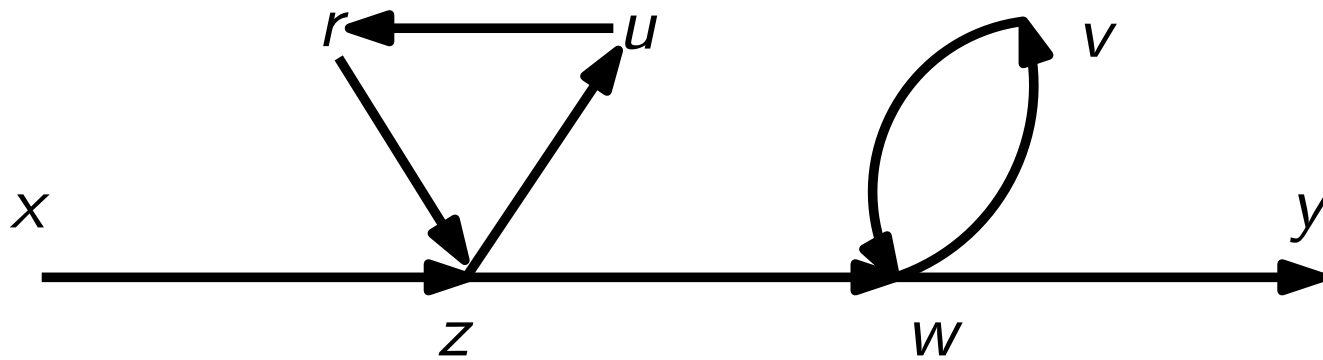
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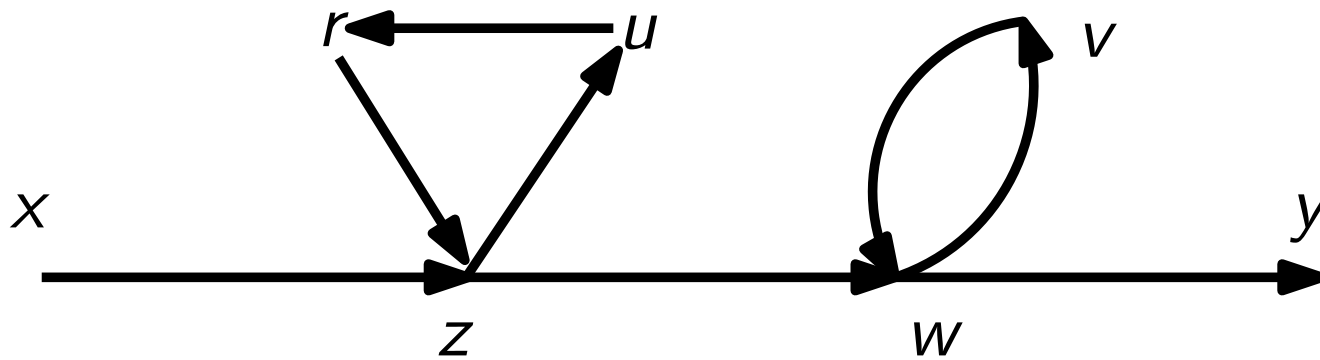
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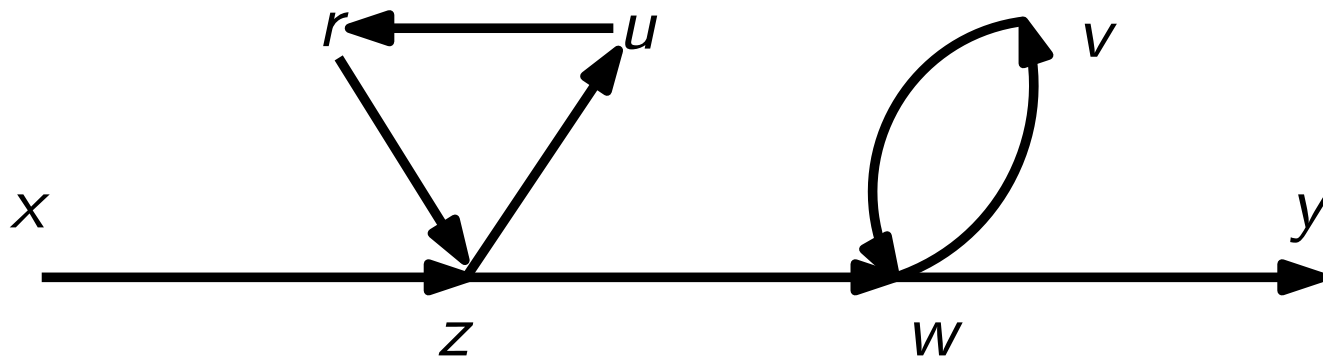
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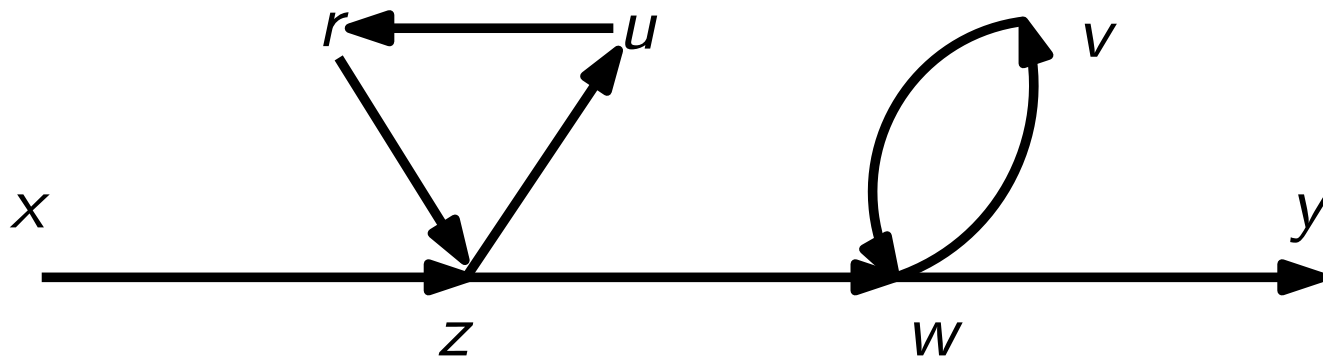
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**Theorem** There is a simple path between every pair of distinct vertices of a connected undirected graph.





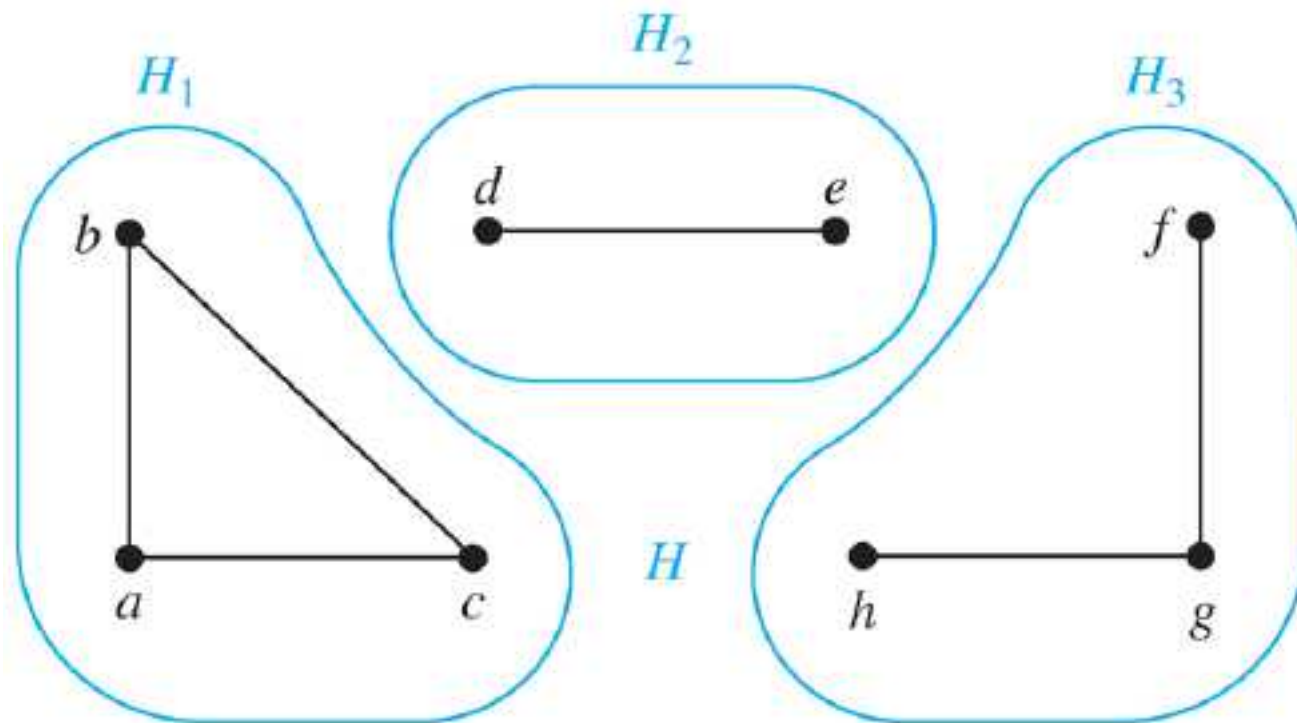
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- **Definition** A **directed graph** is *strongly connected* if there is a path **from  $a$  to  $b$**  and a path **from  $b$  to  $a$**  whenever  $a$  and  $b$  are vertices in the graph.



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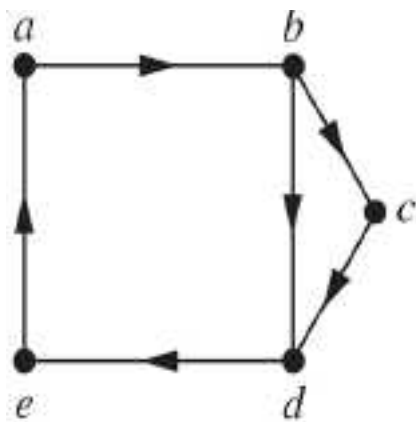
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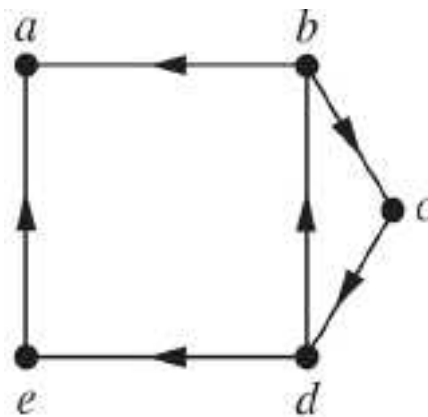
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$G$



$H$



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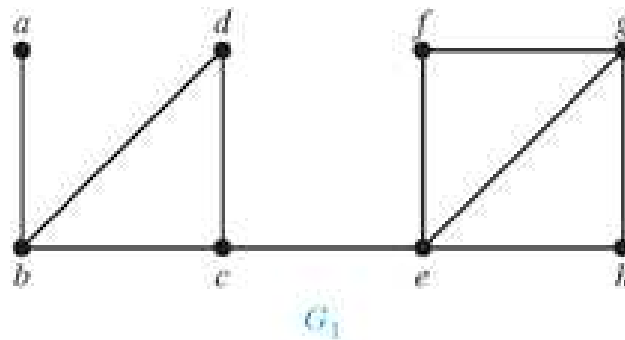
A set of edges  $E'$  is called an **edge cut** of  $G$  if the subgraph  $G - E'$  is **disconnected**. The **edge connectivity**  $\lambda(G)$  is the **minimum number** of edges in an edge cut of  $G$ .



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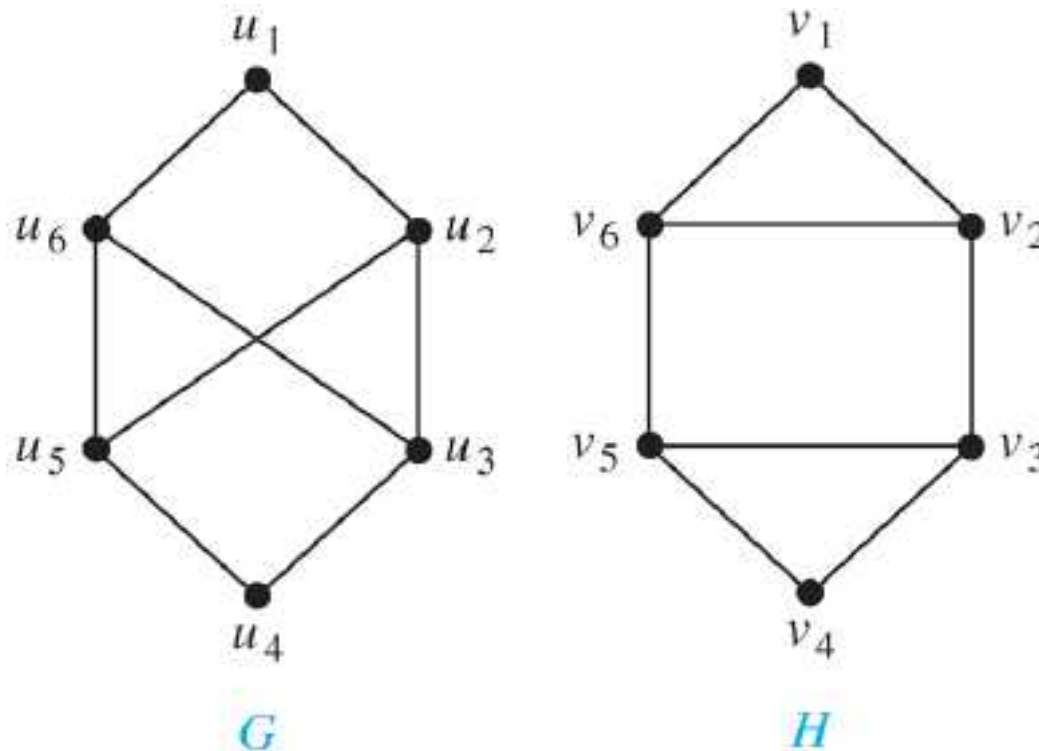
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- The existence of a simple circuit of length  $k$  is **isomorphic invariant**. In addition, **paths** can be used to construct mappings that may be **isomorphisms**.



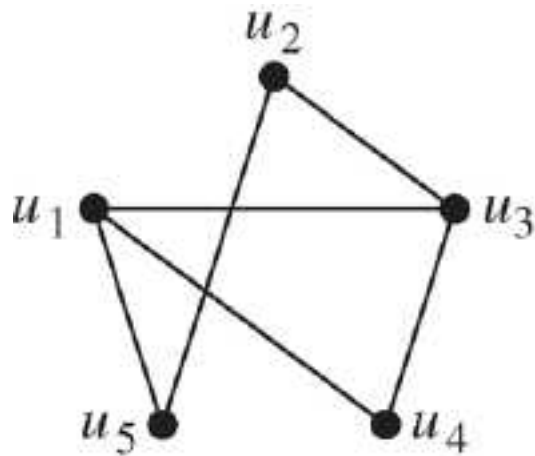
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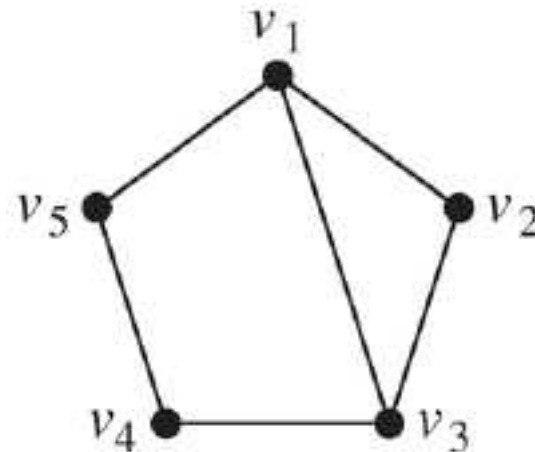


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# Counting Paths between Vertices

- **Theorem** Let  $G$  be a graph with adjacency matrix  $\mathbf{A}$  with respect to the ordering  $v_1, v_2, \dots, v_n$  of vertices. The number of different paths of length  $r$  from  $v_i$  to  $v_j$ , where  $r > 0$  is positive, equals the  $(i, j)$ -th entry of  $\mathbf{A}^r$ .



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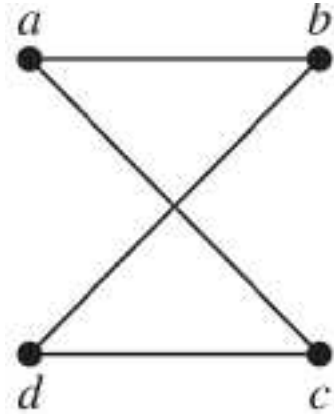
**Proof** (by **induction**)

$\mathbf{A}^{r+1} = \mathbf{A}^r \mathbf{A}$ , the  $(i, j)$ -th entry of  $\mathbf{A}^{r+1}$  equals  $b_{i1}a_{1j} + b_{i2}a_{2j} + \dots + b_{in}a_{nj}$ , where  $b_{ik}$  is the  $(i, k)$ -th entry of  $\mathbf{A}^r$ .



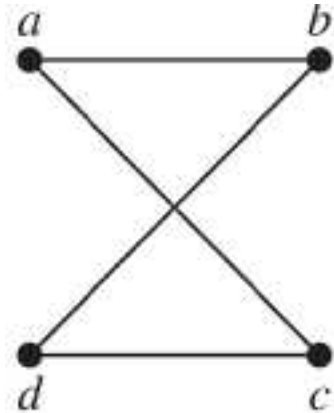
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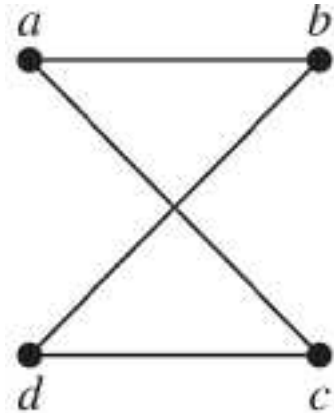
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# Next Lecture

- Graph theory II ...

