

CSC 225 - Summer 2019

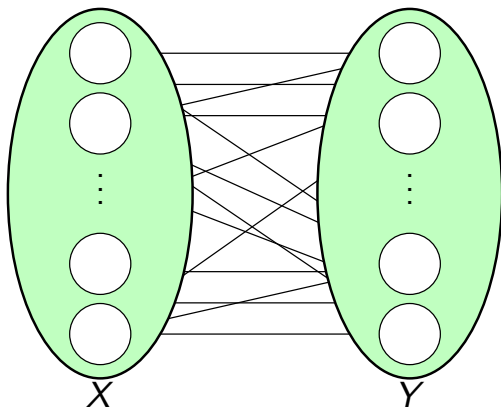
Graphs II

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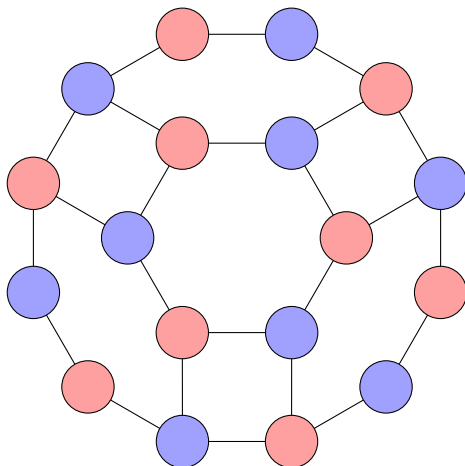
July 8, 2019

Bipartite Graphs (1)



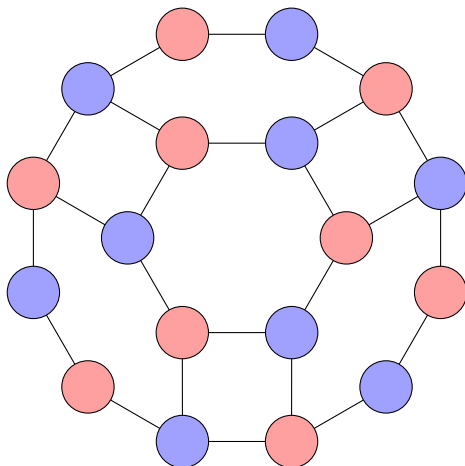
- ▶ A graph is **bipartite** if the vertices of G can be partitioned into sets X and Y such that no edge connects two vertices in X or two vertices in Y .
- ▶ The pair (X, Y) is called a **bipartition**.

Bipartite Graphs (2)



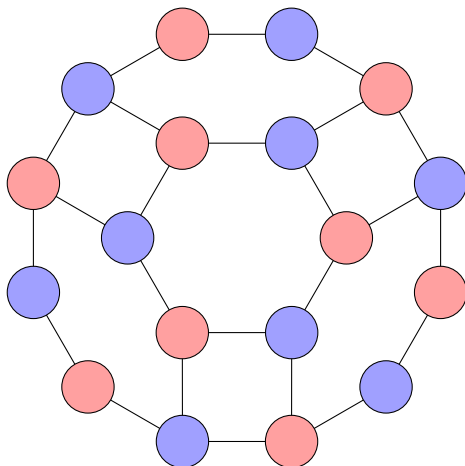
- ▶ The graph above is bipartite, since no pair of blue vertices is adjacent and no pair of red vertices is adjacent.
- ▶ There may be more than one possible bipartition.

Bipartite Graphs (3)



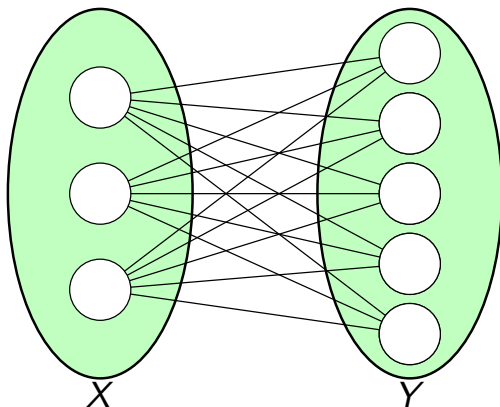
- ▶ A bipartite graph is also called **2-colourable**, since its vertices can be coloured with two colours such that no edge connects two vertices of the same colour.

Bipartite Graphs (4)



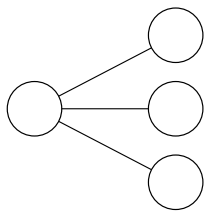
- Similarly, a bipartition can be called a '2-colouring'.

Bipartite Graphs (5)

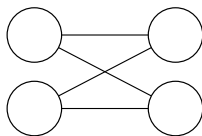


- ▶ A **complete bipartite graph**, denoted $K_{m,n}$, consists of a set X of m vertices and set Y of n vertices, with all possible edges between the two sets.
- ▶ The graph above is $K_{3,5}$.

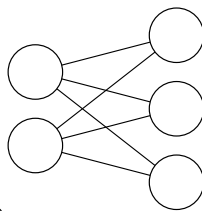
Bipartite Graphs (6)



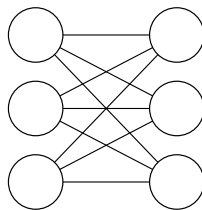
$K_{1,3}$



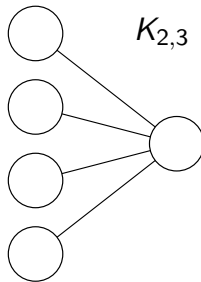
$K_{2,2}$



$K_{2,3}$



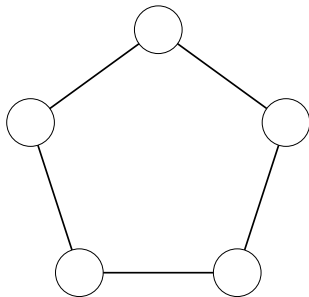
$K_{3,3}$



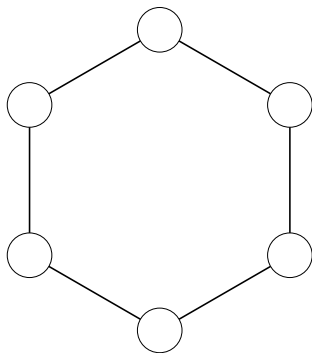
$K_{4,1}$

- The number of edges in $K_{m,n}$ is mn .

Cycles (1)



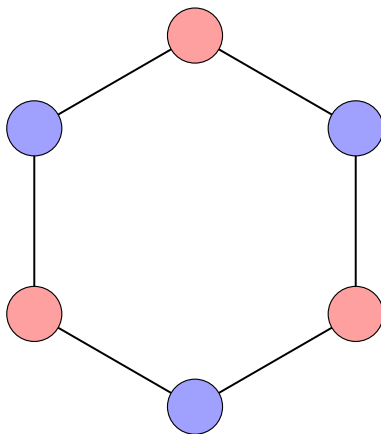
C_5



C_6

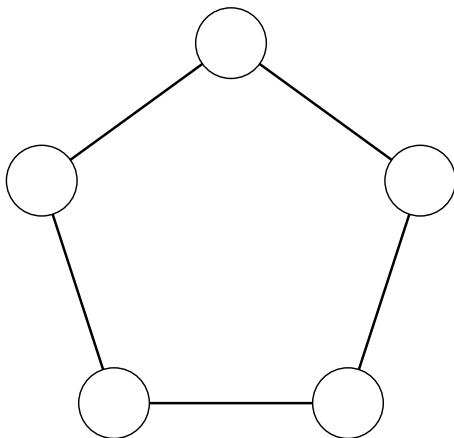
- ▶ A **cycle graph**, denoted C_n , consists of a cycle with n vertices (and n edges).

Cycles (2)



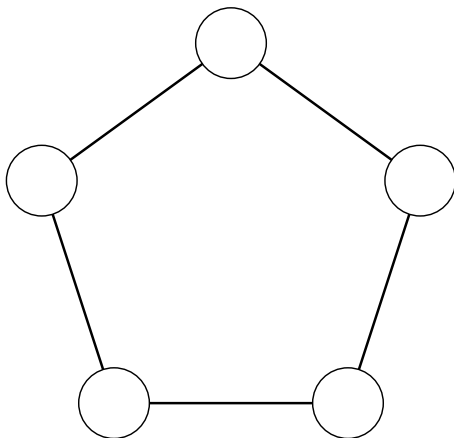
- ▶ The cycle C_6 is bipartite.
- ▶ It is easy to demonstrate that a graph is bipartite by finding a bipartition.

Cycles (3)



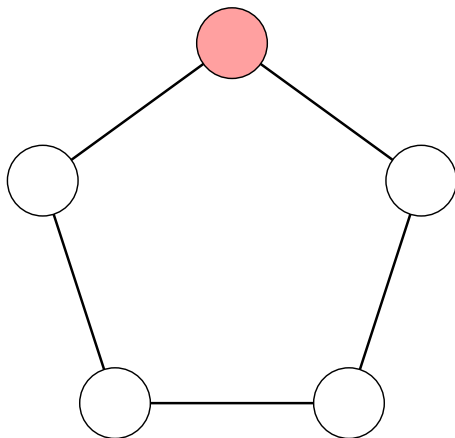
- ▶ **Exercise:** Prove that C_5 is not bipartite.
- ▶ This requires showing that it is impossible for a bipartition to exist.

Cycles (4)



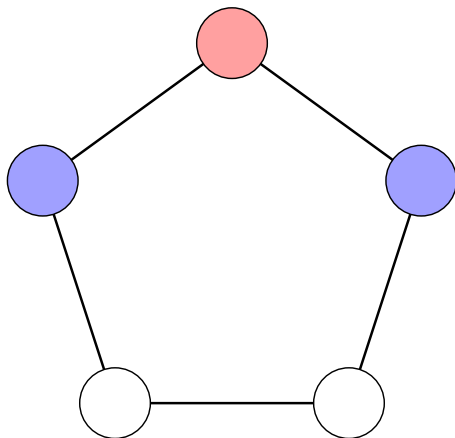
- To prove that C_5 is not bipartite, we will show that any possible 2-colouring leads to a contradiction.

Cycles (5)



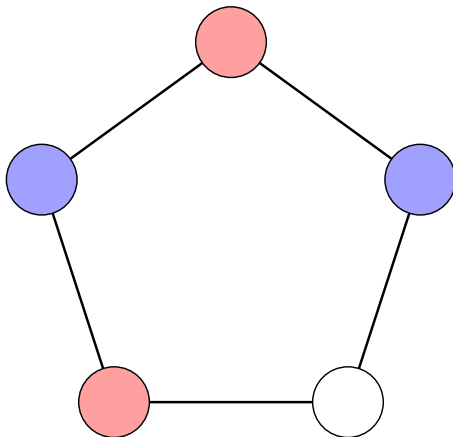
- ▶ First, we can assume that at least one vertex must be red. Without loss of generality, assume that it is the vertex shown.
- ▶ This is a reasonable assumption because all of the vertices are equivalent with respect to the structure of the graph.

Cycles (6)



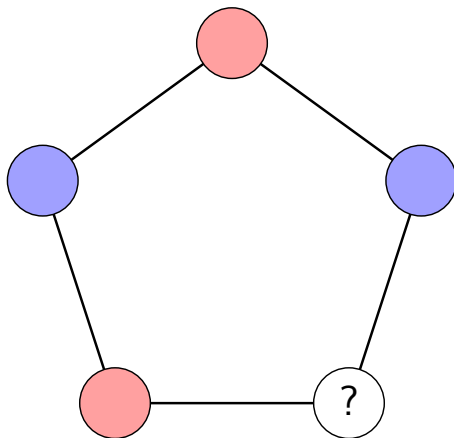
- ▶ Since the top vertex is red, both of its neighbours must be blue in a 2-colouring.

Cycles (7)



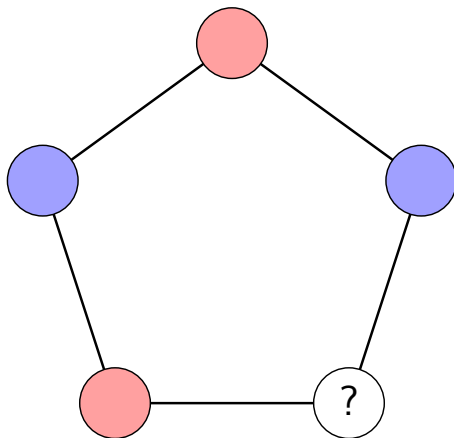
- The neighbour of the left blue vertex must be red.

Cycles (8)



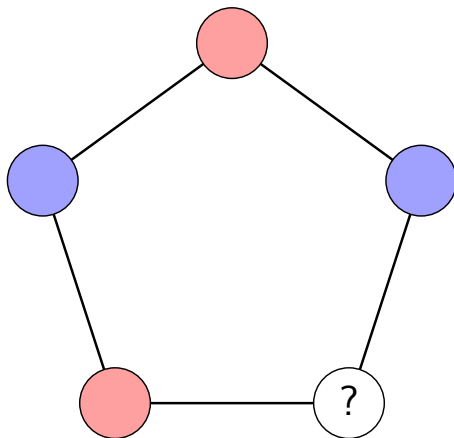
- ▶ The remaining vertex cannot be assigned a colour without breaking the rules, since it is adjacent to both a red and a blue vertex.
- ▶ Therefore, no 2-colouring exists.

Cycles (9)



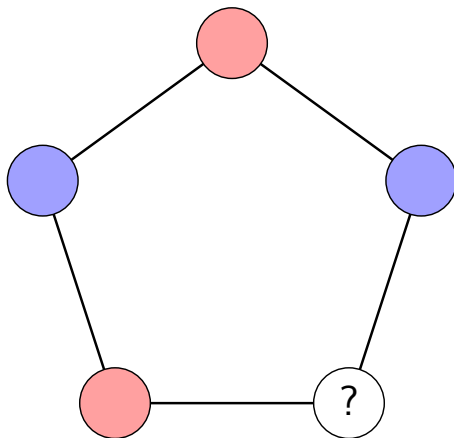
- ▶ The proof made exactly one assumption: There must be at least one red vertex.

Cycles (10)



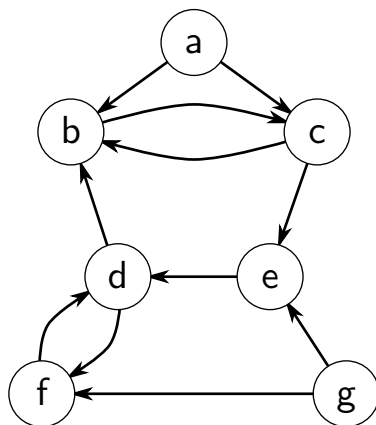
- ▶ All of the other steps in the proof were the result of mathematical requirements based on the definition of a 2-colouring.

Cycles (11)



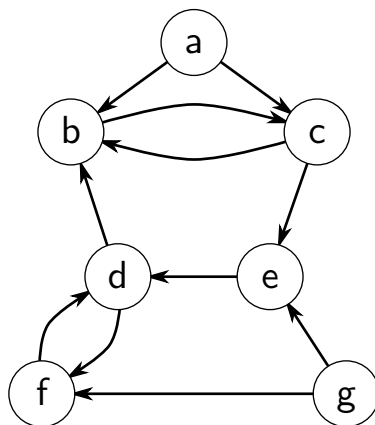
- **Warning:** The proof is only correct because the assumption is justified and sound. Without justifying the assumption, the 'proof' would just be an example, which proves nothing.

Directed Graphs (1)



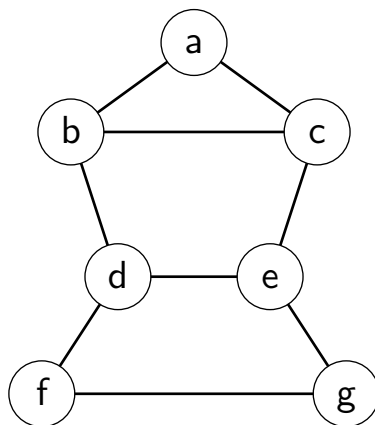
- ▶ A **directed graph** (or digraph) is a graph in which edges have a direction.
- ▶ Formally, edges in a directed graph correspond to ordered pairs of vertices.

Directed Graphs (2)



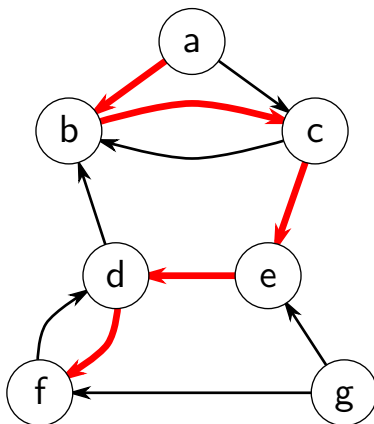
- ▶ Since edges are ordered pairs, the edges (b, c) and (c, b) are distinct.
- ▶ Edges in directed graphs are often called **directed edges** or **arcs**.

Directed Graphs (3)



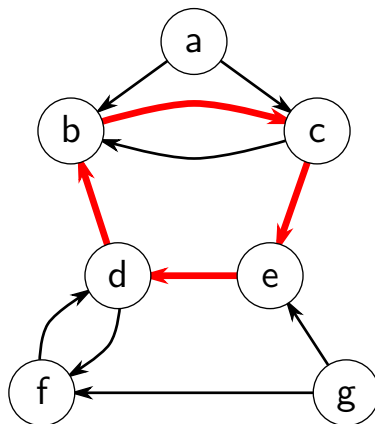
- ▶ Every directed graph can be collapsed into an undirected graph by removing the directions from edges (and combining bi-directional edge pairs into single undirected edges).

Directed Graphs (4)



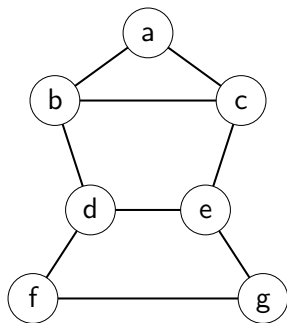
- ▶ A **directed path** is a path in a directed graph which respects the direction of edges.
- ▶ It is natural to refer to the endpoints of a directed path as the 'source' and the 'destination'.

Directed Graphs (5)



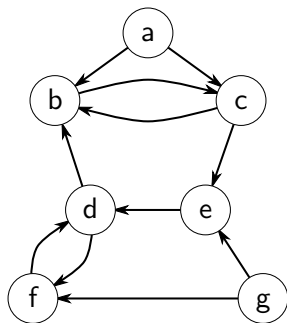
- A **directed cycle** is a directed path whose source and destination are the same.

Graph Representation (1)



$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{\{a, b\}, \{a, c\}, \{b, c\}, \\ \{b, d\}, \{c, e\}, \{d, e\}, \\ \{d, f\}, \{e, g\}, \{f, g\}\}$$

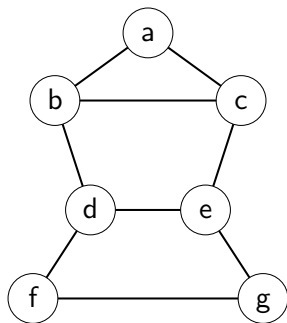


$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{(a, b), (a, c), (b, c), \\ (c, b), (c, e), (d, b), \\ (d, f), (e, d), (f, d), \\ (g, e), (g, f)\}$$

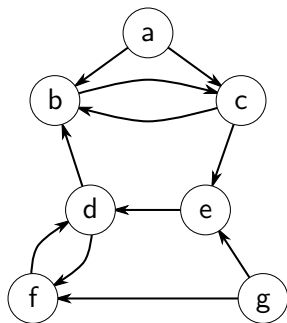
- The most obvious way to represent a graph is to explicitly store the sets V and E .

Graph Representation (2)



$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{\{a, b\}, \{a, c\}, \{b, c\}, \\ \{b, d\}, \{c, e\}, \{d, e\}, \\ \{d, f\}, \{e, g\}, \{f, g\}\}$$

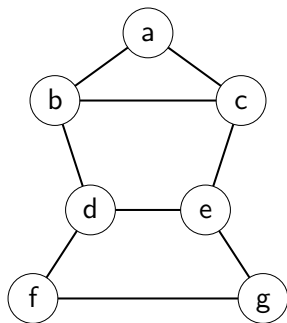


$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{(a, b), (a, c), (b, c), \\ (c, b), (c, e), (d, b), \\ (d, f), (e, d), (f, d), \\ (g, e), (g, f)\}$$

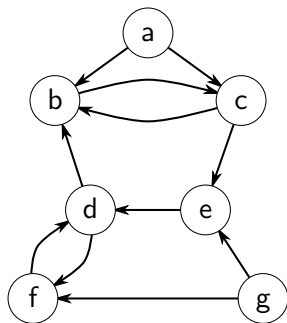
- For lack of a better term, this representation is called an **edge list**.

Graph Representation (3)



$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{\{a, b\}, \{a, c\}, \{b, c\}, \\ \{b, d\}, \{c, e\}, \{d, e\}, \\ \{d, f\}, \{e, g\}, \{f, g\}\}$$



$$V = \{a, b, c, d, e, f, g\}$$

$$E = \{(a, b), (a, c), (b, c), \\ (c, b), (c, e), (d, b), \\ (d, f), (e, d), (f, d), \\ (g, e), (g, f)\}$$

- In an implementation, each edge could be represented by an object containing pointers to its endpoint vertices.

Graph Representation (4)

Count edges in G

Find all neighbours of v

Test if edge vu exists

Add edge vu

Delete edge vu

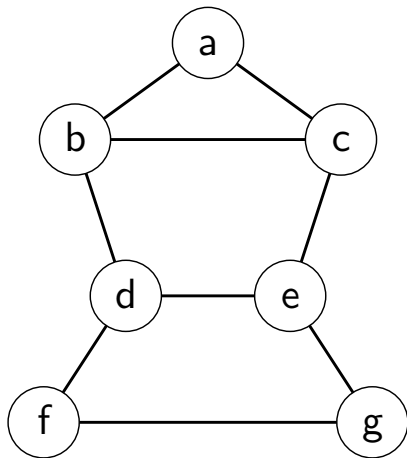
Edge List	Adj. List (unsorted)	Adj. List (sorted)	Adjacency Matrix
$\Theta(m)$	$\Theta(m)$	$\Theta(m)$	$\Theta(n^2)$
$\Theta(m)$	$\Theta(\deg(v))$	$\Theta(\deg(v))$	$\Theta(n)$
$\Theta(m)$	$\Theta(\deg(v))$	$\Theta(\log(\deg(v)))$	$\Theta(1)$
$\Theta(1)$	$\Theta(1)$	$\Theta(\deg(v))$	$\Theta(1)$
$\Theta(1)^1$	$\Theta(\deg(v))$	$\Theta(\log(\deg(v)))$	$\Theta(1)$

Space Complexity

$\Theta(n + m)$	$\Theta(n + m)$	$\Theta(n + m)$	$\Theta(n^2)$
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¹If it is necessary to search for edge vu first, deletion takes $\Theta(m)$ time.

Adjacency Lists (1)

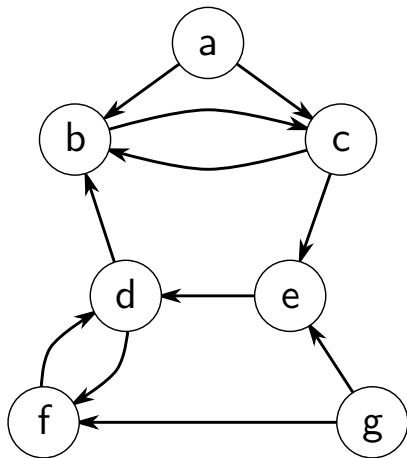


Adjacency List:

Vertex	Neighbours
a	b, c
b	a, c, d
c	a, b, e
d	b, e, f
e	c, d, g
f	d, g
g	e, f

- ▶ An **adjacency list** structure represents a graph with lists of each vertex's neighbours.

Adjacency Lists (2)

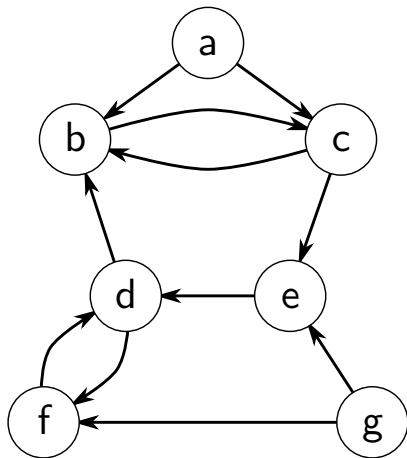


Adjacency List:

Vertex	Neighbours
a	b, c
b	c
c	b, e
d	b, f
e	d
f	d
g	e, f

- ▶ In a directed graph, the neighbour list normally only stores 'outbound' neighbours.
- ▶ A second set of lists can be created to store inbound neighbours if necessary.

Adjacency Lists (3)

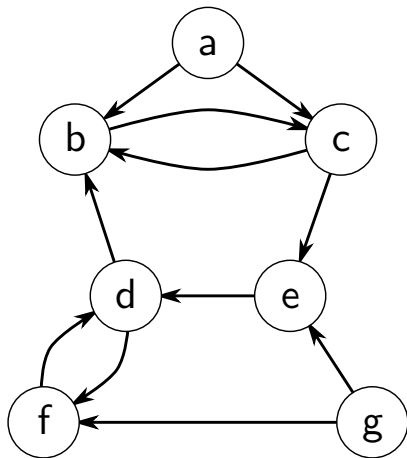


Adjacency List:

Vertex	Neighbours
a	b, c
b	c
c	b, e
d	b, f
e	d
f	d, g
g	f

- ▶ The total size of all neighbour lists together is $2m$, since each edge is represented twice (once for each endpoint).

Adjacency Lists (4)

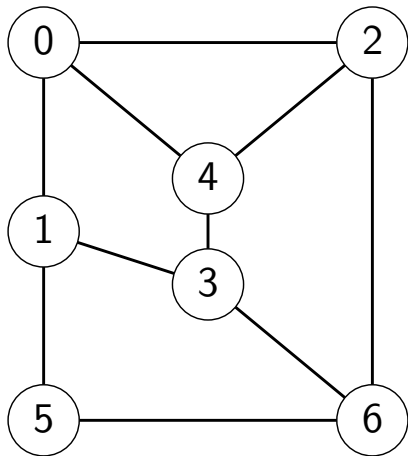


Adjacency List:

Vertex	Neighbours
a	b, c
b	c
c	b, e
d	b, f
e	d
f	d
g	e, f

- ▶ One advantage of adjacency lists is their compactness: Graphs with few edges will require less space than more dense graphs.

Adjacency Lists (5)

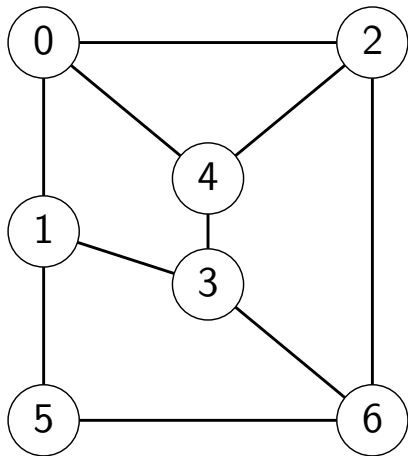


Adjacency List:

Vertex	Neighbours
0	1, 2, 4
1	0, 3, 5
2	0, 4, 6
3	1, 4, 6
4	0, 2, 3
5	1, 6
6	2, 3, 5

- In most cases, it is helpful to number vertices starting at 0 and use their indices to construct an adjacency list.

Adjacency Lists (6)

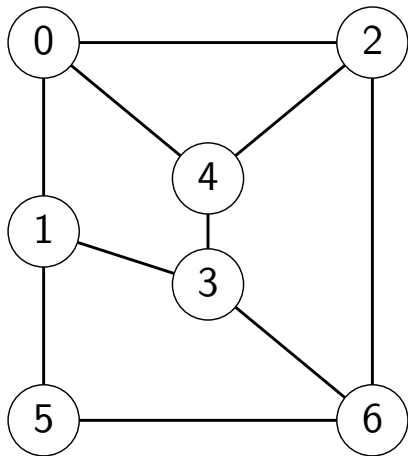


Adjacency List:

Vertex	Neighbours
0	1, 2, 4
1	0, 3, 5
2	0, 4, 6
3	1, 4, 6
4	0, 2, 3
5	1, 6
6	2, 3, 5

- ▶ Assigning an index to each vertex allows the data structure to be stored in an array of n lists.

Adjacency Lists (7)

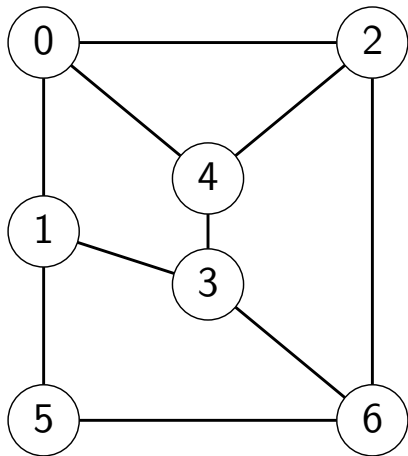


Adjacency List:

Vertex	Neighbours
0	1, 2, 4
1	0, 3, 5
2	0, 4, 6
3	1, 4, 6
4	0, 2, 3
5	1, 6
6	2, 3, 5

- ▶ Alternatively, the data structure can be represented by a set of vertex objects, each with a list of pointers to its neighbours.

Adjacency Lists (8)

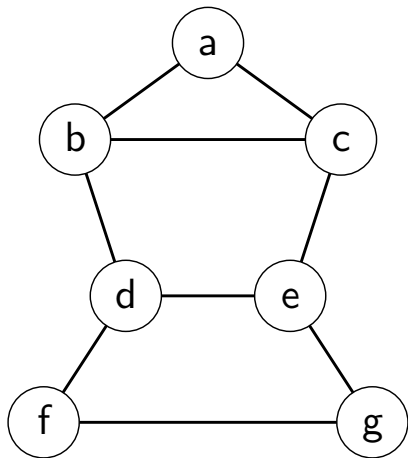


Adjacency List:

Vertex	Neighbours
0	1, 2, 4
1	0, 3, 5
2	0, 4, 6
3	1, 4, 6
4	0, 2, 3
5	1, 6
6	2, 3, 5

- ▶ The linked representation is preferable when vertices might be added to or removed from the graph after construction.

Adjacency Matrices (1)

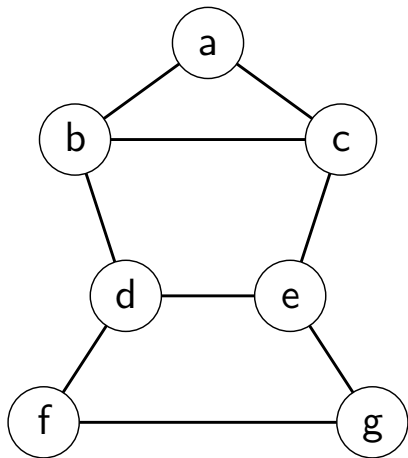


Adjacency Matrix:

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

- ▶ An **adjacency matrix** structure represents a graph on n vertices with an $n \times n$ matrix of 0/1 values.

Adjacency Matrices (2)

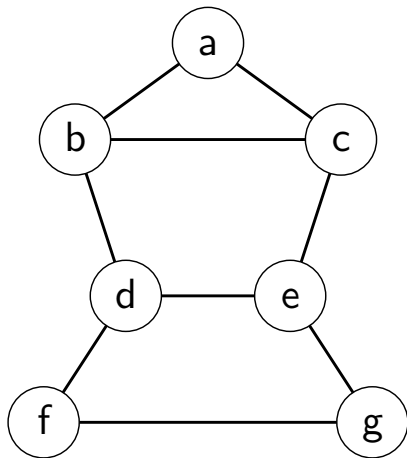


Adjacency Matrix:

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

- ▶ Each vertex receives both a row and column of the matrix.
- ▶ If the entry in row u and column v of the matrix is 1, then there is an edge uv in the graph.

Adjacency Matrices (3)

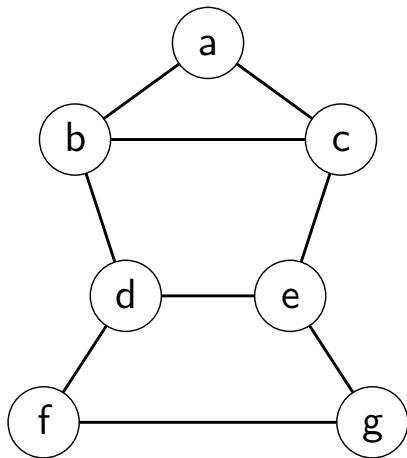


Adjacency Matrix:

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

- ▶ The adjacency matrix of an undirected graph is always symmetric.
- ▶ The space complexity of an adjacency matrix is always $\Theta(n^2)$.

Adjacency Matrices (4)

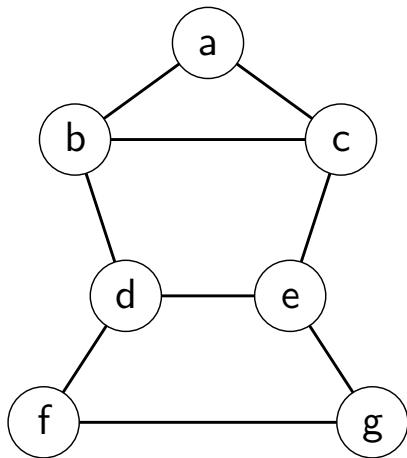


Adjacency Matrix:

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

- ▶ On graphs with a small number of edges, adjacency matrices can be inefficient.

Adjacency Matrices (5)

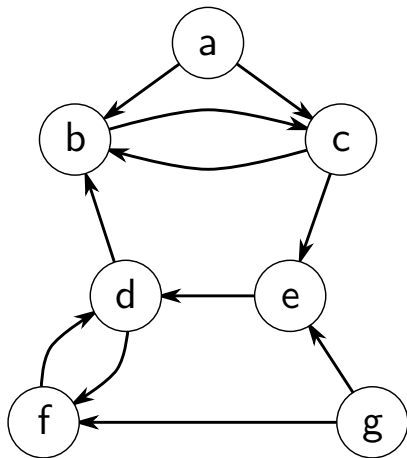


Adjacency Matrix:

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

- ▶ Finding all of the neighbours of a given vertex requires $\Theta(n)$ operations, since every entry in a row (or column) must be inspected.

Adjacency Matrices (6)

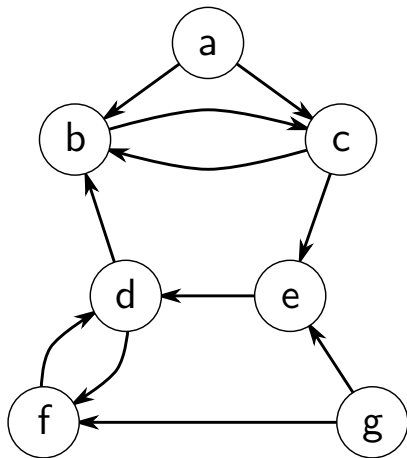


Adjacency Matrix:

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

- For a directed graph, rows represent source vertices and columns represent destination vertices.

Adjacency Matrices (7)

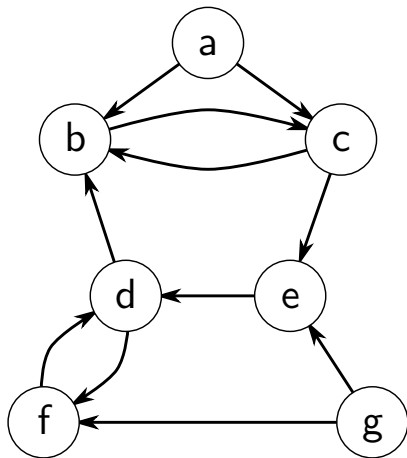


Adjacency Matrix:

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

- If the entry at row u and column v is 1, then there is an edge from u to v .

Adjacency Matrices (8)



Adjacency Matrix:

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

- ▶ As with adjacency lists, it can be helpful to assign vertices an index starting at 0 to allow fast indexing into the matrix.