

CHAPTER 6

GRAPHS

All the programs in this file are selected from

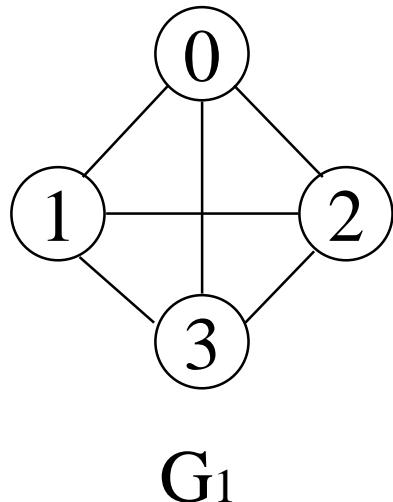
Ellis Horowitz, Sartaj Sahni, and Susan Anderson-Freed
“Fundamentals of Data Structures in C”,

Definition

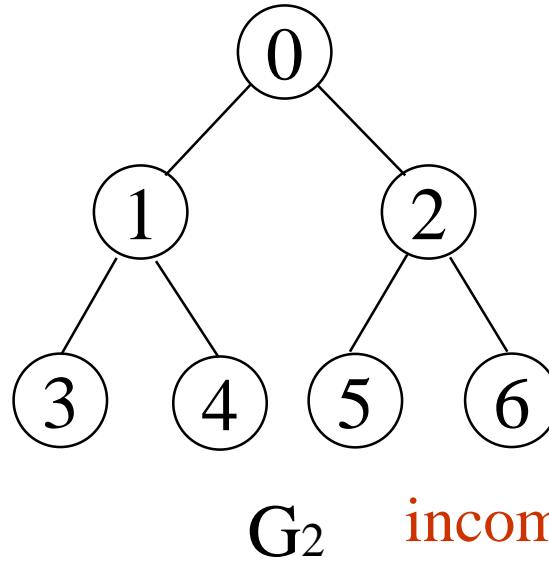
- A graph G consists of two sets
 - a finite, nonempty set of vertices $V(G)$
 - a finite, possible empty set of edges $E(G)$
 - $G(V, E)$ represents a graph
- An undirected graph is one in which the pair of vertices in a edge is unordered, $(v_0, v_1) = (v_1, v_0)$
- A directed graph is one in which each edge is a directed pair of vertices, $\langle v_0, v_1 \rangle \neq \langle v_1, v_0 \rangle$



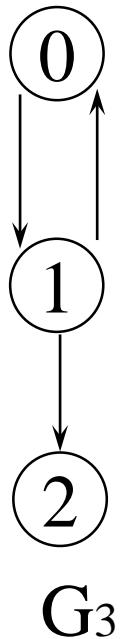
Examples for Graph



complete graph



incomplete graph



$$V(G_1) = \{0, 1, 2, 3\}$$

$$E(G_1) = \{(0,1), (0,2), (0,3), (1,2), (1,3), (2,3)\}$$

$$V(G_2) = \{0, 1, 2, 3, 4, 5, 6\}$$

$$E(G_2) = \{(0,1), (0,2), (1,3), (1,4), (2,5), (2,6)\}$$

$$V(G_3) = \{0, 1, 2\}$$

$$E(G_3) = \{<0,1>, <1,0>, <1,2>\}$$

complete undirected graph: $n(n-1)/2$ edges

complete directed graph: $n(n-1)$ edges

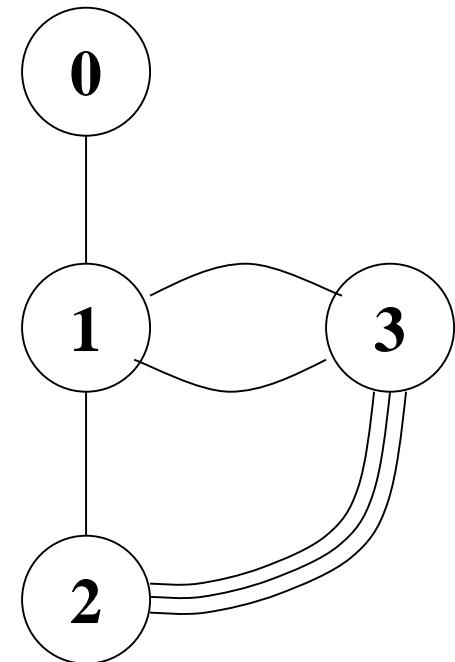
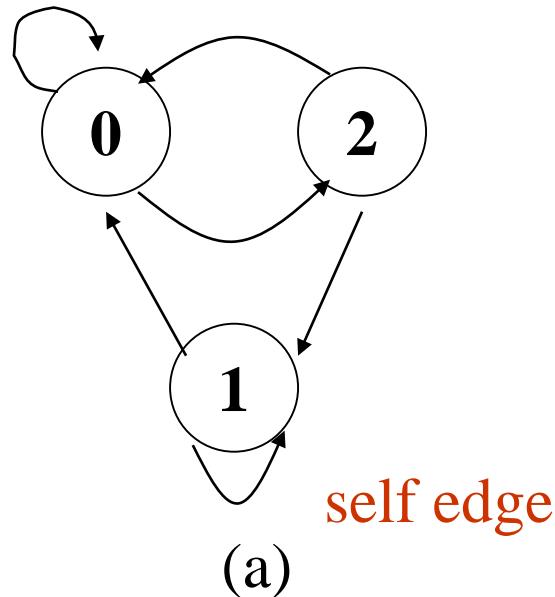
Complete Graph

- A complete graph is a graph that has the maximum number of edges
 - for **undirected graph** with n vertices, the maximum number of edges is $n(n-1)/2$
 - for **directed graph** with n vertices, the maximum number of edges is $n(n-1)$
 - example: G_1 is a complete graph

Adjacent and Incident

- If (v_0, v_1) is an edge in an undirected graph,
 - v_0 and v_1 are adjacent
 - The edge (v_0, v_1) is incident on vertices v_0 and v_1
- If $\langle v_0, v_1 \rangle$ is an edge in a directed graph
 - v_0 is adjacent to v_1 , and v_1 is adjacent from v_0
 - The edge $\langle v_0, v_1 \rangle$ is incident on v_0 and v_1

***Figure 6.3:** Example of a graph with feedback loops and a multigraph

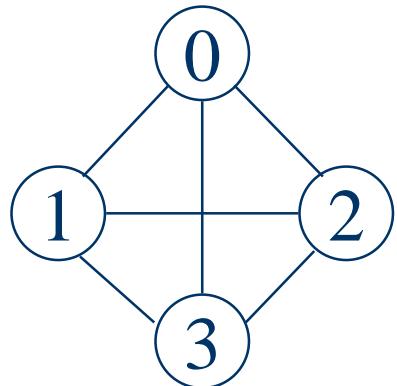


multiple occurrences of the same edge

Subgraph and Path

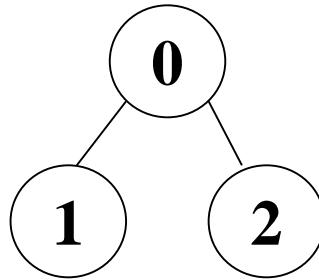
- A **subgraph** of G is a graph G' such that $V(G')$ is a subset of $V(G)$ and $E(G')$ is a subset of $E(G)$
- A **path** from vertex v_p to vertex v_q in a graph G , is a sequence of vertices, $v_p, v_{i1}, v_{i2}, \dots, v_{in}, v_q$, such that $(v_p, v_{i1}), (v_{i1}, v_{i2}), \dots, (v_{in}, v_q)$ are edges in an undirected graph
- The **length of a path** is the number of edges on it

Figure 6.4: subgraphs of G_1 and G_3

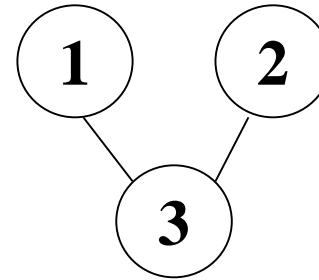


G_1

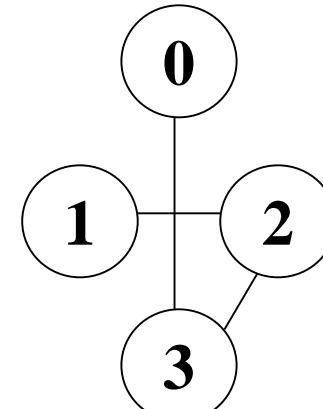
(i)



(ii)



(iii)

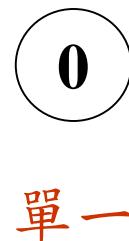


(iv)

(a) Some of the subgraph of G_1

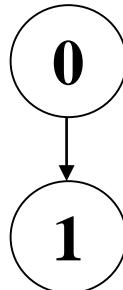


G_3

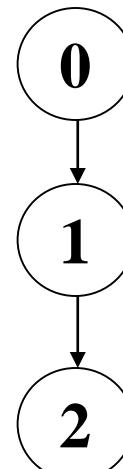


單一

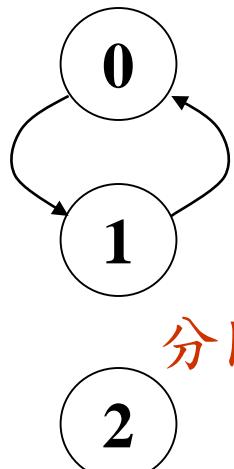
(i)



(ii)



(iii)



分開

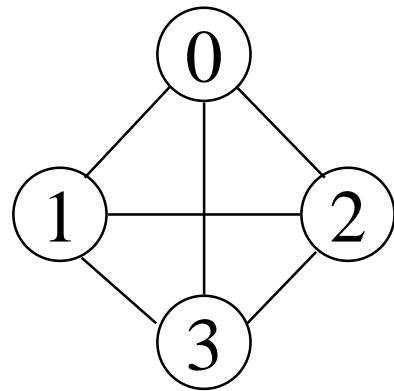
(iv)

(b) Some of the subgraph of G_3

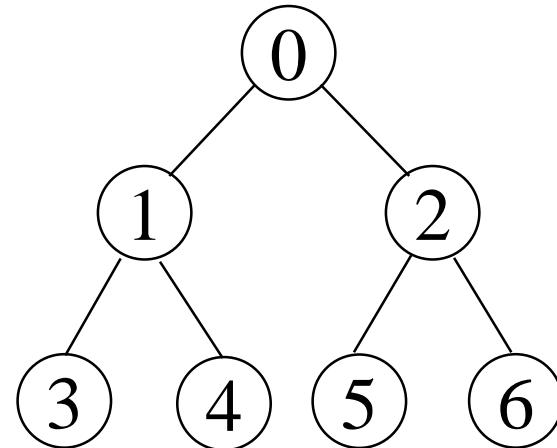
Simple Path and Style

- A **simple path** is a path in which all vertices, except possibly the first and the last, are distinct
- A **cycle** is a simple path in which the first and the last vertices are the same
- In an undirected graph G , two **vertices**, v_0 and v_1 , are **connected** iff there is a path in G from v_0 to v_1
- An undirected **graph** is **connected** iff for every pair of distinct vertices v_i, v_j , there is a path from v_i to v_j

Connected



G_1



G_2

tree (acyclic graph)

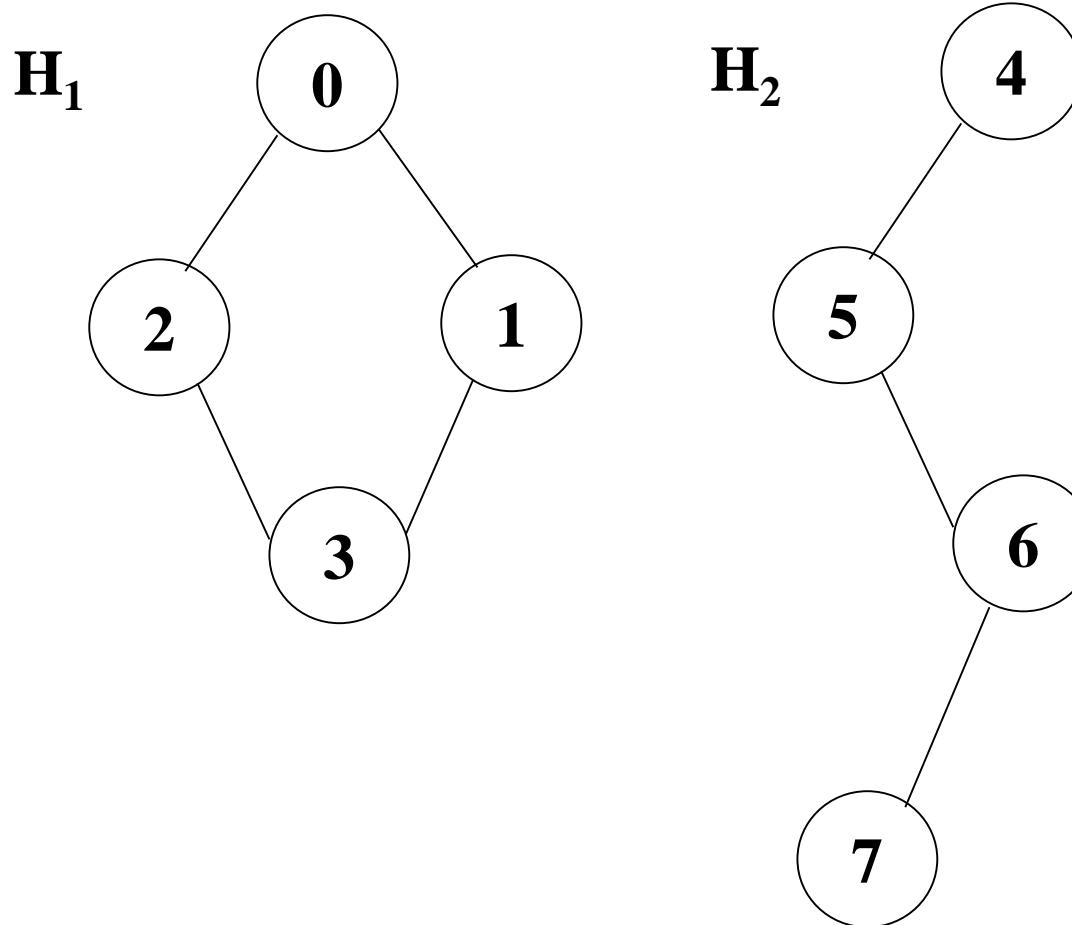
connected component

Connected Component

- A **connected component** of an undirected graph is a maximal connected subgraph.
- A **tree** is a graph that is connected and acyclic (i.e., has no cycles).
- A directed graph is **strongly connected** if there is a directed path from v_i to v_j and also from v_j to v_i .
- A **strongly connected component** is a maximal subgraph that is strongly connected.

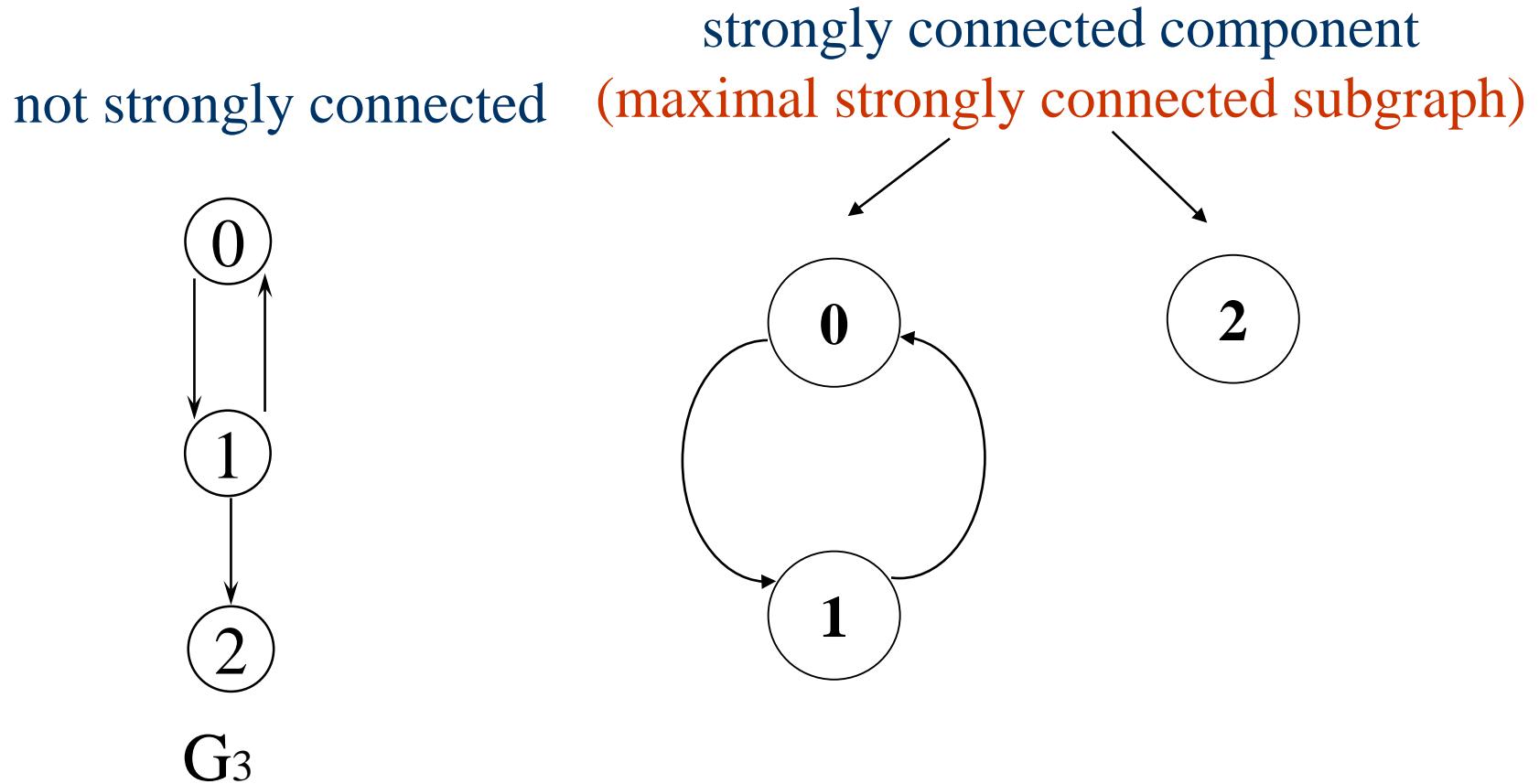
***Figure 6.5: A graph with two connected components (p.262)**

connected component (maximal connected subgraph)



G_4 (not connected)

***Figure 6.6:** Strongly connected components of G_3



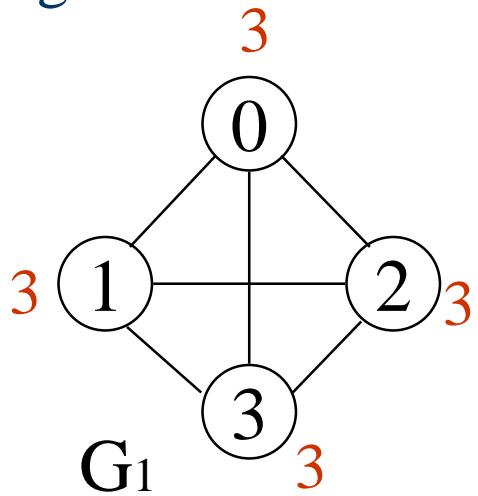
Degree

- The **degree** of a vertex is the number of edges incident to that vertex
- For directed graph,
 - the **in-degree** of a vertex v is the number of edges that have v as the head
 - the **out-degree** of a vertex v is the number of edges that have v as the tail
 - if d_i is the degree of a vertex i in a graph G with n vertices and e edges, the number of edges is

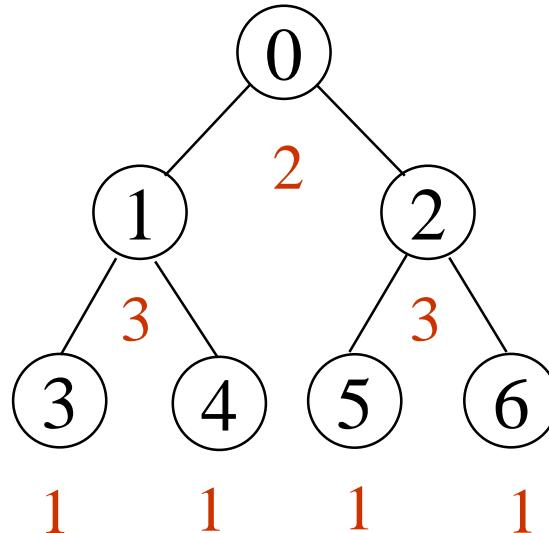
$$e = \left(\sum_0^{n-1} d_i \right) / 2$$

undirected graph

degree



G_2

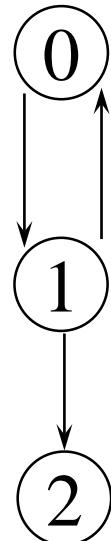


directed graph

in-degree

out-degree

G_3



in:1, out: 1

in: 1, out: 2

in: 1, out: 0

ADT for Graph

structure Graph is

objects: a nonempty set of vertices and a set of undirected edges, where each edge is a pair of vertices

functions: for all $graph \in Graph$, v, v_1 and $v_2 \in Vertices$

$Graph$ Create():=return an empty graph

$Graph$ InsertVertex($graph, v$):=: return a graph with v inserted. v has no incident edge.

$Graph$ InsertEdge($graph, v_1, v_2$):=: return a graph with new edge between v_1 and v_2

$Graph$ DeleteVertex($graph, v$):=: return a graph in which v and all edges incident to it are removed

$Graph$ DeleteEdge($graph, v_1, v_2$):=: return a graph in which the edge (v_1, v_2) is removed

$Boolean$ IsEmpty($graph$):=: if ($graph == empty\ graph$) return TRUE
else return FALSE

$List$ Adjacent($graph, v$):=: return a list of all vertices that are adjacent to v

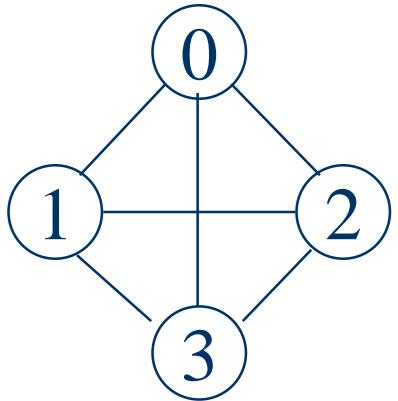
Graph Representations

- Adjacency Matrix
- Adjacency Lists
- Adjacency Multilists

Adjacency Matrix

- Let $G=(V,E)$ be a graph with n vertices.
- The **adjacency matrix** of G is a two-dimensional $n * n$ array, say adj_mat
- If the edge (v_i, v_j) is in $E(G)$, $\text{adj_mat}[i][j]=1$
- If there is no such edge in $E(G)$, $\text{adj_mat}[i][j]=0$
- The adjacency matrix for an undirected graph is symmetric; the adjacency matrix for a digraph need not be symmetric

Examples for Adjacency Matrix



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

G_1



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

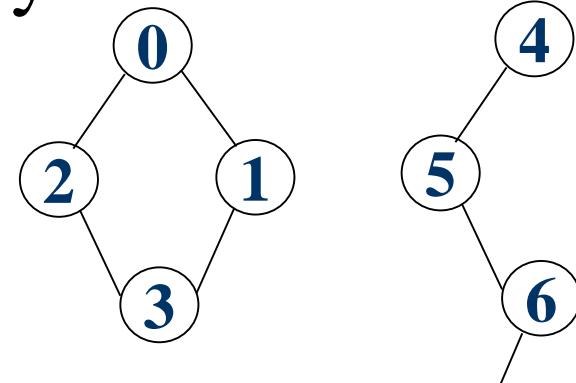
G_2

symmetric

undirected: $n*(n+1)/2$

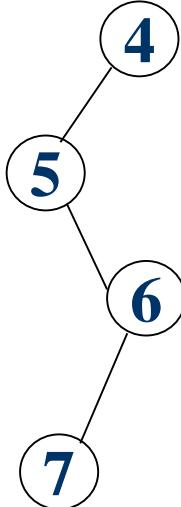
directed: n^2

Memory request



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

G_3



Merits of Adjacency Matrix

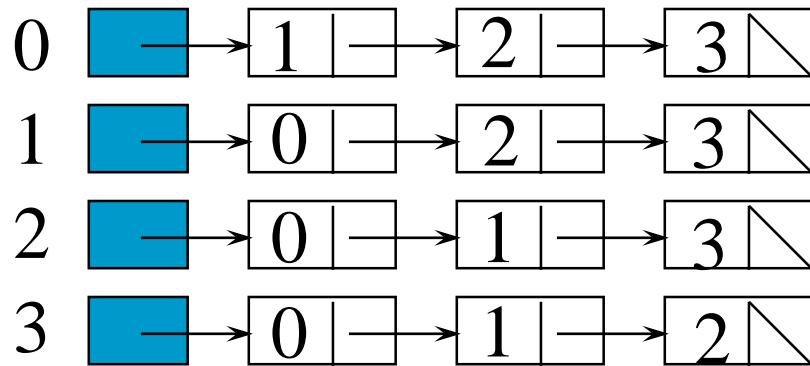
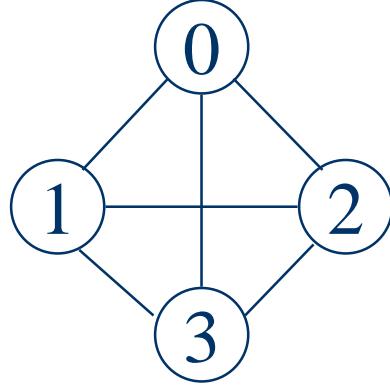
- From the adjacency matrix, to determine the connection of vertices is easy
- The degree of a vertex is $\sum_{j=0}^{n-1} adj_mat[i][j]$
- For a directed graph, the row sum is the out_degree, while the column sum is the in_degree

$$ind(vi) = \sum_{j=0}^{n-1} A[j, i] \quad outd(vi) = \sum_{j=0}^{n-1} A[i, j]$$

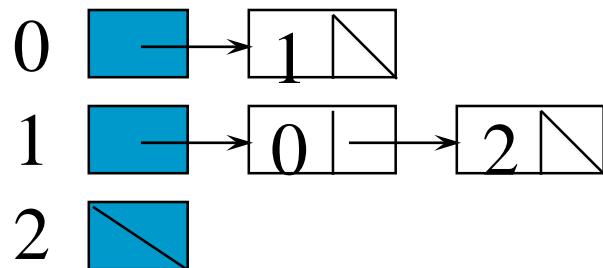
Data Structures for Adjacency Lists

Each row in adjacency matrix is represented as an adjacency list.

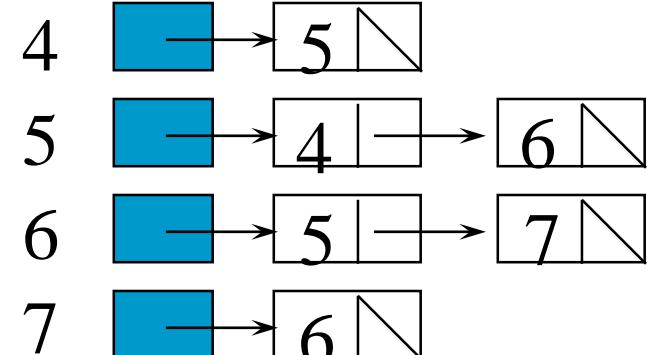
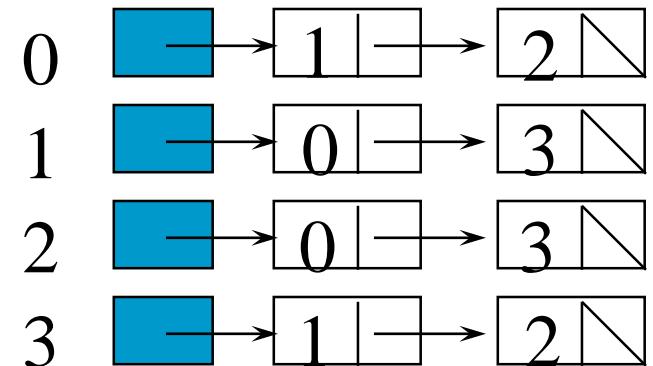
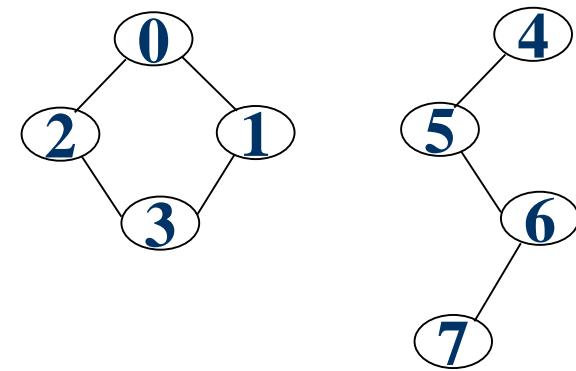
```
#define MAX_VERTICES 50
typedef struct node *node_pointer;
typedef struct node {
    int vertex;
    struct node *link;
} ;
node_pointer graph[MAX_VERTICES];
int n=0; /* vertices currently in use */
```



G_1



G_3

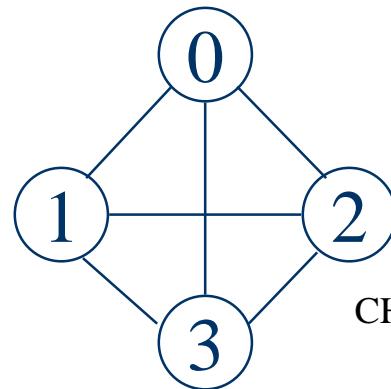
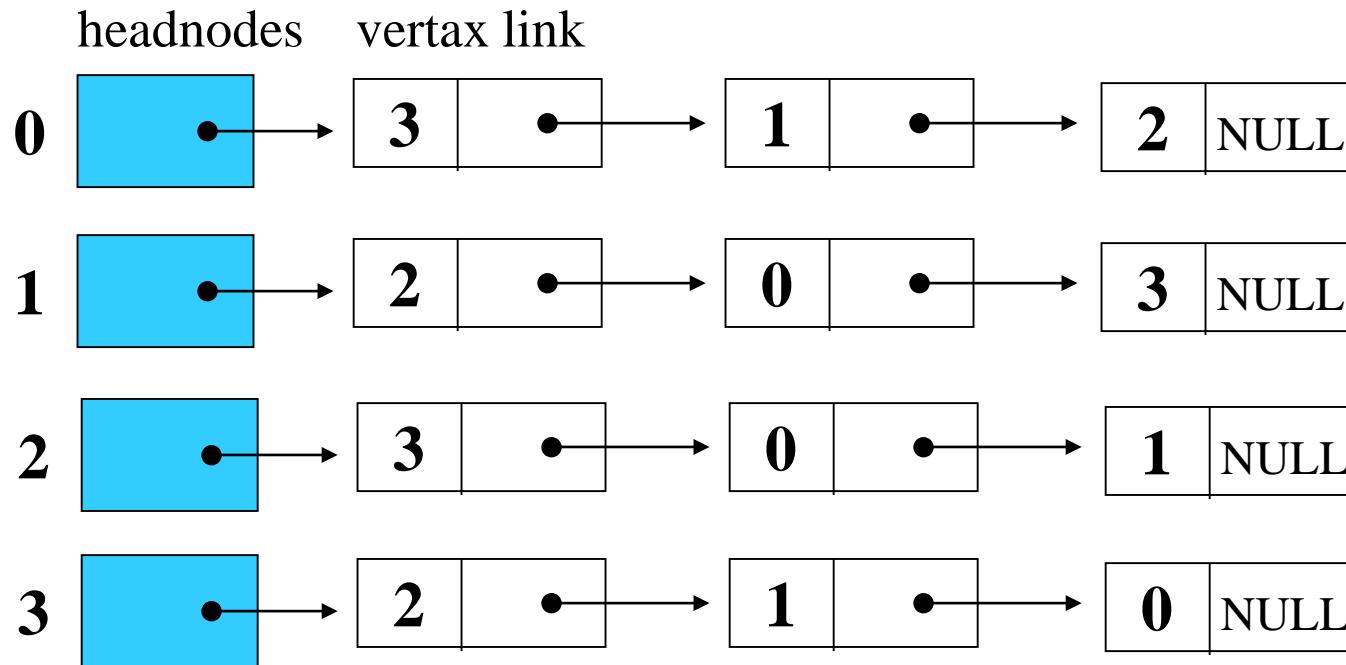


G_4

An undirected graph with n vertices and e edges $\Rightarrow n$ head nodes and $2e$ list nodes

Alternate order adjacency list for G_1

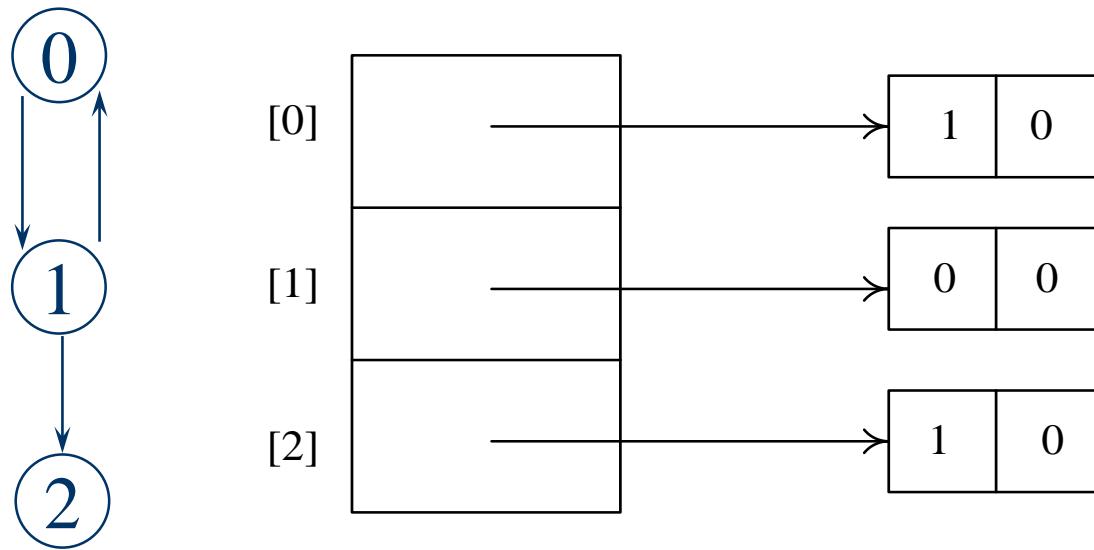
Order is of no significance.



Interesting Operations

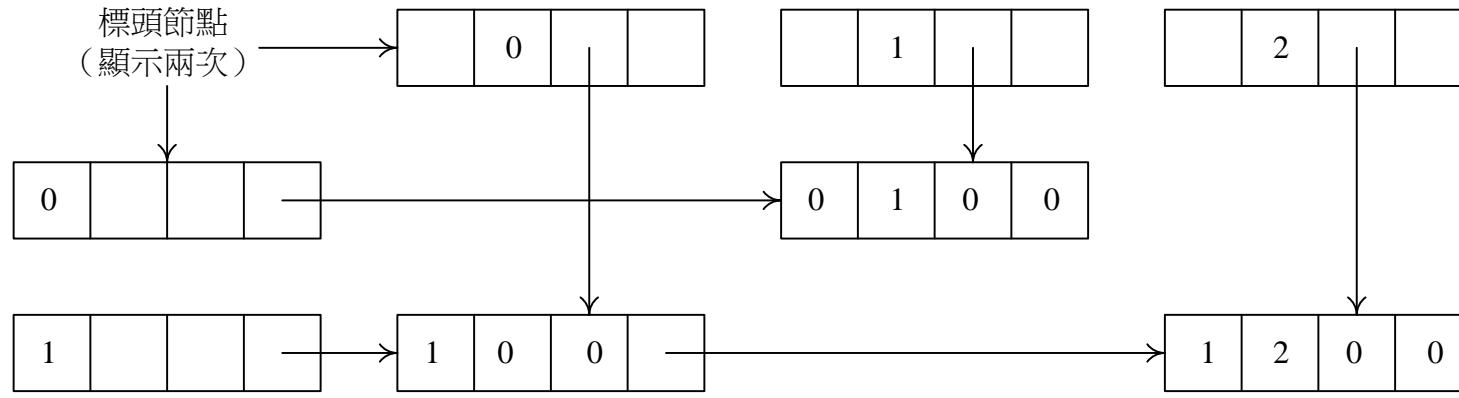
- **degree of a vertex** in an undirected graph
 - # of nodes in adjacency list
- **# of edges** in a graph
 - determined in $O(n+e)$
- **out-degree** of a vertex in a directed graph
 - # of nodes in its adjacency list
- **in-degree** of a vertex in a directed graph
 - traverse the whole data structure

Figure 6.10: Inverse adjacency list for G_3



Determine in-degree of a vertex in a fast way.

Figure 6.11: Orthogonal representation for graph



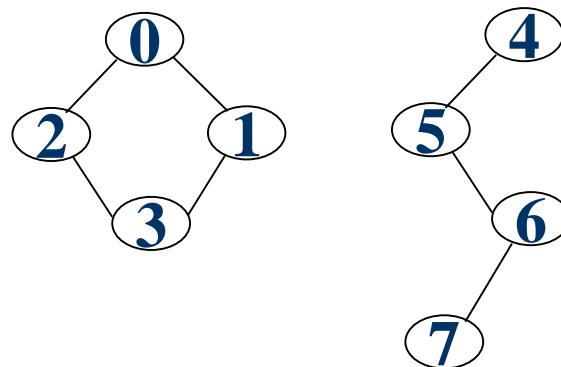
2			0
---	--	--	---

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

row	col	column link for head	row link for tail



Compact Representation



node[0] ... node[n-1]: starting point for vertices

node[n]: $n+2e+1$

node[n+1] ... node[n+2e]: head node of edge

[0]	9		[8]	23		[16]	2	
[1]	11	0	[9]	1	4	[17]	5	
[2]	13		[10]	2	5	[18]	4	
[3]	15	1	[11]	0		[19]	6	
[4]	17		[12]	3	6	[20]	5	
[5]	18	2	[13]	0		[21]	7	
[6]	20		[14]	3	7	[22]	6	
[7]	22	3	[15]	1				

Adjacency Multilists

- An edge in an undirected graph is represented by two nodes in adjacency list representation.
- Adjacency Multilists
 - nodes may be shared among several lists.
(a node is shared by two different paths)

marked	vertex1	vertex2	path1	path2
--------	---------	---------	-------	-------

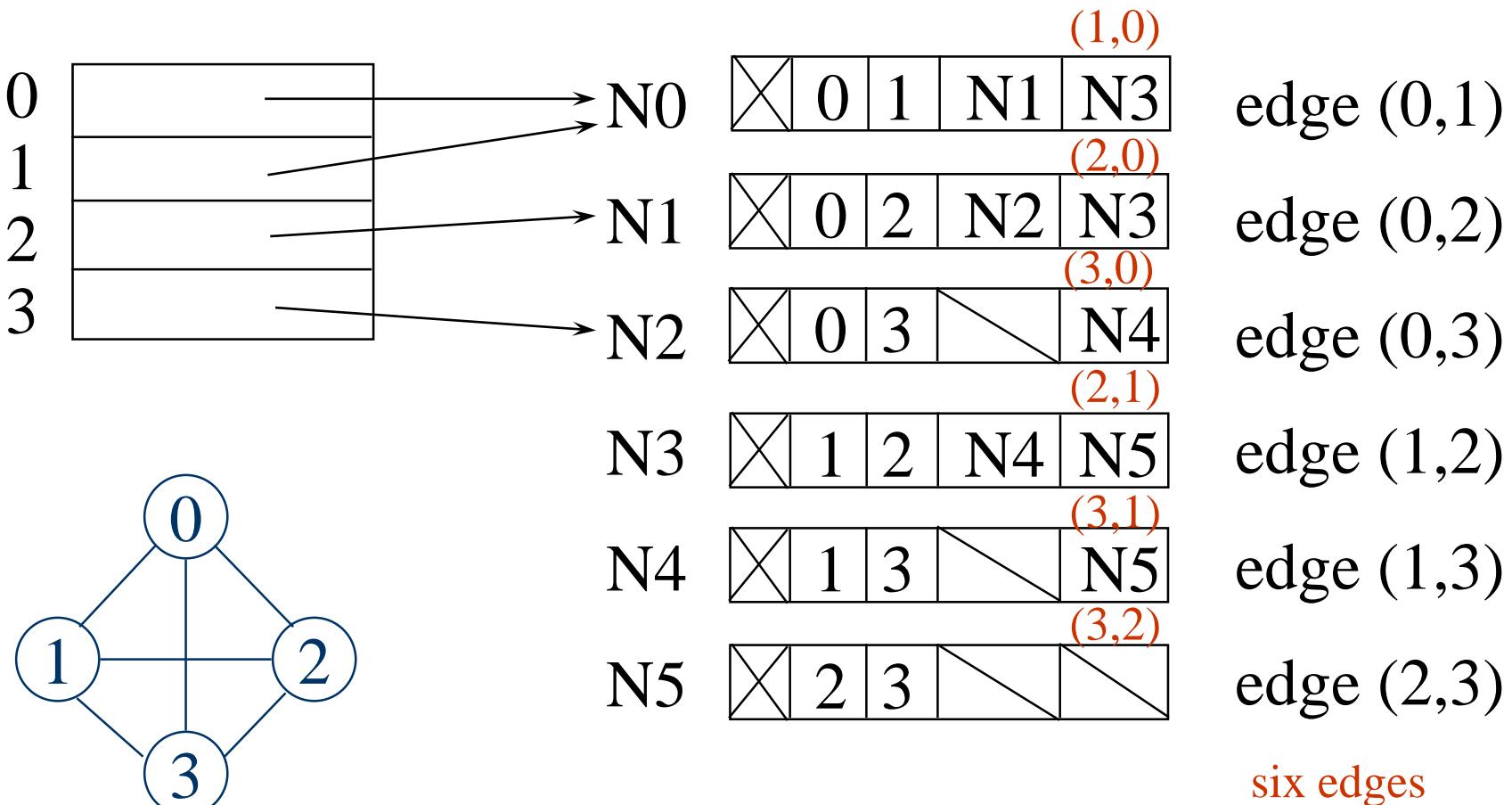
Adjacency Multilists

```
typedef struct edge *edge_pointer;
typedef struct edge {
    short int marked;
    int vertex1, vertex2;
    edge_pointer path1, path2;
} ;
edge_pointer graph[MAX_VERTICES];
```

marked	vertex1	vertex2	path1	path2
--------	---------	---------	-------	-------

Example for Adjacency Multilists

Lists: vertex 0: N0->N1->N2, vertex 1: N0->N3->N4
vertex 2: N1->N3->N5, vertex 3: N2->N4->N5



Some Graph Operations

□ Traversal

Given $G=(V,E)$ and vertex v , find all $w \in V$, such that w connects v .

- Depth First Search (DFS)
preorder tree traversal
- Breadth First Search (BFS)
level order tree traversal

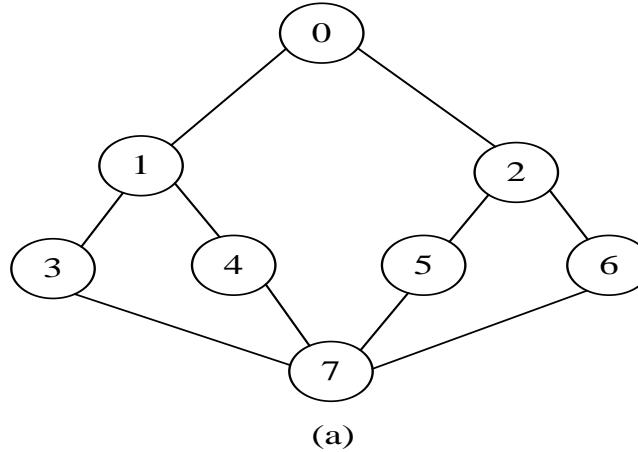
□ Connected Components

□ Spanning Trees

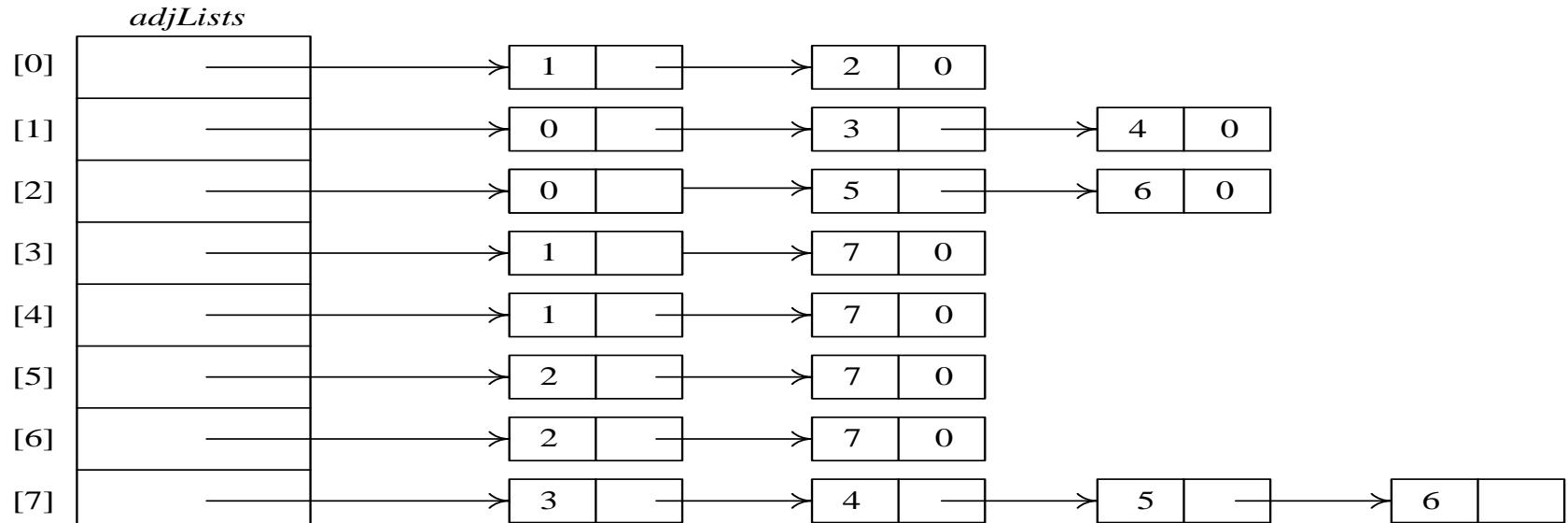
*Figure 6.16: Graph G and its adjacency lists

depth first search: v0, v1, v3, v7, v4, v5, v2, v6

breadth first search: v0, v1, v2, v3, v4, v5, v6, v7



(a)



(b)

Depth First Search

```
#define FALSE 0
#define TRUE 1
short int visited[MAX_VERTICES];
```

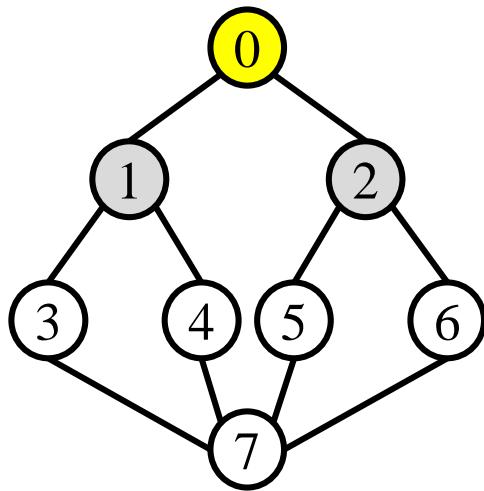
```
void dfs(int v)
{
    node_pointer w;
    visited[v] = TRUE;
    printf("%5d", v);
    for (w=graph[v]; w; w=w->link)
        if (!visited[w->vertex])
            dfs(w->vertex);
}
```

CHAPTER 6

Data structure
adjacency list: $O(e)$
adjacency matrix³³: $O(n^2)$

DFS and BFS

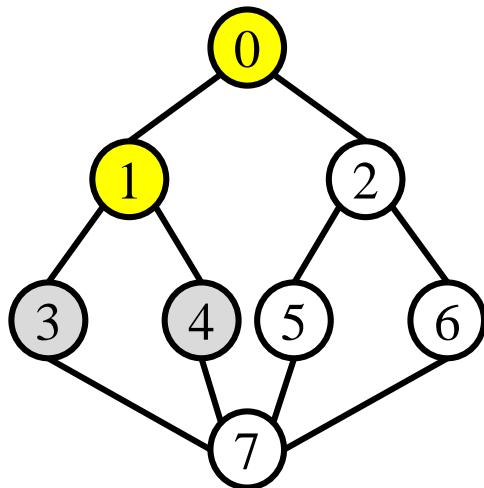
□ DFS



從頂點0開始
有頂點1, 2可以選
選數字小的頂點1

DFS and BFS

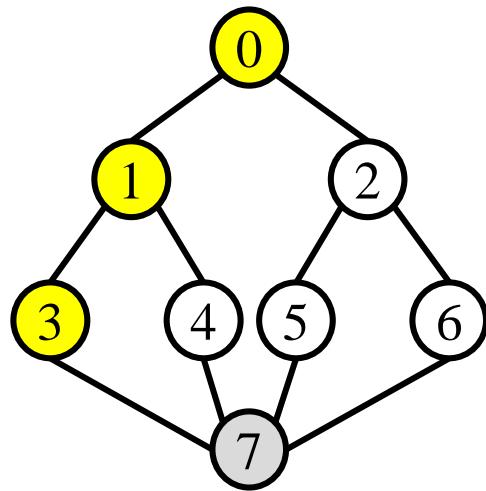
□ DFS



頂點1有頂點3, 4可以選
選數字小的頂點3

DFS and BFS

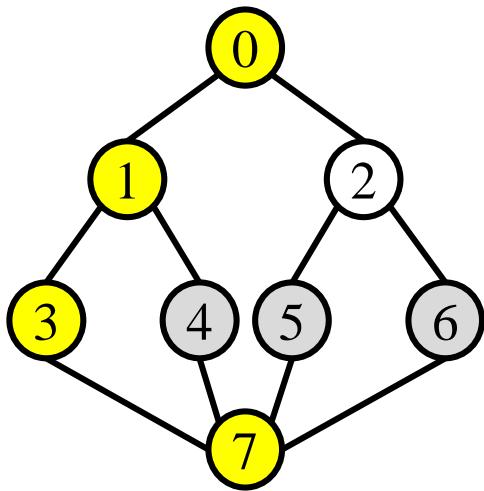
□ DFS



頂點3有頂點1, 7可以選
1選過了，所以選頂點7

DFS and BFS

□ DFS



頂點7有頂點3, 4, 5, 6可以選

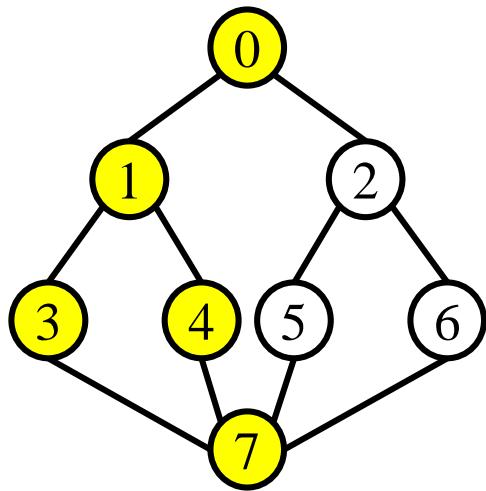
3選過了

有頂點4, 5, 6可以選

選數字小的頂點4

DFS and BFS

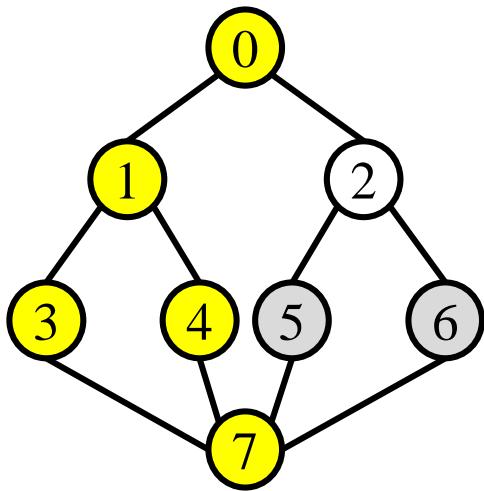
□ DFS



頂點4有頂點1, 7可以選
1, 7都選過了
回上一個頂點，也就是頂
點7

DFS and BFS

□ DFS



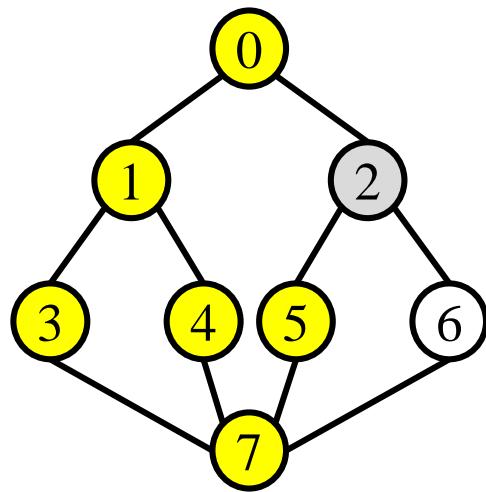
頂點7有頂點3, 4, 5, 6可以選

3, 4選過了

有頂點 5, 6可以選
選數字小的頂點5

DFS and BFS

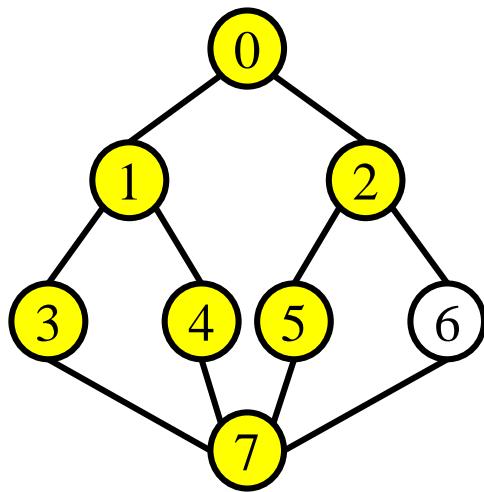
□ DFS



頂點5有頂點2, 7可以選
7選過了，所以選頂點2

DFS and BFS

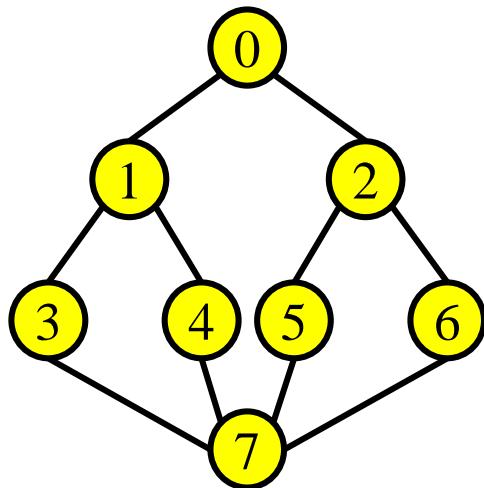
□ DFS



頂點2有頂點0, 5, 6可以選
0, 5選過了，所以選頂點6

DFS and BFS

□ DFS



Breadth First Search

```
typedef struct queue *queue_pointer;
typedef struct queue {
    int vertex;
    queue_pointer link;
} ;
void addq(int);
int deleteq();
```

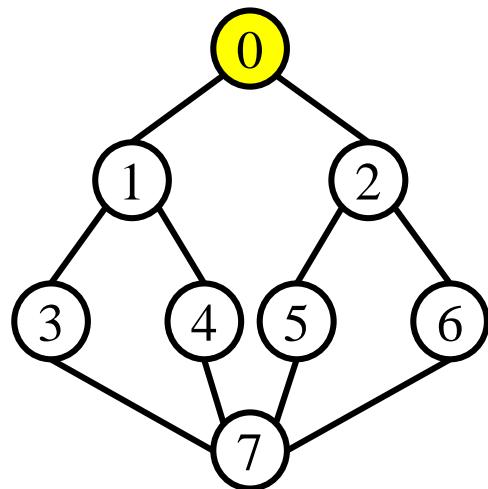
Breadth First Search (*Continued*)

```
void bfs(int v)
{
    node_pointer w;
    queue_pointer front, rear;
    front = rear = NULL;
    printf("%5d", v);                                adjacency list: O(e)
    visited[v] = TRUE;                               adjacency matrix: O(n2)
    addq(v);
```

```
while (front) {  
    v= deleteq( );  
    for (w=graph[v]; w; w=w->link)  
        if (!visited[w->vertex]) {  
            printf("%5d", w->vertex);  
            addq(w->vertex);  
            visited[w->vertex] = TRUE;  
        } /* unvisited vertices */  
    }  
}
```

DFS and BFS

□ BFS

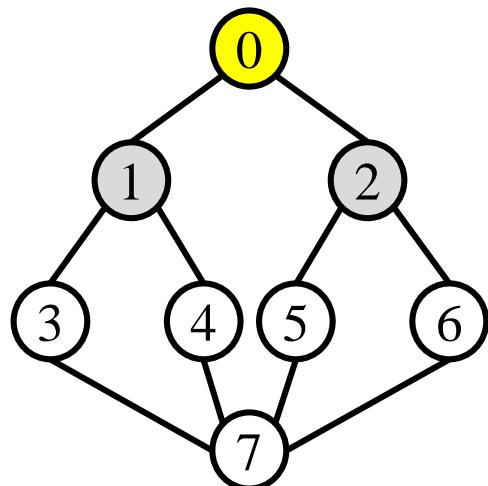


Queue

0

DFS and BFS

□ 步驟1



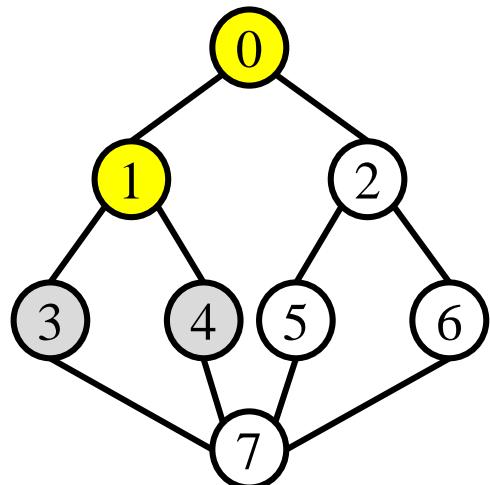
Queue

0 1 2

輸出：

DFS and BFS

□ 步驟2



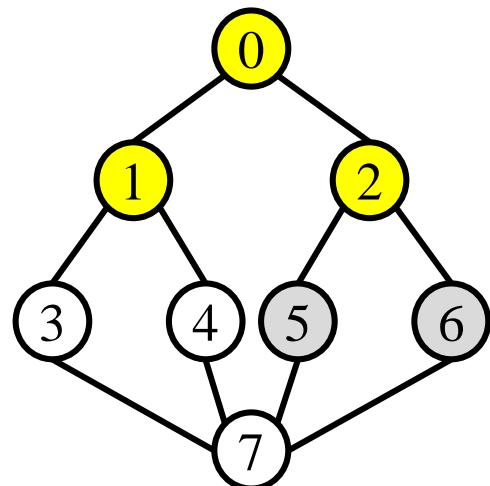
Queue

1 2 3 4

輸出: 0

DFS and BFS

□ 步驟3



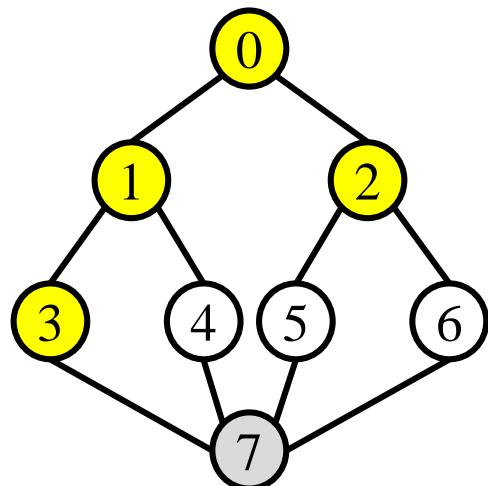
Queue

2 3 4 5 6

輸出: 0 1

DFS and BFS

□ 步驟4



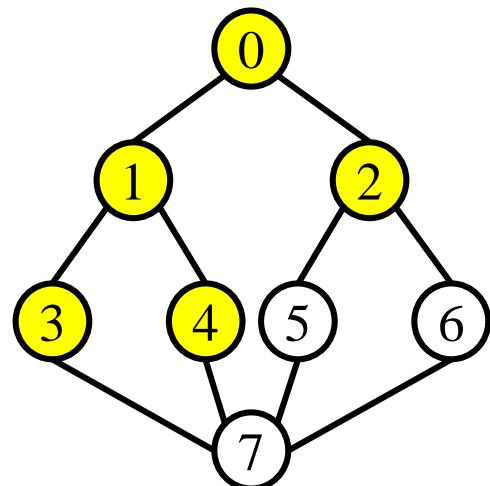
Queue



輸出: 0 1 2

DFS and BFS

□ 步驟5



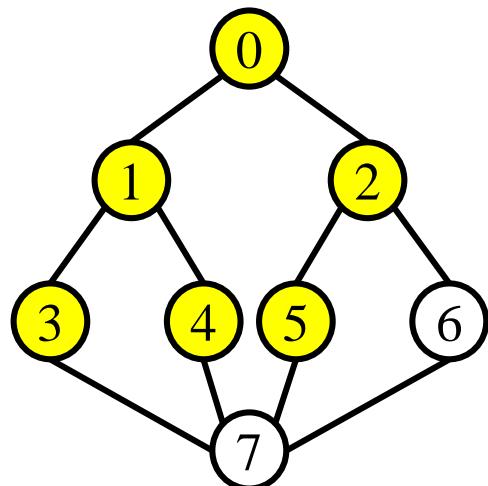
Queue

4 5 6 7

輸出: 0 1 2 3

DFS and BFS

□ 步驟6



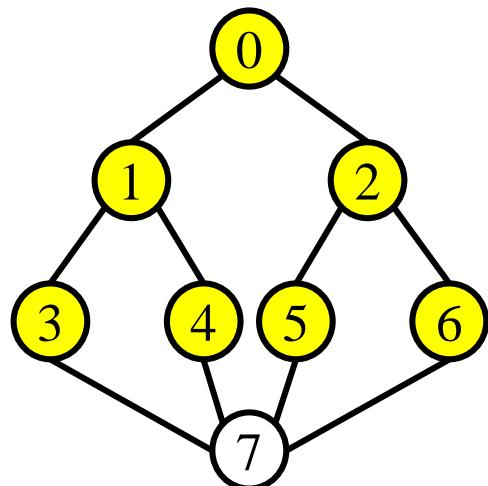
Queue

5 6 7

輸出: 0 1 2 3 4

DFS and BFS

□ 步驟7



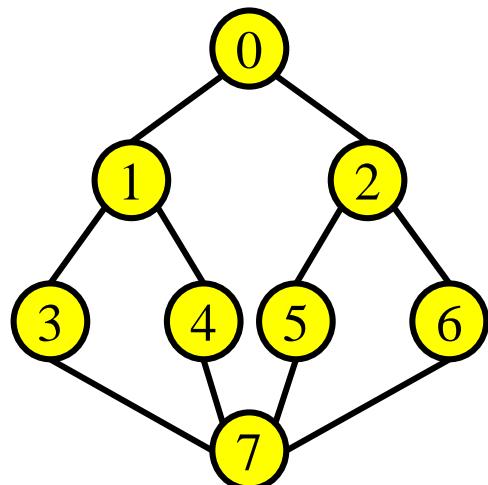
Queue

6 7

輸出: 0 1 2 3 4 5

DFS and BFS

□ 步驟8



Queue

7

輸出: 0 1 2 3 4 5 6

Connected Components

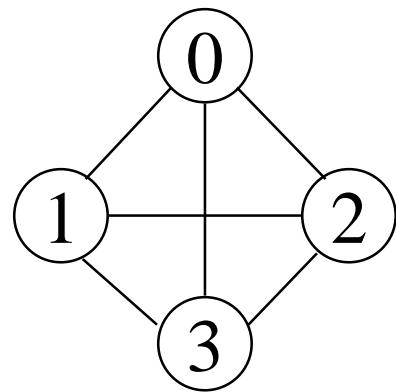
```
void connected(void)
{ /*determine the connected components of
a graph */
    for (i=0; i<n; i++) {
        if (!visited[i]) {
            dfs(i); //dfs→O(n)
            printf("\n");
        }
    }
}
```

adjacency list: $O(n+e)$
adjacency matrix: $O(n^2)$

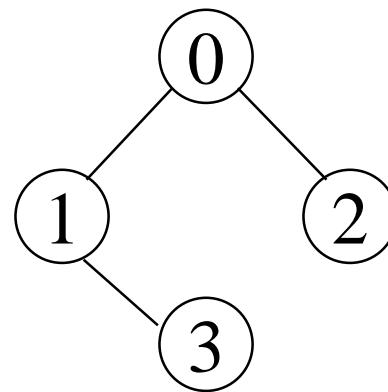
Spanning Trees

- When graph G is connected, a depth first or breadth first search starting at any vertex will visit all vertices in G
- A **spanning tree** is any tree that consists solely of edges in G and that includes all the vertices
- $E(G) = T$ (**tree edges**) + N (**nontree edges**)
where T : set of edges used during search
 N : set of remaining edges

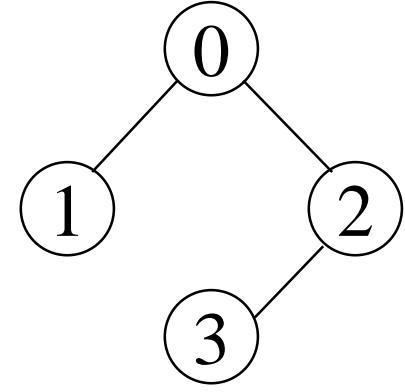
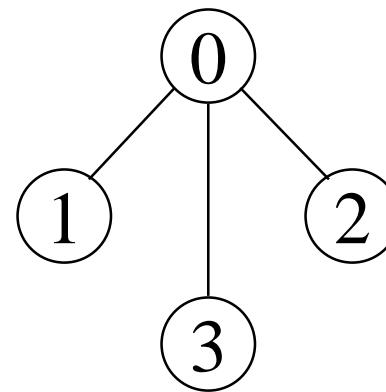
Examples of Spanning Tree



G_1



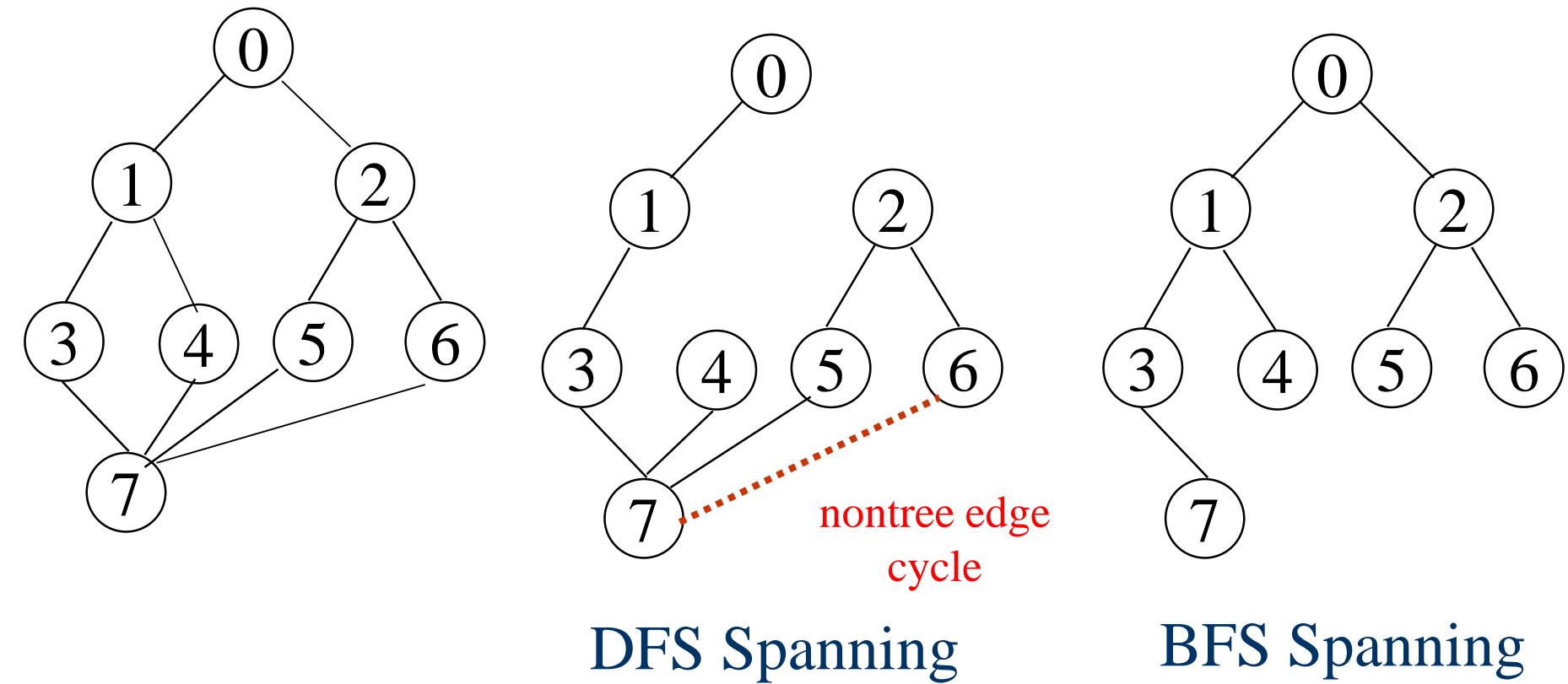
Possible spanning trees



Spanning Trees

- Either dfs or bfs can be used to create a spanning tree
 - When *dfs* is used, the resulting spanning tree is known as a **depth first spanning tree**
 - When *bfs* is used, the resulting spanning tree is known as a **breadth first spanning tree**
- While adding a nontree edge into any spanning tree, this will create a cycle

DFS vs BFS Spanning Tree



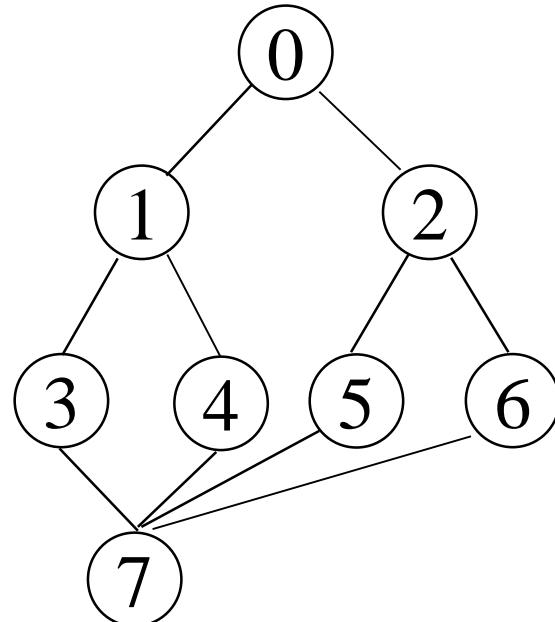
DFS Spanning

BFS Spanning

A spanning tree is a **minimal subgraph**, G' , of G such that $V(G')=V(G)$ and G' is connected.

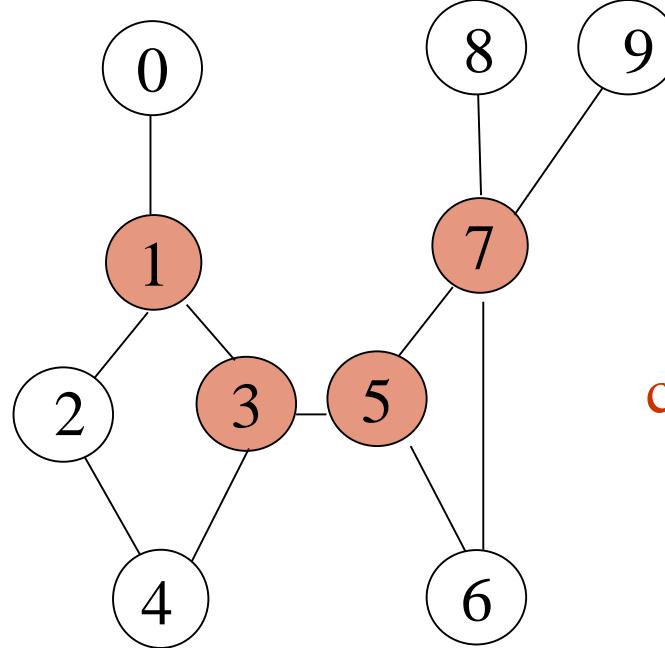
Any connected graph with n vertices must have at least $n-1$ edges.

A **biconnected graph** is a connected graph that has no **articulation points**.



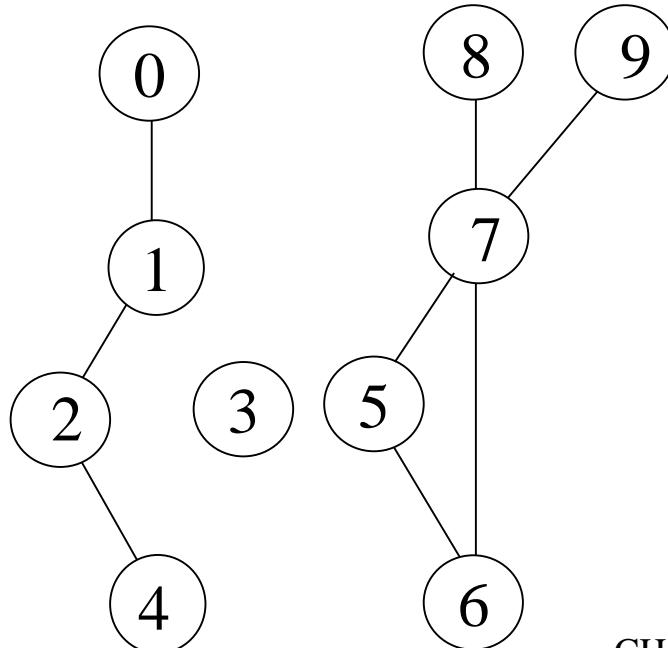
Articulation points

關節點是讓一張無向圖維持連通，不可或缺的點。只要從一張無向圖上移除了關節點（以及與之相連的邊），就會讓這張圖分離成更多部分，呈現不連通的狀態。

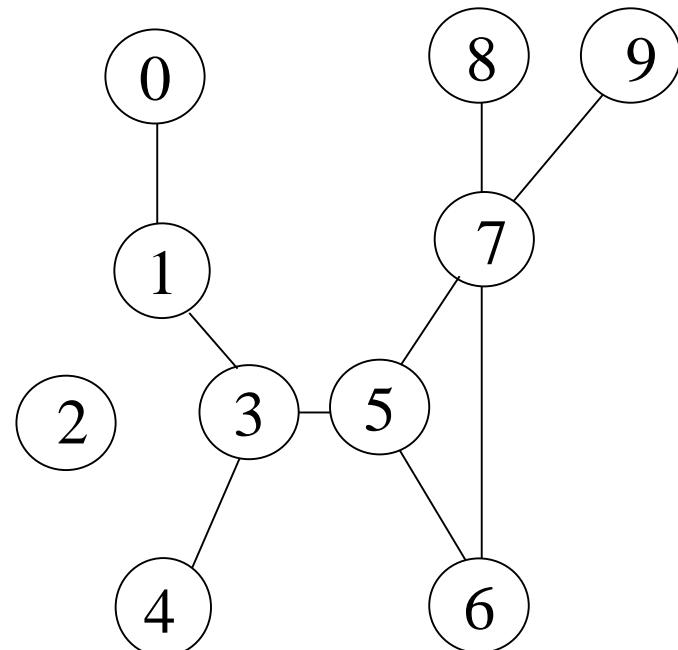


connected graph

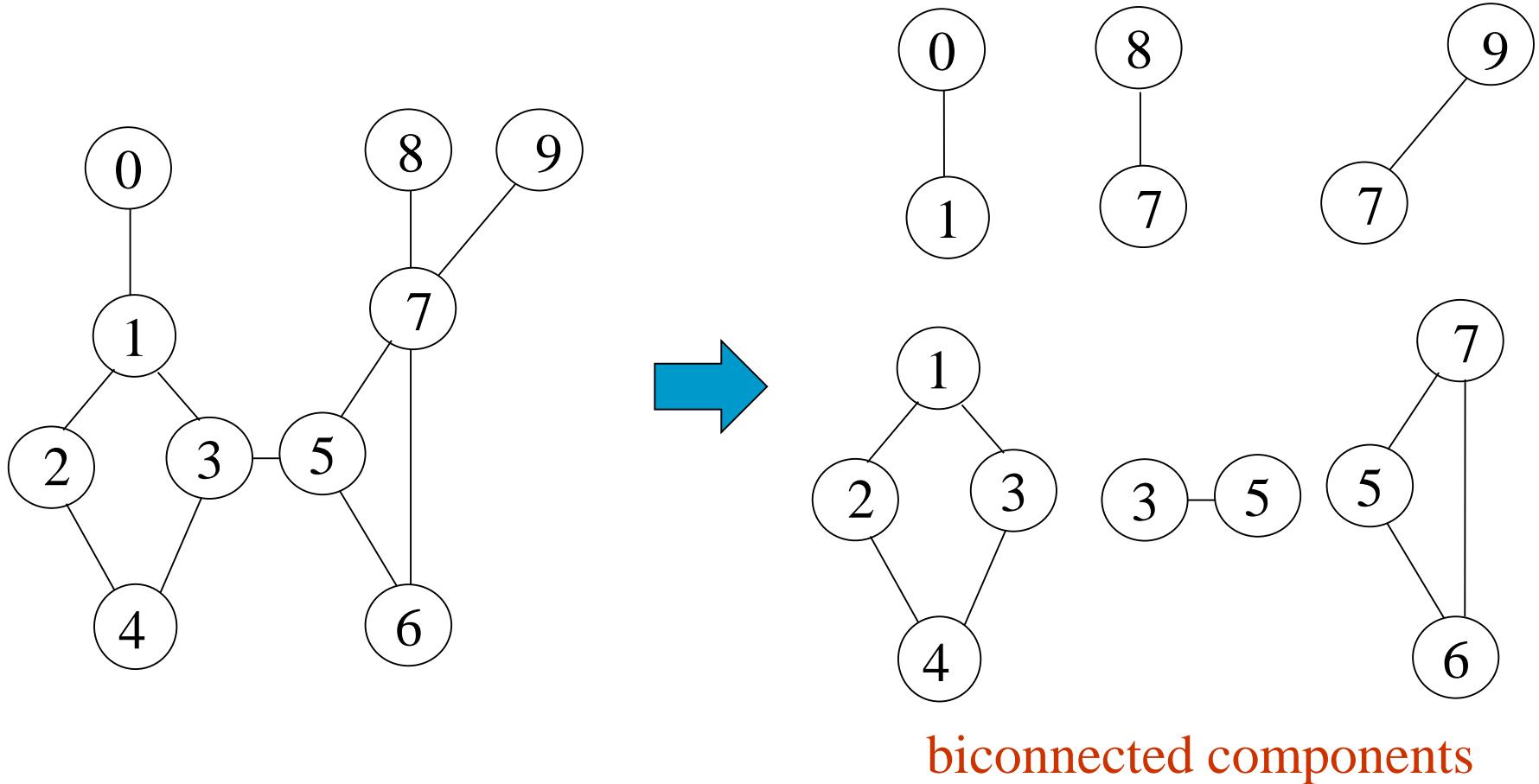
two connected components



one connected graph

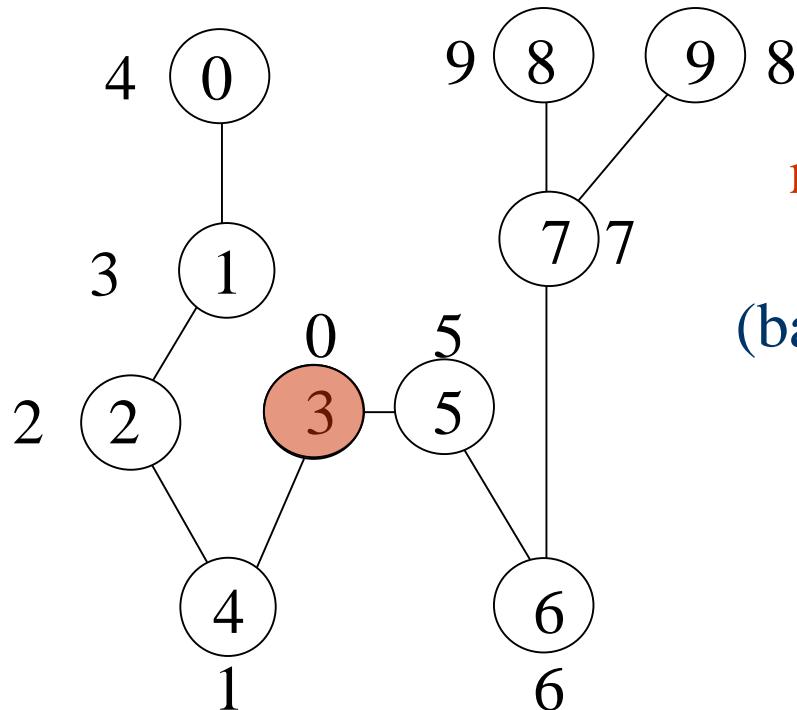


biconnected component: a maximal connected subgraph H
(no subgraph that is both biconnected and properly contains H)

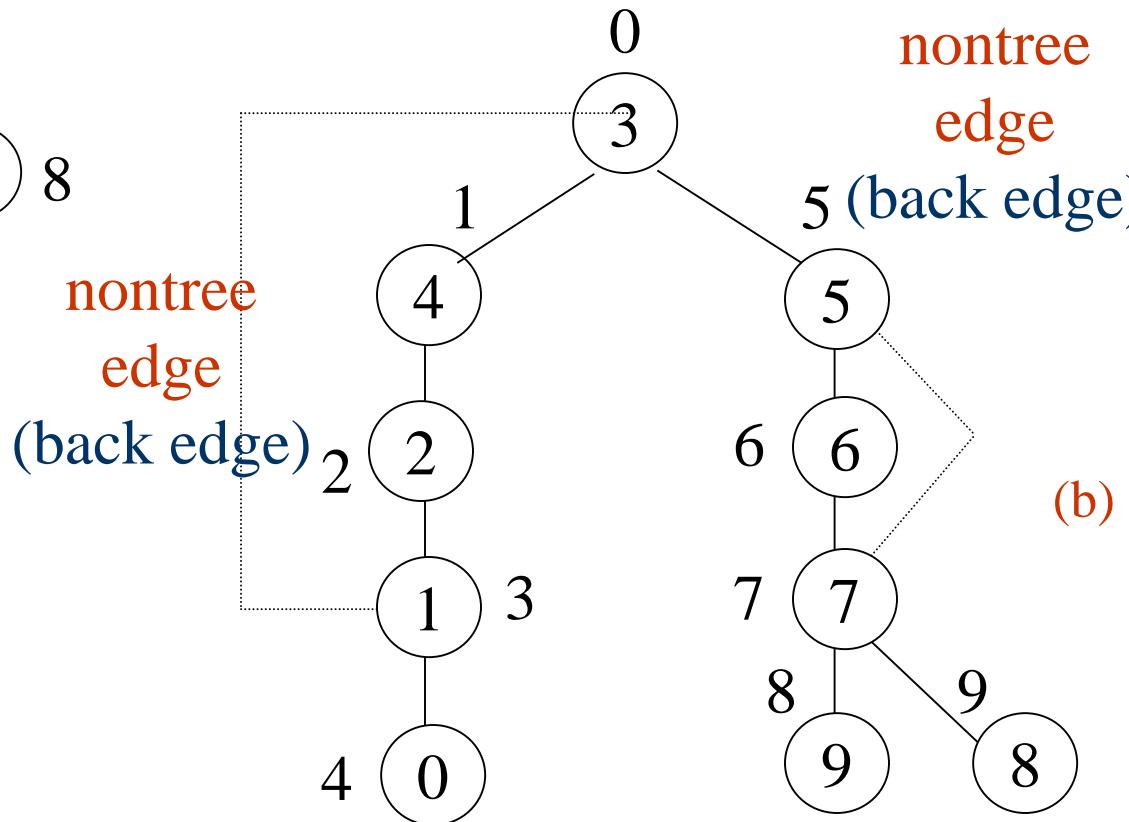


Find biconnected component of a connected undirected graph by depth first spanning tree

depth first number (dfn)



(a) depth first spanning tree



Any other vertex u is an articulation point iff it has at least one child w such that we cannot reach an ancestor of u using a path

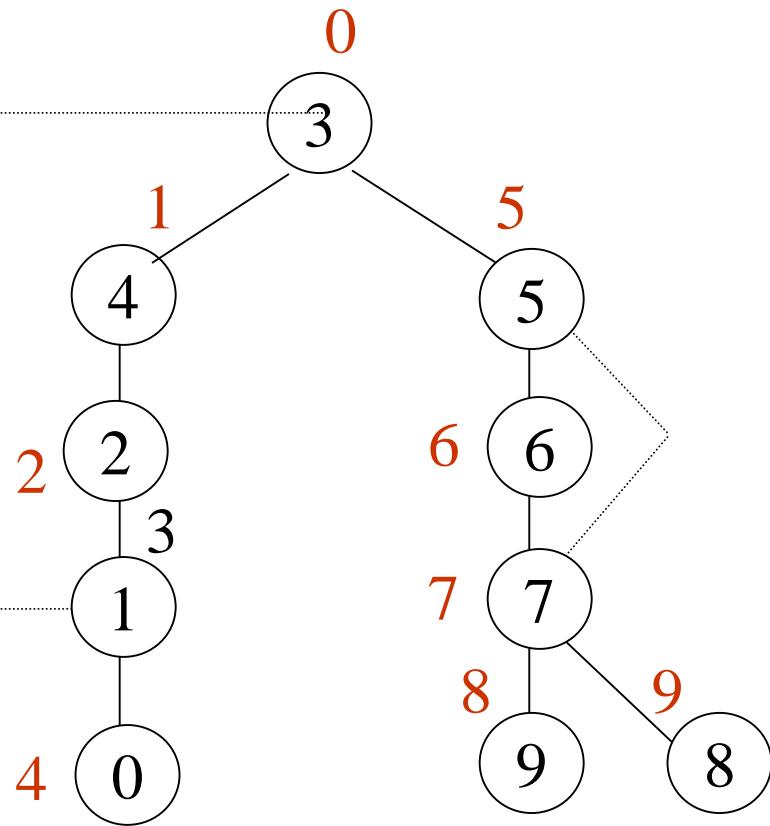
If u is an ancestor of v then $\text{dfn}(u) < \text{dfn}(v)$.

***Figure 6.21:** dfn and low values for dfs spanning tree with $root = 3$

Vertax	0	1	2	3	4	5	6	7	8	9
dfn	4	3	2	0	1	5	6	7	9	8
low	4	0	0	0	0	5	5	5	9	8

$low(u) = \min\{ dfn(u), \min\{ low(w) | w \text{ is a child of } u \}, \min\{ dfn(w) | (u, w) \text{ is a back edge} \}$

u: articulation point
 $low(\text{child}) \geq dfn(u)$



*The root of a depth first spanning tree is an **articulation point** iff it has **at least two children**.

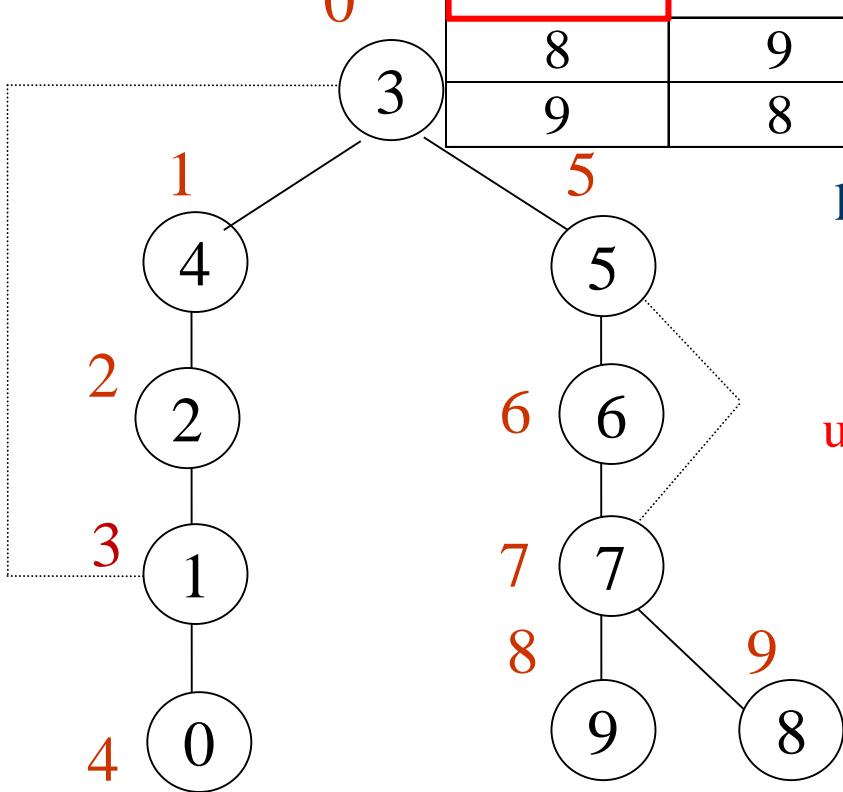
*Any other vertex u is an **articulation point** iff it has at least one child w such that we cannot reach an ancestor of u using a path that consists of

- (1) only w ;
- (2) descendants of w ;
- (3) single back edge.

$$\text{low}(u) = \min \{ \text{dfn}(u), \min \{ \text{low}(w) | w \text{ is a child of } u \}, \min \{ \text{dfn}(w) | (u, w) \text{ is a back edge} \} \}$$

u : articulation point $\rightarrow \text{low}(\text{child}) \geq \text{dfn}(u)$

vertex	dfn	low	child	low (child)	low:dfn
0	4	4 (4,n,n)	null	null	null:4
1	3	0 (3,4,0)	0	4	$4 \geq 3$ •
2	2	0 (2,0,n)	1	0	$0 < 2$
3	0	0 (0,0,n)	4,5	0,5	$0,5 \geq 0$ •
4	1	0 (1,0,n)	2	0	$0 < 1$
5	5	5 (5,5,n)	6	5	$5 \geq 5$ •
6	6	5 (6,5,n)	7	5	$5 < 6$
7	7	5 (7,8,5)	8,9	9,8	$9,8 \geq 7$ •
8	9	9 (9,n,n)	null	null	null, 9
9	8	8 (8,n,n)	null	null	null, 8



$\text{low}(u) = \min \{ \text{dfn}(u),$
 $\min \{ \text{low}(w) | w \text{ is a child of } u \},$
 $\min \{ \text{dfn}(w) | (u,w) \text{ is a back edge} \}$

u: articulation point → $\text{low}(\text{child}) \geq \text{dfn}(u)$

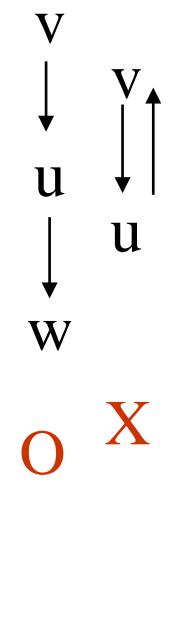
```
#define MIN2(x,y) ( (x)<(y) ? (x):(y) )
```

```
void init(void)
{
    int i;
    for (i = 0; i < n; i++) {
        visited[i] = FALSE;
        dfn[i] = low[i] = -1;
    }
    num = 0;
}
```

***Program 6.5:** Initializaiton of *dfn* and *low*

*Program 6.4: Determining dfn and low num has been initialized to 0

```
void dfnlow(int u, int v)          Initial call: dfn(x,-1)
{
/* compute dfn and low while performing a dfs search
beginning at vertex u, v is the parent of u (if any) */
    node_pointer ptr;
    int w;
    dfn[u] = low[u] = num++;      low[u]=min{dfn(u), ... }
    for (ptr = graph[u]; ptr; ptr = ptr ->link) {
        w = ptr ->vertex;
        if (dfn[w] < 0) { /*w is an unvisited vertex */
            dfnlow(w, u);
            low[u] = MIN2(low[u], low[w]);
        }      low[u]=min{..., min{low(w)|w is a child of u}, ... }
        else if (w != v) dfn[w]≠0 非第一次，表示藉back edge
            low[u] =MIN2(low[u], dfn[w] );
    }
}                                low[u]=min{..., ..., min{dfn(w)|(u,w) is a back edge}
```



*Program 6.6: Biconnected components of a graph

```
void bicon(int u, int v)
{
/* compute dfn and low, and output the edges of G by their
   biconnected components , v is the parent ( if any) of the u
   (if any) in the resulting spanning tree. It is assumed that all
   entries of dfn[ ] have been initialized to -1, num has been
   initialized to 0, and the stack has been set to empty */
node_pointer ptr;
int w, x, y;
dfn[u] = low[u] = num++;
for (ptr = graph[u]; ptr; ptr = ptr->link) {
    w = ptr ->vertex;
    if ( v != w && dfn[w] < dfn[u] ) { (1) dfn[w]=-1 第一次
        push(u, w);      /* add edge to stack */ (2) dfn[w]!=-1 非第一次，藉back
                           edge
                           } 
```

```

if(dfn[w] < 0) /* w has not been visited */
    bicon(w, u); low[u]=min{..., min{low(w)|w is a child of u}, . .
    low[u] = MIN2(low[u], low[w]);
    if (low[w] >= dfn[u] ){      articulation point
        printf("New biconnected component: ");
        do { /* delete edge from stack */
            pop(&x, &y);
            printf("<%d, %d>", x, y);
        } while (!(x == u) && (y == w)));
        printf("\n");
    }
}
else if (w != v) low[u] = MIN2(low[u], dfn[w]);
}   low[u]=min{..., ..., min{dfn(w)|(u,w) is a back edge} }
}

```

Minimum Cost Spanning Tree

- The cost of a spanning tree of a weighted undirected graph is the sum of the costs of the edges in the spanning tree
- A minimum cost spanning tree is a spanning tree of least cost
- Three different algorithms can be used
 - Kruskal
 - Prim
 - Sollin

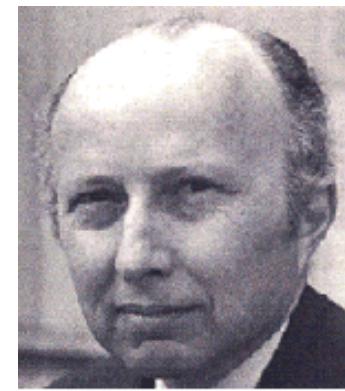
Select $n-1$ edges from a weighted graph of n vertices with minimum cost.

Greedy Strategy

- An optimal solution is constructed in stages
- At each stage, the best decision is made at this time
- Since this decision cannot be changed later,
we make sure that the decision will result in a feasible solution
- Typically, the selection of an item at each stage is based on a least cost or a highest profit criterion

Kruskal's Idea

- Build a minimum cost spanning tree T by adding edges to T one at a time
- Select the edges for inclusion in T in nondecreasing order of the cost
- An edge is added to T if it does not form a cycle
- Since G is connected and has $n > 0$ vertices, exactly $n-1$ edges will be selected



Examples for Kruskal's Algorithm

1 $0 \xrightarrow{10} 5$

2 $2 \xrightarrow{12} 3$

3 $1 \xrightarrow{14} 6$

4 $1 \xrightarrow{16} 2$

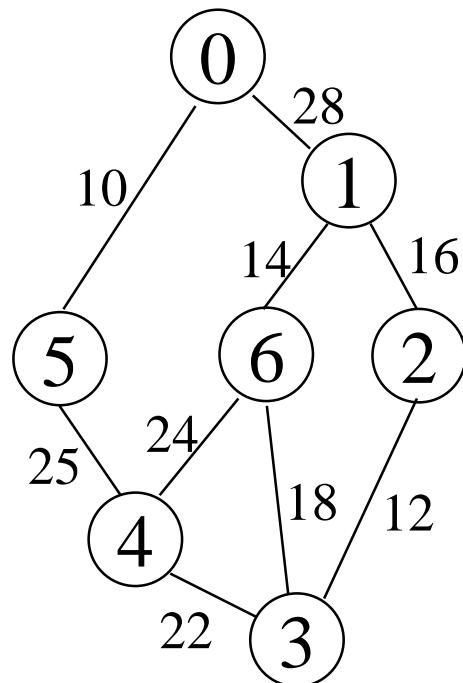
5 $3 \xrightarrow{18} 6$

6 $3 \xrightarrow{22} 4$

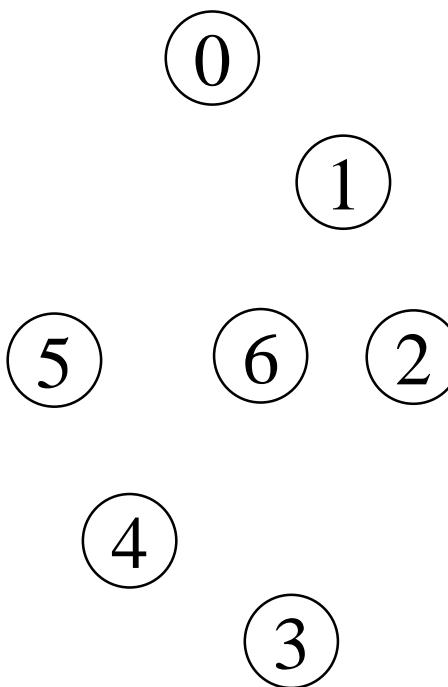
7 $4 \xrightarrow{24} 6$

8 $4 \xrightarrow{25} 5$

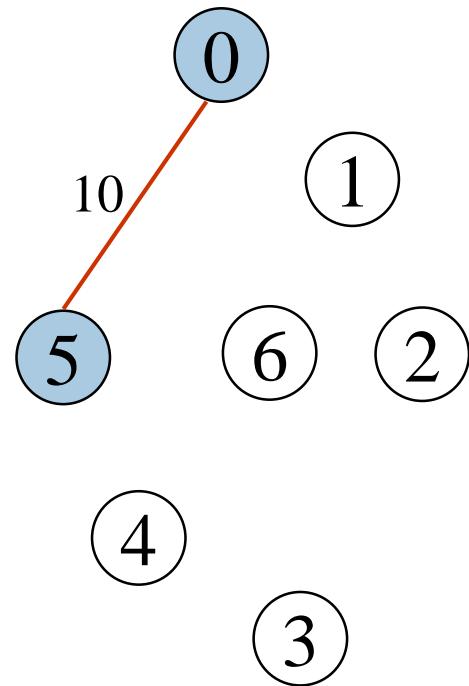
9 $0 \xrightarrow{28} 1$



(a)



(b)



(c)

$$1 \quad 0 \xrightarrow{10} 5$$

$$2 \quad 2 \xrightarrow{12} 3$$

$$3 \quad 1 \xrightarrow{14} 6$$

$$4 \quad 1 \xrightarrow{16} 2$$

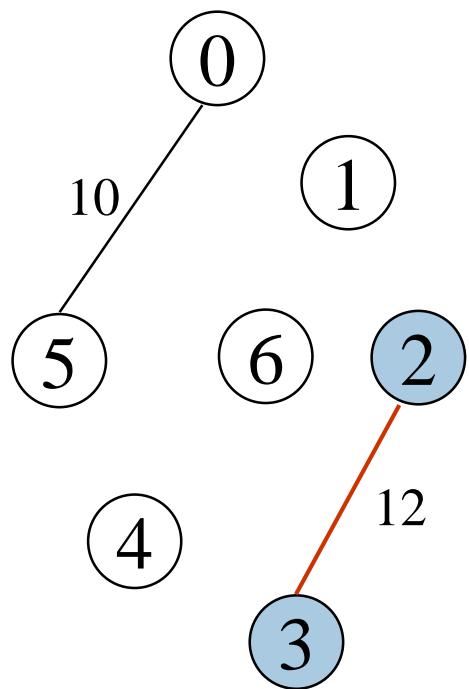
$$5 \quad 3 \xrightarrow{18} 6$$

$$6 \quad 3 \xrightarrow{22} 4$$

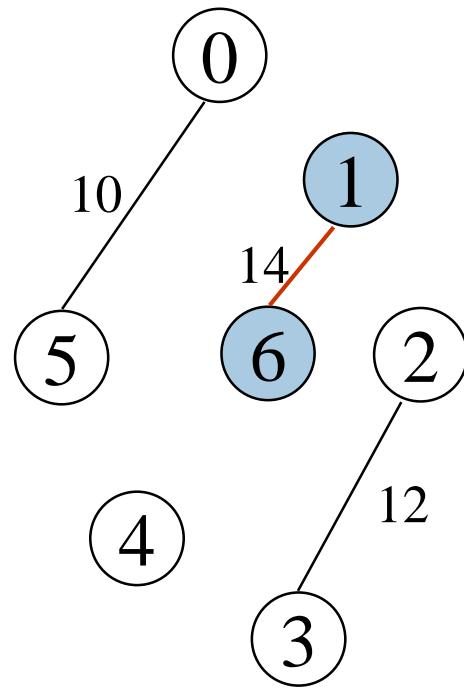
$$7 \quad 4 \xrightarrow{24} 6$$

$$8 \quad 4 \xrightarrow{25} 5$$

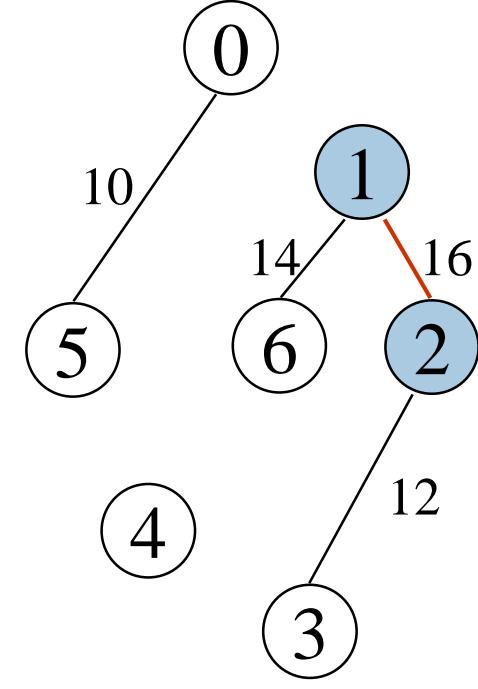
$$9 \quad 0 \xrightarrow{28} 1$$



(d)



(e)



(f)

1 0 $\xrightarrow{10}$ 5

2 2 $\xrightarrow{12}$ 3

3 1 $\xrightarrow{14}$ 6

4 1 $\xrightarrow{16}$ 2

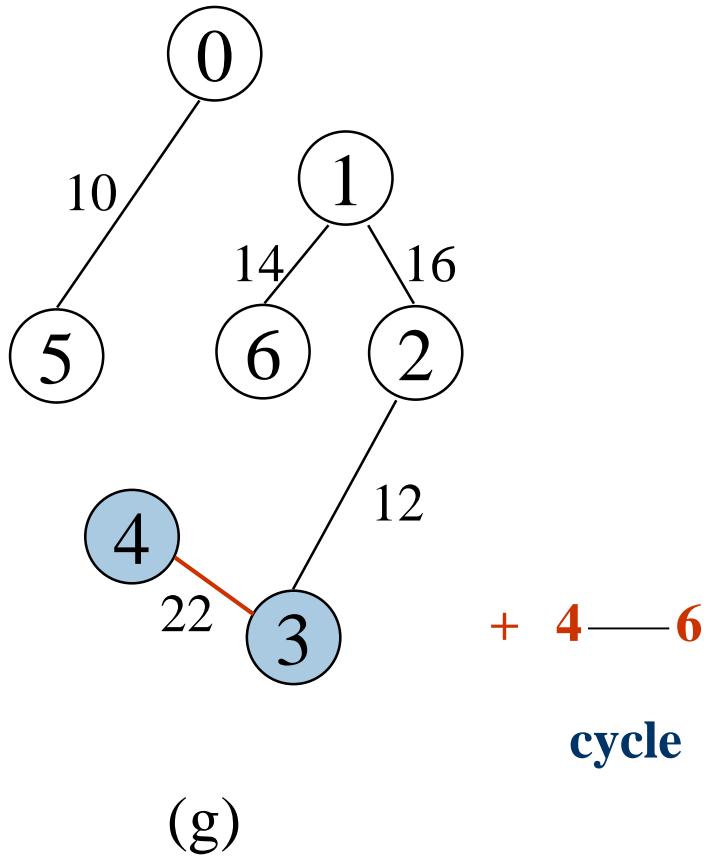
5 3 $\xrightarrow{18}$ 6

6 3 $\xrightarrow{22}$ 4

7 4 $\xrightarrow{24}$ 6

8 4 $\xrightarrow{25}$ 5

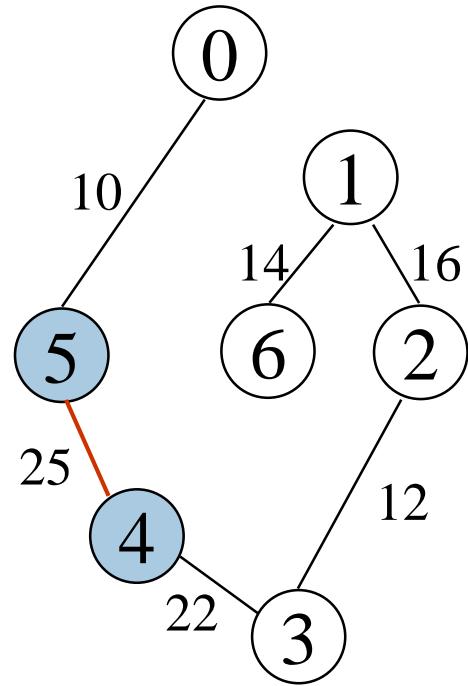
9 0 $\xrightarrow{28}$ 1



$$\text{cost} = 10 + 25 + 22 + 12 + 16 + 14$$

+ 4 $\xrightarrow{ }$ 6

cycle



Kruskal's Algorithm

目標：取出n-1條edges

```
T= {} ;  
while (T contains less than n-1 edges  
      && E is not empty) {  
    choose a least cost edge (v,w) from E;  
    delete (v,w) from E; → min heap construction time O(e)  
    if ((v,w) does not create a cycle in T)  
        add (v,w) to T → choose and delete O(log e)  
    else discard (v,w); → find find & union O(log e)  
}  
if (T contains fewer than n-1 edges)  
    printf("No spanning tree\n");
```

$O(e \log e)$

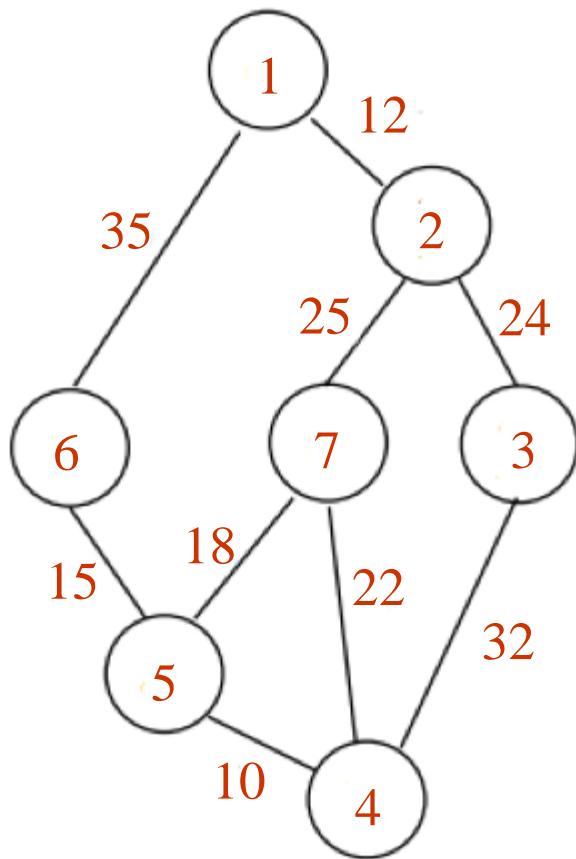
CHAPTER 6

Kruskal's Algorithm

- 取得graph的Adjacency Matrix來實作 Kruskal's Algorithm
- 實作步驟
 1. 取得圖的關係
 2. 每次找出最低cost的邊且不會使圖產生 cycle
 3. 直到產生($\text{頂點數}-1$)個邊的最小花費生成樹

Kruskal's Algorithm

□ 圖解步驟



1 --12--> 2
2 --24--> 3
1 --35--> 6
2 --25--> 7
5 --15--> 6
4 --10--> 5
5 --18--> 7
4 --22--> 7
3 --32--> 4

Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

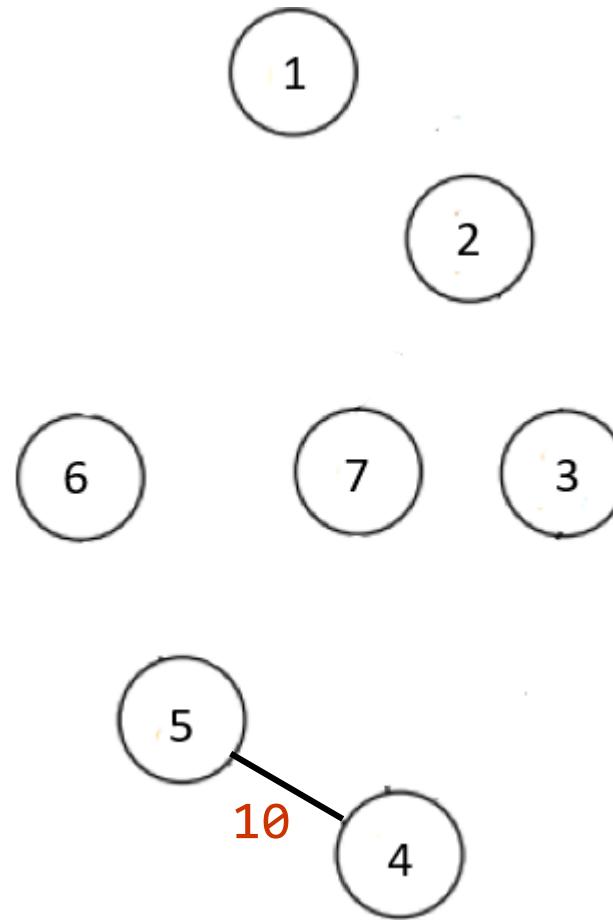
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

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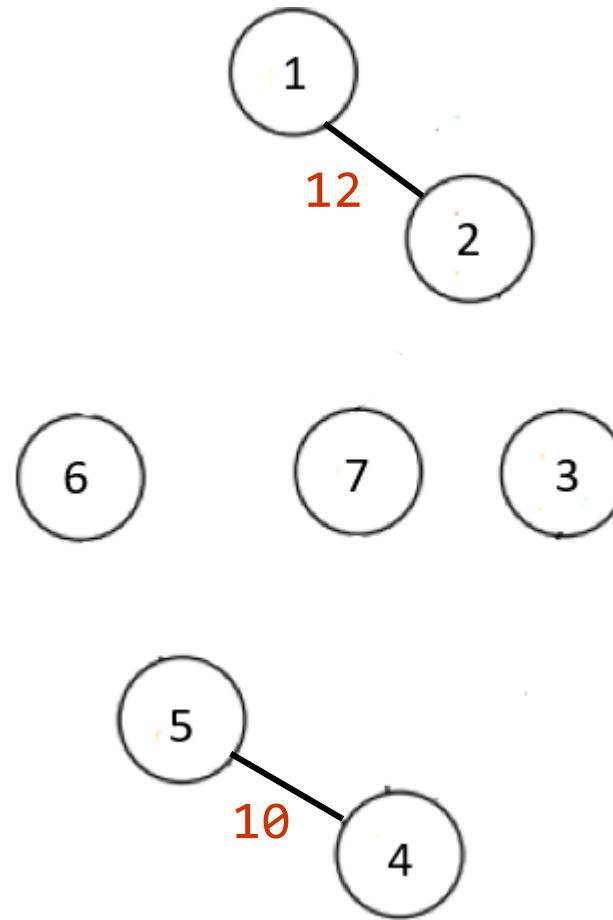
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

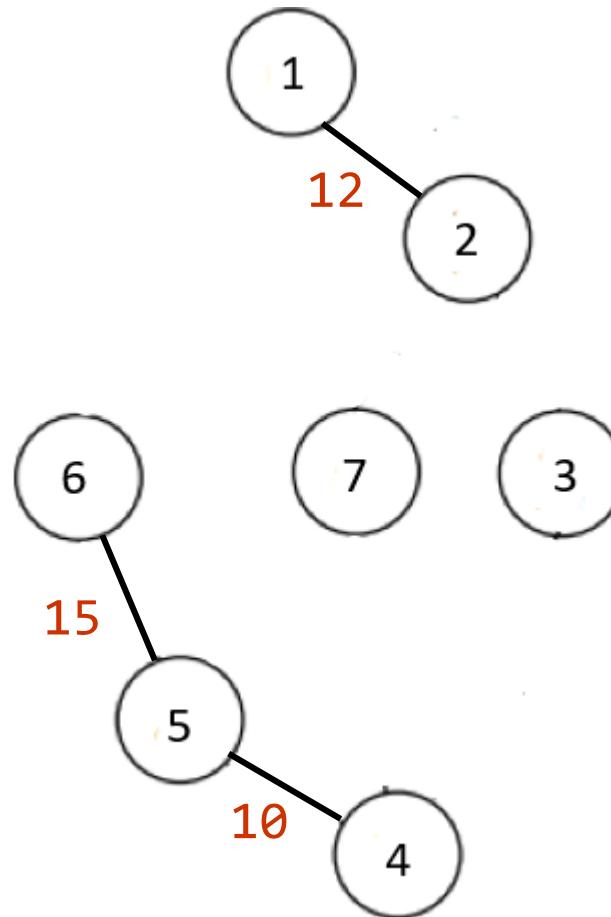
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

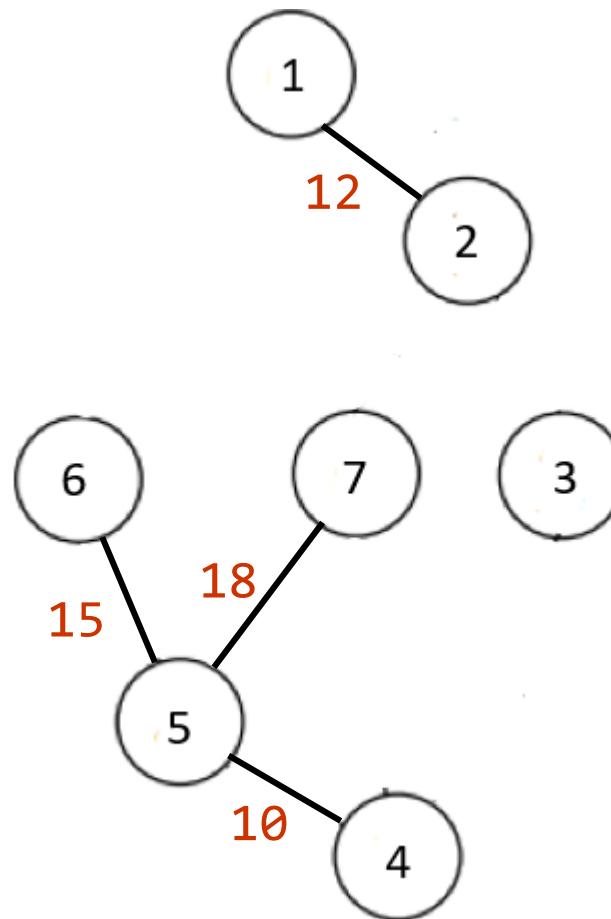
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

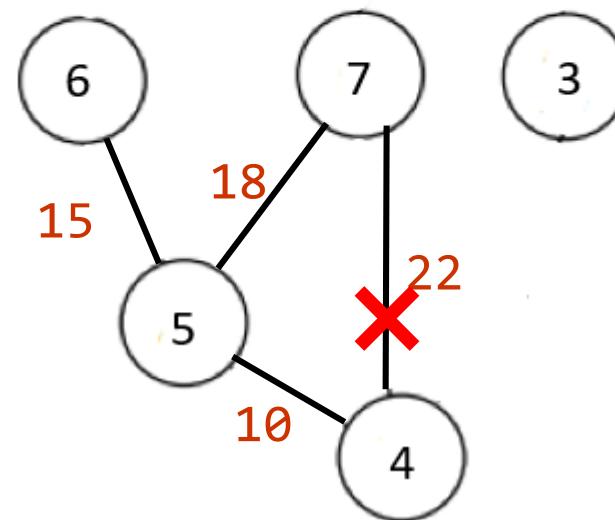
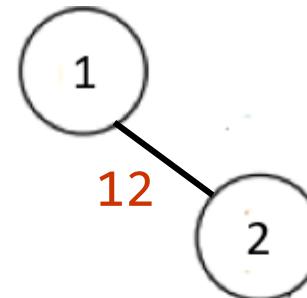
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

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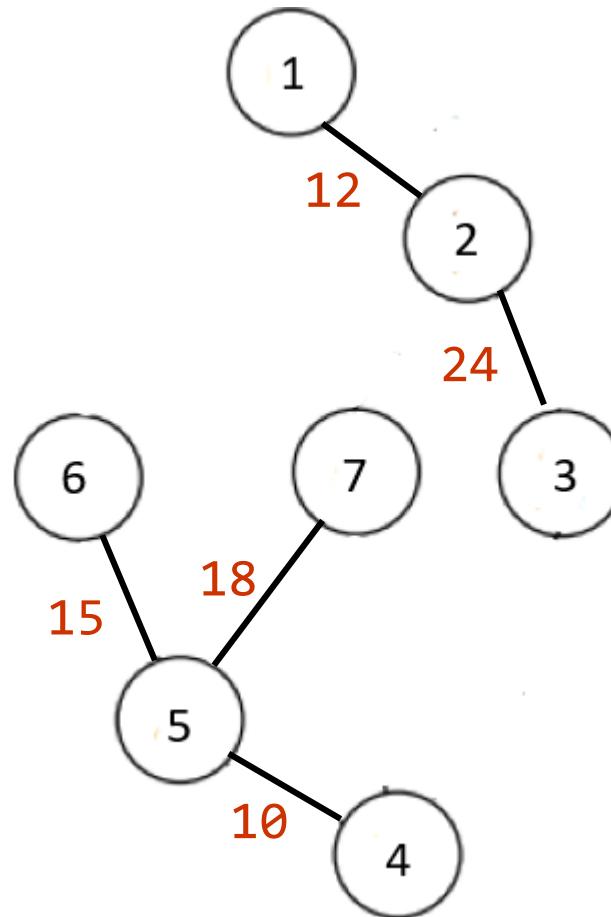
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Kruskal's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

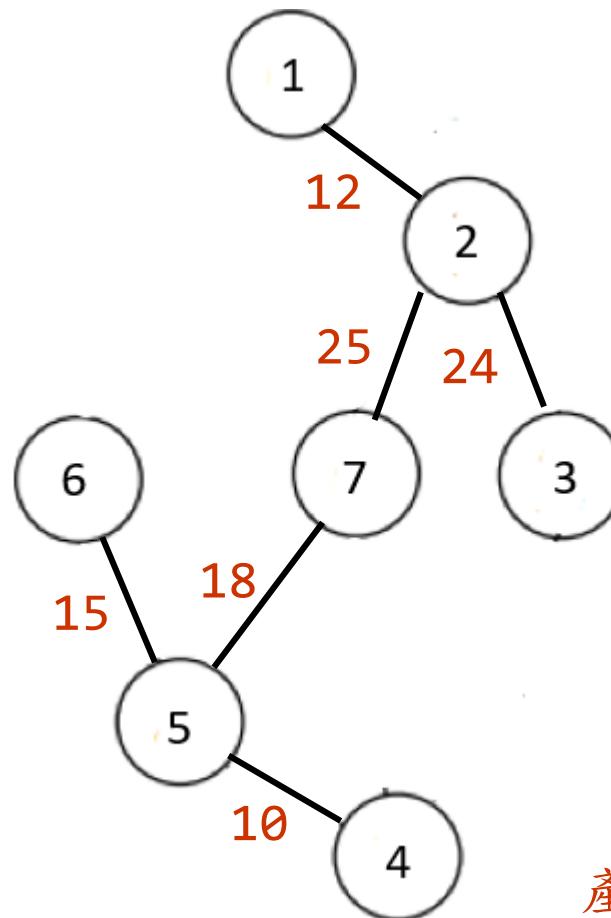
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



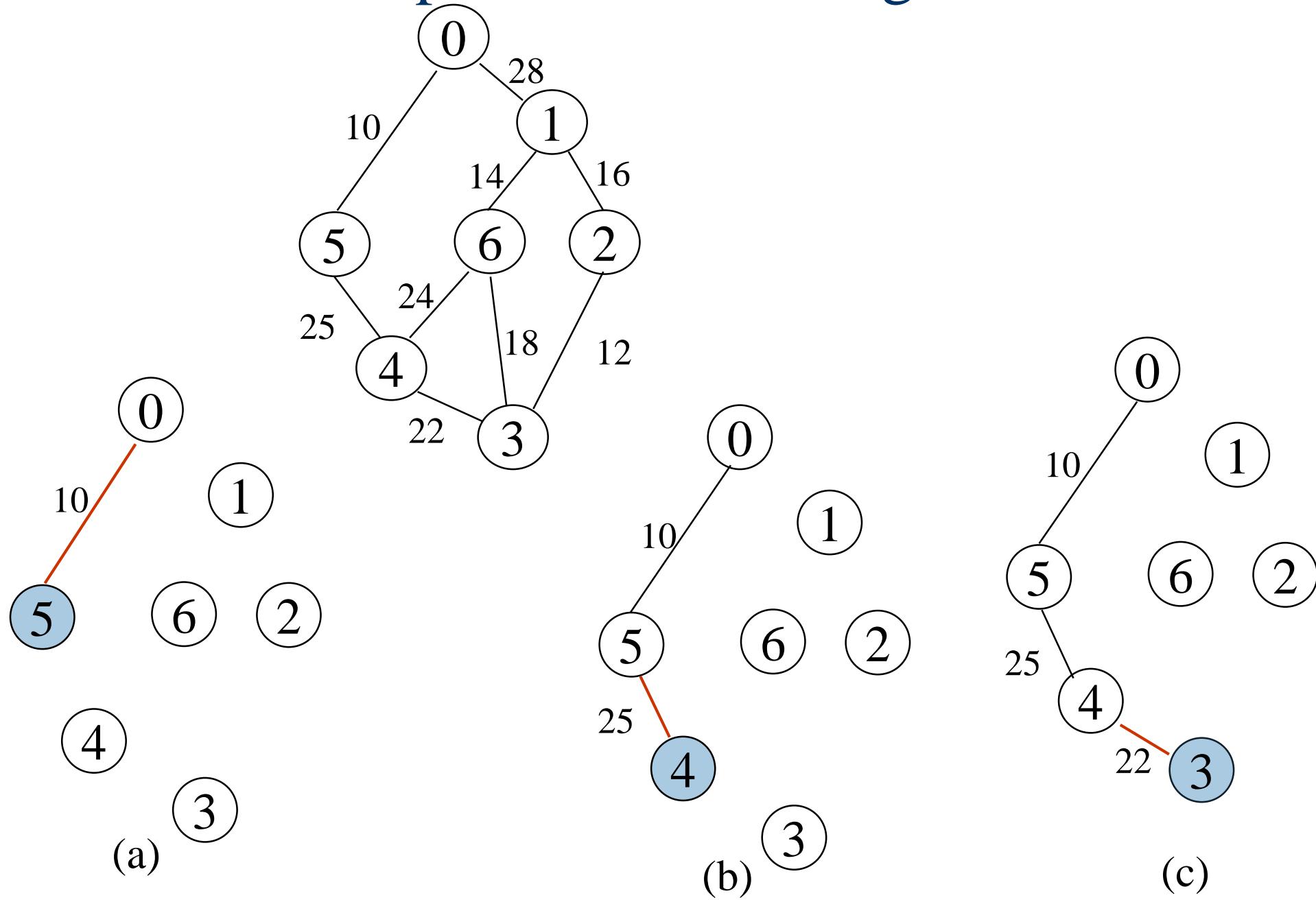
產生(頂點數-1)個邊

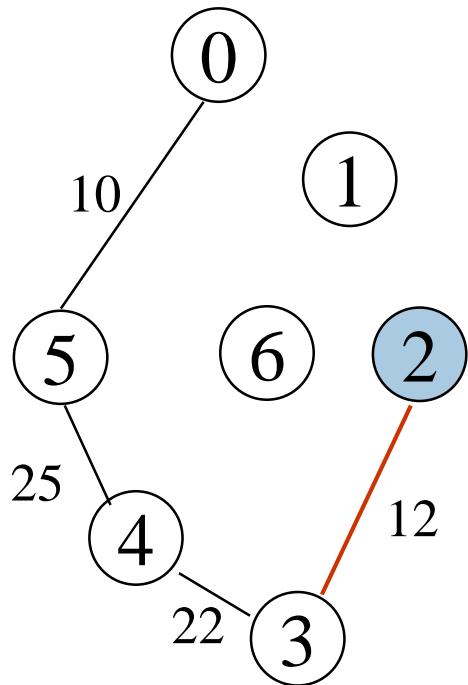
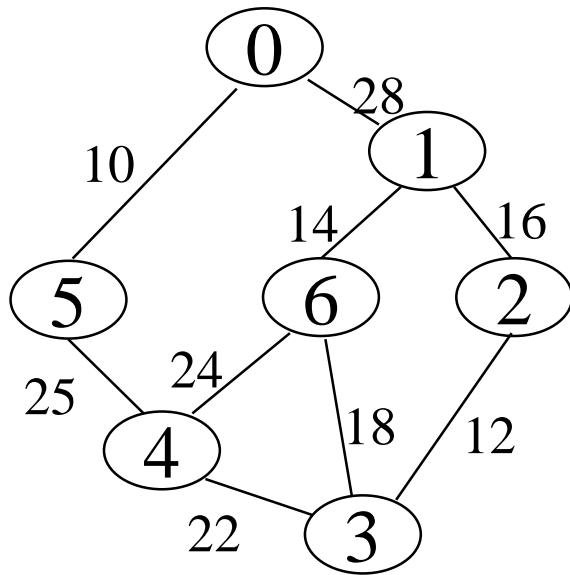
Prim's Algorithm

(tree all the time vs. forest)

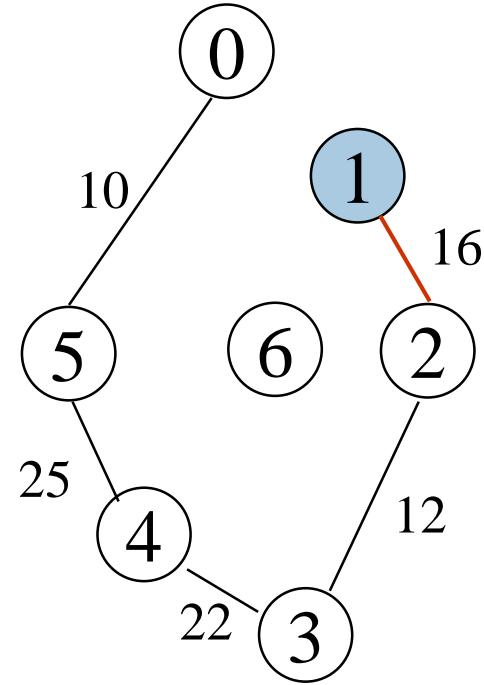
```
T={ } ;  
TV={ 0 } ;  
while (T contains fewer than n-1 edges)  
{  
    let (u,v) be a least cost edge such  
        that u ∈ TV and v ∉ TV  
    if (there is no such edge ) break;  
    add v to TV;  
    add (u,v) to T;  
}  
if (T contains fewer than n-1 edges)  
printf("No spanning tree\n");
```

Examples for Prim's Algorithm

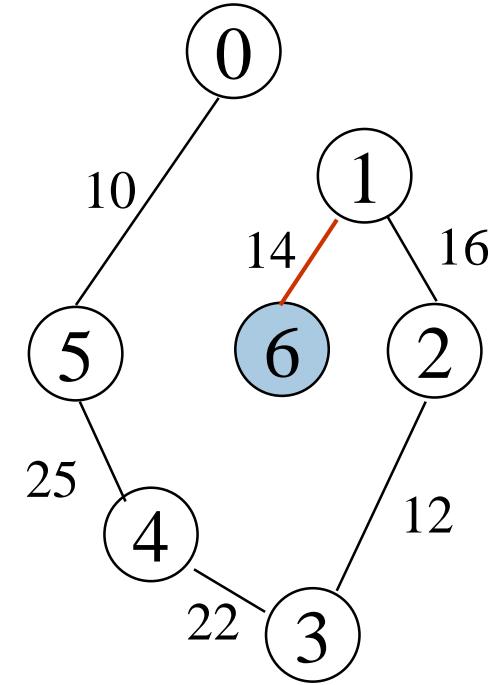




(d)



(e)



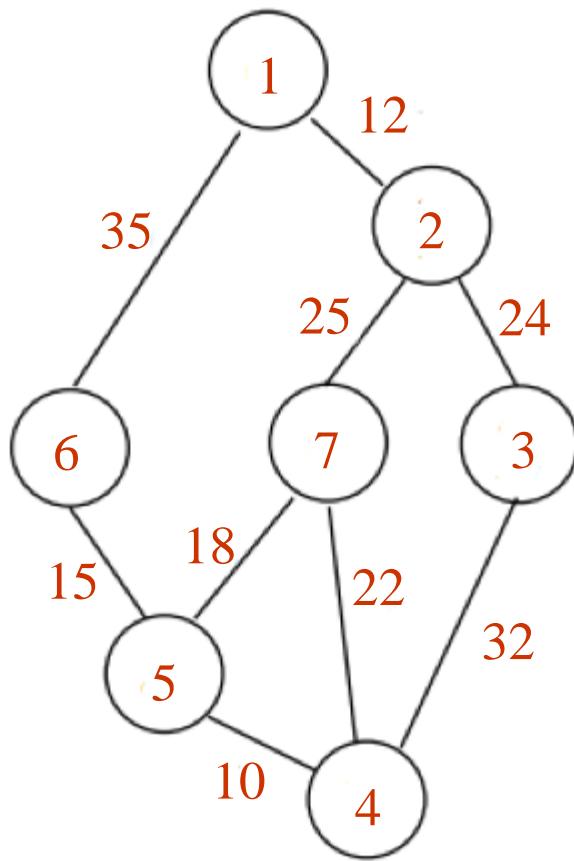
(f)

Prim's Algorithm

- 取得graph的Adjacency Matrix來實作 Prim's Algorithm
- 實作步驟
 1. 取得圖的關係
 2. 從頂點1開始選cost最小的邊來建立最小花費生成樹且不會使圖產生cycle

Prim's Algorithm

□ 圖解步驟



1 --12--> 2
2 --24--> 3
1 --35--> 6
2 --25--> 7
5 --15--> 6
4 --10--> 5
5 --18--> 7
4 --22--> 7
3 --32--> 4

Prim's Algorithm

□ 圖解步驟

1 --12--> 2 選cost小的

2 --24--> 3

1 --35--> 6

2 --25--> 7

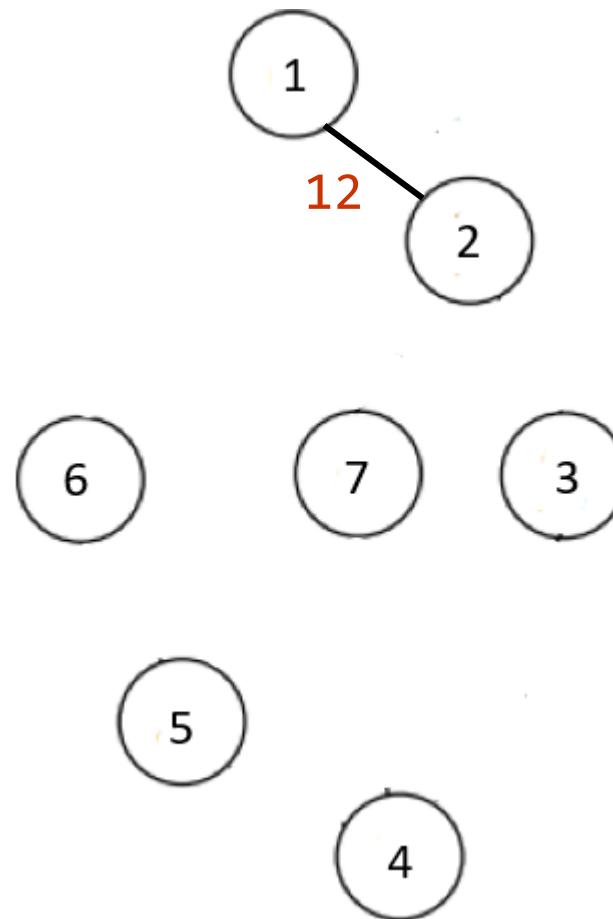
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3 選cost小的

1 --35--> 6

2 --25--> 7

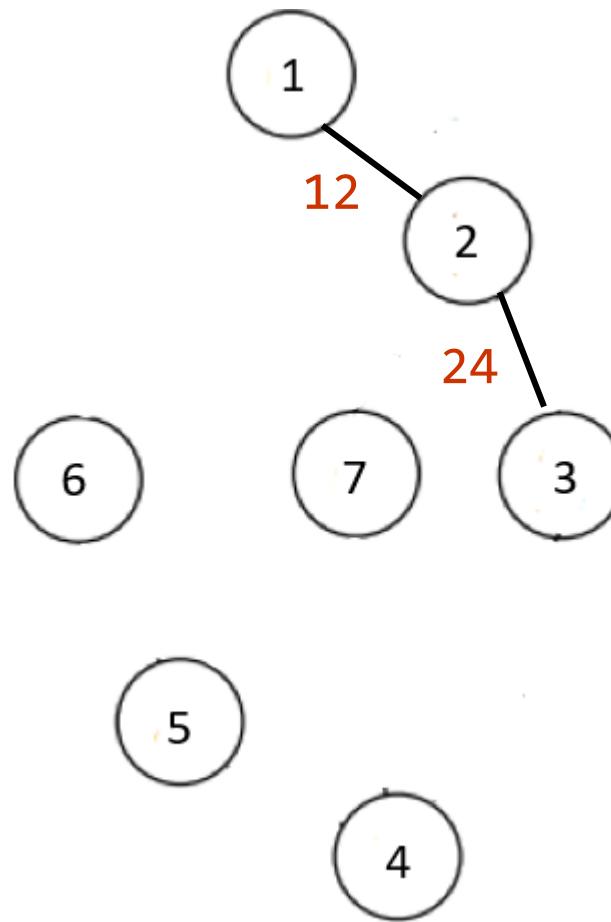
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7 選cost小的

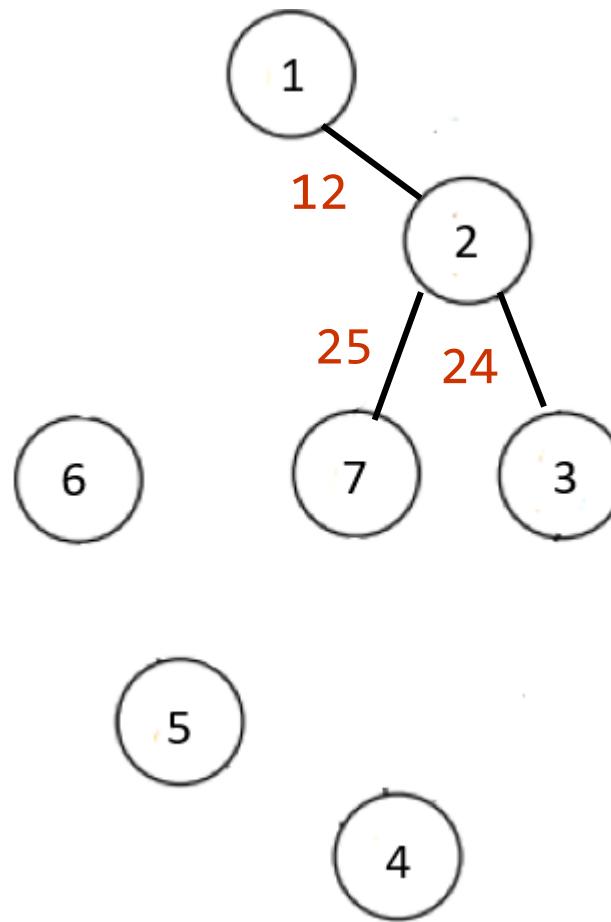
5 --15--> 6

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

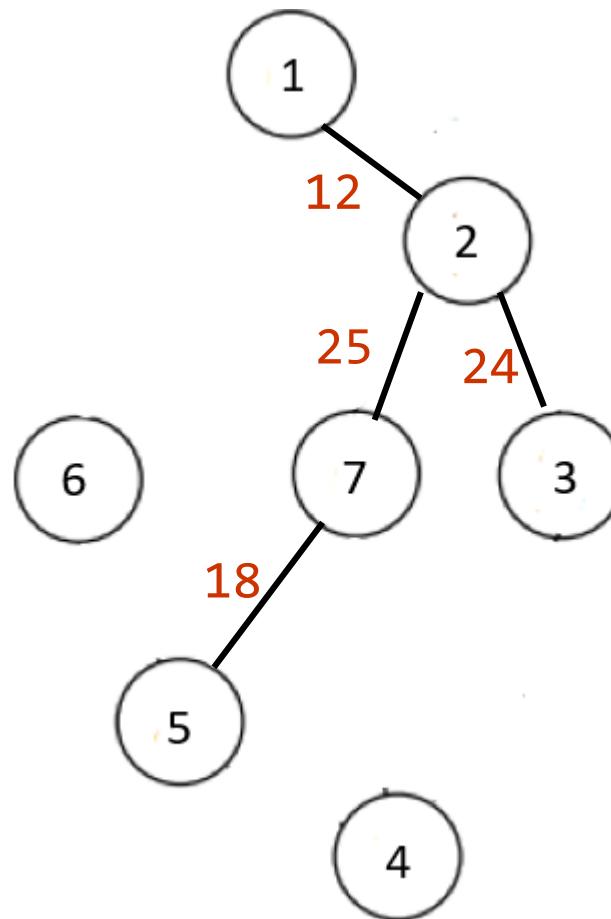
5 --15--> 6

4 --10--> 5

5 --18--> 7 選cost小的

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

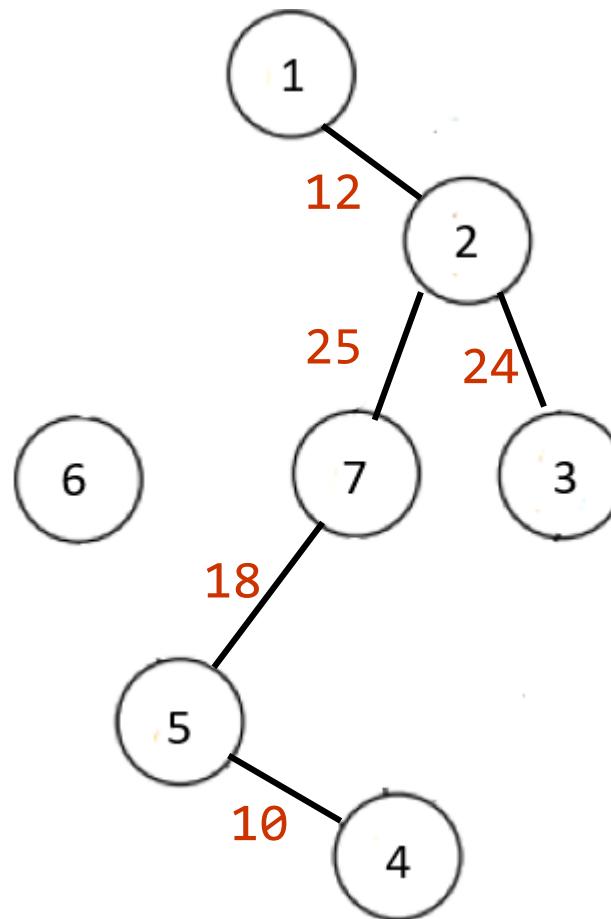
5 --15--> 6

4 --10--> 5 選cost小的

5 --18--> 7

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

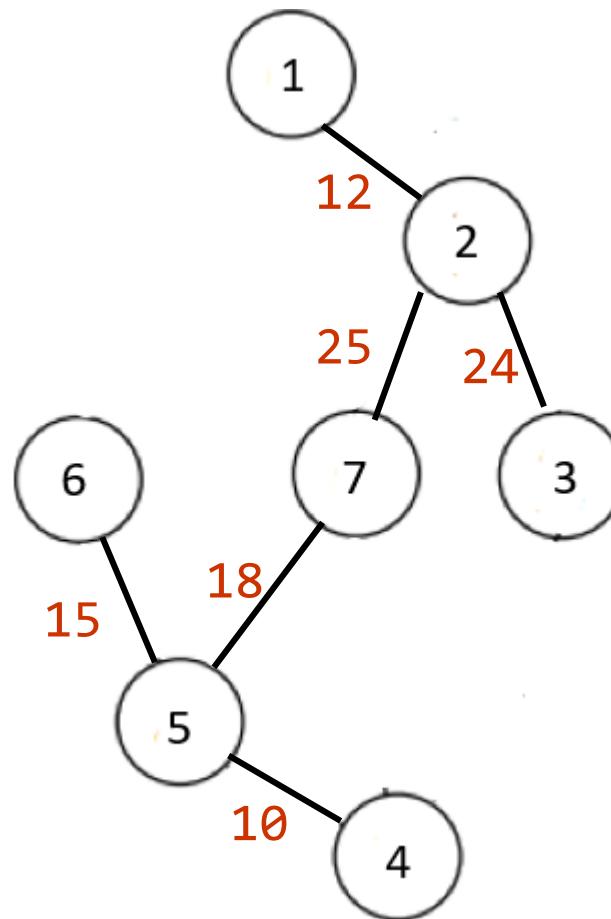
5 --15--> 6 選cost小的

4 --10--> 5

5 --18--> 7

4 --22--> 7

3 --32--> 4



Prim's Algorithm

□ 圖解步驟

1 --12--> 2

2 --24--> 3

1 --35--> 6

2 --25--> 7

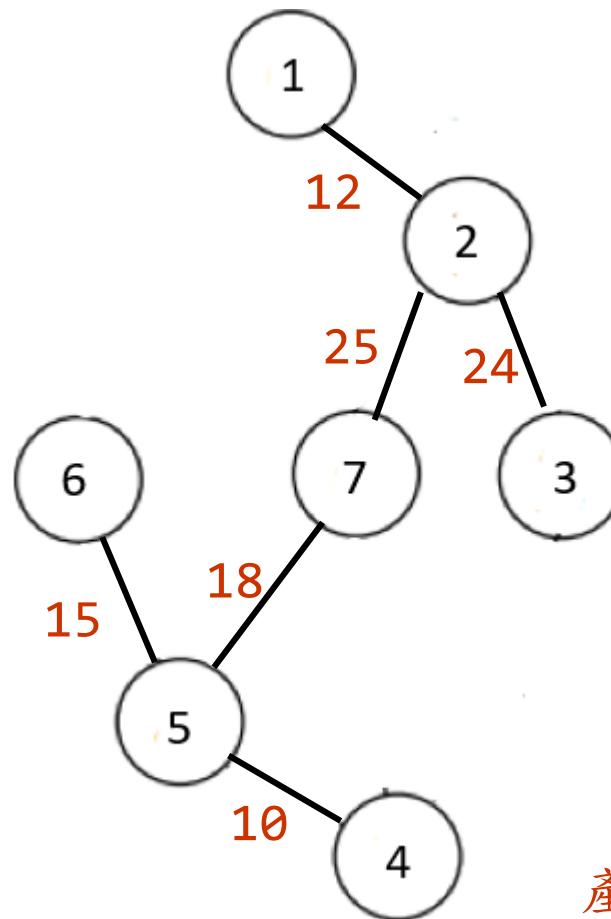
5 --15--> 6

4 --10--> 5

5 --18--> 7

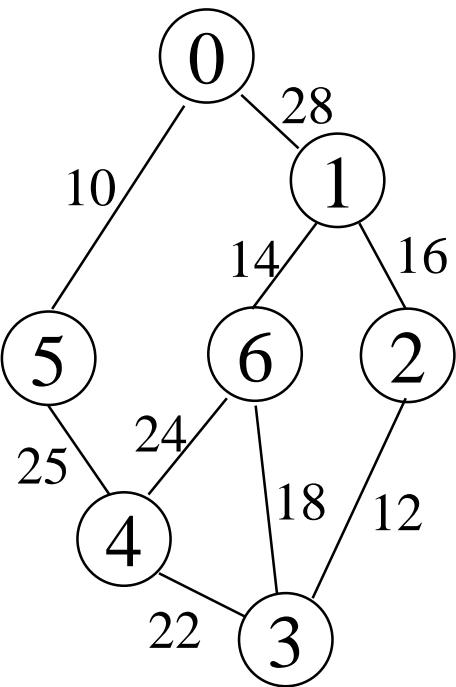
4 --22--> 7

3 --32--> 4

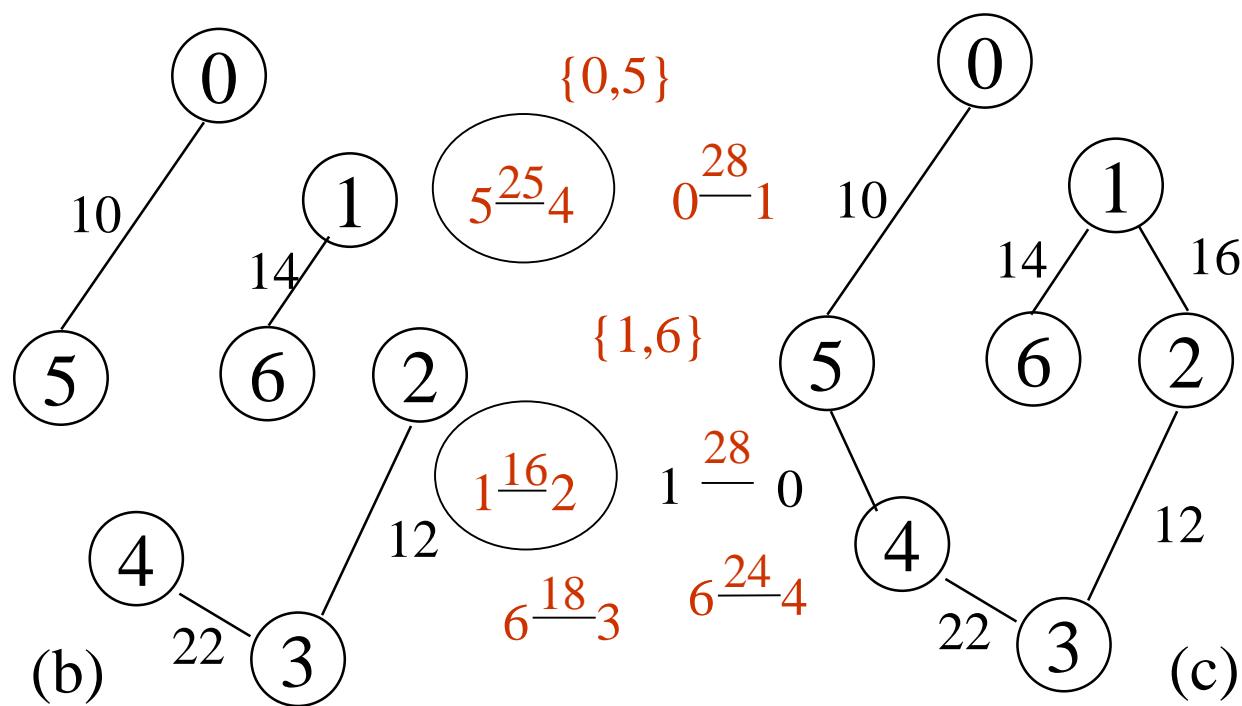
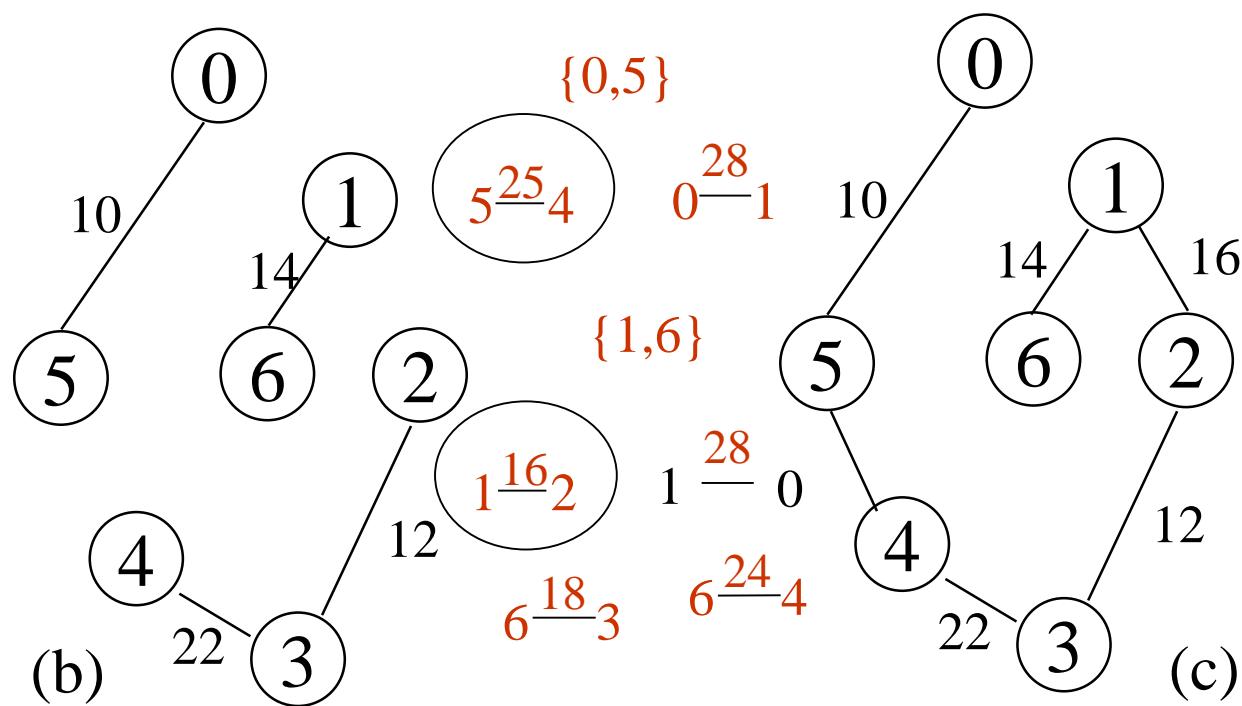
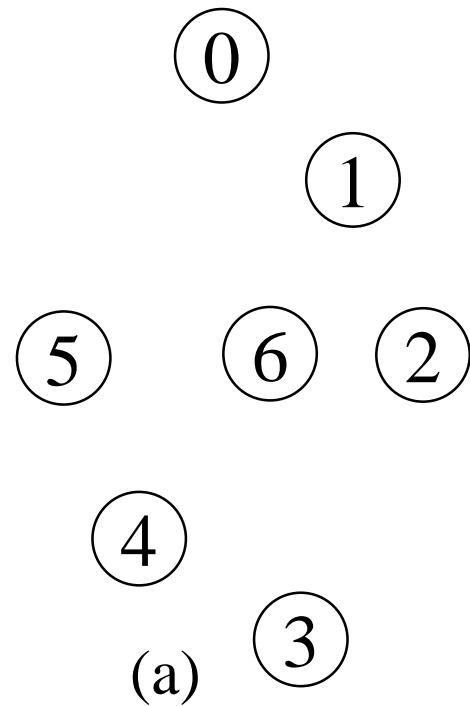


產生(頂點數-1)個邊

Sollin's Algorithm



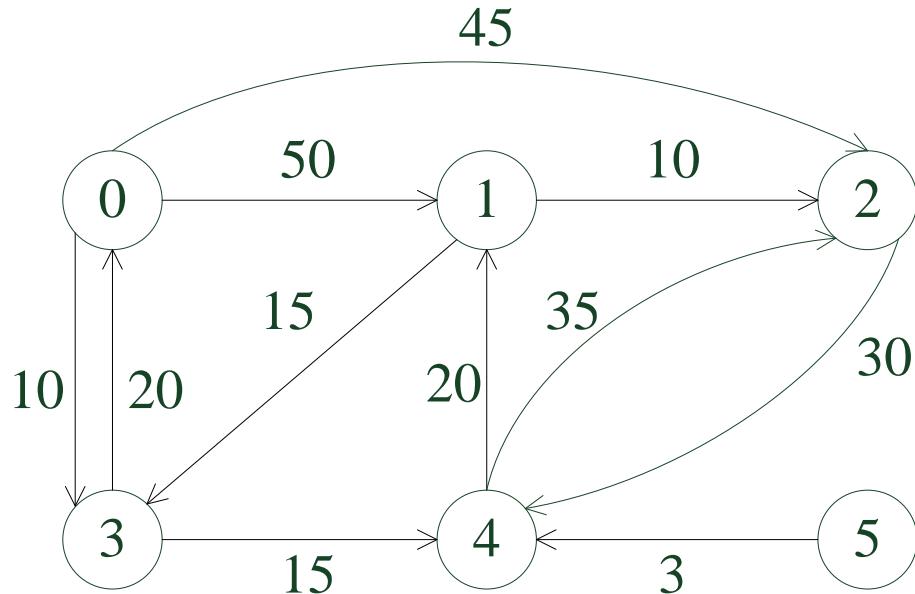
vertex	edge
0	0 -- 10 --> 5, 0 -- 28 --> 1
1	1 -- 14 --> 6, 1-- 16 --> 2, 1 -- 28 --> 0
2	2 -- 12 --> 3, 2 -- 16 --> 1
3	3 -- 12 --> 2, 3 -- 18 --> 6, 3 -- 22 --> 4
4	4 -- 22 --> 3, 4 -- 24 --> 6, 4 -- 25 --> 5
5	5 -- 10 --> 0, 5 -- 25 --> 4
6	6 -- 14 --> 1, 6 -- 18 --> 3, 6 -- 24 --> 4



Single Source to All Destinations



Determine the shortest paths from v_0 to all the remaining vertices.



(a) 圖

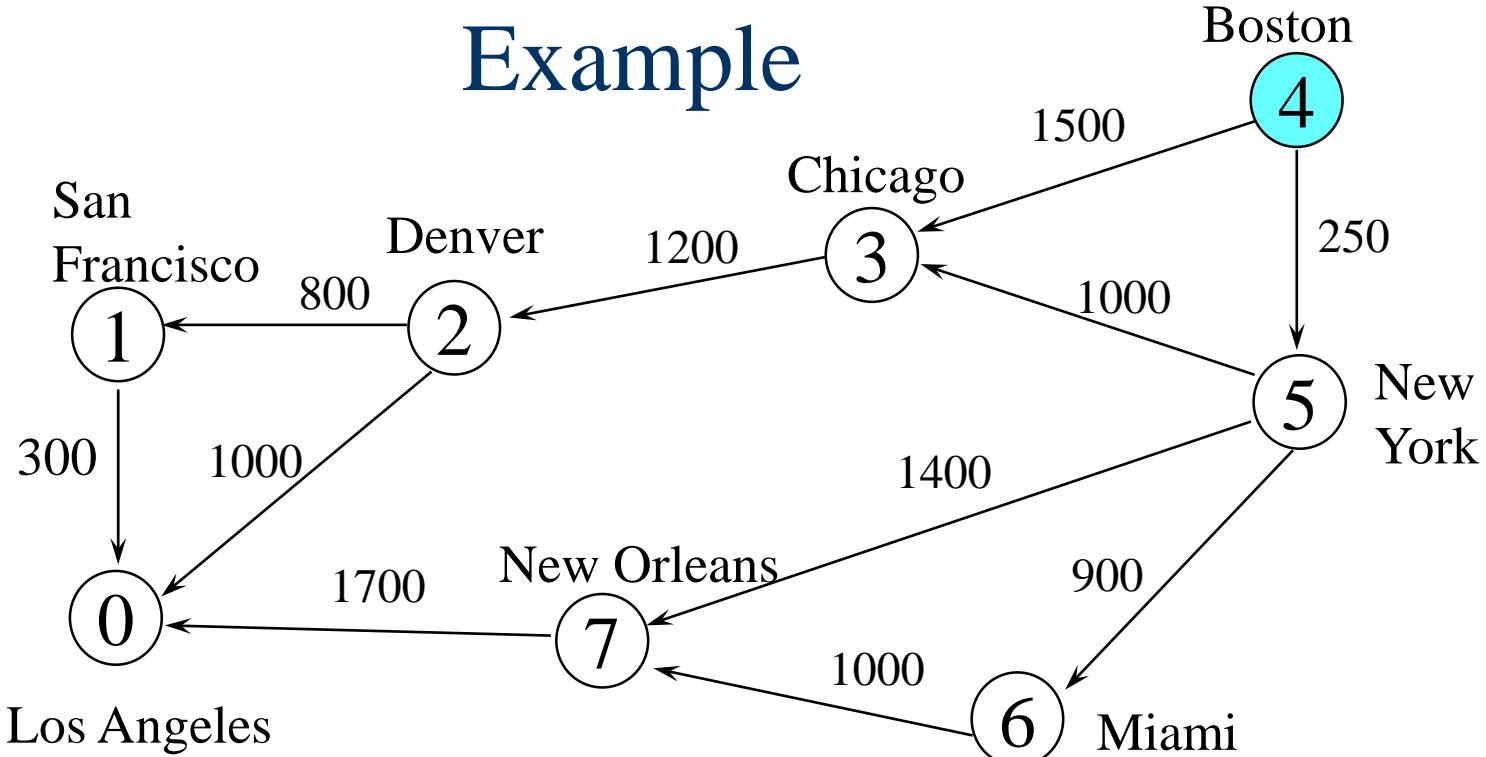
Dijkstra's algorithm

路徑	長度
1) 0, 3	10
2) 0, 3, 4	25
3) 0, 3, 4, 1	45
4) 0, 2	45

(b) 從 0 出發的最短路徑

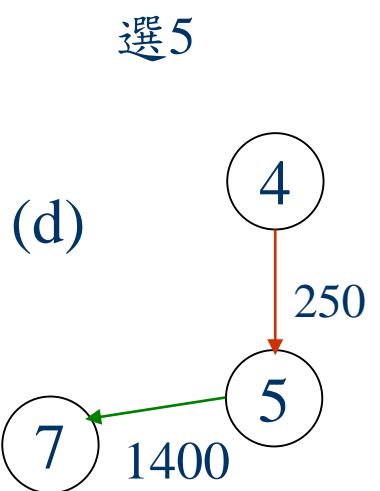
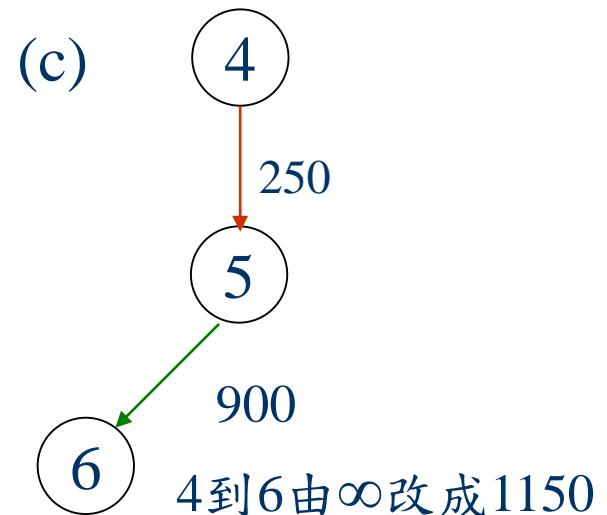
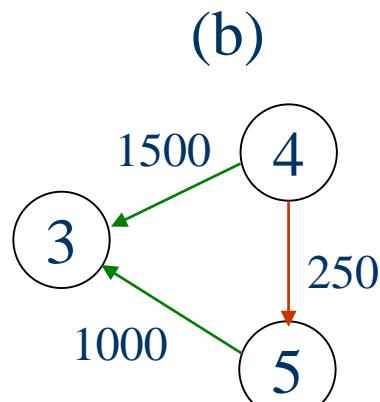
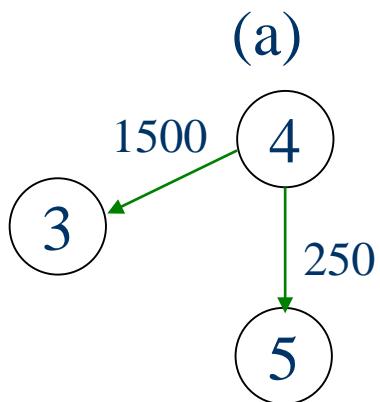
*Figure 6.26: Graph and shortest paths from v_0

Example

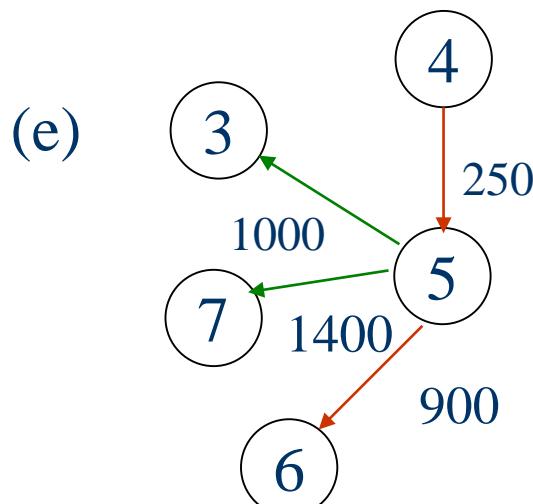


	0	1	2	3	4	5	6	7
0	0							
1	300	0						
2	1000	800	0					
3			1200	0				
4				1500	0	250		
5				1000	0	900	1400	
6					1000	0	1400	
7	1700					0		0

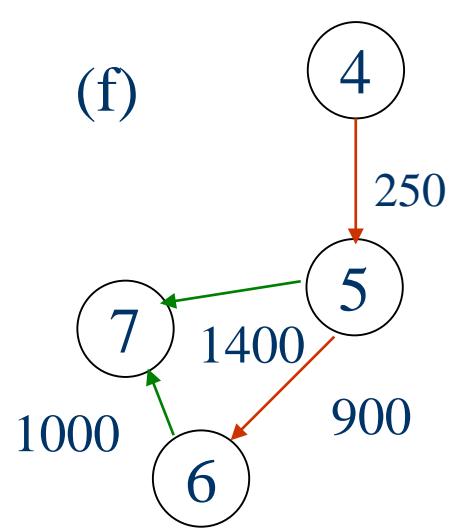
Cost adjacency matrix



4到7由 ∞ 改成1650

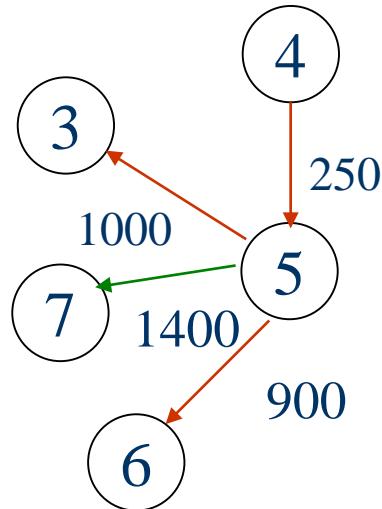


選6



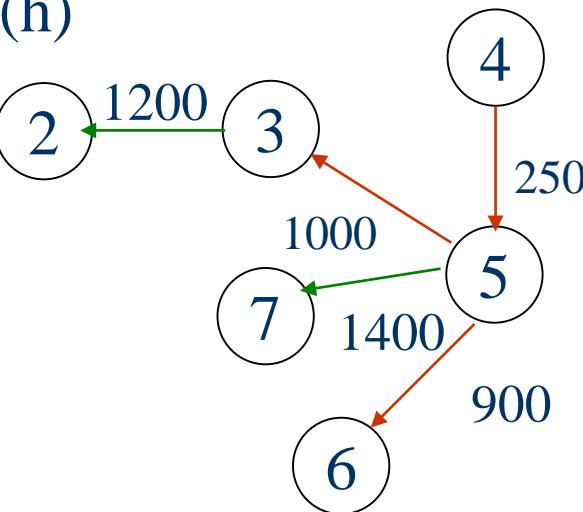
4-5-6-7比4-5-7長

(g)



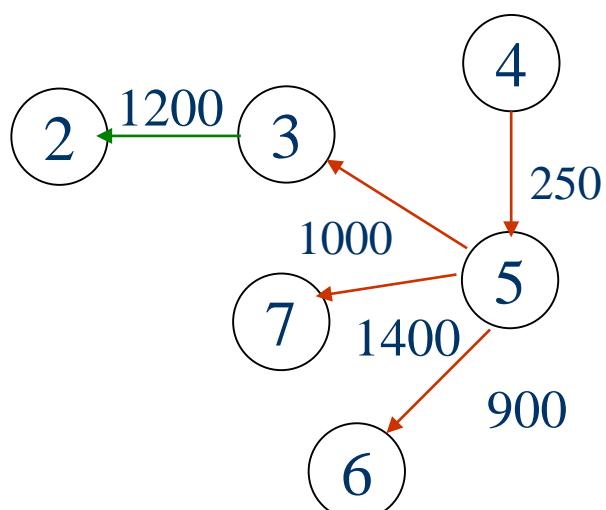
選3

(h)



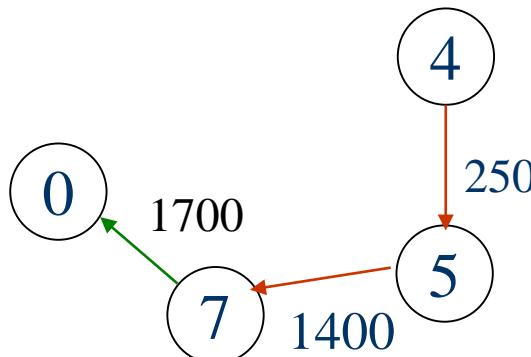
4到2由 ∞ 改成2450

(i)



選7

(j)



4到0由 ∞ 改成3350

Example for the Shortest Path

(Continued)

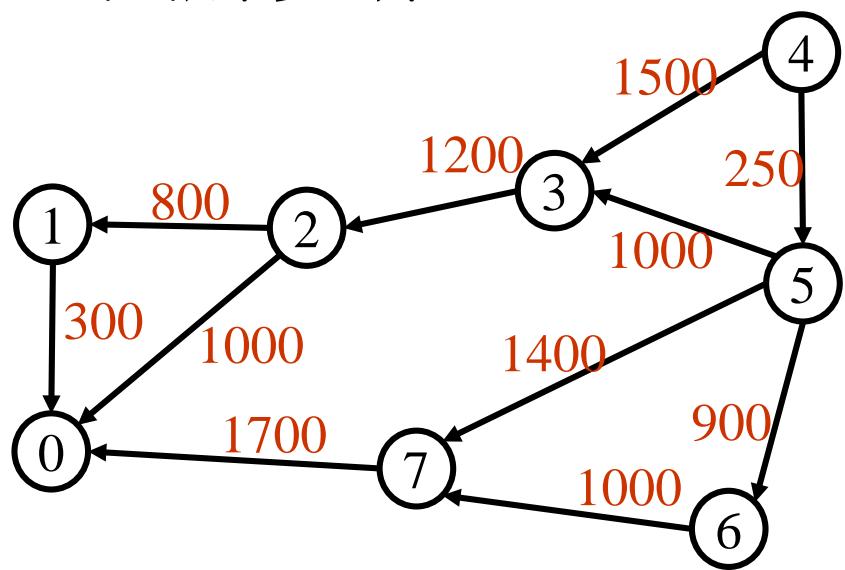
Iteration	S	Vertex Selected	LA [0]	SF [1]	DEN [2]	CHI [3]	BO [4]	NY [5]	MIA [6]	NO
Initial	--	----	$+\infty$	$+\infty$	$+\infty$	(b) 1500	0	250	(c) $+\infty$	(d) $+\infty$
1	{4}	(a) 5	$+\infty$	$+\infty$	$+\infty$	1250	0	250	1150	1650
2	{4,5}	(e) 6	$+\infty$	$+\infty$	$+\infty$	1250	0	250	1150	(f) 1650
3	{4,5,6}	(g) 3	$+\infty$	$+\infty$	2450	1250	0	250	1150	1650
4	{4,5,6,3}	(i) 7	3350	$+\infty$	2450	1250	0	250	1150	1650
5	{4,5,6,3,7}	2	3350	3250	2450	1250	0	250	1150	1650
6	{4,5,6,3,7,2}	1	3350	3250	2450	1250	0	250	1150	1650
7	{4,5,6,3,7,2,1}									

Dijkstra's algorithm

- 取得graph的Adjacency Matrix來實作 Dijkstra's algorithm
- 實作步驟
 1. 取得圖的關係
 2. 算出從起點到其他頂點的最短距離

Dijkstra's algorithm

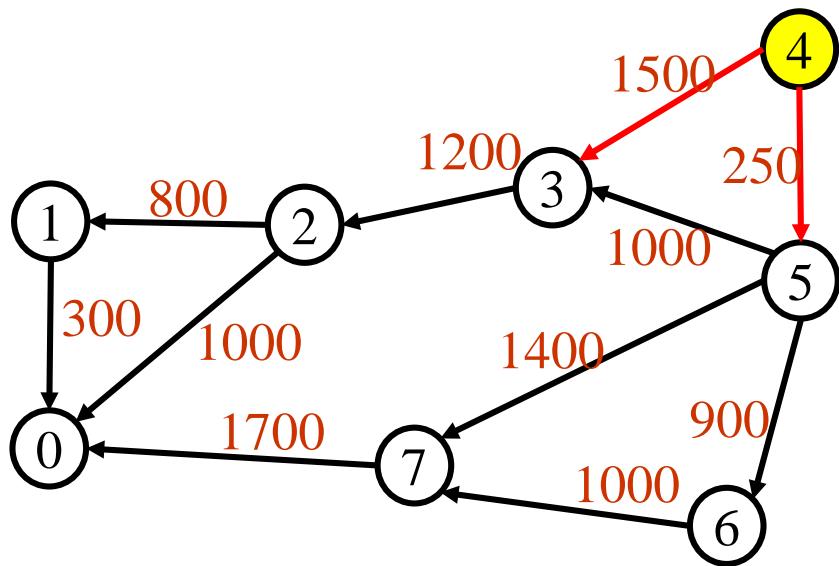
□ 圖解步驟



步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	∞	0	∞	∞

Dijkstra's algorithm

□ 步驟1

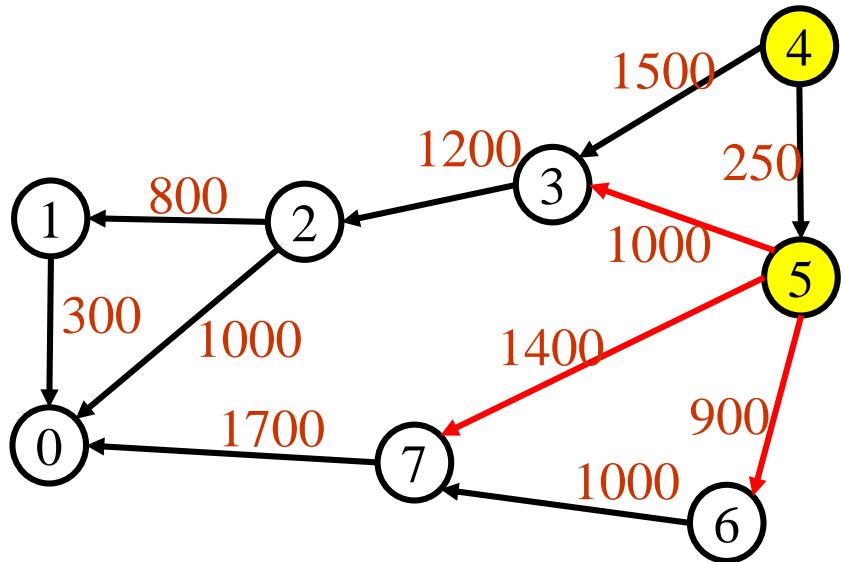


步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	∞	0	∞	∞
1	∞	∞	∞	∞	1500	0	250	∞
2	300	800	1200	1500	1500	250	∞	∞

從頂點4開始，與其相連的有頂點3和頂點5
頂點4→頂點5，距離250（新增）
頂點4→頂點3，距離1500（新增）

Dijkstra's algorithm

□ 步驟2



步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	∞	0	∞	∞
1	∞	∞	∞	∞	1500	0	250	∞
2	∞	∞	∞	1250	0	250	1150	1650

加入頂點5，與其相連的有頂點3, 6, 7

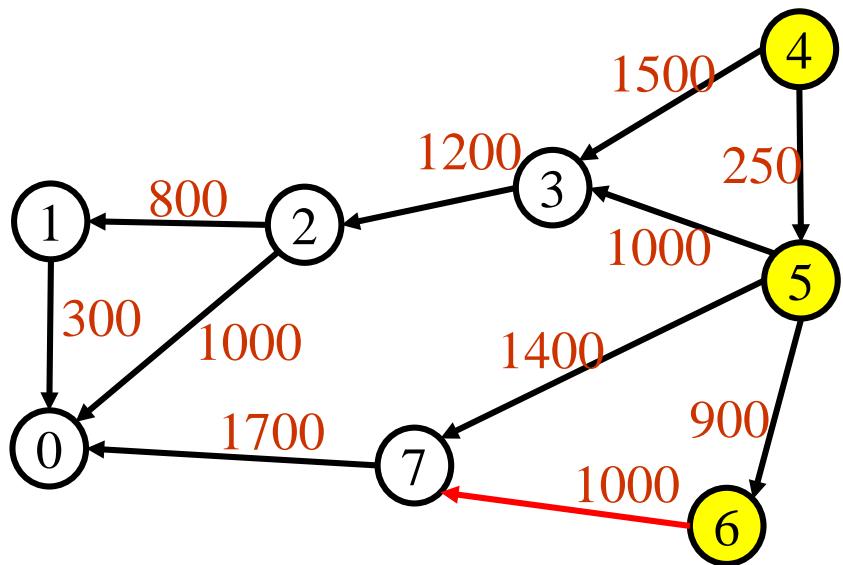
頂點5→頂點3，距離 $250+1000=1250$ （更新）

頂點5→頂點6，距離 $250+900=1150$ （新增）

頂點5→頂點7，距離 $250+1400=1650$ （新增）

Dijkstra's algorithm

□ 步驟3

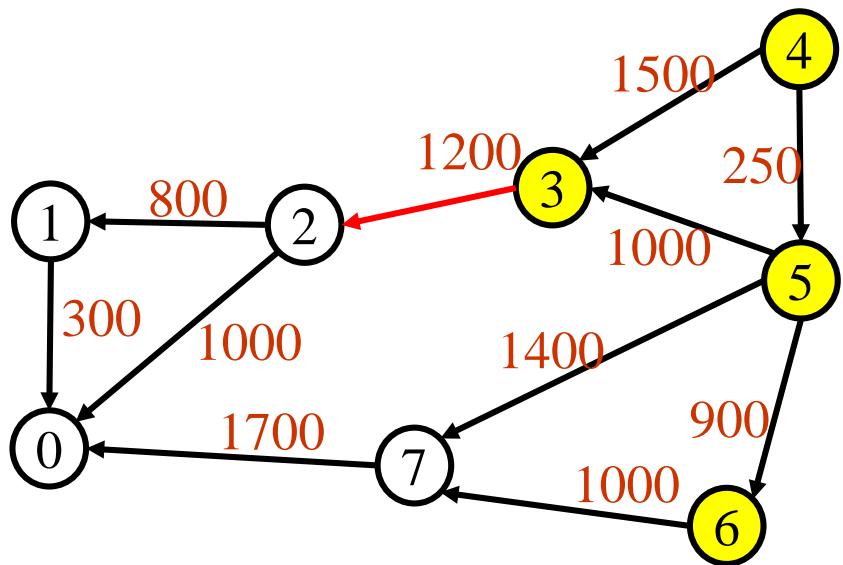


步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	∞	0	∞	∞
1	∞	∞	∞	∞	1500	0	250	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650

加入頂點6，與其相連的有頂點7
頂點6->頂點7，距離 $1150+1000=2150$ （不更新）

Dijkstra's algorithm

□ 步驟4

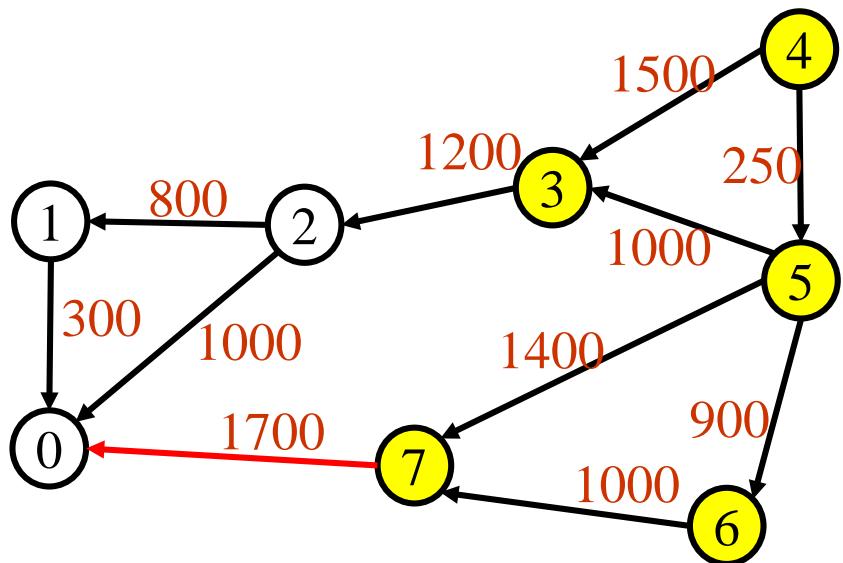


步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	0	∞	∞	∞
1	∞	∞	∞	1500	0	250	∞	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650
4	∞	∞	2450	1250	0	250	1150	1650

加入頂點3，與其相連的有頂點2
頂點3->頂點2，距離 $1250+1200=2450$ （新增）

Dijkstra's algorithm

□ 步驟5

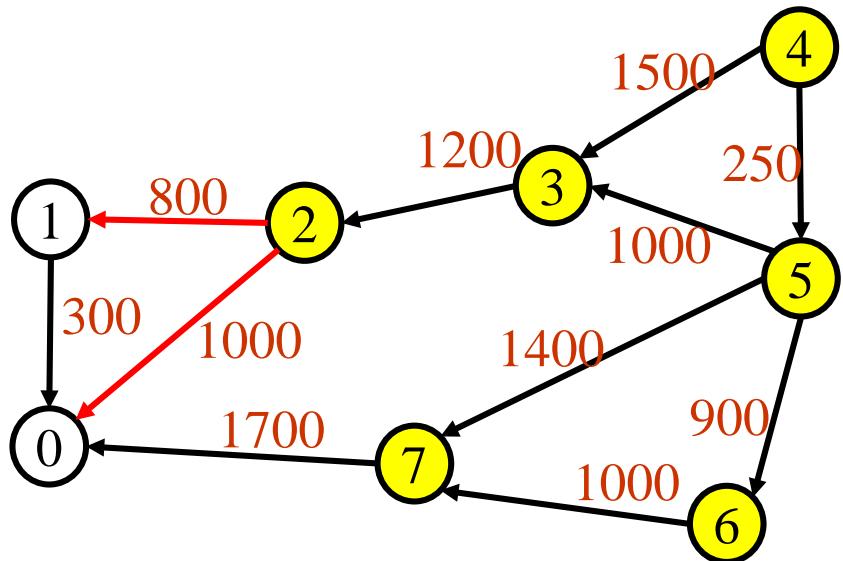


步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	0	∞	∞	∞
1	∞	∞	∞	1500	0	250	∞	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650
4	∞	∞	2450	1250	0	250	1150	1650
5	3350	∞	2450	1250	0	250	1150	1650

加入頂點7，與其相連的有頂點0
頂點7→頂點0，距離 $1650+1700=3350$ （新增）

Dijkstra's algorithm

□ 步驟6



步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	0	∞	∞	∞
1	∞	∞	∞	1500	0	250	∞	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650
4	∞	∞	2450	1250	0	250	1150	1650
5	3350	∞	2450	1250	0	250	1150	1650
6	3350	3250	2450	1250	0	250	1150	1650

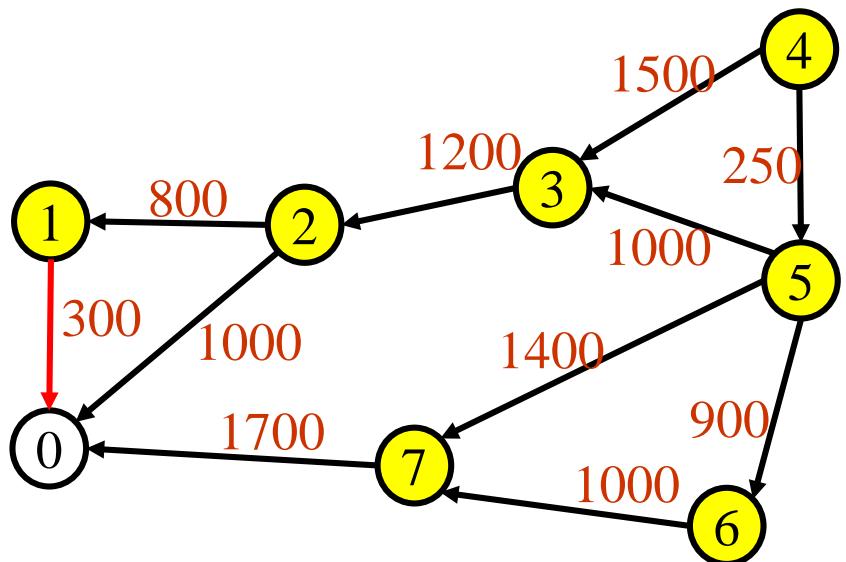
加入頂點2，與其相連的有頂點1, 0

頂點2→頂點0，距離 $2450+1000=3450$ （不更新）

頂點2→頂點1，距離 $2450+800=3250$ （新增）

Dijkstra's algorithm

□ 步驟7

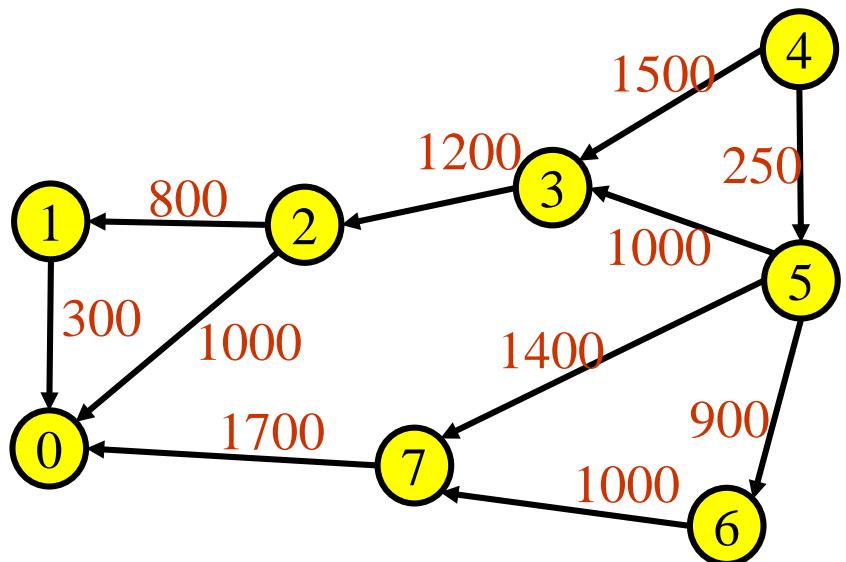


加入頂點1，與其相連的有頂點0
頂點1->頂點0，距離 $3250+300=3550$ （不更新）

步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	0	∞	∞	∞
1	∞	∞	∞	1500	0	250	∞	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650
4	∞	∞	2450	1250	0	250	1150	1650
5	3350	∞	2450	1250	0	250	1150	1650
6	3350	3250	2450	1250	0	250	1150	1650
7	3350	3250	2450	1250	0	250	1150	1650

Dijkstra's algorithm

□ 步驟8



加入頂點0

步驟	0	1	2	3	4	5	6	7
init	∞	∞	∞	∞	0	∞	∞	∞
1	∞	∞	∞	1500	0	250	∞	∞
2	∞	∞	∞	1250	0	250	1150	1650
3	∞	∞	∞	1250	0	250	1150	1650
4	∞	∞	2450	1250	0	250	1150	1650
5	3350	∞	2450	1250	0	250	1150	1650
6	3350	3250	2450	1250	0	250	1150	1650
7	3350	3250	2450	1250	0	250	1150	1650

Single Source to All Destinations

```
void shortestpath(int v, int  
    cost[ ][MAX_ERXTICES], int distance[], int n,  
    short int found[] )  
{  
    int i, u, w;  
    for (i=0; i<n; i++) {  
        found[i] = FALSE;  
        distance[i] = cost[v][i];          O(n)  
    }  
    found[v] = TRUE;  
    distance[v] = 0;
```

```

for (i=0; i<n-2; i++) {  

    determine n-1 paths from v  

    u = choose(distance, n, found);  

    // return u which is minimum distance to v  

    found[u] = TRUE;  

    for (w=0; w<n; w++)  

        if (!found[w]) 與u相連的端點w  

            if (distance[u]+cost[u][w]<distance[w])  

                distance[w] = distance[u]+cost[u][w];  

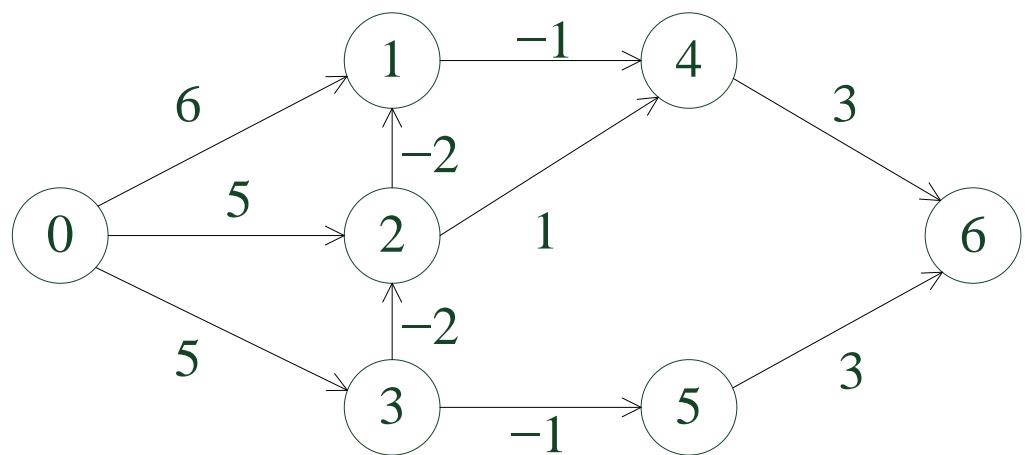
}
}

```

$O(n^2)$

```
int choose(int distance[ ], int n, short int
{found[])
{
    /* 找出還沒確認最短距離的點 */
    int i, min, minpos;
    min = INT_MAX;
    minpos = -1;
    for (i = 0; i < n; i++) {
        if (distance[i] < min && !found[i])
        {
            min = distance[i];
            minpos = i;
        }
    }
    return minpos;
}
```

Shortest paths with negative edge lengths

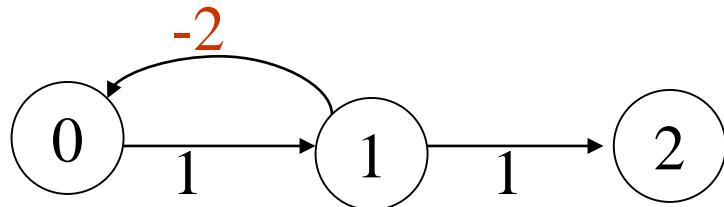


(a) 有向圖

	$dist^k[7]$						
k	0	1	2	3	4	5	6
1	0	6	5	5	∞	∞	∞
2	0	3	3	5	5	4	∞
3	0	1	3	5	2	4	7
4	0	1	3	5	0	4	5
5	0	1	3	5	0	4	3
6	0	1	3	5	0	4	3

(b) $dist^k$

Graph with Negative Cycle



(a) Directed graph

$$\begin{bmatrix} 0 & 1 & \infty \\ -2 & 0 & 1 \\ \infty & \infty & 0 \end{bmatrix}$$

(b) A^{-1}

0, 1, 0, 1, 0, 1, ..., 0, 1, 2

The length of the shortest path from vertex 0 to vertex 2 is $-\infty$.

Bellman and Ford algorithm to compute shortest paths

```
void BellmanFord(int n, int v)
{
    /* 計算單一起點/所有終點的最短路徑，其中邊長允許是負值 */
    for (int i = 0; i < n; i++)
        dist[i] = length[v][i];
    /* 對dist做初始化 */

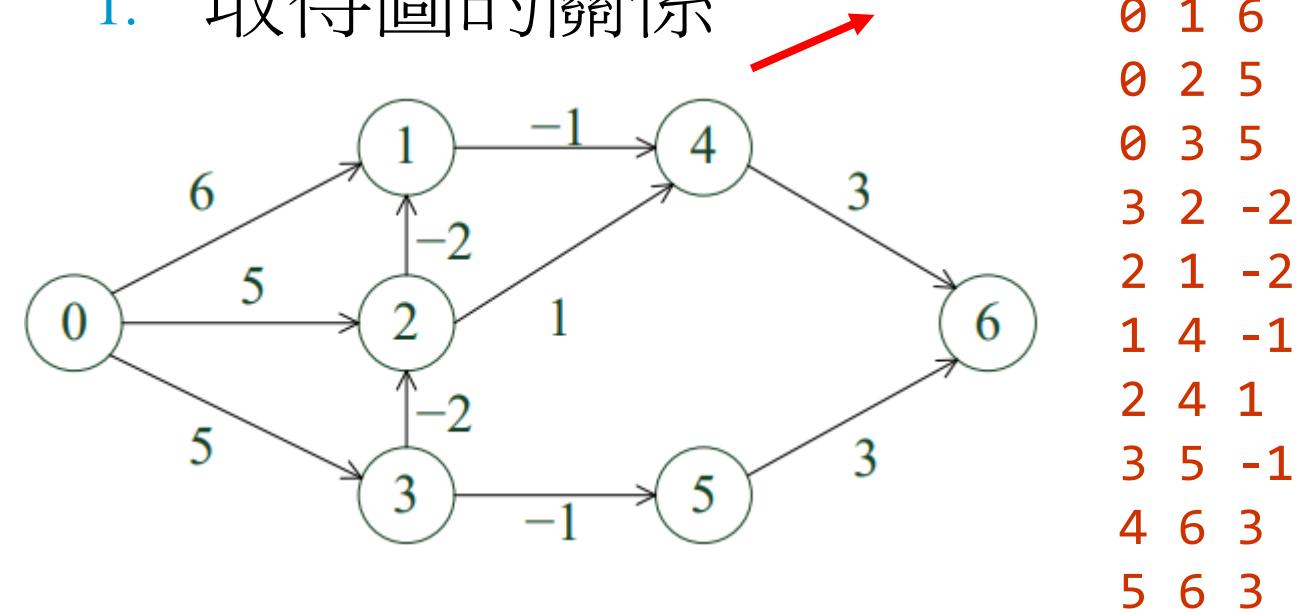
    for (int k = 2; k <= n-1; k++)
        for (每個u滿足u!=v 且u至少有一個進到它的邊)
            for(每個圖上的邊<i,u>)
                if(dist[u] > dist[i] + length[i][u])
                    dist[u] = dist[i] + length[i][u];
}
```

Bellman and Ford algorithm

- 取得graph的關係後來實作Bellman and Ford algorithm

- 實作步驟

1. 取得圖的關係



Number of vertices in graph: 7
Number of edges in graph: 10
Source vertex number: 0
Source, destination, weight of edge

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

Bellman and Ford algorithm

init

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	∞	∞	∞	∞	∞	∞

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

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



$$0 + 6 < c$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 0 + 6$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	6	∞	∞	∞	∞	∞



0	6	5	∞	∞	∞	∞
0	6	5	∞	∞	∞	∞

$$0 + 5 < c$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 0 + 5$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	6	5	∞	∞	∞	∞



0	6	5	5	∞	∞	∞
0	6	5	5	∞	∞	∞

$$0 + 5 < \infty$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 0 + 5$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	6	5	5	∞	∞	∞



0	6	3	5	∞	∞	∞
---	---	---	---	----------	----------	----------

$$5 + -2 < 5$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 5 + -2$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	6	3	5	∞	∞	∞



0	1	3	5	∞	∞	∞
0	1	3	5	∞	∞	∞

$$3 + -2 < 6$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 3 + -2$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	1	3	5	∞	∞	∞



0	1	3	5	0	∞	∞
0	1	3	5	0	∞	∞

$$1 + -1 < \infty$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 1 + -1$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	1	3	5	0	∞	∞

$$3 + 1 < 0$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

不成立

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	1	3	5	0	∞	∞



0	1	3	5	0	4	∞
0	1	3	5	0	4	∞

$$5 + -1 < \infty$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 5 + -1$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

0	1	2	3	4	5	6
0	1	3	5	0	4	∞



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

$$0 + 3 < \infty$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

$$\text{StoreDistance}[v] = 0 + 3$$

Bellman and Ford algorithm

Source =u	Destination =v	weight
0	1	6
0	2	5
0	3	5
3	2	-2
2	1	-2
1	4	-1
2	4	1
3	5	-1
4	6	3
5	6	3

StoreDistance[]

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

$$4 + 3 < 3$$

```
if (StoreDistance[u] + weight < StoreDistance[v])
    StoreDistance[v] = StoreDistance[u] + weight;
```

不成立

All Pairs Shortest Paths

□ Find the shortest paths between all pairs of vertices.

□ Solution 1

- Apply **shortest path** n times with each vertex as source.

$O(n^3)$

□ Solution 2

- Represent the graph G by its cost adjacency matrix with $\text{cost}[i][j]$
- If the edge $\langle i, j \rangle$ is not in G , the $\text{cost}[i][j]$ is set to some sufficiently large number
- $A[i][j]$ is the cost of the shortest path from i to j , **using only those intermediate vertices with an index $\leq k$**

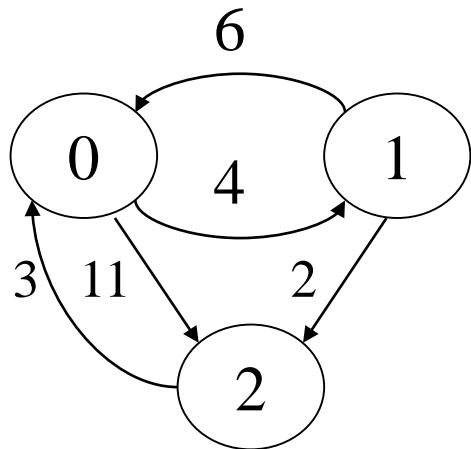
All Pairs Shortest Paths (*Continued*)

- The cost of the shortest path from i to j is $A^{n-1}[i][j]$, as no vertex in G has an index greater than $n-1$
- $A^{-1}[i][j] = \text{cost}[i][j]$
- Calculate the $A^0, A^1, A^2, \dots, A^{n-1}$ from A^{-1} iteratively
- $A^k[i][j] = \min\{A^{k-1}[i][j], A^{k-1}[i][k] + A^{k-1}[k][j]\}, k >= 0$

Algorithm for All Pairs Shortest Paths

```
void allcosts(int cost[ ][MAX_VERTICES] ,
              int distance[ ][MAX_VERTICES] , int n)
{
    int i, j, k;
    for (i=0; i<n; i++)
        for (j=0; j<n; j++)
            distance[i][j] = cost[i][j];
    for (k=0; k<n; k++)
        for (i=0; i<n; i++)
            for (j=0; j<n; j++)
                if (distance[i][k]+distance[k][j]
                    < distance[i][j])
                    distance[i][j]=
                        distance[i][k]+distance[k][j];
}
```

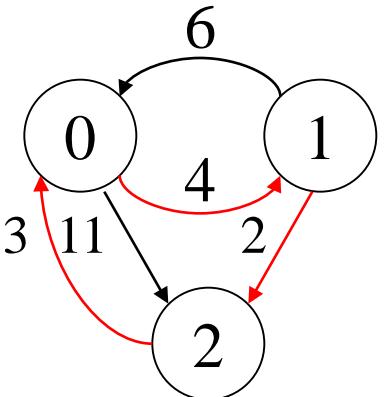
*** Figure 6.33: Directed graph and its cost matrix**



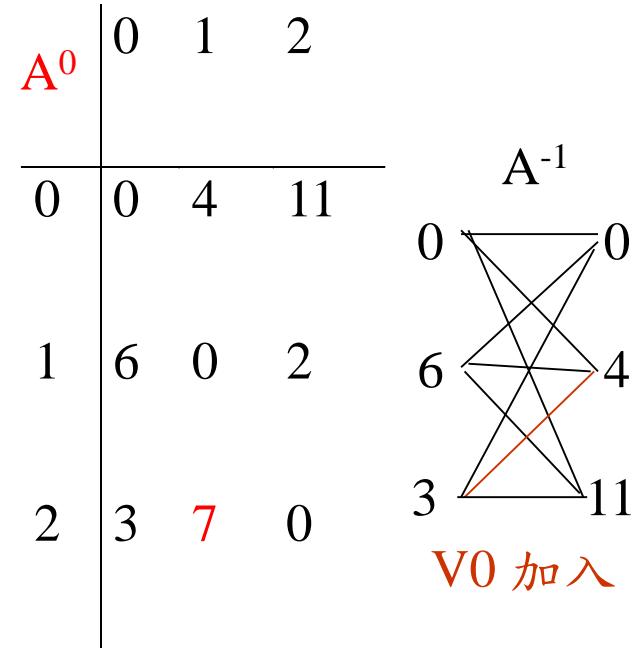
(a)Directed graph G

	0	1	2
0	0	4	11
1	6	0	2
2	3	∞	0

(b)Cost adjacency matrix for G



A^{-1}	0	1	2
0	0	4	11
1	6	0	2
2	3	∞	0
A^1	0	1	2
0	0	4	6
1	6	0	2
2	3	7	0



A²

0 1 2

0 0 4 6

1 5 0 2

2 3 7 0

A¹

6 3

2 7

0 0

V2 加入

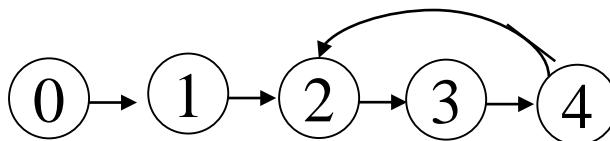
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Transitive Closure

Goal: given a graph with unweighted edges, determine if there is a path from i to j for all i and j .

(1) Require positive path (> 0) lengths. transitive closure matrix

(2) Require nonnegative path (≥ 0) lengths. reflexive transitive closure matrix



(a) Digraph G

$$\begin{matrix} 0 & \begin{bmatrix} 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \end{bmatrix} \\ 1 & \text{cycle} \end{matrix}$$

(c) transitive closure matrix A^+

There is a path of length > 0

$$\begin{matrix} 0 & \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 \end{bmatrix} \\ 1 & \\ 2 & \\ 3 & \\ 4 & \end{matrix}$$

(b) Adjacency matrix A for G

$$\begin{matrix} 0 & \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 \end{bmatrix} \\ 1 & \text{reflexive} \end{matrix}$$

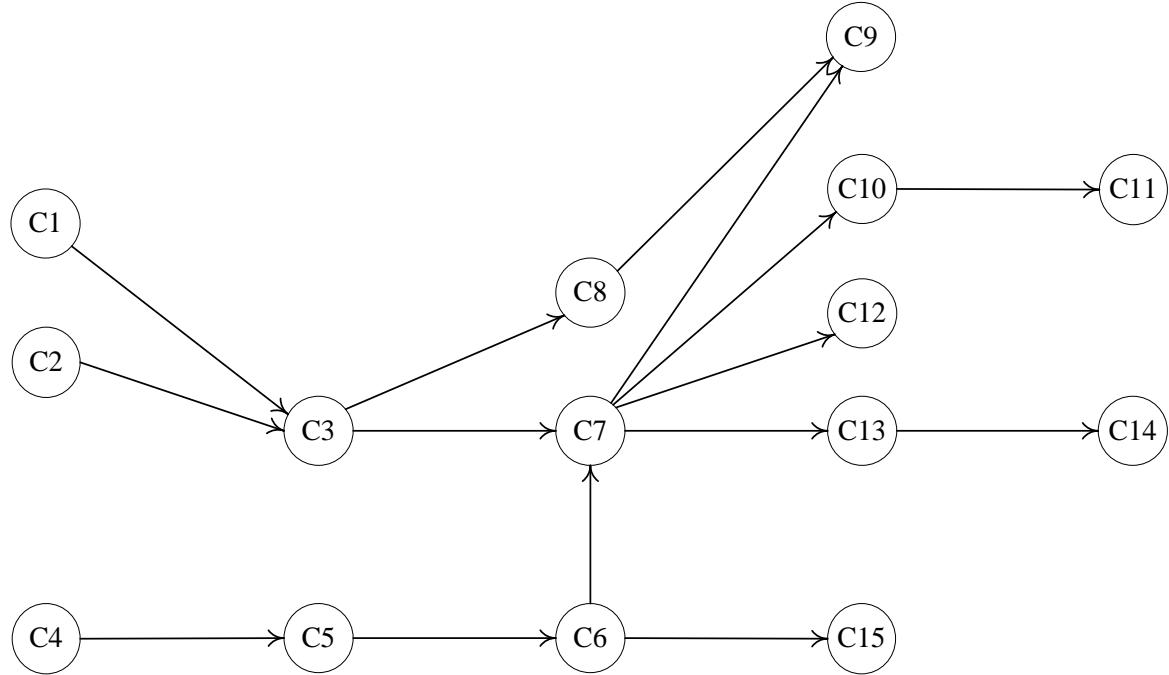
(d) reflexive transitive closure matrix A^*

There is a path of length ≥ 0

Activity on Vertex (AOV) Network

- Definition: A directed graph in which the vertices represent tasks or activities and the edges represent precedence relations between tasks.
- Predecessor (successor): vertex i is a predecessor of vertex j iff there is a directed path from i to j .
 - j is a successor of i .
- Partial order: a precedence relation which is both transitive ($\forall i, j, k, i \bullet j \text{ & } j \bullet k \Rightarrow i \bullet k$) and irreflexive ($\forall x \neg x \bullet x$).
- Acyclic graph: a directed graph with no directed cycles

*Figure 6.37: An AOV network



課程編號	課程名稱	先修課程
C1	程式I	無
C2	離散數學	無
C3	資料結構	C1, C2
C4	微積分I	無
C5	微積分II	C4
C6	線性代數	C5
C7	演算法分析	C3, C6
C8	組合語言	C3
C9	作業系統	C7, C8
C10	程式語言	C7
C11	編譯器設計	C10
C12	人工智慧	C7
C13	計算機理論	C7
C14	平行演算法	C13
C15	數值分析	C5

Topological order:
linear ordering of vertices
of a graph

$\forall i, j \text{ if } i \text{ is a predecessor of } j, \text{ then } i \text{ precedes } j \text{ in the linear ordering}$

C1, C2, C4, C5, C3, C6, C8,
C7, C10, C13, C12, C14, C15,
C11, C9

C4, C5, C2, C1, C6, C3, C8,
C15, C7, C9, C10, C11, C13,
C12, C14

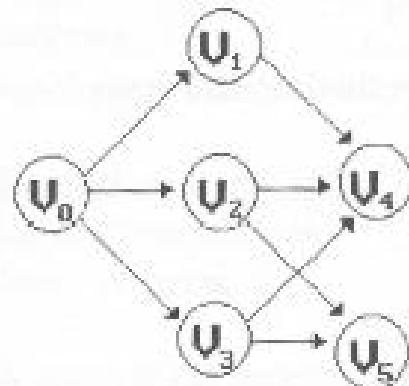
***Program 6.13: Topological sort**

```
for (i = 0; i <n; i++) {  
    if every vertex has a predecessor {  
        fprintf(stderr, "Network has a cycle. \n ");  
        exit(1);  
    }  
    pick a vertex v that has no predecessors;  
    output v;  
    delete v and all edges leading out of v  
    from the network;  
}
```

*Figure 6.38: Simulation of Program 6.13 on an AOV network

1. v0 no predecessor

delete v0->v1, v0->v2, v0->v3



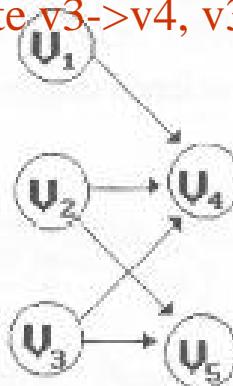
(a) initial

2. v1, v2, v3 no

predecessor

select v3

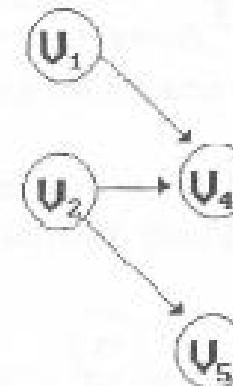
delete v3->v4, v3->v5



(b) v0

3. select v2

delete v2->v4, v2->v5



(c) v3

4. select v5



(d) v2

5. select v1
delete v1->v4



(e) v5

(f) v1

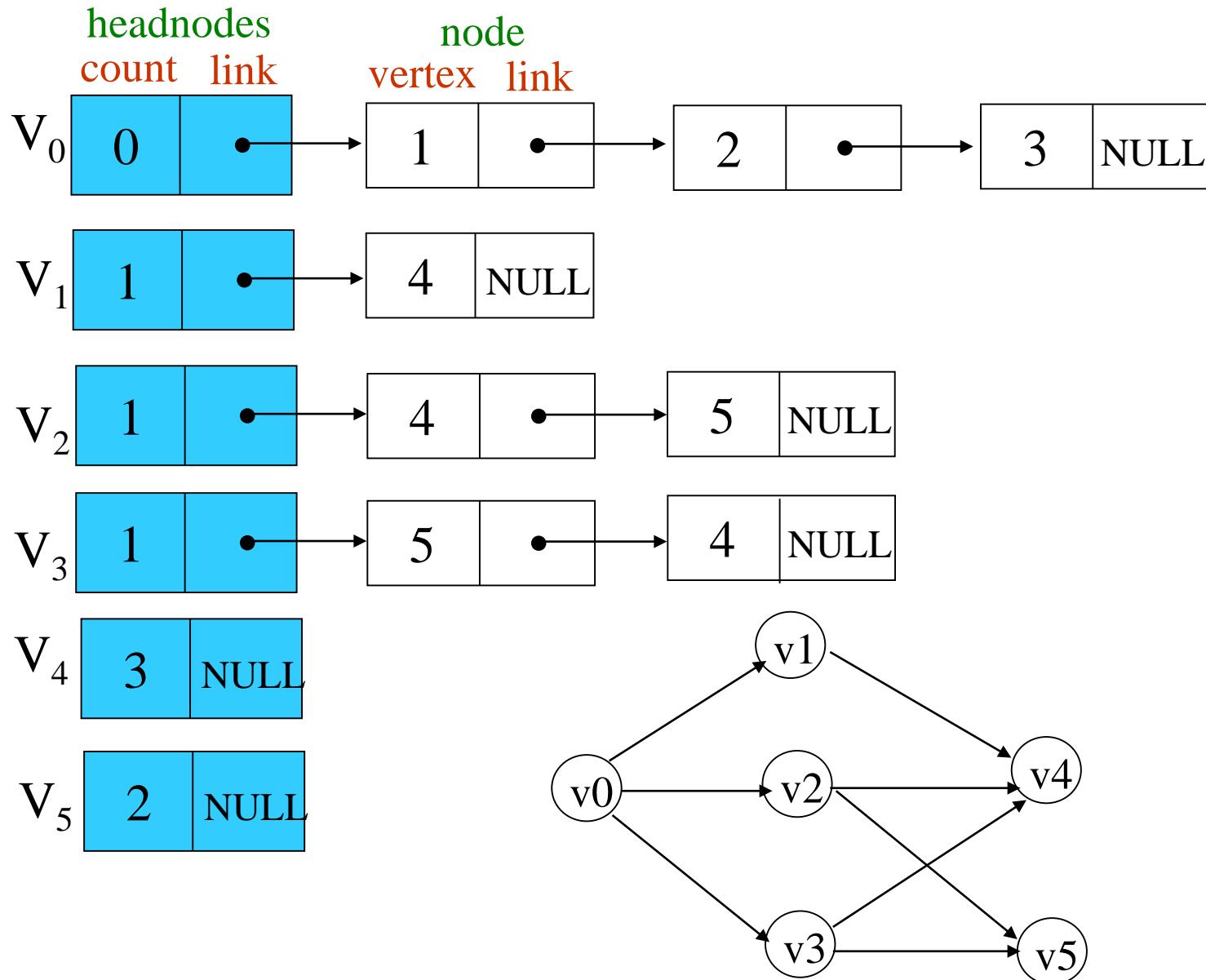
(g) v4

Topological order generated: v0, v3, v2, v5, v1, v4

Issues in Data Structure Consideration

- Decide whether a vertex has any predecessors.
 - Each vertex has a count.
- Decide a vertex together with all its incident edges.
 - Adjacency list

***Figure 6.39:** Internal representation used by topological sorting algorithm



```
typedef struct node *node_pointer;
typedef struct node {
    int vertex;
    node_pointer link;
};
typedef struct {
    int count;
    node_pointer link;
} hdnodes;
hdnodes graph[MAX_VERTICES];
```

*Program 6.14: Topological sort

```
void topsort (hdnodes graph [] , int n)
{
    int i, j, k, top;
    node_pointer ptr;
    /* create a stack of vertices with no predecessors */
    top = -1;
    for (i = 0; i < n; i++)
        if (!graph[i].count) {  no predecessors, stack is linked through count field
            graph[i].count = top;
            top = i;
        }
    for (i = 0; i < n; i++)
        if (top == -1) {
            fprintf(stderr, "\n Network has a cycle. Sort terminated. \n");
            exit(1);
        }
```

O(n)

```

}

else {
    j = top; /* unstack a vertex */
    top = graph[top].count;
    printf("v%d, ", j);
    for (ptr = graph[j].link; ptr ; ptr = ptr ->link ){
        /* decrease the count of the successor vertices of j */
        k = ptr ->vertex;
        graph[k].count--;
        if (!graph[k].count) {
            /* add vertex k to the stack*/
            graph[k].count = top;
            top = k;
        }
    }
}

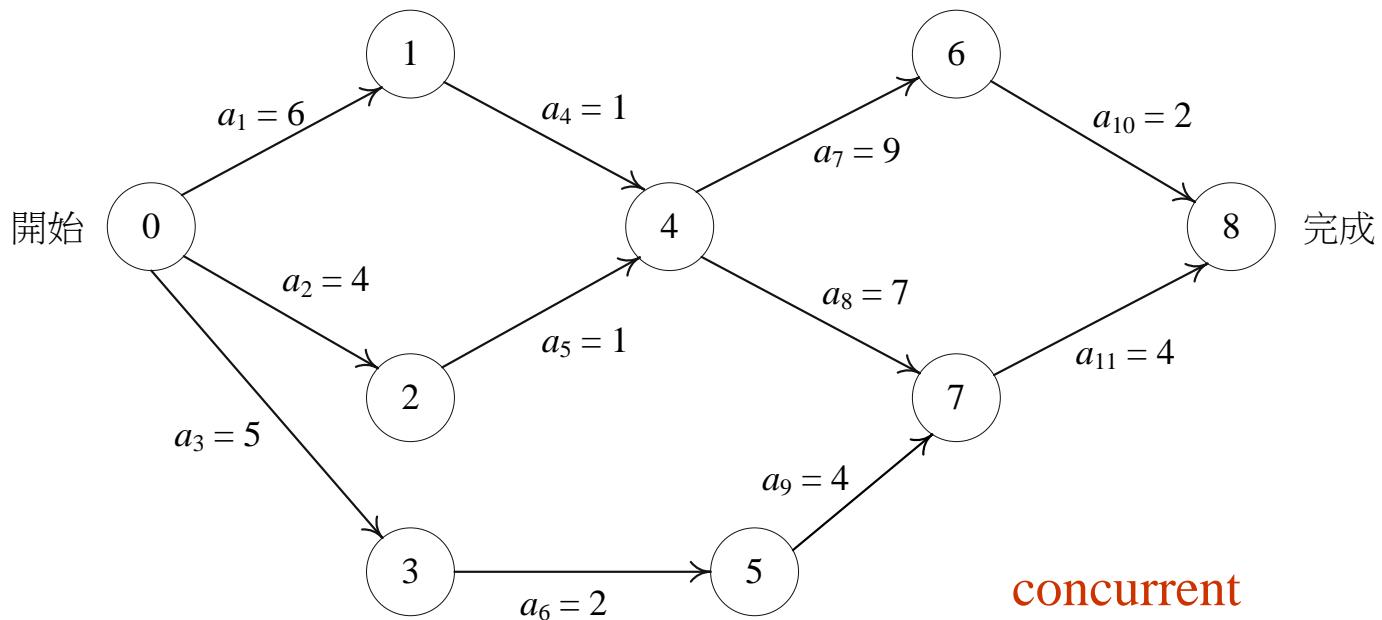
```

$O(e)$

$O(e+n)$

Activity on Edge (AOE) Networks

- Directed edge
 - tasks or activities to be performed
- Vertex
 - events which signal the completion of certain activities
- Number
 - time required to perform the activity



(Fig. 6.40)

事件	解釋
0	計劃開始
1	活動 a_1 完成
4	活動 a_4 和 a_5 完成
7	活動 a_8 和 a_9 完成
8	計畫完成

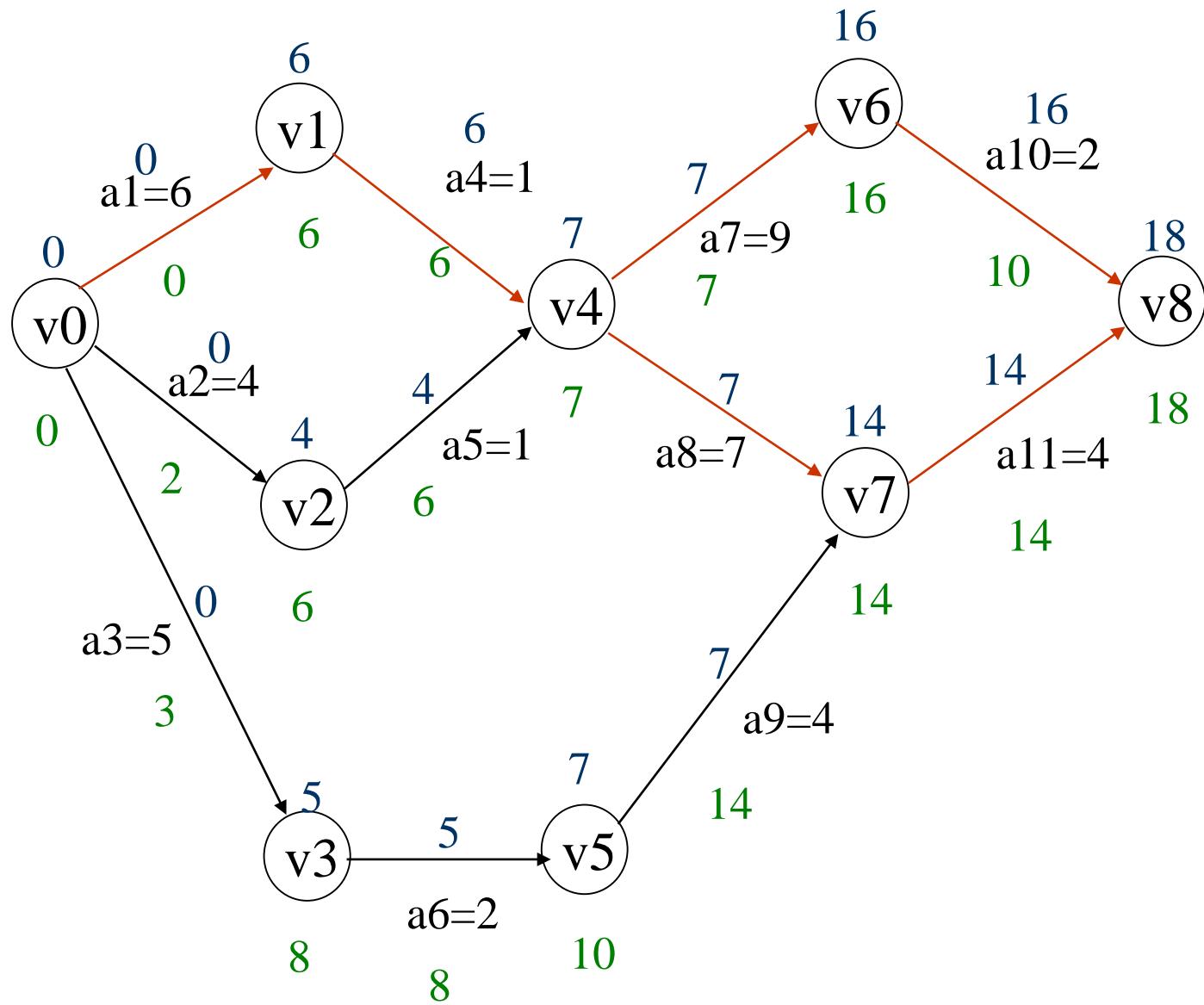
Application of AOE Network

- Evaluate performance
 - minimum amount of time
 - activity whose duration time should be shortened
 - ...
- Critical path
 - **a path that has the longest length**
 - minimum time required to complete the project
 - v0, v1, v4, v7, v8 or v0, v1, v4, v6, v8

AOE

- Earliest time that vi can occur
 - the length of the longest path from v0 to vi
 - the earliest start time for all activities leaving vi
 - $\text{early}(7) = \text{early}(8) = 7$
- Latest time of **activity**
 - the latest time the activity may start without increasing the project duration
 - $\text{late}(6) = 8, \text{late}(8) = 7$
- Critical activity
 - an activity for which $\text{early}(i) = \text{late}(i)$
 - $\text{early}(7) = \text{late}(7) = 14$
- $\text{late}(i)-\text{early}(i)$
 - measure of how critical an activity is
 - $\text{late}(5)-\text{early}(5) = 10-7=3$

earliest, early, latest, late



Determine Critical Paths

- Delete all noncritical activities
- Generate all the paths from the start to finish vertex.

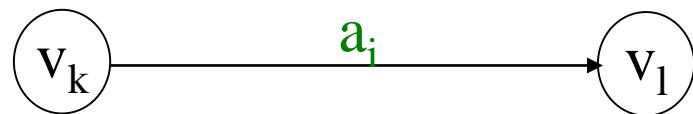
Calculation of Earliest Times

- **earliest[j]**
 - the earliest event occurrence time

earliest[0]=0

earliest[j]= $\max_{i \in p(j)} \{ \text{earliest}[i] + \text{duration of } <i,j> \}$

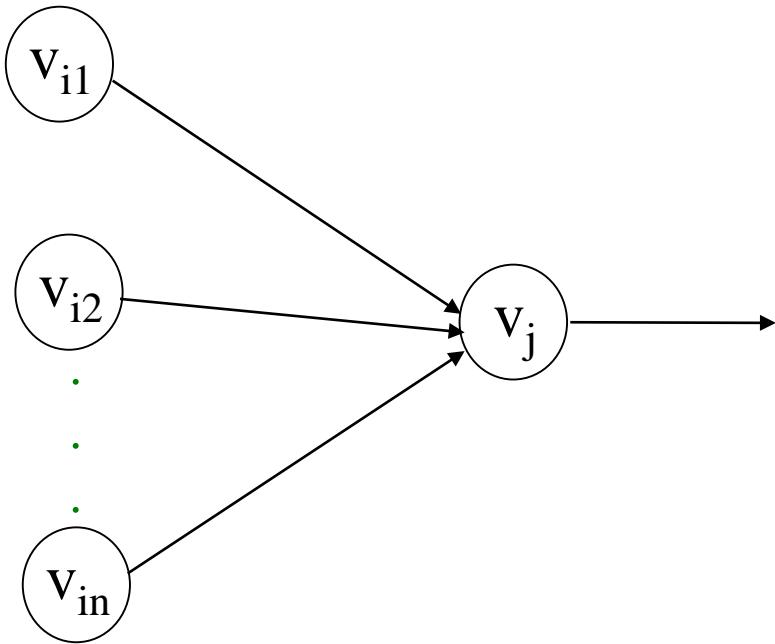
- **latest[j]**
 - the latest event occurrence time



early(i)=earliest(k)

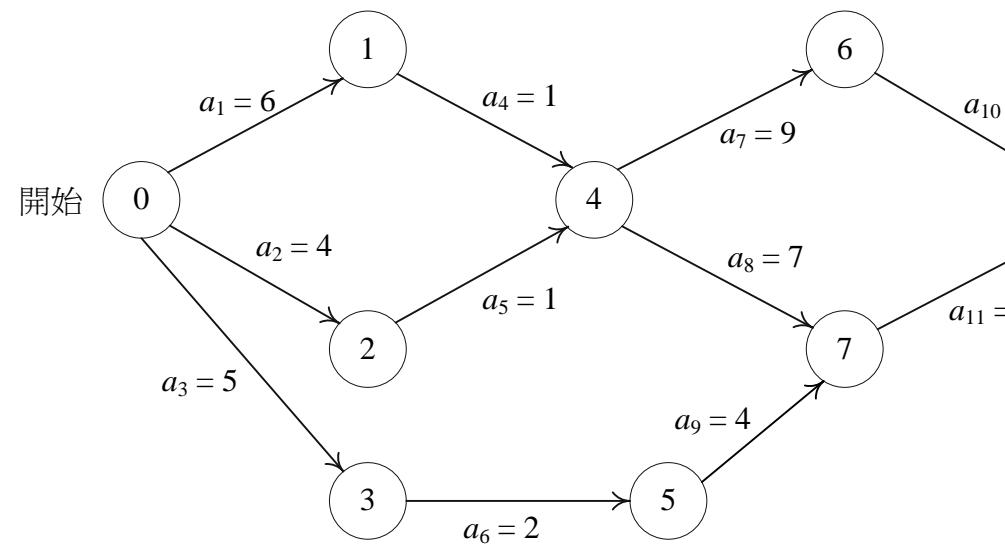
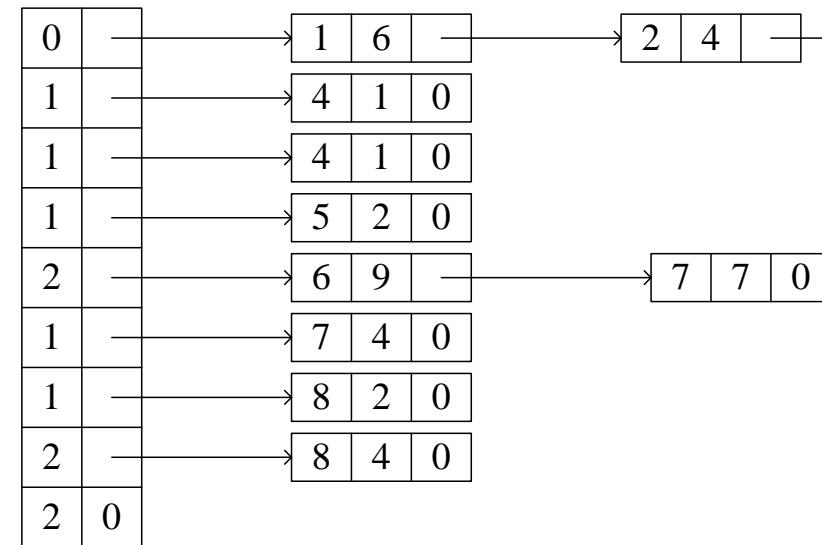
late(i)=latest(l)-duration of a_i

forward stage



```
if (earliest[k] < earliest[j]+ptr->duration)  
earliest[k]=earliest[j]+ptr->duration
```

count first



(a) 圖 6.40(a) 的相鄰串列

ee	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]	[8]	堆疊
起始	0	0	0	0	0	0	0	0	0	[0]
輸出 0	0	6	4	5	0	0	0	0	0	[3, 2, 1]
輸出 3	0	6	4	5	0	7	0	0	0	[5, 2, 1]
輸出 5	0	6	4	5	0	7	0	11	0	[2, 1]
輸出 2	0	6	4	5	5	7	0	11	0	[1]
輸出 1	0	6	4	5	7	7	0	11	0	[4]
輸出 4	0	6	4	5	7	7	16	14	0	[7, 6]
輸出 7	0	6	4	5	7	7	16	14	18	[6]
輸出 6	0	6	4	5	7	7	16	14	18	[8]
輸出 8										

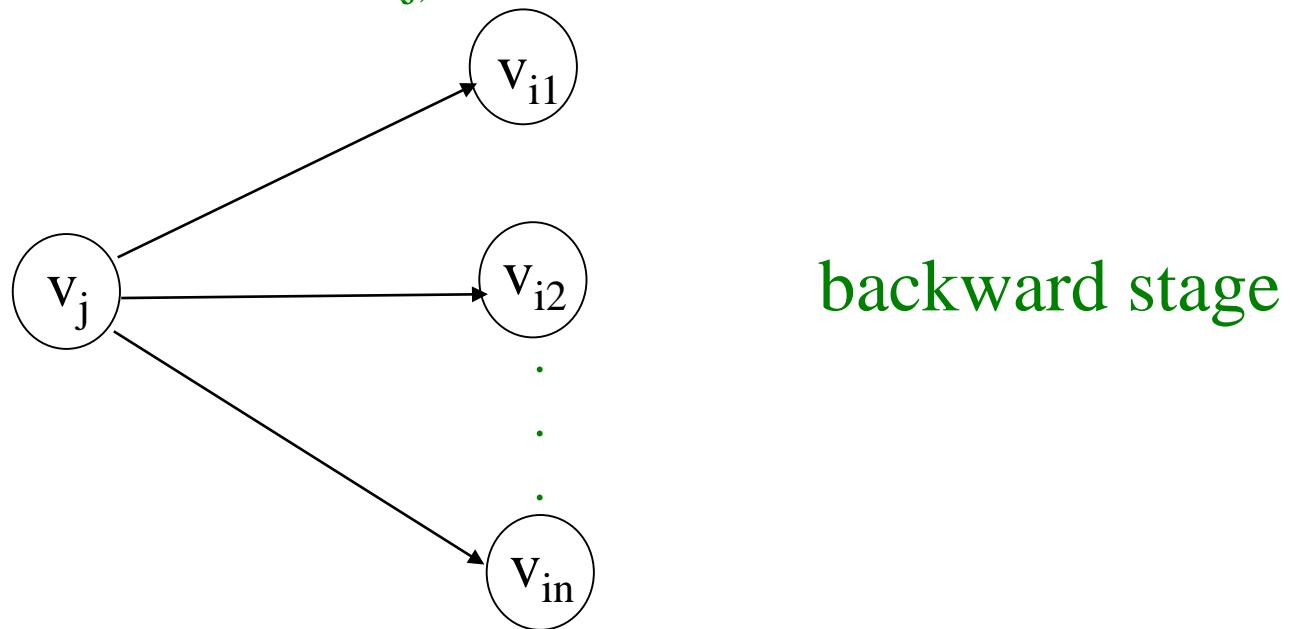
(b) ee 的計算

Calculation of Latest Times

□ $\text{latest}[j]$

– the latest event occurrence time

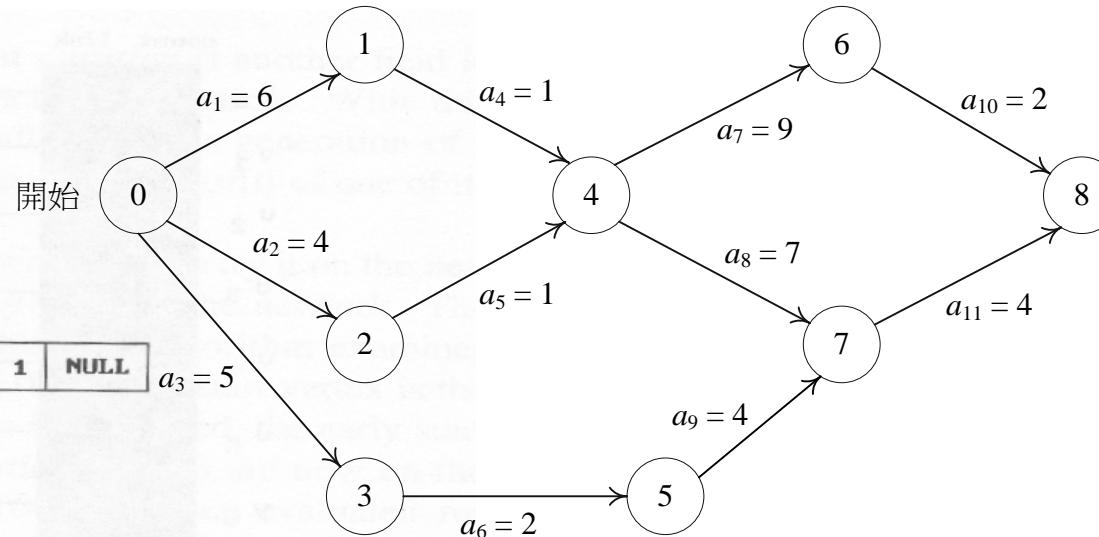
$$\text{latest}[j] = \min_{i \in s(j)} \{\text{latest}[i] - \text{duration of } <j,i>\}$$



```
if (latest[k] > latest[j]-ptr->duration)  
    latest[k]=latest[j]-ptr->duration
```

*Figure 6.43: Computing latest for AOE network of Figure 6.41(a)

	count	link	vertex	dur	link
v ₀	3	NULL			
v ₁	1	•	0	6	NULL
v ₂	1	•	0	4	NULL
v ₃	1	•	0	5	NULL
v ₄	2	•	1	1	• → [2, 1, NULL]
v ₅	1	•	3	2	NULL
v ₆	1	•	4	9	NULL
v ₇	1	•	4	7	• → [5, 4, NULL]
v ₈	0	•	6	2	• → [7, 4, NULL]



(a) Inverted adjacency lists for AOE network of Figure 6.41(a)

Latest	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]	[8]	Stack
initial	18	18	18	18	18	18	18	18	18	[8]
output v ₈	18	18	18	18	18	18	16	14	18	[7, 6]
output v ₇	18	18	18	18	7	10	16	14	18	[5, 6]
output v ₅	18	18	18	18	7	10	16	14	18	[3, 6]
output v ₃	3	18	18	8	7	10	16	14	18	[6]
output v ₆	3	18	18	8	7	10	16	14	18	[4]
output v ₄	3	6	6	8	7	10	16	14	18	[2, 1]
output v ₂	2	6	6	8	7	10	16	14	18	[1]
output v ₁	0	6	6	8	7	10	16	14	18	[0]

(b) Computation of *latest*

***Figure 6.43(continued):** Computing latest of AOE network of Figure 6.41(a)

latest[8]=earliest[8]=18

latest[6]= $\min\{le[8] - 2\} = 16$

latest[7]= $\min\{le[8] - 4\} = 14$

latest[4]= $\min\{le[6] - 9; le[7] - 7\} = 7$

latest[1]= $\min\{le[4] - 1\} = 6$

latest[2]= $\min\{le[4] - 1\} = 6$

latest[5]= $\min\{le[7] - 4\} = 10$

latest[3]= $\min\{le[5] - 2\} = 8$

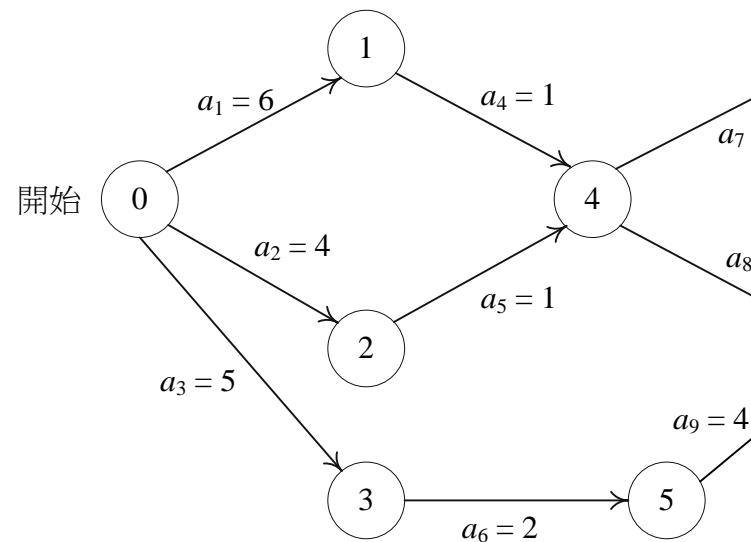
latest[0]= $\min\{le[1] - 6; le[2] - 4; le[3] - 5\} = 0$

(c) Computation of latest from Equation (6.3) using a reverse topological order

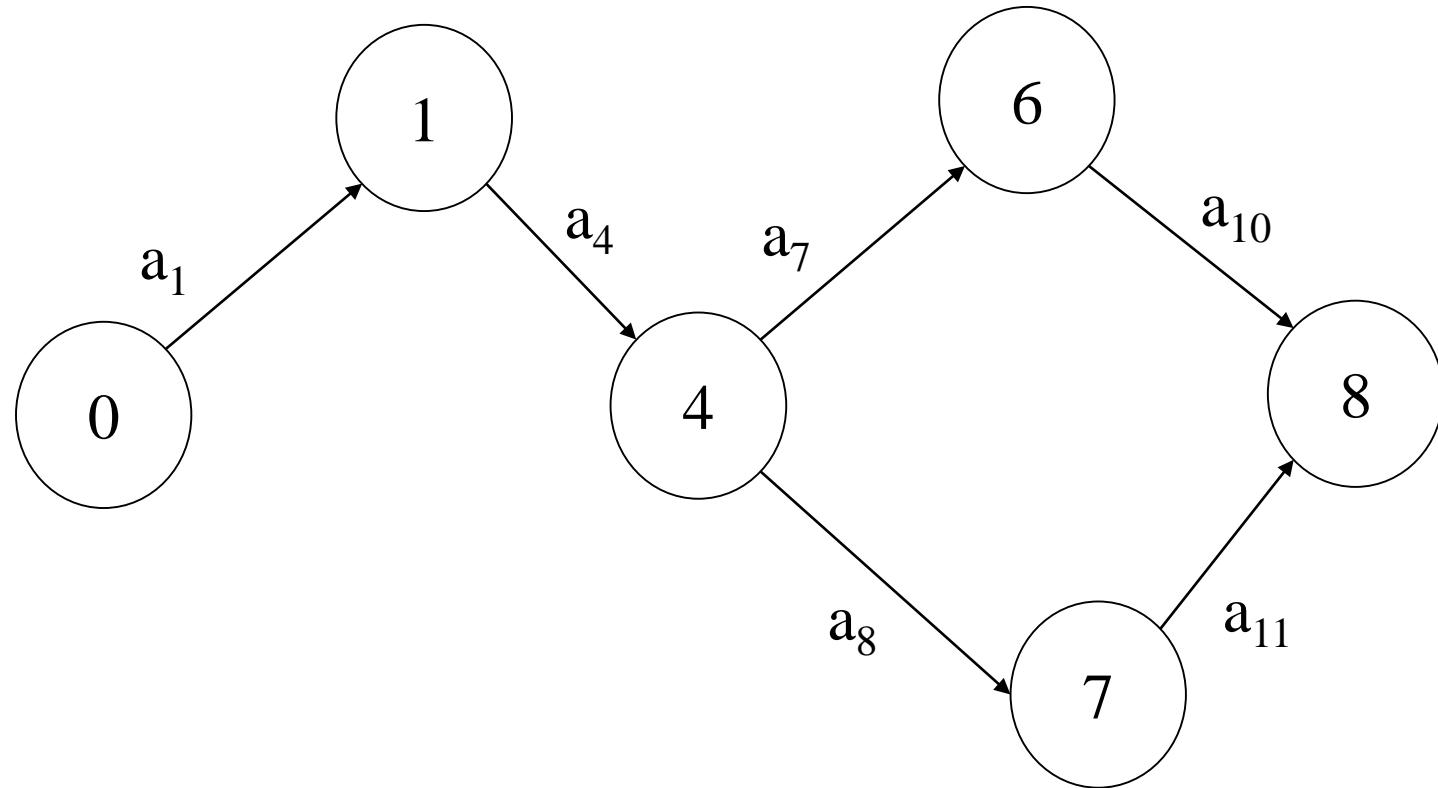
*Figure 6.42: Early, late and critical values

Activity	Early	Late	Late-Early	Critical
a ₁	0	0	0	Yes
a ₂	0	2	2	No
a ₃	0	3	3	No
a ₄	6	6	0	Yes
a ₅	4	6	2	No
a ₆	5	8	3	No
a ₇	7	7	0	Yes
a ₈	7	7	0	Yes
a ₉	7	10	3	No
a ₁₀	16	16	0	Yes
a ₁₁	14	14	0	Yes

$$1 - e = 0$$



***Figure 6.43:**Graph with noncritical activities deleted



***Figure 6.45:** AOE network with unreachable activities

