

Fu-Yin Cherng

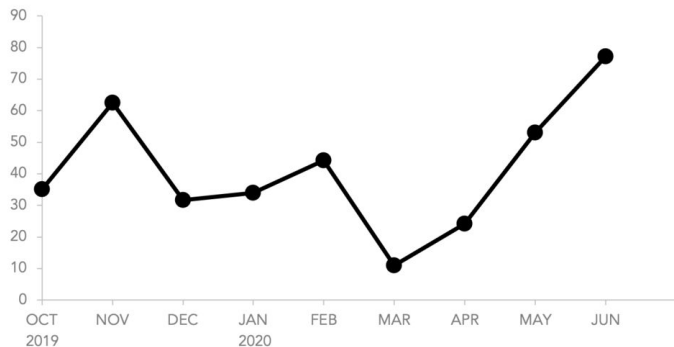
National Taiwan University

Graph

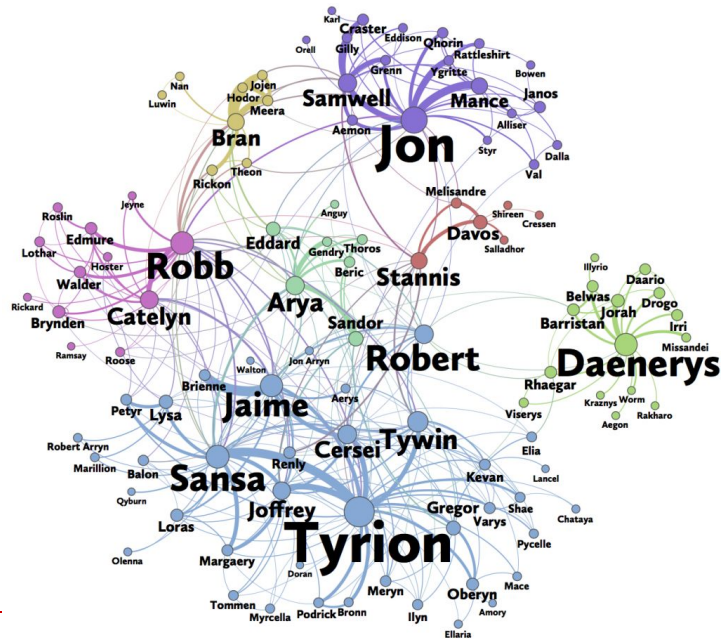
What is graph?

- a way to **illustrate** data & represent the **relationships** among data items

Produce sales
IN THOUSANDS (USD)



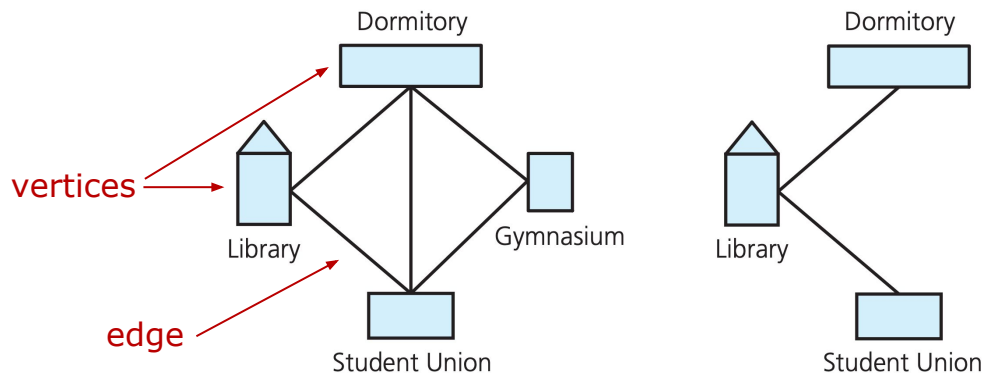
<https://www.storytellingwithdata.com/blog/2020/3/24/what-is-a-line-graph>



<https://predictivehacks.com/social-network-analysis-of-game-of-thrones/>

What is graph?

- General **definition** of graph G consists of **2 sets**
 - V : a set of vertices/nodes
 - E : a set of edges that connect the vertices
- **Subgraph**: a subset of a graph vertices and its edges



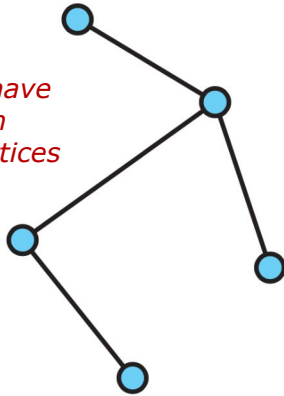
subgraph

Terminology of Graph

- **Adjacent vertices**
 - 2 vertices are adjacent if **joined by an edge**
- **Path**
 - a path between two vertices is a **sequence of edges** the begins at one vertex and ends at another vertex
- **Simple Path**
 - a path passes through a vertex only **once**
- **Cycle**
 - a path that begins and ends at the **same vertex**

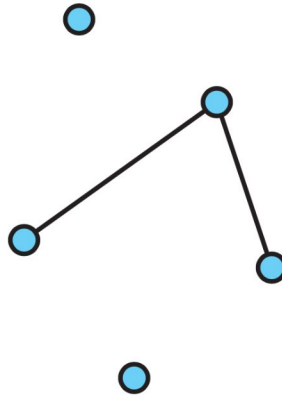
Terminology of Graph

*not necessarily have
an edge between
every pair of vertices*



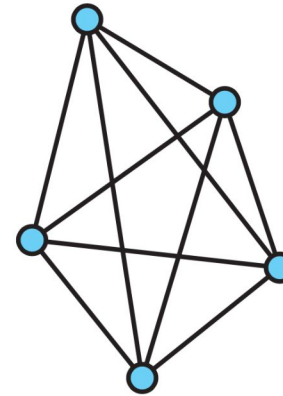
(a)

Connected Graph
each pair of distinct vertices has
a **path** between them



(b)

Disconnected Graph

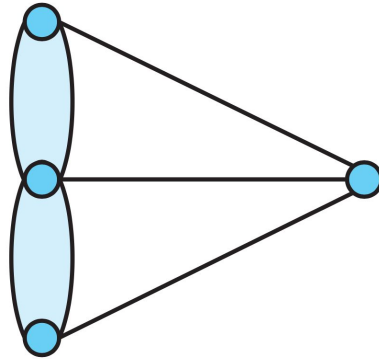


(c)

Complete Graph
each pair of distinct vertices has
an **edge** between them

Terminology of Graph

- A graph
 - edge cannot begin and end at the same vertex (self edge)
 - cannot have **duplicate edges** between vertices
- **Multigraph is not a graph** and can have multiple edges



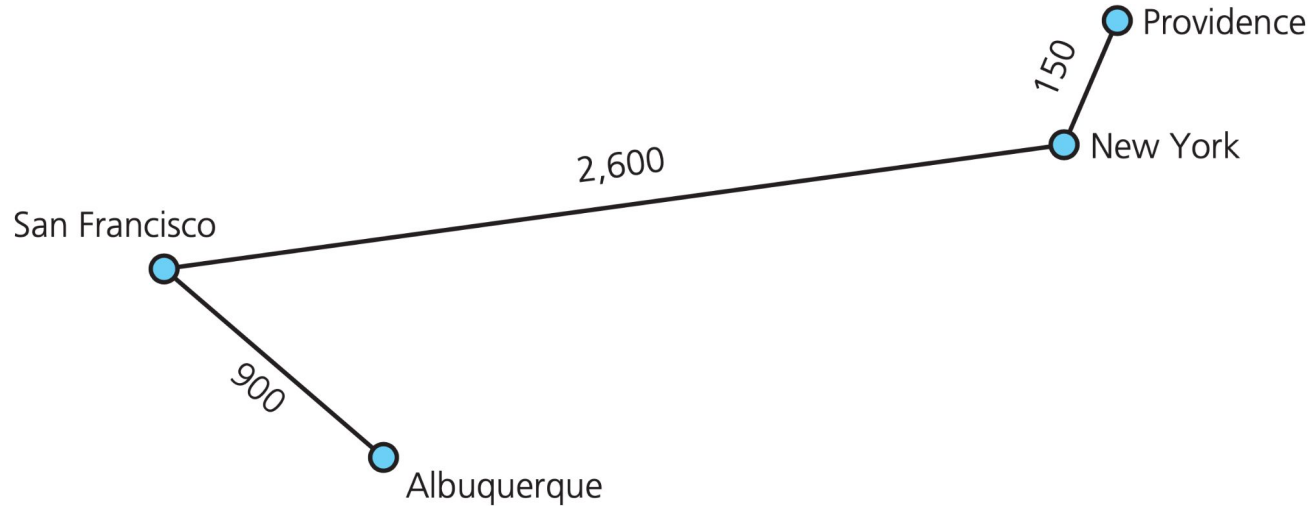
Multigraph



self edge
not allowed in a graph

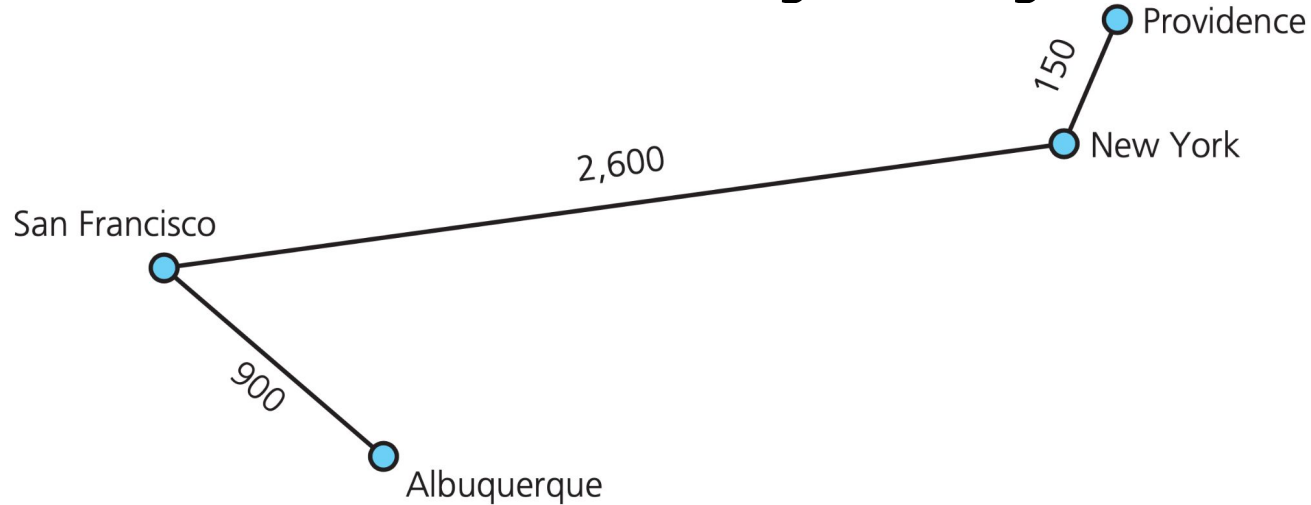
Terminology of Graph

- Weighted graph: edges with numeric labels



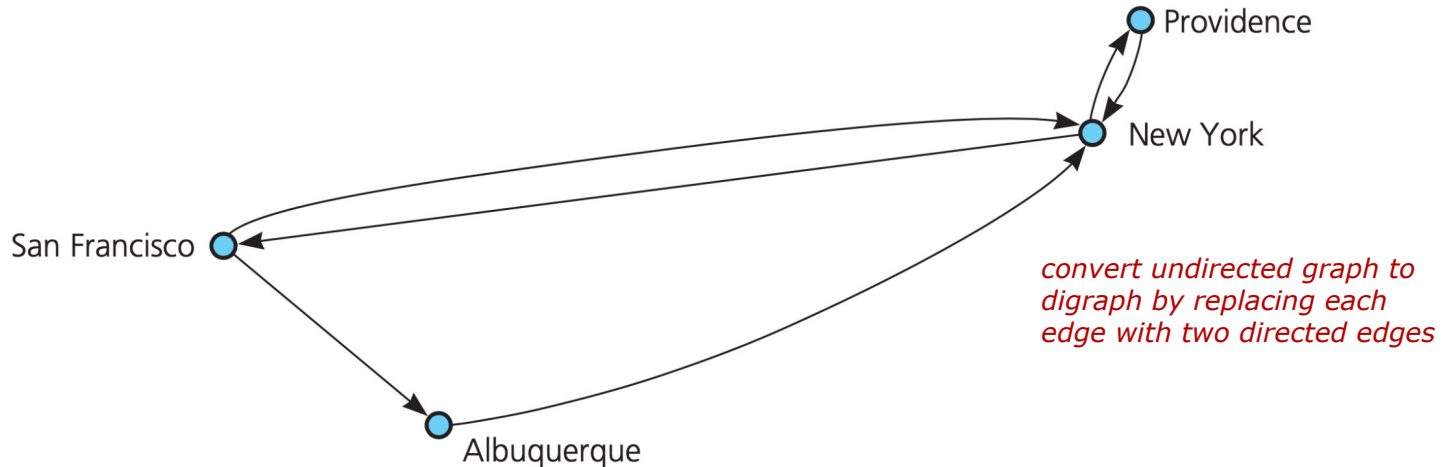
Terminology of Graph

- Undirected graphs: edges **do not indicate direction**, can **travel either direction** along the edges



Terminology of Graph

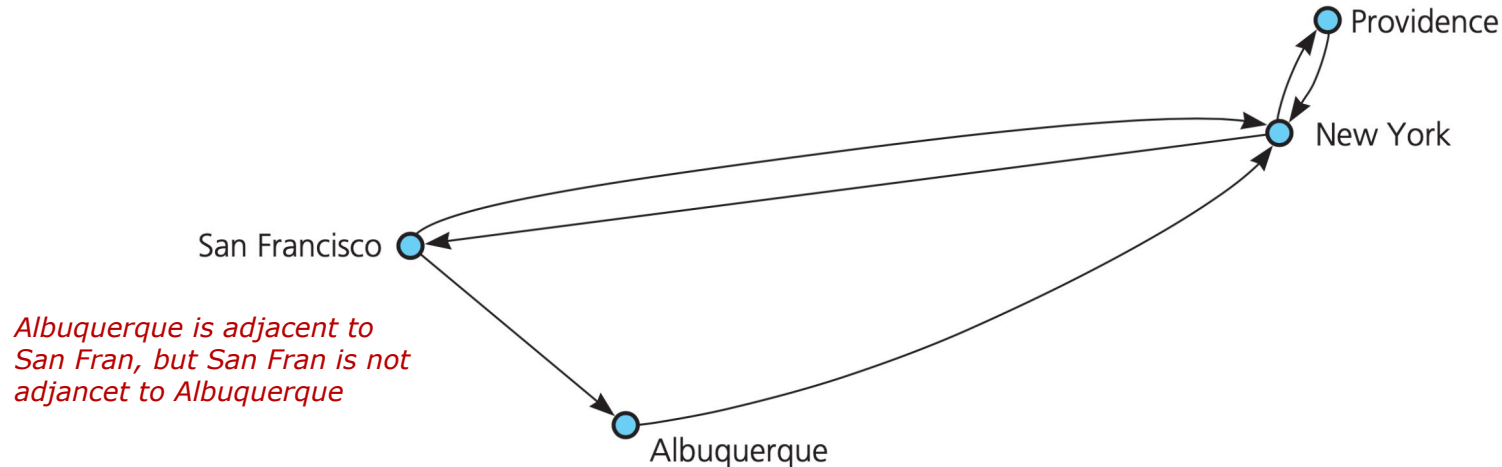
- Directed graph or digraph: graph with directed edge
 - can have 2 edges with different directions between a pair of vertices



Terminology of Graph

□ Directed graph or digraph

- **adjacent vertices**: if there is a directed edge from vertex x to vertex y , then y is adjacent to x ($x \rightarrow y$)
- y is successor of x , x is predecessor of y



Abstract data type: Graph

- define following **ADT graph's vertices contain value**
 - you can also design another ADT graph whose vertices don't contain value.
- many operations, for example
 - **Test** whether a graph is empty.
 - **Get** the number of vertices in a graph.
 - **Insert** an edge between two given vertices in a graph
 - **Retrieve** from a graph the vertex that contains a given value
 - ...

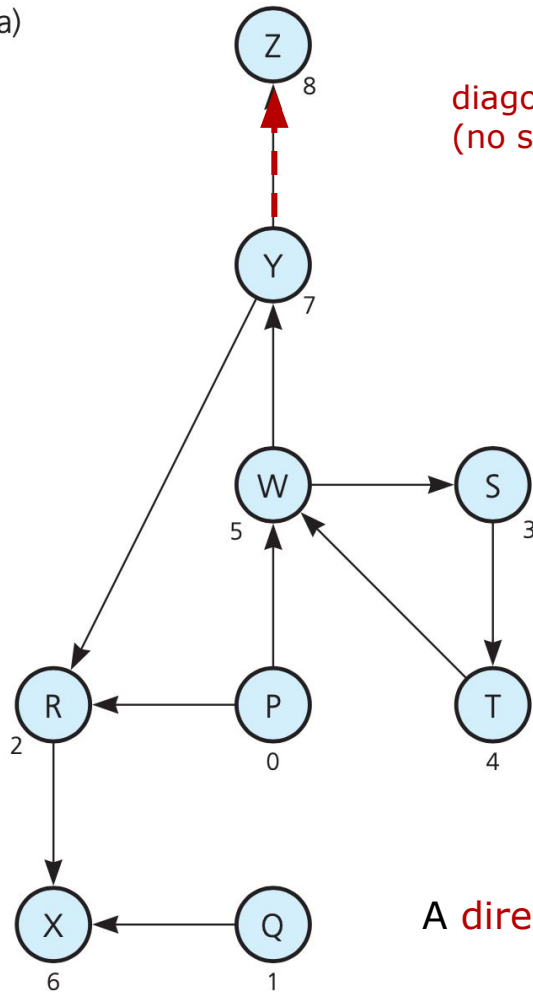
Abstract data type: Graph

```
//An interface for the ADT undirected, connected graph
template<class LabelType>
class GraphInterface{
    public:
        virtual int getNumVertices() const = 0;
        virtual int getNumEdges() const = 0;
        virtual bool add(LabelType start, LabelType end, int edgeWeight) = 0;
        virtual bool remove(LabelType start, LabelType end) = 0;
        virtual int getEdgeWeight(LabelType start, LabelType end) const = 0;
        virtual void depthFirstTraversal(LabelType start, void visit(LabelType&)) = 0;
        virtual void breadthFirstTraversal(LabelType start, void visit(LabelType&)) =
0;
}
```

Implementing Graphs: adjacency matrix

- use **adjacency matrix** or adjacency list
- for a graph with n vertices numbered $0, 1, \dots, n-1$
 - adjacent matrix: `matrix[n][n]`
 - unweighted:
 - `matrix[i][j]` is **1 (true)** if there is an edge from vertex i to vertex j , and **0 (false)** otherwise
 - weighted
 - `matrix[i][j]` is the **weight** that labels the edge from vertex i to vertex j , and **∞** otherwise

(a)



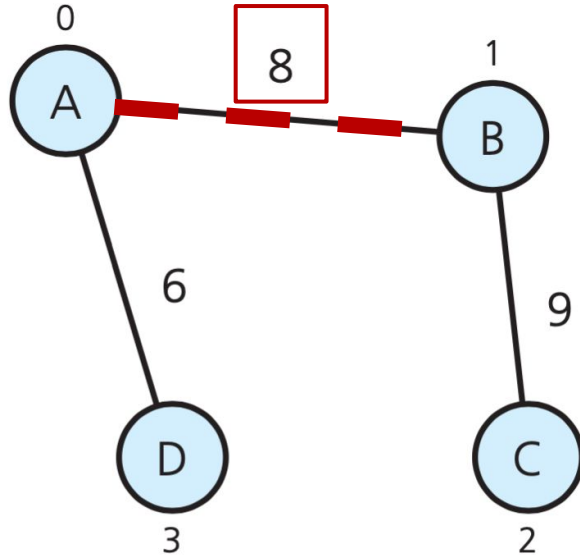
diagonal entries is 0
(no self edge)

(b)

| | | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
|---|---|---|---|---|---|---|---|---|---|---|
| | | P | Q | R | S | T | W | X | Y | Z |
| 0 | P | 0 | 0 | 1 | 0 | 0 | 1 | 0 | 0 | 0 |
| 1 | Q | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| 2 | R | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| 3 | S | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 0 |
| 4 | T | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 0 |
| 5 | W | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 | 0 |
| 6 | X | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 7 | Y | 0 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 1 |
| 8 | Z | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

A **directed + unweighted** graph and its adjacent matrix

(a)



(b)

undirected graph is symmetrical

| | | 0 | 1 | 2 | 3 |
|---|---|----------|----------|----------|----------|
| | | A | B | C | D |
| 0 | A | ∞ | 8 | ∞ | 6 |
| 1 | B | 8 | ∞ | 9 | ∞ |
| 2 | C | ∞ | 9 | ∞ | ∞ |
| 3 | D | 6 | ∞ | ∞ | ∞ |

A **undirected + weighted** graph and its adjancet matrix

Implementing Graphs: adjacency matrix

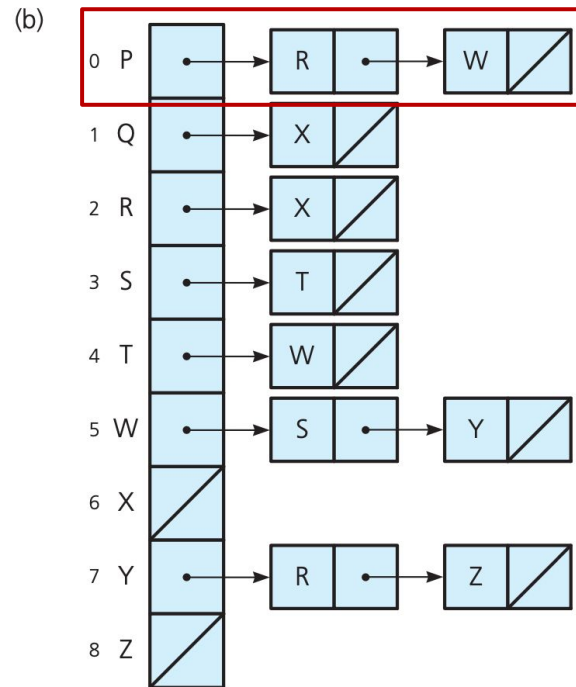
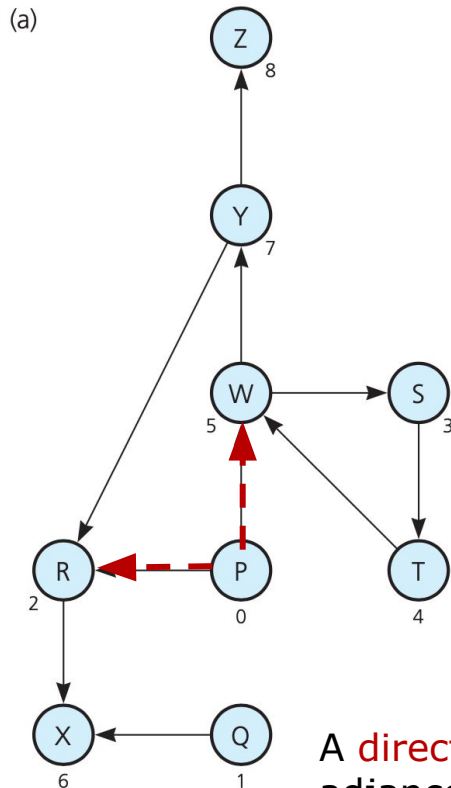
- use **second array** to represent the **n vertex values**
- `values[i]` is the value in vertex `i`

Implementing Graphs: adjacency list

- use adjacency matrix or adjacency list
- for a graph with n vertices numbered $0, 1, \dots, n-1$
 - adjacency list: consists of n linked chains
 - see example directly

Implementing Graphs: adjacency list

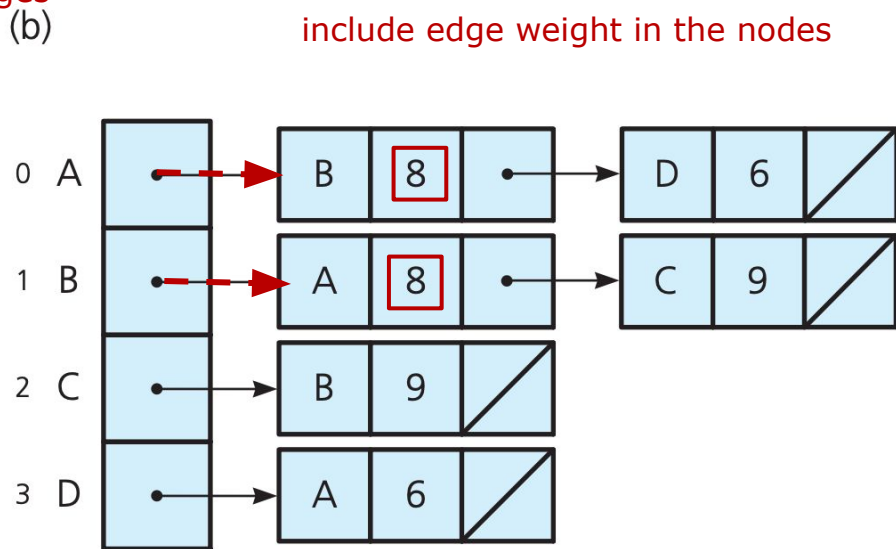
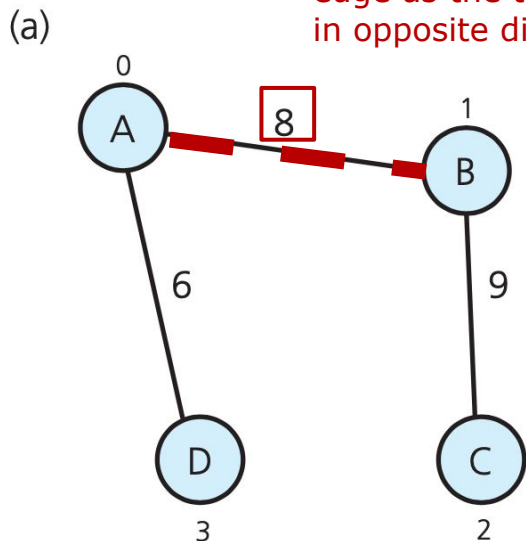
- the i th linked chain has nodes for vertices with edge to vertex i
- If the vertex has no value, the node needs to contain some indication of the vertex's identity



A directed + unweighted graph and its adjancet list

Implementing Graphs: adjacency list

for undirected graph, treat each edge as the two directed edges in opposite directions



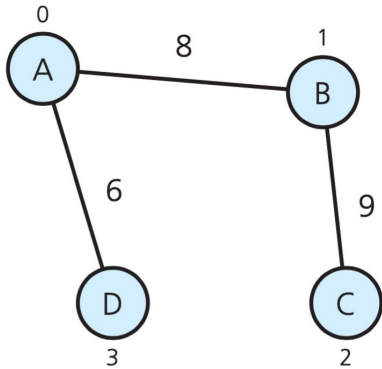
A undirected + weighted graph and its adjancet list

Implementing Graphs: comparison

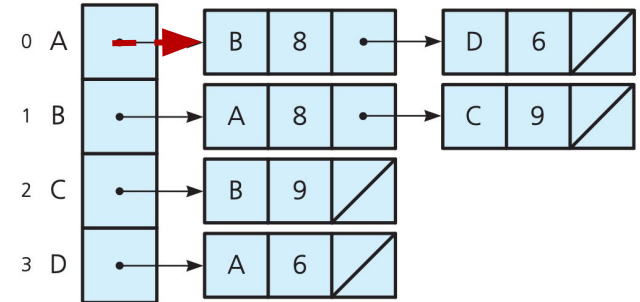
- Which one is better? adjacency matrix or list?
- depend on **applications** and operations

Implementing Graphs: comparison

- For example, **adjacency matrix is more efficient** for determining **whether there is an edge** from vertex i to vertex j
 - only need to check the value of `matrix[i][j]`
 - if an adjacency list, need to traverse i th linked chain



| | 0 | 1 | 2 | 3 |
|-----|----------|----------|----------|----------|
| | A | B | C | D |
| 0 A | ∞ | 8 | ∞ | 6 |
| 1 B | 8 | ∞ | 9 | ∞ |
| 2 C | ∞ | 9 | ∞ | ∞ |
| 3 D | 6 | ∞ | ∞ | ∞ |

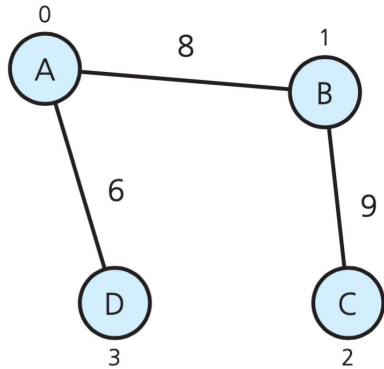


- edge between A and C? ——— check value in `[A][C]` or `[C][A]`
- traverse 2th linked chain to check if there is node for C

Implementing Graphs: comparison

- On the other hand, **adjacency list** is more efficient for finding all **vertices** adjacent to a given **vertex i**
 - directly traverse the *i*th linked chain.

(a)

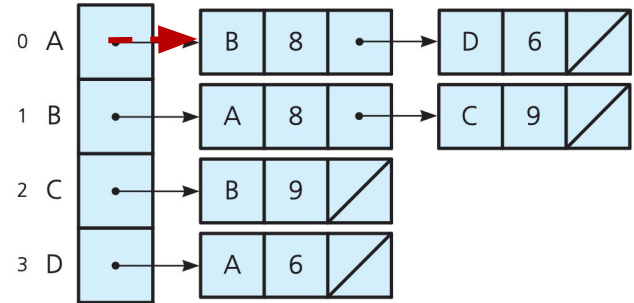


— find adjacency of A

(b)

| | 0 | 1 | 2 | 3 |
|-----|----------|----------|----------|----------|
| | A | B | C | D |
| 0 A | ∞ | 8 | ∞ | 6 |
| 1 B | 8 | ∞ | 9 | ∞ |
| 2 C | ∞ | 9 | ∞ | ∞ |
| 3 D | 6 | ∞ | ∞ | ∞ |

— traverse all columns in 0th row



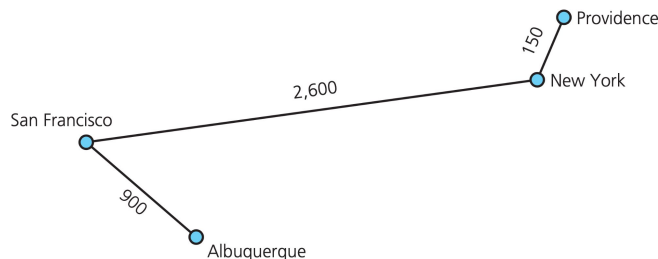
— traverse 0th linked chain with 2 nodes

Implementing Graphs: comparison

- storage space requirement of a directed graph with n node
 - adjacency matrix always has $n*n$ entries
 - adjacency list has number of nodes equals the number of edges in a directed graph
- adjacency list often requires less storage than an adjacency matrix.

Implementing Graphs: comparison

- choosing a graph implementation for a particular application
 - **what operations** you will perform most **frequently**
- for example, the **flight map** problem, most frequent operation was to **find all** cities (vertices) **adjacent** to given city
 - **adjacency list** would be more efficient



Graph Traversal

- **traverse** all vertices that **it can reach**
- **mark** each vertex during visit and **only visit a vertex once**
 - to **prevent loop indefinitely** due to a **cycle** in a graph
- **2 basic graph-traversal algorithms**
 - Depth-First Search (DFS)
 - Breadth-First Search (BFS)

Depth-First Search (DFS)

- 深度優先
- From vertex v , DFS traversal goes **as far (deep) as** possible from the vertex v before backing up
- After visiting a vertex, a **DFS visits an unvisited adjacent** vertex if possible.

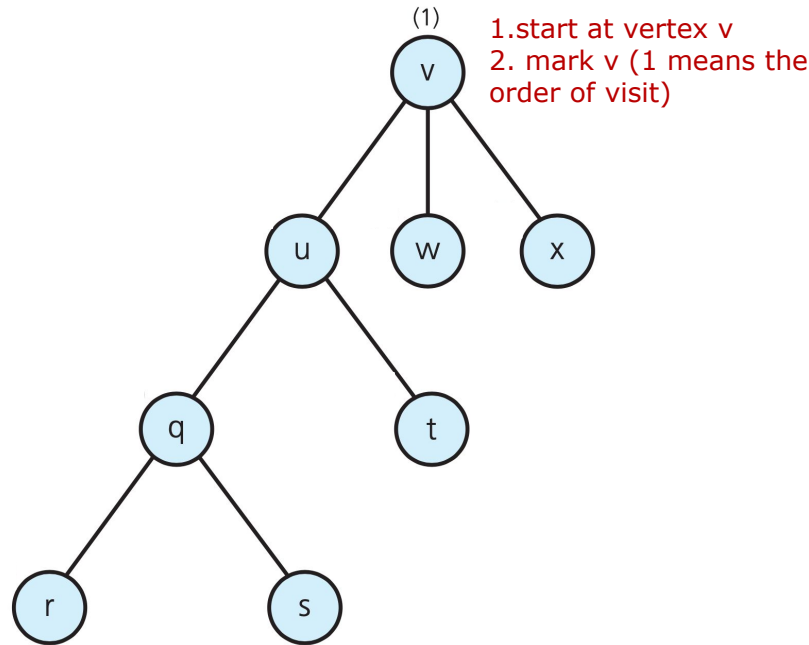
Depth-First Search (DFS)

DFS strategy has a recursive form

```
// Traverses a graph beginning at vertex v by using a  
// depth-first search: Recursive version.  
dfs(v: Vertex)  
    Mark v as visited  
    for (each unvisited vertex u adjacent to v)  
        dfs(u)
```

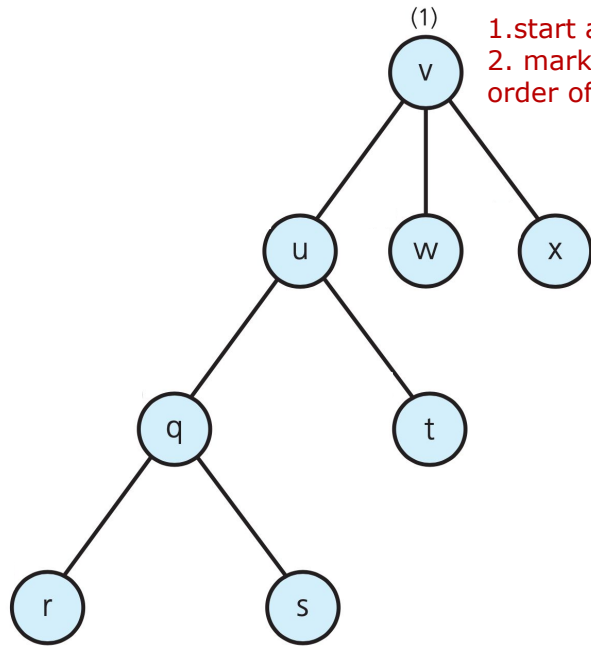
- don't specify the order of which it should visit the vertices adjacent to v
- Could visit the adjacent vertices in sorted order alphabetically or numerically

Depth-First Search (DFS): example

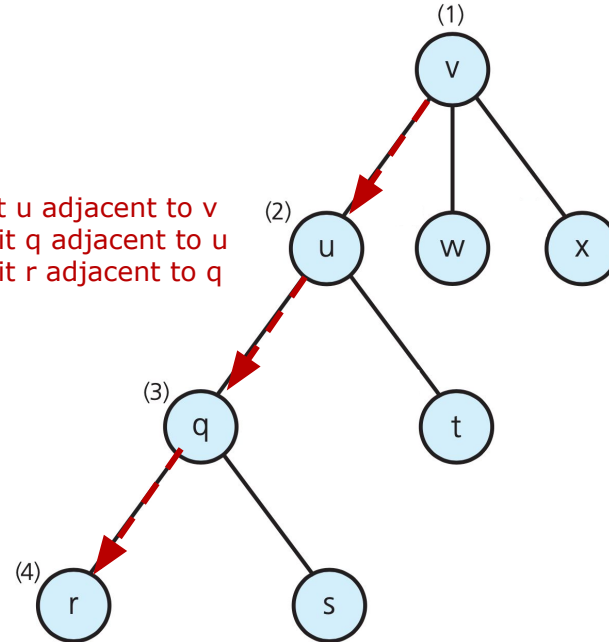


- undirected + unweighted **graph**

Depth-First Search (DFS): example



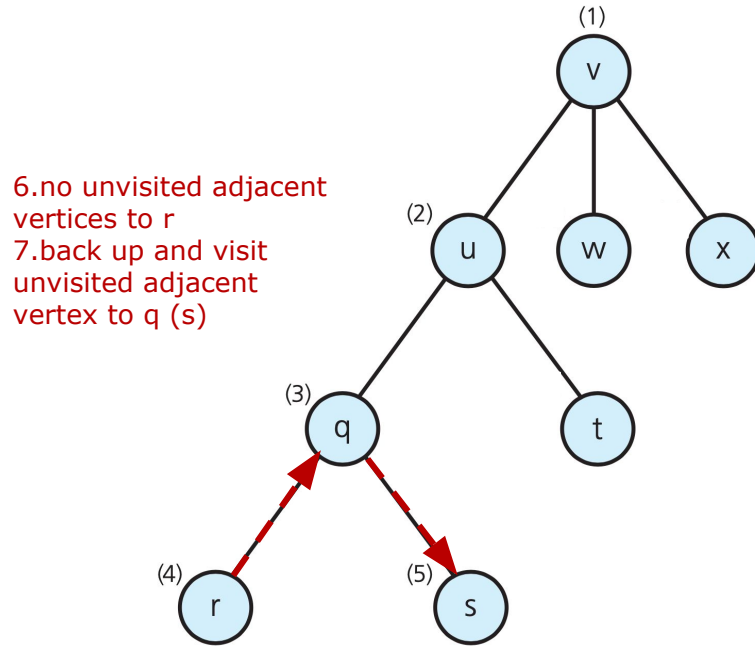
1. start at vertex v
2. mark v (1 means the order of visit)



3. visit u adjacent to v
4. visit q adjacent to u
5. visit r adjacent to q

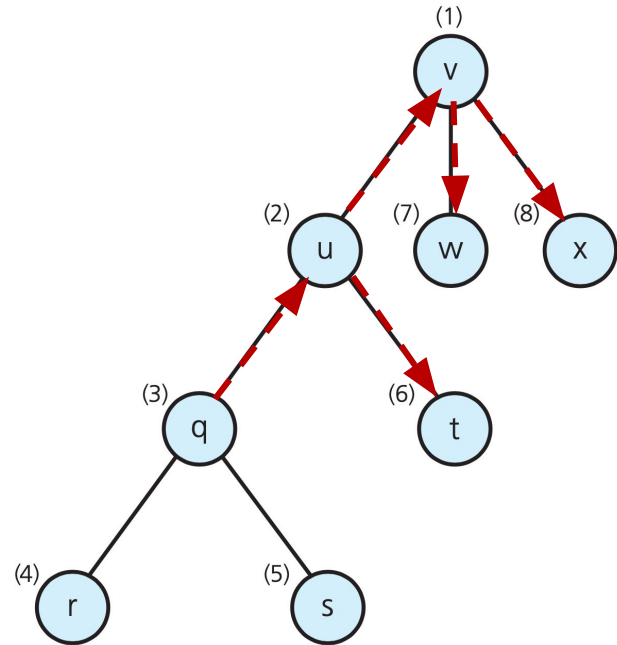
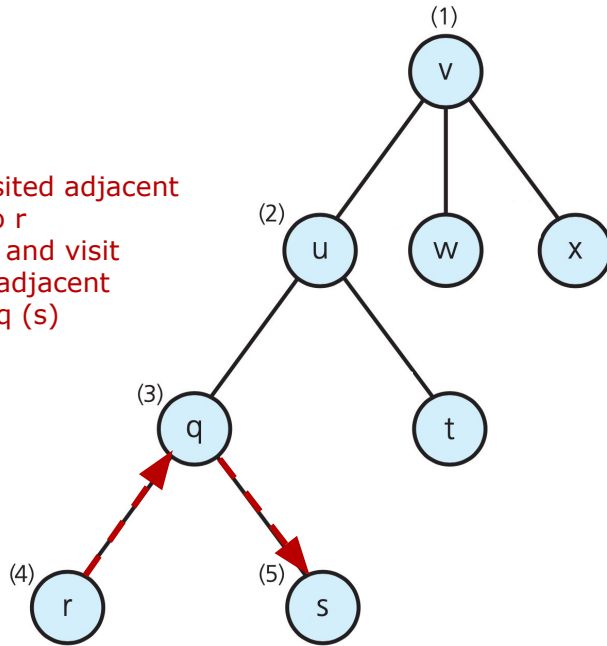
- undirected + unweighted **graph**

Depth-First Search (DFS): example



Depth-First Search (DFS): example

6.no unvisited adjacent vertices to r
7.back up and visit unvisited adjacent vertex to q (s)

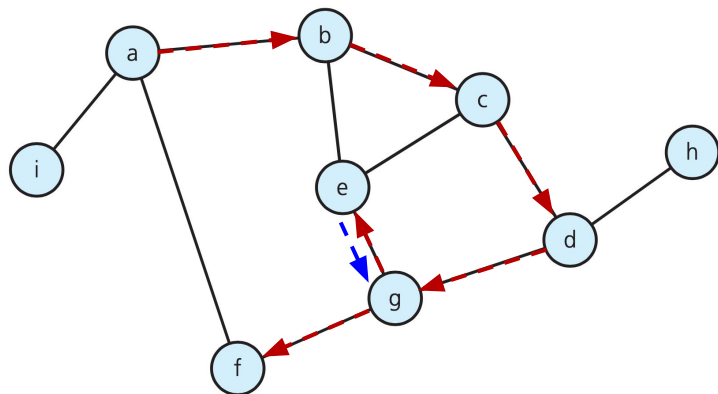


Depth-First Search (DFS)

DFS strategy has a **iterative** version using a **stack**

```
// Traverses a graph beginning at vertex v by using a
// depth-first search: Iterative version.
dfs(v: Vertex)
    s = a new empty stack
    s.push(v)
    Mark v as visited
    while (!s.isEmpty()){
        if (no unvisited vertices are adjacent to the vertex on the top of the stack)
            s.pop() // Backtrack
        else{
            Select an unvisited vertex u adjacent to vertex on the top of the stack
            s.push(u) // vertex u on the top of the stack
            Mark u as visited
        }
    }
}
```


Depth-First Search (DFS): example 2



order of visit: a, b, c, d, e, f, h, i

| Node visited | Stack (bottom to top) |
|--------------|---|
| a | a |
| b | a b |
| c | a b c |
| d | a b c d |
| g | a b c d g |
| e | a b c d g e <i>no unvisited vertex adjacent to e</i> |
| (backtrack) | a b c d g <i>pop e, so the top vertex in stack is g</i> |
| f | a b c d g f <i>visit f adjacent to g</i> |
| (backtrack) | a b c d g <i>pop f, so the top vertex is g</i> |
| (backtrack) | a b c d <i>pop g, so the top vertex is d</i> |
| h | a b c d h |
| (backtrack) | a b c d |
| (backtrack) | a b c |
| (backtrack) | a b |
| (backtrack) | a |
| i | a i |
| (backtrack) | a |
| (backtrack) | (empty) |

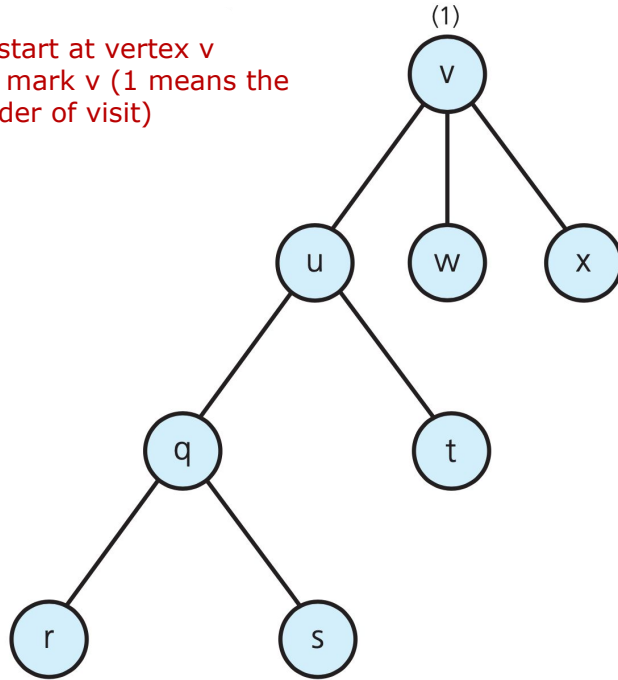
The results of a DFS traversal,
beginning at vertex a

Breadth-First Search (BFS)

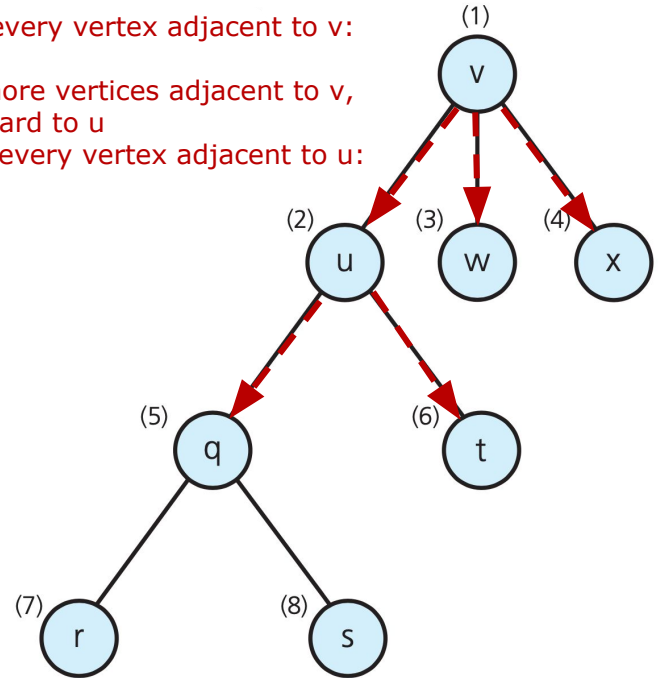
- 廣度優先
- BFS traversal visits all vertices adjacent to a vertex before going forward (deeper)

Breadth-First Search (BFS): example

1. start at vertex v
2. mark v (1 means the order of visit)



3. visit every vertex adjacent to v: u, w, x
4. no more vertices adjacent to v, go forward to u
5. visit every vertex adjacent to u: q, t



undirected + unweighted **graph**

Breadth-First Search (BFS)

- ❑ BFS strategy has a **recursion** version using a **queue**
- ❑ not as simple as the recursive version of DFS traversal

```
// Traverses a graph beginning at vertex v by using a
// breadth-first search: Recursive version.
bfs(q: Queue)
    if (q.empty()) return

    v = q.dequeue(the front node)

    for(each unvisited vertex u adjacent to v){ // do for every edge `v -> u
        Mark u as visited
        q.enqueue(u)
    }
    bfs(q)
```

Breadth-First Search (BFS)

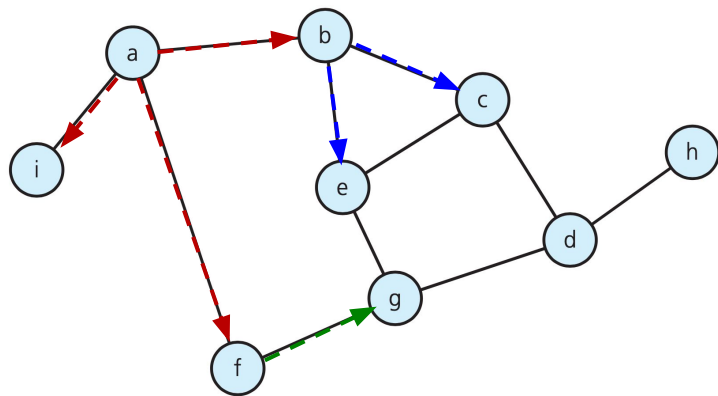
BFS strategy has a **iterative** version using a **queue**

```
// Traverses a graph beginning at vertex v by using a
// breadth-first search: Iterative version.
bfs(v: Vertex)
    q = a new empty queue
    q.enqueue(v) // Add v to queue and mark it
    Mark v as visited

    while (!q.isEmpty()){
        q.dequeue(the front node)

        for (each unvisited vertex u adjacent to v){ // do for every edge v → u
            Mark u as visited
            q.enqueue(u)
        }
    }
```

Breadth-First Search (BFS): example 2



order of visit: a, b, f, i, c, e, g, d, h

| <u>Node visited</u> | <u>Queue (front to back)</u> |
|---------------------|--|
| a | a |
| | (empty) |
| b | b |
| f | b f <i>put unvisited vertices adjacent to a in</i> |
| i | b f i <i>queue: b, f, i</i> |
| | f i <i>dequeue b</i> |
| c | f i c <i>put unvisited vertices adjacent to b in</i> |
| e | f i c e <i>queue: c, e</i> |
| | i c e <i>dequeue f</i> |
| g | i c e g <i>put unvisited vertices adjacent to f in</i> |
| | c e g <i>queue: g</i> |
| | e g |
| d | e g d |
| | g d |
| | d |
| | (empty) |
| h | h |
| | (empty) |

The results of a BFS traversal,
beginning at vertex a

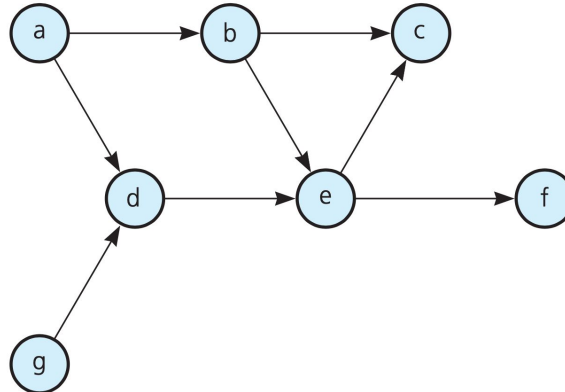
Application of Graphs

- Some common applications of graphs
 - Topological Sorting
 - Shortest Paths

Topological Sorting

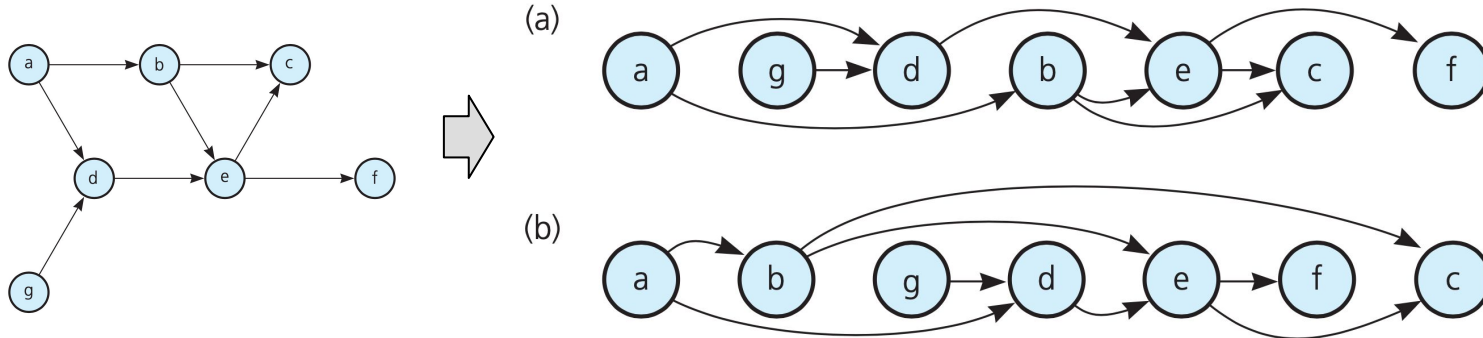
- A **directed graph without cycle** has a natural **order**
- If the vertices represent **courses**, the graph represent the **prerequisite structure** for the courses
- In what **order** should you **take all seven courses** so that you will satisfy all prerequisites? **Topological order**

*a is prerequisite of
b, b is prerequisite
of c and e*



Topological Sorting

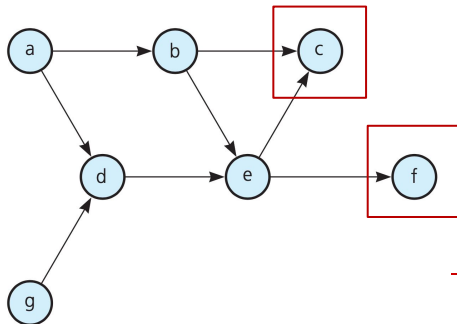
- The vertices in a given graph may have several topological orders
 - a, g, d, b, e, c, f
 - a, b, g, d, e, f, c



arrange the vertices of a directed graph in a topological order, edges will all point in one direction

Topological Sorting

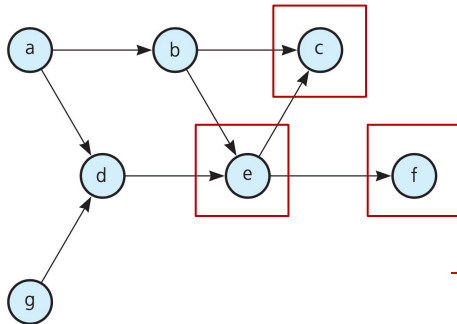
- Arranging the **vertices** into a **topological order** is called **topological sorting**
- Several ways to do topological sorting
 - find a **vertex that has no successor**
 - **remove** the vertex and all edges lead to it from graph
 - add the vertex **to the beginning of the list**
 - **repeatedly** the above steps **until the graph is empty**
 - the list of vertices is in topological order





list: 

Topological Sorting

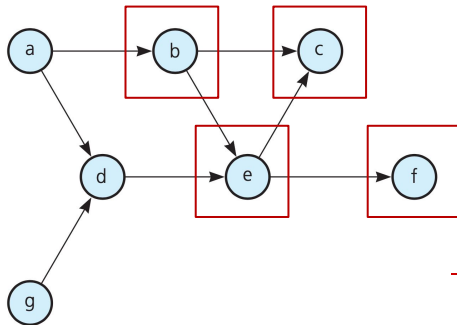
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list:  

Topological Sorting

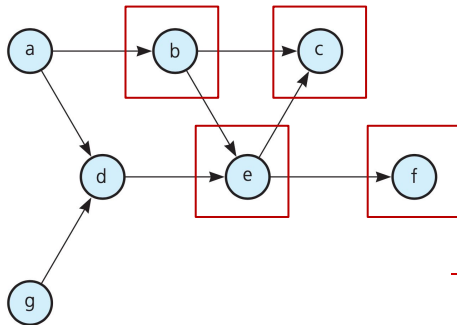
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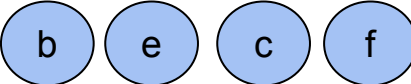


list: e c f

Topological Sorting

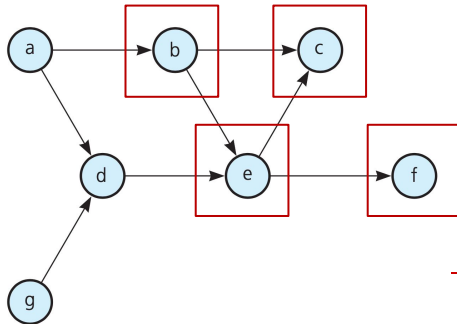
- Arranging the **vertices** into a topological order is called **topological sorting**
- Several ways to do topological sorting
 - find a **vertex that has no successor**
 - **remove** the vertex and all edges lead to it from graph
 - add the vertex **to the beginning of the list**
 - **repeatedly** the above steps **until the graph is empty**
 - the list of vertices is in topological order



list: 

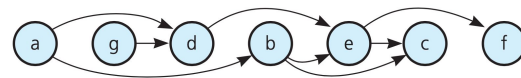
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 - **repeatedly** the above steps **until the graph is empty**
 - the list of vertices is in topological order



list: a g d b e c f

(a)



Topological Sorting: pseudocode

```
// Arranges the vertices in graph the Graph into a
// topological order and places them in list aList

topSort1(theGraph: Graph, aList: List)
    n = number of vertices in theGraph
    for (step = 1 through n){ //all vertices
        Select a vertex v that has no successors
        aList.insert(1, v) //insert at the beginning of list
        Remove from theGraph vertex v and its edges
    }
```

Topological Sorting: DFS topological sorting

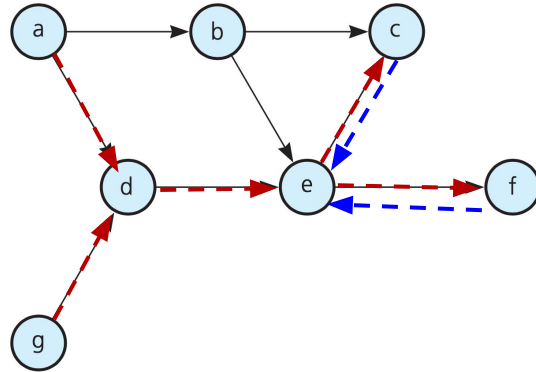
- depth-first search (DFS) topological sorting
 - push all vertices that have no predecessor onto a stack
 - Each time you pop a vertex from the stack
 - add it to the beginning of a list of vertices
- modified from DFS traversal

Topological Sorting: DFS pseudocode

```
topSort2(theGraph: Graph, aList: List)
    s = a new empty stack
    for (all vertices v in the graph){
        if (v has no predecessors){
            s.push(v)
            Mark v as visited
        }
    }

    while(!s.isEmpty()){
        if (all vertices adjacent to the vertex on the top of the stack have been visited)
        {
            s.pop(v) //backtrace
            aList.insert(1, v)
        }
        else{
            Select an unvisited vertex u adjacent to the vertex on the top of the stack
            s.push(u)
            Mark u as visited
        }
    }
}
```

Topological Sorting: DFS



Action

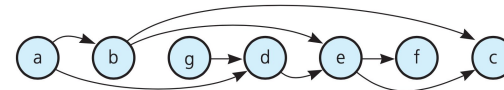
Stack s (bottom to top)

List aList (beginning to end)

Push a
 Push g
 Push d
 Push e
 Push c
 Pop c, add c to aList
 Push f
 Pop f, add f to aList
 Pop e, add e to aList
 Pop d, add d to aList
 Pop g, add g to aList
 Push b
 Pop b, add b to aList
 Pop a, add a to aList

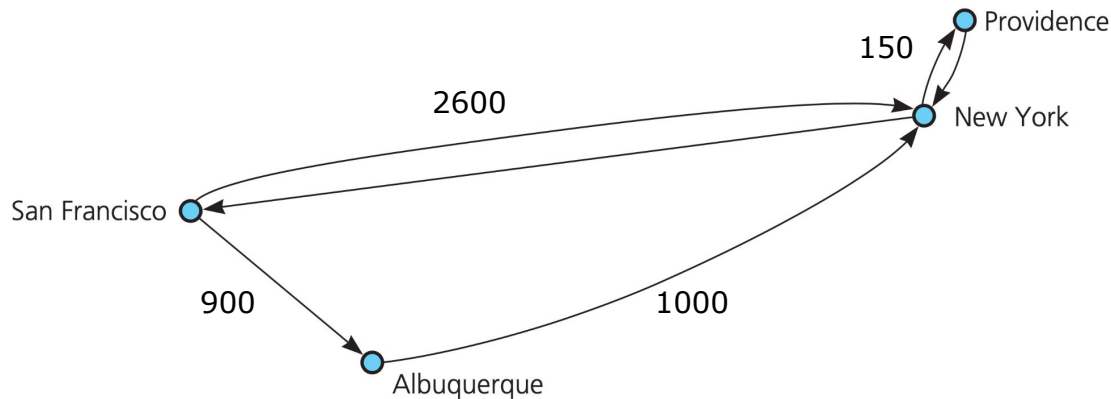
a
 a, g has no predecessors, g at the top of stack
 ag
 agd Select an unvisited vertex d adjacent to g
 agde Select an unvisited vertex e adjacent to d
 agdec Select an unvisited vertex c adjacent to e
 agde c c has no adjacent & unvisited vertex
 agdef c
 agde fc add f at the beginning of list
 agd efc
 ag defc
 a gdefc
 ab gdefc
 a bgdefc
 (empty) abgdefc

resulting topological order: ^(b)



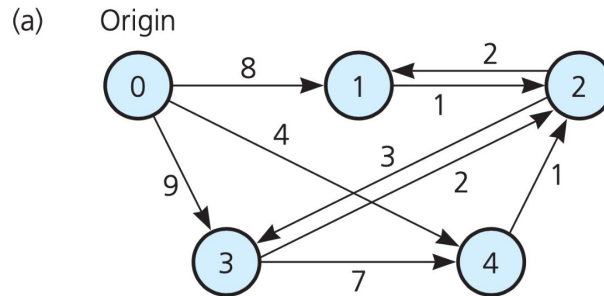
Shortest Path

- find the **shortest path** between 2 cities (vertices)
 - the path has the **smallest edge-weight sum**
- the weight can also be the dollars or duration (nonnegative)
- the sum of weights of edges of a path called **path's length/weight/cost**



Shortest Paths: example

- shortest path between **vertex 0 to 1**
 - **not edge** between 0 and 1 (**cost 8**)
 - path: 0 to 4 to 2 to 1 (**cost 7**)
- origin labeled 0, other vertices labeled from 1 to n-1



directed + weighted graph

(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

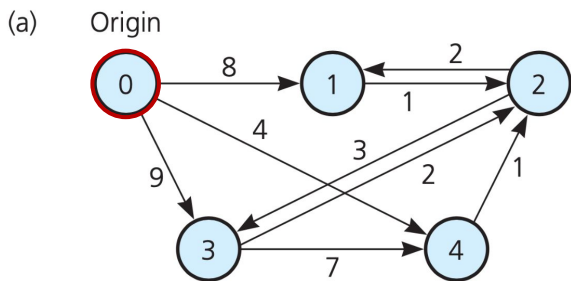
matrix of the left graph

Shortest Paths: Dijkstra's shortest-path algorithm

- Dijkstra's shortest-path algorithm
 - determines the **shortest paths** between a **given origin** and **all other** vertices
- `vertexSet`
 - a **set** of selected vertices
- `weight`
 - **array** where `weight[v]` is the weight of shortest path from vertex 0 (origin) to vertex `v` that **passes through vertices in** `vertexSet`

Shortest Paths: Dijkstra's shortest-path algorithm

- Initially,
 - vertexSet contains only vertex 0 (origin), and weight contains the weights of the single-edge paths from vertex 0 to all other vertices v
 - $weights[v] = matrix[0][v]$



(b)

initial weight is the first row of adjacency matrix

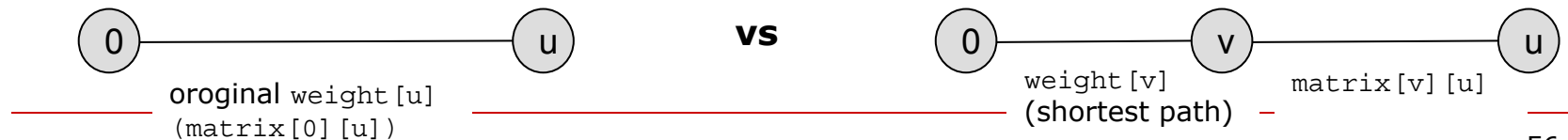
| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

Shortest Paths: Dijkstra's shortest-path algorithm

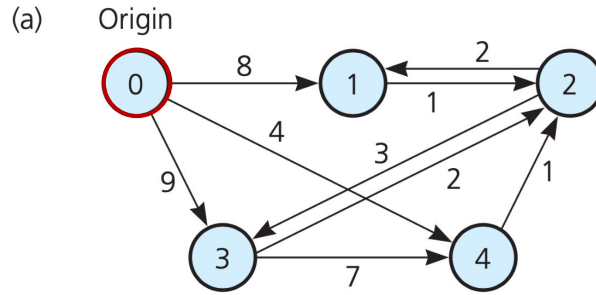
- After this initialization step
 - find a **vertex v**
 - not in `vertexSet`
 - the smallest value in `weight[v]`
 - You **add v** to `vertexSet`
- For all **unselected vertices u not in `vertexSet`**
 - check **value of `weight[u]`** (cost of path from 0 to u) to **ensure** that they are **minimums**
 - can the value of `weight[u]` be **reduced** by **passing newly selected vertex v** ?

Shortest Paths: Dijkstra's shortest-path algorithm

- can the value of $\text{weight}[u]$ be reduced by passing newly selected vertex v ?
- how to ensure value of $\text{weight}[u]$ is minimums?
 - break the path from 0 to u into two pieces and find their weights as follows
 - $\text{weight}[v]$ = weight of the shortest path from 0 to v
 - $\text{matrix}[v][u]$ = weight of the edge from v to u
- $\text{weight}[u] = \min\{\text{weight}[u], \text{weight}[v] + \text{matrix}[v][u]\}$



Shortest Paths: Dijkstra's shortest-path algorithm example



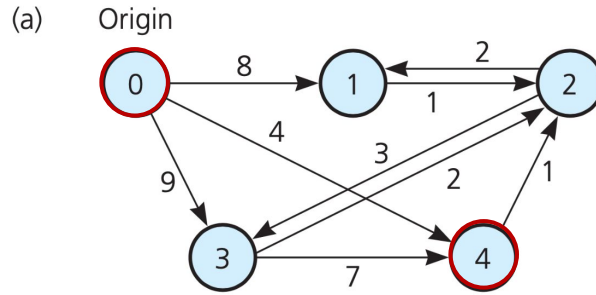
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | <u>v</u> | <u>vertexSet</u> | <u>weight</u> | | | | |
|------|----------|------------------|---------------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |

initially, vertexSet contains 0,
weight = adjacency matrix[0][]

Shortest Paths: Dijkstra's shortest-path algorithm example



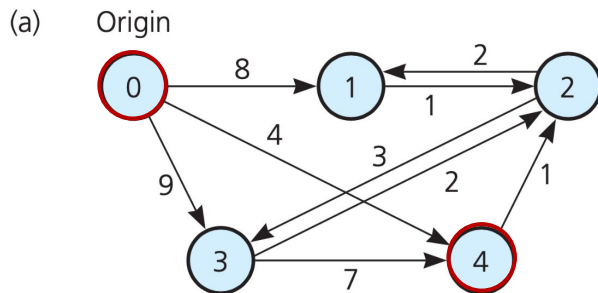
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| | | weight | | | | | |
|------|---|-----------|-----|-----|----------|-----|-----|
| Step | v | vertexSet | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | | | | | |

weight[4] is the smallest value in weight and vertex 4 didn't in vertexSet. So, $v = 4$ & add 4 to vertexSet

Shortest Paths: Dijkstra's shortest-path algorithm example



(b)

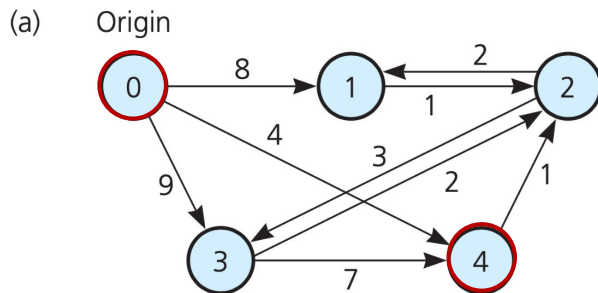
| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | <u>v</u> | <u>vertexSet</u> | weight | | | | |
|------|----------|------------------|--------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | | | | 4 |

For vertices not in `vertexSet`
(for $u = 1, 2, 3$), check if
`weight[u]` (path from 0 to u)
is minimums

$\text{weight}[u] = \min\{\text{weight}[u],$
 $\text{weight}[v] + \text{matrix}[v][u]\}$

Shortest Paths: Dijkstra's shortest-path algorithm example



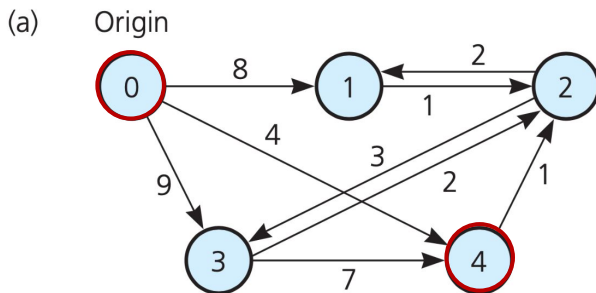
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|---|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | <u>v</u> | <u>vertexSet</u> | <u>weight</u> | | | | |
|------|----------|------------------|---------------|-----|-----|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | | | 4 |

For vertex 1 ($u=1$),
 $\text{weight}[1] = 8$
 $\text{weight}[4] + \text{matrix}[4][1] = 4 + \infty$

Shortest Paths: Dijkstra's shortest-path algorithm example



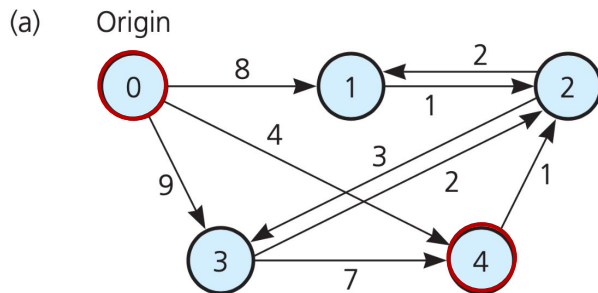
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | <u>v</u> | <u>vertexSet</u> | <u>weight</u> | | | | |
|------|----------|------------------|---------------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | | 4 |

For vertex 2 ($u=2$),
 $\text{weight}[2] = \infty$
 $\text{weight}[4] + \text{matrix}[4][2] = 4 + 1$

Shortest Paths: Dijkstra's shortest-path algorithm example



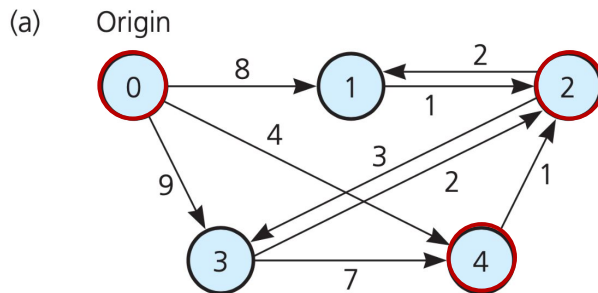
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|---|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | v | vertexSet | weight | | | | |
|------|---|-----------|--------|-----|-----|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |

For vertex 3 ($u=3$),
 $\text{weight}[3] = 9$
 $\text{weight}[4] + \text{matrix}[4][3] = 4 + \infty$

Shortest Paths: Dijkstra's shortest-path algorithm example



(b)

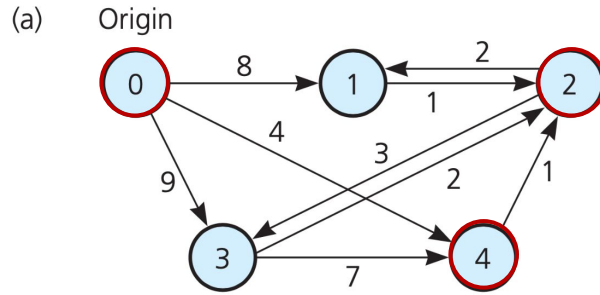
| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | <u>v</u> | <u>vertexSet</u> | <u>weight</u> | | | | |
|------|----------|------------------|---------------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | | 5 | | 4 |

weight[2] is the smallest value in weight and vertex 2 didn't in vertexSet.

So, $v = 2$ & add 2 to vertexSet

Shortest Paths: Dijkstra's shortest-path algorithm example



(b)

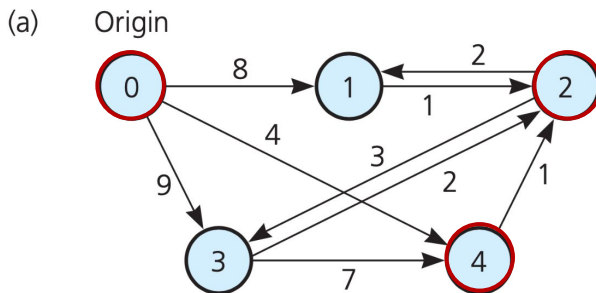
| | 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|---|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | v | vertexSet | weight | | | | |
|------|---|-----------|--------|-----|-----|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | | 5 | | 4 |

For vertices not in vertexSet (for $u = 1, 3$), check if $\text{weight}[u]$ (path from 0 to u) is minimums

$\text{weight}[u] = \min\{\text{weight}[u], \text{weight}[v] + \text{matrix}[v][u]\}$

Shortest Paths: Dijkstra's shortest-path algorithm example



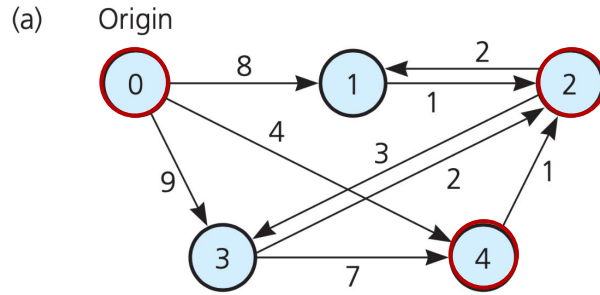
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | v | vertexSet | weight | | | | |
|------|---|-----------|--------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | 7 | 5 | | 4 |

For vertex 1 ($u=1$),
 $\text{weight}[1] = 8$
 $\text{weight}[2] + \text{matrix}[2][1] = 5 + 2$

Shortest Paths: Dijkstra's shortest-path algorithm example



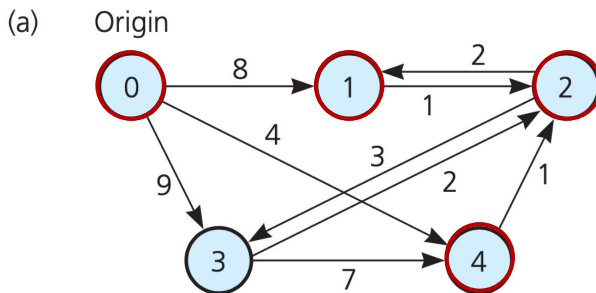
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | v | vertexSet | weight | | | | |
|------|---|-----------|--------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | 7 | 5 | 8 | 4 |

For vertex 3 ($u=3$),
 $\text{weight}[3] = 9$
 $\text{weight}[2] + \text{matrix}[2][3] = 5 + 3$

Shortest Paths: Dijkstra's shortest-path algorithm example



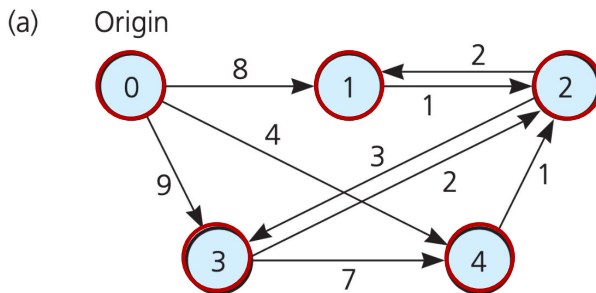
(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|---|
| 0 | 0 | 8 | ∞ | 9 | 4 |
| 1 | ∞ | 0 | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | 0 | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | 0 | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | 0 |

| Step | v | vertexSet | weight | | | | |
|------|---|------------|--------|-----|-----|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | 7 | 5 | 8 | 4 |
| 4 | 1 | 0, 4, 2, 1 | 0 | 7 | 5 | 8 | 4 |

weight[1] is the smallest value in weight and vertex 1 didn't in vertexSet.
So, v = 1 & add 1 to vertexSet

Shortest Paths: Dijkstra's shortest-path algorithm example



(b)

| | 0 | 1 | 2 | 3 | 4 |
|---|----------|----------|----------|----------|----------|
| 0 | ∞ | 8 | ∞ | 9 | 4 |
| 1 | ∞ | ∞ | 1 | ∞ | ∞ |
| 2 | ∞ | 2 | ∞ | 3 | ∞ |
| 3 | ∞ | ∞ | 2 | ∞ | 7 |
| 4 | ∞ | ∞ | 1 | ∞ | ∞ |

| Step | v | vertexSet | weight | | | | |
|------|---|---------------|--------|-----|----------|-----|-----|
| | | | [0] | [1] | [2] | [3] | [4] |
| 1 | – | 0 | 0 | 8 | ∞ | 9 | 4 |
| 2 | 4 | 0, 4 | 0 | 8 | 5 | 9 | 4 |
| 3 | 2 | 0, 4, 2 | 0 | 7 | 5 | 8 | 4 |
| 4 | 1 | 0, 4, 2, 1 | 0 | 7 | 5 | 8 | 4 |
| 5 | 3 | 0, 4, 2, 1, 3 | 0 | 7 | 5 | 8 | 4 |

repeat the same process, update the weight[u] ($u = 3$)

the final values in weight are the weight of shortest path from vertex 0 to other vertices

Shortest Paths: Dijkstra's shortest-path algorithm pseudocode

```
// Finds the minimum-cost paths between an origin vertex
// (vertex 0) and all other vertices in a weighted directed
shortestPath(theGraph: Graph, weight: WeightArray)

    // Step 1: initialization
    Create a set vertexSet that contains only vertex 0
    n = number of vertices in theGraph
    for (v = 0 through n - 1)
        weight[v] = matrix[0][v]

    // Steps 2:
    for (step = 2 through n){ //for the rest vertex 1 to n-1
        Find the smallest weight[v] such that v is not in vertexSet
        Add v to vertexSet

        for (all vertices u not in vertexSet){ // update weight[u] for all u not in vertexSet
            if (weight[u] > weight[v] + matrix[v][u])
                weight[u] = weight[v] + matrix[v][u]
        }
    }
```

Summary of Graph

- Graph implementation
 - adjacency matrix and the adjacency list
 - advantages and disadvantages
- Graph Traversal
 - Depth-first search (DFS)
 - Breadth-first search (BFS)
 - the order of visit
- Applications
 - Topological sorting
 - Dijkstra's shortest-path algorithm