

# ENG 346

# Data Structures and Algorithms for Artificial Intelligence

## Graphs

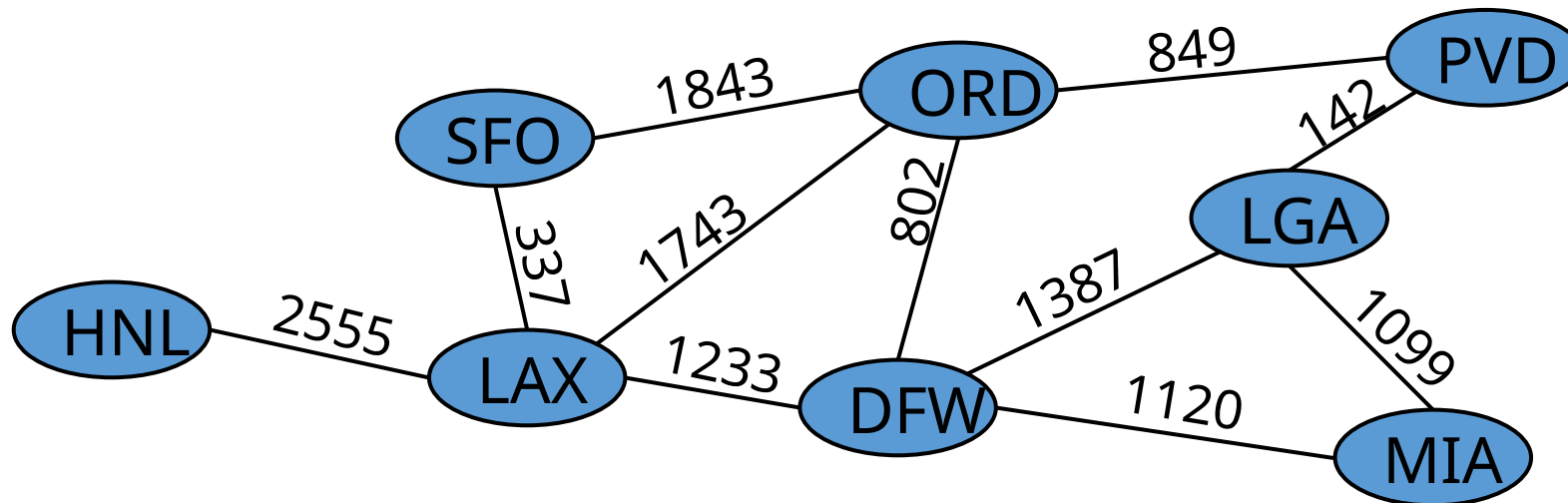
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<https://github.com/mehmetpekmezci/GTU-ENG-346>

# Graphs

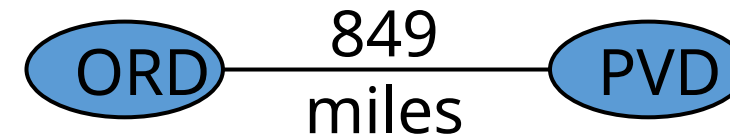
- A graph is a pair  $(V, E)$ , where
  - $V$  is a set of nodes, called **vertices**
  - $E$  is a collection of pairs of vertices, called **edges**
  - Vertices and edges are positions and store elements
- Example:
  - A vertex represents an airport and stores the three-letter airport code
  - An edge represents a flight route between two airports and stores the mileage of the route





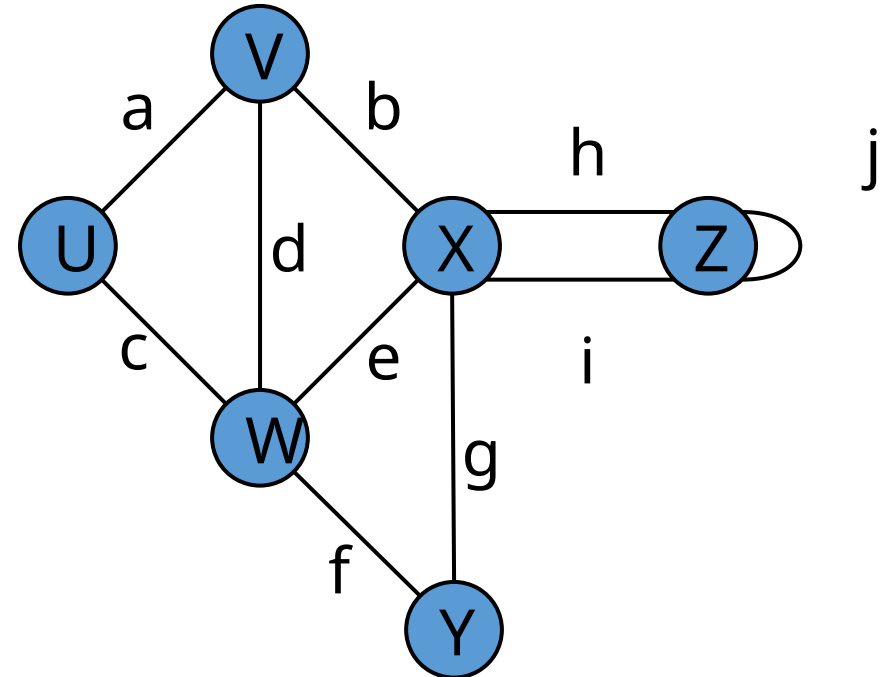
# Edge Types

- Directed edge
  - ordered pair of vertices  $(u, v)$
  - first vertex  $u$  is the origin
  - second vertex  $v$  is the destination
  - e.g., a flight
- Undirected edge
  - unordered pair of vertices  $(u, v)$
  - e.g., a flight route
- Directed graph
  - all the edges are directed
  - e.g., route network
- Undirected graph
  - all the edges are undirected
  - e.g., flight network



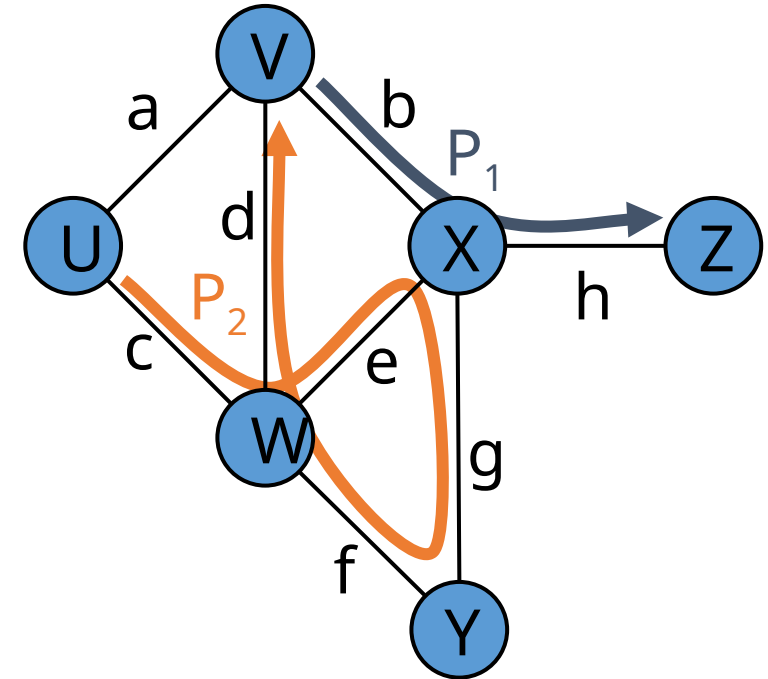
# Terminology

- End vertices (or endpoints) of an edge
  - U and V are the endpoints of a
- Edges incident on a vertex
  - a, d, and b are incident on V
- Adjacent/Neighbor vertices
  - U and V are adjacent
- Degree of a vertex
  - X has degree 5
- Parallel edges
  - h and i are parallel edges
- Self-loop
  - j is a self-loop



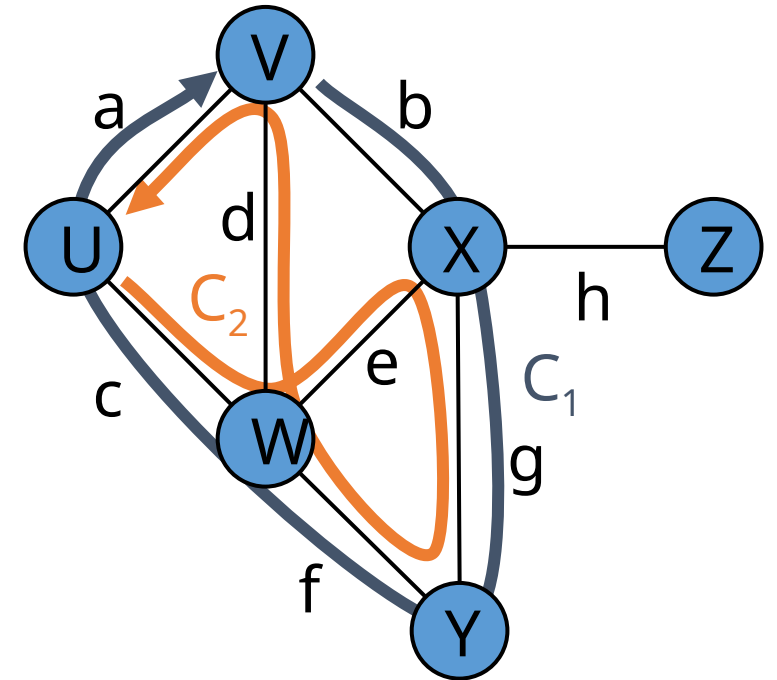
# Terminology – continued

- Path
  - sequence of alternating vertices and edges
  - begins with a vertex
  - ends with a vertex
  - each edge is preceded and followed by its endpoints
- Simple path
  - path such that all its vertices and edges are distinct
- Examples
  - $P_1 = (V, b, X, h, Z)$  is a simple path
  - $P_2 = (U, c, W, e, X, g, Y, f, W, d, V)$  is a path that is not simple



# Terminology – continued

- Cycle
  - circular sequence of alternating vertices and edges
  - each edge is preceded and followed by its endpoints
- Simple cycle
  - cycle such that all its vertices and edges are distinct
- Examples
  - $C_1 = (V, b, X, g, Y, f, W, c, U, a, \underline{U, V})$  is a simple cycle
  - $C_2 = (U, c, W, e, X, g, Y, f, W, d, V, a, \underline{U, U})$  is a cycle that is not simple



# Properties

## Property 1

$$\sum_v \deg(v) = 2m$$

Proof: each edge is counted twice

## Property 2

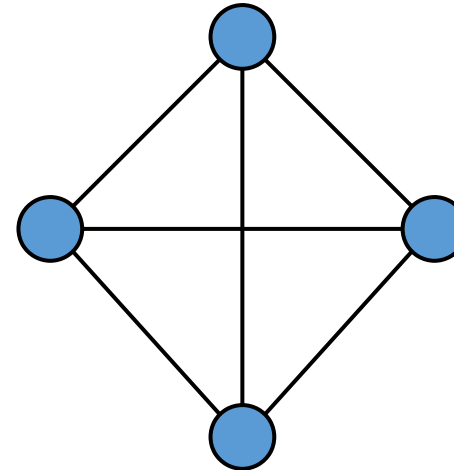
In an undirected graph with no self-loops and no multiple edges

$$m \leq n(n-1)/2$$

Proof: each vertex has degree at most  $(n-1)$

## Notation

$n$	number of vertices
$m$	number of edges
$\deg(v)$	degree of vertex $v$



## Example

- $n = 4$
- $m = 6$
- $\deg(v) = 3$



# Vertices and Edges

- A **graph** is a collection of **vertices** and **edges**.
- We model the abstraction as a combination of three data types: Vertex, Edge, and Graph.
- A **Vertex** is a lightweight object that stores an arbitrary element provided by the user (e.g., an airport code)
  - We assume it supports a method, `element()`, to retrieve the stored element.
- An **Edge** stores an associated object (e.g., a flight number, travel distance, cost), retrieved with the `element()` method.
- In addition, we assume that an Edge supports the following methods:

`endpoints()`: Return a tuple  $(u, v)$  such that vertex  $u$  is the origin of the edge and vertex  $v$  is the destination; for an undirected graph, the orientation is arbitrary.

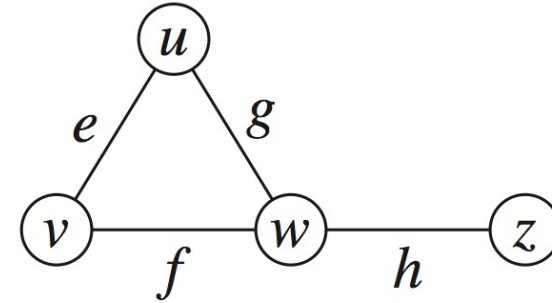
`opposite(v)`: Assuming vertex  $v$  is one endpoint of the edge (either origin or destination), return the other endpoint.

# Graph ADT

- `vertex_count()`: Return the number of vertices of the graph.
- `vertices()`: Return an iteration of all the vertices of the graph.
- `edge_count()`: Return the number of edges of the graph.
- `edges()`: Return an iteration of all the edges of the graph.
- `get_edge(u,v)`: Return the edge from vertex  $u$  to vertex  $v$ , if one exists; otherwise return None. For an undirected graph, there is no difference between `get_edge(u,v)` and `get_edge(v,u)`.
- `degree(v, out=True)`: For an undirected graph, return the number of edges incident to vertex  $v$ . For a directed graph, return the number of outgoing (resp. incoming) edges incident to vertex  $v$ , as designated by the optional parameter.
- `incident_edges(v, out=True)`: Return an iteration of all edges incident to vertex  $v$ . In the case of a directed graph, report outgoing edges by default; report incoming edges if the optional parameter is set to False.
- `insert_vertex(x=None)`: Create and return a new Vertex storing element  $x$ .
- `insert_edge(u, v, x=None)`: Create and return a new Edge from vertex  $u$  to vertex  $v$ , storing element  $x$  (None by default).
- `remove_vertex(v)`: Remove vertex  $v$  and all its incident edges from the graph.
- `remove_edge(e)`: Remove edge  $e$  from the graph.

# Edge List Structure

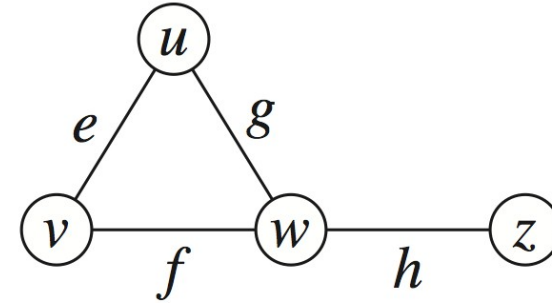
- Vertex object
  - element
  - reference to position in vertex sequence
- Edge object
  - element
  - origin vertex object
  - destination vertex object
  - reference to position in edge sequence
- Vertex sequence
  - sequence of vertex objects
- Edge sequence
  - sequence of edge objects



edges = [(u,v), (u,w), (v,w), (w,z)]

# Adjacency List Structure

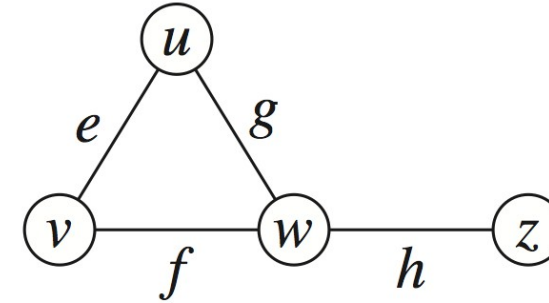
- Lists neighbors for each vertex



```
{  
  u: [v, w],  
  v: [u, w],  
  w: [u, v, z],  
  z: [w]  
}
```

# Adjacency Matrix Structure

- Edge list structure
- Augmented vertex objects
  - Integer key (index) associated with vertex
- 2D-array adjacency array
  - Reference to edge object for adjacent vertices
  - Null for non nonadjacent vertices
- The “old fashioned” version just has 0 for no edge and 1 for edge



		0	1	2	3
$u$	→ 0		$e$	$g$	
$v$	→ 1	$e$		$f$	
$w$	→ 2	$g$	$f$		$h$
$z$	→ 3			$h$	

# Performance

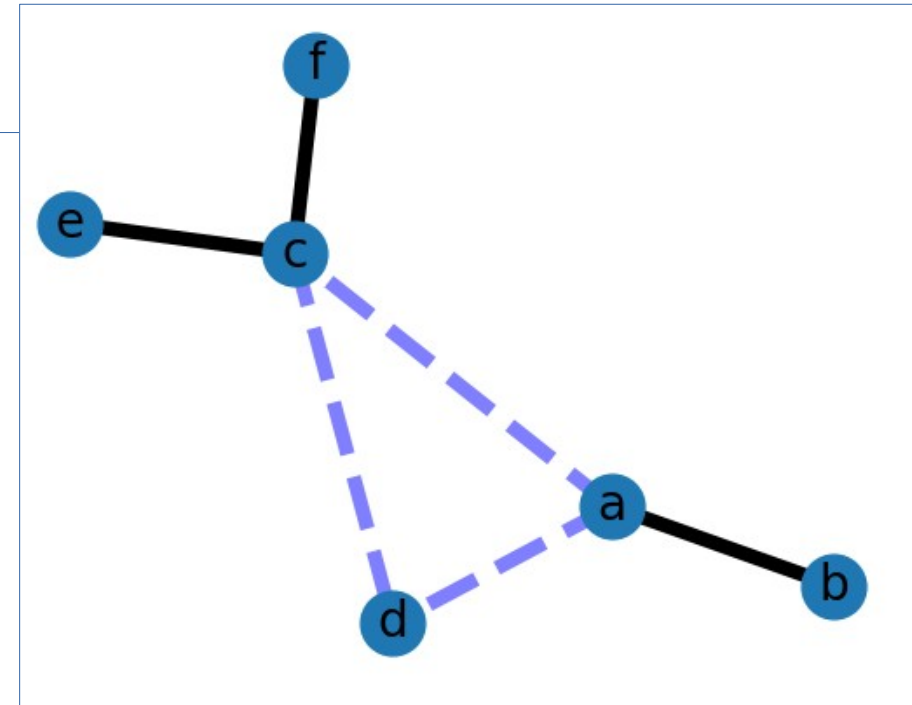
<ul style="list-style-type: none"> <li>▪ <math>n</math> vertices, <math>m</math> edges</li> <li>▪ no parallel edges</li> <li>▪ no self-loops</li> </ul>	Edge List	Adjacency List	Adjacency Matrix
Space	$n + m$	$n + m$	$n^2$
incidentEdges( $v$ )	$m$	deg( $v$ )	$n$
areAdjacent ( $v$ , $w$ )	$m$	min(deg( $v$ ), deg( $w$ ))	1
insertVertex( $o$ )	1	1	$n^2$
insertEdge( $v$ , $w$ , $o$ )	1	1	1
removeVertex( $v$ )	$m$	deg( $v$ )	$n^2$
removeEdge( $e$ )	1	1	1

# Python Graph Libs

- NetworkX : General graph implementation.
- Pytorch Geometric : Generally used in Graph Neural Network implementations.

# NetworkX Example

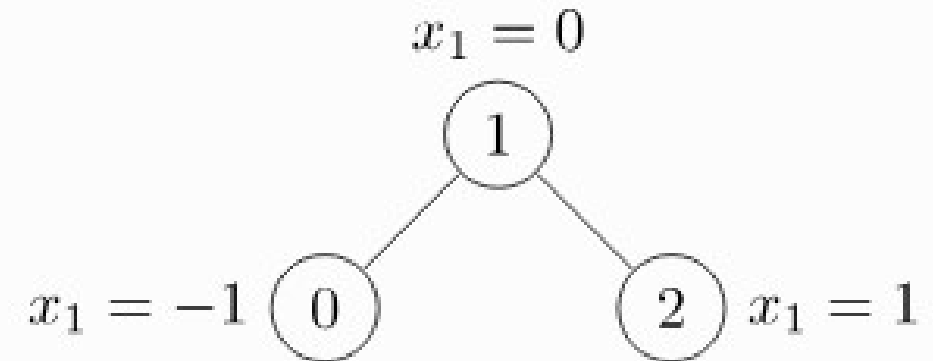
```
import matplotlib.pyplot as plt
import networkx as nx
G = nx.Graph()
G.add_edge('a', 'b', weight=0.6)
G.add_edge('a', 'c', weight=0.2)
G.add_edge('c', 'd', weight=0.1)
G.add_edge('c', 'e', weight=0.7)
G.add_edge('c', 'f', weight=0.9)
G.add_edge('a', 'd', weight=0.3)
elarge = [(u, v) for (u, v, d) in G.edges(data=True) if d['weight'] > 0.5]
esmall = [(u, v) for (u, v, d) in G.edges(data=True) if d['weight'] <= 0.5]
pos = nx.spring_layout(G) # positions for all nodes
nx.draw_networkx_nodes(G, pos, node_size=700)
nx.draw_networkx_edges(G, pos, edgelist=elarge, width=6)
nx.draw_networkx_edges(G, pos, edgelist=esmall, width=6, alpha=0.5, edge_color='b', style='dashed')
nx.draw_networkx_labels(G, pos, font_size=20, font_family='sans-serif')
plt.axis('off')
plt.show()
```





# Pytorch Geometric Example

```
import torch
from torch_geometric.data import Data
edge_index = torch.tensor([[0, 1, 1, 2],
                           [1, 0, 2, 1]], dtype=torch.long)
x = torch.tensor([[ -1], [ 0], [ 1]], dtype=torch.float)
data = Data(x=x, edge_index=edge_index)
>>> Data(edge_index=[2, 4], x=[3, 1])
```



# Python Graph Implementation

- We use a variant of the *adjacency map* representation.
- For each vertex  $v$ , we use a Python dictionary to represent the secondary incidence map  $I(v)$ .
- The list  $V$  is replaced by a top-level dictionary  $D$  that maps each vertex  $v$  to its incidence map  $I(v)$ .
  - Note that we can iterate through all vertices by generating the set of keys for dictionary  $D$ .
- A vertex does not need to explicitly maintain a reference to its position in  $D$ , because it can be determined in  $O(1)$  expected time.
- Running time bounds for the adjacency-list graph ADT operations, given above, become *expected* bounds.

# Vertex Class

```

1  #----- nested Vertex class -----
2  class Vertex:
3      """Lightweight vertex structure for a graph."""
4      __slots__ = '_element'
5
6      def __init__(self, x):
7          """Do not call constructor directly. Use Graph's insert_vertex(x)."""
8          self._element = x
9
10     def element(self):
11         """Return element associated with this vertex."""
12         return self._element
13
14     def __hash__(self):                # will allow vertex to be a map/set key
15         return hash(id(self))

```

# Edge Class

```
17 #----- nested Edge class -----
18 class Edge:
19     """ Lightweight edge structure for a graph. """
20     __slots__ = '_origin', '_destination', '_element'
21
22     def __init__(self, u, v, x):
23         """ Do not call constructor directly. Use Graph's insert_edge(u,v,x). """
24         self._origin = u
25         self._destination = v
26         self._element = x
27
28     def endpoints(self):
29         """ Return (u,v) tuple for vertices u and v. """
30         return (self._origin, self._destination)
31
32     def opposite(self, v):
33         """ Return the vertex that is opposite v on this edge. """
34         return self._destination if v is self._origin else self._origin
35
36     def element(self):
37         """ Return element associated with this edge. """
38         return self._element
39
40     def __hash__(self): # will allow edge to be a map/set key
41         return hash( (self._origin, self._destination) )
```

# Graph, Part 1

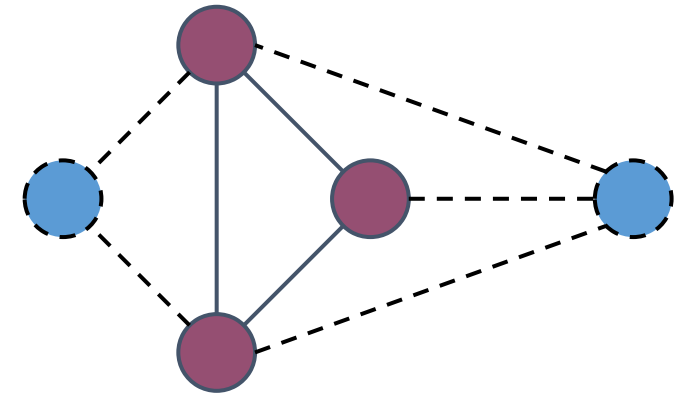
```
1 class Graph:
2     """Representation of a simple graph using an adjacency map."""
3
4     def __init__(self, directed=False):
5         """Create an empty graph (undirected, by default).
6
7         Graph is directed if optional paramter is set to True.
8         """
9         self._outgoing = { }
10        # only create second map for directed graph; use alias for undirected
11        self._incoming = { } if directed else self._outgoing
12
13    def is_directed(self):
14        """Return True if this is a directed graph; False if undirected.
15
16        Property is based on the original declaration of the graph, not its contents.
17        """
18        return self._incoming is not self._outgoing # directed if maps are distinct
19
20    def vertex_count(self):
21        """Return the number of vertices in the graph."""
22        return len(self._outgoing)
23
24    def vertices(self):
25        """Return an iteration of all vertices of the graph."""
26        return self._outgoing.keys()
27
28    def edge_count(self):
29        """Return the number of edges in the graph."""
30        total = sum(len(self._outgoing[v]) for v in self._outgoing)
31        # for undirected graphs, make sure not to double-count edges
32        return total if self.is_directed( ) else total // 2
33
34    def edges(self):
35        """Return a set of all edges of the graph."""
36        result = set( ) # avoid double-reporting edges of undirected graph
37        for secondary_map in self._outgoing.values():
38            result.update(secondary_map.values()) # add edges to resulting set
39        return result
```

# Graph, Part

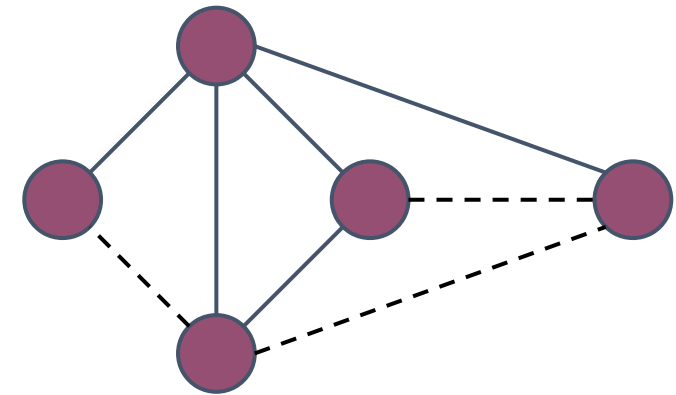
```
40 def get_edge(self, u, v):
41     """Return the edge from u to v, or None if not adjacent."""
42     return self._outgoing[u].get(v)          # returns None if v not adjacent
43
44 def degree(self, v, outgoing=True):
45     """Return number of (outgoing) edges incident to vertex v in the graph.
46
47     If graph is directed, optional parameter used to count incoming edges.
48     """
49     adj = self._outgoing if outgoing else self._incoming
50     return len(adj[v])
51
52 def incident_edges(self, v, outgoing=True):
53     """Return all (outgoing) edges incident to vertex v in the graph.
54
55     If graph is directed, optional parameter used to request incoming edges.
56     """
57     adj = self._outgoing if outgoing else self._incoming
58     for edge in adj[v].values():
59         yield edge
60
61 def insert_vertex(self, x=None):
62     """Insert and return a new Vertex with element x."""
63     v = self.Vertex(x)
64     self._outgoing[v] = { }
65     if self.is_directed():
66         self._incoming[v] = { }          # need distinct map for incoming edges
67     return v
68
69 def insert_edge(self, u, v, x=None):
70     """Insert and return a new Edge from u to v with auxiliary element x."""
71     e = self.Edge(u, v, x)
72     self._outgoing[u][v] = e
73     self._incoming[v][u] = e
```

# Subgraphs

- A subgraph  $S$  of a graph  $G$  is a graph such that
  - The vertices of  $S$  are a subset of the vertices of  $G$
  - The edges of  $S$  are a subset of the edges of  $G$
- A spanning subgraph of  $G$  is a subgraph that contains all the vertices of  $G$



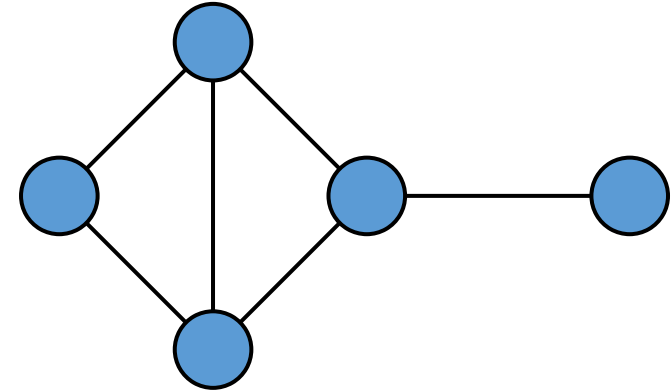
Subgraph



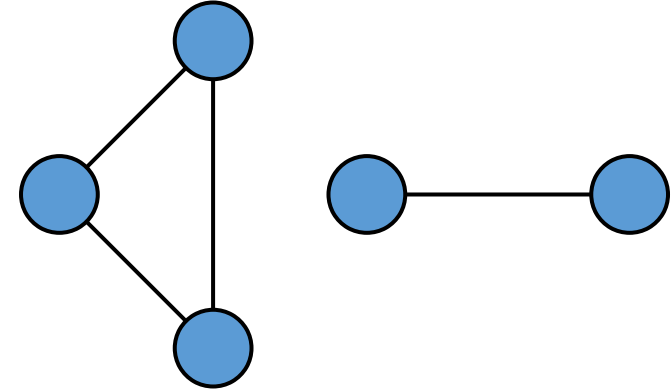
Spanning subgraph

# Connectivity

- A graph is connected if there is a path between every pair of vertices
- A connected component of a graph  $G$  is a maximal connected subgraph of  $G$



Connected graph



Non connected graph with two connected components

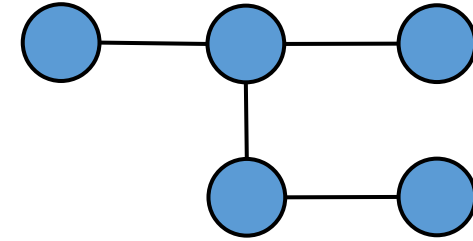


# Trees and Forests

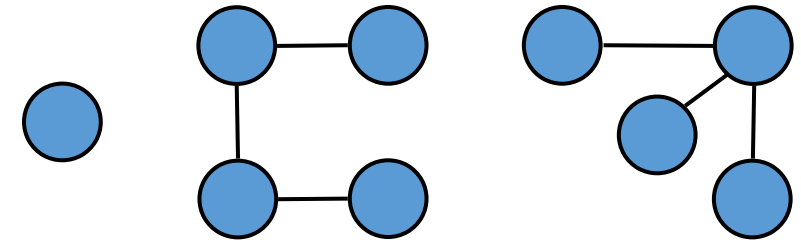
- A (free) tree is an undirected graph  $T$  such that
  - $T$  is connected
  - $T$  has no cycles

This definition of tree is different from the one of a rooted tree

- A forest is an undirected graph without cycles
- The connected components of a forest are trees



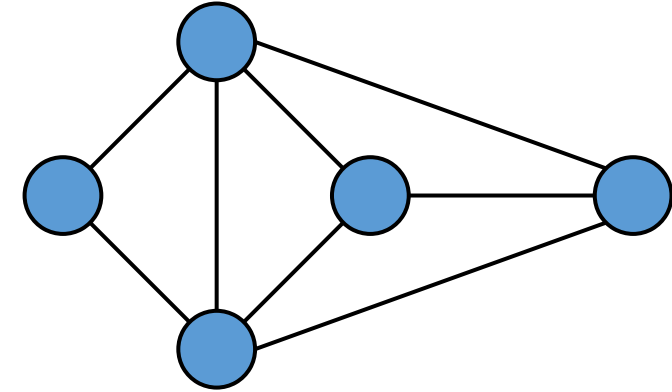
Tree



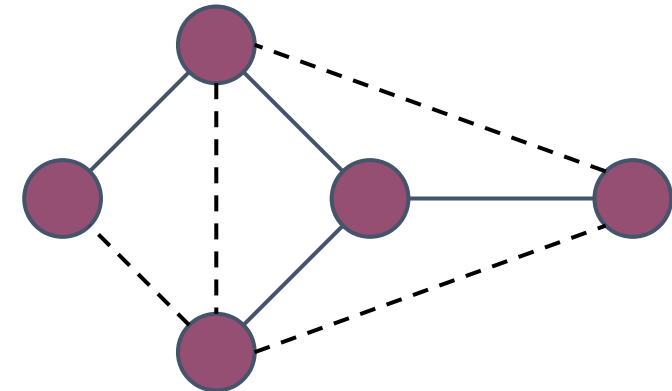
Forest

# Spanning Trees and Forests

- A spanning tree of a connected graph is a spanning subgraph that is a tree
- A spanning tree is not unique unless the graph is a tree
- Spanning trees have applications to the design of communication networks
- A spanning forest of a graph is a spanning subgraph that is a forest



Graph



Spanning tree

# Depth-First Search

- Depth-first search (DFS) is a general technique for traversing a graph
- A DFS traversal of a graph  $G$ 
  - Visits all the vertices and edges of  $G$
  - Determines whether  $G$  is connected
  - Computes the connected components of  $G$
  - Computes a spanning forest of  $G$
- DFS on a graph with  $n$  vertices and  $m$  edges takes  $O(n + m)$  time
- DFS can be further extended to solve other graph problems
  - Find and report a path between two given vertices
  - Find a cycle in the graph
- Depth-first search is to graphs what Euler tour is to binary trees

# DFS Algorithm

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

## Algorithm *DFS(G)*

**Input** graph *G*

**Output** labeling of the edges of *G*  
as discovery edges and  
back edges

```
for all  $u \in G.vertices()$ 
     $setLabel(u, UNEXPLORED)$ 
for all  $e \in G.edges()$ 
     $setLabel(e, UNEXPLORED)$ 
for all  $v \in G.vertices()$ 
    if  $getLabel(v) = UNEXPLORED$ 
         $DFS(G, v)$ 
```

## Algorithm *DFS(G, v)*

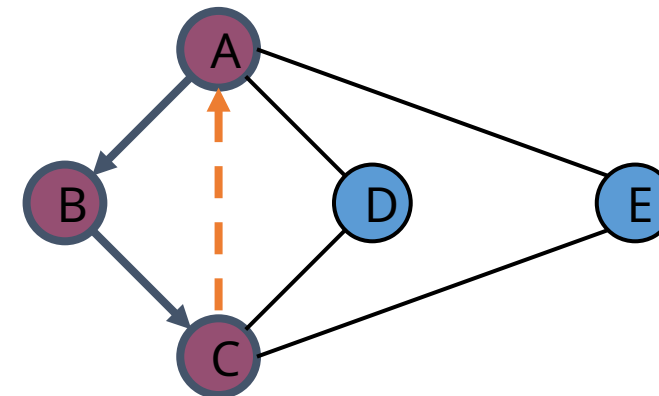
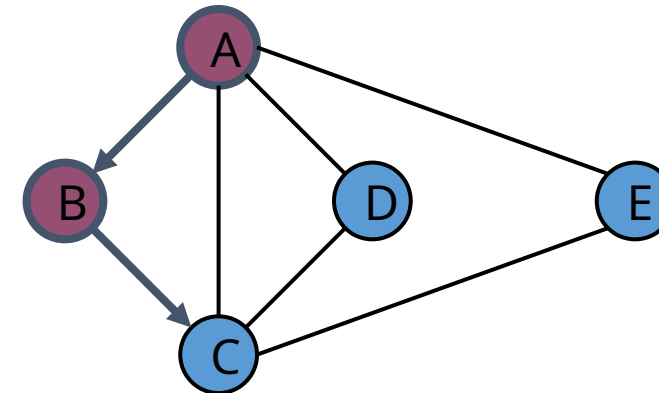
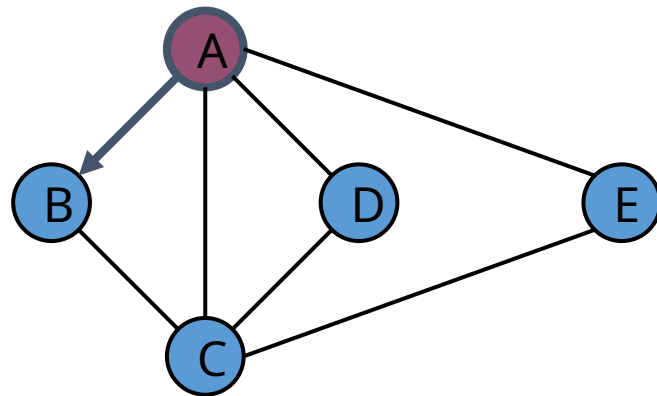
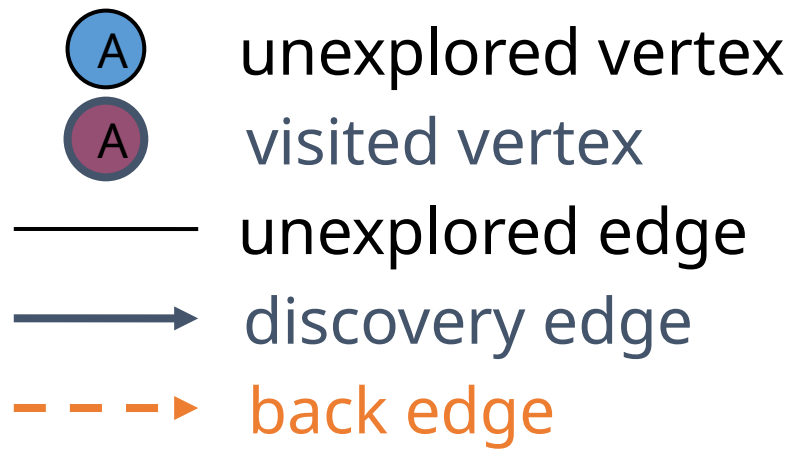
**Input** graph *G* and a start vertex *v* of *G*  
**Output** labeling of the edges of *G*  
in the connected component of *v*  
as discovery edges and back edges

```
 $setLabel(v, VISITED)$ 
for all  $e \in G.incidentEdges(v)$ 
    if  $getLabel(e) = UNEXPLORED$ 
         $w \leftarrow opposite(v, e)$ 
        if  $getLabel(w) = UNEXPLORED$ 
             $setLabel(e, DISCOVERY)$ 
             $DFS(G, w)$ 
        else
             $setLabel(e, BACK)$ 
```

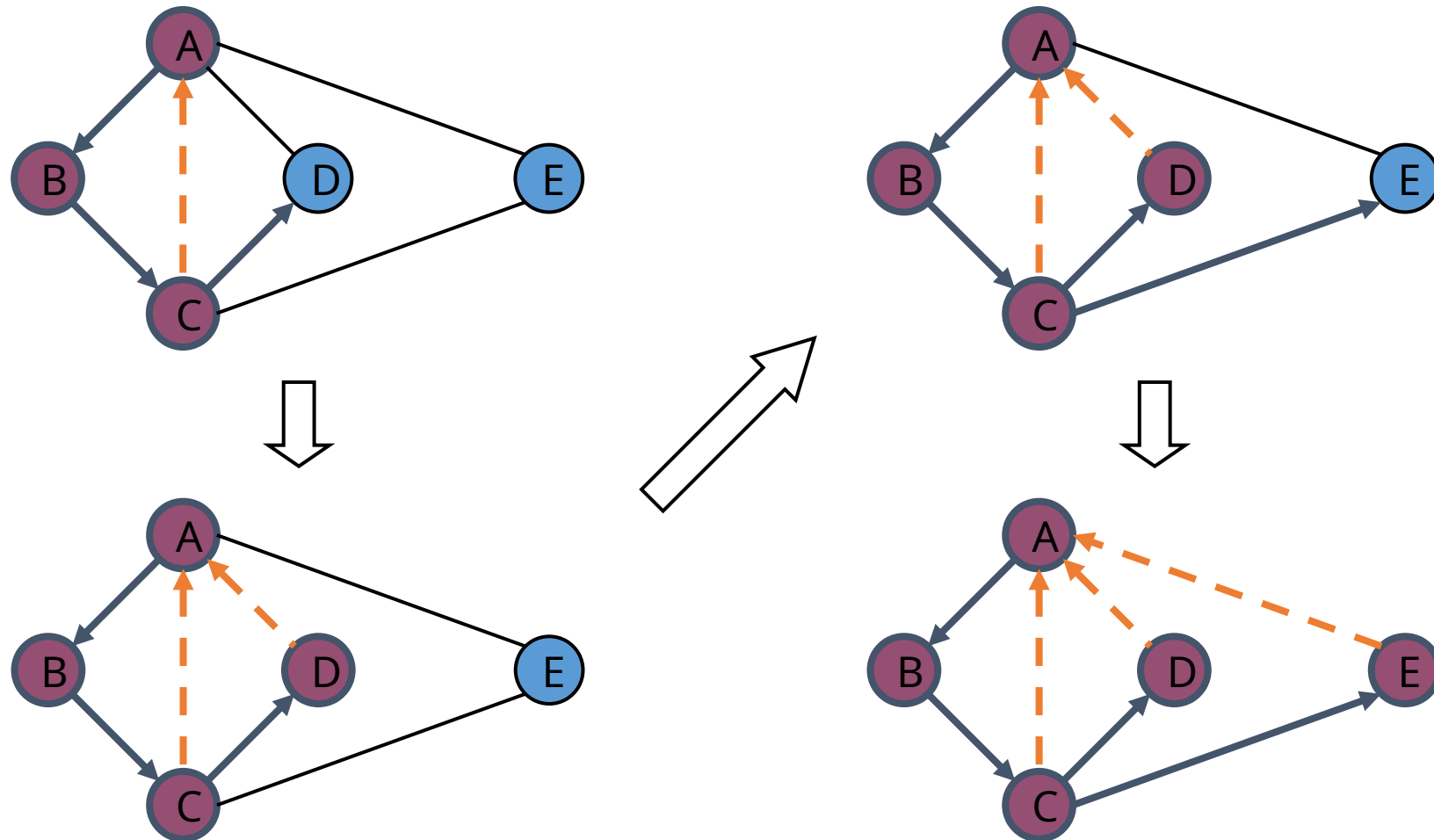
# Python Implementation

```
1 def DFS(g, u, discovered):
2     """Perform DFS of the undiscovered portion of Graph g starting at Vertex u.
3
4     discovered is a dictionary mapping each vertex to the edge that was used to
5     discover it during the DFS. (u should be "discovered" prior to the call.)
6     Newly discovered vertices will be added to the dictionary as a result.
7     """
8     for e in g.incident_edges(u):           # for every outgoing edge from u
9         v = e.opposite(u)
10        if v not in discovered:             # v is an unvisited vertex
11            discovered[v] = e               # e is the tree edge that discovered v
12            DFS(g, v, discovered)           # recursively explore from v
```

# Example

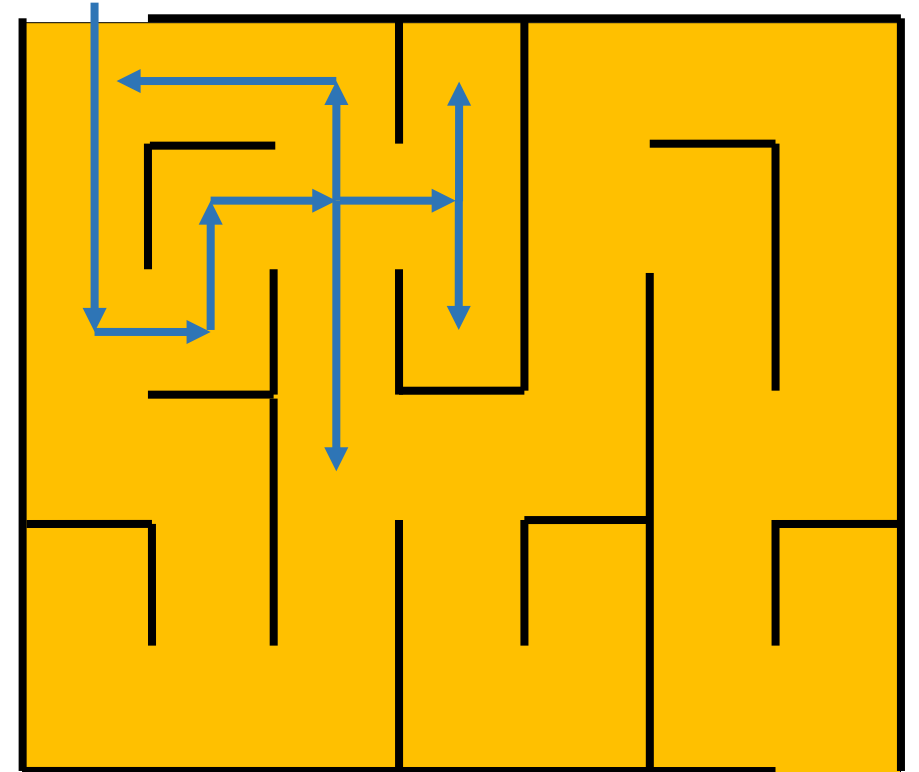


# Example (cont.)



# DFS and Maze Traversal

- The DFS algorithm is similar to a classic strategy for exploring a maze
  - We mark each intersection, corner and dead end (vertex) visited
  - We mark each corridor (edge ) traversed
  - We keep track of the path back to the entrance (start vertex) by means of a rope (recursion stack)





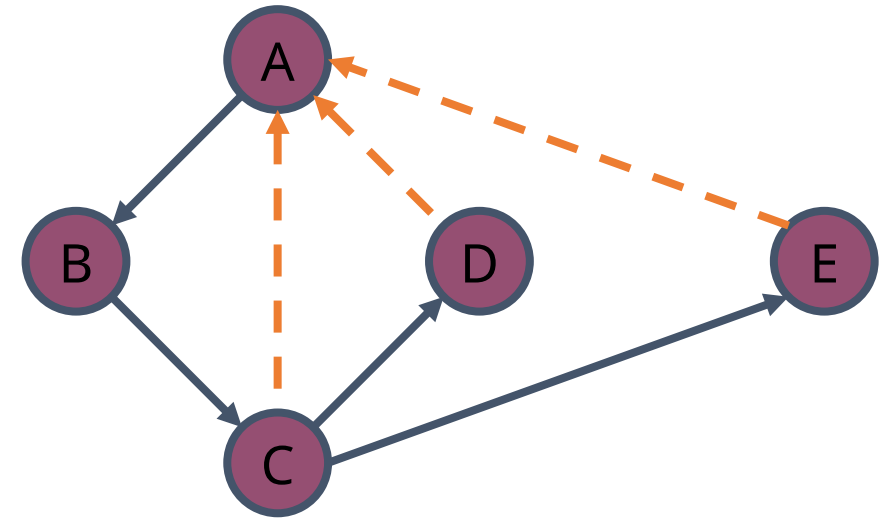
# Properties of DFS

## Property 1

$DFS(G, v)$  visits all the vertices and edges in the connected component of  $v$

## Property 2

The discovery edges labeled by  $DFS(G, v)$  form a spanning tree of the connected component of  $v$



# Analysis of DFS

- Setting/getting a vertex/edge label takes  $O(1)$  time
- Each vertex is labeled twice
  - once as UNEXPLORED
  - once as VISITED
- Each edge is labeled twice
  - once as UNEXPLORED
  - once as DISCOVERY or BACK
- Method incidentEdges is called once for each vertex
- DFS runs in  $O(n + m)$  time provided the graph is represented by the adjacency list structure
  - Recall that  $\sum_v \deg(v) = 2m$

# Path Finding

- We can specialize the DFS algorithm to find a path between two given vertices  $u$  and  $z$  using the template method pattern
- We call  $DFS(G, u)$  with  $u$  as the start vertex
- We use a stack  $S$  to keep track of the path between the start vertex and the current vertex
- As soon as destination vertex  $z$  is encountered, we return the path as the contents of the stack

```

Algorithm pathDFS( $G, v, z$ )
    setLabel( $v, VISITED$ )
     $S.push(v)$ 
    if  $v = z$ 
        return  $S.elements()$ 
    for all  $e \in G.incidentEdges(v)$ 
        if getLabel( $e$ ) = UNEXPLORED
             $w \Leftarrow opposite(v, e)$ 
            if getLabel( $w$ ) = UNEXPLORED
                setLabel( $e, DISCOVERY$ )
                 $S.push(e)$ 
                pathDFS( $G, w, z$ )
                 $S.pop(e)$ 
            else
                setLabel( $e, BACK$ )
     $S.pop(v)$ 
    
```

# Cycle Finding

- We can specialize the DFS algorithm to find a simple cycle using the template method pattern
- We use a stack  $S$  to keep track of the path between the start vertex and the current vertex
- As soon as a back edge  $(v, w)$  is encountered, we return the cycle as the portion of the stack from the top to vertex  $w$

```

Algorithm cycleDFS( $G, v, z$ )
    setLabel( $v, VISITED$ )
     $S.push(v)$ 
    for all  $e \in G.incidentEdges(v)$ 
        if getLabel( $e$ ) = UNEXPLORED
             $w \Leftarrow opposite(v, e)$ 
             $S.push(e)$ 
            if getLabel( $w$ ) = UNEXPLORED
                setLabel( $e, DISCOVERY$ )
                pathDFS( $G, w, z$ )
                 $S.pop(e)$ 
            else
                 $T \Leftarrow$  new empty stack
                repeat
                     $o \Leftarrow S.pop()$ 
                     $T.push(o)$ 
                until  $o = w$ 
                return  $T.elements()$ 
     $S.pop(v)$ 
  
```

# Breadth-First Search

- Breadth-first search (BFS) is a general technique for traversing a graph
- A BFS traversal of a graph  $G$ 
  - Visits all the vertices and edges of  $G$
  - Determines whether  $G$  is connected
  - Computes the connected components of  $G$
  - Computes a spanning forest of  $G$
- BFS on a graph with  $n$  vertices and  $m$  edges takes  $O(n + m)$  time
- BFS can be further extended to solve other graph problems
  - Find and report a path with the minimum number of edges between two given vertices
  - Find a simple cycle, if there is one

# BFS Algorithm

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

## Algorithm *BFS(G)*

**Input** graph *G*

**Output** labeling of the edges  
and partition of the  
vertices of *G*

```
for all  $u \in G.vertices()$ 
     $setLabel(u, UNEXPLORED)$ 
for all  $e \in G.edges()$ 
     $setLabel(e, UNEXPLORED)$ 
for all  $v \in G.vertices()$ 
    if  $getLabel(v) =$ 
        $UNEXPLORED$ 
        $BFS(G, v)$ 
```

## Algorithm *BFS(G, s)*

$L_0 \Leftarrow$  new empty sequence

$L_0.addLast(s)$

$setLabel(s, VISITED)$

$i \Leftarrow 0$

**while**  $\neg L_i.isEmpty()$

$L_{i+1} \Leftarrow$  new empty sequence

**for all**  $v \in L_i.elements()$

**for all**  $e \in G.incidentEdges(v)$

**if**  $getLabel(e) = UNEXPLORED$

$w \Leftarrow opposite(v, e)$

**if**  $getLabel(w) = UNEXPLORED$

$setLabel(e, DISCOVERY)$

$setLabel(w, VISITED)$

$L_{i+1}.addLast(w)$

**else**

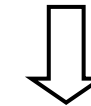
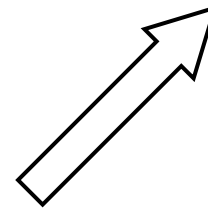
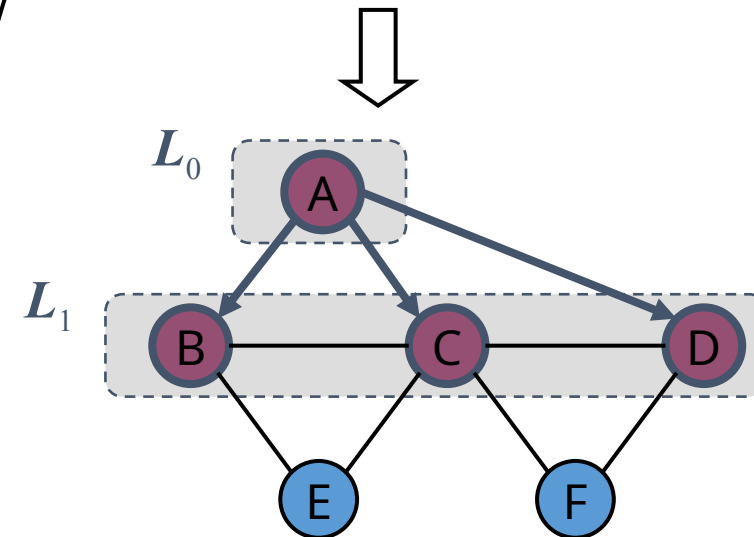
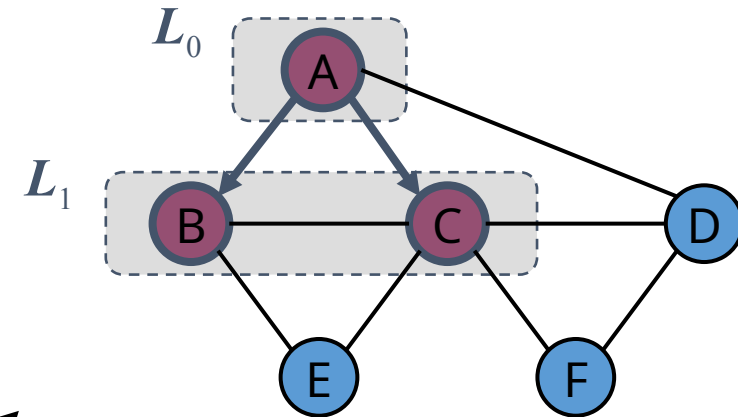
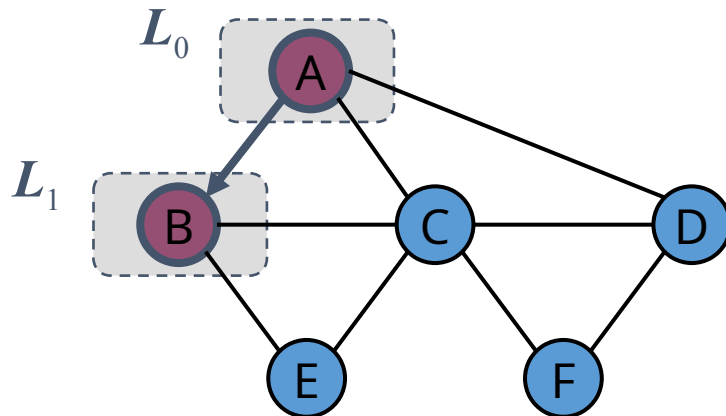
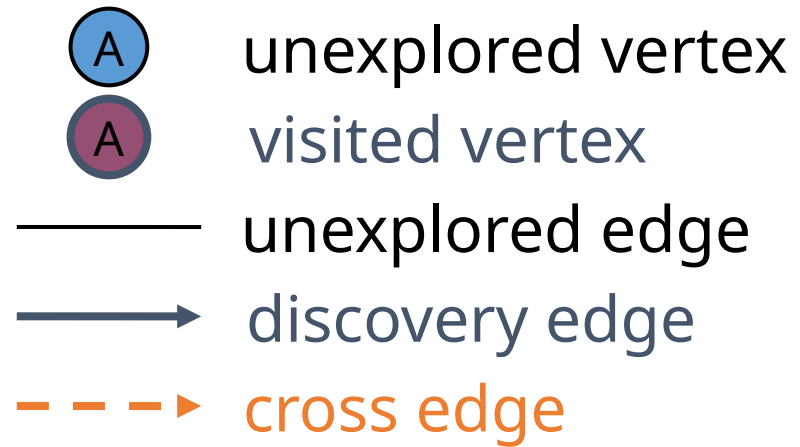
$setLabel(e, CROSS)$

$i \Leftarrow i + 1$

# Python Implementation

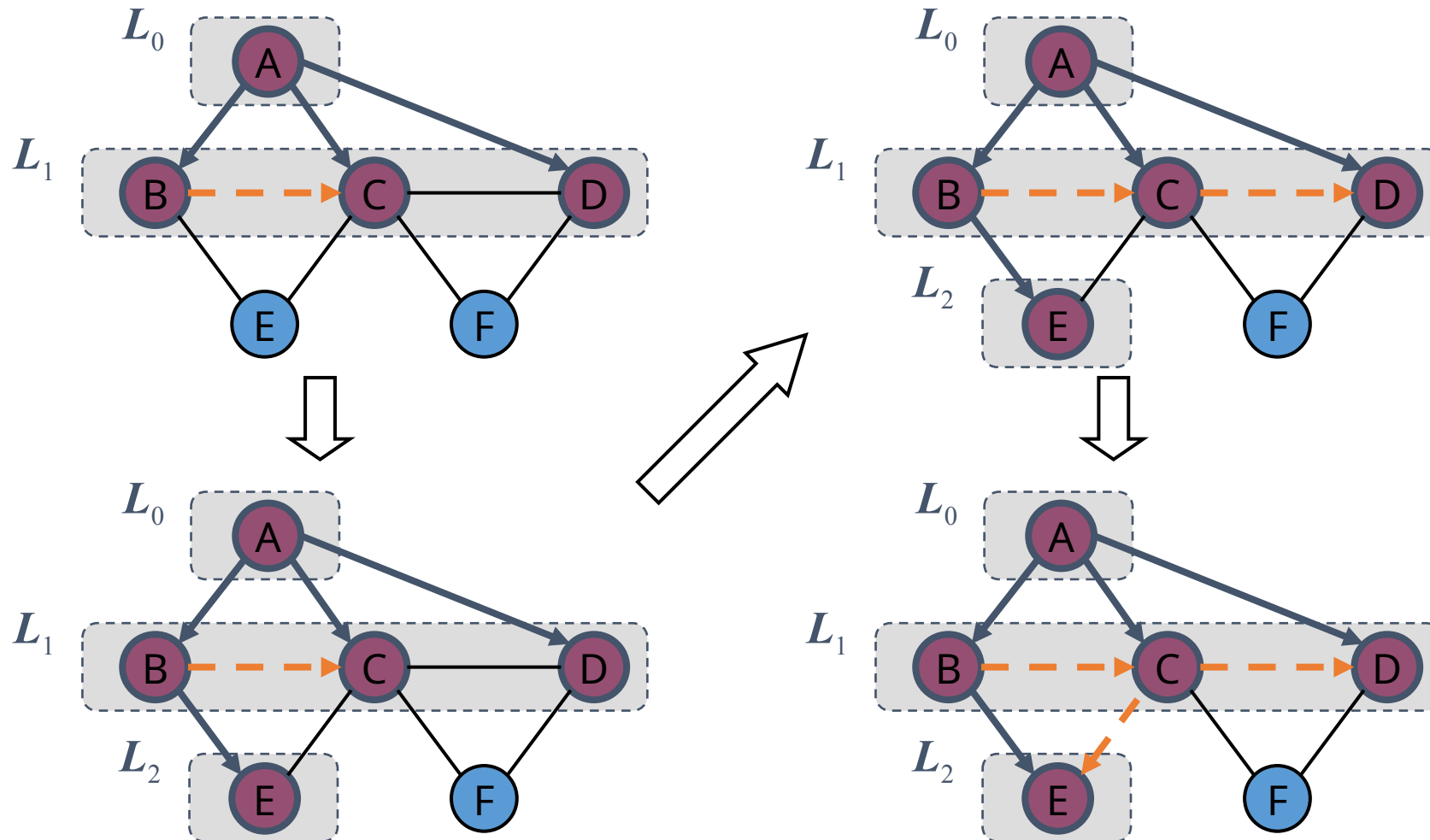
```
1 def BFS(g, s, discovered):
2     """Perform BFS of the undiscovered portion of Graph g starting at Vertex s.
3
4     discovered is a dictionary mapping each vertex to the edge that was used to
5     discover it during the BFS (s should be mapped to None prior to the call).
6     Newly discovered vertices will be added to the dictionary as a result.
7     """
8     level = [s]                # first level includes only s
9     while len(level) > 0:
10         next_level = [ ]      # prepare to gather newly found vertices
11         for u in level:
12             for e in g.incident_edges(u): # for every outgoing edge from u
13                 v = e.opposite(u)
14                 if v not in discovered: # v is an unvisited vertex
15                     discovered[v] = e # e is the tree edge that discovered v
16                     next_level.append(v) # v will be further considered in next pass
17         level = next_level     # relabel 'next' level to become current
```

# Example

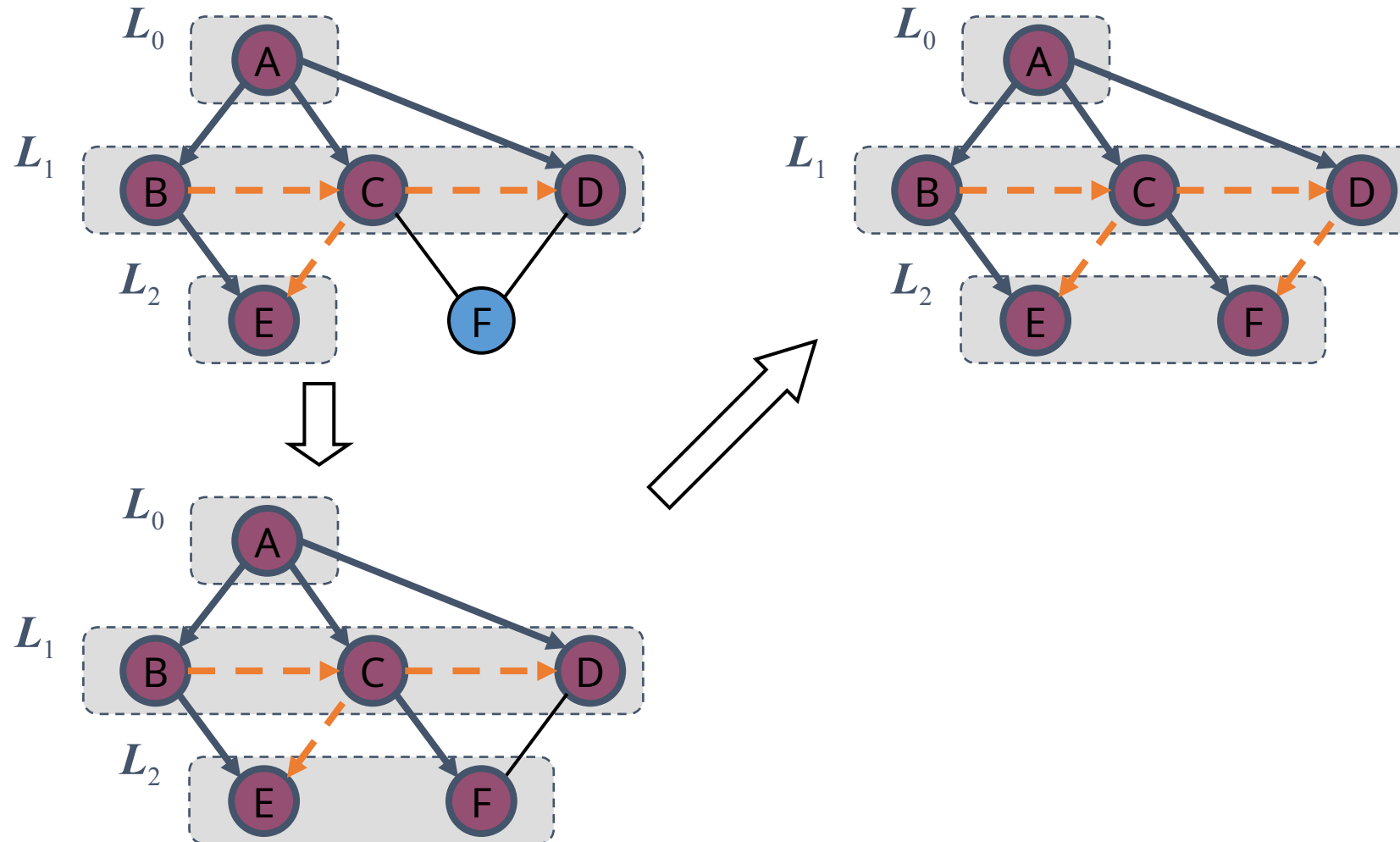




# Example (cont.)



# Example (cont.)



# Properties

## Notation

$G_s$ : connected component of  $s$

## Property 1

$BFS(G, s)$  visits all the vertices and edges of  $G_s$

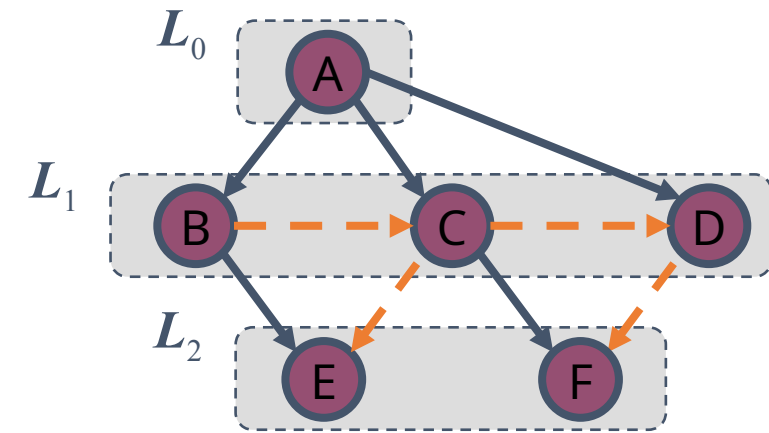
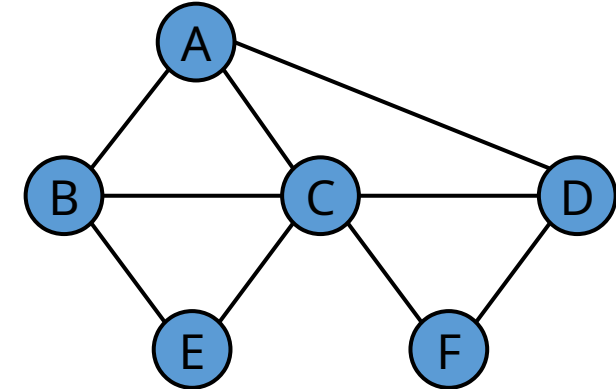
## Property 2

The discovery edges labeled by  $BFS(G, s)$  form a spanning tree  $T_s$  of  $G_s$

## Property 3

For each vertex  $v$  in  $L_i$

- The path of  $T_s$  from  $s$  to  $v$  has  $i$  edges
- Every path from  $s$  to  $v$  in  $G_s$  has at least  $i$  edges



# Analysis

- Setting/getting a vertex/edge label takes  $O(1)$  time
- Each vertex is labeled twice
  - once as UNEXPLORED
  - once as VISITED
- Each edge is labeled twice
  - once as UNEXPLORED
  - once as DISCOVERY or CROSS
- Each vertex is inserted once into a sequence  $L_i$
- Method incidentEdges is called once for each vertex
- BFS runs in  $O(n + m)$  time provided the graph is represented by the adjacency list structure
  - Recall that  $\sum_v \deg(v) = 2m$

# Applications

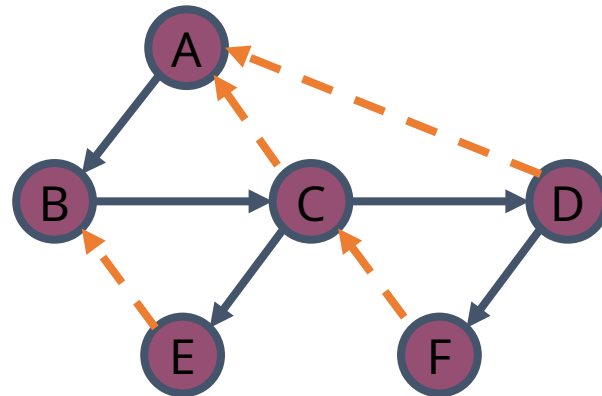
- Using the **template method pattern**, we can specialize the BFS traversal of a graph  $G$  to solve the following problems in  $O(n + m)$  time
  - Compute the connected components of  $G$
  - Compute a spanning forest of  $G$
  - Find a simple cycle in  $G$ , or report that  $G$  is a forest
  - Given two vertices of  $G$ , find a path in  $G$  between them with the minimum number of edges, or report that no such path exists

# DFS vs BFS

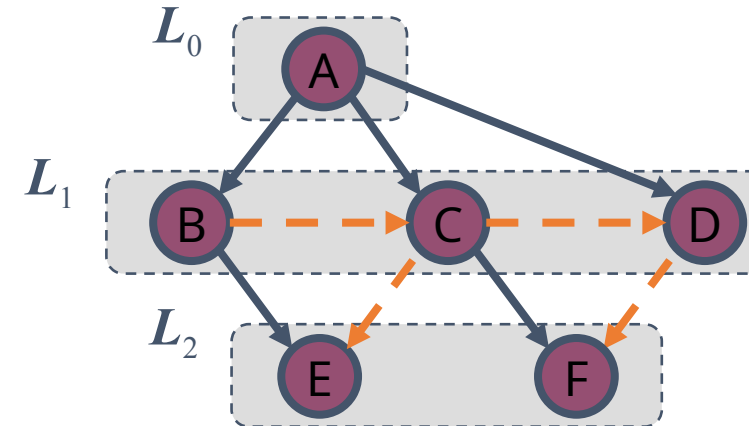
	Depth First Search (DFS)	Breadth First Search (BFS)
Data Structure	DFS uses Stack data structure.	BFS uses Queue data structure
Definition	DFS traversal approach in which the traverse begins at the root node and proceeds through the nodes as far as possible until we reach the node with no unvisited nearby nodes.	BFS is a traversal approach in which we first walk through all nodes on the same level before moving on to the next level.
Conceptual Difference	DFS builds the tree sub-tree by sub-tree.	BFS builds the tree level by level.
Approach used	It works on the concept of <a href="#">LIFO</a> (Last In First Out).	It works on the concept of <a href="#">FIFO</a> (First In First Out).
Suitable for	DFS is more suitable when there are solutions away from source.	BFS is more suitable for searching vertices closer to the given source.
Applications	DFS is used in various applications such as acyclic graphs and finding strongly connected components etc.	BFS is used in various applications such as bipartite graphs, shortest paths, etc.

# DFS vs. BFS

Applications	DFS	BFS
Spanning forest, connected components, paths, cycles	✓	✓
Shortest paths		✓
Biconnected components	✓	



DFS

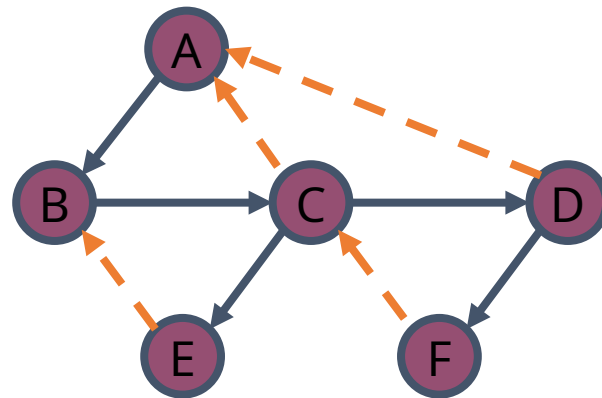


BFS

# DFS vs. BFS (cont.)

## Back edge ( $v, w$ )

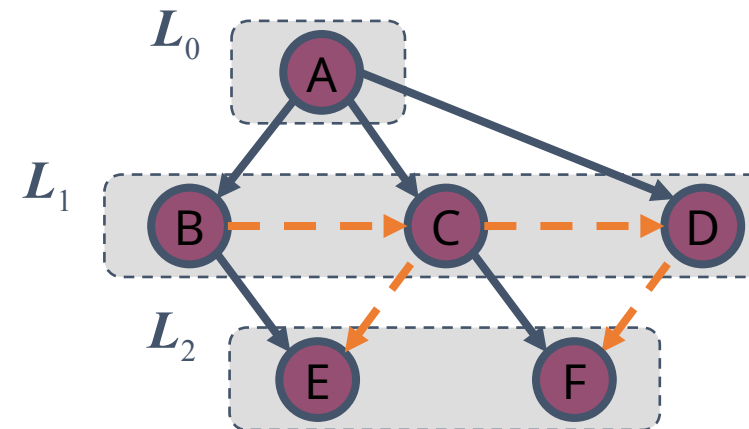
- $w$  is an ancestor of  $v$  in the tree of discovery edges



DFS

## Cross edge ( $v, w$ )

- $w$  is in the same level as  $v$  or in the next level

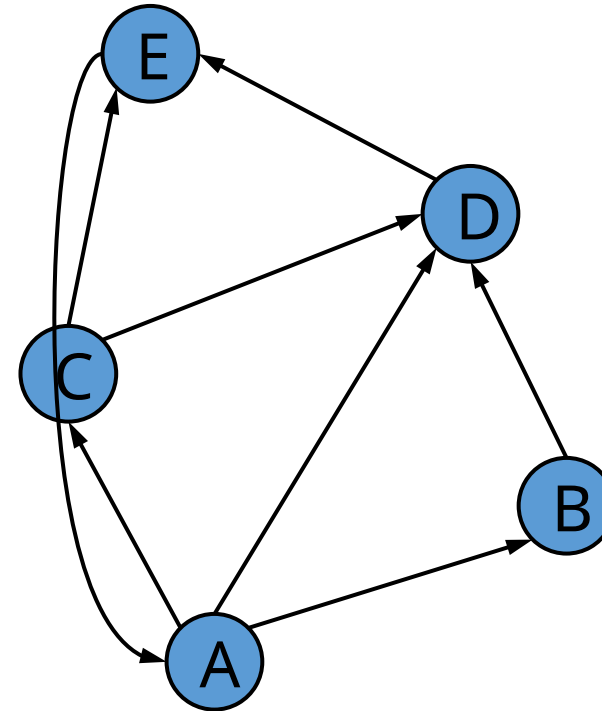


BFS



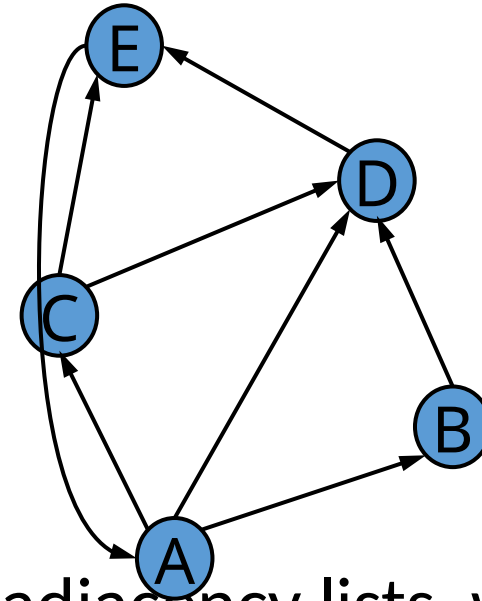
# Directed Graphs

- A **directed graph** is a graph whose edges are all directed
- Applications
  - one-way streets
  - flights
  - task scheduling



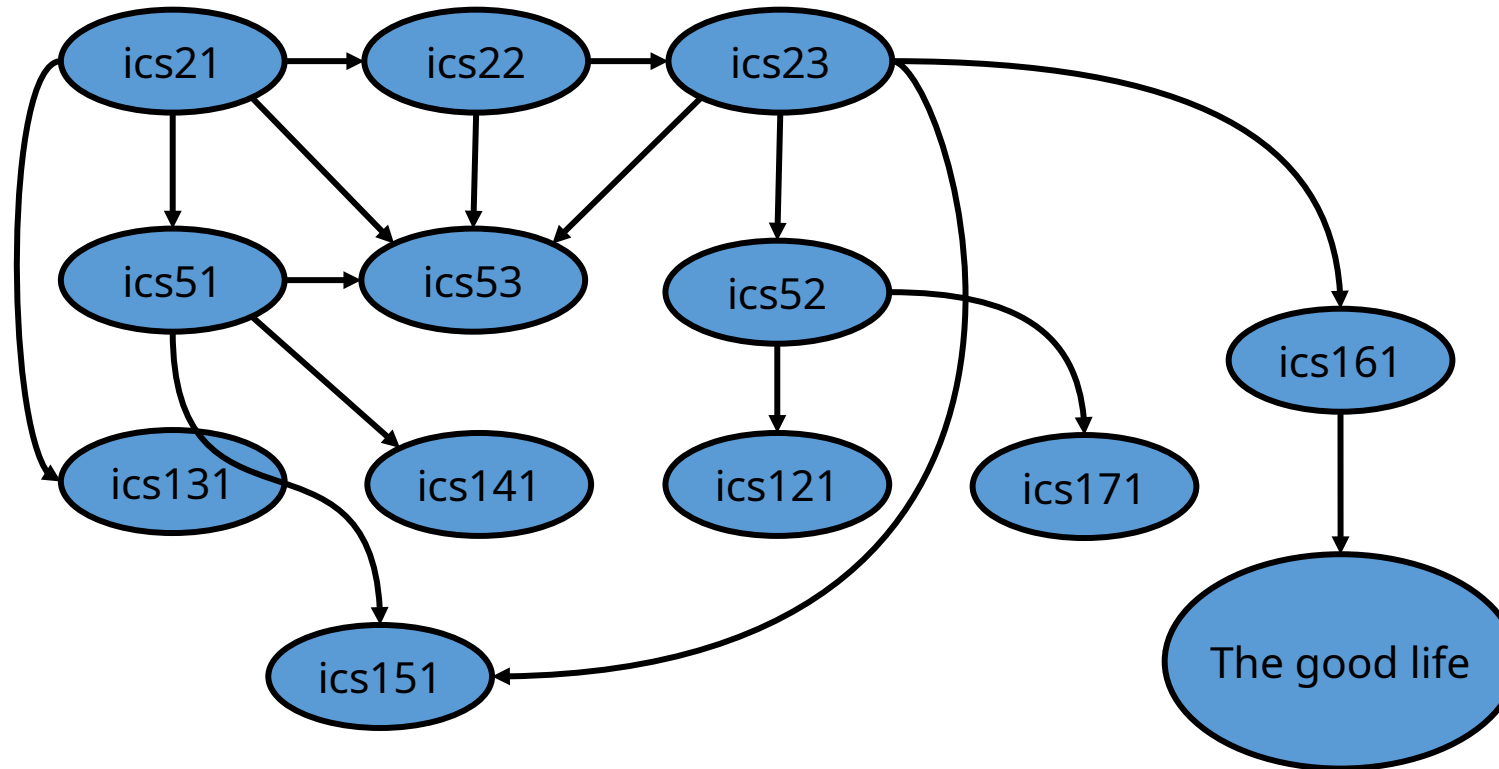
# Directed Graph Properties

- A graph  $G=(V,E)$  such that
  - Each edge goes in **one direction**:
  - Edge  $(a,b)$  goes from  $a$  to  $b$ , but not  $b$  to  $a$
- If  $G$  is simple,  $m \leq n(n-1)$
- If we keep in-edges and out-edges in separate adjacency lists, we can perform listing of incoming edges and outgoing edges in time proportional to their size



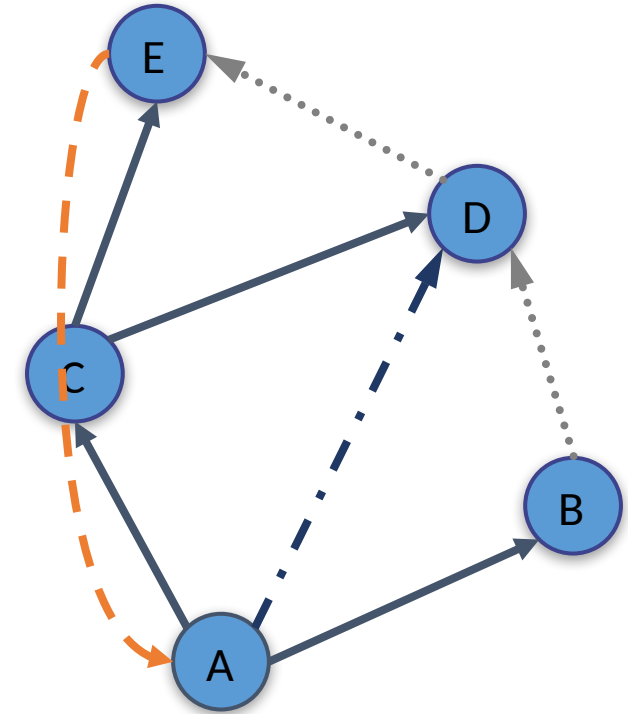
# Directed Graph Application

- **Scheduling:** edge (a,b) means task a must be completed before b can be started



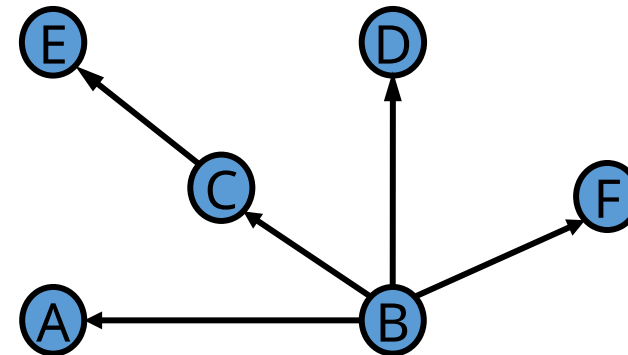
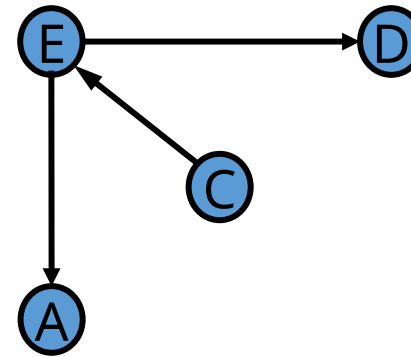
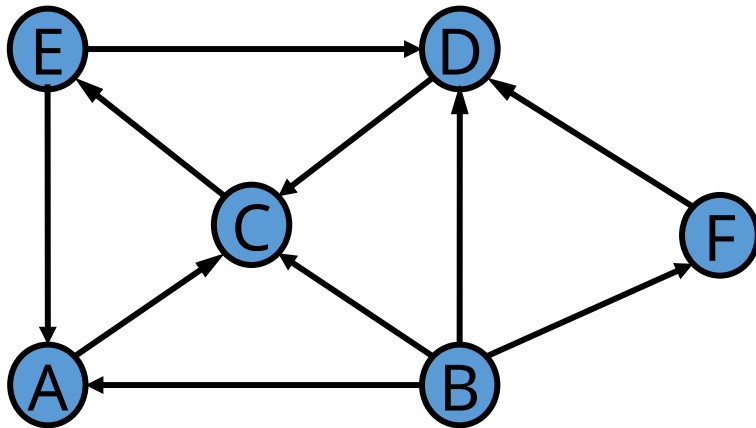
# Directed DFS

- We can specialize the traversal algorithms (DFS and BFS) to digraphs by traversing edges only along their direction
- In the directed DFS algorithm, we have four types of edges
  - discovery edges
  - back edges
  - forward edges
  - cross edges
- A directed DFS starting at a vertex  $s$  determines the vertices reachable from  $s$



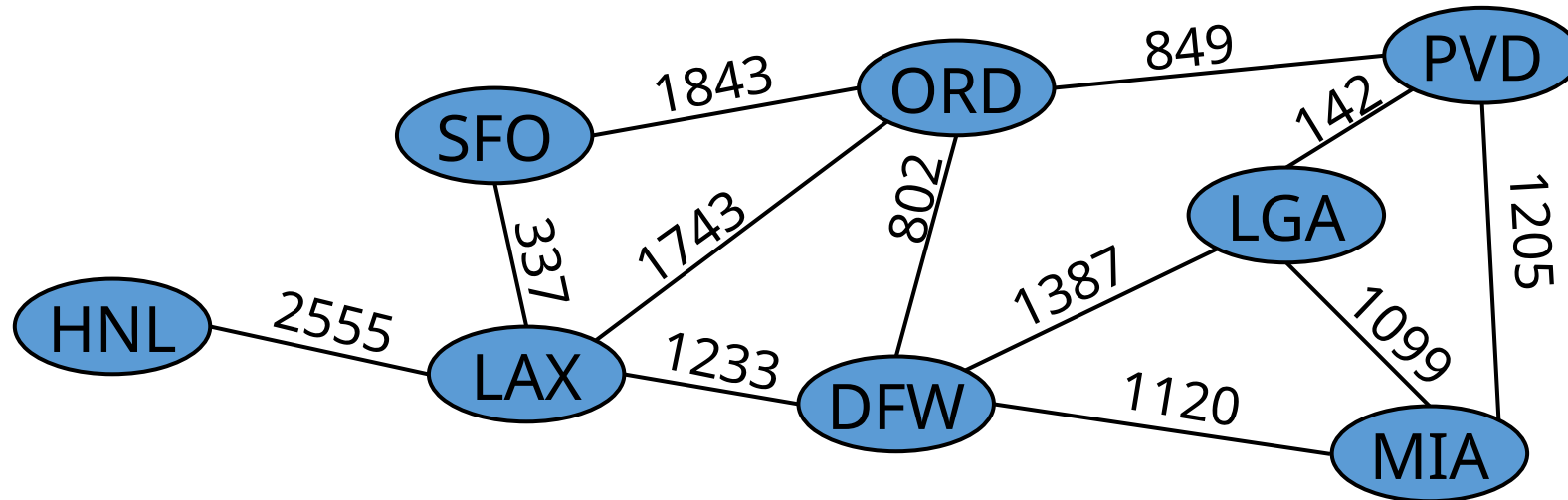
# Reachability

- DFS tree rooted at v: vertices reachable from v via directed paths



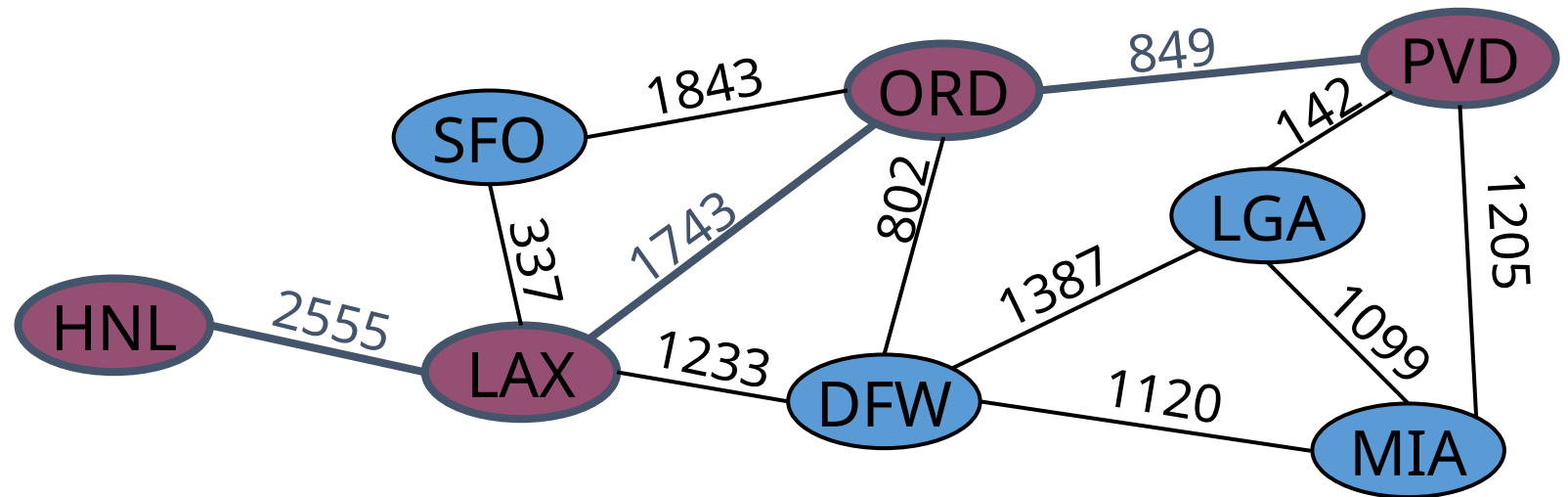
# Weighted Graphs

- In a weighted graph, each edge has an associated numerical value, called the weight of the edge
- Edge weights may represent, distances, costs, etc.
- Example:
  - In a flight route graph, the weight of an edge represents the distance in miles between the endpoint airports



# Shortest Paths

- Given a weighted graph and two vertices  $u$  and  $v$ , we want to find a path of minimum total weight between  $u$  and  $v$ .
  - Length of a path is the sum of the weights of its edges.
- Example:
  - Shortest path between Providence and Honolulu
- Applications
  - Internet packet routing
  - Flight reservations
  - Driving directions



# Shortest Path Properties

Property 1:

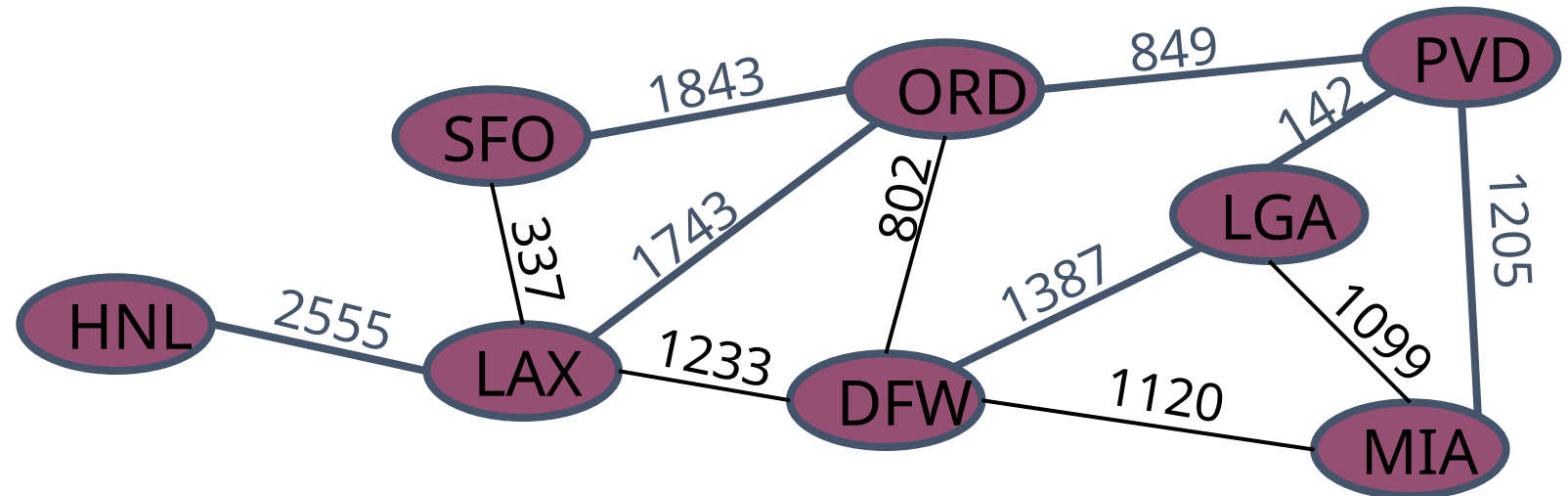
A subpath of a shortest path is itself a shortest path

Property 2:

There is a tree of shortest paths from a start vertex to all the other vertices

Example:

Tree of shortest paths from Providence





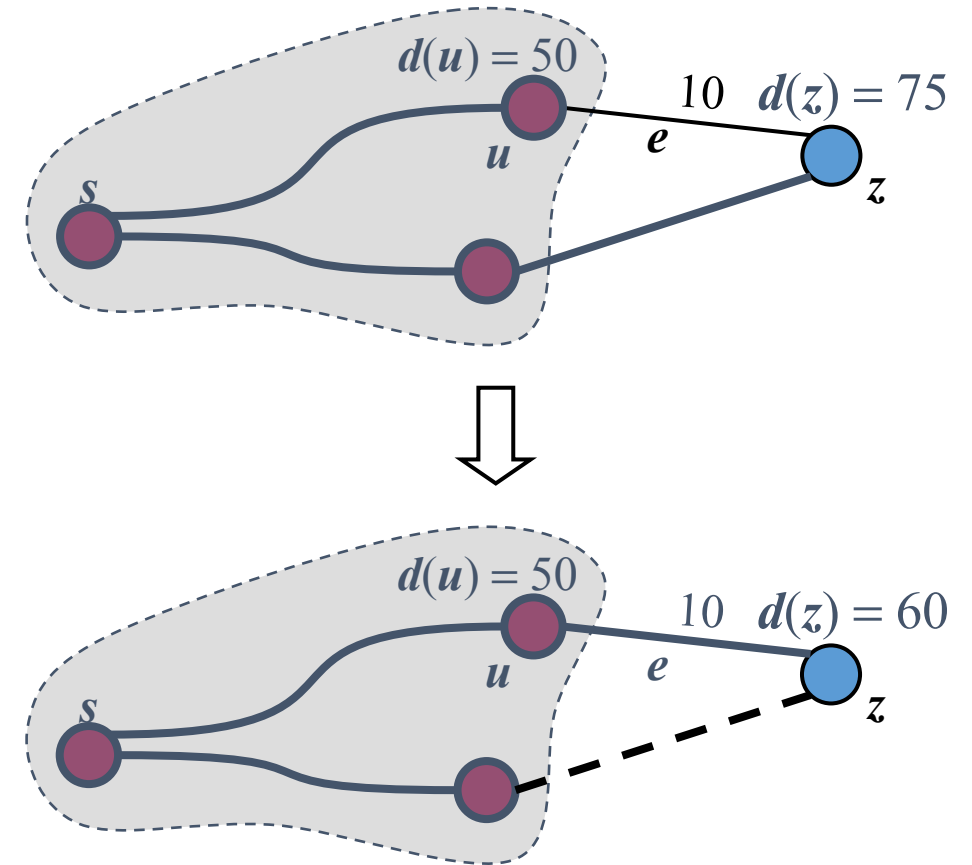
# Dijkstra's Algorithm

- The distance of a vertex  $v$  from a vertex  $s$  is the length of a shortest path between  $s$  and  $v$
- Dijkstra's algorithm computes the distances of all the vertices from a given start vertex  $s$
- Assumptions:
  - the graph is connected
  - the edges are undirected
  - the edge weights are nonnegative
- We grow a “cloud” of vertices, beginning with  $s$  and eventually covering all the vertices
- We store with each vertex  $v$  a label  $d(v)$  representing the distance of  $v$  from  $s$  in the subgraph consisting of the cloud and its adjacent vertices
- At each step
  - We add to the cloud the vertex  $u$  outside the cloud with the smallest distance label,  $d(u)$
  - We update the labels of the vertices adjacent to  $u$

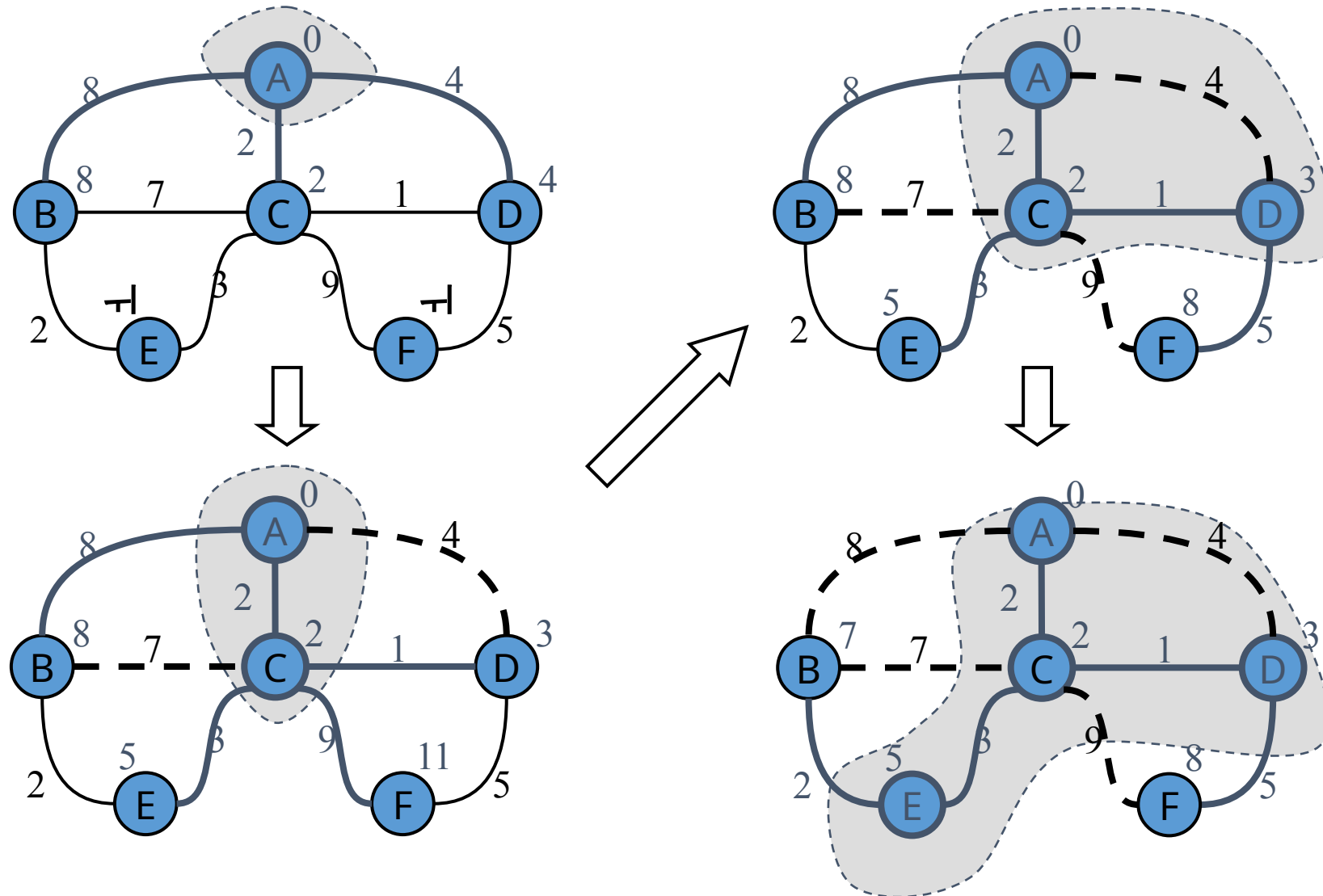
# Edge Relaxation

- Consider an edge  $e = (u, z)$  such that
  - $u$  is the vertex most recently added to the cloud
  - $z$  is not in the cloud
- The relaxation of edge  $e$  updates distance  $d(z)$  as follows:

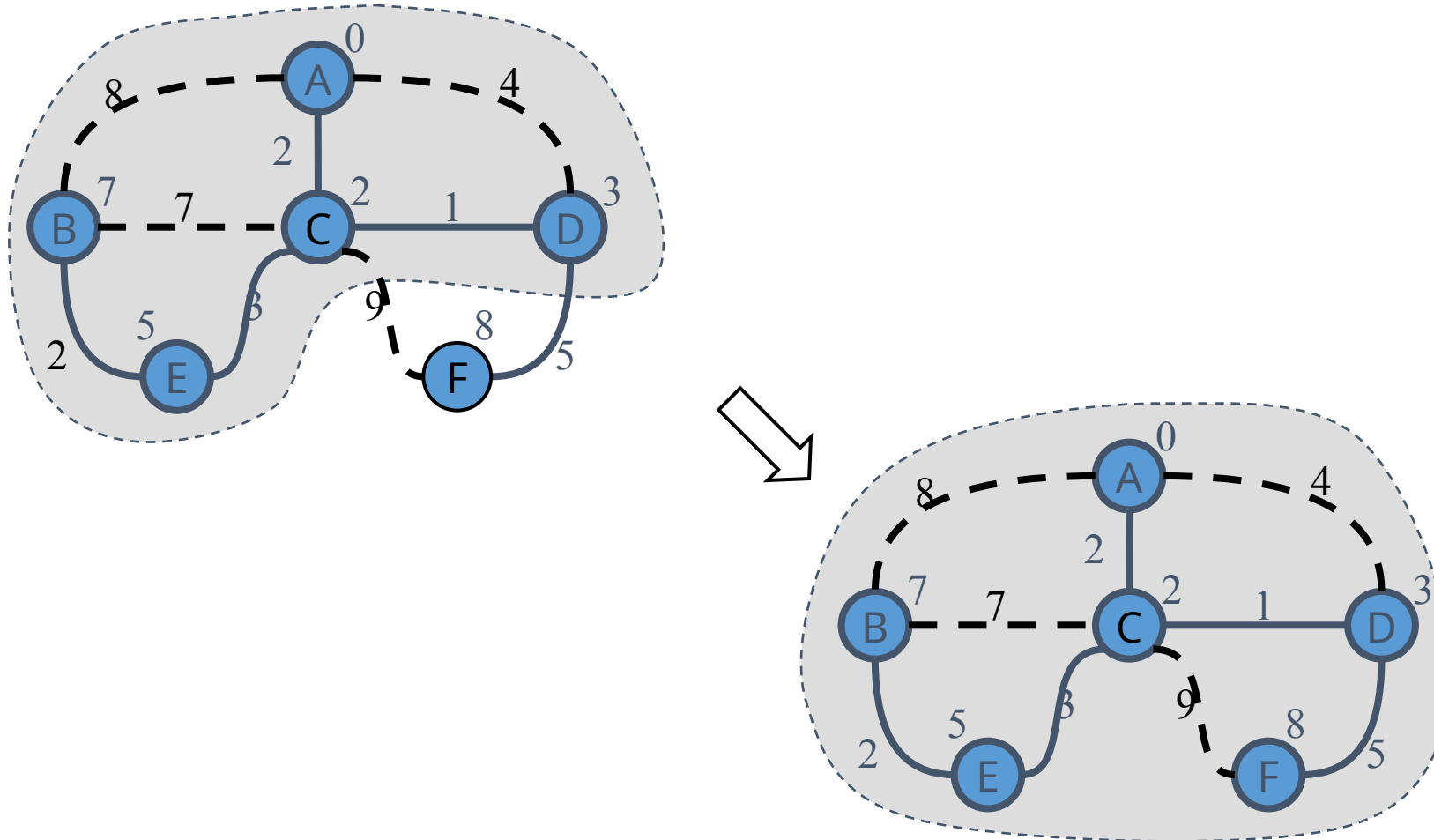
$$d(z) \Leftarrow \min\{d(z), d(u) + \text{weight}(e)\}$$



# Example



# Example (cont.)



# Dijkstra's Algorithm

**Algorithm** ShortestPath( $G, s$ ):

**Input:** A weighted graph  $G$  with nonnegative edge weights, and a distinguished vertex  $s$  of  $G$ .

**Output:** The length of a shortest path from  $s$  to  $v$  for each vertex  $v$  of  $G$ .

Initialize  $D[s] = 0$  and  $D[v] = \infty$  for each vertex  $v \neq s$ .

Let a priority queue  $Q$  contain all the vertices of  $G$  using the  $D$  labels as keys.

**while**  $Q$  is not empty **do**

    {pull a new vertex  $u$  into the cloud}

$u =$  value returned by  $Q.remove\_min()$

**for** each vertex  $v$  adjacent to  $u$  such that  $v$  is in  $Q$  **do**

        {perform the *relaxation* procedure on edge  $(u, v)$ }

**if**  $D[u] + w(u, v) < D[v]$  **then**

$D[v] = D[u] + w(u, v)$

            Change to  $D[v]$  the key of vertex  $v$  in  $Q$ .

**return** the label  $D[v]$  of each vertex  $v$

# Analysis of Dijkstra's Algorithm

- Graph operations
  - We find all the incident edges once for each vertex
- Label operations
  - We set/get the distance and locator labels of vertex  $z$   $O(\deg(z))$  times
  - Setting/getting a label takes  $O(1)$  time
- Priority queue operations
  - Each vertex is inserted once into and removed once from the priority queue, where each insertion or removal takes  $O(\log n)$  time
  - The key of a vertex in the priority queue is modified at most  $\deg(w)$  times, where each key change takes  $O(\log n)$  time
- Dijkstra's algorithm runs in  $O((n + m) \log n)$  time provided the graph is represented by the adjacency list/map structure
  - Recall that  $\sum_v \deg(v) = 2m$
- The running time can also be expressed as  $O(m \log n)$  since the graph is connected

# Python Implementation

```
1 def shortest_path_lengths(g, src):
2     """ Compute shortest-path distances from src to reachable vertices of g.
3
4     Graph g can be undirected or directed, but must be weighted such that
5     e.element() returns a numeric weight for each edge e.
6
7     Return dictionary mapping each reachable vertex to its distance from src.
8     """
9     d = { } # d[v] is upper bound from s to v
10    cloud = { } # map reachable v to its d[v] value
11    pq = AdaptableHeapPriorityQueue( ) # vertex v will have key d[v]
12    pqlocator = { } # map from vertex to its pq locator
13
14    # for each vertex v of the graph, add an entry to the priority queue, with
15    # the source having distance 0 and all others having infinite distance
16    for v in g.vertices():
17        if v is src:
18            d[v] = 0
19        else:
20            d[v] = float('inf') # syntax for positive infinity
21            pqlocator[v] = pq.add(d[v], v) # save locator for future updates
22
23    while not pq.is_empty():
24        key, u = pq.remove_min()
25        cloud[u] = key # its correct d[u] value
26        del pqlocator[u] # u is no longer in pq
27        for e in g.incident_edges(u): # outgoing edges (u,v)
28            v = e.opposite(u)
29            if v not in cloud:
30                # perform relaxation step on edge (u,v)
31                wgt = e.element()
32                if d[u] + wgt < d[v]: # better path to v?
33                    d[v] = d[u] + wgt # update the distance
34                    pq.update(pqlocator[v], d[v], v) # update the pq entry
35
36    return cloud # only includes reachable vertices
```

# Why Dijkstra's Algorithm Works

- Dijkstra's algorithm is based on the greedy method. It adds vertices by increasing distance.
- Suppose it didn't find all shortest distances. Let F be the first wrong vertex the algorithm processed.
- When the previous node, D, on the true shortest path was considered, its distance was correct
- But the edge (D,F) was relaxed at that time!
- Thus, so long as  $d(F) \geq d(D)$ , F's distance cannot be wrong. That is, there is no wrong vertex

