

CS218- Data Structures

Week 14

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Agenda

- Graphs
- Graphs Basic Terminologies
- Graph Data Structures
 - Graph Representation
 - Graph Algorithms
- Breadth First Search
- Depth First Search
- Shortest Path in a Graph
- All Pairs Shortest Paths
- Cycle Detections

Graphs

■ Trees

- Although Trees are very flexible data structures, they have some limitations
 - They can only model only a single type of hierarchical relationship – Parent -> Child
 - Knowledge about Sibling are hard to get through trees.
 - You can not climb back the tree once, you are down from a root pointer.

■ Graphs

- There is no such limitations in Graphs.
- These are more flexible and representational in nature

Graphs

■ Graphs

- Mathematically Graph $G=(V,E)$, where V is a non-empty set and E can be an empty set.

■ Simple Graph

- A simple graph $G = (V, E)$ consists of a nonempty set V of vertices and a possibly empty set E of edges, each edge being a set of two vertices from V .
- The number of vertices and edges is denoted by $|V|$ and $|E|$, respectively.

Graphs

■ Multi Graph

- A multigraph is a graph in which two vertices can be joined by multiple edges.
- Geometric interpretation is very simple Formally, the definition is as follows: a multigraph $G = (V, E, f)$ is composed of a set of vertices V , a set of edges E , and a function $f : E \rightarrow \{\{v_i, v_j\} : v_i, v_j \in V \text{ and } v_i \neq v_j\}$.

■ Pseudograph

- A pseudograph is a multigraph with the condition $v_i \neq v_j$ removed, which allows for loops to occur; in a pseudograph, a vertex can be joined with itself by an edge.

Graphs

■ Directed Graphs

- A directed graph, or a digraph, $G = (V, E)$ consists of a nonempty set V of vertices and a set E of edges (also called arcs), where each edge is a pair of vertices from V .
- The difference is that one edge of a simple graph is of the form $\{v_i, v_j\}$, and for such an edge, $\{v_i, v_j\} = \{v_j, v_i\}$.
- In a digraph, each edge is of the form (v_i, v_j) , and in this case, $(v_i, v_j) \neq (v_j, v_i)$

Graphs

■ Weighted Graphs

- A graph is called a weighted graph if each edge has an assigned number.
- Depending on the context in which such graphs are used, the number assigned to an edge is called its weight, cost, distance, length, or some other name.

Graphs

■ Complete Graphs

- A graph with n vertices is called complete and is denoted K_n if for each pair of distinct vertices there is exactly one edge connecting them; that is, each vertex can be connected to any other vertex.
- The number of edges in such a graph $|E| = |V|^2$.

■ Connected Graph

- All vertices are connected through some edges. A single component.

■ Isolated Graph

- The subsets of vertices may not be connected as a single component.

Graphs

■ Path

- A path from v_1 to v_n is a sequence of edges $\text{edge}(v_1, v_2)$, $\text{edge}(v_2, v_3)$, \dots , $\text{edge}(v_{n-1}, v_n)$ and is denoted as path $v_1, v_2, v_3, \dots, v_{n-1}, v_n$. If $v_1 = v_n$ and no edge is repeated, then the path is called a circuit.

■ Cycle

- If all vertices in a circuit are different, then it is called a *cycle*.

Graph Representations

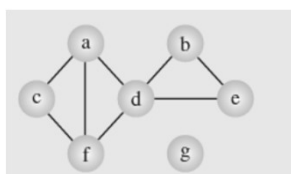
■ Graph Data Structures

- There are several ways in which we can store graphs into memory.
- The representation should be lossless. We can able to reconstruct the unambiguous graph.
- Each representation is good for certain aspect of graph processing.
- All graph processing libraries support internal conversion of these representations.
- It is very common that we may change the representation before opting for an algorithm to solve any given problem on graph.

Graph Representations

■ Adjacency List

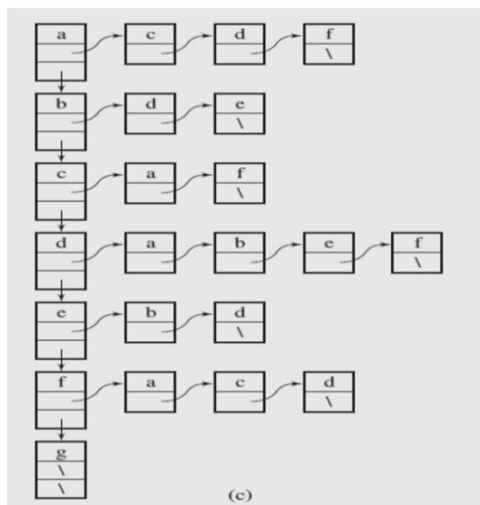
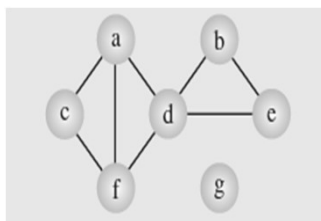
- A simple representation is given by an adjacency list, which specifies all vertices adjacent to each vertex of the graph.
- This list can be implemented as a table, in which case it is called a star representation, which can be forward or reverse.



a	c	d	f
b	d	e	
c	a	f	
d	a	b	e
e	b	d	
f	a	c	d
g			

Graph Representations

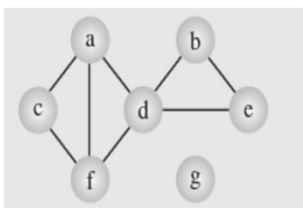
■ Adjacency List



Graph Representations

■ Adjacency Matrix

- An adjacency matrix of graph $G = (V, E)$ is a binary $|V| \times |V|$ matrix.
- Such that each entry of this matrix a_{ij} is 1 if v_i is connected to v_j .

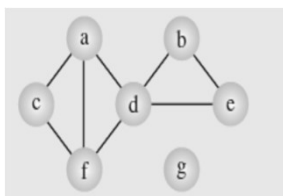


	a	b	c	d	e	f	g
a	0	0	1	1	0	1	0
b	0	0	0	1	1	0	0
c	1	0	0	0	0	1	0
d	1	1	0	0	1	1	0
e	0	1	0	1	0	0	0
f	1	0	1	1	0	0	0
g	0	0	0	0	0	0	0

Graph Representations

■ Incidence Matrix

- An Incidence matrix of graph $G = (V, E)$ is a binary $|V| \times |E|$ matrix.
- Such that each entry of this matrix a_{ij} is 1 if v_i is connected to any edge.



	ac	ad	af	bd	be	cf	de	df
a	1	1	1	0	0	0	0	0
b	0	0	0	1	1	0	0	0
c	1	0	0	0	0	1	0	0
d	0	1	0	1	0	0	1	1
e	0	0	0	0	1	0	1	0
f	0	0	1	0	0	1	0	1
g	0	0	0	0	0	0	0	0

Adjacency List vs. Adjacency Matrix

Operations	Adjacency Matrix	Adjacency List
Storage Space	This representation makes use of $V \times V$ matrix, so space required in worst case is $O(V ^2)$.	In this representation, for every vertex we store its neighbours. In the worst case, if a graph is connected $O(V)$ is required for a vertex and $O(E)$ is required for storing neighbours corresponding to every vertex. Thus, overall space complexity is $O(V + E)$.
Adding a vertex	In order to add a new vertex to $V \times V$ matrix the storage must be increased to $(V +1)^2$. To achieve this we need to copy the whole matrix. Therefore the complexity is $O(V ^2)$.	There are two pointers in adjacency list first points to the front node and the other one points to the rear node. Thus insertion of a vertex can be done directly in $O(1)$ time .

Adjacency List vs. Adjacency Matrix

Operations	Adjacency Matrix	Adjacency List
Adding an edge	To add an edge say from i to j , $matrix[i][j] = 1$ which requires $O(1)$ time .	Similar to insertion of vertex here also two pointers are used pointing to the rear and front of the list. Thus, an edge can be inserted in $O(1)$ time .
Removing a vertex	In order to remove a vertex from $V \times V$ matrix the storage must be decreased to $ V ^2$ from $(V +1)^2$. To achieve this we need to copy the whole matrix. Therefore the complexity is $O(V ^2)$.	In order to remove a vertex, we need to search for the vertex which will require $O(V)$ time in worst case, after this we need to traverse the edges and in worst case it will require $O(E)$ time. Hence, total time complexity is $O(V + E)$.

Adjacency List vs. Adjacency Matrix

Operations	Adjacency Matrix	Adjacency List
Removing an edge	To remove an edge say from i to j , $\text{matrix}[i][j] = 0$ which requires $O(1)$ time.	To remove an edge traversing through the edges is required and in worst case we need to traverse through all the edges. Thus, the time complexity is $O(E)$.
Querying	In order to find for an existing edge the content of matrix needs to be checked. Given two vertices say i and j $\text{matrix}[i][j]$ can be checked in $O(1)$ time.	In an adjacency list every vertex is associated with a list of adjacent vertices. For a given graph, in order to check for an edge we need to check for vertices adjacent to given vertex. A vertex can have at most $O(V)$ neighbours and in worst case we would have to check for every adjacent vertex. Therefore, time complexity is $O(V)$.

Graph Traversals

- Graph traversals are more challenging than Tree traversals
 - Cycles in the graphs
 - Isolated Graphs – disconnected components.
- Graph Search
 - Breadth First Search
 - Depth First Search

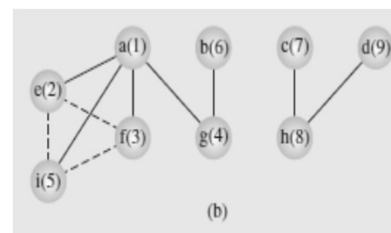
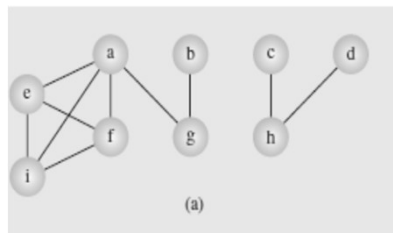
Breadth First Search

```

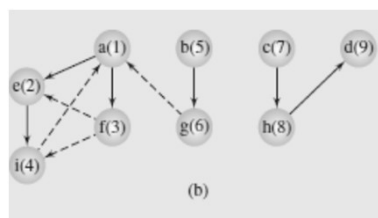
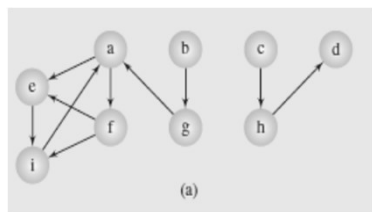
breadthFirstSearch()
  for all vertices u
    num(u) = 0;
  edges = null;
  i = 1;
  while there is a vertex v such that num(v) is 0
    num(v) = i++;
    enqueue(v);
    while queue is not empty
      v = dequeue();
      for all vertices u adjacent to v
        if num(u) is 0
          num(u) = i++;
          enqueue(u);
          attach edge(vu) to edges;
  output edges;

```

Breadth First Search



Breadth First Search



Depth First Search

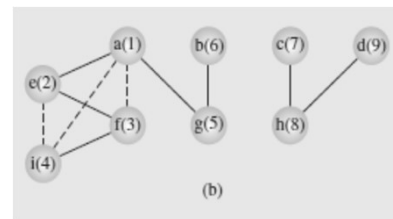
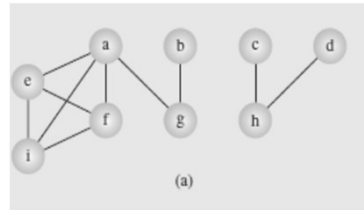
```

DFS(v)
    num(v) = i++;
    for all vertices u adjacent to v
        if num(u) is 0
            attach edge(uv) to edges;
            DFS(u);

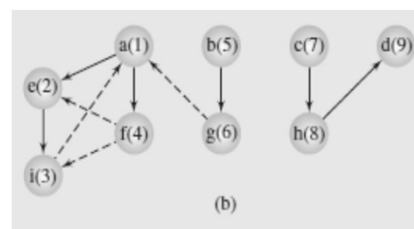
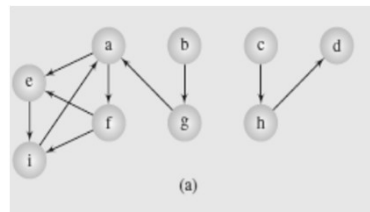
depthFirstSearch()
    for all vertices v
        num(v) = 0;
    edges = null;
    i = 1;
    while there is a vertex v such that num(v) is 0
        DFS(v);
    output edges;

```

Depth First Search



Depth First Search



Shortest Path in a Graph

- Finding the shortest path is a classical problem in graph theory, and a large number of different solutions have been proposed.
 - Edges are assigned certain weights representing different measure like cost of the path, distance, etc.
 - When determining the shortest path from vertex v to vertex u , information about distances between intermediate vertices w has to be recorded.
-

Shortest Path in a Graph

- When determining the shortest path from vertex v to vertex u , information about distances between intermediate vertices w has to be recorded.
 - The methods of finding the shortest path rely on these labels.
 - There are two kinds of methods:
 - Label-setting methods
 - Label correcting methods.
-

Shortest Path in a Graph

■ Label-setting methods

- For label-setting methods, in each pass through the vertices still to be processed, one vertex is set to a value that remains unchanged to the end of the execution.
- These methods limit to processing graphs with only positive weights.

Shortest Path in a Graph

■ Label-Correcting methods

- The label-correcting methods, allows for the changing of any label during application of the method. These methods limit to processing graphs with only positive weights.
- These methods can be applied to graphs with negative weights and with no negative cycle.
- Negative Cycle -a cycle composed of edges with weights adding up to a negative number—but they guarantee that, for all vertices, the current distances indicate the shortest path only after the processing of the graph is finished.

General Shortest Path

```

genericShortestPathAlgorithm(weighted simple digraph, vertex first)
  for all vertices v
    currDist(v) =  $\infty$ ;
  currDist(first) = 0;
  initialize toBeChecked;
  while toBeChecked is not empty
    v = a vertex in toBeChecked;
    remove v from toBeChecked;
    for all vertices u adjacent to v
      if currDist(u) > currDist(v) + weight(edge(vu))
        currDist(u) = currDist(v) + weight(edge(vu));
        predecessor(u) = v;
        add u to toBeChecked if it is not there;

```

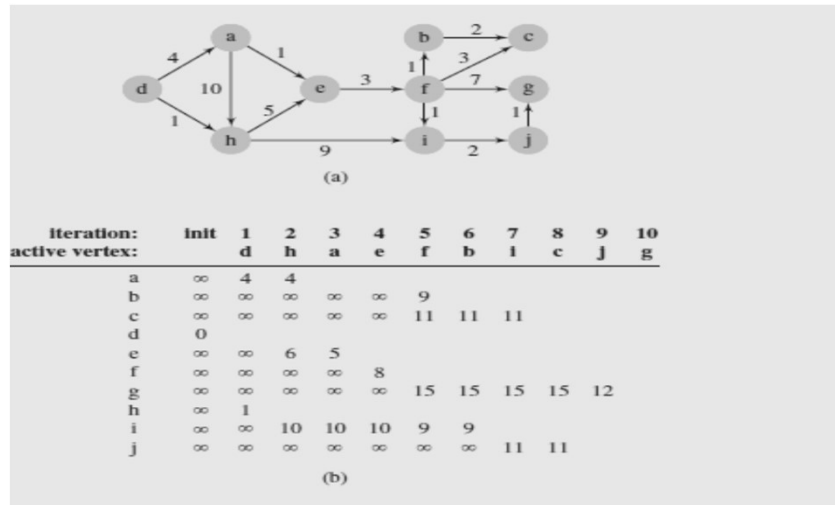
Dijkstra Algorithm

```

DijkstraAlgorithm(weighted simple digraph, vertex first)
  for all vertices v
    currDist(v) =  $\infty$ ;
  currDist(first) = 0;
  toBeChecked = all vertices;
  while toBeChecked is not empty
    v = a vertex in toBeChecked with minimal currDist(v);
    remove v from toBeChecked;
    for all vertices u adjacent to v and in toBeChecked
      if currDist(u) > currDist(v) + weight(edge(vu))
        currDist(u) = currDist(v) + weight(edge(vu));
        predecessor(u) = v;

```

Dijkstra Algorithm

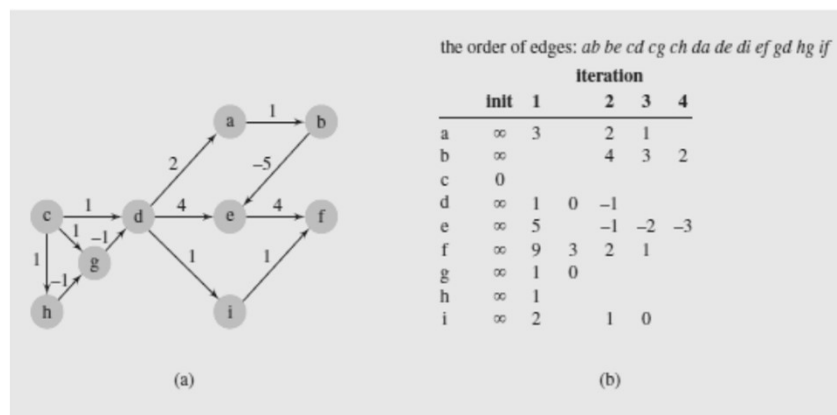


Ford's Algorithm

```

FordAlgorithm(weighted simple digraph, vertex first)
  for all vertices v
    currDist(v) =  $\infty$ ;
  currDist(first) = 0;
  while there is an edge (vu) such that currDist(u) > currDist(v) + weight(edge(vu))
    currDist(u) = currDist(v) + weight(edge(vu));
  
```


Ford's Algorithm



Label-Correcting Algorithm

```

labelCorrectingAlgorithm(weighted simple digraph, vertex first)
  for all vertices v
    currDist(v) =  $\infty$ ;
  currDist(first) = 0;
  toBeChecked = {first};
  while toBeChecked is not empty
    v = a vertex in toBeChecked;
    remove v from toBeChecked;
    for all vertices u adjacent to v
      if currDist(u) > currDist(v) + weight(edge(vu))
        currDist(u) = currDist(v) + weight(edge(vu));
        predecessor(u) = v;
        add u to toBeChecked if it is not there;

```

Label-Correcting Algorithm

FIGURE 8.9 An execution of `labelCorrectingAlgorithm()`, which uses a queue.

	active vertex															
queue	c	d	g	h	a	e	i	d	g	b	f	a	e	i	d	b
	d	g	h	a	e	i	d	g	b	f	a	e	i	d	b	f
	g	h	a	e	i	d	g	b	f	a	e	i	d	b	f	a
	h	a	e	i	d	g	b	f	a	e	i	d	b	f	a	i
	e	i	d	g	b	f	a	e	i	d	b	f	a	i	e	b
	i	d	g	b	f	e	i	d								f
																e
a	∞	∞	3	3	3	3	3	2	2	2	2	2	2	2	1	
b	∞	∞	∞	∞	∞	4	4	4	4	4	4	3	3	3	3	2
c	0															
d	∞	1	1	0	0	0	0	0	-1							
e	∞	∞	5	5	5	5	5	4	4	-1	-1	-1	-1	-1	-2	-2
f	∞	∞	∞	∞	∞	9	3	3	3	3	3	3	3	2	2	2
g	∞	1	1	1	1	0										1
h	∞	1														
i	∞	∞	2	2	2	2	2	1	1	1	1	1	1	1	0	

Label-Correcting Algorithm

FIGURE 8.10 An execution of `labelCorrectingAlgorithm()`, which applies a deque.

	active vertex									
deque	c	d	g	d	h	g	d	a	e	i
	d	g	d	h	g	d	a	e	i	b
	g	h	h	a	a	a	e	i	b	f
	h	a	a	e	e	e	i	b	f	
	e	e	e	i	i	i				
	i	i								
a	∞	∞	3	3	2	2	2	1		
b	∞	∞	∞	∞	∞	∞	∞	2		
c	0									
d	∞	1	1	0	0	0	-1			
e	∞	∞	5	5	4	4	4	3	3	3
f	∞	∞	∞	∞	∞	∞	∞	∞	7	1
g	∞	1	1	1	1	0				
h	∞	1								
i	∞	∞	2	2	1	1	1	0		

All Pairs Shortest Path

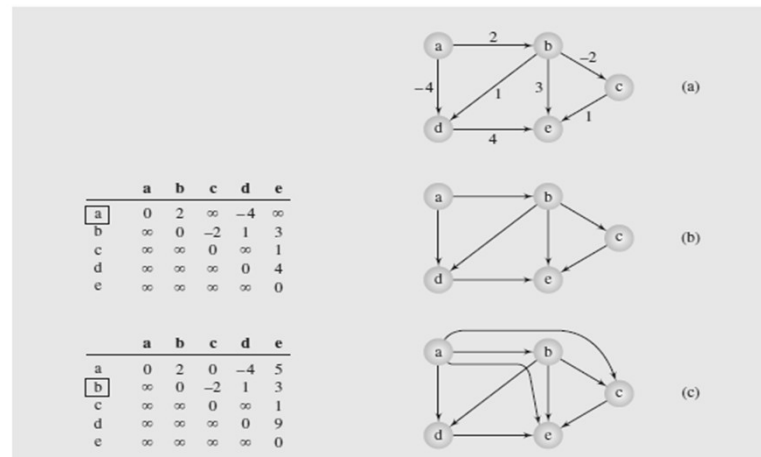
- All pairs shortest path seems more complicated than simple single source shortest path at first.
- W. Floyd algorithm is an elegant implementation of this problem.

All Pairs Shortest Path

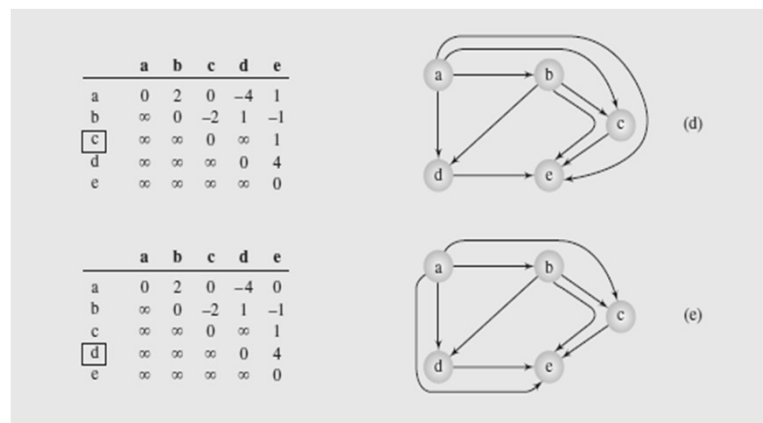
```
WFIalgorithm(matrix weight)
  for i = 1 to |V|
    for j = 1 to |V|
      for k = 1 to |V|
        if weight[j][k] > weight[j][i] + weight[i][k]
          weight[j][k] = weight[j][i] + weight[i][k];
```

All Pairs Shortest Path

FIGURE 8.11 An execution of `WFIalgorithm()`.



All Pairs Shortest Path



Cycle Detection

- As it can be seen with all previous algorithm that cycle is quite challenging in algorithm development for graph.
- A simple cycle detection algorithm is very much required.
 - Undirected Graph
 - Directed Graph

Cycle Detection

```

cycleDetectionDFS(v)
    num(v) = i++;
    for all vertices u adjacent to v
        if num(u) is 0
            pred(u) = v;
            cycleDetectionDFS(u);
        else if edge(vu) is not in edges
            pred(u) = v;
            cycle detected;

DFS(v)
    num(v) = i++;
    for all vertices u adjacent to v
        if num(u) is 0
            attach edge(uv) to edges;
            DFS(u);

depthFirstSearch()
    for all vertices v
        num(v) = 0;
    edges = null;
    i = 1;
    while there is a vertex v such that num(v) is 0
        DFS(v);
    output edges;

```

Di-graph Cycle Detection

```

digraphCycleDetectionDFS(v)
    num(v) = 1++;
    for all vertices u adjacent to v
        if num(u) is 0
            pred(u) = v;
            digraphCycleDetectionDFS(u);
        else if num(u) is not ∞
            pred(u) = v;
            cycle detected;
    num(v) = ∞;

DFS(v)
    num(v) = 1++;
    for all vertices u adjacent to v
        if num(u) is 0
            attach edge(uv) to edges;
            DFS(u);

depthFirstSearch()
    for all vertices v
        num(v) = 0;
    edges = null;
    i = 1;
    while there is a vertex v such that num(v) is 0
        DFS(v);
    output edges;

```