## 8.1 Introduction



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# 8.2 Introduction to Graphs



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#### Graphs

A Graph: 
$$G = (V, E)$$

- 1. Definition: Undirected Graphs; Directed Graphs
- 2. Terms vertices, edges, paths, degrees, connected graphs, cycles etc.
- 3. Facts such as:

Handshaking Lemma : 
$$\sum_{v \in V} \deg(v) = 2|E|$$

4. Representation: Adjacency Matrix; Adjacency List

## Illustrations: Graphs and its basic features I

Illustrations: Graphs and its basic features II

#### **Undirected Graphs**

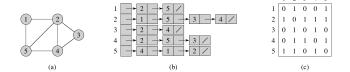


Figure: Representing an un-directed graph

#### **Directed Graphs**

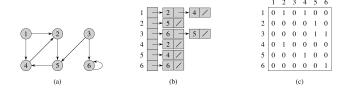


Figure: Representing a directed graph

#### Question

Draw the adjacency matrix for the *directed* graph  $H = (V_H, E_H)$  where

- 1. Vertex set of  $H = V_H = \{1, 2, 3, 4, 5\}$
- 2. The set of edges of *H*:

$$E_H = \{(1\ 2), (2\ 3), (1\ 3), (2\ 4), (5\ 1), (5\ 3)\}$$

Answer:

## Illustrations: Adjacency Lists and its features

## 8.3 Breadth First Search I



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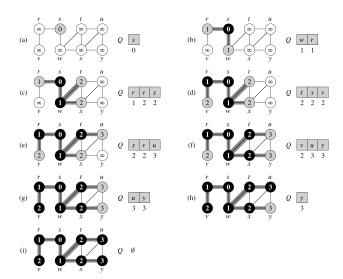
#### Graph Search: Basic Algorithms

- 1. Breadth First Search
- 2. Depth First Search

## Illustrations for Basic Graph Search Strategies

Breadth First Depth First

#### Breadth First Search Example



#### Illustrations: Breadth First Search

#### Question

#### Question Fill in the blanks:

When performing *breadth first search*, we use the \_\_\_\_\_ data structure to help selecting the next vertex to be examined.

#### The Breadth First Search Algorithm

```
BFS(V, E, s)
 for each u \in V - \{s\}
      u.d = \infty
 s.d = 0
 O = \emptyset
 ENQUEUE(Q, s)
 while Q \neq \emptyset
      u = \text{DEQUEUE}(Q)
      for each v \in G.Adj[u]
           if v.d == \infty
                v.d = u.d + 1
                ENQUEUE(O, v)
```

#### Question

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In the pseudocode of the breadth first search algorithm given:

## 8.4 Breadth First Search II



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#### Analyzing Breadth First Search: Background

- 1. For a graph G=(V,E), its 'size' has two parameters: |V| and |E|
- 2. If the graph is a tree, then |E| = |V| 1 = O(|V|)
- 3. If the graph is a complete graph, then

$$|E| = \frac{|V|(|V|-1)}{2} = O(|V|^2)$$

- 4. When  $|E| = \Theta(|V|^2)$ , we may call that graph a *dense* graph.
- 5. When |E| = O(|V|), we may call that graph a *sparse* graph.

#### The Breadth First Search Algorithm

```
BFS(V, E, s)
 for each u \in V - \{s\}
      u.d = \infty
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 ENQUEUE(Q, s)
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      for each v \in G.Adj[u]
           if v.d == \infty
                v.d = u.d + 1
                ENQUEUE(O, v)
```

#### Analyzing Breadth First Search I

**Discussions** What is the worst case running time for Breadth First Search?

## Analyzing Breadth First Search II

#### Question

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Given a graph G with n vertices, the worst case running time for running breadth first search over G is (in Big-O notation):

Case 1: The graph is sparse

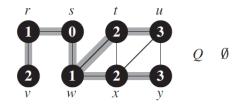
Case 2: The graph is *dense* 

#### Finding Shortest Path via BFS

- 1. When performing BFS, we are building a BFS tree.
- 2. We can remember the predecessor  $\pi$  of each node in the tree.
- 3. We can recover the shortest path from the predecessor relation  $\pi$  after running BFS.

#### Question

#### Fill in the blanks:



vertex	r	s	t	и	V	W	X	у
predecessor								

## Illustrations: Finding Shortest Path

## Finding Shortest Path: Code

```
PRINT-PATH(G, s, v)

1 if v == s

2 print s

3 elseif v.\pi == NIL

4 print "no path from" s "to" v "exists"

5 else PRINT-PATH(G, s, v.\pi)

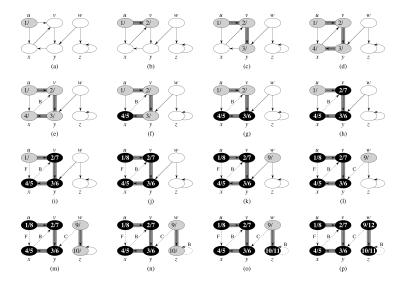
6 print v
```

# 8.5 Depth First Search I



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#### Depth First Search: Example



#### Question

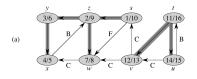
**Question:** Which of the following best describes the key features in the previous sequence of diagrams:

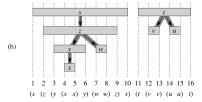
- 1. Each node is colored either in white, grey or black
- 2. Each node is eventually labelled by 2 numbers
- 3. Each edge is either without labelled or eventually labelled by one of these symbols: *C*, *B* or *F*.
- 4. All of the above

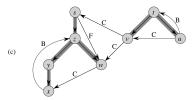
## Illustrations: Depth First Search I

#### Illustrations: Depth First Search II

#### Properties of Depth First Search







## Pseudocode for DFS (Part I)

```
\begin{aligned} \mathsf{DFS}(G) \\ & \textbf{for} \ \mathsf{each} \ u \in G.V \\ & u.color = \mathsf{WHITE} \\ & \mathit{time} = 0 \\ & \textbf{for} \ \mathsf{each} \ u \in G.V \\ & \textbf{if} \ u.color == \mathsf{WHITE} \\ & \mathsf{DFS-VISIT}(G,u) \end{aligned}
```

## Pseudocode for DFS (Part II)

```
DFS-VISIT(G, u)
 time = time + 1
 u.d = time
 u.color = GRAY
                                // discover u
 for each v \in G.Adj[u]
                                /\!\!/ explore (u, v)
      if v.color == WHITE
          DFS-VISIT(\nu)
 u.color = BLACK
 time = time + 1
 u.f = time
                                // finish u
```

# 8.6 Depth First Search II



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#### Pseudocode for DFS

We will re-visit the pseudocode for DFS, which consists of

- 1. DFS-VISIT
- 2. DFS

Pseudocode for DFS	Pseudocode for DFS-VISIT	
$\begin{aligned} DFS(G) \\ & \textbf{for } each \ u \in G.V \\ & u.color = WHITE \\ & \mathit{time} = 0 \\ & \textbf{for } each \ u \in G.V \\ & \mathbf{if} \ u.color == WHITE \\ & DFS-VISIT(G,u) \end{aligned}$	$\begin{aligned} & \text{DFS-VISIT}(G, u) \\ & \textit{time} = \textit{time} + 1 \\ & \textit{u.d} = \textit{time} \\ & \textit{u.color} = \text{GRAY} \\ & \textbf{for each } v \in \textit{G.Adj[u]} \\ & \textbf{if } v.color = \text{WHITE} \\ & \textit{DFS-VISIT}(v) \\ & \textit{u.color} = \text{BLACK} \\ & \textit{time} = \textit{time} + 1 \\ & \textit{u.f} = \textit{time} \end{aligned}$	// discover u // explore (u, v) // finish u

## Worst Case Analysis

**Discussions** It is known that the worst case time complexity for DFS is  $\Theta(|V| + |E|)$ . Why ?

### Question

#### Question Fill in the blanks:

Given a graph G with n vertices, the worst case running time for running depth first search over G is (in Big-O notation):

Case 1: The graph is *sparse* 

Case 2: The graph is dense

## Using DFS I

#### Types of edges:

- 1. Directed graphs: tree, forward, back, cross edges
- 2. Undirected graphs: tree and back edges

#### Question

Fill in the blanks:

After running DFS over a directed graph G, if there is at least one back edge found, then G has a \_\_\_\_\_\_.

#### **Discussions**

**Discussions** If a directed graph G has a cycle, will running DFS over G always produce at least one back edge?

## 8.7 Dijkstra's Algorithm



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## Shortest Path Algorithms I: Introduction

#### Single Source Shortest Path Problem

- G = (V, E); source vertex s
- ▶ Determine a shortest path from s to any vertex v,  $v \in V$ .

#### **Noteworthy Properties**

- Subpaths of shortest paths are shortest paths (Lemma 24.1)
- ▶ If *G* contains no negative weight cycles reachable from the source *s*, then the shortest-path weight remains well defined (can be negative).
- A shortest path can be made cycle free.

## Shortest Path Algorithms II: Variants

#### Variants may include

1. Traversing an edge in E may include a cost.

Directed Graph G = (V, E) & weight function  $w : E \to \mathbb{R}$ 

- 2. The cost of some edges may be negative
- 3. The graphs we consider has specific properties (e.g. acyclic, disconnected etc.)

## Shortest Path Algorithms II: Variants

#### Similar formulations (See CLRS page 644):

- Single-destination shortest-paths problem
   Reverse the direction of each edge may help
- Single-pair shortest-path problem
   Set the source s to be the beginning vertex
- All-pairs shortest-paths problem Run a single source shortest path algorithm multiple times Can we have a faster method?

# Shortest Path Algorithms III: Design Paradigms and new techniques

- Dynamic Programming
- Greedy Choice
- Relaxation Method
- Use properties of shortest paths and relaxation

## Relaxation Method: The Two Steps

INIT-SINGLE-SOURCE $(G, s)$ for each $v \in G$ . $V$ $v.d = \infty$ $v.\pi = \text{NIL}$	RELAX $(u, v, w)$ if $v.d > u.d + w(u, v)$ v.d = u.d + w(u, v)
s.d = 0	$v.\pi = u$

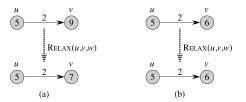


Figure: Relaxing an edge

#### Relaxation Method: Discussions

#### **Discussions**

Can we view relaxation method as a basic step in developing short path algorithms ?

## Breadth First Search (BFS) Revisited

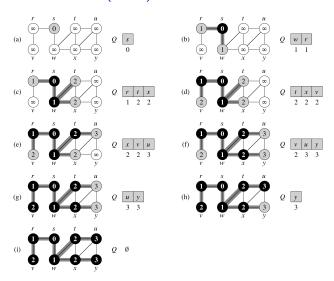


Figure: An illustration for running BFS

## Dijkstra's Algorithm

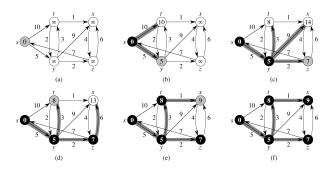


Figure: Running Dijkstra's Algorithm: an example

**Questions** Can you spot an relaxation step? Which data structure may help to locate the next vertex to be explored?

## Dijkstra's Algorithm: Pseudocode

```
DIJKSTRA(G, w, s)

INIT-SINGLE-SOURCE(G, s)
S = \emptyset
Q = G.V  // i.e., insert all vertices into Q
while Q \neq \emptyset
u = \text{EXTRACT-MIN}(Q)
S = S \cup \{u\}
for each vertex v \in G.Adj[u]
\text{RELAX}(u, v, w)
```

Figure : Pseudocode for Dijkstra's Algorithm

#### Discussions

Does Dijkstra's algorithm apply the greedy strategy?



## Dijkstra's Algorithm: Correctness

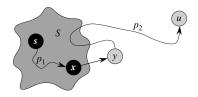


Figure: Why Dijkstra's Algorithm works as expected

#### **Discussions**

Use this diagram to explain why Dijkstra's algorithm work.

## Dijkstra's Algorithm: Complexity

- 1. Choice of Data structure matters
- 2.  $O(V^2)$  if an array is used to maintain the priority queue
- 3. O((|V| + |E|)Ig|V|) if a binary heap is used to maintain the priority queue
- 4. Can you more advanced data structure (Fibonacci Heap) to improve the run time (use amortized analysis)



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