## **Graph Algorithms**

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## **The Königsberg Bridges Problem**



Figure 7.1 The Königsberg bridges problem.

#### Source: Manber 1989

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Can one start from one of the lands, cross every bridge exactly once, and return to the origin?

# The Königsberg Bridges Problem (cont.)



Figure 7.2 The graph corresponding to the Königsberg bridges problem.



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# Graphs

A graph consists of a set of vertices (or nodes) and a set of edges (or links, each normally connecting two vertices) and is commonly denoted as G(V, E), where

- $\bigcirc$  G is the name of the graph,
- V is the set of vertices, and
- $\bigcirc$  E is the set of edges.



# **Modeling with Graphs**

- Reachability
- Shortest Routes
- Scheduling



# Graphs (cont.)

- Undirected vs. Directed Graphs
- Paths, Simple Paths, Trails
- 📀 Circuits, Cycles
- Degrees, In-Degrees, Out-Degrees
- Connected Graphs, Trees
- Subgraphs, Induced Subgraphs, Spanning Trees



**The Problem** Given an undirected connected graph G = (V, E) such that all the vertices have even degrees, find a circuit *P* such that each edge of *E* appears in *P* exactly once.

The circuit *P* in the problem statement is called an *Eulerian circuit*.

#### Theorem

An undirected connected graph has an Eulerian circuit if and only if all of its vertices have even degrees.



## **Depth First Search**



Figure 7.4 A DFS for an undirected graph.



Source: Manber 1989

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## **Depth First Search (cont.)**

# **Algorithm Depth\_First\_Search**(G, v); **begin**

mark v; perform preWORK on v; for all edges (v, w) do if w is unmarked then  $Depth\_First\_Search(G, w)$ ; perform postWORK for (v, w)end



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## **Depth First Search (cont.)**

# Algorithm Refined\_DFS(G, v); begin

mark v; perform preWORK on v; for all edges (v, w) do if w is unmarked then  $Refined\_DFS(G, w)$ ; perform postWORK for (v, w); perform postWORK\_II on vend



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## **Connected Components**

# Algorithm Connected\_Components(G); begin

Component\_Number := 1;
while there is an unmarked vertex v do
 Depth\_First\_Search(G, v)
 (preWORK:
 v.Component := Component\_Number);
 Component\_Number := Component\_Number + 1

end



## **DFS Numbers**

# Algorithm DFS\_Numbering(G, v); begin

 $DFS\_Number := 1;$   $Depth\_First\_Search(G, v)$ (preWORK:  $v.DFS := DFS\_Number;$  $DFS\_Number := DFS\_Number + 1)$ 

end



### **The DFS Tree**

# Algorithm Build\_DFS\_Tree(G, v); begin

 $Depth\_First\_Search(G, v)$ (postWORK: if w was unmarked then add the edge (v, w) to T);

end



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## The DFS Tree (cont.)



Figure 7.9 A DFS tree for a directed graph.



Source: Manber 1989

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#### Lemma 7.2

For an undirected graph G = (V, E), every edge  $e \in E$ either belongs to the *DFS* tree *T*, or connects two vertices of *G*, one of which is the ancestor of the other in *T*.

For undirected graphs, DFS avoids cross edges.

#### Lemma 7.3

For a directed graph G = (V, E), if (v, w) is an edge in *E* such that  $v.DFS\_Number < w.DFS\_Number$ , then *w* is a descendant of *v* in the *DFS* tree *T*.

For directed graphs, cross edges must go "from right to left".



### **Directed Cycles**

**The Problem** Given a directed graph G = (V, E), determine whether it contains a (directed) cycle.

#### Lemma 7.4

G contains a directed cycle if and only if G contains a back edge (relative to the DFS tree).



### **Directed Cycles (cont.)**

# **Algorithm Find\_a\_Cycle**(*G*); **begin**

 $Depth\_First\_Search(G, v)$  /\* arbitrary v \*/ (preWORK:

 $v.on\_the\_path := true;$ 

postWORK:

if w.on\_the\_path then
 Find\_a\_Cycle := true;
 halt:

if w is the last vertex on v's list then
 v.on\_the\_path := false;)

end



### **Directed Cycles (cont.)**

# **Algorithm Refined\_Find\_a\_Cycle**(*G*); **begin**

 $Refined\_DFS(G, v)$  /\* arbitrary v \*/ (preWORK:

 $v.on\_the\_path := true;$ 

postWORK:

end



#### **Breadth-First Search**



Figure 7.12 A BFS tree for a directed graph.



Source: Manber 1989

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### **Breadth-First Search (cont.)**

# **Algorithm Breadth\_First\_Search**(G, v); **begin**

mark v; put v in a queue; while the queue is not empty do remove vertex w from the queue; perform preWORK on w; for all edges (w, x) with x unmarked do mark x; add (w, x) to the BFS tree T; put x in the queue

#### end



#### Lemma 7.5

If an edge (u, w) belongs to a *BFS* tree such that u is a parent of w, then u has the minimal *BFS* number among vertices with edges leading to w.

#### Lemma 7.6

For each vertex w, the path from the root to w in T is a shortest path from the root to w in G.

#### Lemma 7.7

If an edge (v, w) in *E* does not belong to *T* and *w* is on a larger level, then the level numbers of *w* and *v* differ by at most 1.



## **Breadth-First Search (cont.)**

# **Algorithm Simple\_BFS**(G, v); **begin**

put v in a queue; while the queue is not empty do remove vertex w from the queue; if w is unmarked then mark w; perform preWORK on w; for all edges (w, x) with x unmarked do put x in the queue

end



## **Breadth-First Search (cont.)**

# Algorithm Simple\_Nonrecursive\_DFS(G, v); begin

push v to Stack; while Stack is not empty do pop vertex w from Stack; if w is unmarked then mark w; perform preWORK on w; for all edges (w, x) with x unmarked do push x to Stack

end



# **Topological Sorting**

**The Problem** Given a directed acyclic graph G = (V, E) with *n* vertices, label the vertices from 1 to *n* such that, if *v* is labeled *k*, then all vertices that can be reached from *v* by a directed path are labeled with labels > k.

#### Lemma 7.8

A directed acyclic graph always contains a vertex with indegree 0.



# **Topological Sorting (cont.)**

**until** Queue is empty

#### Algorithm Topological\_Sorting(G);

```
initialize v.indegree for all vertices; /* by DFS */
G label := 0;
for i := 1 to n do
    if v_i indegree = 0 then put v_i in Queue;
repeat
    remove vertex v from Queue;
    G \ label := G \ label + 1;
    v.label := G\_label;
    for all edges (v, w) do
        w.indegree := w.indegree - 1;
        if w.indegree = 0 then put w in Queue
```

## **Single-Source Shortest Paths**

**The Problem** Given a directed graph G = (V, E) and a vertex v, find shortest paths from v to all other vertices of G.



## **Shorted Paths: The Acyclic Case**

Algorithm Acyclic\_Shortest\_Paths(G, v, n); {After performing a topological sort on  $G, \ldots$ } begin

let z be the vertex labeled n; if  $z \neq v$  then  $Acyclic\_Shortest\_Paths(G - z, v, n - 1);$ for all w such that  $(w, z) \in E$  do if w.SP + length(w, z) < z.SP then z.SP := w.SP + length(w, z)else v.SP := 0end



## The Acyclic Case (cont.)

#### Algorithm Imp\_Acyclic\_Shortest\_Paths(G, v);

for all vertices w do  $w.SP := \infty$ ;

initialize *v.indegree* for all vertices;

for i := 1 to n do

if  $v_i.indegree = 0$  then put  $v_i$  in Queue;

*v.SP* := 0;

#### repeat

remove vertex w from Queue; for all edges (w, z) do if w.SP + length(w, z) < z.SP then z.SP := w.SP + length(w, z); z.indegree := z.indegree - 1; if z.indegree = 0 then put z in Queueuntil Queue is empty



### **Shortest Paths: The General Case**

# Algorithm Single\_Source\_Shortest\_Paths(G, v); begin

for all vertices w do w.mark := false; $w.SP := \infty;$ v.SP := 0;while there exists an unmarked vertex do let w be an unmarked vertex s.t. w.SP is minimal; w.mark := true;for all edges (w, z) such that z is unmarked do if w.SP + length(w, z) < z.SP then z.SP := w.SP + length(w, z)

end



### The General Case (cont.)



Figure 7.18 An example of the single-source shortest-paths algorithm.



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# **Minimum-Weight Spanning Trees**

**The Problem** Given an undirected connected weighted graph G = (V, E), find a spanning tree *T* of *G* of minimum weight.

#### Theorem

Let  $V_1$  and  $V_2$  be a partition of V and  $E(V_1, V_2)$  be the set of edges connecting nodes in  $V_1$  to nodes in  $V_2$ . The edge with the minimum weight in  $E(V_1, V_2)$  must be in the minimum-cost spanning tree of G.







If cost(u, v) is the smallest among  $E(V_1, V_2)$ , then  $\{u, v\}$ must be in the minimum spanning tree.

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Figure 7.19 Finding the next edge of the MCST.



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# **Algorithm MST**(G); **begin**

initially *T* is the empty set; for all vertices *w* do

 $w.mark := false; w.cost := \infty;$ let (x, y) be a minimum cost edge in G;x.mark := true;for all edges (x, z) do z.edge := (x, z); z.cost := cost(x, z);



while there exists an unmarked vertex do let w be an unmarked vertex with minimal w.cost; if  $w.cost = \infty$  then print "G is not connected"; halt else w.mark := true;add w.edge to T; for all edges (w, z) do if not *z.mark* then if cost(w, z) < z.cost then

z.edge := (w, z); z.cost := cost(w, z)

end



Algorithm Another\_MST(G); begin initially T is the empty set; for all vertices w do  $w.mark := false; w.cost := \infty;$  x.mark := true; /\* x is an arbitrary vertex \*/ for all edges (x, z) do z.edge := (x, z); z.cost := cost(x, z);



while there exists an unmarked vertex do let w be an unmarked vertex with minimal w.cost; if  $w.cost = \infty$  then print "G is not connected"; halt else w.mark := true;add w edge to T:

w.mark := true;add w.edge to T; for all edges (w, z) do if not z.mark then if cost(w, z) < z.cost then z.edge := (w, z);z.cost := cost(w, z)

#### end





	v	а	b	С	d	e	f	g	h
v	-	v(1)	v(6)	~	v(9)	~	~~~~	~~~~	∞
a	-		v(6)	a(2)	v(9)	~	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	~~~~	∞
С	-	-	v(6)	( <b>1</b> 5)	c(4)	~~~	c(10)	00	∞
d	-		v(6)	131	1.0	d(7)	c(10)	<i>d</i> (12)	~
b	-	-	-	547F	-	b(3)	c(10)	<i>d</i> (12)	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~
е	-	-	-			-	c(10)	d(12)	e(5)
h	-	141	-	-	5 <b>4</b> 0	:=):	<i>c</i> (10)	h(11)	
f	-	-						h(11)	
g	-	121	141	-		-	-	-	-

Figure 7.21 An example of the minimum-cost spanning-tree algorithm.



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**The Problem** Given a weighted graph G = (V, E) (directed or undirected) with nonnegative weights, find the minimum-length paths between all pairs of vertices.

# **Algorithm All\_Pairs\_Shortest\_Paths**(W); **begin**

{initialization omitted} for m := 1 to n do {the induction sequence} for x := 1 to n do for y := 1 to n do if W[x,m] + W[m,y] < W[x,y] then W[x,y] := W[x,m] + W[m,y]

end



**The Problem** Given a directed graph G = (V, E), find its transitive closure.

# **Algorithm Transitive\_Closure**(*A*); **begin**

{initialization omitted} for m := 1 to n do for x := 1 to n do for y := 1 to n do if A[x,m] and A[m,y] then A[x,y] := true

end



## **Transitive Closure (cont.)**

# **Algorithm Improved\_Transitive\_Closure**(A); **begin**

{initialization omitted} for m := 1 to n do for x := 1 to n do if A[x, m] then for y := 1 to n do if A[m, y] then A[x, y] := true

end



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## **Biconnected Components**

- An undirected graph is *biconnected* if there are at least two vertex-disjoint paths from every vertex to every other vertex.
- A graph is *not* biconnected if and only if there is a vertex whose removal disconnects the graph. Such a vertex is called an *articulation point*.
- A biconnected component is a maximal subset of the edges such that its induced subgraph is biconnected (namely, there is no other subset that contains it and induces a biconnected graph).





Figure 7.25 The structure of a nonbiconnected graph.



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#### Lemma 7.9

Two distinct edges e and f belong to the same biconnected component if and only if there is a cycle containing both of them.

#### Lemma 7.10 Each edge belongs to exactly one biconnected component.





**Figure 7.26** An edge that connects two different biconnected components. (a) The components corresponding to the graph of Fig. 7.25 with the articulation points indicated. (b) The biconnected component tree.



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Figure 7.27 Computing the High values.



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# Algorithm Biconnected\_Components(G, v, n); begin

for every vertex w do  $w.DFS_Number := 0$ ;  $DFS_N := n$ ; BC(v)end

#### procedure BC(v); begin

 $v.DFS\_Number := DFS\_N;$   $DFS\_N := DFS\_N-1;$ insert v into Stack; $v.high := v.DFS\_Number;$ 



for all edges (v, w) do insert (v, w) into *Stack*; if w is not the parent of v then if w.DFS Number = 0 then BC(w);if  $w.high \leq v.DFS\_Number$  then remove all edges and vertices from Stack until v is reached; insert v back into Stack;  $v.high := \max(v.high, w.high)$ else  $v.high := \max(v.high, w.DFS\_Number)$ 

end



# procedure BC(v); begin

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 $v.DFS\_Number := DFS\_N;$  $DFS \ N := DFS \ N - 1;$  $v.high := v.DFS\_Number;$ for all edges (v, w) do if w is not the parent of v then insert (v, w) into *Stack*; if w.DFS Number = 0 then BC(w);if  $w.high \leq v.DFS\_Number$  then remove all edges from *Stack* until (v, w) is reached;  $v.high := \max(v.high, w.high)$ else  $v.high := \max(v.high, w.DFS\_Number)$ 





Figure 7.29 An example of computing High values and biconnected components.



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## **Even-Length Cycles**

**The Problem** Given a connected undirected graph G = (V, E), determine whether it contains a cycle of even length.

#### Theorem

Every biconnected graph that has more than one edge and is not merely an odd-length cycle contains an even-length cycle.



## **Even-Length Cycles (cont.)**



Figure 7.35 Finding an even-length cycle.



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## **Strongly Connected Components**

- A directed graph is strongly connected if there is a directed path from every vertex to every other vertex.
- A strongly connected component is a maximal subset of the vertices such that its induced subgraph is strongly connected (namely, there is no other subset that contains it and induces a strongly connected graph).



#### Lemma 7.11

Two distinct vertices belong to the same strongly connected component if and only if there is a circuit containing both of them.

#### Lemma 7.12

Each vertex belongs to exactly one strongly connected component.





Figure 7.30 A directed graph and its strongly connected component graph.



#### Source: Manber 1989

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Figure 7.31 Adding an edge connecting two different strongly connected components.



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Figure 7.32 The effect of cross edges.



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# Algorithm Strongly\_Connected\_Components(G, n); begin

for every vertex v of G do
 v.DFS\_Number := 0;
 v.component := 0;
Current\_Component := 0; DFS\_N := n;
while v.DFS\_Number = 0 for some v do
 SCC(v)

end

#### procedure SCC(v); begin

 $v.DFS\_Number := DFS\_N;$   $DFS\_N := DFS\_N-1;$ insert v into Stack; $v.high := v.DFS\_Number;$ 



for all edges (v, w) do if  $w.DFS\_Number = 0$  then SCC(w); $v.high := \max(v.high, w.high)$ else if  $w.DFS\_Number > v.DFS\_Number$ and w.component = 0 then  $v.high := \max(v.high, w.DFS\_Number)$ if  $v.high = v.DFS\_Number$  then  $Current\_Component := Current\_Component + 1;$ repeat remove x from the top of Stack;  $x.component := Current\_Component$ until x = v

end





	а 11	b 10	с 9	d 8	с 7	f 6	8 5	h 4	i 3	j 2	k I
a	11	24	19			э.	$\ast$		$\mathbf{x}$		2
ь	11	10	3	4		•	÷	8	$\mathcal{A}_{i}^{(2)}$	+	
c	11	10	9	121		4	*		-		ŝ
d	11	10	9	8				-		*	1
e	11	10	9	8	10	1		70	32	2	2
d	11	10	9	10	10			65	5	$\approx$	đ
c	11	10	10	10	10		$\hat{\mathbf{r}}$	83	$\mathbf{x}_{i}^{i}$	$\otimes$	÷
ſ	11	10	10	10	10	6		÷.	20		4
g	11	10	10	10	10	6	7		2		ŝ
r	11	10	10	10	10	7	7		-		÷
с	11	10	10	10	10	7	7		\$		1
Э	H	10	10	10	10	7	7	÷.:	$\mathcal{A}$	•	1
a	11	10	10	10	10	7	7	$\frac{1}{2}$	83	*	8
h	11	10	10	10	10	7	7	4	÷.	æ	ž
4	11	10	10	10	10	7	7	4	3		1
j	11	10	10	10	10	7	7	4	3	11	1
1	11	10	10	10	10	7	7	4	11	н	5
k	11	10	10	10	10	7	7	-4	11	11	1
i	11	10	10	10	10	7	7	4	11	11	1
h	11	10	10	10	10	7	7	11	11	п	1
(a)	11	10	10	10	10	7	7	11	11	11	1

Figure 7.34 An example of computing High values and strongly connected components.



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**The Problem** Given a directed graph G = (V, E), determine whether it contains a (directed) cycle of odd length.



### **Network Flows**

- Consider a directed graph, or network, G = (V, E) with two distinguished vertices: s (the source) with indegree 0 and t (the sink) with outdegree 0.
- Each edge e in E has an associated positive weight c(e), called the capacity of e.



## **Network Flows (cont.)**

A flow is a function f on E that satisfies the following two conditions:

**1.** 
$$0 \le f(e) \le c(e)$$
.

**2.** 
$$\sum_{u} f(u, v) = \sum_{w} f(v, w)$$
, for all  $v \in V - \{s, t\}$ .

The network flow problem is to maximize the flow f for a given network G.



### **Network Flows (cont.)**



Figure 7.39 Reducing bipartite matching to network flow (the directions of all the edges are from left to right).

Source: Manber 1989



# **Augmenting Paths**

- An augmenting path w.r.t. a given flow f (of a network G) is a directed path from s to t consisting of edges from G, but not necessarily in the same direction; each of these edges (v, u) satisfies exactly one of:
  - 1. (v, u) is in the same direction as it is in *G*, and f(v, u) < c(v, u). (forward edge)
  - 2. (v, u) is in the opposite direction in G (namely,  $(u, v) \in E$ ), and f(u, v) > 0. (backward edge)
- If there exists an augmenting path w.r.t. a flow f(f) admits an augmenting path), then f is not maximum.



# Augmenting Paths (cont.)



Figure 7.40 An example of a network with a (nonmaximum) flow.



#### Source: Manber 1989

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## Augmenting Paths (cont.)



Figure 7.41 The result of augmenting the flow of Fig. 7.40.



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### **Properties of Network Flows**

**The Augmenting-Path Theorem** A flow *f* is maximum if and only if it admits no augmenting path.

A *cut* is a set of edges that separate *s* from *t*, or more precisely a set of the form  $\{(v, w) \in E \mid v \in A \text{ and } w \in B\}$ , where B = V - A such that  $s \in A$  and  $t \in B$ .

**Max-Flow Min-Cut Theorem** The value of a maximum flow in a network is equal to the minimum capacity of a cut.



# **Properties of Network Flows (cont.)**

**The Integral-Flow Theorem** If the capacities of all edges in the network are integers, then there is a maximum flow whose value is an integer.



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## **Residual Graphs**

- The residual graph with respect to a network
  G = (V, E) and a flow f is the network R = (V, F), where
  F consists of all forward and backward edges and their capacities are given as follows:
  - 1.  $c_R(v, w) = c(v, w) f(v, w)$  if (v, w) is a forward edge and
  - 2.  $c_R(v, w) = f(w, v)$  if (v, w) is a backward edge.
- An augmenting path is thus a regular directed path from s to t in the residual graph.



## **Residual Graphs (cont.)**



Figure 7.42 A bad example of network flow.



Source: Manber 1989

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