# Basic Graph Algorithms (Based on [Manber 1989]) 

Yih-Kuen Tsay<br>Department of Information Management<br>National Taiwan University

## The Königsberg Bridges Problem



Figure: The Königsberg bridges problem.
Source: redrawn from [Manber 1989, Figure 7.1].
Can one start from one of the lands, cross every bridge exactly once, and return to the origin?

## The Königsberg Bridges Problem (cont.)

An abstract model is more convenient to work with:


Figure: The graph corresponding to the Königsberg bridges problem. Source: redrawn from [Manber 1989, Figure 7.2].

## Graphs

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A graph consists of a set of vertices (or nodes) and a set of edges (or links, each normally connecting two vertices).

- A graph is commonly denoted as $G(V, E)$, where
* $G$ is the name of the graph,

在 $V$ is the set of vertices, and
淃 $E$ is the set of edges.
Note: we assume that you have learned from a course on Data Structures the basics of graph theory and the representation of a graph by an adjacency matrix or incidence list.

## Graphs (cont.)

- Undirected vs. Directed Graph
- Simple Graph vs. Multigraph
- Path, Simple Path, Trail
- Cycle, Simple Cycle, Circuit

Degree, In-Degree, Out-Degree

- Connected Graph, Connected Components
- Tree, Forest

Subgraph, Induced Subgraph
Spanning Tree, Spanning Forest

- Weighted Graph


## Modeling with Graphs

- Reachability

Finding program errors
Solving sliding tile puzzles

- Shortest Paths
* Finding the fastest route to a place

Routing messages in networks
Graph Coloring
Coloring maps
Scheduling classes

## Eulerian Graphs

## 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: A DFS for an undirected graph.
Source: redrawn from [Manber 1989, Figure 7.4].

## Depth-First Search (cont.)

Algorithm Depth_First_Search( $G, v$ ); begin<br>mark $v$;<br>perform preWORK on $v$;<br>for all edges $(v, w)$ do<br>if $w$ is unmarked then<br>Depth_First_Search( $G, w)$;<br>perform postWORK for $(v, w)$<br>end

## 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 $v$
end

## A "Metaphor" of DFS

Space: the final frontier. These are the voyages of the starship Enterprise. Its five-year mission: to explore strange new worlds. To seek out new life and new civilizations. To boldly go where no man/one has gone before!

- Captain James T. Kirk, Star Trek


## 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

## 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

Time complexity:

## 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

Time complexity: $O(|E|+|V|)$.

## 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

## 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

Time complexity: $O(|E|)$ (assuming the input graph is connected).

## The DFS Tree

## Algorithm Build_DFS_Tree( $G, v$ ); begin <br> Depth_First_Search(G, v) <br> (postWORK:

if $w$ was unmarked then add the edge ( $v, w$ ) to $T$ ); end

## The DFS Tree (cont.)



Figure: A DFS tree for a directed graph.
Source: redrawn from [Manber 1989, Figure 7.9].

## The DFS Tree (cont.)

## 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 $v . D F S \_$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

## 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);

 beginRefined_DFS (G,v) /* arbitrary v */
(preWORK:
v.on_the_path := true; postWORK:
if w.on_the_path then Refined_Find_a_Cycle := true; halt; postWORK_II:
v.on_the_path := false)
end

## Breadth-First Search



Figure: A BFS tree for a directed graph.
Source: redrawn from [Manber 1989, Figure 7.12].

## 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

## Breadth-First Search (cont.)

## 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 Queue;
while Queue is not empty do
remove vertex w from Queue;
if $w$ is unmarked then
mark $w$;
perform preWORK on $w$; for all edges ( $w, x$ ) with $x$ unmarked do put $x$ in 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

## 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.)

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
until Queue is empty

## Single-Source Shortest Paths

## 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 $(G, v, n)$; $\{$ Initially, $w \cdot S P=\infty$, for every node $w$.\}
$\{$ A topological sort has been performed 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 $w . S P+$ length $(w, z)<z . S P$ then
$z . S P:=w . S P+l$ length $(w, z)$
else $v . S P:=0$
end

## The Acyclic Case (cont.)

Algorithm Imp_Acyclic_Shortest_Paths( $G, v$ );
for all vertices $w$ do $w . S P:=\infty$;
initialize $v$.indegree for all vertices;
for $i:=1$ to $n$ do
if $v_{i}$.indegree $=0$ then put $v_{i}$ in Queue;
$v . S P:=0$;
repeat
remove vertex w from Queue;
for all edges $(w, z)$ do
if $w . S P+$ length $(w, z)<z . S P$ then $z . S P:=w . S P+\operatorname{length}(w, z)$;
z.indegree $:=$ z.indegree -1 ;
if z.indegree $=0$ then put $z$ in Queue
until Queue is empty

## Shortest Paths: The General Case

Algorithm Single_Source_Shortest_Paths( $G, v$ );
// Dijkstra's algorithm
begin
for all vertices $w$ do
w.mark := false;
$w . S P:=\infty$;
$v . S P:=0$;
while there exists an unmarked vertex do
let $w$ be an unmarked vertex s.t. $w . S P$ is minimal;
w.mark := true;
for all edges $(w, z)$ such that $z$ is unmarked do

$$
\begin{gathered}
\text { if } w \cdot S P+\text { length }(w, z)<z . S P \text { then } \\
z . S P:=w \cdot S P+\text { length }(w, z)
\end{gathered}
$$

end

## Shortest Paths: The General Case

Algorithm Single_Source_Shortest_Paths( $G, v$ );
// Dijkstra's algorithm
begin
for all vertices $w$ do
w.mark := false;
$w . S P:=\infty$;
$v . S P:=0$;
while there exists an unmarked vertex do
let $w$ be an unmarked vertex s.t. $w . S P$ is minimal;
w.mark := true;
for all edges $(w, z)$ such that $z$ is unmarked do

$$
\begin{gathered}
\text { if } w . S P+\text { length }(w, z)<z . S P \text { then } \\
z . S P:=w \cdot S P+\text { length }(w, z)
\end{gathered}
$$

end
Time complexity:

## Shortest Paths: The General Case

Algorithm Single_Source_Shortest_Paths( $G, v$ );
// Dijkstra's algorithm
begin
for all vertices $w$ do
w.mark := false;
$w . S P:=\infty$;
$v . S P:=0$;
while there exists an unmarked vertex do
let $w$ be an unmarked vertex s.t. $w . S P$ is minimal;
w.mark := true;
for all edges $(w, z)$ such that $z$ is unmarked do

$$
\begin{gathered}
\text { if } w . S P+\text { length }(w, z)<z . S P \text { then } \\
z . S P:=w \cdot S P+\text { length }(w, z)
\end{gathered}
$$

end
Time complexity: $O((|E|+|V|) \log |V|)$ (using a min heap).

## The General Case (cont.)



Figure: An example of the single-source shortest-paths algorithm. Source: redrawn from [Manber 1989, Figure 7.18].

## Minimum-Weight Spanning Trees

## 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 $E\left(V_{1}, V_{2}\right)$ be the set of edges connecting nodes in $V_{1}$ to nodes in $V_{2}$. The edge with the minimum weight in $E\left(V_{1}, V_{2}\right)$ must be in the minimum-cost spanning tree of $G$.

## Minimum-Weight Spanning Trees (cont.)



If $\operatorname{cost}(u, v)$ is the smallest among $E\left(V_{1}, V_{2}\right)$, then $\{u, v\}$ must be in the minimum spanning tree.

## Minimum-Weight Spanning Trees (cont.)



Figure: Finding the next edge of the MCST. Source: redrawn from [Manber 1989, Figure 7.19].

## Minimum-Weight Spanning Trees (cont.)

Algorithm MST(G);
// A variant of Prim's algorithm
begin
initially $T$ is the empty set;
for all vertices $w$ do

$$
\text { w.mark }:=\text { false; w.cost }:=\infty \text {; }
$$

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

$$
\text { z.edge }:=(x, z) ; \quad \text { z.cost }:=\cos t(x, z) ;
$$

## Minimum-Weight Spanning Trees (cont.)

while there exists an unmarked vertex do
let $w$ be an unmarked vertex with minimal $w$.cost;
if $w \cdot \operatorname{cost}=\infty$ 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 $\operatorname{cost}(w, z)<z$.cost then

$$
z . e d g e ~:=(w, z) ; \quad \text { z.cost }:=\operatorname{cost}(w, z)
$$

end

## Minimum-Weight Spanning Trees (cont.)

Algorithm Another_MST(G);
// Prim's algorithm
begin
initially $T$ is the empty set;
for all vertices $w$ do

$$
\text { w.mark }:=\text { false; w.cost }:=\infty ;
$$

$x . m a r k:=$ true; $/^{*} x$ is an arbitrary vertex */
for all edges $(x, z)$ do
z.edge $:=(x, z) ; \quad z . \cos t:=\operatorname{cost}(x, z) ;$

## Minimum-Weight Spanning Trees (cont.)

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 $\operatorname{cost}(w, z)<z$.cost then
z.edge $:=(w, z)$;
$z . \cos t:=\operatorname{cost}(w, z)$
end

## Minimum-Weight Spanning Trees (cont.)

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 $\operatorname{cost}(w, z)<z$.cost then
z.edge $:=(w, z)$;
$z . \cos t:=\operatorname{cost}(w, z)$
end

Time complexity:

## Minimum-Weight Spanning Trees (cont.)

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 $\operatorname{cost}(w, z)<z$.cost then
z.edge $:=(w, z)$;
$z . \cos t:=\operatorname{cost}(w, z)$
end

Time complexity: same as that of Dijkstra's algorithm.

## Minimum-Weight Spanning Trees (cont.)



|  | v | a | b | c | d | e | f | g | h |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| v | - | $\mathrm{v}(1)$ | $\mathrm{v}(6)$ | $\infty$ | $\mathrm{v}(9)$ | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| a | - | - | $\mathrm{v}(6)$ | $a(2)$ | $\mathrm{v}(9)$ | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| c | - | - | $\mathrm{v}(6)$ | - | $\mathrm{c}(4)$ | $\infty$ | $c(10)$ | $\infty$ | $\infty$ |
| d | - | - | $\mathrm{v}(6)$ | - | - | $\mathrm{d}(7)$ | $c(10)$ | $\mathrm{d}(12)$ | $\infty$ |
| b | - | - | - | - | - | $\mathrm{b}(3)$ | $c(10)$ | $\mathrm{d}(12)$ | $\infty$ |
| e | - | - | - | - | - | - | $c(10)$ | $\mathrm{d}(12)$ | $\mathrm{e}(5)$ |
| h | - | - | - | - | - | - | $c(10)$ | $\mathrm{h}(11)$ | - |
| f | - | - | - | - | - | - | - | $\mathrm{h}(11)$ | - |
| g | - | - | - | - | - | - | - | - | - |

Figure: An example of the minimum-cost spanning-tree algorithm. Source: redrawn from [Manber 1989, Figure 7.21].

## All Shortest Paths

## Problem

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

## All Shortest Paths

## Problem

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

Basic ideas (of Floyd's algorithm):

- Introduce the notion of a $k$-path, where the largest number of the intermediate vertices is $k$.
- Induct over the sequence of numbers of the vertices.

The best $m$-path from $u$ to $v$ is the best $(<m)$-path from $u$ to $m$ combined with the best $(<m)$-path from $m$ to $v$.

## Floyd's Algorithm

Algorithm All_Pairs_Shortest_Paths(W); begin
\{initialization\}
for $i:=1$ to $n$ do
for $j:=1$ to $n$ do
if $(i, j) \in E$ then $W[i, j]:=\operatorname{length}(i, j)$
else $W[i, j]:=\infty$;
for $i:=1$ to $n$ do $W[i, i]:=0$;
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

## Transitive Closure

## 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

$$
\text { for } y:=1 \text { to } n \text { do }
$$

if $A[m, y]$ then

$$
A[x, y]:=\text { true }
$$

end

