

G205

Fundamentals of Computer Engineering

CLASSES 18, Wed. Nov. 10 2004

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M-W, 1:30pm-3:10pm

All-Pairs Shortest Paths

- ◆ Finding shortest path between all pairs of vertices in a graph
- ◆ Input:
 - $G=(V,E)$
 - $w:E \rightarrow \mathbf{R}$
- ◆ Output: For each pair of vertices u and v in V we want the least weight path from u to v

Representation of Input

- ◆ For APSP graph represents by an adjacency matrix $W = (w_{ij})$
 - $w_{ij} = 0$ if $i = j$
 - $w_{ij} = w(i,j)$ if $i \neq j$ and $(i,j) \in E$
 - $w_{ij} = \infty$ if $i \neq j$ and $(i,j) \notin E$
- ◆ Negative-weight edges \rightarrow OK
- ◆ Negative-weight cycles \rightarrow not OK

Representation of Output

◆ $n \times n$ matrix $D = (d_{ij})$

- d_{ij} = shortest path weight from i to j
- At termination: $d_{ij} = d(i,j)$

◆ Actual shortest paths: Predecessor matrix $\Pi = (\pi_{ij})$

- $\pi_{ij} = \text{NIL}$ if $i=j$ or there is no path from i to j
- π_{ij} = predecessor of j on some path from i to j otherwise

Predecessor Subgraph

- ◆ The subgraph induced by the i -th row of the matrix Π is a shortest path tree with root i
- ◆ For each vertex $i \in V$ we define the predecessor subgraph $G_{\pi,i} = (V_{\pi,i}, E_{\pi,i})$:
 - $V_{\pi,i} = \{j \in V: \pi_{ij} \neq \text{NIL}\} \cup \{i\}$
 - $E_{\pi,i} = \{(\pi_{ij}, j): j \in V_{\pi,i} \setminus \{i\}\}$

Printing APSPs

Print-APSP(Π, i, j)

if $i = j$ then print i

else if $\pi_{ij} = \text{NIL}$

then print no i - j path

else Print-APSP(Π, i, π_{ij})

print j

Some Notation

- ◆ Graph G has $|V|=n$ vertices
- ◆ Matrix are denoted in uppercase D, W, L
- ◆ Matrix elements: d_{ij}, w_{ij}, l_{ij}
- ◆ Iterates of matrices: $D^{(m)} = (d^{(m)}_{ij})$

Shortest Paths and Matrix Multiplication

- ◆ **Dynamic Programming approach:**
 - Characterize the structure of an optimal solution
 - Define its value recursively
 - Compute a solution in a bottom-up fashion
 - Constructing an optimal solution

Structure of a Shortest Path

- ◆ All subpaths of a shortest paths are shortest paths
- ◆ Graph represented by adjacency matrix $W = (w_{ij})$
- ◆ Let p be a shortest path with at most m edges
- ◆ If $i=j$ then p has no edges and weight 0
- ◆ If $i \neq j$ then $p = i \rightsquigarrow k \rightarrow j$ with $i \rightsquigarrow k$ with at most $m-1$ edges and $d(i,j) = d(i,k) + w_{kj}$

Recursive Solution

◆ $l^{(m)}_{ij}$ is the minimum weight of any path from i to j with at most m edges

◆ When $m=0$

■ $l^{(0)}_{ij} = 0$ if $i = j$

■ $l^{(0)}_{ij} = \infty$ if $i \neq j$

◆ When $m \geq 1$

■ $l^{(m)}_{ij} = \min(l^{(m-1)}_{ij}, \min_{1 \leq k \leq n} \{l^{(m-1)}_{ik} + w_{ki}\}) = \min_{1 \leq k \leq n} \{l^{(m-1)}_{ik} + w_{ki}\}$ (since $w_{jj} = 0$ for each j)

The Shortest Path Weight

- ◆ No negative-weight cycles \rightarrow shortest paths have at most $n-1$ edges
- ◆ A path from i to j for which $d(i,j) < \infty$ is simple and with $\leq n-1$ edges or otherwise it cannot have weight $\leq d(i,j)$
- ◆ Actual shortest path weight:

$$d(i,j) = l^{(n-1)}_{ij} = l^{(n)}_{ij} = l^{(m+1)}_{ij} = \dots$$

Computing weights bottom-up

- ◆ Input: $W = (w_{ij})$
- ◆ We compute: $L^{(1)}, L^{(2)}, \dots, L^{(n-1)}$ with $L^{(m)} = (l^{(m)}_{ij}), m = 1, 2, \dots, n-1$
- ◆ $L^{(n-1)}$ contains the shortest-path weights
- ◆ By definition of $L^{(m)}$ is $L^{(1)} = W$
- ◆ Basic step: Extending shortest paths edge by edge: Given $L^{(m-1)}$ and W we obtain $L^{(m)}$

Extending Shortest Paths

Extend-Shortest-Paths(L, W)

for $i = 1$ to n do

for $j = 1$ to n do

$l'_{ij} = \infty$

for $k = 1$ to n do

$l'_{ij} = \min(l'_{ij}, l_{ik} + w_{kj})$

return L'

Matrix Multiplication Analogy

- ◆ Extend-Shortest-Paths costs $O(n^3)$
- ◆ It is like multiplying $n \times n$ matrices:
- ◆ $C = A \times B \rightarrow c_{ij} = \text{SUM}_{(k=1,n)} a_{ik} b_{kj}$
- ◆ Here:
 - $l^{(m-1)} = a$
 - $w = b$
 - $l^{(m)} = c$
 - $\min = +$
 - $+ = *$
 - $\infty = 0$

Computing Shortest-Paths Weights

- ◆ We extend shortest paths edge by edge
- ◆ We compute the sequence:
 - $L^{(1)} = L^{(0)} \times W = W$
 - $L^{(2)} = L^{(1)} \times W = W^2$
 - $L^{(3)} = L^{(2)} \times W = W^3$
 - ...
 - $L^{(n-1)} = L^{(n-2)} \times W = W^{n-1}$

A Slow APSP algorithm

Slow-APSP(W)

$L^{(1)} = W$

for $m = 2$ to $n - 1$ do

$L^{(m)} = \text{Extend-Shortest-Paths}(L^{(m-1)}, W)$

return $L^{(n-1)}$

- ◆ Since $\text{Extend-Shortest-Paths}$ is $O(n^3)$
Slow-APSP(W) is $O(n^4)$

Improving the Running Time

- ◆ Goal: Compute $L^{(n-1)}$ and not the whole sequence $L^{(1)}, L^{(2)}, \dots, L^{(n-1)}$
- ◆ Recall: $L^{(m)} = L^{(n-1)}$ for each $m \geq n - 1$
- ◆ Repeated Squaring:
 - $L^{(1)} = L^{(0)} \times W = W$
 - $L^{(2)} = W^2 = W \times W$
 - $L^{(4)} = W^4 = W^2 \times W^2$
 - ...
 - $L^{(2^{\log(n-1)})} = W^{(2^{\log(n-1)})} = W^{(2^{\log(n-1)-1})} \times W^{(2^{\log(n-1)-1})} = L^{(n-1)}$

A Faster APSP Algorithm

Faster-APSP(W)

$L^{(1)} = W$

$m = 1$

while $m \leq n-1$ do

$L^{(2m)} = \text{Extend-Shortest-Paths}(L^{(m)}, L^{(m)})$

$m = 2m$

return $L^{(m)}$

◆ Faster-APSP(W) is $O(n^3 \log n)$

The Floyd-Warshall Algorithm

- ◆ Another dynamic programming formulation for All-Pairs Shortest Path
 - The structure of a shortest path
 - ◆ Uses intermediate vertices of a shortest path
 - Recursive solution to the ASPSP problem
 - Computing the shortest path weights bottom up

Structure of a Shortest Path, 1

- ◆ An intermediate vertex of a simple path $p = \langle v_1, v_2, \dots, v_l \rangle$ is any vertex of p in $\{v_2, \dots, v_{l-1}\}$
- ◆ Consider $G=(V,E)$ with $V=\{1, \dots, n\}$
- ◆ Consider $K=\{1, \dots, k\}$, for some k
- ◆ For each $i, j \in V$ consider paths with vertices only from K
- ◆ Let p be a shortest paths among them

Structure of a Shortest Path, 2

- ◆ Relationship of p and the $i \rightsquigarrow j$ shortest path with vertices from $K-1 = \{1, \dots, k-1\}$
 - k is not intermediate in $p \rightarrow$ the int. vertices of p are in $K-1 \rightarrow$ shortest path $i \rightsquigarrow j$ with vertices in $K-1$ has also vertices in K
 - If k is an intermediate vertex in p then:
 $p = p_1 p_2$ where $p_1 = i \rightsquigarrow k$ and $p_2 = k \rightsquigarrow j$
where p_1 and p_2 have int. vertices in $K-1$

Recursive Solution to ASPSP

- ◆ Let $d^{(k)}_{ij}$ the weight of a $i \rightsquigarrow j$ shortest path with all int. vertices in K
 - $k=0 \rightarrow d^{(0)}_{ij} = w_{ij}$
 - $k \geq 1 \rightarrow \min\{d^{(k-1)}_{ij}, d^{(k-1)}_{ik} + d^{(k-1)}_{kj}\}$
- ◆ Since for any path all intermediate vertices are in V the matrix $D^{(n)} = (d^{(n)}_{ij})$ is such that $d^{(n)}_{ij} = d(i,j)$, for each $i, j \in V$

Computing the Shortest-Paths Weights Bottom Up

Floyd-Warshall(W)

$D^{(0)} = W$

for k = 1 to n do

 for i = 1 to n do

 for j = 1 to n do

$d_{ij}^{(k)} = \min\{d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\}$

return $D^{(n)}$

Running Time and Space

- ◆ The running time is clearly $\Theta(n^3)$ since the min operation and the sum takes $O(1)$ time
- ◆ Space needed is $\Theta(n^3)$: Each of the n $D^{(k)}$ needs $\Theta(n^2)$ space
- ◆ Dropping all superscript leads to a solution that works in $\Theta(n^2)$ space

C++ implementation of Floyd-Warshall

```
void FW( int n, matrix< int > &fw ) {  
    matrix< int > t( n, n ) = fw;  
    for( int k = 0; k < n; k++ ) {  
        for( int i = 0; i < n; i++ )  
            for( int j = 0; j < n; j++ )  
                fw[ i ][ j ] = min( t[ i ][ j ],  
                                     t[ i ][ k ] + t[ k ][ j ] );  
        t = fw;  
    }  
}
```

Assignments

- ◆ Textbook, Chapter 25, pages 620—640
- ◆ Updated information on the class web page:

www.ece.neu.edu/courses/eceg205/2004fa