## 25.2 The Floyd-Warshall algorithm

In this section, we shall use a different dynamic-programming formulation to solve the all-pairs shortest-paths problem on a directed graph G = (V, E). The resulting algorithm, known as the **Floyd-Warshall algorithm**, runs in  $\Theta(V^3)$  time. As before, negative-weight edges may be present, but we assume that there are no negative-weight cycles. As in Section 25.1, we shall follow the dynamic-programming process to develop the algorithm. After studying the resulting algorithm, we shall present a similar method for finding the transitive closure of a directed graph.

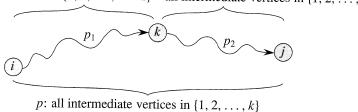
### The structure of a shortest path

In the Floyd-Warshall algorithm, we use a different characterization of the structure of a shortest path than we used in the matrix-multiplication-based all-pairs algorithms. The algorithm considers the "intermediate" vertices of a shortest path, where an *intermediate* vertex of a simple path  $p = \langle v_1, v_2, \ldots, v_l \rangle$  is any vertex of p other than  $v_1$  or  $v_l$ , that is, any vertex in the set  $\{v_2, v_3, \ldots, v_{l-1}\}$ .

The Floyd-Warshall algorithm is based on the following observation. Under our assumption that the vertices of G are  $V = \{1, 2, ..., n\}$ , let us consider a subset  $\{1, 2, ..., k\}$  of vertices for some k. For any pair of vertices  $i, j \in V$ , consider all paths from i to j whose intermediate vertices are all drawn from  $\{1, 2, ..., k\}$ , and let p be a minimum-weight path from among them. (Path p is simple.) The Floyd-Warshall algorithm exploits a relationship between path p and shortest paths from i to j with all intermediate vertices in the set  $\{1, 2, ..., k-1\}$ . The relationship depends on whether or not k is an intermediate vertex of path p.

- If k is not an intermediate vertex of path p, then all intermediate vertices of path p are in the set  $\{1, 2, ..., k-1\}$ . Thus, a shortest path from vertex i to vertex j with all intermediate vertices in the set  $\{1, 2, ..., k-1\}$  is also a shortest path from i to j with all intermediate vertices in the set  $\{1, 2, ..., k\}$ .
- If k is an intermediate vertex of path p, then we break p down into  $i \stackrel{p_1}{\longleftrightarrow} k \stackrel{p_2}{\longleftrightarrow} j$  as shown in Figure 25.3. By Lemma 24.1,  $p_1$  is a shortest path from i to k with all intermediate vertices in the set  $\{1, 2, \ldots, k\}$ . Because vertex k is not an intermediate vertex of path  $p_1$ , we see that  $p_1$  is a shortest path from i to k with all intermediate vertices in the set  $\{1, 2, \ldots, k-1\}$ . Similarly,  $p_2$  is a shortest path from vertex k to vertex j with all intermediate vertices in the set  $\{1, 2, \ldots, k-1\}$ .

all intermediate vertices in  $\{1, 2, \dots, k-1\}$  all intermediate vertices in  $\{1, 2, \dots, k-1\}$ 



**Figure 25.3** Path p is a shortest path from vertex i to vertex j, and k is the highest-numbered intermediate vertex of p. Path  $p_1$ , the portion of path p from vertex i to vertex k, has all intermediate vertices in the set  $\{1, 2, \ldots, k-1\}$ . The same holds for path  $p_2$  from vertex k to vertex j.

# A recursive solution to the all-pairs shortest-paths problem

Basedon the above observations, we define a recursive formulation of shortest-path estimates that is different from the one in Section 25.1. Let  $d_{ij}^{(k)}$  be the weight of a shortest path from vertex i to vertex j for which all intermediate vertices are in the set  $\{1,2,\ldots,k\}$ . When k=0, a path from vertex i to vertex j with no intermediate vertex numbered higher than 0 has no intermediate vertices at all. Such a path has at most one edge, and hence  $d_{ij}^{(0)}=w_{ij}$ . A recursive definition following the above discussion is given by

$$d_{ij}^{(k)} = \begin{cases} w_{ij} & \text{if } k = 0, \\ \min\left(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\right) & \text{if } k \ge 1. \end{cases}$$
(25.5)

Because for any path, all intermediate vertices are in the set  $\{1, 2, \ldots, n\}$ , the matrix  $D^{(n)} = (d_{ij}^{(n)})$  gives the final answer:  $d_{ij}^{(n)} = \delta(i, j)$  for all  $i, j \in V$ .

# Computing the shortest-path weights bottom up

Based on recurrence (25.5), the following bottom-up procedure can be used to compute the values  $d_{ij}^{(k)}$  in order of increasing values of k. Its input is an  $n \times n$  matrix W defined as in equation (25.1). The procedure returns the matrix  $D^{(n)}$  of shortest-path weights.

```
FLOYD-WARSHALL(W)

\begin{array}{lll}
1 & n \leftarrow rows[W] \\
2 & D^{(0)} \leftarrow W \\
3 & \text{for } k \leftarrow 1 \text{ to } n \\
4 & \text{do for } i \leftarrow 1 \text{ to } n \\
5 & \text{do for } j \leftarrow 1 \text{ to } n \\
6 & \text{do } d_{ij}^{(k)} \leftarrow \min\left(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\right) \\
7 & \text{return } D^{(n)}
\end{array}
```

$$D^{(0)} = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & \infty & -5 & 0 & \infty \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix} \qquad \Pi^{(0)} = \begin{pmatrix} \text{NIL} & 1 & 1 & \text{NIL} & 1 \\ \text{NIL} & \text{NIL} & \text{NIL} & \text{NIL} & \text{NIL} \\ 4 & \text{NIL} & 4 & \text{NIL} & \text{NIL} \\ \text{NIL} & \text{NIL} & \text{NIL} & \text{NIL} \\ 4 & \text{NIL} & 4 & \text{NIL} & \text{NIL} \\ \text{NIL} & \text{NIL} & \text{NIL} \\ \text{NIL} & \text{NIL} & \text{NIL} & \text{NIL} \\ \text{NIL} & \text{NIL}$$

**Figure 25.4** The sequence of matrices  $D^{(k)}$  and  $\Pi^{(k)}$  computed by the Floyd-Warshall algorithm for the graph in Figure 25.1.

Figure 25.4 shows the matrices  $D^{(k)}$  computed by the Floyd-Warshall algorithm for the graph in Figure 25.1.

The running time of the Floyd-Warshall algorithm is determined by the triply nested for loops of lines 3–6. Because each execution of line 6 takes O(1) time, the algorithm runs in time  $\Theta(n^3)$ . As in the final algorithm in Section 25.1, the

code is tight, with no elaborate data structures, and so the constant hidden in the  $\Theta$ -notation is small. Thus, the Floyd-Warshall algorithm is quite practical for even moderate-sized input graphs.

## Constructing a shortest path

There are a variety of different methods for constructing shortest paths in the Floyd-Warshall algorithm. One way is to compute the matrix D of shortest-path weights and then construct the predecessor matrix  $\Pi$  from the D matrix. This method can be implemented to run in  $O(n^3)$  time (Exercise 25.1-6). Given the predecessor matrix  $\Pi$ , the PRINT-ALL-PAIRS-SHORTEST-PATH procedure can be used to print the vertices on a given shortest path.

We can compute the predecessor matrix  $\Pi$  "on-line" just as the Floyd-Warshall algorithm computes the matrices  $D^{(k)}$ . Specifically, we compute a sequence of matrices  $\Pi^{(0)}$ ,  $\Pi^{(1)}$ , ...,  $\Pi^{(n)}$ , where  $\Pi = \Pi^{(n)}$  and  $\pi_{ij}^{(k)}$  is defined to be the predecessor of vertex j on a shortest path from vertex i with all intermediate vertices in the set  $\{1, 2, \ldots, k\}$ .

We can give a recursive formulation of  $\pi_{ij}^{(k)}$ . When k = 0, a shortest path from i to j has no intermediate vertices at all. Thus,

$$\pi_{ij}^{(0)} = \begin{cases} \text{NIL} & \text{if } i = j \text{ or } w_{ij} = \infty, \\ i & \text{if } i \neq j \text{ and } w_{ij} < \infty. \end{cases}$$
(25.6)

For  $k \ge 1$ , if we take the path  $i \leadsto k \leadsto j$ , where  $k \ne j$ , then the predecessor of j we choose is the same as the predecessor of j we chose on a shortest path from k with all intermediate vertices in the set  $\{1, 2, \ldots, k-1\}$ . Otherwise, we choose the same predecessor of j that we chose on a shortest path from i with all intermediate vertices in the set  $\{1, 2, \ldots, k-1\}$ . Formally, for  $k \ge 1$ ,

$$\pi_{ij}^{(k)} = \begin{cases} \pi_{ij}^{(k-1)} & \text{if } d_{ij}^{(k-1)} \le d_{ik}^{(k-1)} + d_{kj}^{(k-1)}, \\ \pi_{kj}^{(k-1)} & \text{if } d_{ij}^{(k-1)} > d_{ik}^{(k-1)} + d_{kj}^{(k-1)}. \end{cases}$$
(25.7)

We leave the incorporation of the  $\Pi^{(k)}$  matrix computations into the FLOYD-WARSHALL procedure as Exercise 25.2-3. Figure 25.4 shows the sequence of  $\Pi^{(k)}$  matrices that the resulting algorithm computes for the graph of Figure 25.1. The exercise also asks for the more difficult task of proving that the predecessor subgraph  $G_{\pi,i}$  is a shortest-paths tree with root i. Yet another way to reconstruct shortest paths is given as Exercise 25.2-7.

## Transitive closure of a directed graph

Given a directed graph G = (V, E) with vertex set  $V = \{1, 2, ..., n\}$ , we may wish to find out whether there is a path in G from i to j for all vertex pairs  $i, j \in V$ . The *transitive closure* of G is defined as the graph  $G^* = (V, E^*)$ , where

 $E^* = \{(i, j) : \text{there is a path from vertex } i \text{ to vertex } j \text{ in } G\}$ .

One way to compute the transitive closure of a graph in  $\Theta(n^3)$  time is to assign a weight of 1 to each edge of E and run the Floyd-Warshall algorithm. If there is a path from vertex i to vertex j, we get  $d_{ij} < n$ . Otherwise, we get  $d_{ij} = \infty$ .

There is another, similar way to compute the transitive closure of G in  $\Theta(n^3)$  time that can save time and space in practice. This method involves substitution of the logical operations  $\vee$  (logical OR) and  $\wedge$  (logical AND) for the arithmetic operations min and + in the Floyd-Warshall algorithm. For  $i, j, k = 1, 2, \ldots, n$ , we define  $t_{ij}^{(k)}$  to be 1 if there exists a path in graph G from vertex i to vertex j with all intermediate vertices in the set  $\{1, 2, \ldots, k\}$ , and 0 otherwise. We construct the transitive closure  $G^* = (V, E^*)$  by putting edge (i, j) into  $E^*$  if and only if  $t_{ij}^{(n)} = 1$ . A recursive definition of  $t_{ij}^{(k)}$ , analogous to recurrence (25.5), is

$$t_{ij}^{(0)} = \begin{cases} 0 & \text{if } i \neq j \text{ and } (i, j) \notin E, \\ 1 & \text{if } i = j \text{ or } (i, j) \in E, \end{cases}$$
and for  $k \geq 1$ ,
$$t_{ij}^{(k)} = t_{ij}^{(k-1)} \vee \left( t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)} \right). \tag{25.8}$$

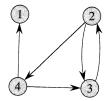
As in the Floyd-Warshall algorithm, we compute the matrices  $T^{(k)} = (t_{ij}^{(k)})$  in order of increasing k.

```
Transitive-Closure (G)
          n \leftarrow |V[G]|
  2
          for i \leftarrow 1 to n
  3
                     do for j \leftarrow 1 to n
                                     do if i = j or (i, j) \in E[G]

then t_{ij}^{(0)} \leftarrow 1

else t_{ij}^{(0)} \leftarrow 0
  4
  5
  6
  7
          for k \leftarrow 1 to n
  8
                     do for i \leftarrow 1 to n
                                     \begin{array}{c} \textbf{do for } j \leftarrow 1 \textbf{ to } n \\ \textbf{do } t_{ij}^{(k)} \leftarrow t_{ij}^{(k-1)} \vee \left(t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)}\right) \end{array}
  9
10
          return T^{(n)}
11
```

Figure 25.5 shows the matrices  $T^{(k)}$  computed by the TRANSITIVE-CLOSURE procedure on a sample graph. The TRANSITIVE-CLOSURE procedure, like the Floyd-Warshall algorithm, runs in  $\Theta(n^3)$  time. On some computers, though, logical operations on single-bit values execute faster than arithmetic operations on integer words of data. Moreover, because the direct transitive-closure algorithm uses only boolean values rather than integer values, its space requirement is less than the



$$T^{(0)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \end{pmatrix} \quad T^{(1)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \end{pmatrix} \quad T^{(2)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \end{pmatrix}$$

Figure 25.5 A directed graph and the matrices  $T^{(k)}$  computed by the transitive-closure algorithm.

Floyd-Warshall algorithm's by a factor corresponding to the size of a word of computer storage.

### **Exercises**

#### 25.2-1

Run the Floyd-Warshall algorithm on the weighted, directed graph of Figure 25.2. Show the matrix  $D^{(k)}$  that results for each iteration of the outer loop.

#### *25.2-2*

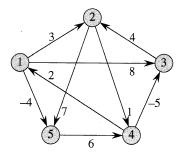
Show how to compute the transitive closure using the technique of Section 25.1.

#### *25.2-3*

Modify the FLOYD-WARSHALL procedure to include computation of the  $\Pi^{(k)}$  matrices according to equations (25.6) and (25.7). Prove rigorously that for all  $i \in V$ , the predecessor subgraph  $G_{\pi,i}$  is a shortest-paths tree with root i. (Hint: To show that  $G_{\pi,i}$  is acyclic, first show that  $\pi^{(k)}_{ij} = l$  implies  $d^{(k)}_{ij} \geq d^{(k)}_{il} + w_{lj}$ , according to the definition of  $\pi^{(k)}_{ij}$ . Then, adapt the proof of Lemma 24.16.)

### 25.2-4

As it appears above, the Floyd-Warshall algorithm requires  $\Theta(n^3)$  space, since we compute  $d_{ij}^{(k)}$  for i, j, k = 1, 2, ..., n. Show that the following procedure, which simply drops all the superscripts, is correct, and thus only  $\Theta(n^2)$  space is required.



$$L^{(1)} = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & \infty & -5 & 0 & \infty \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix} \quad L^{(2)} = \begin{pmatrix} 0 & 3 & 8 & 2 & -4 \\ 3 & 0 & -4 & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & \infty & 1 & 6 & 0 \end{pmatrix}$$

$$L^{(3)} = \begin{pmatrix} 0 & 3 & -3 & 2 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 11 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix} \qquad L^{(4)} = \begin{pmatrix} 0 & 1 & -3 & 2 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix}$$

**Figure 25.1** A directed graph and the sequence of matrices  $L^{(m)}$  computed by SLOW-ALL-PAIRS-SHORTEST-PATHS. The reader may verify that  $L^{(5)} = L^{(4)} \cdot W$  is equal to  $L^{(4)}$ , and thus  $L^{(m)} = L^{(4)}$  for all  $m \ge 4$ .

Since  $2^{\lceil \lg(n-1) \rceil} \ge n-1$ , the final product  $L^{(2^{\lceil \lg(n-1) \rceil})}$  is equal to  $L^{(n-1)}$ .

The following procedure computes the above sequence of matrices by using this technique of *repeated squaring*.