

# Lecture 15: The Floyd-Warshall Algorithm

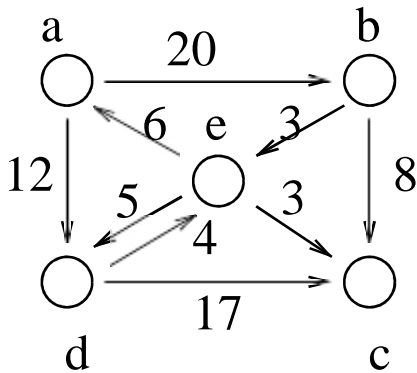
CLRS section 25.2

## Outline of this Lecture

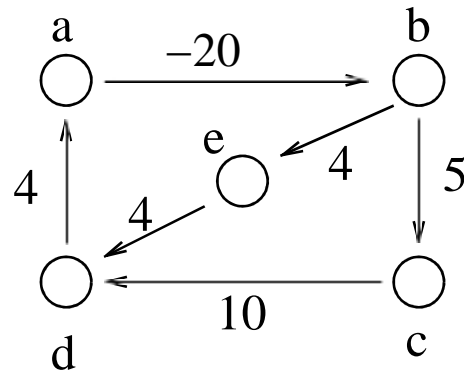
- Recalling the all-pairs shortest path problem.
- Recalling the previous two solutions.
- The Floyd-Warshall Algorithm.

## The All-Pairs Shortest Paths Problem

Given a weighted digraph  $G = (V, E)$  with a weight function  $w : E \rightarrow \mathbf{R}$ , where  $\mathbf{R}$  is the set of real numbers, determine the **length of the shortest path** (i.e., **distance**) between all pairs of vertices in  $G$ . Here we assume that there are no cycle with **zero or negative cost**.



without negative cost cycle



with negative cost cycle

## Solutions Covered in the Previous Lecture

**Solution 1:** Assume no negative *edges*.

Run Dijkstra's algorithm,  $n$  times, once with each vertex as source.

$O(n^3 \log n)$ .  $O(n^3)$  with more sophisticated data structures.

**Solution 2:** Assume no negative *cycles*.

Dynamic programming solution, based on a natural decomposition of the problem.

$O(n^4)$ .  $O(n^3 \log n)$  using “repeated squaring”.

**This lecture:** Assume no negative *cycles*.

develop another dynamic programming algorithm, the

*Floyd-Warshall algorithm*, with time complexity  $O(n^3)$ .

Also illustrates that there can be more than one way of developing a dynamic programming algorithm.

### Solution 3: the Input and Output Format

As in the previous dynamic programming algorithm, we assume that the graph is represented by an  $n \times n$  matrix with the weights of the edges:

$$w_{ij} = \begin{cases} 0 & \text{if } i = j, \\ w(i, j) & \text{if } i \neq j \text{ and } (i, j) \in E, \\ \infty & \text{if } i \neq j \text{ and } (i, j) \notin E. \end{cases}$$

**Output Format:** an  $n \times n$  distance  $D = [d_{ij}]$  where  $d_{ij}$  is the distance from vertex  $i$  to  $j$ .

## Step 1: The Floyd-Warshall Decomposition

**Definition:** The vertices  $v_2, v_3, \dots, v_{l-1}$  are called the *intermediate vertices* of the path  $p = \langle v_1, v_2, \dots, v_l \rangle$ .

- Let  $d_{ij}^{(k)}$  be the **length of the shortest path** from  $i$  to  $j$  such that *all* intermediate vertices on the path (**if any**) are in set  $\{1, 2, \dots, k\}$ .

$d_{ij}^{(0)}$  is set to be  $w_{ij}$ , i.e., no intermediate vertex.

Let  $D^{(k)}$  be the  $n \times n$  matrix  $[d_{ij}^{(k)}]$ .

- Claim:  $d_{ij}^{(n)}$  is the distance from  $i$  to  $j$ . So our aim is to compute  $D^{(n)}$ .
- Subproblems:** compute  $D^{(k)}$  for  $k = 0, 1, \dots, n$ .

## Step 2: Structure of shortest paths

### Observation 1:

A shortest path does not contain the same vertex twice.

Proof: A path containing the same vertex twice contains a cycle. Removing cycle gives a shorter path.

**Observation 2:** For a shortest path from  $i$  to  $j$  such that any intermediate vertices on the path are chosen from the set  $\{1, 2, \dots, k\}$ , there are two possibilities:

1.  $k$  is not a vertex on the path,

The shortest such path has length  $d_{ij}^{(k-1)}$ .

2.  $k$  is a vertex on the path.

The shortest such path has length  $d_{ik}^{(k-1)} + d_{kj}^{(k-1)}$ .

## Step 2: Structure of shortest paths

Consider a **shortest path** from  $i$  to  $j$  containing the vertex  $k$ . It consists of a subpath from  $i$  to  $k$  and a subpath from  $k$  to  $j$ .

Each subpath can only contain intermediate vertices in  $\{1, \dots, k-1\}$ , and must be as short as possible, namely they have lengths  $d_{ik}^{(k-1)}$  and  $d_{kj}^{(k-1)}$ .

Hence the path has length  $d_{ik}^{(k-1)} + d_{kj}^{(k-1)}$ .

Combining the two cases we get

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

### Step 3: the Bottom-up Computation

- Bottom:  $D^{(0)} = [w_{ij}]$ , the weight matrix.
- Compute  $D^{(k)}$  from  $D^{(k-1)}$  using

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

for  $k = 1, \dots, n$ .



## The Floyd-Warshall Algorithm: Version 1

**Floyd-Warshall**(  $w, n$  )

```
{ for  $i = 1$  to  $n$  do                                initialize
    for  $j = 1$  to  $n$  do
        {  $D^0[i, j] = w[i, j];$ 
           $pred[i, j] = nil;$ 
        }

    for  $k = 1$  to  $n$  do                                dynamic programming
        for  $i = 1$  to  $n$  do
            for  $j = 1$  to  $n$  do
                if  $(d^{(k-1)}[i, k] + d^{(k-1)}[k, j] < d^{(k-1)}[i, j])$ 
                    {  $d^{(k)}[i, j] = d^{(k-1)}[i, k] + d^{(k-1)}[k, j];$ 
                       $pred[i, j] = k;$  }
                else  $d^{(k)}[i, j] = d^{(k-1)}[i, j];$ 
    return  $d^{(n)}[1..n, 1..n];$ 
}
```

## Comments on the Floyd-Warshall Algorithm

- The algorithm's running time is clearly  $\Theta(n^3)$ .
- The predecessor pointer  $\text{pred}[i, j]$  can be used to extract the final path (see later ).
- Problem: the algorithm uses  $\Theta(n^3)$  space.  
It is possible to reduce this down to  $\Theta(n^2)$  space by keeping only one matrix instead of  $n$ .  
Algorithm is on next page. Convince yourself that it works.

## The Floyd-Warshall Algorithm: Version 2

**Floyd-Warshall**(  $w, n$  )

```
{ for  $i = 1$  to  $n$  do                                initialize
    for  $j = 1$  to  $n$  do
        {  $d[i, j] = w[i, j];$ 
           $pred[i, j] = nil$ 
        }

    for  $k = 1$  to  $n$  do                                dynamic programming
        for  $i = 1$  to  $n$  do
            for  $j = 1$  to  $n$  do
                if ( $d[i, k] + d[k, j] < d[i, j]$ )
                    { $d[i, j] = d[i, k] + d[k, j];$ 
                      $pred[i, j] = k;$ }
    return  $d[1..n, 1..n];$ 
}
```

## Extracting the Shortest Paths

The predecessor pointers  $\text{pred}[i, j]$  can be used to extract the final path. The idea is as follows.

Whenever we discover that the shortest path from  $i$  to  $j$  passes through an intermediate vertex  $k$ , we set  $\text{pred}[i, j] = k$ .

If the shortest path does not pass through any intermediate vertex, then  $\text{pred}[i, j] = \text{nil}$ .

To find the shortest path from  $i$  to  $j$ , we consult  $\text{pred}[i, j]$ . If it is nil, then the shortest path is just the edge  $(i, j)$ .

Otherwise, we recursively compute the shortest path from  $i$  to  $\text{pred}[i, j]$  and the shortest path from  $\text{pred}[i, j]$  to  $j$ .

## The Algorithm for Extracting the Shortest Paths

```
Path(  $i, j$  )
{
    if (  $pred[i, j] = nil$  )    single edge
        output (  $i, j$  );
    else                        compute the two parts of the path
    {
        Path(  $i, pred[i, j]$  );
        Path(  $pred[i, j], j$  );
    }
}
```

### Example of Extracting the Shortest Paths

Find the shortest path from vertex 2 to vertex 3.

2..3	Path (2, 3)	$pred[2, 3] = 4$	
2..4..3	Path (2, 4)	$pred[2, 4] = 5$	
2..5..4..3	Path (2, 5)	$pred[2, 5] = nil$	<i>Output(2,5)</i>
25..4..3	Path (5, 4)	$pred[5, 4] = nil$	<i>Output(5,4)</i>
254..3	Path (4, 3)	$pred[4, 3] = 6$	
254..6..3	Path (4, 6)	$pred[4, 6] = nil$	<i>Output(4,6)</i>
2546..3	Path (6, 3)	$pred[6, 3] = nil$	<i>Output(6,3)</i>
25463			