

# Shortest Paths Problems

Input: a directed graph  $G = (V, E)$  and a **weight** function  $w : E \rightarrow R$ .

The weight of a path  $p = v_0, v_1, v_2, \dots, v_k$  is

$$w(p) = \sum_{i=1}^k w(v_{i-1}, v_i).$$

The weight of the **shortest path** from  $u$  to  $v$ ,  $\delta(u, v)$  is the minimum of  $w(p)$  for all  $p$  connecting  $u$  to  $v$ , and  $\infty$  if there is no such path in  $G$ .

# All-Pairs Shortest Paths

## Single-Source Shortest Paths:

Compute shortest paths from a given source to all vertices in the graph.

## All-Pairs Shortest Paths:

Given a graph  $G = (V, E)$ ,  $|V| = n$ , and a weight function  $w$  on the edges, compute the shortest paths between all pairs of vertices.

The problem is not well defined in the presence of a **negative weight cycle** in the graph.

# All-Pairs Shortest Path Algorithms

We can solve an all-pairs shortest-paths problem by running a single source shortest-paths algorithm  $n$  times, once for each vertex as a source. Then the running time is:

- $O(n^3)$  if we use the Dijkstra algorithm (assuming no negative edge weights);
- $O(n^2|E|)$  if we use Bellman-Ford algorithm - i.e.,  $O(n^4)$  if the graph is dense.

Instead we will give a direct approach to finding the shortest paths between all pairs of vertices. We assume that negative weights exist, but no negative weight cycle.

First we will give an  $O(n^4)$  algorithm, then improve it to  $O(n^3 \log n)$ . Later we will give a even better algorithm running in  $O(n^3)$  time.

# All Pairs Shortest Paths

Input: an adjacency matrix  $W$  where  $w_{i,j}$  is the weight of the edge  $(i, j)$ :

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

Output:

A matrix  $D$  where  $d_{i,j}$  is the shortest path from  $i$  to  $j$ .

## The Basic Idea

Define  $d_{i,j}^{(m)}$  to be the shortest path between  $i$  and  $j$  using paths of up to  $m$  edges. When  $m = 0$ , we have

$$d_{ij}^{(0)} = \begin{cases} 0 & : i = j \\ \infty & : i \neq j \end{cases}$$

Recursively we define,

$$d_{i,j}^{(m)} = \min_{1 \leq k \leq n} [d_{i,k}^{(m-1)} + w_{k,j}].$$

If there are no negative weight cycles then no shortest path has more than  $n - 1$  edges.

How to compute these matrices?

## Computing the Matrices

Define a sequence of matrices  $D^{(1)}, D^{(2)}, \dots, D^{(n-1)}$ , where for  $m = 1, 2, \dots, n-1$ , we have  $D^{(m)} = \left( d_{ij}^{(m)} \right)$ .

Note that  $D^{(1)} = W$ .

The key procedure is to compute the matrix  $D^{(m)}$  given  $D^{(m-1)}$  and  $W$ : extending the shortest paths computed so far by one more edge.

**Theorem 1.** *The procedure  $EXTEND-SHORTEST-PATHS(D^{(m-1)}, W)$  returns the matrix  $D' = D^{(m)}$ .*

**Theorem 2.** *For  $n \times n$  matrices  $D$  and  $W$ , the procedure  $EXTEND-SHORTEST-PATHS(D^{(m-1)}, W)$  takes  $\Theta(n^3)$  steps.*

# Slow All-Pairs Algorithm

**Theorem 3.** *The SLOW-ALL-PAIRS-SHORTEST-PATHS algorithm computes the correct shortest paths and terminates in  $\Theta(n^4)$  steps.*



# Shortest Paths and Matrix Multiplication

EXTEND-SHORTEST-PATHS procedure is closely related to MATRIX-MULTIPLY.

Let  $C = A \cdot B$  be the product of two  $n \times n$  matrices  $A$  and  $B$ . For  $i, j = 1, 2, \dots, n$ , we compute

$$c_{ij} = \sum_{k=1}^n a_{ik} \cdot b_{kj}$$

Note that by substituting,

$$d^{(m-1)} \rightarrow a$$

$$w \rightarrow b$$

$$d^{(m)} \rightarrow c$$

$$\min \rightarrow +$$

$$+ \rightarrow \cdot$$

in

$$d_{i,j}^{(m)} = \min_{1 \leq k \leq n} [d_{i,k}^{(m-1)} + w_{k,j}]$$

we get matrix multiplication.

## Repeated Squaring

Let  $D' = D \cdot W$ , where the operation  $\cdot$  is the output of the procedure EXTEND-SHORTEST-PATHS( $D, W$ ).

The SLOW-ALL-PAIRS-SHORTEST-PATHS algorithm starts with  $D^{(1)} = W$  and computes  $D^{(m)} = W^m$  for  $m = 2, \dots, n - 1$ .

Like the “product” operation our “ $\cdot$ ” operation is **associative**, i.e.

$$A \cdot (B \cdot C) = (A \cdot B) \cdot C$$

Assume that  $n - 1 = 2^k$ . A faster method for computing  $D^{n-1}$  is:

For  $t = 1$  to  $k$  do

$$D^{2^t} = D^{2^{t-1}} \cdot D^{2^{t-1}}$$

# Fast All-Pairs Shortest Paths Algorithm

We need to compute  $D^{(m)}$  for some  $m \geq n - 1$ .

Let  $m = 2^{\lceil \log_2(n-1) \rceil}$

**Theorem 4.** *The FAST-ALL-PAIRS-SHORTEST-PATHS algorithm computes the correct distances in  $\Theta(n^3 \log n)$  steps.*

# The Floyd-Warshall algorithm

The previous algorithm extends in each iteration the number of edges used by the paths.

This algorithm extends the set of vertices that can be used as **intermediate** vertices on the paths.

For a path  $P = v_1, v_2, \dots, v_{k-1}, v_k$ , the edges  $v_2, \dots, v_{k-1}$  are intermediate edges.

Let  $V = \{1, \dots, n\}$ .

In iteration  $k$ , the algorithm computes all pairs shortest paths with intermediate vertices in  $\{1, \dots, k\}$ .

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Let  $d_{i,j}^{(k)}$  = the distance of a shortest path from  $i$  to  $j$  using only vertices of  $\{1, \dots, k\}$ .

**Lemma 1.**

$$d_{i,j}^{(k)} = \begin{cases} w_{i,j} & \text{if } k = 0 \\ \text{MIN} \left[ d_{i,j}^{(k-1)}, d_{i,k}^{(k-1)} + d_{k,j}^{(k-1)} \right] & \text{if } k \geq 1 \end{cases}$$

**Proof.** Let  $P$  be a shortest path from  $i$  to  $j$  using vertices in  $\{1, \dots, k\}$ .

If  $P$  does not use  $k$  then  $d_{i,j}^{(k)} = d_{i,j}^{(k-1)}$ .

Otherwise  $P$  consists of a path  $P_1$  from  $i$  to  $k$ , followed by a path  $P_2$  from  $k$  to  $j$ .

$P_1$  is a shortest path from  $i$  to  $k$  in  $\{1, \dots, k-1\}$  and  $P_2$  is a shortest path from  $k$  to  $j$  in  $\{1, \dots, k-1\}$ .  $\square$

**Theorem 5.** *The run-time of the Floyd-Warshall algorithm is  $O(n^3)$  steps.*

# Transitive Closure

Given a directed graph  $G$ , the **transitive closure** of  $G$  is a directed graph  $G^* = (V, E^*)$ , where

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

Given  $G$ , we can compute  $G^*$  by computing all pairs shortest paths with all edges having weight 1.

More efficiently:

Let

$$t_{i,j}^{(0)} = \begin{cases} 0 & \text{if } i \neq j \text{ and } (i, j) \notin E \\ 1 & \text{if } i = j \text{ or } (i, j) \in E \end{cases}$$

For  $k \geq 1$

$$t_{i,j}^{(k)} = t_{i,j}^{(k-1)} \vee (t_{i,k}^{(k-1)} \wedge t_{k,j}^{(k-1)}).$$