

# Data Structures and Algorithms

Lecture 33

Aniket Basu Roy

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## 1 Agenda

### 1.1 All-Pairs Shortest Paths

- Problem definition
- Dynamic programming via edge-count subproblems
- Matrix multiplication analogy: the  $(\min, +)$  semiring
- Slow APSP:  $O(n^4)$
- Fast APSP via repeated squaring:  $O(n^3 \log n)$

## 2 Problem Definition

**Input:** Weighted directed graph  $G = (V, E, w)$ ,  $|V| = n$ ; assume no negative-weight cycles.

**Output:**  $\delta(i, j)$  for all pairs  $i, j \in V$ ; represented as an  $n \times n$  matrix  $\Delta$ .

### 2.1 Why not simply run SSSP $n$ times?

- $n \times$  Bellman-Ford:  $O(n \cdot VE) = O(n^4)$  for dense graphs.
- $n \times$  Dijkstra (non-negative weights):  $O(n \cdot m \log n) = O(n^3 \log n)$  for dense graphs.

Today's approach also achieves  $O(n^3 \log n)$  but works with negative weights (no negative cycles).

## 3 Weight Matrix

Represent the graph as an  $n \times n$  matrix  $W = (w_{ij})$ :

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

## 4 Dynamic Programming Formulation

### 4.1 Subproblem

Let  $\ell_{ij}^{(m)}$  = minimum weight of any path from  $i$  to  $j$  using **at most**  $m$  edges.

Denote the matrix  $L^{(m)} = (\ell_{ij}^{(m)})$ .

### 4.2 Base Case

A path using  $\leq 1$  edge:

$$\ell_{ij}^{(1)} = w_{ij} \quad \Rightarrow \quad L^{(1)} = W.$$

(Recall  $w_{ii} = 0$ , so trivial zero-length paths are captured.)

### 4.3 Recurrence

Extend paths of  $\leq m - 1$  edges by one more edge ( $k \rightarrow j$ ):

$$\ell_{ij}^{(m)} = \min_{1 \leq k \leq n} \{ \ell_{ik}^{(m-1)} + w_{kj} \}.$$

**Proof:** Any path  $i \rightsquigarrow j$  using  $\leq m$  edges either uses  $\leq m - 1$  edges (captured by  $k = j$ ,  $w_{jj} = 0$ ), or has last edge  $(k, j)$  with prefix using  $\leq m - 1$  edges. By optimal substructure, the prefix weight is  $\ell_{ik}^{(m-1)}$ . ■

### 4.4 Stopping Condition

Since there are no negative-weight cycles, all shortest paths are simple. A simple path in a graph with  $n$  vertices uses at most  $n - 1$  edges, so:

$$L^{(n-1)} = \Delta \quad (\text{the true all-pairs shortest-path matrix}).$$

## 5 The $(\min, +)$ Matrix Product

### 5.1 Definition

For two  $n \times n$  matrices  $A$  and  $B$ , define their  $(\min, +)$  **product**  $A \otimes B$  (pronounced as A op B):

$$(A \otimes B)_{ij} = \min_{1 \leq k \leq n} (a_{ik} + b_{kj}).$$

Replace the standard (sum,  $\times$ ) semiring with (min, +).

## 5.2 Observation

$$L^{(m)} = L^{(m-1)} \otimes W = \underbrace{W \otimes W \otimes \cdots \otimes W}_{m \text{ times}}.$$

Computing  $L^{(n-1)}$  is equivalent to raising  $W$  to the  $(n-1)$ -st power in the  $(\min, +)$  semiring.

## 6 EXTEND: One $(\min, +)$ Matrix Multiplication

```
EXTEND(A, B, n)           // computes C = A op B
1. let C be a new n×n matrix
2. for i = 1 to n
3.     for j = 1 to n
4.         c[i][j] = infinity
5.         for k = 1 to n
6.             c[i][j] = min(c[i][j], a[i][k] + b[k][j])
7. return C
```

Time:  $\Theta(n^3)$  — identical structure to ordinary matrix multiplication.

## 7 Slow APSP

Compute  $L^{(1)}, L^{(2)}, \dots, L^{(n-1)}$  iteratively:

```
SLOW-APSP(W, n)
1. L = W           // L = L^(1)
2. for m = 2 to n-1
3.     L = EXTEND(L, W, n) // L <- L op W
4. return L       // L = L^(n-1) = Delta
```

Each of the  $n - 2$  calls to EXTEND costs  $\Theta(n^3)$ :

$$T_{\text{Slow-APSP}} = \Theta(n^4).$$

## 8 Fast APSP: Repeated Squaring

**Key observation:**  $(\min, +)$  multiplication is associative, so:

$$L^{(2m)} = L^{(m)} \otimes L^{(m)}.$$

We do not need every intermediate matrix. Compute only  $L^{(1)}, L^{(2)}, L^{(4)}, \dots, L^{(2^{\lceil \lg(n-1) \rceil})}$ .

```

FASTER-APSP(W, n)
1. L = W // L = L^(1) = L^(2^0)
2. m = 1
3. while m < n-1
4.     L = EXTEND(L, L, n) // L <- L op L (doubles m)
5.     m = 2*m
6. return L

```

### 8.1 Correctness

The while-loop runs  $\lceil \lg(n-1) \rceil$  times. After termination,  $m \geq n-1$ , so  $L = L^{(m)}$ . Since  $L^{(m)} = L^{(n-1)} = \Delta$  for all  $m \geq n-1$  (shortest paths are simple), the result is correct. ✓

### 8.2 Time Complexity

$O(\log n)$  calls to EXTEND, each  $\Theta(n^3)$ :

$$T_{\text{Faster-APSP}} = O(n^3 \log n).$$

## 9 Summary

Algorithm	Handles Neg. Weights?	Time Complexity
$n \times$ Dijkstra	No (non-negative weights only)	$O(n^3 \log n)$
Slow APSP (DP)	Yes (no neg. cycles)	$\Theta(n^4)$
Faster APSP (repeated sq.)	Yes (no neg. cycles)	$O(n^3 \log n)$

## 10 Questions

- Verify  $L^{(2)} = W \otimes W$  gives correct shortest paths using  $\leq 2$  edges on a small example.
- Why is it safe to stop at  $m \geq n-1$  even if  $m > n-1$ ? What property of  $L^{(m)}$  ensures this?