# **All-Pairs Shortest Paths**

Tao Hou

## **Why study it?**

We can solve an all-pairs shortest-paths problem by running a single-source shortest-paths algorithm *for each vertex*:

- $\blacksquare$  Use Dijkstra's algorithm:  $O(|V||E|\log(|V|))$ 
	- ▶ For sparse graph,  $|E| = \Theta(|V|)$ :  $O(|V|^2 \log(|V|))$  (not too bad)
	- ▶ For dense graph,  $|E| = \Theta(|V|^2)$ :  $O(|V|^3 \log(|V|))$  (can do better)
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We introduce *Floyd-Warshall* algorithm:

Run in  $O(|V|^3)$  time and allow negative weights

#### **Floyd-Warshall: Setting**

- Assume that the vertices are numbered  $1, 2, \ldots, n$  where  $n = |V|$
- **n** The input is an  $n \times n$  matrix  $W = (w_{i,j})$  representing the edge weights (an *augmentation* of adjacency matrix):

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w_{i,j} = \begin{cases} 0 & \text{if } i = j \\ \text{weight of edge } (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}
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- Allow negative-weight edges, but *assume that the input graph contains no negative-weight cycles*
- Returns an  $n \times n$  matrix  $D = (d_{i,j})$ , where  $d_{i,j} = \delta(i,j)$
- **Also returns a** *predecessor matrix*  $\Pi = (\pi_{i,j})$

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\pi_{ij} = \begin{cases} \text{ Nil} & i = j \text{ or no path from } i \text{ to } j \\ \text{Predecessor of } j \text{ on a shortest path from } i \text{ to } j \quad \text{otherwise} \end{cases}
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▶ *i*-th row of Π defines a shortest-paths tree rooted at *i* (the procedure to print a shortest path from *i* should be evident from previous contents)

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- A dynamic-programming approach utilizing the *optimal substructure property* of shortest paths
- As can be imagined, the parameter of the *OPT* function contains: *i and j, the start and end vertices*
- However, if your *OPT* contains only *i*, *j*, then:

*d*(*i*, *j*) = min {{*d*(*i*,  $\ell$ ) + *d*( $\ell$ , *j*) |  $\ell \in V$ } ∪ {*d*(*i*, *j*)}}

- It would be nearly *impossible* to find a valid **evaluation order** 
	- $\triangleright$  There is no natural definition of 'size' for the problems  $d(i, j)$ : they are all 'equal'; no one is a natural 'subproblem' of another
	- ▶ Also no natural *base cases*

- The solution is that, we introduce **another parameter** *k*, and consider all paths from *i* to *j* whose *intermediate vertices* are ≤ *k*
	- ▶ E.g., path  $p = (v_1 = i, v_2, \ldots, v_{q-1}, v_q = j)$  where  $v_2, \ldots, v_{q-1} \leq k$

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- *OPT function*: Let *d* (*k*)  $\hat{a}^{(n)}_{ij} := d(i,j,k)$  be the minimum weight of all paths from  $i$  to  $j$  with intermediate vertices ≤ *k*
- $\blacksquare$  We have the following immediate evidence why this definition makes sense: (1) We could easily identify the *base case:*  $d_{i,j}^{(0)} = w_{i,j}$ (2) There is a natural notion of "size": *k*

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 $\blacksquare$  So,

$$
d_{i,j}^{(k)} = \begin{cases} w_{i,j} & k=0\\ \min\left\{d_{i,j}^{(k-1)}, d_{i,k}^{(k-1)} + d_{k,j}^{(k-1)}\right\} & k>0 \end{cases}
$$

#### **Floyd-Warshall: Algorithm**

**FLOYD-WARSHALL**(*W*) 1  $D^{(0)} = W$ 2 **for**  $k = 1, ..., n$ 3  $D^{(k)} := (d_{ii}^{(k)})$ *i*,*j*  $\Big)$  be a new  $n \times n$  matrix 4 **for**  $i = 1, ..., n$ 5 **for**  $j = 1, ..., n$ 6  $d_{ij}^{(k)} = \min \left\{ d_{ij}^{(k-1)} \right\}$ *i*,*j* , *d* (*k*−1)  $\frac{d^{(k-1)}}{d^{(k)}} + d^{(k-1)}}_{k,j}$ *k*,*j*  $\overline{)}$ 

Time complexity:  $\Theta(|V|^3)$ , or  $\Theta(n^3)$ 

Recall that we also need to compute a *predecessor matrix*  $\Pi = (\pi_{i,j})$ 

 $\pi_{i,j} =$  $i = j$  or no path from *i* to *j* Predecessor of *j* on a shortest path from *i* to *j* otherwise

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 $\pi_{i,j} =$  $i = j$  or no path from *i* to *j* Predecessor of *j* on a shortest path from *i* to *j* otherwise ▶ *i*-th row of Π defines a shortest-paths tree rooted at *i* We have  $\Pi^{(k)} = \left(\pi^{(k)}_{ii}\right)$ *i*,*j* ) corresponding to each *D* (*k*)  $\pi^{(k)}_{i,j} =$  $\begin{array}{c} \hline \end{array}$  $\begin{array}{c} \n\end{array}$ Nil *<sup>i</sup>* <sup>=</sup> *<sup>j</sup>* or no path from *<sup>i</sup>* to *<sup>j</sup>* with intermediate vertices ≤ *k* Predecessor of *j* on a shortest path from *i* to *j* with intermediate vertices  $\leq k$ otherwise

**We simply let**  $\Pi = \Pi^{(n)}$ 

**Base case** 

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\pi_{i,j}^{(0)} = \begin{cases} \text{Nil} & \text{if } i = j \text{ or } (i,j) \notin E \\ i & \text{if } i \neq j \text{ and } (i,j) \in E \end{cases}
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\pi_{i,j}^{(k)} = \begin{cases}\n\text{if } d_{i,j}^{(k-1)} \leq d_{i,k}^{(k-1)} + d_{k,j}^{(k-1)} \\
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#### **Floyd-Warshall: With** Π **matrix update**

