

Lecture 19

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Shortest paths

Single-source shortest paths

- Nonnegative edge weights
 - Dijkstra's algorithm: $O(E + V \lg V)$
- General
 - Bellman-Ford: $O(VE)$
- DAG
 - One pass of Bellman-Ford: $O(V + E)$

All-pairs shortest paths

- Nonnegative edge weights
 - Dijkstra's algorithm $|V|$ times: $O(VE + V^2 \lg V)$
- General
 - Three algorithms today.

All-pairs shortest paths

Input: Digraph $G = (V, E)$, where $|V| = n$, with edge-weight function $w : E \rightarrow \mathbb{R}$.

Output: $n \times n$ matrix of shortest-path lengths $\delta(i, j)$ for all $i, j \in V$.

IDEA #1:

- Run Bellman-Ford once from each vertex.
- Time = $O(V^2E)$.
- Dense graph $\Rightarrow O(V^4)$ time.

Good first try!

Dynamic programming

Consider the $n \times n$ adjacency matrix $A = (a_{ij})$ of the digraph, and define

$d_{ij}^{(m)}$ = weight of a shortest path from i to j that uses at most m edges.

Claim: We have

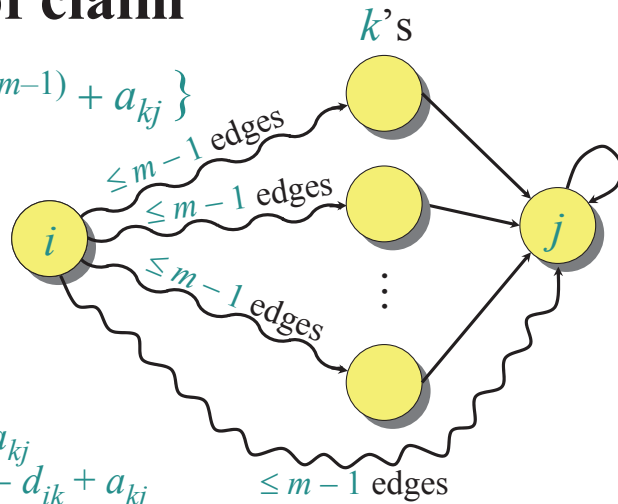
$$d_{ij}^{(0)} = \begin{cases} 0 & \text{if } i = j, \\ \infty & \text{if } i \neq j, \end{cases}$$

and for $m = 1, 2, \dots, n - 1$,

$$d_{ij}^{(m)} = \min_k \{d_{ik}^{(m-1)} + a_{kj}\}.$$

Proof of claim

$$d_{ij}^{(m)} = \min_k \{ d_{ik}^{(m-1)} + a_{kj} \}$$



Relaxation!

for $k \leftarrow 1$ to n

do if $d_{ij} > d_{ik} + a_{kj}$
then $d_{ij} \leftarrow d_{ik} + a_{kj}$

Note: No negative-weight cycles implies

$$\delta(i, j) = d_{ij}^{(n-1)} = d_{ij}^{(n)} = d_{ij}^{(n+1)} = \dots$$

Matrix multiplication

Compute $C = A \cdot B$, where C , A , and B are $n \times n$ matrices:

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

Time = $\Theta(n^3)$ using the standard algorithm.

What if we map “+” \rightarrow “min” and “.” \rightarrow “+”?

$$c_{ij} = \min_k \{ a_{ik} + b_{kj} \}.$$

Thus, $D^{(m)} = D^{(m-1)}$ “ \times ” A .

$$\text{Identity matrix} = I = \begin{pmatrix} 0 & \infty & \infty & \infty \\ \infty & 0 & \infty & \infty \\ \infty & \infty & 0 & \infty \\ \infty & \infty & \infty & 0 \end{pmatrix} = D^0 = (d_{ij}^{(0)}).$$

Matrix multiplication (continued)

The $(\min, +)$ multiplication is *associative*, and with the real numbers, it forms an algebraic structure called a *closed semiring*.

Consequently, we can compute

$$\begin{aligned} D^{(1)} &= D^{(0)} \cdot A = A^1 \\ D^{(2)} &= D^{(1)} \cdot A = A^2 \\ &\vdots \\ D^{(n-1)} &= D^{(n-2)} \cdot A = A^{n-1}, \end{aligned}$$

yielding $D^{(n-1)} = (\delta(i, j))$.

Time = $\Theta(n \cdot n^3) = \Theta(n^4)$. No better than $n \times$ B-F.

Improved matrix multiplication algorithm

Repeated squaring: $A^{2k} = A^k \times A^k$.

Compute $A^2, A^4, \dots, A^{2^{\lceil \lg(n-1) \rceil}}$.

$O(\lg n)$ squarings

Note: $A^{n-1} = A^n = A^{n+1} = \dots$.

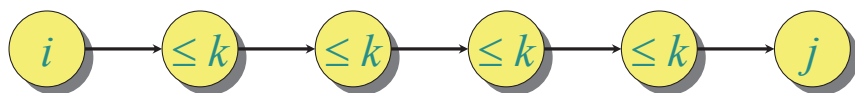
Time = $\Theta(n^3 \lg n)$.

To detect negative-weight cycles, check the diagonal for negative values in $O(n)$ additional time.

Floyd-Warshall algorithm

Also dynamic programming, but faster!

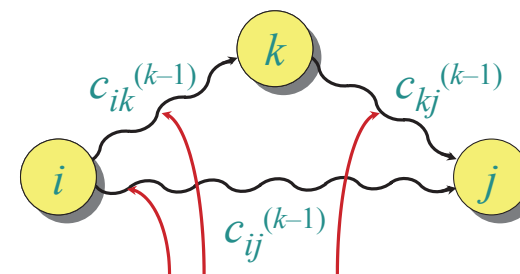
Define $c_{ij}^{(k)}$ = weight of a shortest path from i to j with intermediate vertices belonging to the set $\{1, 2, \dots, k\}$.



Thus, $\delta(i, j) = c_{ij}^{(n)}$. Also, $c_{ij}^{(0)} = a_{ij}$.

Floyd-Warshall recurrence

$$c_{ij}^{(k)} = \min_k \{c_{ij}^{(k-1)}, c_{ik}^{(k-1)} + c_{kj}^{(k-1)}\}$$



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

Pseudocode for Floyd-Warshall

```

for  $k \leftarrow 1$  to  $n$ 
  do for  $i \leftarrow 1$  to  $n$ 
    do for  $j \leftarrow 1$  to  $n$ 
      do if  $c_{ij} > c_{ik} + c_{kj}$ 
        then  $c_{ij} \leftarrow c_{ik} + c_{kj}$  } relaxation
    
```

Notes:

- Okay to omit superscripts, since extra relaxations can't hurt.
- Runs in $\Theta(n^3)$ time.
- Simple to code.
- Efficient in practice.

Transitive closure of a directed graph

Compute $t_{ij} = \begin{cases} 1 & \text{if there exists a path from } i \text{ to } j, \\ 0 & \text{otherwise.} \end{cases}$

IDEA: Use Floyd-Warshall, but with (\vee, \wedge) instead of $(\min, +)$:

$$t_{ij}^{(k)} = t_{ij}^{(k-1)} \vee (t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)}).$$

Time = $\Theta(n^3)$.

Graph reweighting

Theorem. Given a label $h(v)$ for each $v \in V$, *reweight* each edge $(u, v) \in E$ by

$$\hat{w}(u, v) = w(u, v) + h(u) - h(v).$$

Then, all paths between the same two vertices are reweighted by the same amount.

Proof. Let $p = v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$ be a path in the graph.

Then, we have

$$\begin{aligned}\hat{w}(p) &= \sum_{i=1}^{k-1} \hat{w}(v_i, v_{i+1}) \\ &= \sum_{i=1}^{k-1} (w(v_i, v_{i+1}) + h(v_i) - h(v_{i+1})) \\ &= \sum_{i=1}^{k-1} w(v_i, v_{i+1}) + h(v_1) - h(v_k) \\ &= w(p) + h(v_1) - h(v_k). \quad \square\end{aligned}$$

Johnson's algorithm

1. Find a vertex labeling h such that $\hat{w}(u, v) \geq 0$ for all $(u, v) \in E$ by using Bellman-Ford to solve the difference constraints

$$h(v) - h(u) \leq w(u, v),$$

or determine that a negative-weight cycle exists.

- Time = $O(VE)$.
2. Run Dijkstra's algorithm from each vertex using \hat{w} .
 - Time = $O(VE + V^2 \lg V)$.
 3. Reweight each shortest-path length $\hat{w}(p)$ to produce the shortest-path lengths $w(p)$ of the original graph.
 - Time = $O(V^2)$.

Total time = $O(VE + V^2 \lg V)$.