

12.0

Last week we gave Kruskal algorithms by Kruskal and Prim for constructing a MST

We then started discussing Single Source Shortest Path problem in which we are given a weighted directed graph an a vertex as input and we have to compute, for each vertex, the length of the shortest directed path from the input vertex

Important observation:

Optimal substructure: If $p = \langle v_0, \dots, v_k \rangle$, $s = v_0$, is a shortest path from v_0 to v_k then $\forall 0 \leq i < j \leq k$ $\langle v_i, \dots, v_j \rangle$ is a shortest path from v_i to v_j .

Negative cycles: If \exists a negative cycle then no solution

Relax(u, v, w):

1. If $v.d > u.d + w(u, v)$
2. $v.d = u.d + w(u, v)$
3. $v.\pi = u$

Initialize-Single-Source(G, s)

1. For each $v \in G.V$
2. $v.d = \infty$, $v.\pi = \text{NIL}$
3. $s.d = 0$

Bellman-Ford alg: Bellman-Ford(G, w, s)

1. Initialize-Single-Source(G, s)
2. For $i = 1$ to $|G.V| - 1$
3. For each $(u, v) \in G.E$
4. Relax(u, v, w)
5. For each edge $(u, v) \in E$
6. If $v.d > u.d + w(u, v)$
7. Return False (neg. cycle exist).
8. Return True

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Worst Run time is $\mathcal{O}(V \cdot E)$

Analysis

Main Lemma:

If $p = \langle v_0, \dots, v_k \rangle$ is a shortest path from v_0 to v_k and we relax edges of p in the order $(v_0, v_1), \dots, (v_{k-1}, v_k)$ then $v_k.d = \delta(v_0, v_k)$ (the shortest length of the shortest path). This remains true even if other relaxation steps occur. (Assuming no negative cycles).

Pf:

By induction on k . For $k=0$ (trivial) this is true. Also easy to see for $k=1$.

For $k > 1$ by induction $\langle v_0, \dots, v_{k-1} \rangle$ is a s.p. and by ind. hyp. $v_{k-1}.d = \delta(v_0, v_{k-1})$.

Regardless of other relaxations, when we relax (v_{k-1}, v_k) we get that $v_k.d \leq v_{k-1}.d + w(v_{k-1}, v_k) = \delta(v_0, v_k)$.

Comment: we should have also argued by induction that $v_i.d \geq \delta(v_0, v_i)$ for all i .

Corollary: if no negative cycles, since each path of length $\leq n-1$ after $n-1$ steps of relaxing all edges all distances are computed correctly.

Also need to show that we get a tree of shortest paths.

Corollary: if neg cycle exists we will catch it!

Pf: if cycle is $\langle v_0, \dots, v_k \rangle$ then $\sum_{i=0}^{k-1} (v_i, v_{i+1}) < 0$.

if we fail to find it in lines 5-7 then

$$v_i.d = v_{i-1}.d + w(v_{i-1}, v_i)$$

$$\Rightarrow \sum_{i=1}^k v_i.d \geq \sum_{i=1}^k v_{i-1}.d + \sum_{i=1}^k w(v_{i-1}, v_i)$$

As $v_0 = v_k$ we get a contradiction

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Next we are going to see another classical alg. due to Dijkstra in ^{the} case where all the weights are non-negative.

The basic idea is at each step to resolve the distance to one vertex and adjust the rest of $u.d$ accordingly.

For example at the first step we are assumed that for some neighbor of s , v
 $v.d = \delta(s, v) = w(s, v)$ etc.

We shall use a min-priority-queue again, sorted according to $v.d$.

Dijkstra(G, s, w)

1. Initialize-Single-Source(G, s)

2. ~~$Q = G.V$~~ min-priority-queue

3. while $Q \neq \emptyset$

4. $u = \text{Extract-min}(Q)$

5. For each $v \in G.\text{Adj}[u]$

6. $\text{Relax}(u, v, w)$

The running time is $O((|E| + |V|) \cdot \lg |V|)$. Indeed, we need $O(|V|)$ at line 2 and for line 3. For each edge we have to decrease key which costs $O(\lg |V|)$. For each v we also have Extract-min which costs $O(\lg |V|)$.

Analysis: At each iteration, all vertices in $G.V \setminus Q$ satisfy $v.d = \delta(s, v)$.
~~and for each $v \in Q$, $v.d$ is its minimal distance to $G.V \setminus Q$~~
~~(distance measured by nearest neighbor).~~

Since at the end $Q = \emptyset$ this ~~proves~~ gives correctness.

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Proof: The invariant holds after Relax iteration.

Assume it does not hold ~~at some point~~ and let u be first such vertex.

By assumption, for all $v \in Q$ $v.d = \delta(s, v)$.

~~A minimal path from s to u~~ Consider the minimal path from s to u and let y be the first vertex there from Q (maybe $y = u$).

Let x be y 's predecessor on that path. Thus the path is $s \rightsquigarrow x - y \rightsquigarrow u$.

We now note that we must have $y.d = \delta(s, y)$ as we relaxed (x, y) after ~~not~~ removing x from Q . (and by assumption $x.d = \delta(s, x)$).

By optimal-substructure $\delta(s, u) \geq \delta(s, y)$, but also $u.d \geq \delta(s, u)$ (as before)

hence $u.d \geq \delta(s, u) \geq \delta(s, y) = y.d \geq u.d$

Thus, we only have equalities and hence $u.d = \delta(s, u)$ and no contradiction. \square

All pairs shortest paths (APSP).

This is the problem of finding shortest paths between every pair of vertices in a directed graph.

Comments:

If graph has more than $(6 \cdot V^2 / 5) \cdot V$ many edges, we can just keep all into in an array (identifying vertices with the numbers $1, 2, \dots, (6 \cdot V)$).

Extract-min takes $O(V)$ and we get a run time of $O(VV^2 + |E|) = O(VV^2)$

If we use Fibonacci-heap can reduce run time to $O(MV, V \sim 15V)$

All Pairs Shortest Paths (APSP)

The goal is to compute, for every u, v , the shortest path between them ($u \rightsquigarrow v$, $u \rightsquigarrow v$).

We can run the SSSP alg. $|V|$ times to get $O(V^3)$ or $O(VE \log V)$ time alg.

(or $O(V^2 \log V + VE)$ using Fib. heap), if all wts ≥ 0 .

If this is not the case we must run Bellman-Ford and get $O(V^3/E)$.

We will see an alg. that does better.

Here it will be convenient to assume the input is given by adjacency matrix

W , where $W_{ij} = w(i, j)$, where we think of G.V. as $1, 2, \dots, |V|$.

If there is no edge (i, j) we have $w(i, j) = \infty$.

The output will be represented by a matrix D where $D_{ij} = d(i, j)$ and a

matrix Π where $\Pi_{ij} = \text{NIL}$ if $i=j$ or there is no $i \rightsquigarrow j$ path and Π_{ij} is j 's predecessor in a shortest $i \rightsquigarrow j$ path.

We will present the Floyd-Warshall alg.

The idea is for every (i, j) and k to compute the s.p. from i to j that only goes through vertices in $\{1, \dots, k\}$. Then we increase k until we get $k=n$.

The reasoning is that if p is a minimal path $i \rightsquigarrow j$ using only $\{1, \dots, k\}$

then it must be of the form $i \xrightarrow{k} \dots \xrightarrow{k} j$ or $i \xrightarrow{k} \dots \xrightarrow{k} k \xrightarrow{k} \dots \xrightarrow{k} j$.

Thus, it can be solved by using the information computed so far.

Formal:

$$d_{ij}^{(k)} = \begin{cases} w_{ij} & k=0 \\ \min(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}) & k>0 \end{cases}$$

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Floyd-Warshall (W)

1. $n = W.rows$
2. $D^{(0)} = W$
3. For $k=1$ to n
4. Let $D^{(k)} = (d_{ij}^{(k)})$ be a new matrix
5. For $i=1$ to n
6. For $j=1$ to n
7. $d_{ij}^{(k)} = \min(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)})$
8. Return $D^{(n)}$

Run time: $O(n^3) = O(|V|^3)$

How to compute π ?

Notice that in the k 'th step we should take

$$\pi_{ij}^{(k)} = \begin{cases} \pi_{ij}^{(k-1)} & \text{if min is } d_{ij}^{(k-1)} \\ \pi_{kj}^{(k-1)} & \text{if min is } d_{ik}^{(k-1)} + d_{kj}^{(k-1)} \end{cases}$$

Simple to incorporate to alg.

~~Consider the following~~