

Bellman-Ford algorithm for Single-Source Shortest Path

Given $G = (V, E)$ and source $s \in V$. Find shortest distance from s to every reachable node $v \in V$. Compute $dist[v]$ for every $v \in V$. Also compute $parent[v]$ for all $v \in V$, where $parent[v]$ is the predecessor of v on a shortest path from s to v .

```

1 Function Initialize-Single-Source ( $G, s$ )
2   for  $v \in V$  do
3      $dist[v] \leftarrow \infty$ 
4      $parent[v] \leftarrow NIL$ 
5      $dist[v] \leftarrow 0$ 
6 Function Relax ( $u, v, w$ )
7   if  $dist[v] > dist[u] + w(u, v)$  then
8      $dist[v] \leftarrow dist[u] + w(u, v)$ 
9      $parent[v] \leftarrow u$ 
10 Function Bellman-Ford ( $G, s, w$ )
11   Initialize-Single-Source ( $G, s$ )
12   for  $i = 1$  to  $|V| - 1$  do
13     for  $(u, v) \in E$  do
14       Relax ( $u, v, w$ )
15   for  $(u, v) \in E$  do
16     if  $dist[v] > dist[u] + w(u, v)$  then
17       return False
18   return True

```

Proposition. *Properties of shortest paths of relaxation*

1. **Triangular Inequality**

$$\delta(s, v) \leq \delta(s, u) + w(u, v) \quad \text{for all } (u, v) \in E$$

2. **Upper bound property**

$$dist[v] \geq \delta(s, v) \quad \text{for all } v \in V$$

Implies if $dist[v] = \delta(s, v)$ then $dist[v]$ stays the same forever

3. **No-path property** *If there is no path from s to v then $dist[v] = \delta(s, v) = \infty$*

4. **Convergence property** *If $s \xrightarrow{p} u \rightarrow v$ is a shortest path from s to v and $dist[v] = \delta(s, u)$, then if you relax the edge (u, v) , you obtain $dist[v] = \delta(s, v)$*

5. **Path-Relaxation property** *If $p = \langle v_0, v_1, \dots, v_k \rangle$ is a shortest path from $v_0 = s$ to v_k , and we relax the edges $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ in that order, then*

$$dist[v_k] = \delta(s, v_k)$$

1. **Complexity** $\Theta(|V||E|) = \Theta(|V|^3)$
2. **Discussion** works well on negative weights as well. Only restriction is there cannot be any negative cycle.

Dijkstra's Algorithm for Shortest Path

Given $G = (V, E)$ and source $s \in V$ and weight $w : E \rightarrow \mathbb{R}^+$, nonnegative. Find shortest distance from s to every reachable node $v \in V$. Compute $dist[v]$ for every $v \in V$. Also compute $parent[v]$ for all $v \in V$, where $parent[v]$ is the predecessor of v on a shortest path from s to v . Let $S \subseteq V$ be set of all vertices where shortest distance from s has already been found. At any iteration, we have a set S whose shortest distances from s we know, and we have $V \setminus S$

```

1 Function Initialize-Single-Source ( $G, s$ )
2   for  $v \in V$  do
3      $dist[v] \leftarrow \infty$ 
4      $parent[v] \leftarrow NIL$ 
5    $dist[s] \leftarrow 0$ 
6 Function Relax ( $u, v, w$ )
7   if  $dist[v] > dist[u] + w(u, v)$  then
8      $dist[v] \leftarrow dist[u] + w(u, v)$ 
9      $parent[v] \leftarrow u$ 
10 Function Dijkstra-Shortest-Path ( $G, s, w$ )
11   Initialize-Single-Source ( $G, s$ )
12    $S \leftarrow \emptyset$ 
13    $Q \leftarrow V$ 
14   while  $Q \neq \emptyset$  do
15      $u \leftarrow \text{Extract-Min}(Q)$ 
16      $S \leftarrow S \cup \{u\}$ 
17     for  $v \in Adj[u]$  do
18       Relax( $u, v, w$ )

```

Complexity

1. $|V|$ iterations for while loop, EXTRACT-MIN takes $O(\lg |V|)$ each time
2. Inner for loop runs for $|E|$ times in total
3. In summary, if Q implemented with,

(a) **array** $\Theta(|V|^2)$

- (b) **binary heap** $\Theta(|E| \lg |V| + |V| \lg |V|) = \Theta(|E| \lg |V|)$ because in relax step in worst case every vertices has $dist$ updated which takes $O(\lg |V|)$ each time for a total of $|E|$ vertices
- (c) **Fibonacci heap** $\Theta(|V| \lg |V| + |E|)$

Proof of correctness

Proposition. *At the start of each iteration in the while loop, we have*

$$dist[v] = \delta(s, v) \quad \text{for all } v \in S$$

Proof. Proof by contradiction. Assume there is a first vertex u such that $u \in S$ and $dist[u] > \delta(s, u)$. Note $u \neq s$ (since it is $dist[s] = 0$ is shortest) and $\delta[u] \neq \infty$ (i.e. reachable from s). In particular there is a shortest path $s \xrightarrow{p} u$ in S . Consider $s \xrightarrow{p_1} x \rightarrow y \xrightarrow{p_2} u$. Note here $x \in S$, $y \in Q$, and all vertices in p_1 are in S . Since $x \in S$ and u is the first vertex to defy the condition we have $dist[x] = \delta(s, x)$. When we added x to S we relaxed the edge (x, y) . By convergence property we have $dist[y] = \delta(s, y)$. Since y occurs before u in that path $\delta(s, y) \leq \delta(s, u)$. By upper-bound property, we have $\delta(s, u) \leq dist[u]$.

$$dist[y] = \delta(s, y) \leq \delta(s, u) \leq dist[u]$$

Since $y, u \in Q$ and we chose u to add to S , so

$$dist[y] \geq dist[u]$$

Then in particular

$$dist[y] = \delta(s, y) = \delta(s, u) = dist[u]$$

The latte equality contradicts with assumption. □

All-pairs Shortest Path

Given $G = (V, E)$ and weight w . Want to find $dist[i, j]$ which is the shortest distance between vertices $v_i, v_j \in V$.

```

1 Function All-Pairs ( $G, w$ )
2   for  $s \in V$  do
3     Single-Source( $G, w, s$ )

```

Complexity

1. Non-negative weights: Use *Dijkstra* for single source

$$\Theta(|V|^2 \lg |V| + |E||V|) \leq \Theta(|V|^3)$$

2. Potentially negative weights: Use Bellman-Ford for single source

$$\Theta(|V|^2 |E|) \leq \Theta(|V|^4)$$

Dynamic programming approach

1. **Optimal substructure** Let l_{ij}^m be shortest distance from i to j using at most m edges

$$l_{ij}^{(m)} = \text{Min}\{l_{ij}^{(m)}, \text{Min}_{1 \leq k \leq |V|} \{l_{ik}^{(m+1)} + w_{kj}\}\}$$

where

$$l_{ij}^{(0)} = \begin{cases} 0 & i = j \\ \infty & \text{otherwise} \end{cases}$$

2. **bottom up approach**

```

1 Function Extend-Shortest-Path ( $L, w$ )
2    $n \leftarrow$  number of rows of  $L$ 
3    $L' = (l'_{ij}) \leftarrow n \times n$  matrix
4   for  $i = 1$  to  $|V|$  do
5     for  $j = 1$  to  $|V|$  do
6        $l'_{ij} \leftarrow \infty$ 
7       for  $k = 1$  to  $|V|$  do
8          $l'_{ij} \leftarrow \{l'_{ij}, l_{jk} + w_{kj}\}$ 
9   return  $L'$ 

```

Complexity is $\Theta(|V|^3)$

another approach

1. **Optimal substructure** Let $d_{ij}^{(k)}$ is length of a shortest path $i \xrightarrow{p} j$ where for all $v \in p$ we have $v \subseteq \{1, \dots, k\}$

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

So if k is on the shortest path then $d_{ij}^{(k)} = d_{ij}^{(k-1)}$, otherwise it is summation of shortest distance with vertices up to but no including k where

$$d_{ij}^{(0)} = \begin{cases} w_{ij} & (i, j) \in E \\ \infty & \text{otherwise} \end{cases}$$

2. Algorithm

```
1 Function Floyd-Warshall ( $W$ )
2    $n \leftarrow$  number of rows of  $W$ 
3    $D^{(0)} \leftarrow W$ 
4   for  $k = 1$  to  $|V|$  do
5      $D^{(k)} = (d_{ij}^{(k)}) \leftarrow$  be  $n \times n$  matrix
6     for  $i = 1$  to  $|V|$  do
7       for  $j = 1$  to  $|V|$  do
8          $d_{ij}^{(k)} \leftarrow \text{Min}\{d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\}$ 
9   return  $D$ 
```

Transitive closure of G

Given $W = (w_{ij})_{n \times n}$ where w_{ij} is weight of edge (i, j) . Transitive closure is a matrix consisting of 1 and 0, where 1 represent if there is path from i to j , whereas 0 represent if there is i is not reachable from j . One approach, compute W^n with $\Theta(n^3 \lg n)$

1. optimal substructure let

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

where

$$t_{ij}^{(0)} = \begin{cases} 1 & i = j \text{ or } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$