

Shortest paths

- ▶ Generalization of BFS to handle weighted graphs
- ▶ Directed graph $G = (V, E)$,
- ▶ Weight function $w : E \longrightarrow \mathbf{R}$
- ▶ Weight of path $p = v_0 \rightarrow v_1 \rightarrow \cdots \rightarrow v_k$:

$$w(p) = \sum_{i=1}^k w(v_{i-1}, v_i)$$

- ▶ Shortest-path weight $u \rightsquigarrow v$

$$\delta(u, v) = \begin{cases} \min\{w(p) : u \rightsquigarrow v\} & \text{if there exists a path } u \rightsquigarrow v \\ \infty & \text{otherwise} \end{cases}$$

- ▶ Shortest-path $u \rightsquigarrow v$
any path p such that $w(p) = \delta(u, v)$

Shortest paths

Single-source shortest path problem (SSSP): find shortest-paths from a given source vertex $s \in V$ to every vertex $v \in V$

Variants:

- ▶ **Single-destination:** find shortest-paths to a given destination vertex
(*reverse the direction of each edge to become the single-source problem*)
- ▶ **Single-pair:** find shortest-path from u to v
(*no way know that's better in worst case than solving single-source*)
- ▶ **All-pairs:** find shortest-paths from u to v for all $u, v \in V$.
(*skip, if interested, see algorithms in Chapter 25 of CLRS, 3ed*)

Shortest paths

Negative-weight edges and well-definedness

- ▶ Negative-weight edges are OK, as long as **no negative-weight cycles** reachable from the source.
... otherwise, can always get a shorter path by going around the cycle again.
- ▶ The shortest path problem is **ill-posed** in graph with negative-weight cycle
- ▶ Bellman-Ford algorithm can detect and report the existence of negative-weight cycle

Shortest paths

- ▶ **Optimal substructure property:**
subpaths of shortest-paths are shortest-paths.

Proof. If some subpath were not a shortest path, could substitute it and create a shorter total path.

Thus, will see greedy and dynamical programming algorithms.

Shortest paths

- ▶ Notation: $d[v]$: shortest-path estimate
 $\pi[v]$: predecessor of v

- ▶ **Output** of SSSP algorithms

$d[v] = \delta(s, v) = \text{shortest-path weight } s \rightsquigarrow v$

$\pi[v] = \text{predecessor of } v \text{ on a shortest path from } s.$

Shortest paths

Two key components of shortest-path algorithms

- Initialization

```
for every vertex  $v$  in  $V$ 
     $d[v] = \text{infty}$ 
     $\text{pi}[v] = \text{nil}$ 
endfor
 $d[s] = 0$       //  $s$  = source vertex
```

- Relaxing an edge (u, v) : can we improve the shortest-path estimate $d[v]$ by going through u and taking the edge (u, v) ?

```
if  $d[v] > d[u] + w(u, v)$ 
     $d[v] = d[u] + w(u, v)$ 
     $\text{pi}[v] = u$ 
endif
```

Shortest paths

Basic properties:

1. Triangular inequality

for all $(u, v) \in E$, $\delta(u, v) \leq \delta(u, x) + \delta(x, v)$

2. Upper-bound property

Always have $d[v] \geq \delta(s, v)$ for all v .

Once $d[v] = \delta(s, v)$, it never changes

3. No-path property

If $\delta(s, v) = \infty$, then $d[v] = \infty$ always

4. Convergence property

If $s \rightsquigarrow u \rightarrow v$ is a shortest-path, and $d[u] = \delta(s, u)$. Then after “Relax $u \rightarrow v$ ”, $d[v] = \delta(s, v)$

5. Path relaxation property

Let $p = v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_k$ be a shortest-path. If we relax in order, $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$, even intermixed with other relaxations, then $d[v_k] = \delta(v_0, v_k)$