# Lecture 7 Single-Source Shortest Paths Problems

- 1. Related Notions
- 2. Variants of Shortest-Paths Problems
- 3. Properties of Shortest-Paths and Relaxation
- 4. The Bellman-Ford Algorithm

#### **Shortest Paths—Notions**

- 1. Given a weighted, directed graph G = (V, E; W), the weight of path  $p = \langle v_0, v_1, ..., v_k \rangle$  is the sum of the weights of its constituent edges, that is,  $w(p) = \sum_{i=1}^{k} w(v_{i-1}, v_i)$
- 2. The shortest-path weight from u to v is defined as

$$\delta(u,v) = \begin{cases} \min\{w(p) : u \xrightarrow{p} v\} & \text{if there is a path from } u \text{ to } v, \\ \infty & \text{otherwise.} \end{cases}$$

3. A **shortest path** from vertex u to v is any path p with weight  $w(p) = \delta(u, v)$ .

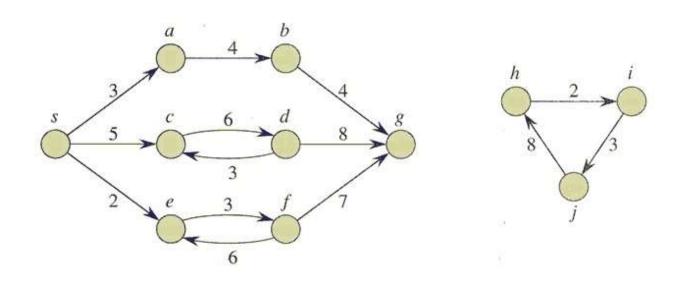
## Single-source shortest path problem

- Input: A weighted directed graph G=(V, E; W) and a source vertex s.
- Output: Shortest-path weight from s to each vertex v in V, and a shortest path from s to each vertex v in V if v is reachable from s.

## Variants of Single-source shortest-path problem

- Single-source shortest-paths problem
- Single-destination shortest-paths problem
  - --- reverse the direction of edge
- Single-pair shortest-path problem
  - --- consider u as source vertex
- All-pairs shortest-paths problem
  - --- topic of next chapter

## Shortest paths – an example



Please caculate  $\delta(s,a)$ ,  $\delta(s,b)$ , ...

## Optimal Substructure of Shortest Paths

- Lemma 24.1 (Subpaths of shortest paths are shortest paths)
- Given a weighted, directed graph G = (V, E; W), let  $p = \langle v_1, v_2, ..., v_k \rangle$  be a shortest path from vertex  $v_1$  to vertex  $v_k$  and, for any i and j such that  $1 \le i \le j \le k$ , let  $p_{ij} = \langle v_i, v_{i+1}, ..., v_j \rangle$  be the subpath of p from vertex  $v_i$  to vertex  $v_j$ . Then,  $p_{ij}$  is a shortest path from  $v_i$  to  $v_j$ .

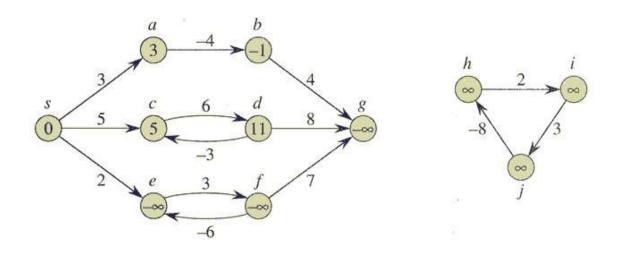
#### **Proof**

• The path p can be decomposed into

$$v_1 \xrightarrow{p_{1i}} v_i \xrightarrow{p_{ij}} v_j \xrightarrow{p_{jk}} v_k$$

- Then we have  $w(p)=w(p_{1i})+w(p_{ij})+w(p_{jk})$ .
- Assume that there is a path  $P_{ij}$  from  $v_i$  to  $v_j$  with weight  $w(p_{ij}) < w(p_{ij})$ .
- Then  $v_1 \xrightarrow{p_{1i}} v_i \xrightarrow{p_{ij}} v_j \xrightarrow{p_{jk}} v_k$  is a path from  $v_1$  to  $v_k$  whose weight is less than w(p), which contradicts the assumption that p is a shortest path from  $v_1$  to  $v_k$ .

## Negative-weight edge and negative-weight cycle



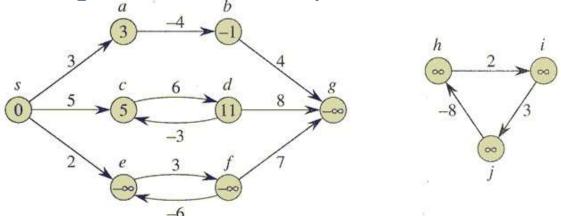
Please caculate  $\delta(s,a)$ ,  $\delta(s,b)$ ,  $\delta(s,c)$ ,...

If there is a negative-weight cycle on some path from s to v, we define  $\delta(s, v) = -\infty$ .

## Cycles

Three typies of cycle: negative-weight cycle, positive-weight cycle, 0-weight cycle

Can a shortest path contains a cycle?



We can assume that when we are finding shortest paths, they are cycle-free. Any acyclic path contains at most |V| distinct vertices, it also contains at most |V|-1 edges. Thus, we can restrict our attention to shortest paths of at most |V|-1 edges.

### Representing Shortest-Paths

- Given a graph G = (V, E), maintain for each vertex  $v \in V$  a **predecessor**  $\pi[v]$  that is either another vertex or NIL.
- Given a vertex v for which  $\pi[v] \neq \text{NIL}$ , the procedure PRINT-PATH(G, s, v) prints a shortest path from s to v.

```
PRINT-PATH(G, s, v)

1 if v = s

2 then print s

3 else if \pi[v] = \text{NIL}

4 then print "no path from" s "to" v "exists"

5 else PRINT-PATH(G, s, \pi[v])

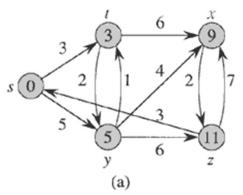
6 print v
```

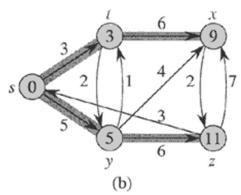
### Predecessor subgraph

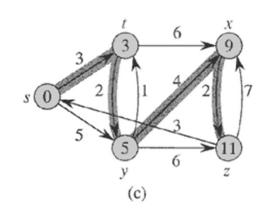
- Predecessor subgraph  $G_{\pi} = (V_{\pi}, E_{\pi})$ :
  - $V_{\pi} = \{ v \in V : \pi[v] \neq \text{NIL} \} \cup \{ s \}$
  - $E_{\pi} = \{ (\pi[v], v) \in E: v \in V_{\pi} \{s\} \}$
- We shall prove that  $\pi$  values produced by the algorithms in this chapter have the property that at termination  $G_{\pi}$  is a "shortest-paths tree"

#### **Shortest-Paths Tree**

- Let G = (V, E; W) be a weighted, directed graph, and assume that G contains no negative cycles that are reachable from s. A **shortest-paths tree** rooted at s is a directed sub-graph G' = (V', E'), where  $V' \subseteq V$  and  $E' \subseteq E$ , such that
  - V' is the set of vertices reachable from s in G,
  - G' forms a rooted tree with root s, and
  - For all  $v \in V'$ , the unique simple path from s to v in G' is a shortest path from s to v in G.
- The shortest-path tree may not be unique.







## The technique of Relaxation

- In our shortest-paths problem, we maintain an attribute d[v], which is an upper bound on the weight of a shortest path from source s to v.
- We call d[v] a shortest-path estimate. i.e., d[v] is an estimate of  $\delta(s, v)$ .
- The term Relaxation is used here for an operation that tightens an upper bound.

#### Relaxation--Initialization

• The initial estimate of  $\delta(s, v)$  can be given by:

```
INITIALIZE-SINGLE-SOURCE (G, s)
```

```
1 for each vertex v \in V[G]
2 do d[v] \leftarrow \infty
3 \pi[v] \leftarrow \text{NIL}
4 d[s] \leftarrow 0
```

### Relaxation Process

- Relaxing an edge (u, v) consists:
  - Testing whether we can improve the shortest path from s to v found so far by going through u to v, and if so,
  - Updating d[v] and  $\pi[v]$ .
- A relaxation step may either decrease the value of the shortest-path estimate d[v] and update v's predecessor  $\pi[v]$ , or cause no change.

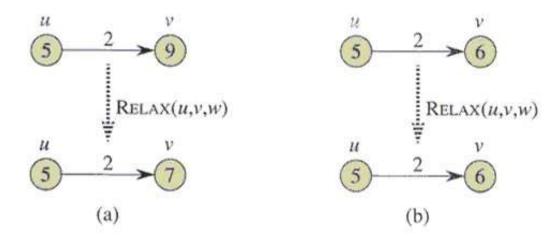
### A Relaxation Step

```
RELAX(u, v, \underline{w})

1 if d[v] > d[u] + w(u, v)

2 then d[v] \leftarrow d[u] + w(u, v)

3 \pi[v] \leftarrow u
```



#### Some notes about Relaxation

#### • Note:

- Each algorithm in this chapter calls INITIALIZE-SINGLE-SOURCE and then repeatedly relaxes edges.
- Relaxation is the only means by which shortest-path estimates d[] and predecessors  $\pi[]$  change.
- The algorithms differ in **how many times** they relax each edge and **the order** in which they relax edges.
- Bellman-Ford algorithm relaxes each edge many times, while Dijkstra's algorithm and the algorithm for directed acyclic graphs relax each edge exactly once.

## Properties of Shortest Paths and Relaxation

- Triangle inequality (Lemma 24.10)
  - For any edge  $(u, v) \in E$ , we have  $\delta(s, v) \leq \delta(s, u) + w(u, v)$
- Upper-bound property (Lemma 24.11)
  - We always have  $d[v] \ge \delta(s, v)$  for all vertices  $v \in V$ , and once d[v] achieves the value  $\delta(s, v)$ , it never changes.
- No-path property (Lemma 24.12)
  - If there is no path from s to v, then we always have  $d[v] = \delta(s, v) = \infty$ .

## Properties of Shortest Paths and Relaxation

#### Convergence property (Lemma 24.14)

• If  $s \sim \to u \to v$  is a shortest path from s to v in G, and if  $d[u] = \delta(s, u)$  at any time prior to relaxing edge (u, v), then  $d[v] = \delta(s, v)$  at all times afterward.

#### Path-relaxation property (Lemma 24.15)

- If  $p = \langle s, v_1, ..., v_k \rangle$  is a shortest path from s to  $v_k$ , and the edges of p are relaxed in the order  $(s, v_1), (v_1, v_2), ..., (v_{k-1}, v_k)$ , then after relaxing all the edges,  $d[v_k] = \delta(s, v_k)$ .
- Note that other relaxations may take place among these relaxations.

## Properties of Shortest Paths and Relaxation

- Predecessor-subgraph property (Lemma 24.17)
  - Once  $d[v] = \delta(s, v)$  for all  $v \in V$ , the predecessor subgraph is a shortest-paths tree rooted at s.

Note: Proofs for all these properties can be found in your textbook.

## Single-source shortest path problem

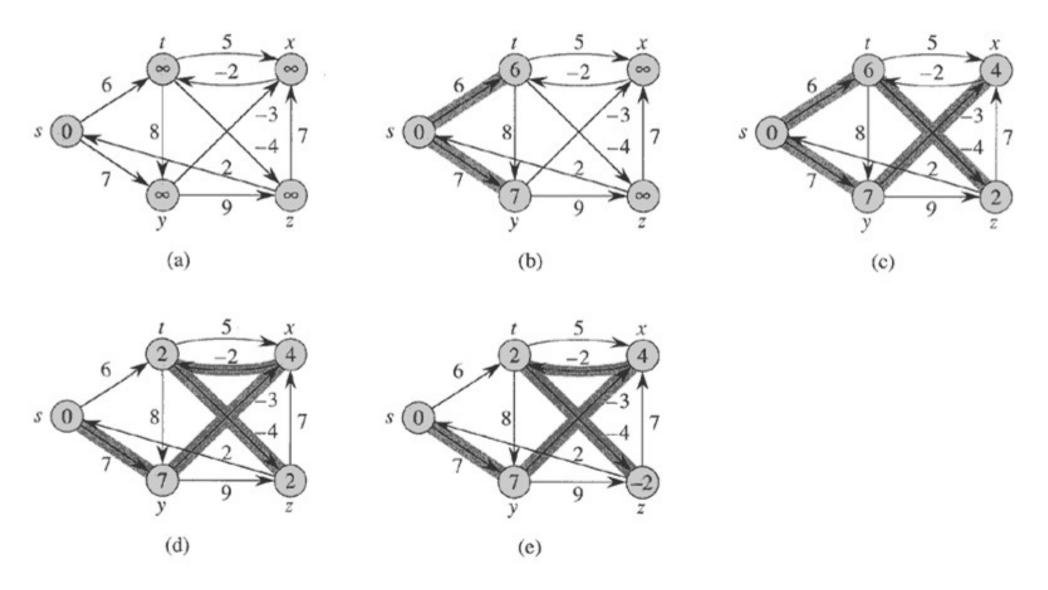
- Input: A weighted directed graph G=(V, E; W) and a source vertex s.
- Output: Shortest-path weight from s to each vertex v in V, and a shortest path from s to each vertex v in V if v is reachable from s.

### The Bellman-Ford Algorithm

- To solve the single-source shortest-paths problem
  - The algorithm returns a **boolean** value indicating whether or not there is a negative-weight cycle that is reachable from the source.
  - If there is a negative-weight cycle reachable from the source *s*, the algorithm indicates no solution exists.
  - If there are no such cycles, the algorithm produces the shortest paths and their weights.

### The Bellman-Ford O(VE)Algorithm

```
BELLMAN-FORD(G, w, s)
   INITIALIZE-SINGLE-SOURCE (G, s)
   for i \leftarrow 1 to |V[G]| - 1
        do for each edge (u, v) \in E[G]
               do RELAX(u, v, w)
   for each edge (u, v) \in E[G]
        do if d[v] > d[u] + w(u, v)
            then return FALSE
   return TRUE
```



Note: each pass relaxes the edges in the order: (t,x), (t,y), (t,z), (x,t), (y,x), (y,z), (z,x), (z,s), (s,t), (s,y)

#### Time Complexity of The Bellman-Ford

```
BELLMAN-FORD (G, w, s)
   INITIALIZE-SINGLE-SOURCE (G, s)
   for i \leftarrow 1 to |V[G]| - 1
        do for each edge (u, v) \in E[G]
               do RELAX(u, v, w)
   for each edge (u, v) \in E[G]
        do if d[v] > d[u] + w(u, v)
            then return FALSE
   return TRUE
```

The time complexity is O(VE)

#### Correctness of The Bellman-Ford

#### • Lemma 24.2

Let G = (V, E) be a weighted, directed graph with source s and weight function  $w: E \rightarrow \mathbb{R}$ , and assume that G contains **no negative cycles that are reachable from** s. Then, after the |V| - 1 iterations of the **for** loop of lines 2-4 of BELLMAN-FORD, we have  $d[v] = \delta(s, v)$  for all vertices v reachable from s.

#### Proof of Lemma 24.2

• Proof: Consider any vertex v that is reachable from s, and let  $p = \langle v_0, v_1, ..., v_k \rangle$ , where  $v_0 = s$ and  $v_k = v$ , be any acyclic shortest path from s to v. Path p has at most | V |-1 edges, and so  $k \le |V|$ -1. Each of the |V|-1 iterations of the **for** loop of lines 2-4 relaxes all |E| edges. Among the edges relaxed in the *i*th iteration, the edge in P is  $(v_{i-1},$  $v_i$ ), for i = 1, 2, ..., k. By the path-relaxation property,  $d[v] = d[v_k] = \delta(s, v_k) = \delta(s, v)$ .

#### Correctness of The Bellman-Ford

#### Corollary 24.3

- Let G = (V, E) be a weighted, directed graph with source s and weight function  $w: E \rightarrow \mathbb{R}$ . Then for each vertex  $v \in V$ , there is a path from s to v if and only if BELLMAN-FORD terminates with  $d[v] < \infty$  when it is run on G.
- Proof : Exercise 24.1-2.
- ( $\Rightarrow$ )If there is a path P= $\langle v_0, v_1, ..., v_k \rangle$  from s to v, where  $s=v_0$ ,  $v=v_k$ , we show by induction that after the ith iteration,  $d[v_i]<\infty$ .
- ( $\Leftarrow$ ) By the *no-path property*.

#### Correctness of The Bellman-Ford

#### Theorem 24.4

Let BELLMAN-FORD be run on a weighted, directed graph G = (V, E) with source s and weight function w:  $E \rightarrow \mathbf{R}$ . If G contains no negative cycles that are reachable from s, then the algorithm returns TRUE, we have  $d[v] = \delta(s, v)$  for all vertices  $v \in V$ , and the predecessor subgraph  $G_{\pi}$  is a shortest-paths tree rooted at s. If G does contain a negative weight cycle reachable from s, then the algorithm returns FALSE.

#### **Proof of Theorem 24.4**

- 1. Suppose *G* contains no negative-weight cycles that are reachable from *s*.
  - Prove that at termination,  $d[v] = \delta(s, v)$ , for all vertices v in V.
    - $\triangleright$  Case 1: v is reachable from s; --> Lemma~24.2
    - $\triangleright$  Case 2: v is not reachable from s; --> No Path Property
  - Prove that the algorithm returns TRUE For each edge (u,v),  $d[v] = \delta(s, v) \le \delta(s,u) + w(u,v) = d[u] + w(u,v)$ .
- 2. Suppose that *G* contains a negative-weight cycle that is reachable from *s*.
  - Prove that the algorithm returns FALSE For each edge on a negative-weight cycle  $C = \langle v_0, v_1, ..., v_k \rangle$ , where  $v_0 = v_k$   $d[v_i] \le d[v_{i-1}] + w(v_{i-1}, v_i)$ , for  $1 \le i \le k$ .
    - Summing both sides of equations yields  $0 \le w(C)$ , contradiction.

### Conclusion

- Relaxation
- The Bellman-Ford Algorithm

### Homework

• 24.1-1,24.1-6