#### 24.1 The Bellman-Ford Algorithm

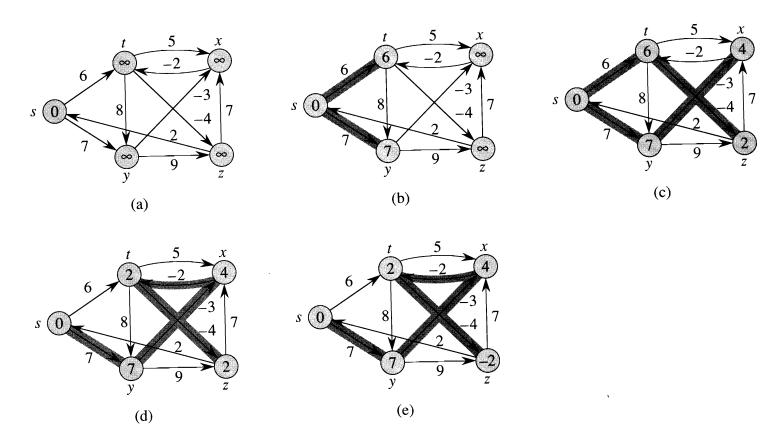
The Bellman-Ford algorithm solves the single-source shortest paths problem in more general settings.

- ➤ Unlike Dijkstra's algorithm, it allows edges of negative length. However, it takes much longer time.
- ➤ Unlike Dijkstra's algorithm that adopts the greedy policy, the Bellman-Ford algorithm adopts Dynamic Programming technique, progressively decreasing the estimate of *v*. *d* the distance from *s* to node *v* until the estimate is precise.

The algorithm returns **true** if and only if the graph does not contain any negative cycles that are reachable from the source.

```
Bellman_Ford(G, w, s)
            s.d \leftarrow 0;
     2 s.\pi \leftarrow NIL;
     3
           for all v \in V \setminus \{s\} do
                 v.d \leftarrow \infty;
     5
                 v.\pi \leftarrow NIL;
           for i \leftarrow 1 to |V| - 1 do
     6
                  for each edge (u, v) \in E do
     8
                       Relax(u, v, w);
            for each edge (u, v) \in E do
     9
                  if v.d > u.d + w(u,v) then /* i.e., (u,v) can still be relaxed */
     10
                       return false
     11
     12
            return true.
```

The running time of algorithm Bellman\_Ford is O(|V||E|).



This example is from page 652 of the textbook. Here, each iteration 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). Figures (b)-(e) show the result after each iteration.

#### **Proof of the correctness.**

As usual with relaxation, v.d can only decrease, and if  $\delta(s,v)$  is defined (there are no negative cycles reachable from the source), we always have  $v.d \geq \delta(s,v)$ . A phase or a pass is one iteration of the **for** loop of lines 6–8, where each edge is relaxed once.

#### **Case (1):** Suppose there is no negative cycle reachable from *s*.

Consider some shortest path  $s \to v_1 \to v_2 \to \cdots \to v_{k-1} \to v_k$ . We can assume  $k \le |V| - 1$ .

- \* When  $(s, v_1)$  is relaxed in the 1st phase,  $v_1$ . d is set to  $\delta(s, v_1)$  if it isn't already.
- \* When  $(v_1, v_2)$  is relaxed in the 2nd phase,  $v_2$ . d is set to  $\delta(s, v_2)$  if it isn't already.
- \* • •
- \* When  $(v_{k-1}, v_k)$  is relaxed in the kth phase,  $v_k$ . d is set to  $\delta(s, v_k)$  if it isn't already.

So,  $v \cdot d = \delta(s, v)$  for all v after |V| - 1 phases, and no edges are still relaxable.

**Case (2):** Suppose there is a negative cycle reachable from s.

Say the cycle is  $v_0 \rightarrow v_1 \rightarrow v_2 \rightarrow \cdots \rightarrow v_k = v_0$ .

Consider the situation after |V|-1 phases. Note that at this point, all the v.d values of vertices in the cycle are **finite**.

By contradiction: suppose that none of the edges on the cycle are relaxable. That is,

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

Summing this inequality over i = 1, ..., k, we find a contradiction since the sum of  $w(v_{i-1}, v_i)$  is negative by the assumption.

Therefore, some edge of the negative cycle is still relaxable.

# **24.8 Special Shortest Paths Problems**

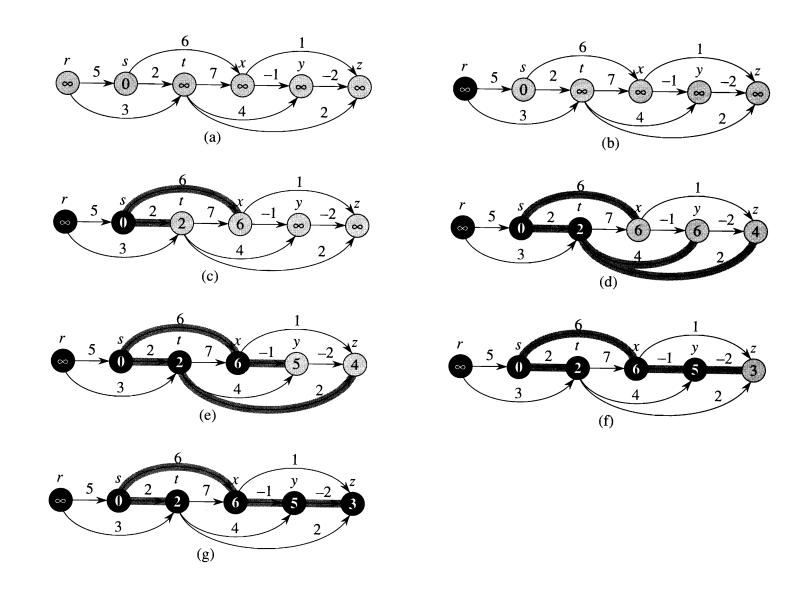
- > SSP in a DAG
- > Special linear programming

## 24.2 Single-Source Shortest Paths in DAGs

For a directed acyclic graph (DAG), we can relax the edges according to the topological order of their start vertices, from left to right.

```
DAG\_Shortest\_Paths(G, w, s)
            s.d \leftarrow 0;
           s.\pi \leftarrow NIL;
            for all v \in V - \{s\} do
     3
     4
                v.d \leftarrow \infty;
                 v.\pi \leftarrow NIL;
     5
            determine the topological order of each vertex v \in V, using the DFS technique;
     6
            for each vertex u in increasing topological order do
     8
                 for v \in G. Adj[u] do
     9
                      Relax(u, v, w).
```

The time complexity of algorithm DAG\_Shortest\_Paths is O(|V| + |E|).



This example is from page 656 of our textbook.

#### 24.2 Shortest Paths in DAGs (continued)

#### **Proof of the correctness.**

Consider any shortest path  $s \to v_1 \to v_2 \to \cdots \to v_{k-1} \to v_k$ . The algorithm relaxes the edges from left to right.

- ▶ When  $(s, v_1)$  is relaxed,  $v_1.d$  is set to  $\delta(s, v_1)$  (note that it was  $\infty$  before this point).
- $\blacktriangleright$  When  $(v_1, v_2)$  is relaxed,  $v_2$ . d is set to  $\delta(s, v_2)$  if it isn't already.
- **>** ...
- When  $(v_{k-1}, v_k)$  is relaxed,  $v_k$ . d is set to  $\delta(s, v_k)$  if it isn't already.

So  $v.d = \delta(s, v)$  for all v by the time all edges are relaxed.

#### 24.4 Difference constraints

Suppose we have to schedule n tasks  $T_1, T_2, \ldots, T_n$ , and we have a set of constraints like these:

 $T_3$  must be done at least 15 minutes after  $T_7$ 

 $T_2$  must be done before  $T_9$ 

 $T_2$  must be done at least 5 minutes before  $T_4$ 

 $T_5$  must be done at most 10 minutes after  $T_1$ 

: We wish to know if this arrangement is possible, and if so, find a schedule.

If  $T_i$  is scheduled at time  $x_i$ , then the above constraints can be written as:

$$x_7 - x_3 \le -15$$

$$x_9 - x_2 \le 0$$

$$x_2 - x_4 \le -5$$

$$x_5 - x_1 \le 10$$

## 24.4 Difference constraints (continued)

In general, we have real variables  $x_1, x_2, \dots, x_n$ , and some numbers of constraints of the form  $x_i - x_i \le b_k$ .

We will define a weighted graph G = (V, E, w), called the constraint graph.

There are n + 1 vertices  $V = \{v_0, v_1, v_2, ..., v_n\}$ .

There is a directed edge  $(v_0, v_i)$  of length 0 from  $v_0$  to  $v_i$  for all i with i = 1, 2, ..., n.

For each constraint  $x_j - x_i \le b_k$ , there is a directed edge  $(v_i, v_j)$  from  $v_i$  to  $v_j$  with length  $b_k$ .

**Interesting fact:** If the constraint graph has a negative cycle, there is no solution. Otherwise, an example of a solution is

$$x_i = \delta(v_0, v_i)$$
 for  $i = 1, 2, ..., n$ .

#### Is the solution is unique?

### 24.4 Example system of difference constraints

$$x_1 - x_2 \le 0,$$
 (1)  
 $x_1 - x_5 \le -1,$  (2)  
 $x_2 - x_5 \le 1,$  (3)  
 $x_3 - x_1 \le 5,$  (4)  
 $x_4 - x_1 \le 4,$  (5)  
 $x_4 - x_3 \le -1,$  (6)  
 $x_5 - x_3 \le -3,$  (7)  
 $x_5 - x_4 \le -3.$  (8)

# 24.4 Example constraint graph and solution

