

①

## (24) Single Source Shortest Paths

Let  $G = (V, E)$  be a ~~WEIGHTED~~ GRAPH with weight function  $w: E \rightarrow \mathbb{R}$ .

Let  $x, y \in V$  and let  $P$  denote an  $x-y$  PATH in  $G$ , i.e. A SEQUENCE

$$P: x = v_0, v_1, \dots, v_k = y$$

where  $(v_{i-1}, v_i) \in E$  for  $i = 1, \dots, k$ .  
The weight of the path  $P$  is the sum of the weights of all the edges in  $P$ .

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

The shortest path weight (or distance) from  $x$  to  $y$  is

$$\delta(x, y) = \begin{cases} \min\{w(p) : p \text{ is an } x-y \text{ PATH}\} & \text{IF such a path exists} \\ \infty & \text{IF NO such path exists} \end{cases}$$

A shortest path from  $x$  to  $y$  is any  $x-y$  PATH  $P$  with  $w(P) = \delta(x, y)$ .

(2)

## Problem: Single Source Shortest Path (SSSP)

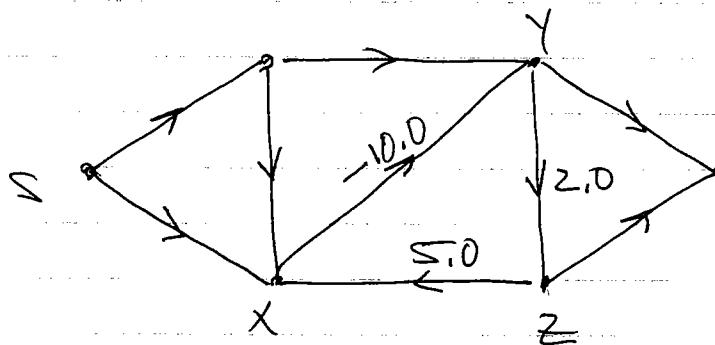
Given a vertex  $s \in V$  (called the source)

Determining a shortest  $s-y$  path ~~exists~~ (if one exists) for all  $y \in V$ .

This Problem makes sense for both Directed and Undirected Graphs. But we will assume throughout this chapter that  $G$  is a Directed Graph.

Note that if  $G$  has a negative weight cycle particular from  $s$ , they take notion of a shortest path from  $s$  to any vertex on that cycle is not well defined.

Ex



cycle :  $x \rightarrow y \rightarrow z \rightarrow x$  weight = -3.0

Given any path from  $s$  to  $y$  say, we can always find a 'shorter' path by traversing this cycle one more time.

~~GIVEN ANY PATH FROM S TO T, WE CAN FIND A "SHORTEST" PATH BY  
TRVERSING P.~~

Thus we will assume NO such cycle exists,  
THOUGH WE ALLOW NEGATIVE EDGE WEIGHT,  
IN GENERAL.

WE CONSIDER TWO ALGORITHMS FOR SOLVING  
THE SSSP PROBLEM

- DIJKSTRA : NEGATIVE EDGES NOT ALLOWED (FASTER)
- BELLMAN-FORD : NEGATIVE EDGES ALLOWED. (SLOWER)

THESE ALGORITHMS UTILIZE THE PARENT, OR  
PREDECESSOR FIELD  $P[u]$  OF A VERTEX  
 $u \in V$ . AS IN BFS & DFS THEY CREATE  
A PREDECESSOR SUBGRAPH  $G_p = (V_p, E_p)$   
WHERE

$$V_p = \{v \in V : P[v] \neq \text{NIL}\} \cup \{s\}$$

$$E_p = \{(P[v], v) : v \in V_p - \{s\}\}$$

A SHORTEST PATH, FROM SOURCE S TO  
 $v \in V$  CAN THEN BE PRINTED BY :

PrintPath( $G, s, v$ ) on p. 538

### Print Path ( $G, s, v$ )

- 1.) If  $v = s$
- 2.) Print  $s$
- 3.) Else If  $p[v] = \text{NIL}$
- 4.) Print "NO PATH from"  $s$  "TO"  $v$  "EXISTS"
- 5.) ELSE
- 6.) PrintPath ( $G, s, p[v]$ )
- 7.) Print  $v$

THE SUBGRAPH  $G_p$  PRODUCED BY DIJKSTRA'S AND BELLMAN-FORD'S ALGORITHMS HAVE THE FOLLOWING PROPERTIES

- (1)  $V_p$  IS THE SET OF VERTICES REACHABLE FROM  $s$ .
- (2)  $G_p$  FORMS A ROOTED TREE WITH ROOT  $s$ .
- (3) FOR ALL  $v \in V_p$ , THE UNIQUE SIMPLE PATH IN  $G_p$  FROM  $s$  TO  $v$  IS A SHORTEST PATH IN  $G$  FROM  $s$  TO  $v$ .

SUCH A SUBGRAPH IS CALLED A SHORTEST-PATHS TREE. NOTE THAT SHORTEST-PATHS ARE NOT NECESSARILY UNIQUE, AND NEITHER ARE SHORTEST-PATHS TREES.

THE KEY TO DIJKSTRA AND BELLMAN-FORD IS A TECHNIQUE CALLED RELAXATION, WE MAINTAIN AN ATTRIBUTE  $d[v]$  FOR EACH VERTEX  $v \in V$ , WHICH IS AN UPPERBOUND ON THE SHORTEST-PATH WEIGHT FROM  $s$  TO  $v$ , i.e.

(\*)

$$d(s, v) \leq d[v]$$

THROUGHOUT THE ALGORITHM'S EXECUTION, AND  $d(s, v) = d[v]$  UPON COMPLETION.

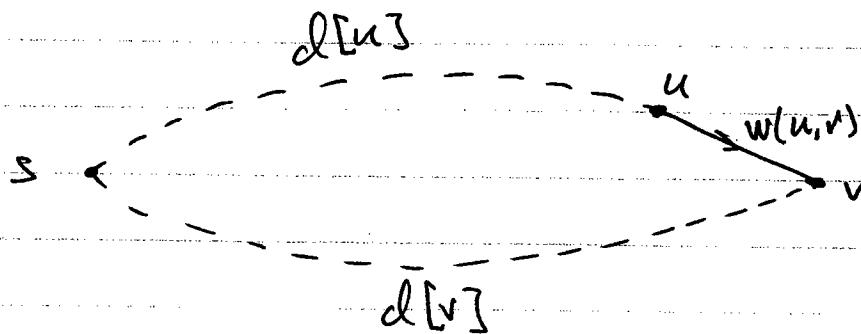
$d[v]$  IS CALLED A SHORTEST-PATH ESTIMATE.  $d[v]$  AND  $p[v]$  ARE INITIALIZED BY THE SUBROUTINE

Initialize ( $G, s$ )

- 1.) For All  $v \in V$
- 2.)  $d[v] \leftarrow \infty$
- 3.)  $p[v] \leftarrow \text{NIL}$
- 4.)  $d[s] \leftarrow 0$

OBSERVE THAT AFTER Initialize IS CALLED,  
(\*) IS CERTAINLY TRUE.

THE PROCESS OF RELAXING AN EDGE  $(u, v)$  CONSISTS OF TESTING WHETHER WE CAN IMPROVE THE SHORTEST PATH FROM  $S$  TO  $v$  FOUND SO FAR BY GOING THROUGH  $u$ , AND IF SO, UPDATING  $P[v]$  AND  $d[v]$  ACCORDINGLY.



Relax  $(u, v)$  (PRE:  $v \in \text{Adj}[u]$ )

- 1.) If  $d[v] > d[u] + w(u, v)$
- 2.)  $d[v] \leftarrow d[u] + w(u, v)$
- 3.)  $P[v] \leftarrow u$

OBSERVE THAT IMMEDIATELY AFTER RELAXING EDGE  $(u, v)$  WE HAVE

$$d[v] \leq d[u] + w(u, v).$$

ALSO NOTE  $\text{Relax}(u, v)$  CHANGES (AT MOST) THE FIELDS OF VERTEX  $v$ , NOT OF  $u$ .

3. (24.5-4)

Let  $G = (V, E)$  be a weighted, directed graph with source vertex  $s$  and let  $G$  be initialized by INITIALIZE-SINGLE-SOURCE( $G, s$ ). Prove that if a sequence of relaxation steps sets  $\pi[s]$  to a non-NIL value, then  $G$  contains a negative-weight cycle.

**Proof:**

RELAX( $u, v$ )

1. if  $d[v] > d[u] + w(u, v)$
2.  $d[v] \leftarrow d[u] + w(u, v)$
3.  $\pi[v] \leftarrow u$

Examination of the above pseudocode for RELAX( $u, v$ ) shows that whenever the algorithm sets the predecessor  $\pi[v]$  for some vertex  $v$ , it also reduces that vertex's d-value  $d[v]$ . Thus if  $\pi[s]$  is at some point set to a non-NIL value, then  $d[s]$  is reduced from its initial value of 0 to a negative number. But  $d[s]$  is necessarily the weight of some existing path from  $s$  to  $s$ , by the lemma below. This path is the required negative weight cycle in  $G$ .

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(1) **Lemma:**

Let  $v \in V[G]$  and suppose that after INITIALIZE-SINGLE-SOURCE( $G, s$ ), some sequence of relaxation steps causes  $d[v]$  to be set to a finite value. Then  $G$  contains an  $s-v$  path of weight  $d[v]$ .

**Proof:**

We use induction on the length  $n$  of the relaxation sequence. If  $n = 0$ , then the only  $d$ -value which is finite is that of the source  $s$ . Indeed, there is a path in  $G$  from  $s$  to  $s$  of weight  $d[s] = 0$ , namely the trivial path, and so the base case is verified.

Let  $n > 0$ , and assume that for any vertex  $u$ , if  $d[u]$  achieves some finite value during a sequence of fewer than  $n$  relaxations, then there exists an  $s-u$  path in  $G$  of weight  $d[u]$ . Now consider a sequence of  $n$  relaxations in which  $d[v]$  becomes finite. Then an edge of the form  $(u, v)$  must have been relaxed during this sequence, and on that relaxation step  $d[v]$  was set to  $d[u] + w(u, v)$ . Since we suppose that this number is finite,  $d[u]$  must have been finite before that relaxation step was executed. Thus  $d[u]$  became finite during a sequence of fewer than  $n$  relaxations, and by our induction hypothesis, there must exist an  $s-u$  path in  $G$  of weight  $d[u]$ . That path, followed by the edge  $(u, v)$ , constitutes a path in  $G$  from  $s$  to  $v$  of weight  $d[v] = d[u] + w(u, v)$ .

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4. (12.2-1)

Suppose that we have numbers between 1 and 1000 in a binary search tree and want to search for the number 363. Which of the following sequences could not be the sequence of nodes examined?

- a. 2, 252, 401, 398, 330, 344, 397, 363.
- b. 924, 220, 911, 244, 898, 258, 362, 363.
- c. 925, 202, 911, 240, 912, 245, 363.
- d. 2, 399, 387, 219, 266, 382, 381, 278, 363.
- e. 935, 278, 347, 621, 299, 392, 358, 363.

(2) LEMMA

AFTER Initialize( $G, s$ ) is EXECUTED,  
 THE INEQUALITY  $\delta(s, v) \leq d[v]$  IS MAINTAINED  
 OVER ANY SEQUENCE OF CALLS TO Relax  
 ON EDGES OF  $G$ .

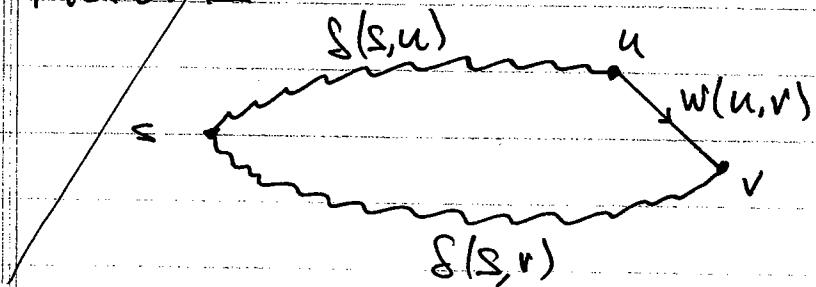
PROOF:

SUPPOSE (TO GET A CONTRADICTION) THAT (\*)  
 BECOMES FALSE DURING A SEQUENCE OF  
 CALLS TO Relax. LET  $v$  BE THE FIRST  
 VERTEX FOR WHICH THE CALL Relax( $u, v$ )  
 CAUSES  $d[v] < \delta(s, v)$  FOR SOME  $u$ .

THEN JUST AFTER THIS CALL

$$\begin{aligned} d[u] + w(u, v) &= d[v] && (\text{WHY Relax}(u, v) \text{ DOES}) \\ &< \delta(s, v) && (\text{OUR ASSUMPTION}) \\ &\leq \delta(s, u) + w(u, v) && (\text{TRIANGLE INEQUALITY}) \end{aligned}$$

THIS LAST INEQUALITY IS CLEAR FROM THE  
 PICTURE



$$\delta(s, v) \leq \delta(s, u) + w(u, v)$$

Proof of ② : Contradiction.

If  $d[v] < \delta(s, v)$  were to become true after some seq. of calls to Relax(), then  $d[v]$  is finite, and by lemma ① there exists an  $s-v$  path in  $G$  of length  $d[v]$ , contradicting the defn of  $\delta(s, v)$  as the largest w.r.t. a shortest  $s-v$  path. 111.

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### Lemma (Path Relaxation Property)

If  $P = (v_0, v_1, \dots, v_k)$  is a shortest path from  $s = v_0$  to  $v_k$ , and the edges of  $P$  are relaxed in the order  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$  then  $d[v_k] = \delta(s, v_k)$ . This property holds regardless of any other relaxation steps that occur, even if they are interleaved with relaxations of the edges of  $P$ .

Proof: (lemma 24.15, p. 609)

Induction on  $k$ , the # of edges in a shortest path.

Thus  $d[u] < f(s, u)$ . But  $\text{Relax}(u, v)$  does nothing to  $u$ , so this inequality must have been true just before the relaxation step. This contradicts our choice of  $v$  as the first vertex for which  $d[v] < f(s, v)$ .

This contradiction shows that  $*$  is minimized over any sequence of relaxation steps.

III

Note that if at some point in such a sequence  $d[v] = f(s, v)$ , then  $d[v]$  never changes, since Relax does not increase  $d$  values.

In an arbitrary sequence of calls to Relax, the equality  $d[v] = f(s, v)$  may never be achieved! The strategy of Dijkstra and Bellman-Ford is to structure the calls to Relax so as to guarantee that  $d[v]$  converges to  $f(s, v)$  for all  $v \in V$ .

The following result is key:

## 24.1 Bellman-Ford

Bellman-Ford( $G, s$ ) Returns A Boolean Value indicating whether or not there is a negative weight cycle in  $G$ . Relaxation from  $s$ , and hence whether a solution to SLP is possible.

The Algorithm Returns True iff SLP is Solvable from source  $s$ .

### Bellman-Ford( $G, s$ )

- 1.) Initialize( $G, s$ )
- 2.) for  $i \leftarrow 1$  to  $|V|-1$
- 3.) for each  $(u, v) \in E$
- 4.)     Relax( $u, v$ )
- 5.) for each  $(u, v) \in E$
- 6.)     if  $d[v] > d[u] + w(u, v)$
- 7.)         return false
- 8.) return true

The Correctness of Bellman-Ford follows from the Path Relaxation Property and the fact that no shorter  $s-v$  path contains more than  $|V|-1$  edges (since otherwise some vertex is visited twice). (P. P. 589)

The Rationale of loop 5-7 is easy to see. If  $G$  contained a negative weight cycle Relaxation from  $s$ , then some shorter path estimate can still be relaxed, i.e.  $d[v] > d[u] + w(u, v)$ , in which case false is returned.

Run time

Initialize( $G, s$ ) cost:  $\Theta(|V|)$ .

Relax costs  $\Theta(1)$ , Each edge is relaxed  $|V|-1$  times so loop  $z=4$  costs:

$$\Theta(|E|(|V|-1)) = \Theta(|E||V|).$$

loop  $s\rightarrow$  costs:  $\Theta(|E|)$

$\therefore$  TOTAL COST:  $\Theta(|E|\cdot|V|)$ .

Exercise: Run Bellman-Ford on some example.

### 24.3 Dijkstra's Algorithm

$\text{Dijkstra}(G, s)$  Requires that all edge weights be non-negative.

It maintains a set  $S$  of vertices whose shortest path weights have been determined. i.e.

$$v \in S \Rightarrow d[v] = \delta(s, v)$$

The algorithm maintains a min-Priority Queue  $Q$  which contains vertices in  $V - S$ , keyed by their  $d$ -values. It selects  $u \in Q$  with minimum  $d[u]$ , inserts  $u$  into  $S$ , then relaxes all edges leaving  $u$ .

### Dijkstra( $G, s$ )

- 1.) Initialize( $G, s$ )
- 2.)  $S \leftarrow \emptyset$
- 3.)  $Q \leftarrow V(G)$
- 4.) while  $Q \neq \emptyset$
- 5.)      $u \leftarrow \text{extractMin}(Q)$
- 6.)      $S \leftarrow S \cup u$
- 7.)     for all  $v \in \text{adj}[u]$   
        Relax( $u, v$ )

- NOTE: when  $u$  is extracted from  $Q$  on line 5,  $d[u] = \delta(s, u)$ .

RELL PROOF ON P. 597. NOTICE THAT NON-NEGATIVE EDGE WEIGHTS ARE NECESSARY FOR PROOF.

- OBSERVE SET  $S$  is not Really Needy, i.e. may Remove line 2 & 6.
- Dijkstra is a Greedy Algorithm: it Selects the closest  $v \in Q$  to Process and insert into  $S$ .
- The call to Relax one line & contains an implicit call to the Priority Queue Operation DecreaseKey. Explicitly:

Relax( $u, v$ ) (Pre:  $v \in adj[u]$ )

- 1.) if  $d[v] > d[u] + w(u, v)$
- 2.)  $\text{DecreaseKey}(v, d[u] + w(u, v))$
- 3.)  $P[v] \leftarrow u$

### Run time:

This depends on how Priority Queue  $Q$  is implemented.  
Assume  $Q$  is a min HEAP. (Binary)

COST: line 3 (BuildHeap) :  $\Theta(|V|)$

line 5 (ExtractMin) :  $\Theta(\lg |V|)$

∴ All calls to ExtractMin cost  $\Theta(|V| \cdot \lg |V|)$ .

line 8 (Relax) :  $\Theta(\lg |V|)$

∴  $|E|$  calls to Relax cost  $\Theta(|E| \cdot \lg |V|)$

∴ TOTAL COST :  $\Theta((|V| + |E|) \log |V|)$ .

Exercise Run Dijkstra one Examples.