

Shortest Paths

- Input: weighted, directed graph $G = (V, E)$, with weight function $w : E \rightarrow \mathbf{R}$.
- The **weight** of path $p = \langle v_0, v_1, \dots, v_k \rangle$ is the sum of the weights of its constituent edges:

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

- The **shortest-path weight** from u to v is

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

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All Pairs Shortest Paths

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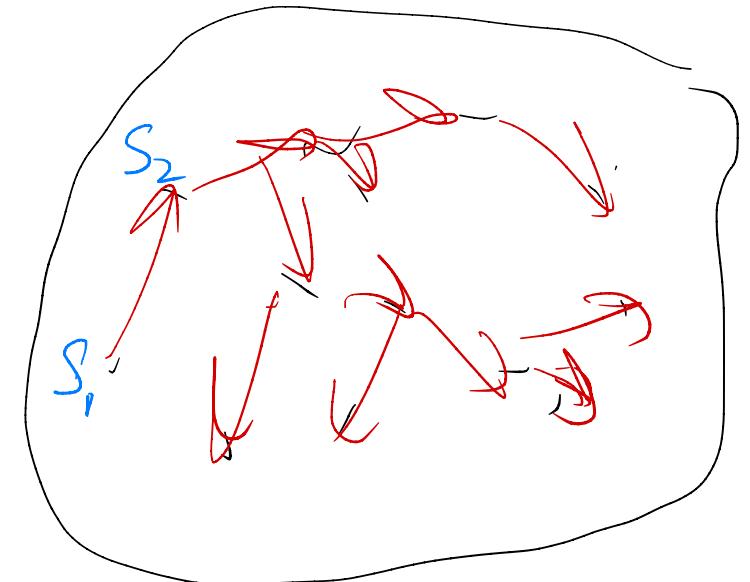
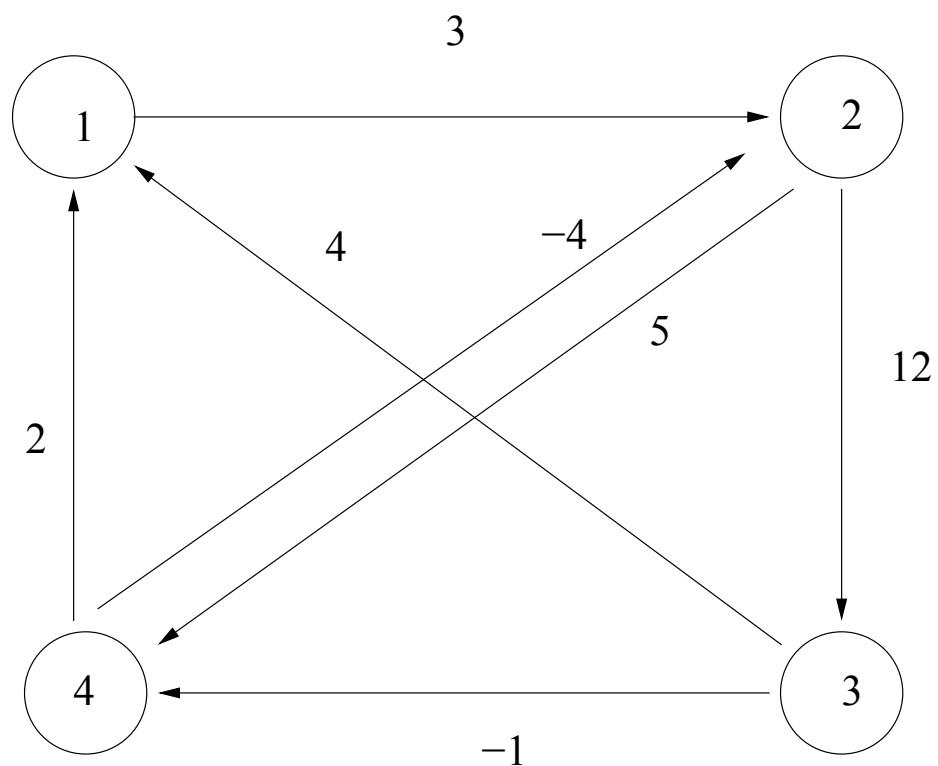
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All Pairs Shortest Paths: Compute $d(u, v)$ the shortest path distance from u to v for all pairs of vertices u and v .

Example



Solution

$$\begin{pmatrix} 0 & 3 & 15 & 8 \\ 7 & 0 & 12 & 5 \\ 1 & -5 & 0 & -1 \\ 2 & -4 & 8 & 0 \end{pmatrix}$$

Approach 1

Run Single source shortest paths V times

- $O(V^2E)$ for general graphs
- $O(VE + V^2 \log V)$ for graphs with non-negative edge weights

Other approaches : Share information between the various computations

Floyd-Warshall, Dynamic Programming

- Let $d_{ij}^{(k)}$ be the weight of a shortest path from vertex i to vertex j for which all intermediate vertices are in the set $\{1, 2, \dots, k\}$.
- When $k = 0$, a path from vertex i to vertex j with no intermediate vertex numbered higher than 0 has no intermediate vertices at all, hence $d_{ij}^{(0)} = w_{ij}$.

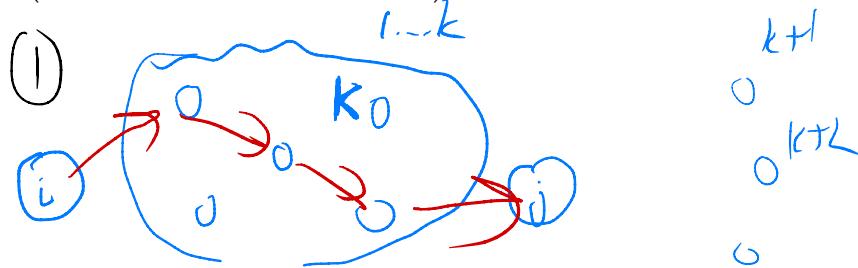
$$d_{ij}^{(k)} = \begin{cases} \min \left(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)} \right) & \text{if } k \geq 1 . \\ w_{ij} & \text{if } k = 0 , \end{cases} \quad (1)$$

Floyd-Warshall(W)

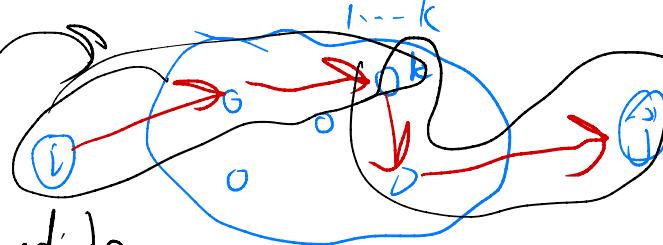
```

1   $n \leftarrow \text{rows}[W]$ 
2   $D^{(0)} \leftarrow W$ 
3  for  $k \leftarrow 1$  to  $n$ 
4      do for  $i \leftarrow 1$  to  $n$ 
5          do for  $j \leftarrow 1$  to  $n$ 
6              do  $d_{ij}^{(k)} \leftarrow \min \left( d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)} \right)$ 
7  return  $D^{(n)}$ 

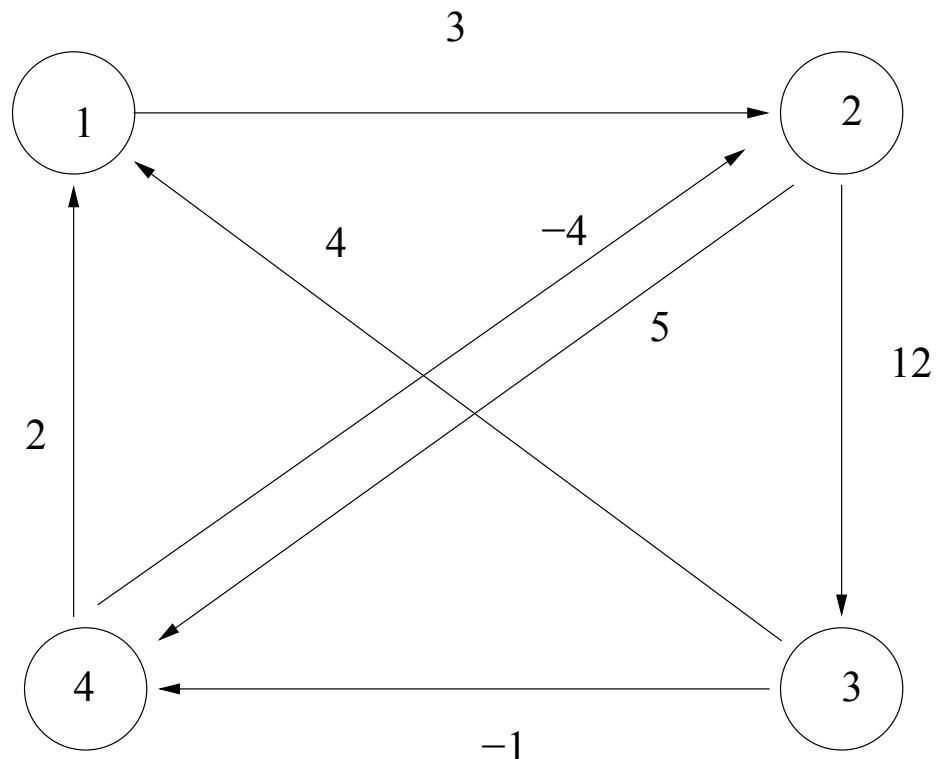
```



Running time $O(V^3)$
shortest path from
 i to k using $1 \dots k-1$ intermediate



Example



$$d_{32}^{(1)} = \min(d_{32}^{(0)}, d_{30}^{(0)} + d_{12}^{(0)})$$

$$4 + 3 = 7$$

$$D^0 = \begin{pmatrix} 0 & 3 & \infty & \infty \\ \infty & 0 & 12 & 5 \\ 4 & \infty & 0 & -1 \\ 2 & -4 & \infty & 0 \end{pmatrix} \quad D^1 = \begin{pmatrix} 0 & 3 & \infty & \infty \\ \infty & 0 & 12 & 5 \\ 4 & 7 & 0 & -1 \\ 2 & -4 & \infty & 0 \end{pmatrix} \quad D^2 = \begin{pmatrix} 0 & 3 & 15 & 8 \\ \infty & 0 & 12 & 5 \\ 4 & 7 & 0 & -1 \\ 2 & -4 & 8 & 0 \end{pmatrix}$$

$$D^3 = \begin{pmatrix} 0 & 3 & 15 & 8 \\ 16 & 0 & 12 & 5 \\ 4 & 7 & 0 & -1 \\ 2 & -4 & 8 & 0 \end{pmatrix} \quad D^4 = \begin{pmatrix} 0 & 3 & 15 & 8 \\ 7 & 0 & 12 & 5 \\ 1 & -5 & 0 & -1 \\ 2 & -4 & 8 & 0 \end{pmatrix}$$

$$d_{14}^{(3)} = \min(d_{14}^{(2)}, d_{13}^{(2)} + d_{34}^{(2)})$$

$$8, 15 + 12 = 17$$

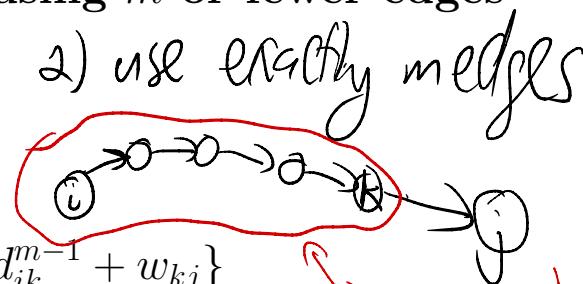
Another Algorithm

RESET ALL DEFINITIONS OF D.

- Let w_{ij} be the length of edge ij
- Let $w_{ii} = 0$
- Let d_{ij}^m be the shortest path from i to j using m or fewer edges
 - 1) uses $< m$ edges
 - 2) use exactly m edges

$$d_{ij}^1 = w_{ij}$$

$$d_{ij}^m = \min\{d_{ij}^{m-1}, \min_{1 \leq k \leq n, k \neq j} d_{ik}^{m-1} + w_{kj}\}$$



shortest path from
 i to k using
 $m-1$ or fewer edges
 d_{ik}^{m-1}

Combining these two, we get

$$d_{ij}^m = \min_{1 \leq k \leq n} \{d_{ik}^{m-1} + w_{kj}\}$$

This would give an $O(V^4)$ algorithm

loops m
 i j k

D_{ij}^m

Using matrix multiplication analogy

Note the similarity of

$$d_{ij}^m = \min_{1 \leq k \leq n} \{d_{ik}^{m-1} + w_{kj}\}$$

with matrix multiplication:

$$c_{ij} = \sum_{1 \leq k \leq n} \{a_{ik} \cdot b_{kj}\}$$

Closed
Smirking

Make the following substitutions (which have the right algebraic properties):

floor wax
desk topping
fl \rightarrow des
or \rightarrow est
wa \rightarrow t
 $\times \rightarrow$ op²

$$\begin{aligned} \text{sum} &\rightarrow \min \\ a_{ik} &\rightarrow d_{ik}^{m-1} \\ \cdot &\rightarrow + \\ b_{kj} &\rightarrow w_{kj} \\ c &\rightarrow d^m \end{aligned}$$

for i
for j
for k
 $c_{ij} = (c_{ij} + a_{ik} \cdot b_{kj})$
 $d_{ij} = \min(d_{ij}, d_{ik} + a_{ik} \cdot b_{kj})$

Using this matrix multiplication terminology, we have

$$\begin{aligned} D^1 &= W \\ D^2 &= D^1 \cdot W = W^2 \\ D^3 &= D^2 \cdot W = W^3 \\ \dots &\quad \dots \quad \dots \\ D^m &= D^{m-1}W = W^m \end{aligned}$$

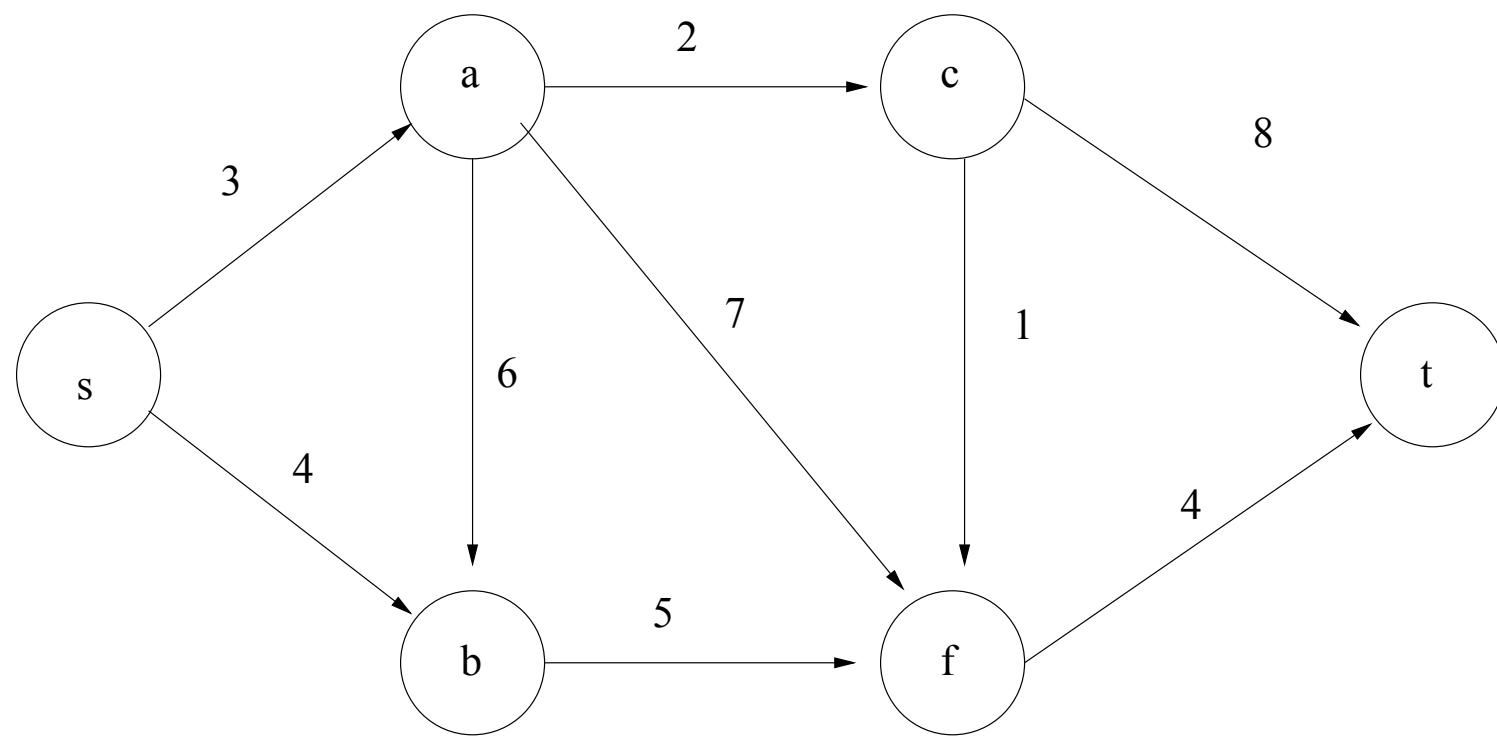
$$\sqrt{mat, mat} \quad \sqrt{3} = \sqrt{4}$$

But we can execute W^m be repeated squaring and get $\tilde{O}(V^3 \log V)$ time.

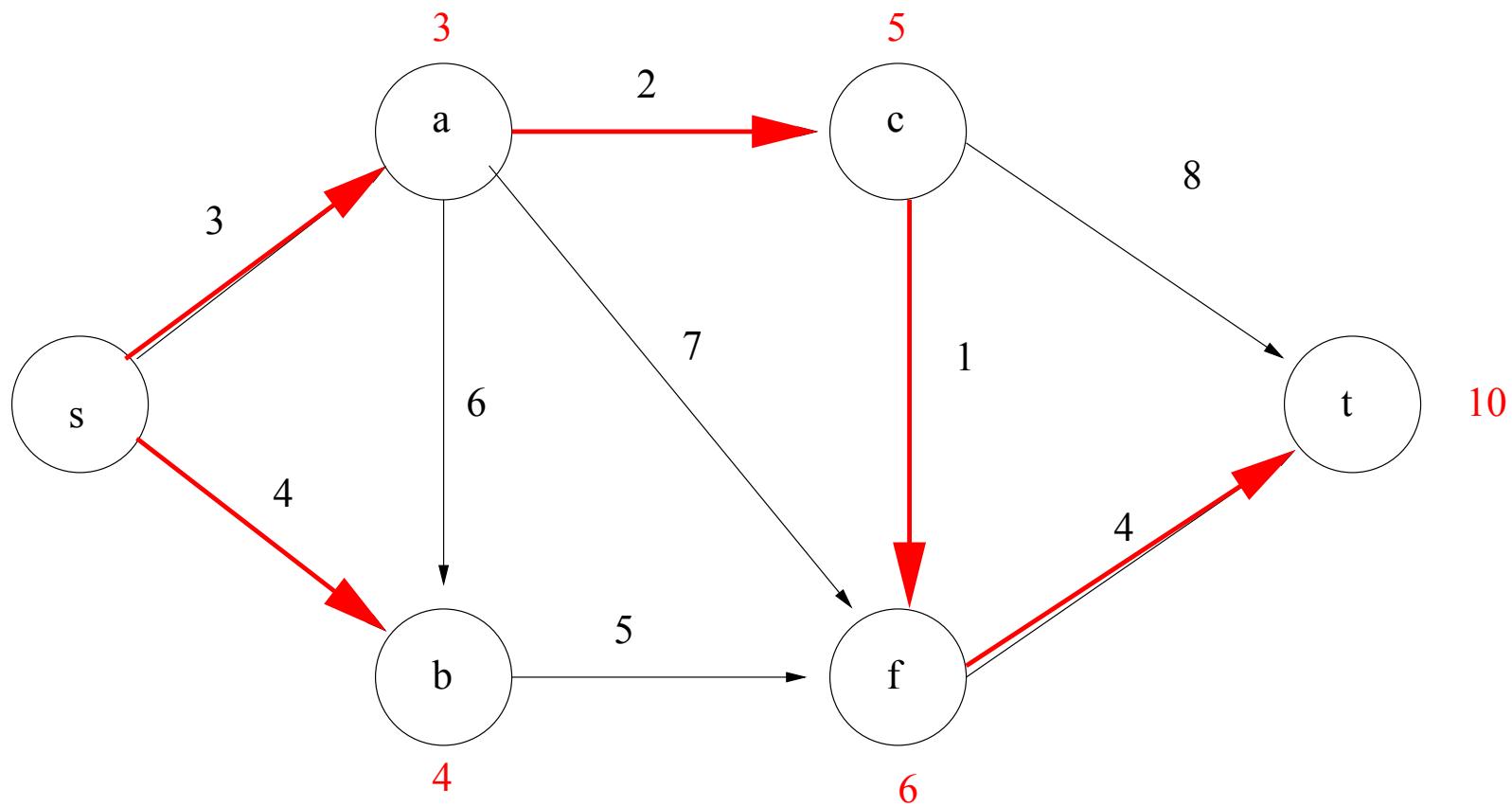
$$W^m \xrightarrow{\text{repeated squaring}}$$
$$W^{27} = W^{16} \cdot W^8 \cdot W^2 \cdot W^1$$

$$\begin{aligned} W^2 &= W \cdot W \\ W^4 &= W^2 \cdot W^2 \\ W^8 &= W^4 \cdot W^4 \\ W^{16} &= W^8 \cdot W^8 \\ W^{27} &= W^{16} \cdot W^8 \cdot W^2 \cdot W^1 \end{aligned} \quad O(\lg m) \text{ matrix mult.}$$

Example



Solution



Shortest Paths

Shortest Path Variants

- Single Source-Single Sink
- Single Source (all destinations from a source s)
- All Pairs

Defs:

- Let $\delta(v)$ be the real shortest path distance from s to v
- Let $d(v)$ be a value computed by an algorithm

Edge Weights

- All non-negative
- Arbitrary

Note: Must have no negative cost cycles

Single Source Shortest Paths

Key Property: Subpaths of shortest paths are shortest paths Given a weighted, directed graph $G = (V, E)$ with weight function $w : E \rightarrow \mathbf{R}$, let $p = \langle v_1, v_2, \dots, v_k \rangle$ be a shortest path from vertex v_1 to vertex v_k and, for any i and j such that $1 \leq i \leq j \leq k$, let $p_{ij} = \langle v_i, v_{i+1}, \dots, v_j \rangle$ 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 .

Note: this is optimal substructure

Corollary 1 For all edges $(u, v) \in E$,

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

Corollary 2 Shortest paths follow a tree of edges for which

$$\delta(v) = \delta(u) + w(u, v)$$

More precisely, any edge in a shortest path must satisfy

$$\delta(v) = \delta(u) + w(u, v)$$

Relaxation

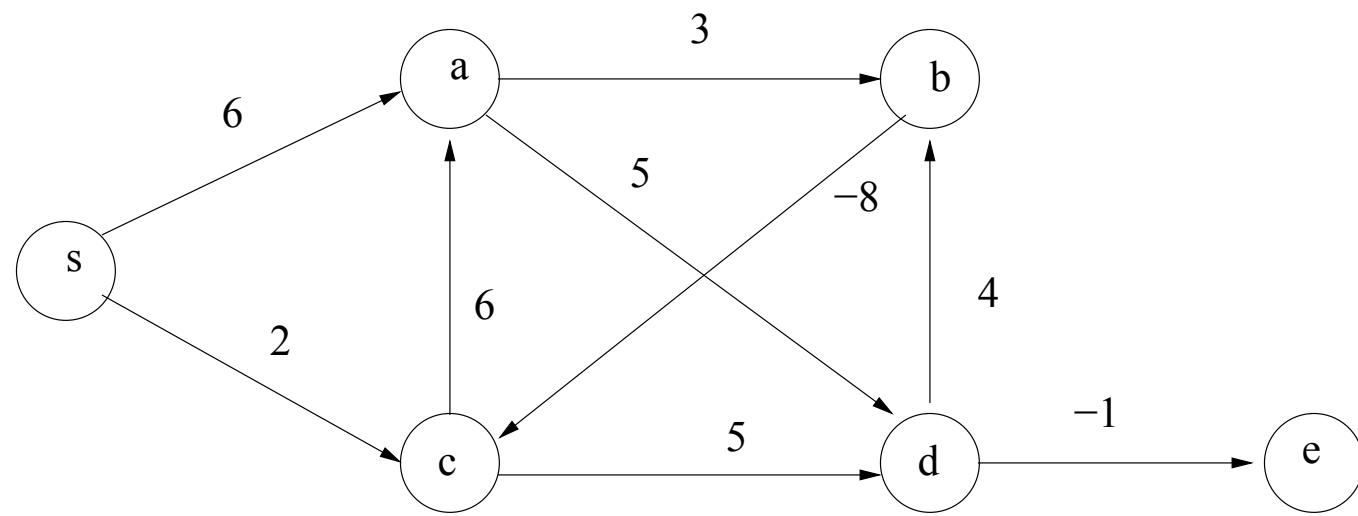
Relax(u, v, w)

- 1 **if** $d[v] > d[u] + w(u, v)$
- 2 **then** $d[v] \leftarrow d[u] + w(u, v)$
- 3 $\pi[v] \leftarrow u$ (keep track of actual path)

Lemma: Assume that we initialize all $d(v)$ to ∞ , $d(s) = 0$ and execute a series of Relax operations. Then for all v , $d(v) \geq \delta(v)$.

Lemma: Let $P = e_1, \dots, e_k$ be a shortest path from s to v . After initialization, suppose that we relax the edges of P in order (but not necessarily consecutively). Then $d(v) = \delta(v)$.

Example



Algorithms

Goal of an algorithm: Relax the edges in a shortest path in order (but not necessarily consecutively).

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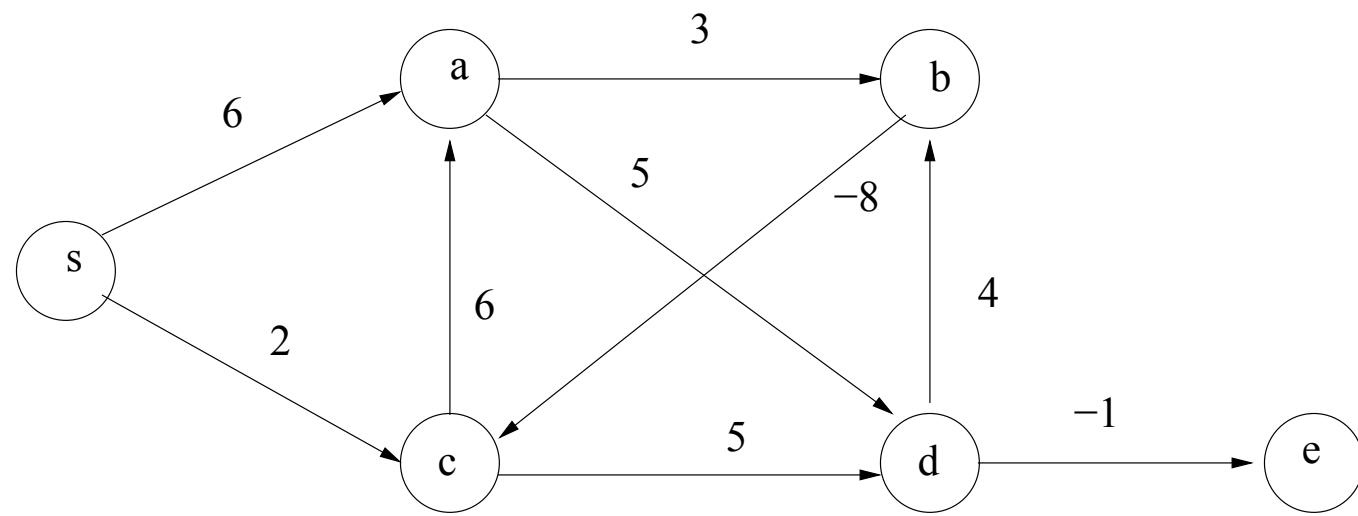
Bellman-Ford(G, w, s)

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2  for  $i \leftarrow 1$  to  $|V[G]| - 1$ 
3      do for each edge  $(u, v) \in E[G]$ 
4          do RELAX( $u, v, w$ )
5  for each edge  $(u, v) \in E[G]$ 
6      do if  $d[v] > d[u] + w(u, v)$ 
7          then return FALSE
8  return TRUE
```

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   $d[s] \leftarrow 0$ 
```

Example



Correctness of Bellman Ford

- Every shortest path must be relaxed in order
- If there are negative weight cycles, the algorithm will return false

Running Time $O(VE)$

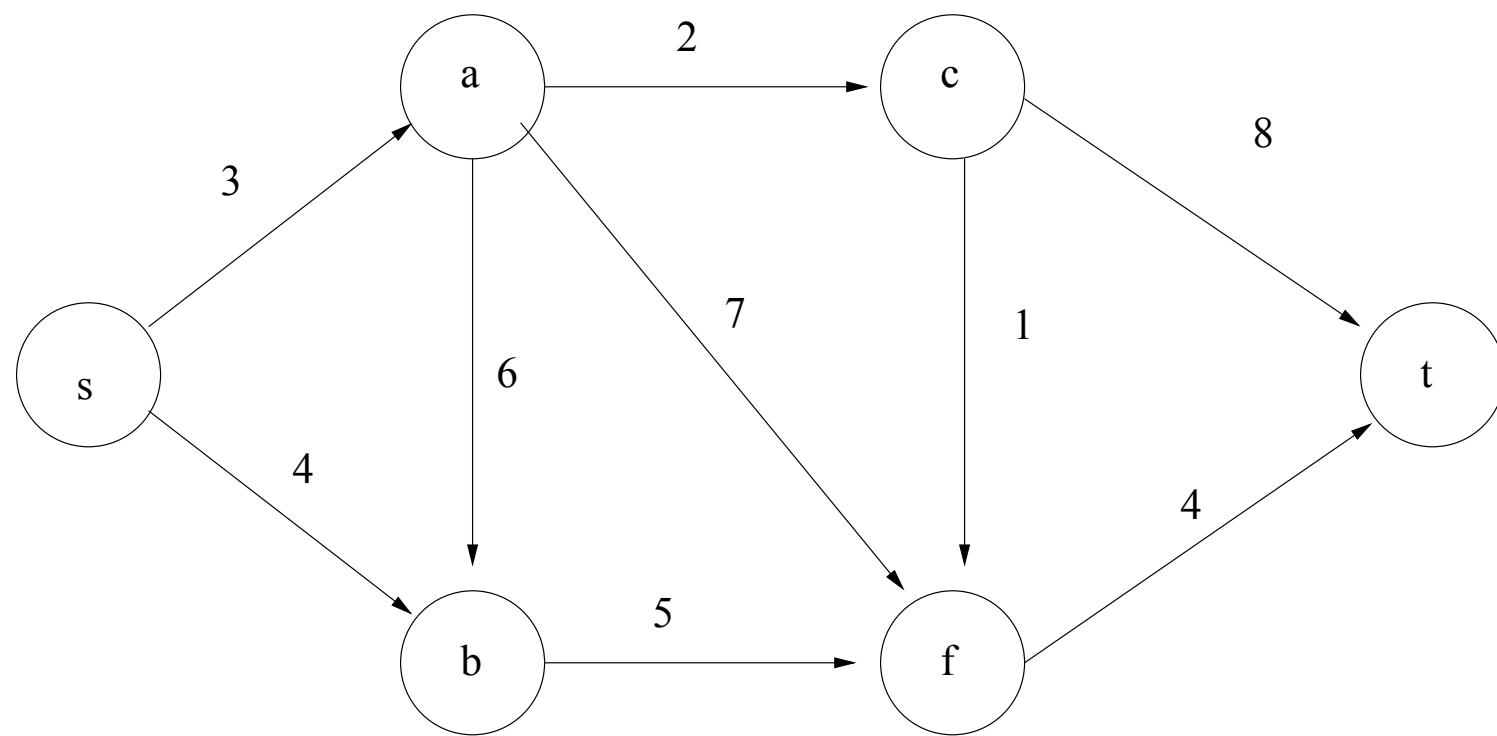
All edges non-negative

- Dijkstra's algorithm, a greedy algorithm
- Similar in spirit to Prim's algorithm
- Idea: Run a discrete event simulation of breadth-first-search. Figure out how to implement it efficiently
- Can relax edges out of each vertex exactly once.

Dijkstra(G, w, s)

```
1  INITIALIZE-SINGLE-SOURCE( $G, s$ )
2   $S \leftarrow \emptyset$ 
3   $Q \leftarrow V[G]$   $\triangleright$  This line does  $V$  INSERTS
4  while  $Q \neq \emptyset$ 
5      do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
6           $S \leftarrow S \cup \{u\}$ 
7          for each vertex  $v \in \text{Adj}[u]$ 
8              do  $\text{RELAX}(u, v, w)$   $\triangleright$  This line does a DECREASE-KEY
```

Example



Correctness

Correctness of Dijkstra's algorithm Dijkstra's algorithm, run on a weighted, directed graph $G = (V, E)$ with nonnegative weight function w and source s , terminates with $d[v] = \delta(s, v)$ for all vertices $u \in V$.

Claim to Prove: When v is put in S , $d(v) = \delta(v)$.

Proof

Claim to Prove: When v is put in S , $d(v) = \delta(v)$.

Proof

- $d(v) \geq \delta(v)$ because any algorithm that does a sequence of Relax calls has this property.
- Assume fpc that $d(v) > \delta(v)$, and that v is the first such vertex that is permanently labelled that has this property.
- Consider the state of the world just before v is put in S
 - shortest path from s to v goes through an edge (x, y) where $x \in S$ and $y \notin S$ (it is possible that $y = v$ and/or $x = s$).
 - $d(x) = \delta(x)$, because $x \in S$
 - $d(y) = \delta(y)$ because (x, y) was relaxed when x was put in S .
- Putting these together with $\delta(y) \leq \delta(v)$ because y is before v on a shortest path, we have

$$d(y) = \delta(y) \leq \delta(v) < d(v)$$

- $d(y) < d(v)$, so the algorithm would have chosen to permanently label y and not v , which is a contradiction.

Running Time

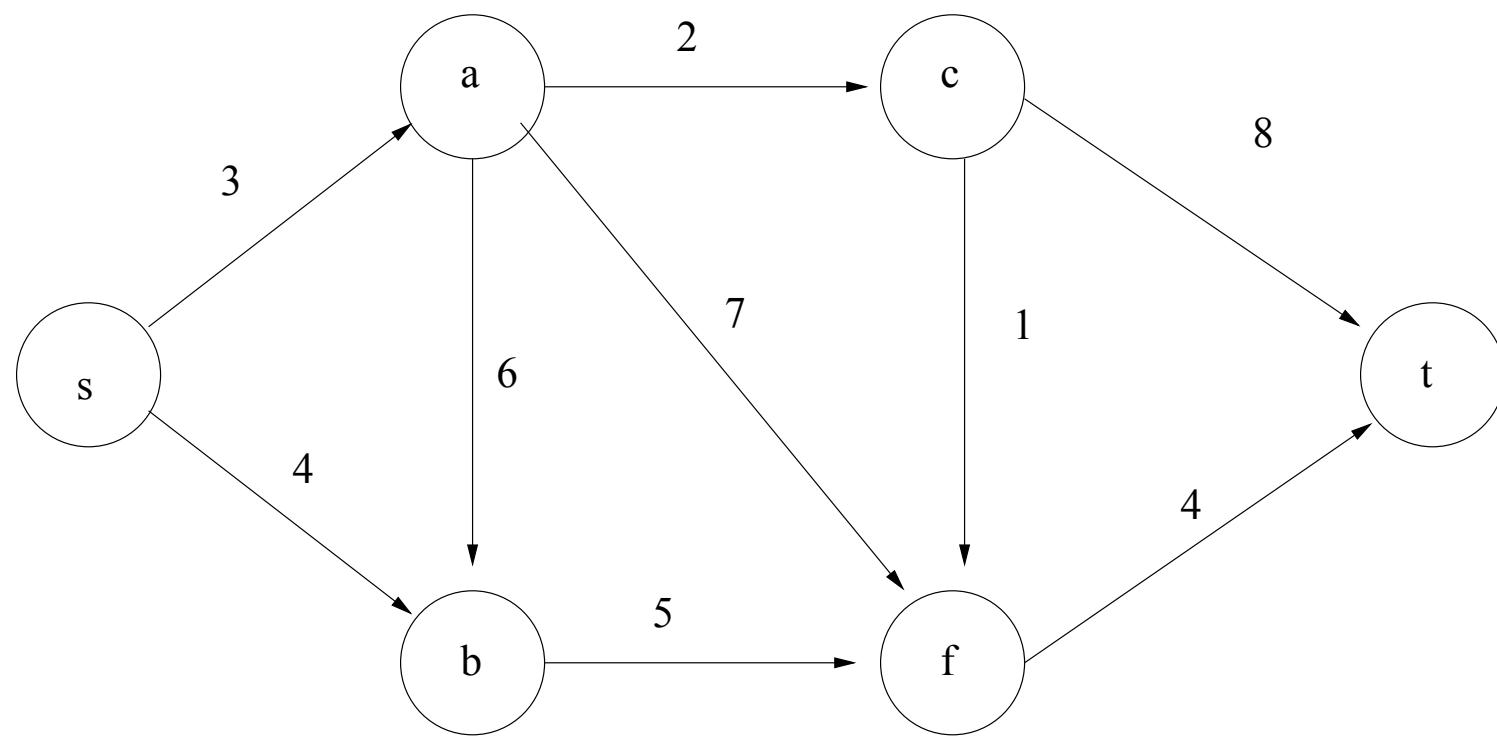
- E decrease keys and V delete-min's
- $O(E \log V)$ using a heap
- $O(E + V \log V)$ using a Fibonacci heap

Shortest Path in a DAG

Dag-Shortest-Paths(G, w, s)

- 1 topologically sort the vertices of G
- 2 INITIALIZE-SINGLE-SOURCE'(G, s)
- 3 for each u taken in topological order
 - 4 do for each $v \in \text{Adj}[u]$
 - 5 do RELAX(u, v, w)

Example



Correctness and Running Time

Correctness If a weighted, directed graph $G = (V, E)$ has source vertex s and no cycles, then at the termination of the DAG-SHORTEST-PATHS procedure, $d[v] = \delta(s, v)$ for all vertices $v \in V$, and the predecessor subgraph G_π is a shortest-paths tree.

Running Time

- Topological sort is linear time
 - Each edge is relaxed once
 - No additional data structure overhead
- $O(V + E)$ time.