Algorithms: Dynamic Programming

Shortest Path Problems:
Floyd-Warshall Algorithm
Johnson's Algorithm

Shortest-Path Problems

- Shortest-Path problems
 - **Single-Source (Single-Destination):** Find a shortest path from a given source (vertex *s*) to each of the vertices.
 - **Single-Pair:** Given two vertices, find a shortest path between them. Solution to single-source problem solves this problem efficiently, too.
 - **All-Pairs:** Find shortest-paths for every pair of vertices. Dynamic programming algorithm.

All-Pairs Shortest Paths

- We want to compute a table giving the length of the shortest path between any two vertices. (We also would like to get the shortest paths themselves.)
- We could just call Dijkstra or Bellman-Ford |V| times, passing a different source vertex each time.
- It can be done in $\Theta(V^3)$, which seems to be as good as you can do on dense graphs.

Doing APSP with SSSP

Dijkstra would take time

$$\Theta(V \times E \lg V) = \Theta(V E \lg V)$$
 [by using ordinary heaps]
 $\Theta(V \times (V \lg V + E)) = \Theta(V^2 \lg V + VE)$ [Fibonacci heaps],
but doesn't work with negative-weight edges.

• Bellman-Ford would take $\Theta(V \times VE) = \Theta(V^2E)$.

The Floyd-Warshall Algorithm

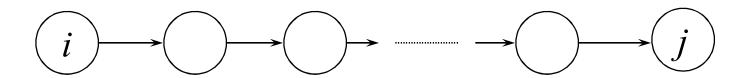
• Represent the directed, edge-weighted graph in adjacency-matrix form.

•
$$W$$
= matrix of weights =
$$\begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \end{bmatrix}$$

- w_{ij} is the weight of edge (i, j), or ∞ if there is no such edge.
- Return a matrix D, where each entry d_{ij} is $\delta(i, j)$.
- Could also return a predecessor matrix, Π , where each entry π_{ij} is the predecessor of j on the shortest path from i.

Floyd-Warshall: Idea

• Consider *intermediate vertices* of a path:

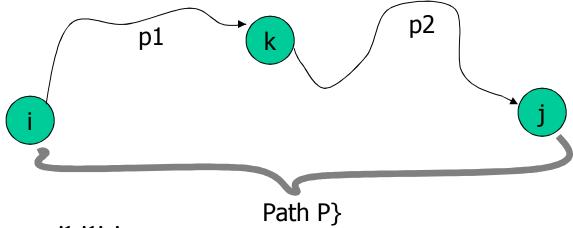


Say we know the length of the shortest path from i to j whose intermediate vertices are only those with numbers 1, 2, ..., k-1. Call this length $d_{ij}^{(k-1)}$

Now how can we extend this from k-1 to k? In other words, we want to compute $d_{ij}^{(k)}$.

Can we use $d_{ij}^{(k-1)}$, and if so how?

Floyd-Warshall: Idea



Two possibilities:

- 1. *k* is not an intermediate vertex on *P*: the path through vertices 1 ... *k*-1 is still the shortest.
- 2. *k* is an intermediate vertex on *P*: there is a shorter path consisting of two subpaths, one from *i* to *k* and one from *k* to *j*.

If the vertex k is not an intermediate vertex on P, then

$$d_{ij}^{(k)} = d_{ij}^{(k-1)}$$

If the vertex k is an intermediate vertex on P, then

$$d_{ij}^{(k)} = d_{ik}^{(k-1)} + d_{kj}^{(k-1)}$$

Floyd-Warshall: Idea

Therefore, we can conclude that

$$d_{ij}^{(k)} = \min\{d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\}$$

If we do not use intermediate nodes, i.e., when k=0, then

$$\mathbf{d}_{ij}^{(0)} = \mathbf{w}_{ij}$$

If k > 0, then

$$d_{ij}^{(k)} = \min\{d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\}$$

When k = |V|, we're done.

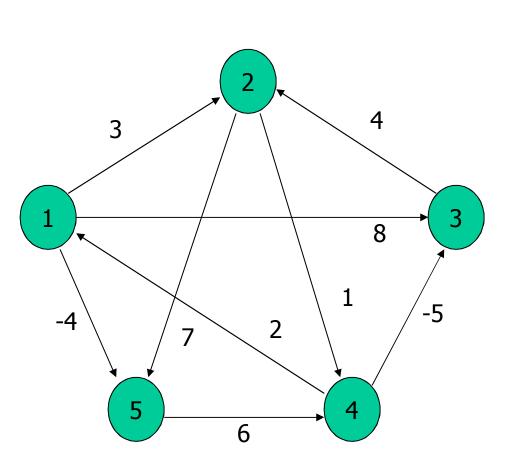
Dynamic Programming

- Floyd-Warshall is a *dynamic programming* algorithm:
- Compute and store solutions to sub-problems. Combine those solutions to solve larger sub-problems.
- Here, the sub-problems involve finding the shortest paths through a subset of the vertices.

Code for Floyd-Warshall

```
Floyd-Warshall(W)
```

```
1 \ n \leftarrow rows[W] // number of vertices
2 D^{(0)} \leftarrow W
3 for k \leftarrow 1 to n do
      for i \leftarrow 1 to n do
          for j \leftarrow 1 to n do
            d_{ii}^{(k)} \leftarrow \min\{d_{ii}^{(k-1)}, d_{ik}^{(k-1)} + d_{ki}^{(k-1)}\}
7 return D^{(n)}
Running time: \theta(V^3).
(Small constant, because operations are simple.)
```



$$D(0) = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & \infty & -5 & 0 & \infty \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

$$D(1) = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

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$$D(1) = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

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$$D(2) = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

$$D(2) = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix} \qquad \Pi(2) = \begin{pmatrix} NIL & 1 & 1 & 2 & 1 \\ NIL & NIL & NIL & 2 & 2 \\ NIL & 3 & NIL & 2 & 2 \\ 4 & 1 & 4 & NIL & 1 \\ NIL & NIL & NIL & 5 & NIL \end{pmatrix}$$

$$D(3) = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & -1 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

$$D(3) = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & -1 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix} \qquad \prod(3) = \begin{pmatrix} NIL & 1 & 1 & 2 & 1 \\ NIL & NIL & NIL & 2 & 2 \\ NIL & 3 & NIL & 2 & 2 \\ 4 & 3 & 4 & NIL & 1 \\ NIL & NIL & NIL & 5 & NIL \end{pmatrix}$$

$$D(4) = \begin{pmatrix} 0 & 3 & -1 & 4 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix}$$

$$D(4) = \begin{pmatrix} 0 & 3 & -1 & 4 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix} \qquad \Pi(4) = \begin{pmatrix} NIL & 1 & 4 & 2 & 1 \\ 4 & NIL & 4 & 2 & 1 \\ 4 & 3 & NIL & 2 & 1 \\ 4 & 3 & 4 & NIL & 1 \\ 4 & 3 & 4 & 5 & NIL \end{pmatrix}$$

$$D(5) = \begin{pmatrix} 0 & 1 & -3 & 2 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix}$$

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Johnson's Algorithm

- Makes clever use of Bellman-Ford and Dijkstra to do All-Pairs-Shortest-Paths efficiently on sparse graphs
- Motivation: By running Dijkstra |V| times, we could do APSP in time
 - $\Theta(V^2 \lg V + VE \lg V)$ (Modified Dijkstra), or
 - $\Theta(V^2 \lg V + VE)$ (Fibonacci Dijkstra).

This beats $\Theta(V^3)$ (Floyd-Warshall) when the graph is sparse

• Problem: negative edge weights

The Basic Idea

- Reweight the edges so that:
 - No edge weight is negative.
 - Shortest paths are preserved (A shortest path in the original graph is still one in the new, reweighted graph)
- An obvious attempt: subtract the minimum weight from all the edge weights.

For example, if the minimum weight is -2:

$$-2 - (-2) = 0$$

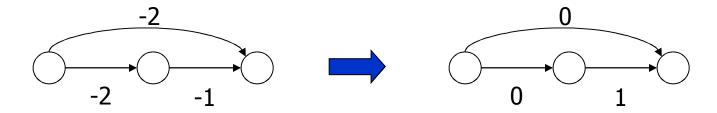
3 - (-2) = 5

etc.

Counter Example

• Subtracting the minimum weight from every weight doesn't work.

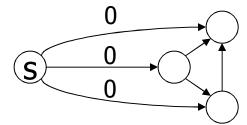
Consider:



• Paths with more edges are unfairly penalized.

Johnson's Insight

• Add a vertex s to the original graph G, with edges of weight 0 to each vertex in G:



• Assign new weights \hat{w} to each edge as follows:

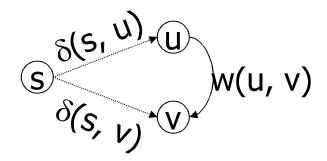
$$\hat{w}(u, v) = w(u, v) + \delta(s, u) - \delta(s, v)$$

Question 1

• Are all the w's non-negative? **Yes:**

$$\delta(s,u) + w(u,v)$$
 must be $\geq \delta(s,v)$

Otherwise, $s \Rightarrow u \rightarrow v$ would be shorter than the shortest path from s to v.



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

Rewriting:

$$\underbrace{w(u,v) + \delta(s,u) - \delta(s,v)}_{\hat{w}(u,v)} \ge 0$$

Question 2

• Does the reweighting preserve shortest paths? Yes:

Consider any path $p = v_1, v_2, ..., v_k$

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

$$= w(v_1, v_2) + \delta(s, v_1) - \delta(s, v_2)$$

$$+ w(v_2, v_3) + \delta(s, v_2) - \delta(s, v_3)$$

$$\vdots$$

$$+ w(v_{k-1}, v_k) + \delta(s, v_{k-1}) - \delta(s, v_k)$$

$$= w(p) + \delta(s, v_1) - \delta(s, v_k)$$

A value that depends only on the endpoints, not on the path.

Because $\delta(s, v_1)$ and $\delta(s, v_k)$ do not depend on the path, if one path from v_1 to v_k is shorter than another using weight function w, then it is also shorter using \hat{w} . Thus, shortest paths will be preserved.

Question 3

• How do we compute the $\delta(s, v)$'s?

Use Bellman-Ford Algorithm.

This also tells us if we have a negative-weight cycle.

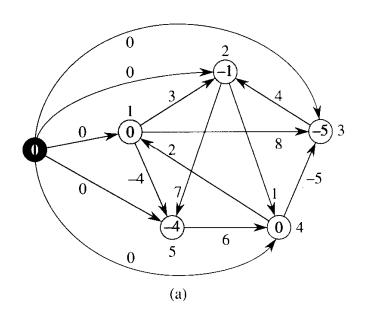
Johnson's Algorithm

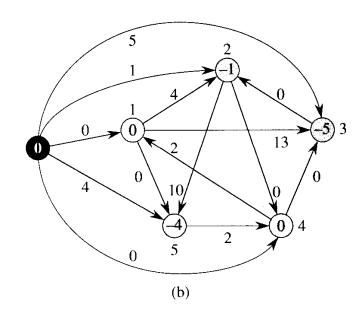
- 1. Compute G', which consists of G augmented with s and a zero-weight edge from s to every vertex in G.
- 2. Run Bellman-Ford(G', w, s) to obtain the $\delta(s, v)$'s
- 3. Reweight by computing ŵ for each edge
- 4. Run Dijkstra on each vertex to compute δ'
- 5. Undo reweighting factors to compute δ

Johnson's Algorithm: CLRS

```
Johnson(G)
     compute G', where V[G'] = V[G] \cup \{s\},
              E[G'] = E[G] \cup \{(s, v) : v \in V[G]\}, \text{ and }
              w(s, v) = 0 for all v \in V[G]
     if Bellman-Ford (G', w, s) = \text{False}
 3
        then print "the input graph contains a negative-weight cycle"
        else for each vertex v \in V[G']
                   do set h(v) to the value of \delta(s, v)
                                computed by the Bellman-Ford algorithm
              for each edge (u, v) \in E[G']
 6
                   do \widehat{w}(u,v) \leftarrow w(u,v) + h(u) - h(v)
              for each vertex u \in V[G]
                   do run DIJKSTRA(G, \widehat{w}, u) to compute \widehat{\delta}(u, v) for all v \in V[G]
10
                       for each vertex v \in V[G]
                            do d_{uv} \leftarrow \widehat{\delta}(u,v) + h(v) - h(u)
11
12
              return D
```

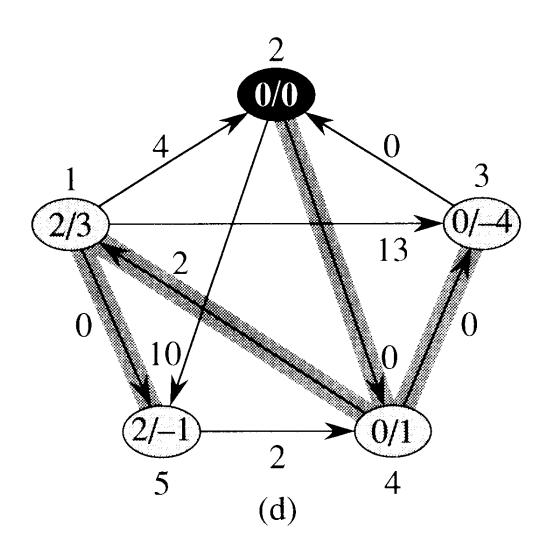
Johnson: reweighting





$$\hat{\mathbf{w}}(u, v) = w(u, v) + \delta(s, u) - \delta(s, v)$$

Johnson using Dijkstra



Johnson's: Running Time

- 1. Computing G': $\Theta(V)$
- 2. Bellman-Ford: $\Theta(VE)$
- 3. Reweighting: $\Theta(E)$
- 4. Running (Modified) Dijkstra: $\Theta(V^2 \lg V + VE \lg V)$
- 5. Adjusting distances: $\Theta(V^2)$

Total is dominated by Dijkstra: $\Theta(V^2 \lg V + VE \lg V)$