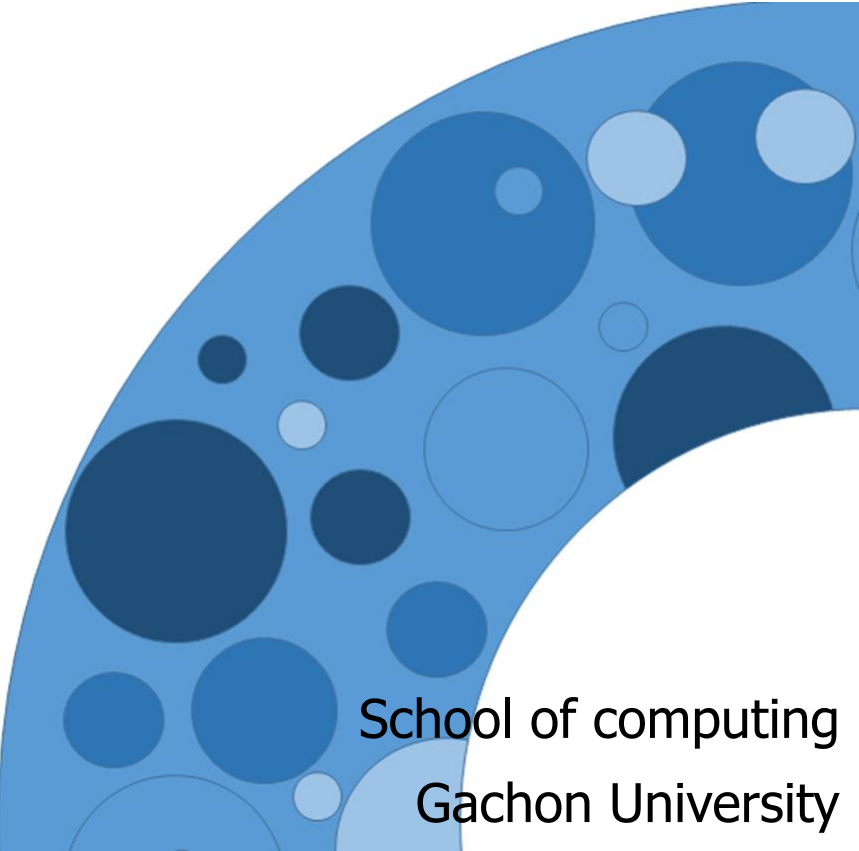


Algorithms

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A decorative graphic in the bottom right corner consisting of a blue curved shape filled with various-sized circles in different shades of blue, creating a bubble-like or cellular pattern.

5. Graph Algorithms III

- The Bellman-Ford algorithm: Single-Source Shortest Paths
- The Floyd-Warshall algorithm: All-Pairs Shortest Paths

Contents

- Single-source shortest paths
 - Bellman-Ford algorithm
- All Pairs Shortest Paths
 - Matrix multiplication
 - Floyd-Warshall algorithm

The Bellman-Ford Algorithm

- Dijkstra's algorithm doesn't work when there are negative edges:
 - Intuition – we can not be greedy any more on the assumption that the lengths of paths will only increase in the future
- Bellman-Ford algorithm detects negative cycles (returns *false*) or returns the shortest path-tree
 - Returns True if no negative-weight cycles reachable from s, False otherwise.

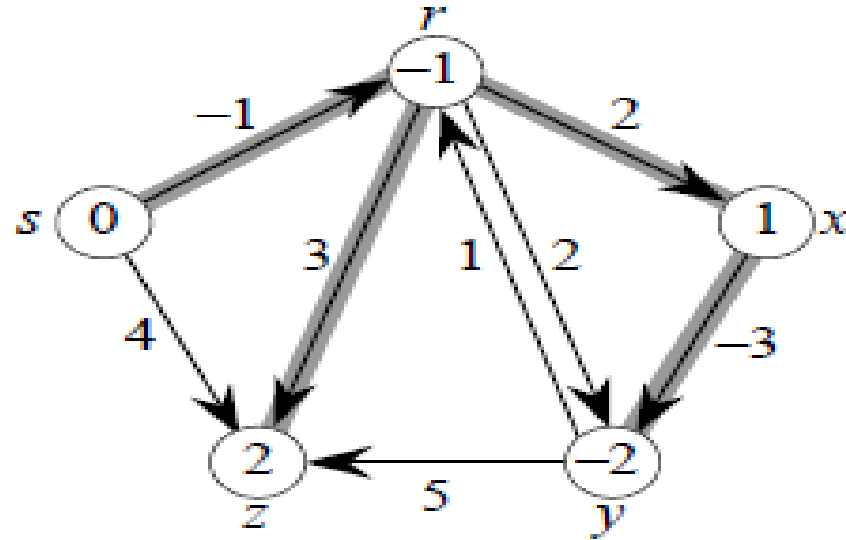
The Bellman-Ford Algorithm (cont'd)

```
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
```

- **Core:** The first **for** loop relaxes all edges $|V| - 1$ times.
- **Time:** $\Theta(V E)$.

The Bellman-Ford Algorithm (cont'd)

- Examples
 - Values you get on each pass and how quickly it converges depends on order of relaxation.
 - But guaranteed to converge after $|V| - 1$ passes, assuming no negative-weight cycles.



The Bellman-Ford Algorithm (cont'd)

Proof Use path-relaxation property.

- Let v be reachable from s , and let $p = v_0, v_1, \dots, v_k$ be a shortest path from s to v , where $v_0 = s$ and $v_k = v$. Since p is acyclic, it has $\leq |V| - 1$ edges, so $k \leq |V| - 1$.

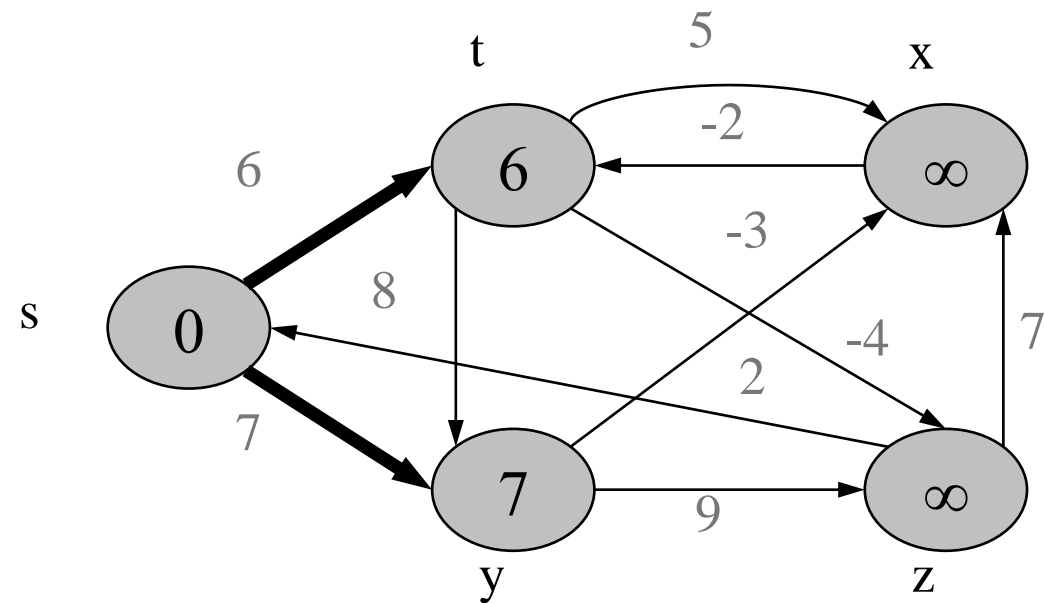
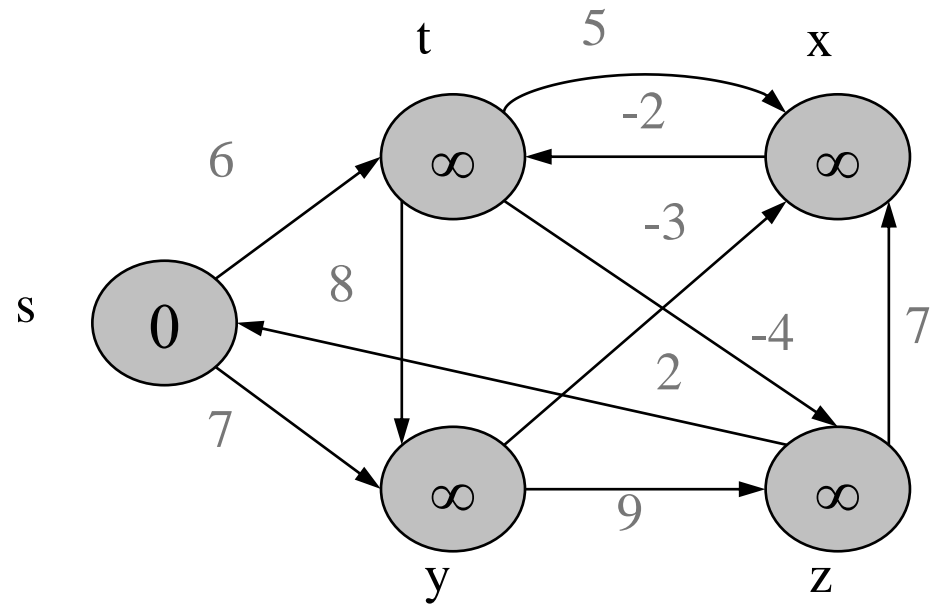
Each iteration of the for loop relaxes all edges:

- First iteration relaxes (v_0, v_1) .
- Second iteration relaxes (v_1, v_2) .
- K^{th} iteration relaxes (v_{k-1}, v_k) .

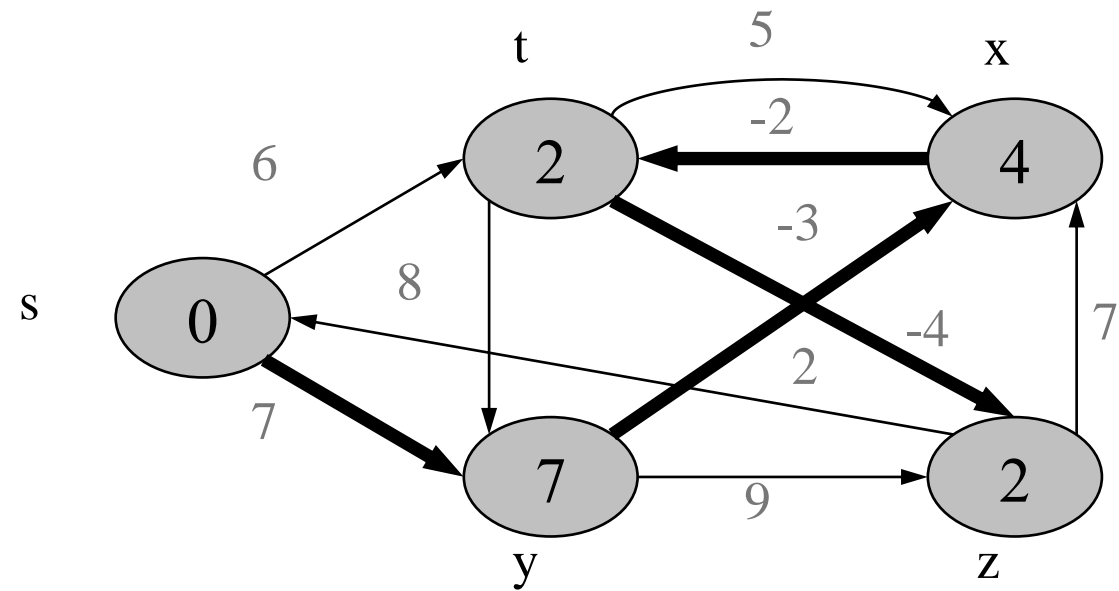
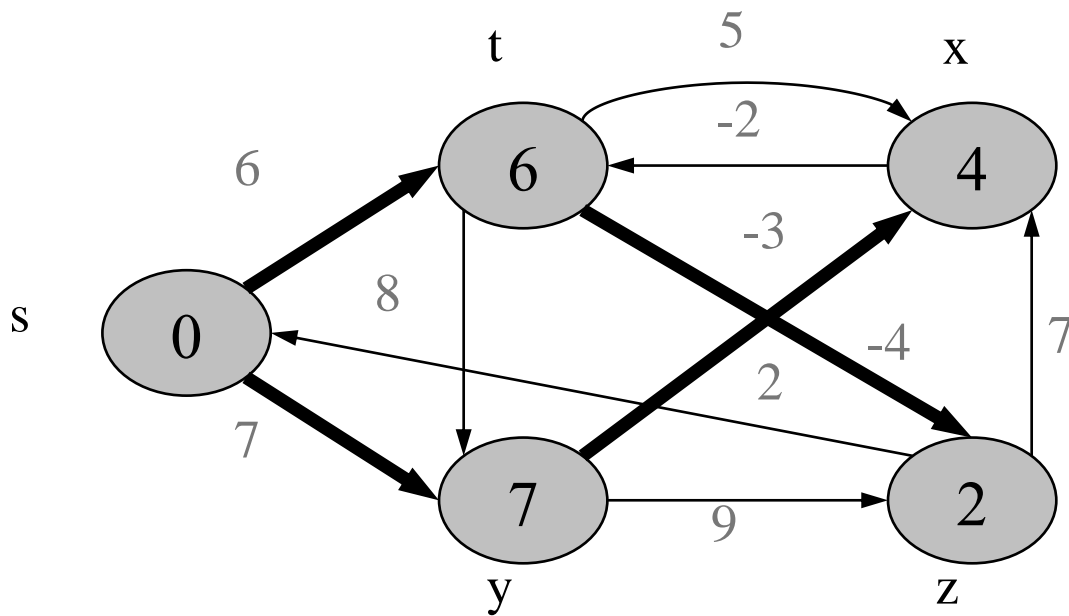
By the path-relaxation property,

$$d[v] = d[v_k] = \delta(s, v_k) = \delta(s, v).$$

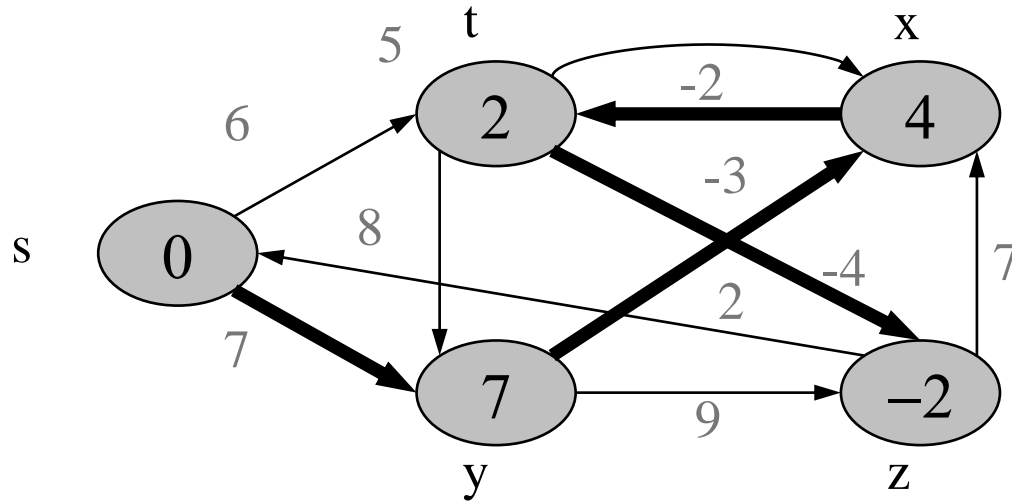
The Bellman-Ford Algorithm (cont'd)



The Bellman-Ford Algorithm (cont'd)



The Bellman-Ford Algorithm (cont'd)



- Bellman-Ford running time:
 - $(|V|-1)|E| + |E| = O(VE)$

All Pairs Shortest Paths

motivation

- computer network
- aircraft network (e.g. flying time, fares)
- railroad network
- table of distances between all pairs of cities for a road atlas

single source shortest path algorithms

- if edges are non-negative:

running the Dijkstra's algorithm n -times, once for each vertex as the source

- *running time:* $O(nm \lg n)$
(using binary-heap)

- negative-weight edges:

Bellman-Ford algorithm

- *running time:* $O(n^2 m)$

All-Pairs Shortest Paths

data structure

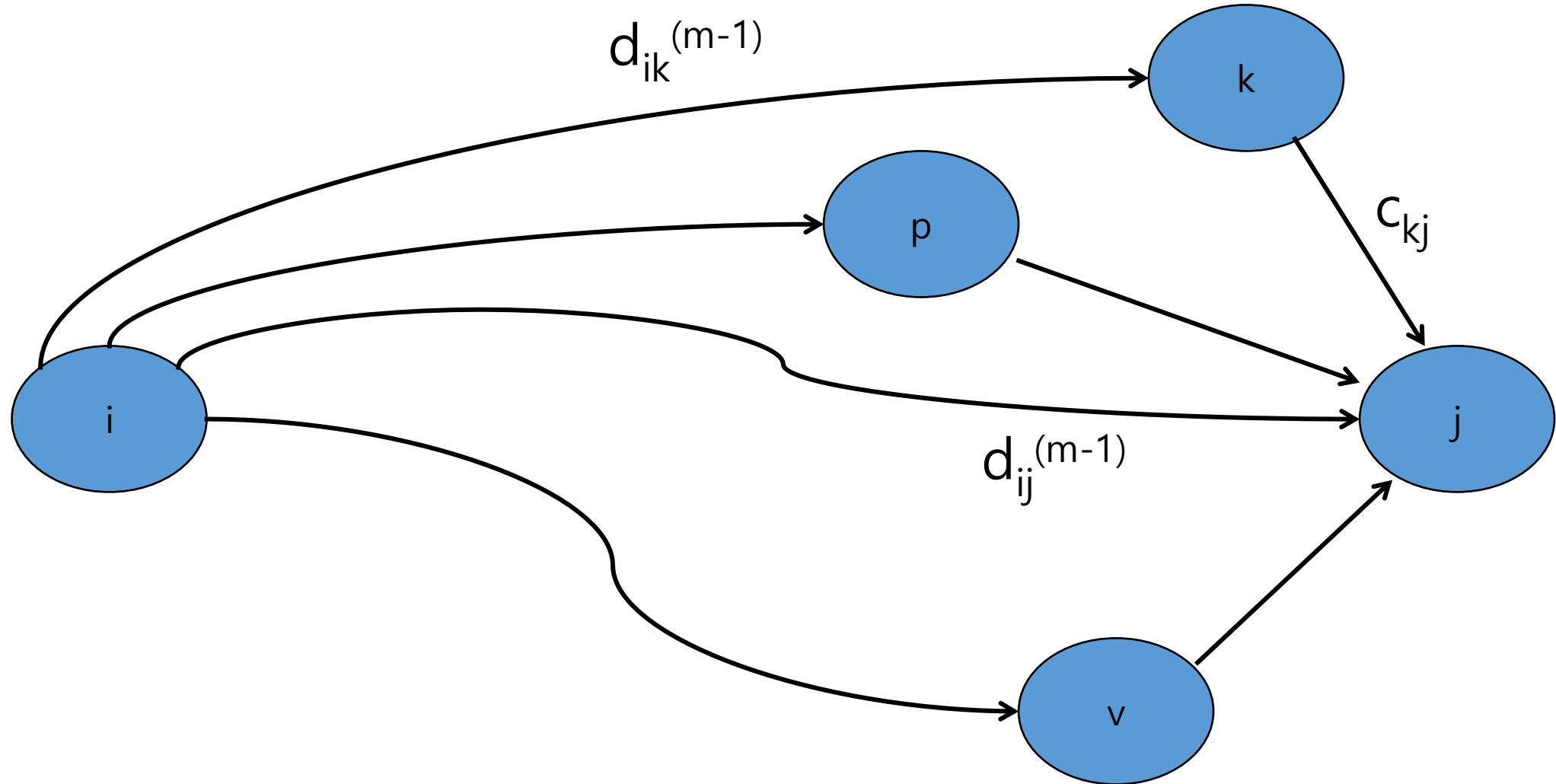
- adjacency matrix
- $c: E \rightarrow \mathbb{R}$ as $n \times n$ matrix C

$$c_{ij} = \begin{cases} 0 & \text{if } i = j, \\ c(i,j) & \text{if } i \neq j \text{ and } (i,j) \in E, \\ \infty & \text{if } i \neq j \text{ and } (i,j) \notin E \end{cases}$$

matrix multiplication (idea)

- $d_{ij}^{(m)}$: minimum weight of any path
from i to j
that contains at most m edges

matrix multiplication (idea)



matrix multiplication (idea)

$$d_{ij}^{(m)} = \min (d_{ij}^{(m-1)}, \min_{1 \leq k \leq n} \{d_{ik}^{(m-1)} + c_{kj}\})$$

↪ look at all possible predecessors k of j and compare

matrix multiplication (structure)

$$d_{ij}^{(1)} = c_{ij}$$

$$\begin{aligned} d_{ij}^{(m)} &= \min (d_{ij}^{(m-1)}, \min_{1 \leq k \leq n} \{d_{ik}^{(m-1)} + c_{kj}\}) \\ &= \min_{1 \leq k \leq n} \{d_{ik}^{(m-1)} + c_{kj}\} \end{aligned}$$

matrix multiplication (structure)

- Compute a series of matrices

$$D^{(1)}, D^{(2)}, \dots, D^{(n-1)}$$

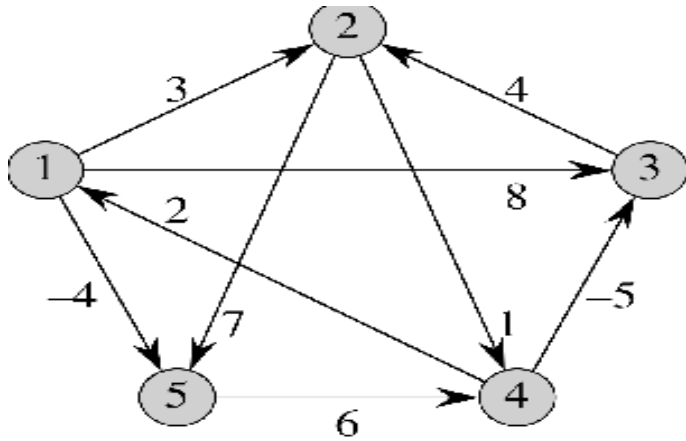
$$D^{(m)} = D^{(m-1)} \cdot C$$

final matrix $D^{(n-1)}$ contains the final shortest-path weights

matrix multiplication (pseudo-code)

```
EXTEND-SHORTEST-PATHS ( $D, C$ )  
 $n \leftarrow \text{rows}[D]$   
let  $D' = (d'_{ij})$  be an  $n \times n$  matrix  
for  $i \leftarrow 1$  to  $n$   
  do for  $j \leftarrow 1$  to  $n$   
    do  $d'_{ij} \leftarrow \infty$   
    for  $k \leftarrow 1$  to  $n$   
      do  $d'_{ij} \leftarrow \min(d'_{ij}, d_{ik} + c_{kj})$   
return  $D'$ 
```

matrix multiplication (example)



$$d_{14}^{(2)} = (0 \ 3 \ 8 \ \infty \ -4) * \begin{pmatrix} \infty \\ 1 \\ \infty \\ 0 \\ 6 \end{pmatrix}$$

$$= \min(\infty, 4, \infty, \infty, 2)$$

$$= 2$$

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

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

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

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

Figure 25.1, p690

matrix multiplication (example)

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

$$\begin{aligned} d_{14}^{(2)} &= (0 \ 3 \ 8 \ \infty \ -4) * \begin{pmatrix} \infty \\ 1 \\ \infty \\ 0 \\ 6 \end{pmatrix} \\ &= \min(\infty, 4, \infty, \infty, 2) \\ &= 2 \end{aligned}$$

relation to matrix multiplication

$$C = A \cdot B$$

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

compare:

$$D^{(m)} = D^{(m-1)} \cdot C$$

$$\Rightarrow d_{ij}^{(m)} = \min_{1 \leq k \leq n} \{d_{ik}^{(m-1)} + c_{kj}\}$$

matrix multiplication

- compute the sequence of $n-1$ matrices:

$$D^{(1)} = D^{(0)} \cdot C = C,$$

$$D^{(2)} = D^{(1)} \cdot C = C^2,$$

$$D^{(3)} = D^{(2)} \cdot C = C^3,$$

...

$$D^{(n-1)} = D^{(n-2)} \cdot C = C^{n-1}$$

matrix multiplication (pseudo-code)

ALL-PAIRS-SHORTEST-PATHS (C)

$n \leftarrow \text{rows}[C]$

$D^{(1)} \leftarrow C$

for $m \leftarrow 2$ **to** $n - 1$

do $D^{(m)} \leftarrow \text{EXTEND-SHORTEST-PATHS}(D^{(m-1)}, C)$

return $D^{(n-1)}$

- *negative cycles:* $d_{ii} < 0$

matrix multiplication (running time)

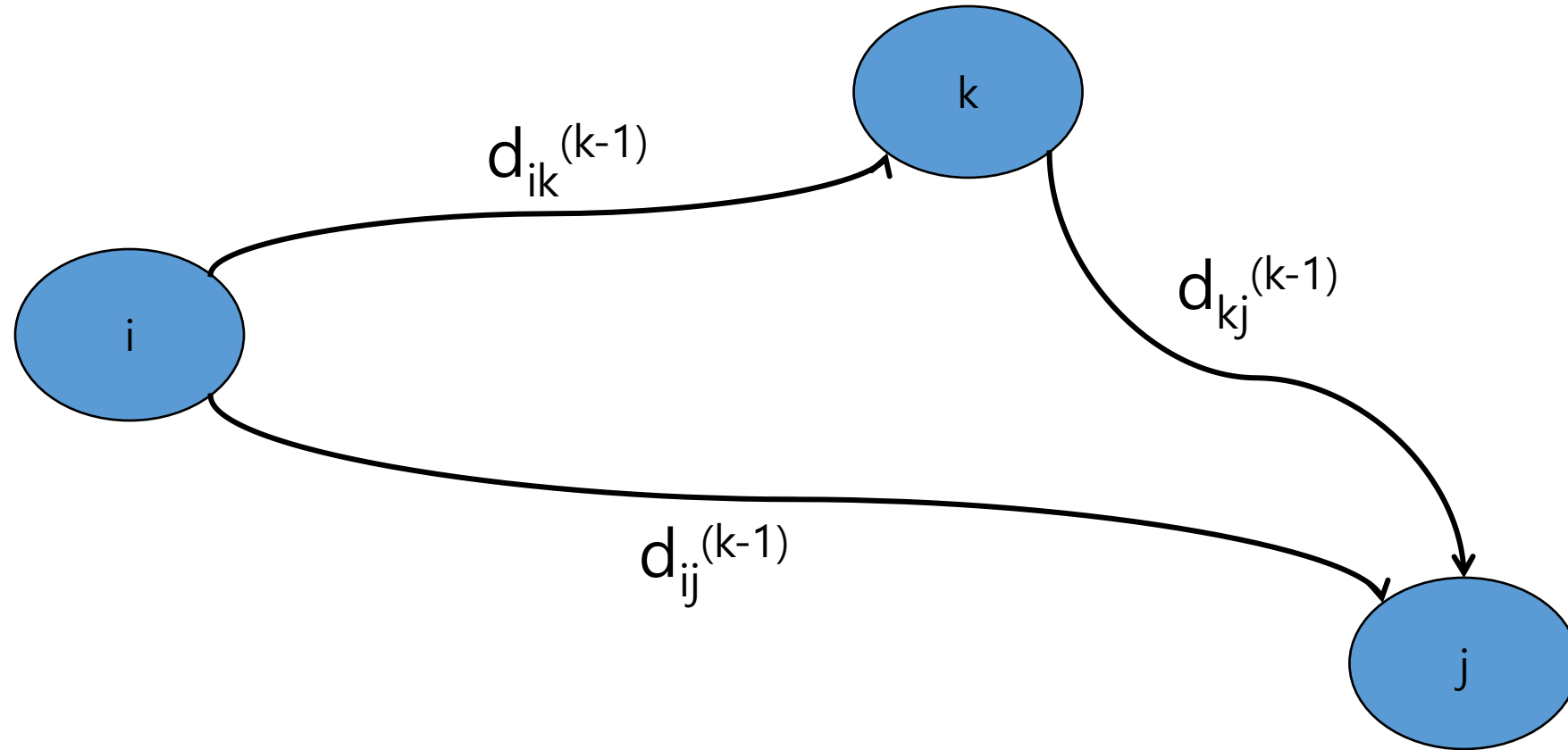
Improving the running time:

- compute not all $D^{(m)}$ matrices
interested only in $D^{(n-1)}$, which is $D^{(m)}$ for all integers $m \geq n-1$

The Floyd-Warshall Algorithm

- Graph analysis algorithm for finding shortest paths in a weighted, directed graph
- Graph can have negative weights, but not negative weight cycles
- Example of dynamic programming, a method of solving problems where one needs to find the best decision one after another

Floyd-Warshall algorithm (cont'd)



Floyd-Warshall algorithm (cont'd)

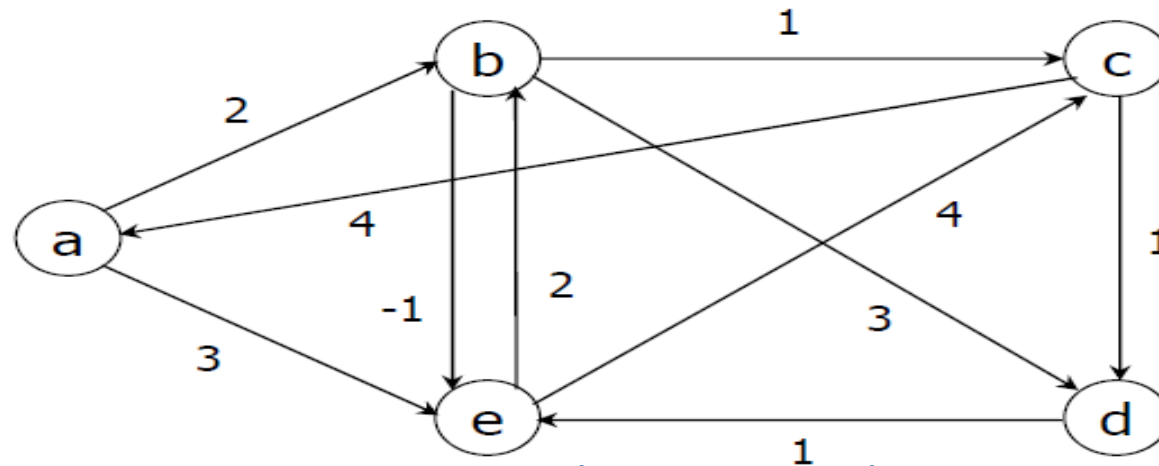
- Consider a graph G with vertices V , each numbered 1 through N .
 - Further consider a function $\text{shortestPath}(i, j, k)$ that returns the shortest possible path from i to j using vertices only from the set $\{1, 2, \dots, k\}$ as intermediate points along the way.
 - Now, given this function, our goal is to find the shortest path from each i to each j using only vertices 1 to $k + 1$.

Floyd-Warshall algorithm (cont'd)

- There are two candidates for each of these paths: either the true shortest path only uses vertices in the set $\{1, \dots, k\}$; or there exists some path that goes from i to $k + 1$, then from $k + 1$ to j that is better.
 - We know that the best path from i to j that only uses vertices 1 through k is defined by $\text{shortest_path}(i, j, k)$, and it is clear that if there were a better path from i to j , then the length of this path would be the concatenation of the shortest path from i to $k + 1$ (using vertices in $\{1, \dots, k\}$) and the shortest path from $k + 1$ to j (also using vertices in $\{1, \dots, k\}$).

Floyd-Warshall algorithm (cont'd)

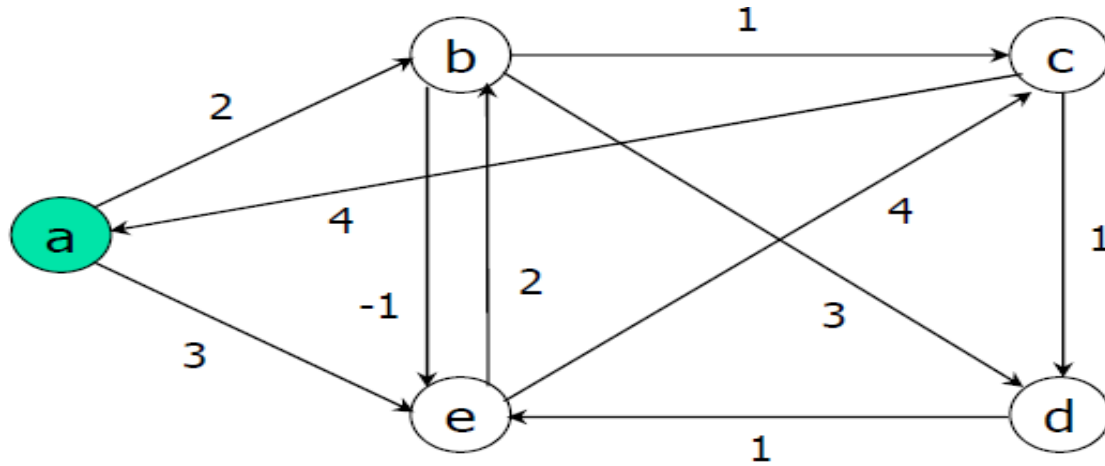
Example : step(1/6), $K=0$



$$D^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & \infty & \infty & 3 \\ \infty & 0 & 1 & 3 & -1 \\ 4 & \infty & 0 & 1 & \infty \\ \infty & \infty & \infty & 0 & 1 \\ \infty & 2 & 4 & \infty & 0 \end{bmatrix} \end{matrix}$$

Floyd-Warshall algorithm (cont'd)

Example : step(2/6), $k=1$

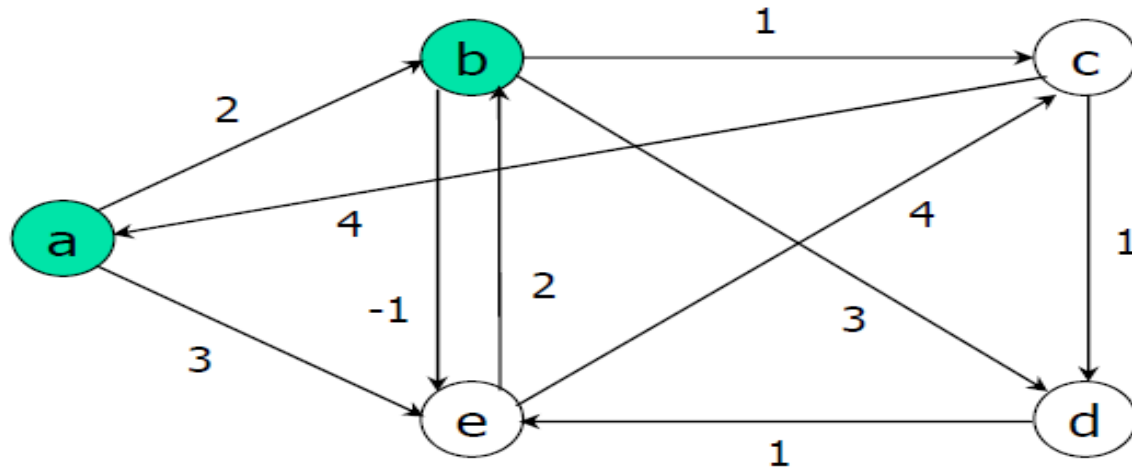


$$D^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & \infty & \infty & 3 \\ \infty & 0 & 1 & 3 & -1 \\ 4 & \underline{6} & 0 & 1 & \underline{7} \\ \infty & \infty & \infty & 0 & 1 \\ \infty & 2 & 4 & \infty & 0 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} d_{35}^{(1)} &= \min(d_{35}^{(0)}, d_{31}^{(0)} + d_{15}^{(0)}) \\ &= \min(\infty, 4 + 3) \\ &= 7 \end{aligned}$$

Floyd-Warshall algorithm (cont'd)

Example : step(3/6), $k=2$

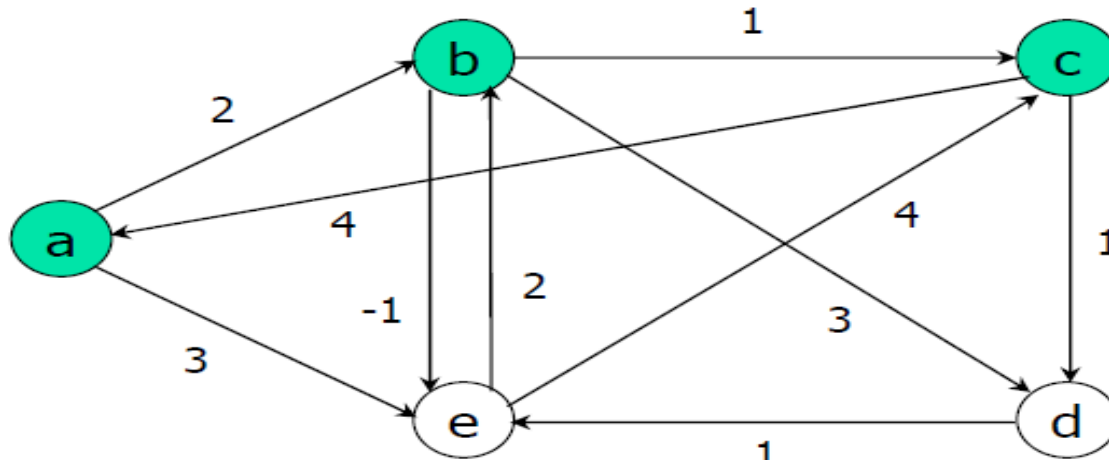


$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & \underline{3} & \underline{5} & \underline{1} \\ \infty & 0 & 1 & 3 & -1 \\ 4 & 6 & 0 & 1 & \underline{5} \\ \infty & \infty & \infty & 0 & 1 \\ \infty & 2 & 3 & \underline{5} & 0 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} d_{35}^{(2)} &= \min(d_{35}^{(1)}, d_{32}^{(1)} + d_{25}^{(1)}) \\ &= \min(7, 6 - 1) \\ &= 5 \end{aligned}$$

Floyd-Warshall algorithm (cont'd)

Example : step(4/6), $k=3$

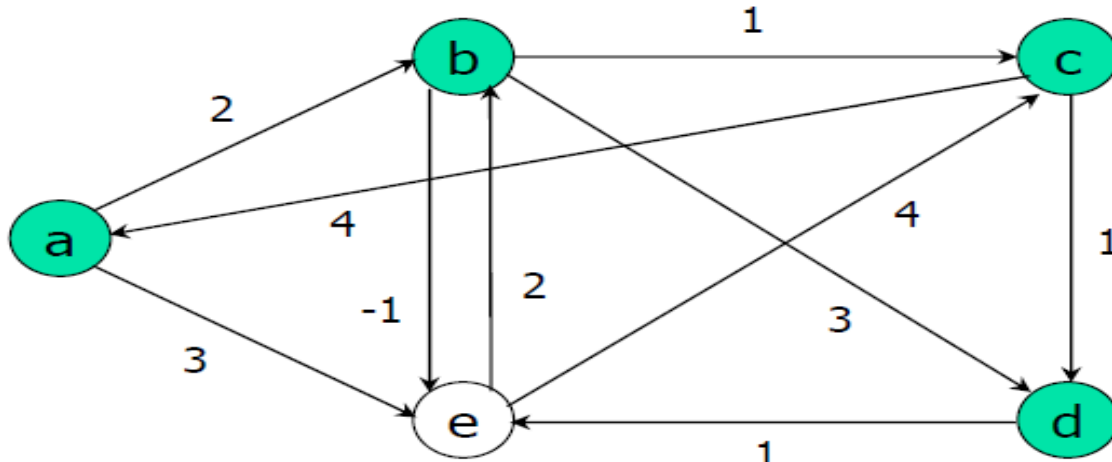


$$D^{(3)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & 3 & \underline{4} & 1 \\ \underline{5} & 0 & 1 & \underline{2} & -1 \\ 4 & 6 & 0 & 1 & \textcolor{red}{5} \\ \infty & \infty & \infty & 0 & 1 \\ \underline{7} & 2 & 3 & \underline{4} & 0 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} d_{35}^{(3)} &= \min(d_{35}^{(2)}, d_{33}^{(2)} + d_{35}^{(2)}) \\ &= \min(5, 0 + 5) \\ &= 5 \end{aligned}$$

Floyd-Warshall algorithm (cont'd)

Example : step(5/6), $k=4$

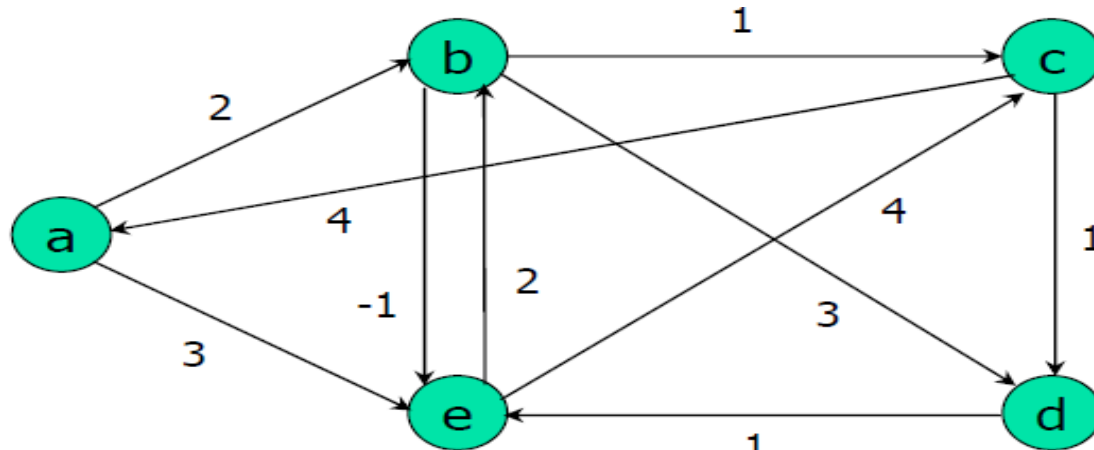


$$D^{(4)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & 3 & 4 & 1 \\ 5 & 0 & 1 & 2 & -1 \\ 4 & 6 & 0 & 1 & \underline{2} \\ \infty & \infty & \infty & 0 & 1 \\ 7 & 2 & 3 & 4 & 0 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} d_{35}^{(4)} &= \min(d_{35}^{(3)}, d_{34}^{(3)} + d_{45}^{(3)}) \\ &= \min(5, 1+1) \\ &= 2 \end{aligned}$$

Floyd-Warshall algorithm (cont'd)

Example : step(6/6) , $k=5$



$$D^{(5)} = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 2 & 3 & 4 & 1 \\ 5 & 0 & 1 & 2 & -1 \\ 4 & \underline{4} & 0 & 1 & 2 \\ \underline{8} & \underline{3} & \underline{4} & 0 & 1 \\ 7 & 2 & 3 & 4 & 0 \end{bmatrix} \end{matrix}$$

$$\begin{aligned} d_{35}^{(5)} &= \min(d_{35}^{(4)}, d_{35}^{(4)} + d_{55}^{(4)}) \\ &= \min(2, 2 + 0) \\ &= 2 \end{aligned}$$

Floyd-Warshall algorithm (cont'd)

Compute bottom-up

Compute in increasing order of k :

FLOYD-WARSHALL(W, n)

$D^{(0)} = W$

for $k = 1$ **to** n

 let $D^{(k)} = (d_{ij}^{(k)})$ be a new $n \times n$ matrix

for $i = 1$ **to** n

for $j = 1$ **to** n

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

return $D^{(n)}$

running time: $O(n^3)$

Floyd-Warshall algorithm (cont'd)

Floyd(A, n) {

/* input: adjacency matrix $A[1..n][1..n]$, $n=|V|$,

output: distance matrix D */

```
    int i, j, k;
1   for (i = 1; i <= n; i++)
2       for (j = 1; j <= n; j++)
3           D[i][j] = A[i][j];
4   for (k = 1; k <= n; k++)
5       for (i = 1; i <= n; i++)
6           for (j = 1; j <= n; j++)
7               if (D[i][j] > D[i][k] + D[k][j])
8                   D[i][j] = D[i][k] + D[k][j];
}
```

Implementation

Implementation of Floyd-Warshall algorithm

Compute bottom-up

Compute in increasing order of k :

FLOYD-WARSHALL(W, n)

$D^{(0)} = W$

for $k = 1$ **to** n

 let $D^{(k)} = (d_{ij}^{(k)})$ be a new $n \times n$ matrix

for $i = 1$ **to** n

for $j = 1$ **to** n

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

return $D^{(n)}$

```
import sys
N = 5 # matrix size NxN
M = 5 # the number of edges

distance = [[sys.maxsize for j in range(N)] for i in range(N)]

# initial edges
for i in range(0, N):
    distance[i][i] = 0

distance[0][1] = 2
distance[0][4] = 3
distance[1][2] = 1
distance[1][3] = 3
distance[1][4] = -1
distance[2][0] = 4
distance[2][3] = 1
distance[3][4] = 1
distance[4][1] = 2
distance[4][2] = 3

# core algorithm
for k in range(0, N):
    for i in range(0, N):
        for j in range(0, N):
            if distance[i][j] > distance[i][k] + distance[k][j]:
                distance[i][j] = distance[i][k] + distance[k][j]

for i in range(0, N):
    for j in range(0, N):
        if distance[i][j] == sys.maxsize:
            print('&', end=' ')
        else:
            print(distance[i][j], end=' ')
    print()
```

```
0 2 3 4 1
5 0 1 2 -1
4 4 0 1 2
8 3 4 0 1
7 2 3 4 0
```

Example code test

- Code test: <https://www.acmicpc.net/problem/11403>
- Solving the problem using Floyd-Warshall algorithm
- Example result of submission

THANK YOU

