

# COMP319 Algorithms 1

## Lecture 16

### Shortest Paths part 2

Instructor: Gil-Jin Jang

School of Electronics Engineering, Kyungpook National University

Textbook Chapters 25 and 26

Slide credits: 홍석원, 명지대학교; 김한준, 서울시립대학교;  
J. Lillis, UIC; Roger Crawfis, CSE 680;

George Bebis, Analysis of Algorithms, CS 477/677

David Luebke, CS332, Virginia University

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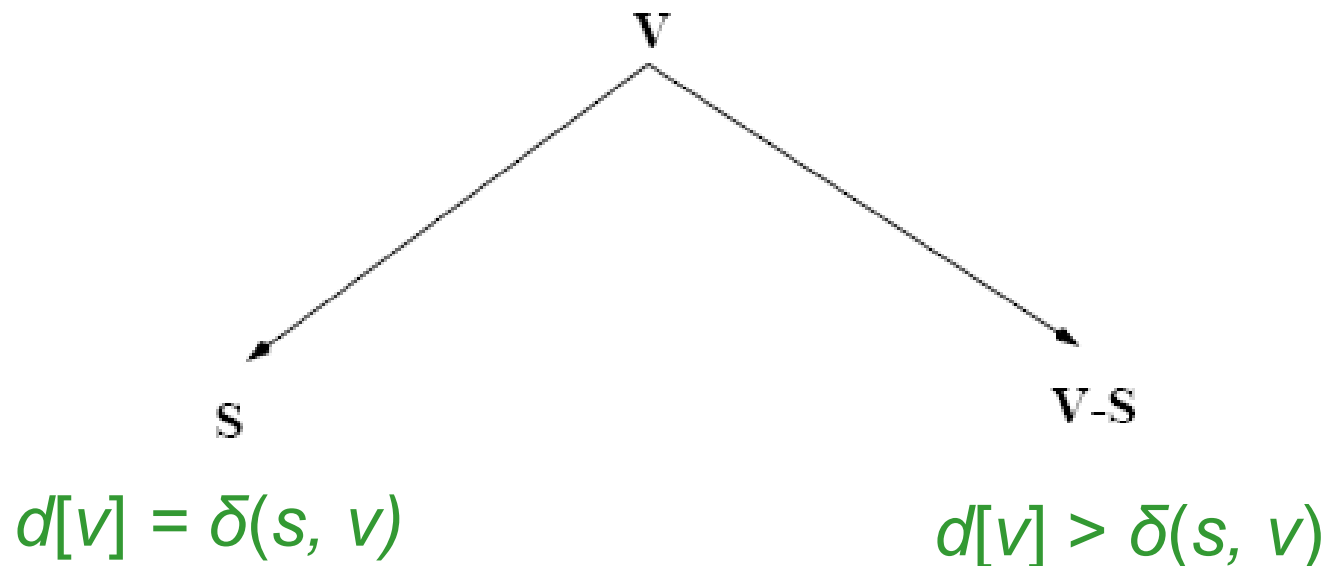
- Definition of shortest path problem
- Single-source shortest path
  - Bellman-Ford Algorithm
  - Dijkstra's Algorithm
- All-source shortest path
  - Floyd-Warshall algorithm for multiple paths

Single-Source Shortest Path without negative edge weight

# DIJKSTRA'S ALGORITHM

# Dijkstra's Algorithm

- Single-source shortest path problem:
  - No negative-weight edges:  $w(u, v) > 0, \forall (u, v) \in E$
- Each edge is relaxed **only once!**
- Maintains two sets of vertices:



# Dijkstra's Algorithm (cont.)

- Vertices in  $V-S$  reside in a min-priority queue
  - Keys in  $Q$  are estimates of shortest-path weights  $d[u]$
- Repeatedly select a vertex  $u \in V-S$ , with the minimum shortest-path estimate  $d[u]$
- Relax all edges leaving  $u$
- Steps
  - Extract a vertex  $u$  from  $Q$  (i.e., with the highest priority)
  - Insert  $u$  to  $S$
  - Relax all edges leaving  $u$
  - Update  $Q$

# Dijkstra Algorithm

Dijkstra( $G, r$ )

▷  $G=(V, E)$ : given graph

▷  $r$ : source vertex, given

*All the edge weights are non-negative*

```
{
    S ← {} ;                ▷ S : set of chosen vertices
    for each u ∈ V
        du ← ∞ ;
    dr ← 0 ;
    while (S ≠ V) {          ▷ repeat n times
        u ← extractMin(V-S, d) ;
        S ← S ∪ {u};
        for each v ∈ L(u) ▷ L(u) : set of vertices connected from u
            if (v ∈ V-S and dv < du + wu,v) then dv ← du + wu,v;
            relaxation
        }
    }
```

extractMin( $Q, d$ )

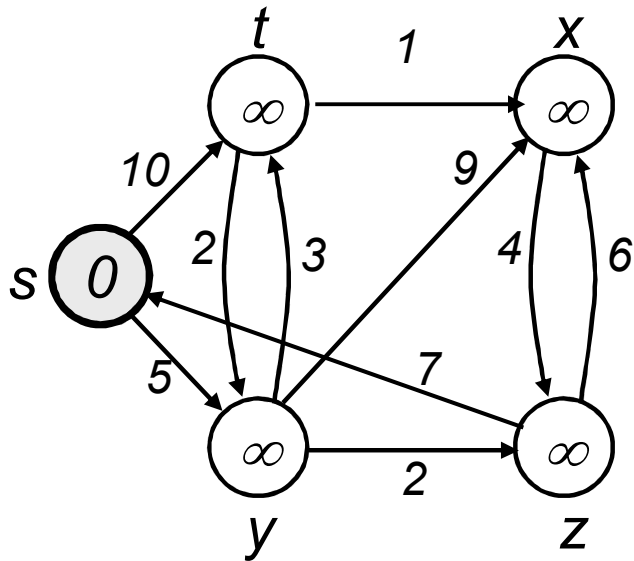
```
{
    Returns a vertex u ∈ Q whose value of d is the smallest;
}
```

✓ Time complexity:  $O(|E| \log |V|)$

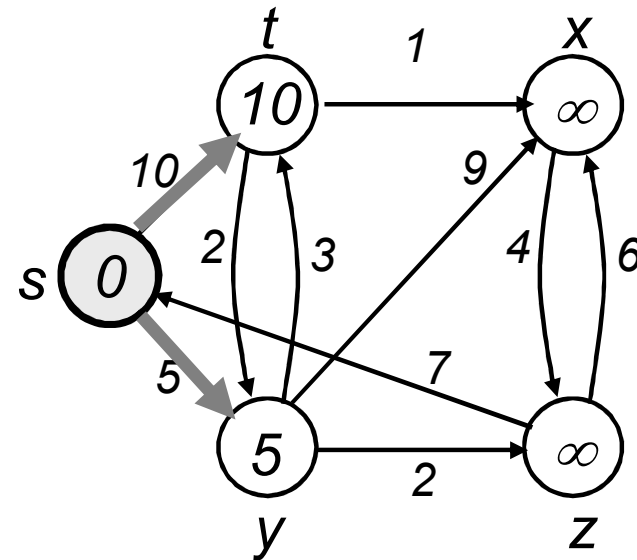
↑  
Using heap

# Dijkstra (G, w, s)

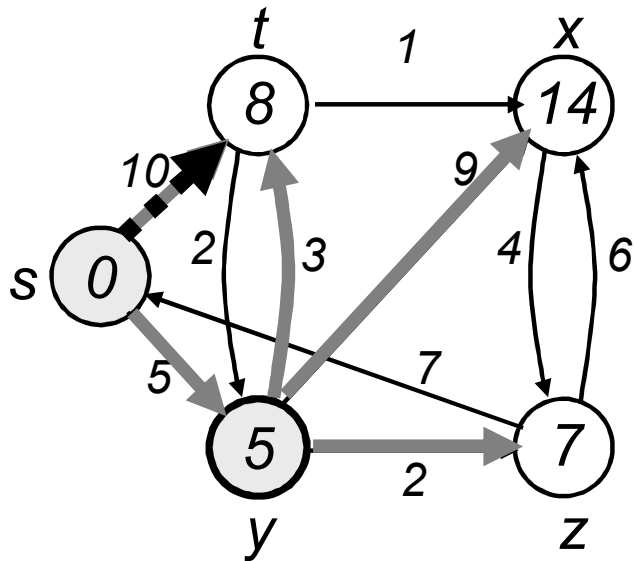
$S = \langle \rangle$   $Q = \langle s, t, x, z, y \rangle$



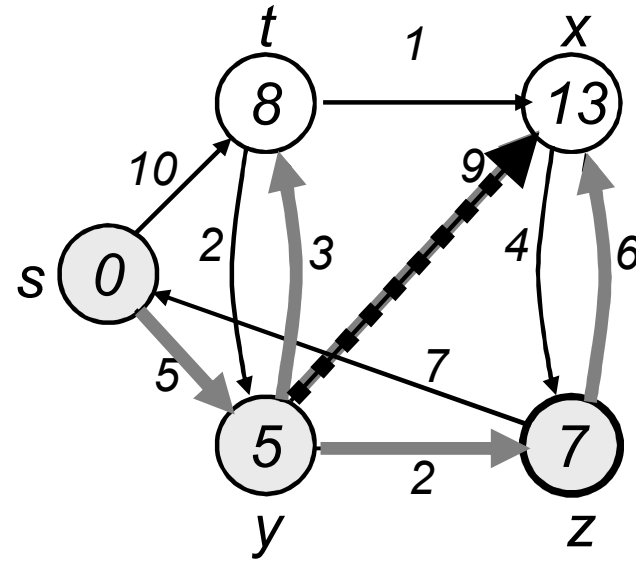
$S = \langle s \rangle$   $Q = \langle y, t, x, z \rangle$



# Example (cont.)



$S = \langle s, y \rangle$   $Q = \langle z, t, x \rangle$

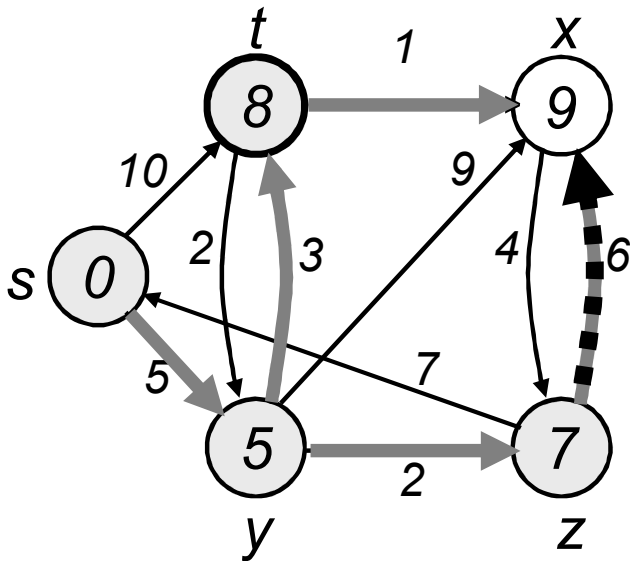


$S = \langle s, y, z \rangle$   $Q = \langle t, x \rangle$

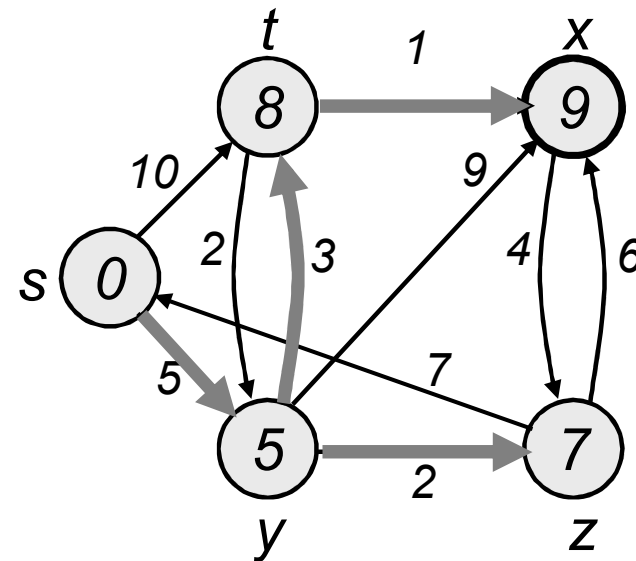


# Example (cont.)

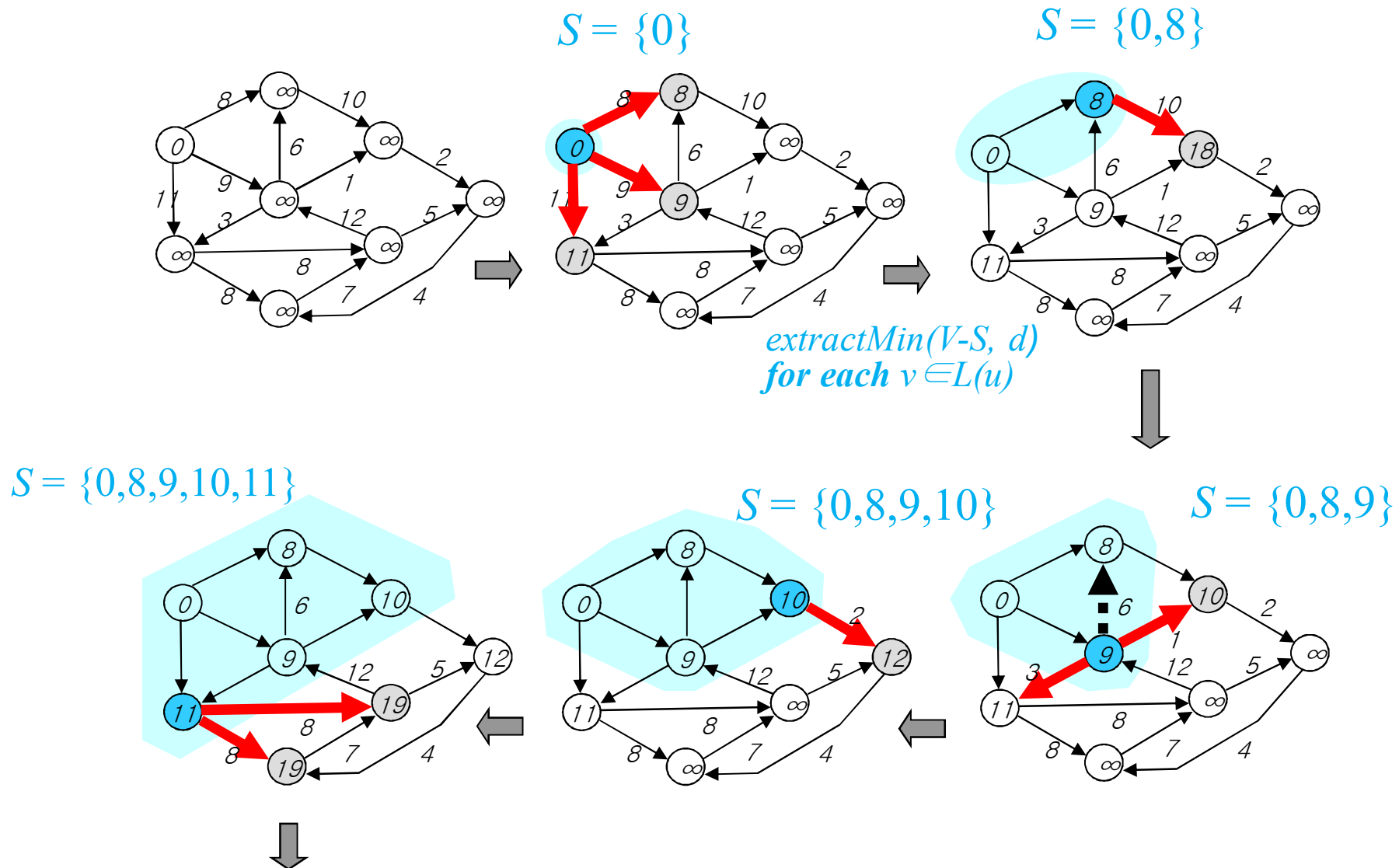
$S = \langle s, y, z, t \rangle$   $Q = \langle x \rangle$



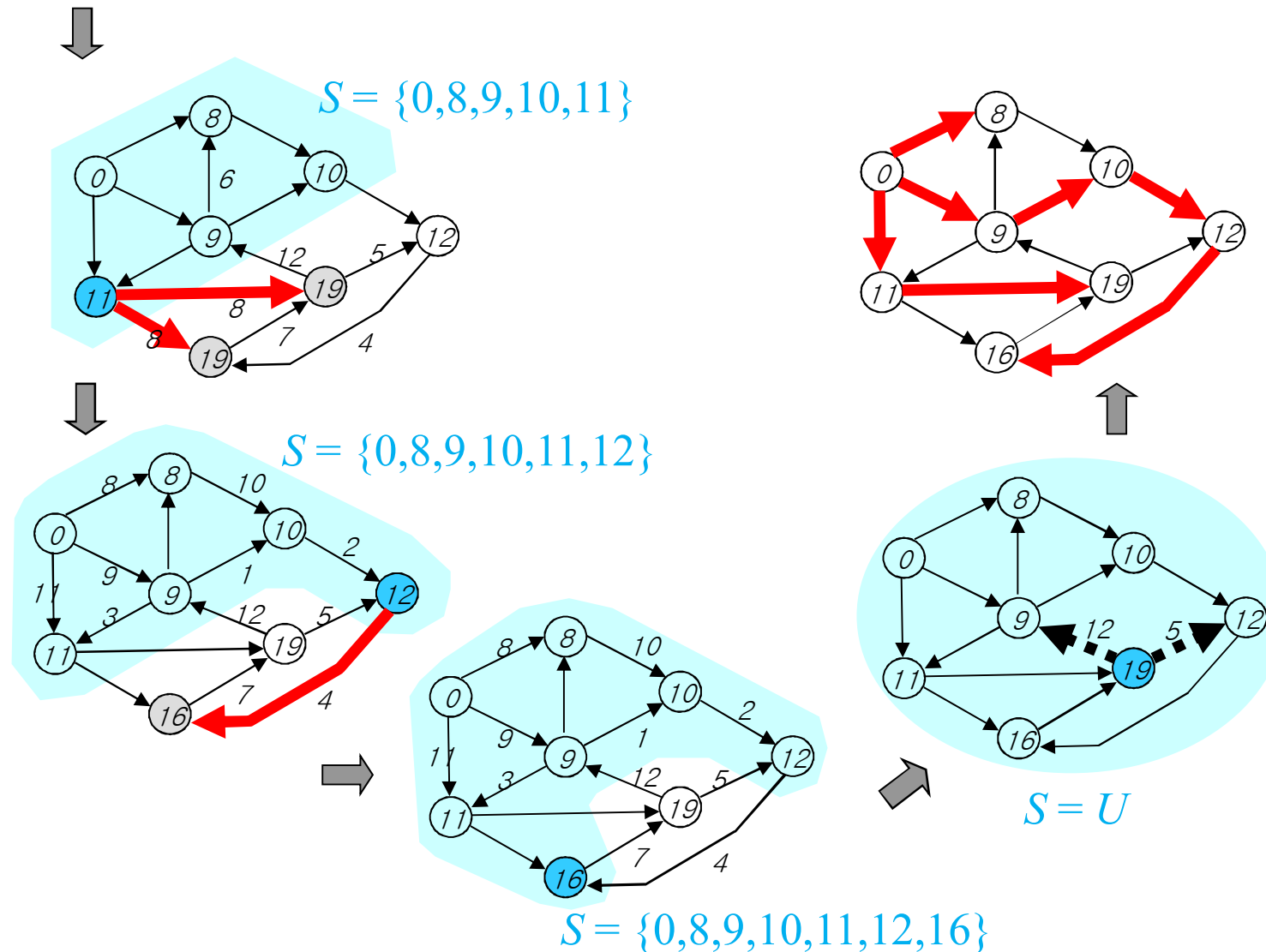
$S = \langle s, y, z, t, x \rangle$   $Q = \langle \rangle$



# Dijkstra Algorithm Illustration



# Dijkstra Algorithm Illustration



# Dijkstra ( $G, w, s$ )

1. INITIALIZE-SINGLE-SOURCE( $V, s$ )  $\leftarrow \Theta(V)$
2.  $S \leftarrow \emptyset$
3.  $Q \leftarrow V[G]$   $\leftarrow O(V)$  build min-heap
4. **while**  $Q \neq \emptyset$   $\leftarrow$  Executed  $O(V)$  times
5.     **do**  $u \leftarrow \text{EXTRACT-MIN}(Q)$   $\leftarrow O(\lg V)$   $\left. \vphantom{\begin{array}{l} 4. \\ 5. \end{array}} \right\} O(V \lg V)$
6.      $S \leftarrow S \cup \{u\}$
7.     **for** each vertex  $v \in \text{Adj}[u]$   $\leftarrow O(E)$  times
8.         **do** RELAX( $u, v, w$ )  $\left. \vphantom{\begin{array}{l} 7. \\ 8. \end{array}} \right\} \begin{array}{l} (total) \\ O(E \lg V) \end{array}$
9.     Update  $Q$  (DECREASE\_KEY)  $\leftarrow O(\lg V)$

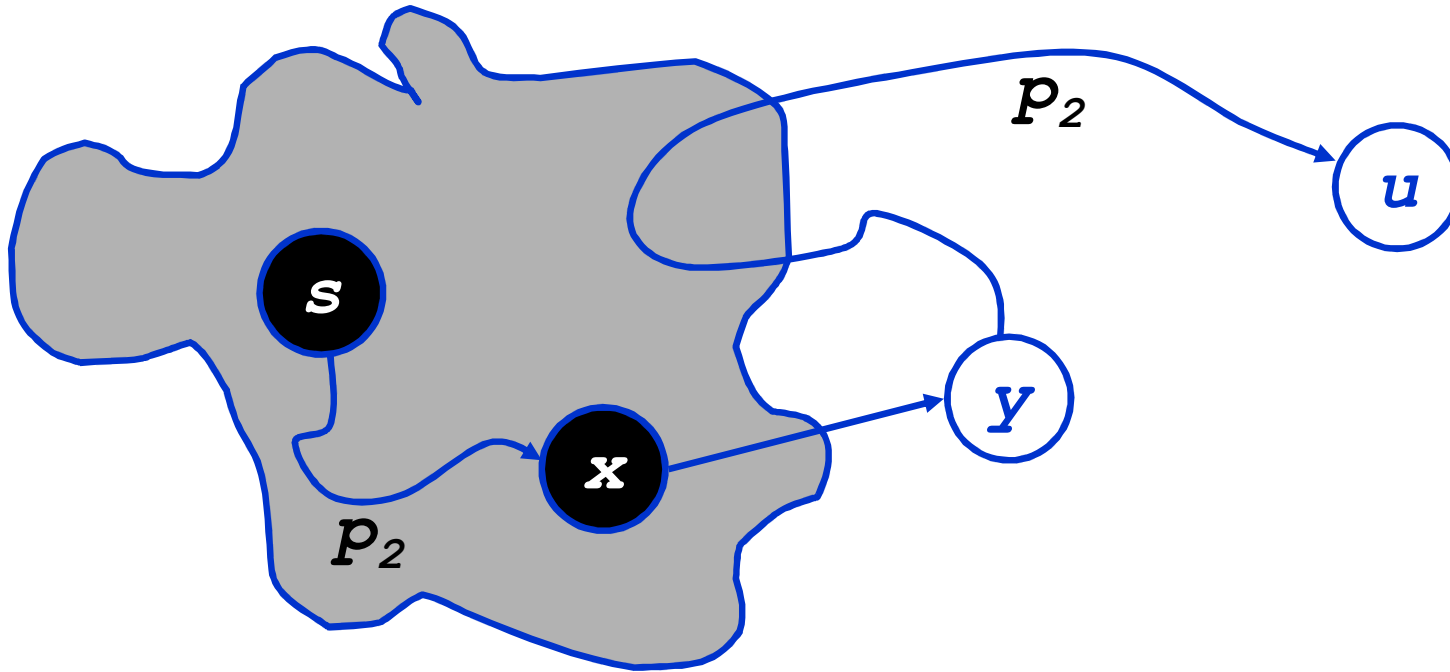
Running time:  $O(V \lg V + E \lg V) = O(E \lg V)$

# Binary Heap vs Fibonacci Heap

*Running time depends on the implementation of the heap*

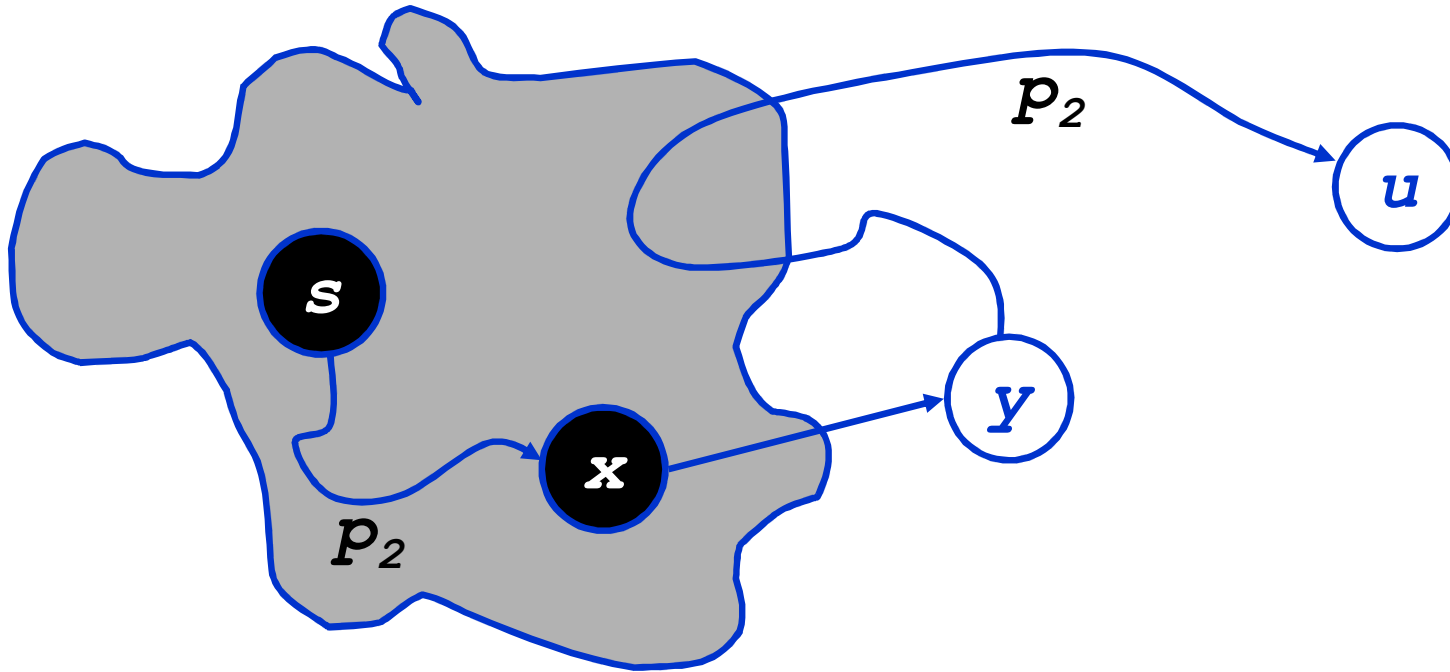
|                | <b>EXTRACT-MIN</b> | <b>DECREASE-KEY</b> | <b>Total</b>     |
|----------------|--------------------|---------------------|------------------|
| binary heap    | $O(\lg V)$         | $O(\lg V)$          | $O(E \lg V)$     |
| Fibonacci heap | $O(\lg V)$         | $O(1)$              | $O(V \lg V + E)$ |

# Correctness Of Dijkstra's Algorithm



- Note that  $d[v] \geq \delta(s,v) \forall v$
- Let  $u$  be first vertex picked s.t.  $\exists$  shorter path than  $d[u]$   $\Rightarrow d[u] > \delta(s,u)$
- Let  $y$  be first vertex  $\in V-S$  on actual shortest path from  $s \rightarrow u$   $\Rightarrow d[y] = \delta(s,y)$ 
  - Because  $d[x]$  is set correctly for  $y$ 's predecessor  $x \in S$  on the shortest path, and
  - When we put  $x$  into  $S$ , we relaxed  $(x,y)$ , giving  $d[y]$  the correct value

# Correctness Of Dijkstra's Algorithm



- Note that  $d[v] \geq \delta(s,v) \forall v$
- Let  $u$  be first vertex picked s.t.  $\exists$  shorter path than  $d[u]$   $\Rightarrow d[u] > \delta(s,u)$
- Let  $y$  be first vertex  $\in V-S$  on actual shortest path from  $s \rightarrow u$   $\Rightarrow d[y] = \delta(s,y)$
- $d[u] > \delta(s,u)$   
 $= \delta(s,y) + \delta(y,u)$  (*Why?*)  
 $= d[y] + \delta(y,u)$   
 $\geq d[y]$

But if  $d[u] > d[y]$ , wouldn't have chosen  $u$ . Contradiction.

문병로, 쉽게 배우는 알고리즘

Cevdet Aykanat and Mustafa Ozdal, CS473,  
Bilkent Univ

Andreas Klappenecker

# THE FLOYD-WARSHALL ALGORITHM



# All-Pairs Shortest Path Problem

- Suppose we are given a directed graph  $G$  and a weight function  $\mathbf{w}$ :
  - $G=(V,E)$ ;  $\omega: E \rightarrow \mathbb{R}$
- We assume that  $G$  does not contain cycles of weight 0 or less.
- The All-Pairs Shortest Path Problem asks to find the length of the shortest path between any pair of vertices in  $G$ .
- Applications
  - Network communications

# All Pairs Shortest Paths (APSP)

**given** : directed graph  $G = (V, E)$ ,  
weight function  $\omega: E \rightarrow R, |V| = n$

**goal** : create an  $n \times n$  matrix  $D = (d_{ij})$  of shortest path distances

$$\text{i.e., } d_{ij} = \delta(v_i, v_j)$$

**trivial solution** : run a SSSP algorithm  $n$  times, one for each vertex as the source.

# All Pairs Shortest Paths (APSP)

► all edge weights are nonnegative : use **Dijkstra's algorithm**

- **PQ = linear array** :  $O(V^3 + VE) = O(V^3)$
- **PQ = binary heap** :  $O(V^2 \lg V + EV \lg V) = O(V^3 \lg V)$  for dense graphs
  - better only for sparse graphs
- **PQ = fibonacci heap** :  $O(V^2 \lg V + EV) = O(V^3)$  for dense graphs
  - better only for sparse graphs

► negative edge weights : use **Bellman-Ford algorithm**

- $O(V^2 E) = O(V^4)$  on dense graphs

# Floyd-Warshall Algorithm

- A dynamic programming solution that solved the APSP problem with time complexity  $O(n^3)$  for a graph with  $n$  vertices.
  - Dynamic programming

# Adjacency Matrix Representation of Graphs

►  $n \times n$  matrix  $\mathbf{W} = (\omega_{ij})$  of edge weights :

$$\omega_{ij} = \begin{cases} \omega(\mathbf{v}_i, \mathbf{v}_j) & \text{if } (\mathbf{v}_i, \mathbf{v}_j) \in E \\ \infty & \text{if } (\mathbf{v}_i, \mathbf{v}_j) \notin E \end{cases}$$

► assume  $\omega_{ii} = 0$  for all  $\mathbf{v}_i \in \mathbf{V}$ , because

- no neg-weight cycle

$\Rightarrow$  shortest path to itself has no edge,

i.e.,  $\delta(\mathbf{v}_i, \mathbf{v}_i) = 0$

# Intermediate Vertices

- Without loss of generality, we will assume that  $V=\{1,2,\dots,n\}$ , i.e., that the vertices of the graph are numbered from 1 to  $n$ .
- Given a path  $p=(v_1, v_2, \dots, v_m)$  in the graph, we will call the vertices  $v_k$  with index  $k$  in  $\{2, \dots, m-1\}$  the **intermediate vertices** of path  $p$ .
- If  $k$  is an intermediate vertex of path  $p$ , then we break  $p$  down into:
  - $i \rightarrow k \rightarrow j$  ( $p_1$  and  $p_2$ )
  - $p_1$  is a shortest path from  $i$  to  $k$  with all intermediate vertices in the set  $\{1, 2, \dots, k-1\}$
  - $p_2$  is a shortest path from  $k$  to  $j$  with all intermediate vertices in the set  $\{1, 2, \dots, k-1\}$

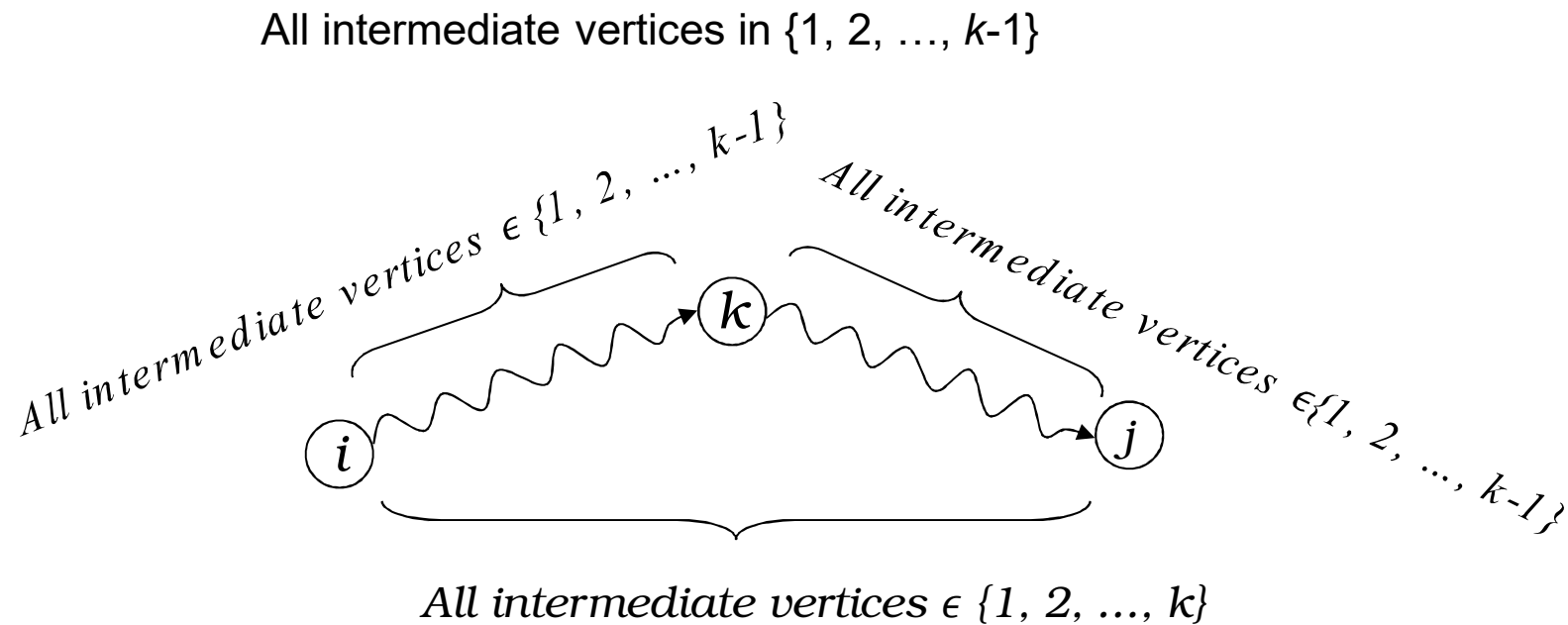


Figure 2. Path  $p$  is a shortest path from vertex  $i$  to vertex  $j$ , and  $k$  is the highest-numbered intermediate vertex of  $p$ . Path  $p_1$ , the portion of path  $p$  from vertex  $i$  to vertex  $k$ , has all intermediate vertices in the set  $\{1, 2, \dots, k-1\}$ . The same holds for path  $p_2$  from vertex  $k$  to vertex  $j$ .

# Key Idea

- Let  $d_{ij}^{(k)}$  denote the length of the shortest path from  $i$  to  $j$  such that all intermediate vertices are contained in the set  $\{1, \dots, k\}$ .
- Consider a shortest path  $p$  from  $i$  to  $j$  such that the intermediate vertices are from the set  $\{1, \dots, k\}$ .
  - 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  $d_{ij}^{(k)} = d_{ik}^{(k-1)} + d_{kj}^{(k-1)}$ .
  - Interestingly, in either case, the subpaths contain merely nodes from  $\{1, \dots, k-1\}$ .



# Recursive Formulation

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

$$d_{ij}^{(0)} = w_{ij}$$

If  $k>0$ , then

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

# The Floyd-Warshall Algorithm

Floyd-Warshall( $W$ )

$n$  = # of rows of  $W$ ;

$D^{(0)} = W$ ;     /\* initialization, equivalent to the next nested loop \*/

/\* for  $(i,j) = (1, 1)$  to  $(n, n)$  do;  $d_{ij}^{(k)} = w_{ij}$ ; od; \*/

for  $k = 1$  to  $n$  do

    for  $i = 1$  to  $n$  do

        for  $j = 1$  to  $n$  do

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

        od;

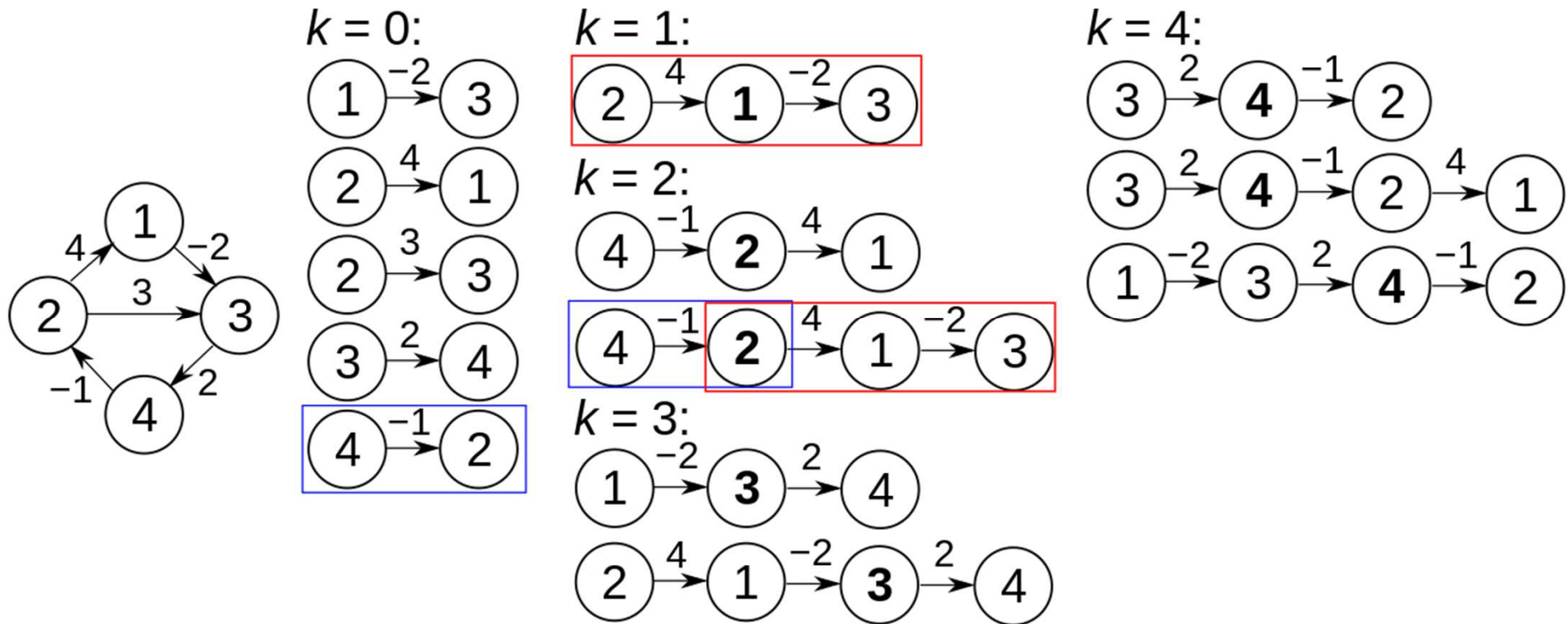
    od;

od;

return  $D^{(n)}$ ;

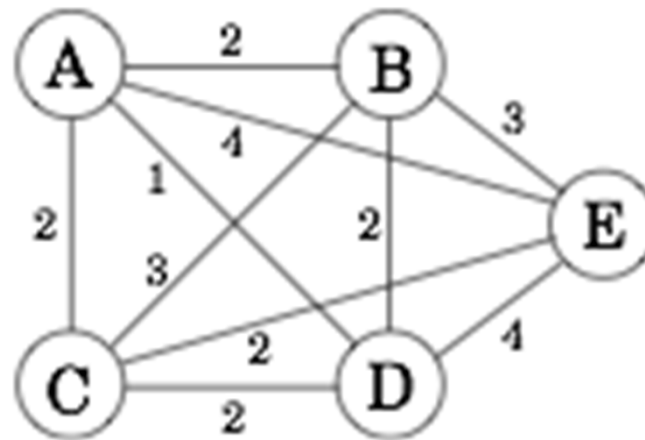
✓  $d_{ij}^{(k)}$  : shortest path from  $i$  to  $j$  using intermediate edges in  $\{1, 2, \dots, k\}$

# Example: Floyd-Warshall Algorithm



# Quiz

Run Floyd-Warshall algorithm for the following graph:



# Alternate Implementation of the Floyd-Warshall

**EXTEND** (  $D$  ,  $W$  )

►  $D = (d_{ij})$  is an  $n \times n$  matrix

**for**  $i \leftarrow 1$  **to**  $n$  **do**

**for**  $j \leftarrow 1$  **to**  $n$  **do**

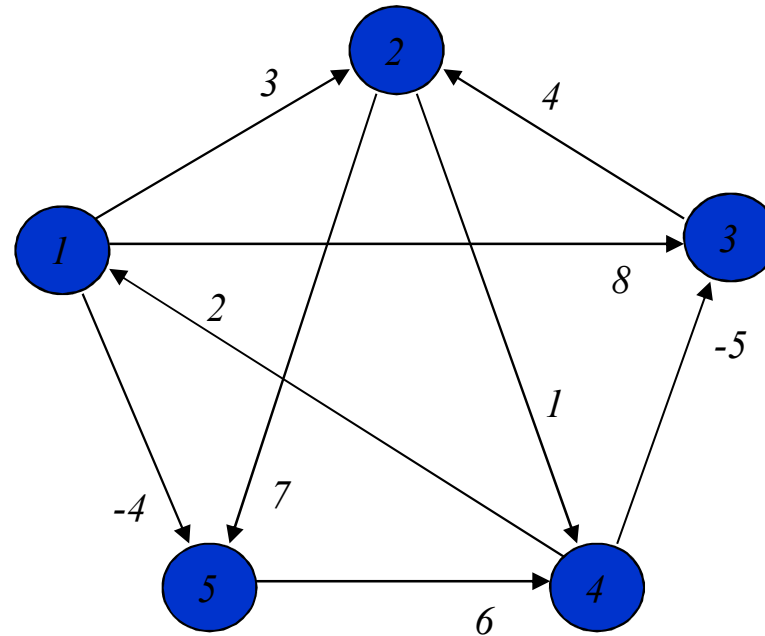
$d_{ij} \leftarrow \infty$

**for**  $k \leftarrow 1$  **to**  $n$  **do**

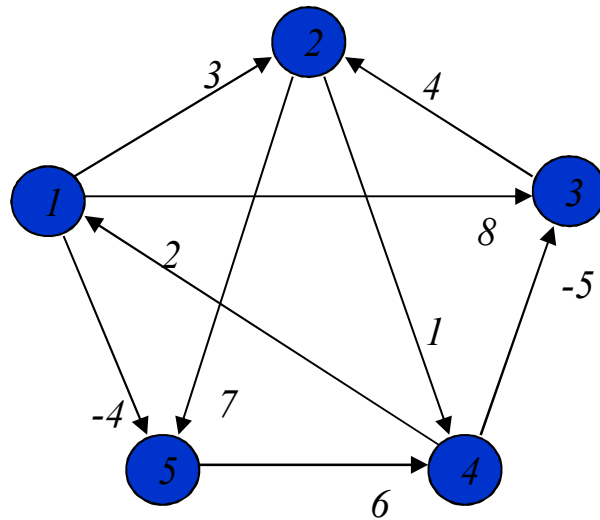
$d_{ij} \leftarrow \min \{d_{ij}, d_{ik} + \omega_{kj}\}$

**return**  $D$

# Floyd-Warshall: EXTEND



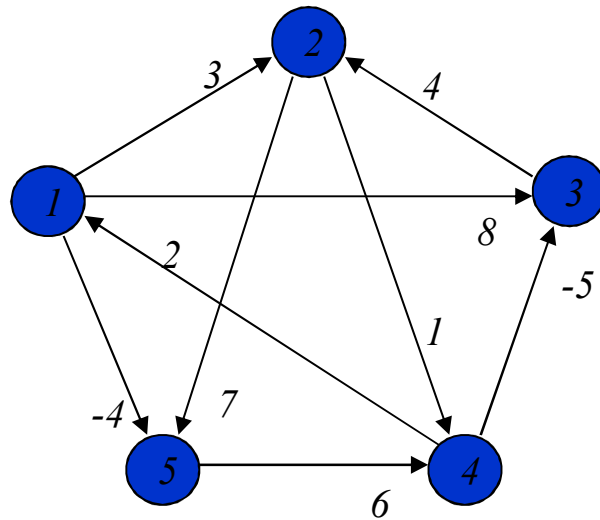
# Floyd-Warshall: EXTEND



|   | 1        | 2        | 3        | 4        | 5        |
|---|----------|----------|----------|----------|----------|
| 1 | 0        | 3        | 8        | $\infty$ | -4       |
| 2 | $\infty$ | 0        | $\infty$ | 1        | 7        |
| 3 | $\infty$ | 4        | 0        | $\infty$ | $\infty$ |
| 4 | 2        | $\infty$ | -5       | 0        | $\infty$ |
| 5 | $\infty$ | $\infty$ | $\infty$ | 6        | 0        |

$$D^1 = D^0 W$$

# Floyd-Warshall: EXTEND

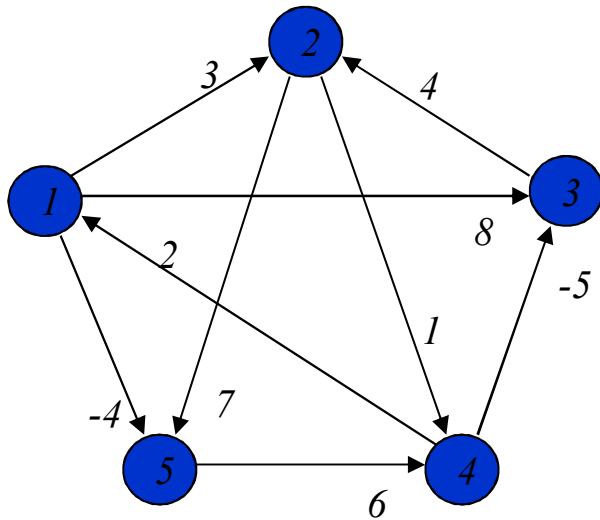


|   | 1        | 2        | 3  | 4 | 5  |
|---|----------|----------|----|---|----|
| 1 | 0        | 3        | 8  | 2 | -4 |
| 2 | 3        | 0        | -4 | 1 | 7  |
| 3 | $\infty$ | 4        | 0  | 5 | 11 |
| 4 | 2        | -1       | -5 | 0 | -2 |
| 5 | 8        | $\infty$ | 1  | 6 | 0  |

$$D^2 = D^1 W$$



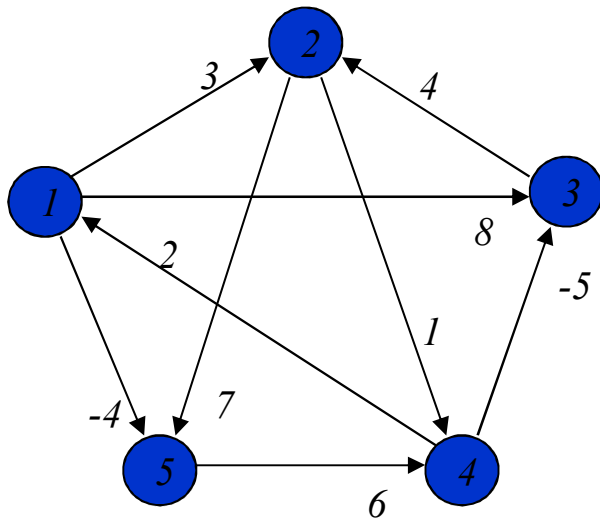
# Floyd-Warshall: EXTEND



|   | 1 | 2  | 3  | 4 | 5  |
|---|---|----|----|---|----|
| 1 | 0 | 3  | -3 | 2 | -4 |
| 2 | 3 | 0  | -4 | 1 | -1 |
| 3 | 7 | 4  | 0  | 5 | 11 |
| 4 | 2 | -1 | -5 | 0 | -2 |
| 5 | 8 | 5  | 1  | 6 | 0  |

$$D^3 = D^2 W$$

# Floyd-Warshall: EXTEND



|   | 1 | 2  | 3  | 4 | 5  |
|---|---|----|----|---|----|
| 1 | 0 | 1  | -3 | 2 | -4 |
| 2 | 3 | 0  | -4 | 1 | -1 |
| 3 | 7 | 4  | 0  | 5 | 3  |
| 4 | 2 | -1 | -5 | 0 | -2 |
| 5 | 8 | 5  | 1  | 6 | 0  |

$$D^4 = D^3 W$$

# Time and Space Requirements

The running time is obviously  $O(n^3)$ .

However, in this version, the space requirements are high. One can reduce the space from  $O(n^3)$  to  $O(n^2)$  by using a single array  $d$ .

# Conclusion

- Negative weights and negative cycles.
  - If no negative cycles, can find shortest paths via Bellman-Ford.
  - If negative cycles, can find one via Bellman-Ford.
- Dijkstra's algorithm.
  - Nearly linear-time when weights are nonnegative.
- All-pair shortest path.
  - can be solved via Floyd-Warshall
  - Floyd-Warshall can also compute the transitive closure of directed graph.

Next topic: NP-Completeness

**END OF LECTURE 17**