

APSP

Computer Algorithm

TY-IT

Agenda

- Review of
 - Prim's algorithm for MST
 - Dijkstra's algorithm
 - Example
- APSP consideration
 - Matrix operation
 - Repeated squaring
- Floyd Warshall algorithm
- Transitive closure

APSP

- Weight matrix
 - X axis- source, Y axis- Destination
 - Matrix element $W=w_{i,j}$ - weight of edge from i to j
- For each pair (i,j) see if there is shortest path through other vertex k
- $L^{(m)}_{i,j}$ be shortest path length from i,j with max m edges
- Compute $L^{(1)}_{i,j}$, $L^{(2)}_{i,j}$ $L^{(m)}_{i,j}$

APSP

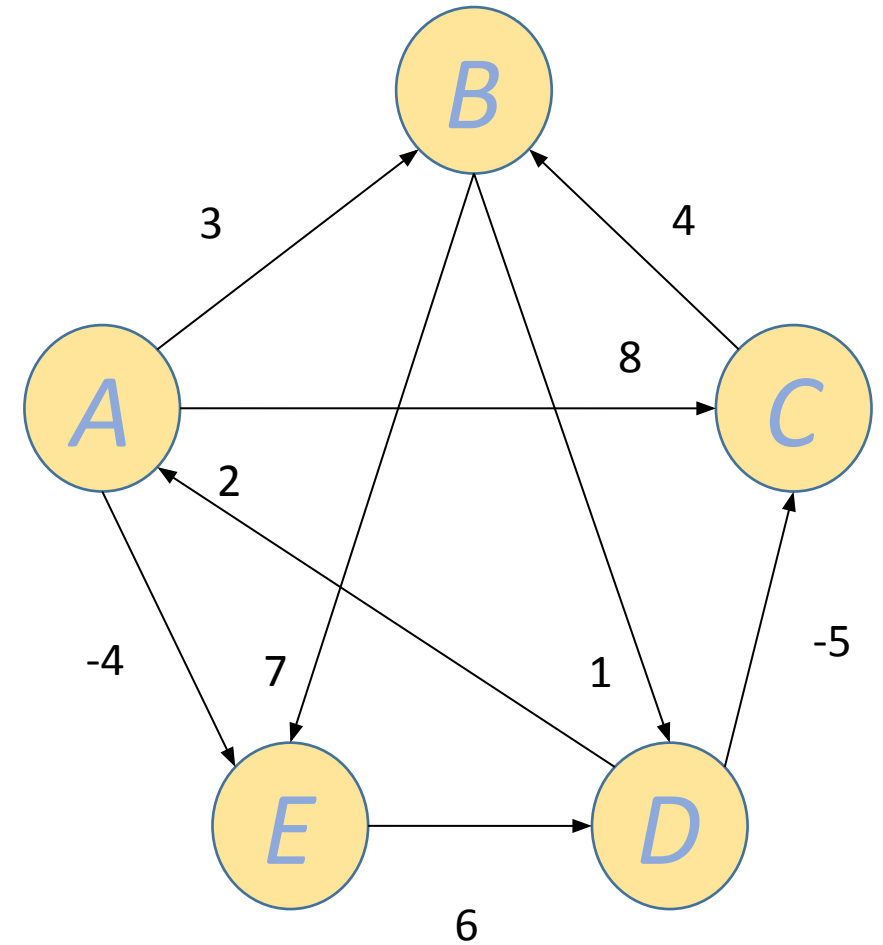
EXTEND-SHORTEST-PATHS (L, W)

```
1   $n \leftarrow \text{rows}[L]$ 
2  let  $L' = (l'_{ij})$  be an  $n \times n$  matrix
3  for  $i \leftarrow 1$  to  $n$ 
4      do for  $j \leftarrow 1$  to  $n$ 
5          do  $l'_{ij} \leftarrow \infty$ 
6              for  $k \leftarrow 1$  to  $n$ 
7                  do  $l'_{ij} \leftarrow \min(l'_{ij}, l_{ik} + w_{kj})$ 
8  return  $L'$ 
```

- Repeat for $n-1$ times
- Complexity n^4
- Optimization- Repeated squaring $-n^3 \log n$

APSP ANIMATED EXAMPLE

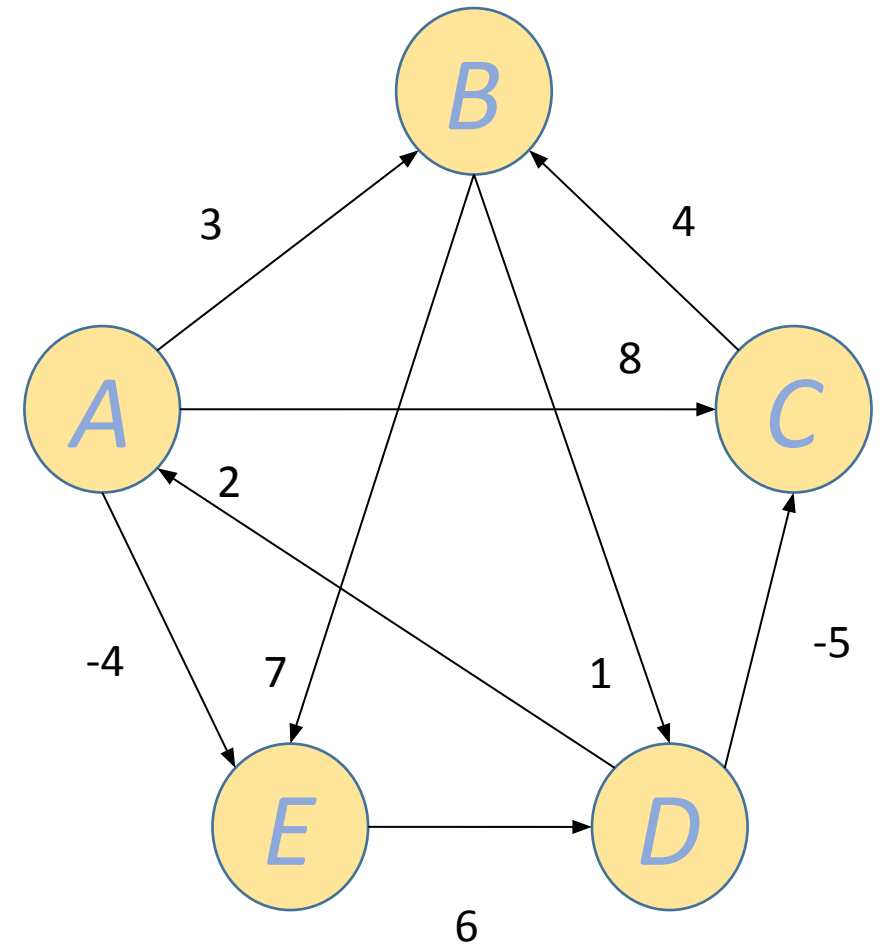
		<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
	<i>A</i>	0	3	8	∞	-4
	<i>B</i>	∞	0	∞	1	7
<i>W</i>	<i>C</i>	∞	4	0	∞	∞
	<i>D</i>	2	∞	-5	0	∞
	<i>E</i>	∞	∞	∞	6	0



APSP ANIMATED EXAMPLE

		<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
	<i>A</i>	0	3	8		-4
	<i>B</i>		0		1	7
$L^{(1)}$	<i>C</i>		4	0		
	<i>D</i>	2		-5	0	
	<i>E</i>				6	0

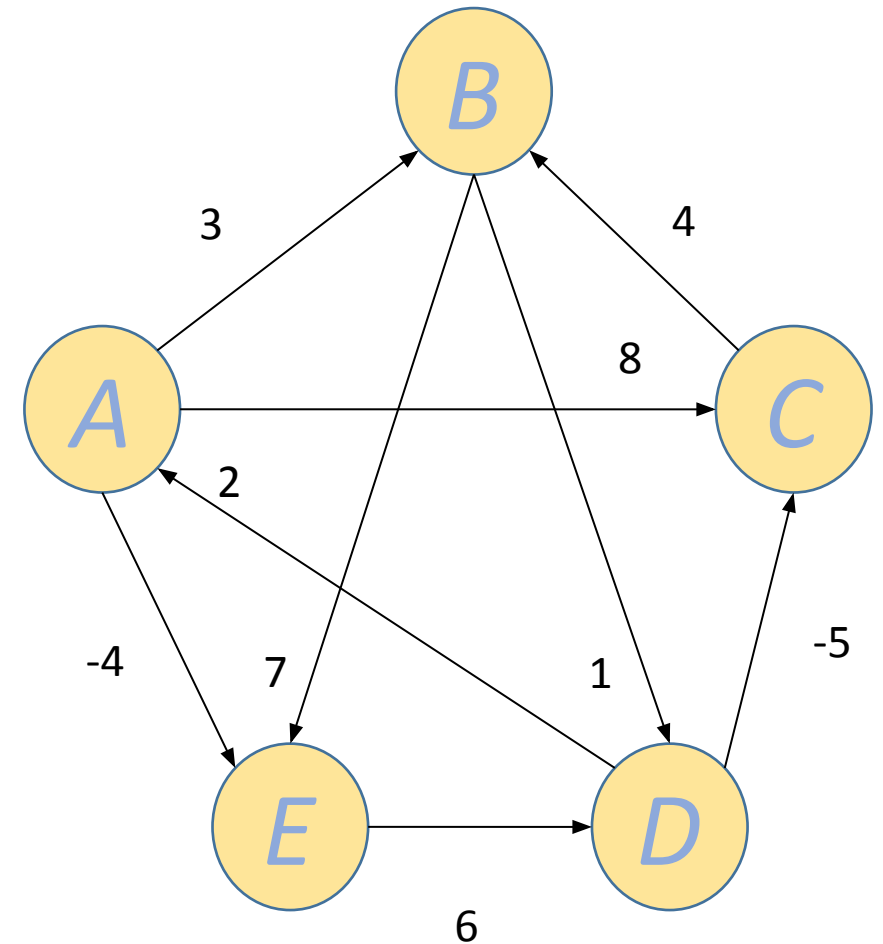
		<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
	<i>A</i>	0	3	8	2	-4
	<i>B</i>	3	0	-4	1	7
$L^{(2)}$	<i>C</i>		4	0	5	11
	<i>D</i>	2	-1	-5	0	-2
	<i>E</i>	8		1	6	0



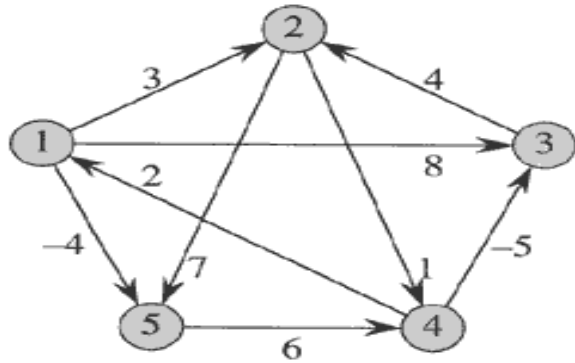
APSP ANIMATED EXAMPLE

		<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
	<i>A</i>	0	3	-3	2	-4
	<i>B</i>	3	0	-4	1	-1
$L^{(3)}$	<i>C</i>	7	4	0	5	11
	<i>D</i>	2	-1	-5	0	-2
	<i>E</i>	8	5	1	6	0

		<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>
	<i>A</i>	0	1	-3	2	-4
	<i>B</i>	3	0	-4	1	-1
$L^{(4)}$	<i>C</i>	7	4	0	5	3
	<i>D</i>	2	-1	-5	0	-2
	<i>E</i>	8	5	1	6	0



APSP: Example



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

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

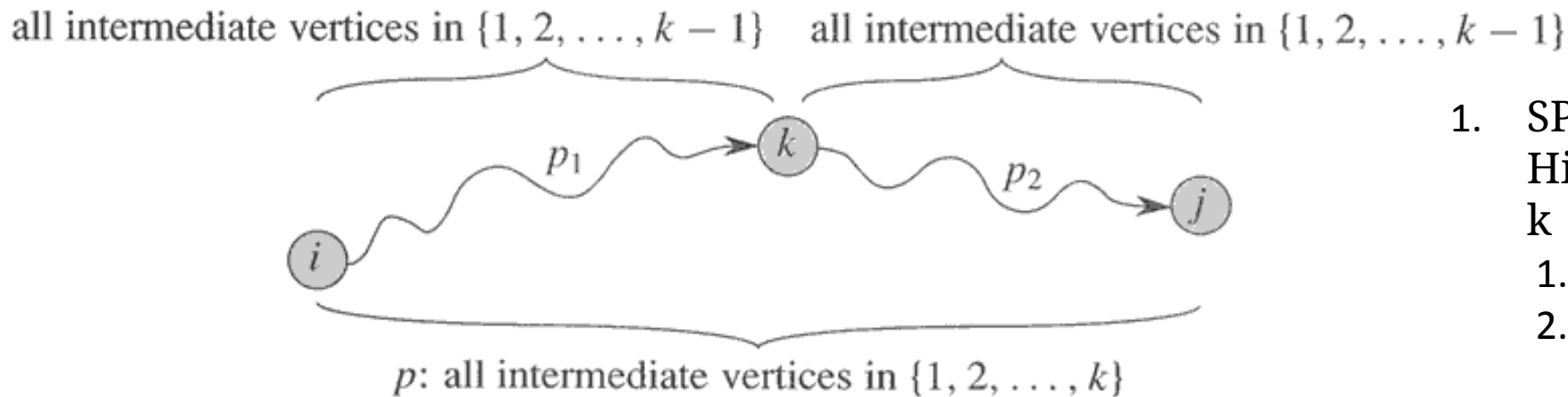
Figure 25.1 A directed graph and the sequence of matrices $L^{(m)}$ computed by SLOW-ALL-PAIRS-SHORTEST-PATHS. The reader may verify that $L^{(5)} = L^{(4)} \cdot W$ is equal to $L^{(4)}$, and thus $L^{(m)} = L^{(4)}$ for all $m \geq 4$.

APSP

- Repeat for $n-1$ times
- Complexity n^4
- Optimization- Repeated squaring $-n^3 \log n$

Floyd Warshall Algorithm

- Index of vertex
- Intermediate vertex



1. SP from i to j may go thr Highest Index Vertex (HIV) k
 1. No – HIV $k-1$
 2. Yes, consider two sub paths $\{i \text{ to } k, k \text{ to } j\}$
2. Compute $d_{ij}^{(k)}$, start from $d_{ij}^{(0)}$

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

If $(d_{ij}^{(k-1)} > d_{ik}^{(k-1)} + d_{kj}^{(k-1)})$ then $\pi_{ij}^{(k)} := \pi_{ik}^{(k-1)} \pi_{kj}^{(k-1)}$

Flyod Warshall (FW) Algorithm

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

7 **return** $D^{(n)}$

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

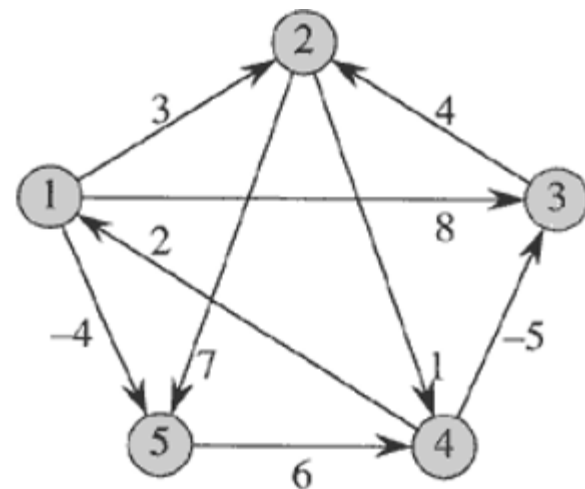
$$D^{(1)} = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix} \quad \Pi^{(1)} = \begin{pmatrix} \text{NIL} & 1 & 1 & \text{NIL} & 1 \\ \text{NIL} & \text{NIL} & \text{NIL}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

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

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

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

$$D^{(5)} = \begin{pmatrix} 0 & 1 & -3 & 2 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \end{pmatrix} \quad \Pi^{(5)} = \begin{pmatrix} \text{NIL} & 3 & 4 & 5 & 1 \\ 4 & \text{NIL} & 4 & 2 & 1 \\ 4 & 3 & \text{NIL} & 2 & 1 \\ 4 & 3 & 4 & \text{NIL} & 1 \end{pmatrix}$$



Transitive closure

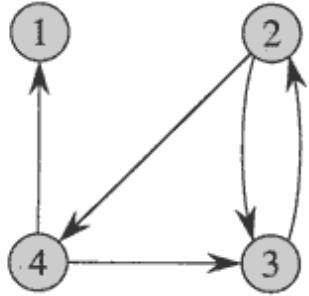
- Adjacency matrix to denote connectivity between nodes
 - 1- Path exists
 - 0- No path
- Floyd Warshall algorithm can be used
- Start from $t_{ij}^{(0)}$
- $t_{ij}^{(k)} := t_{ij}^{(k-1)} \vee (t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)})$

Transitive closure (TC)

TRANSITIVE-CLOSURE(G)

```
1   $n \leftarrow |V[G]|$ 
2  for  $i \leftarrow 1$  to  $n$ 
3      do for  $j \leftarrow 1$  to  $n$ 
4          do if  $i = j$  or  $(i, j) \in E[G]$ 
5              then  $t_{ij}^{(0)} \leftarrow 1$ 
6              else  $t_{ij}^{(0)} \leftarrow 0$ 
7  for  $k \leftarrow 1$  to  $n$ 
8      do for  $i \leftarrow 1$  to  $n$ 
9          do for  $j \leftarrow 1$  to  $n$ 
10             do  $t_{ij}^{(k)} \leftarrow t_{ij}^{(k-1)} \vee (t_{ik}^{(k-1)} \wedge t_{kj}^{(k-1)})$ 
11  return  $T^{(n)}$ 
```

TC: Example



$$T^{(0)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \end{pmatrix} \quad T^{(1)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \end{pmatrix} \quad T^{(2)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \end{pmatrix}$$

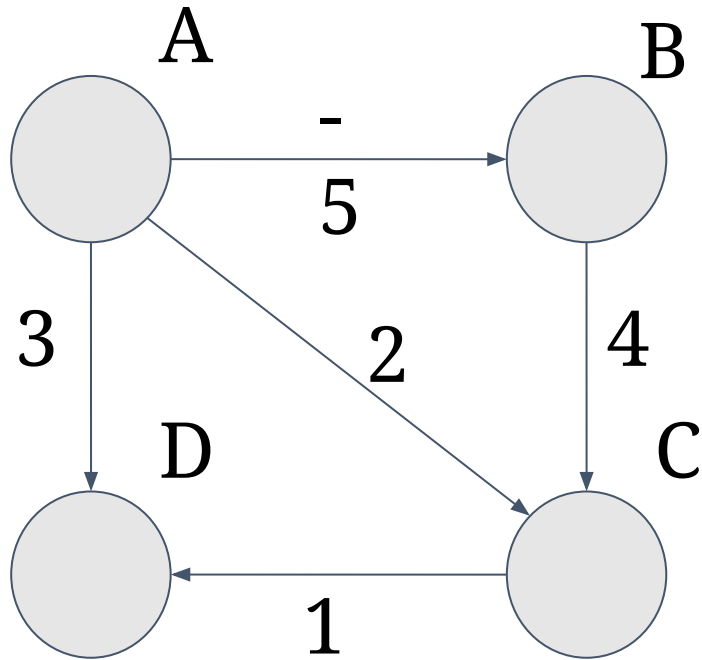
$$T^{(3)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} \quad T^{(4)} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix}$$

Figure 25.5 A directed graph and the matrices $T^{(k)}$ computed by the transitive-closure algorithm.

Johnson algorithm

- Used for sparse graph
- Let the given graph be G . Add a new vertex s to the graph, add edges from s to all vertices of G . Let the modified graph be G' .
- Run [Bellman-Ford algorithm](#) on G' with s as source. Let the distances calculated by Bellman-Ford be $h[0], h[1], \dots, h[V-1]$.
- Reweight : For each edge (u, v) , assign the new weight as “original weight + $h[u] - h[v]$ ”.
- Remove the added vertex s and run [Dijkstra's algorithm](#) for every vertex.

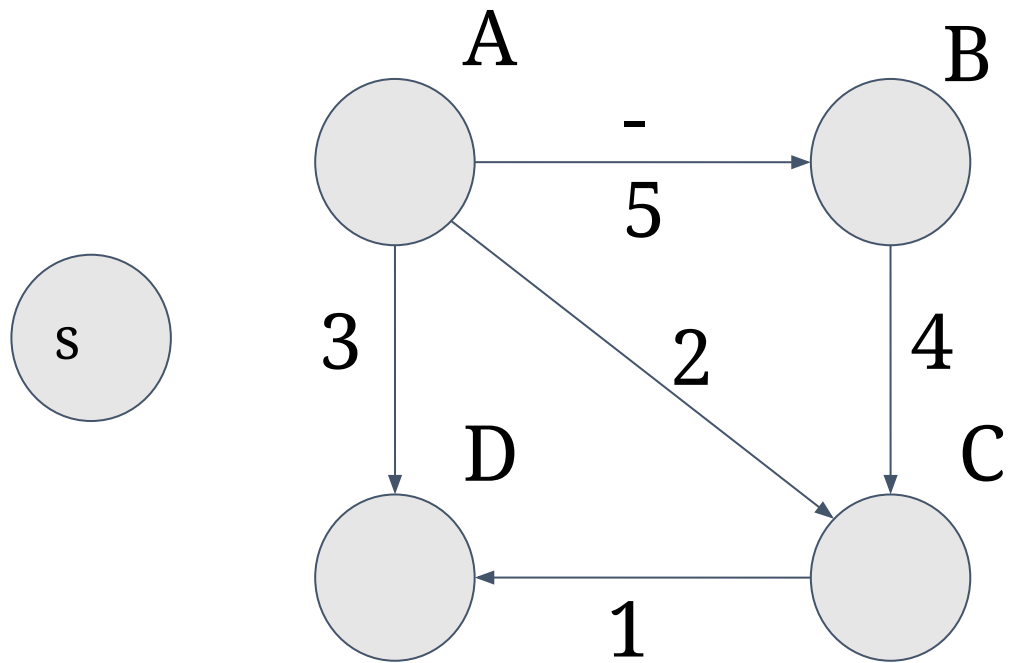
Johnson Algorithm: Example



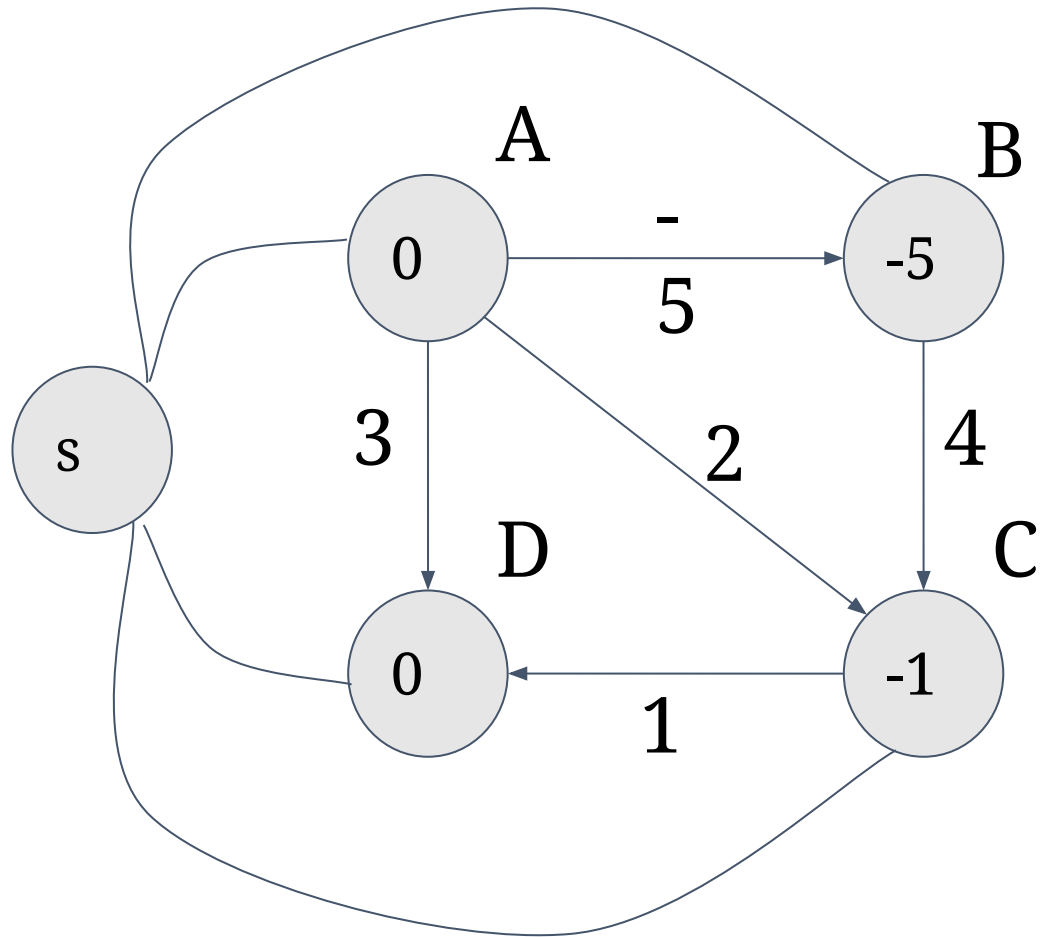
The reason that Johnson's algorithm is better for sparse graphs is that its time complexity depends on the number of edges in the graph, while Floyd-Warshall's does not. Johnson's algorithm runs in $O(V^2 \log(V) + |V| |E|)$ time.

So, if the number of edges is small (i.e. the graph is sparse), it will run faster than the $O(V^3)$ runtime of Floyd-Warshall.

Johnson Algorithm: Example

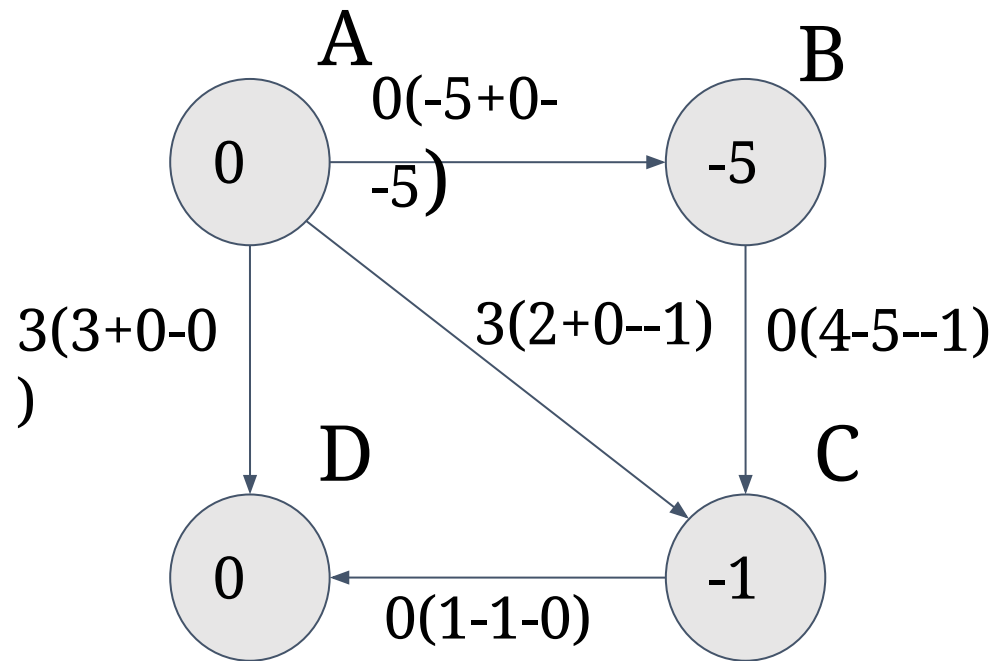


Johnson Algorithm: Example



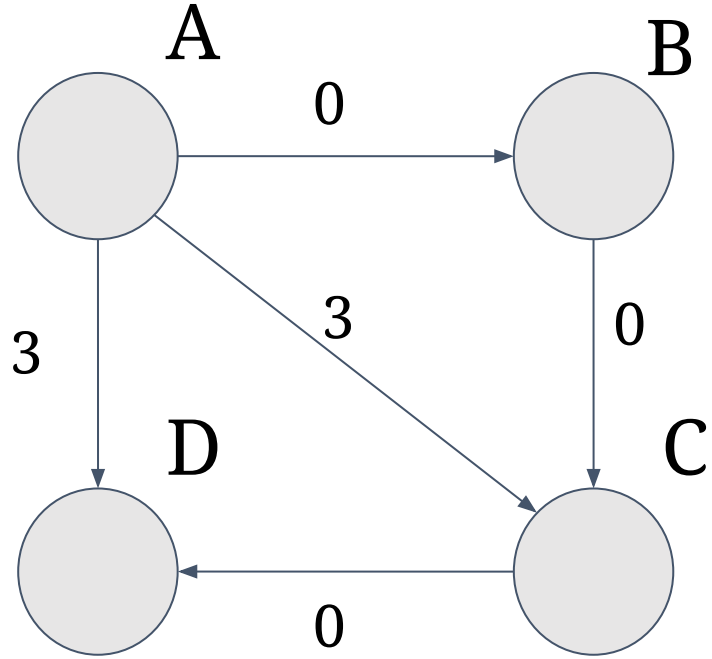
- Apply Bellman Ford Algorithm from s

Johnson Algorithm: Example



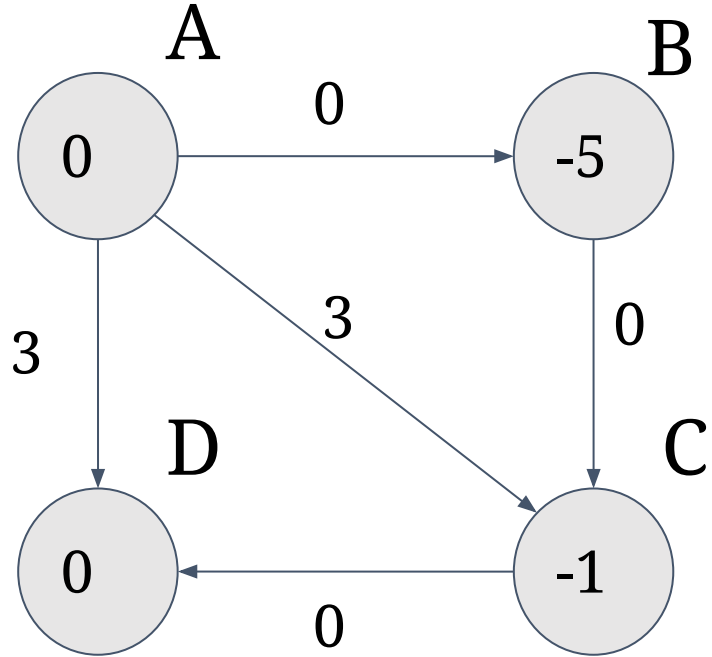
- remove the source vertex s
- reweight the edges using following formula. $w(u, v) = w(u, v) + h[u] - h[v]$

Johnson Algorithm: Example



- Run Dijkstra's algorithm from every vertex

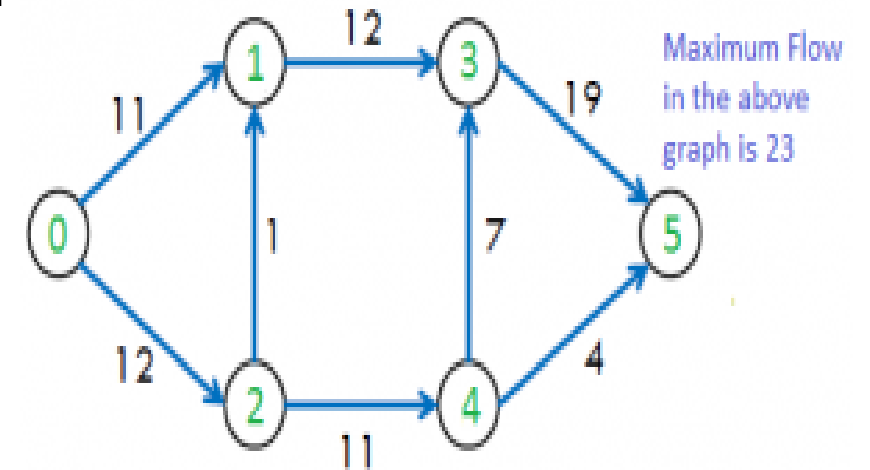
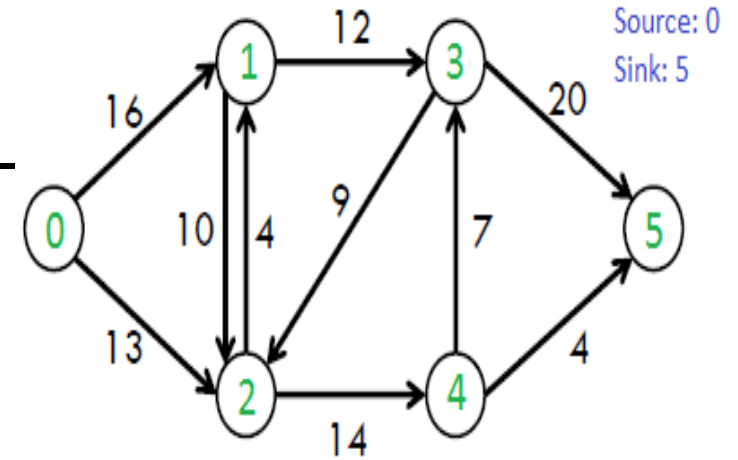
Johnson Algorithm: Example



- Run Dijkstra's algorithm from every vertex

Maximum flow: Example

- Maximum flow problems involve finding a feasible flow through a single-source, single-sink flow network that is maximum.
- Each edge is labeled with capacity, the maximum amount of stuff that it can carry. The goal is to figure out how much stuff can be pushed from the vertex s (source) to the vertex t (sink).
- maximum flow possible is : **23**



- Flow network
 - Directed graph
 - One source, sink
 - Non negative weights edge
- Residual graph G_f
 - Edges from G that have some non zero flow
 - $C_f(u,v) = c(u,v) - f(u,v)$
 - Add the edges: Reverse edges to decrease the flow
 - $C_f(v,u) = c(v,u) + f(u,v)$
- Augmenting path (p)
 - Path from s to t
 - Residual capacity: Smallest capacity of path p

Ford Fulkerson (FF) Algorithm

- $f_m = 0$
- While there exists augmenting path
 - Identify augmenting path
 - $C_f(p)$ = smallest capacity of p
 - $f_m = f_m + C_f(p)$
 - For each edge in p
 - $C_f(u,v) = c(u,v) - f(u,v)$
 - $C_f(v,u) = c(v,u) + f(u,v)$

FF Algorithm: Example

- <https://www.youtube.com/watch?v=TI90tNtKvxs>

FF Algorithm: Complexity

- Flow increased by at least one unit in every iteration
 - max value of f_m -max flow times
 - Finding augmenting path- $O(E)$
- Total complexity: $O(E * f_m)$

Min cut

- **Minimal Cut** : It is one which replacement of any of its member reconnects the Graph.
- **Minimum Cut** : In a weighted graph each cut has capacity. A cut with minimum capacity is minimum cut.
- **Max Flow Min Cut Theorem** : The maximum flow between any two arbitrary nodes in any graph cannot exceed the capacity of the minimum cut separating those two nodes.

- Thank you