

Transitive closure of graphs and all-pairs shortest paths

# Transitive closure (accessibility)

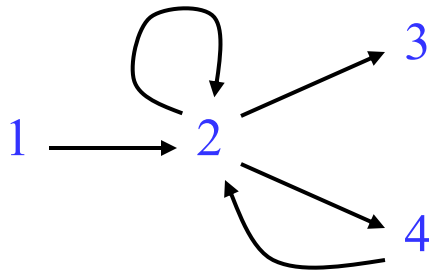
## Problem:

$G = (V, E)$  (unweighted) directed graph

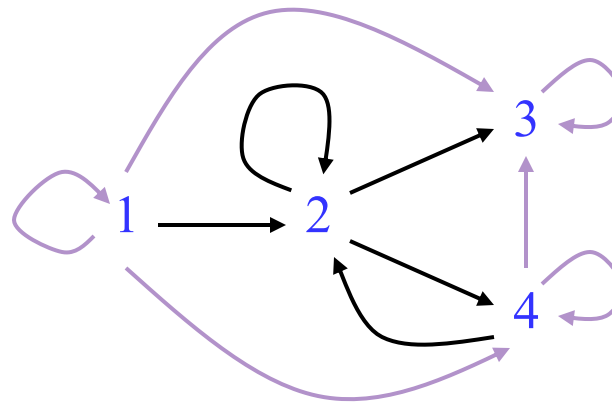
Compute  $H = (V, B)$  where  $B$  is the reflexive and transitive closure of  $E$

**Remark:**  $(s, t) \in B$  iff there exists a path from  $s$  to  $t$  in  $G$

graph  $G$ :



graph  $H$ :



# Matrix representation

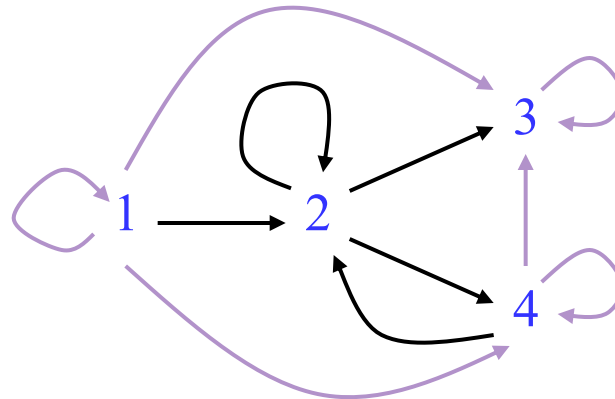
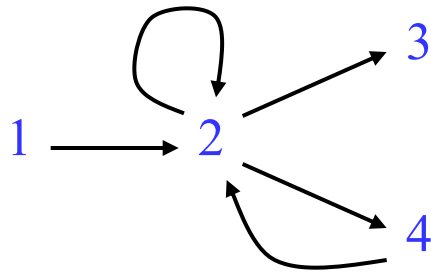
Matrix  $n \times n$  where  $n = |V|$

$A$  adjacency matrix of  $G$  (= matrix of paths of length 1)

$B$  adjacency matrix of  $H$  (= matrix of paths of  $H$ )

$$A = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

$$B = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 \end{pmatrix}$$



# Boolean matrix multiplication

- ▶  $A = (a_{ij}) \in \{0,1\}^{n \times k}$ ,  $B = (b_{ij}) \in \{0,1\}^{k \times m}$
- ▶  $AB = C$ ,  $C = \{0,1\}^{n \times m}$
- ▶  $c_{ij} = (a_{i1} \wedge b_{1j}) \vee \cdots \vee (a_{ik} \wedge b_{kj}) = \bigvee_{l=1}^k (a_{il} \wedge b_{lj})$

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}.$$

$$\mathbf{AB} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}.$$

# Closure by matrix multiplication

## Notation

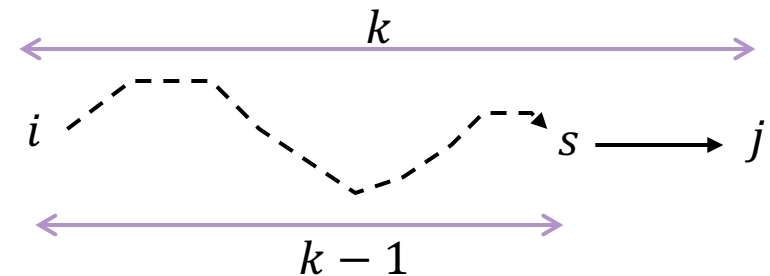
$A_k$  = matrix of paths of length  $k$  in  $G$

$A_0$  =  $I$  (identity matrix)

$A_1$  =  $A$  (matrix of paths of length 1)

## Lemma

For all  $k \geq 0$ ,  $A_k = A^k$   
(boolean matrix multiplication)



## Proof:

$A_k[i, j] = 1$  iff there exists  $s \in V$ :  $A_{k-1}[i, s] = 1$  and  $A[s, j] = 1$

that is,  $A_k[i, j] = \bigvee_s (A_{k-1}[i, s] \wedge A[s, j])$

that is,  $A_k = A_{k-1} \cdot A$  and  $A_0 = I$

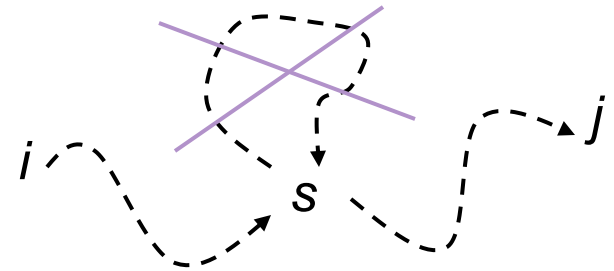
then  $A_k = A^k$

# Closure by matrix multiplication

there exists path from  $i$  to  $j$  in  $G \Leftrightarrow$

there exists a path from  $i$  to  $j$  without cycle (*simple path*)  $\Leftrightarrow$

there exists a path from  $i$  to  $j$  of length  $\leq n - 1$



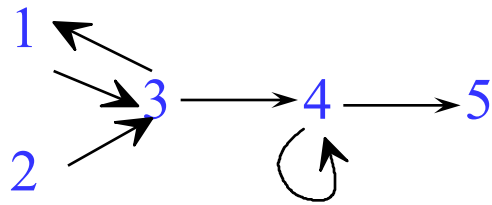
$$B[i, j] = 1 \quad \text{iff } \exists k, 0 \leq k \leq n - 1, A^k[i, j] = 1$$

therefore  $B = I + A + A^2 + \dots + A^{n-1}$  where  $+$  is  $\vee$

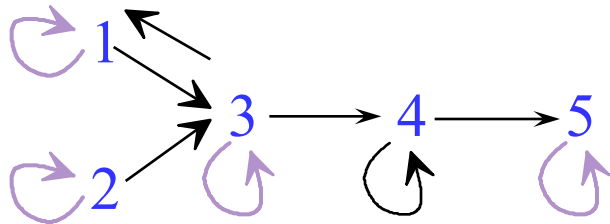
Computation of  $B$  using Horner's rule:

$$B_0 = I,$$

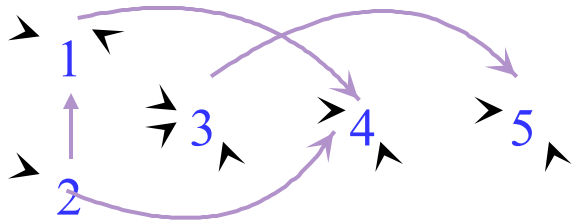
$$B_i = I + B_{i-1}A \quad \text{for } i = 1..n - 1. \quad \text{Then } B = B_{n-1}$$



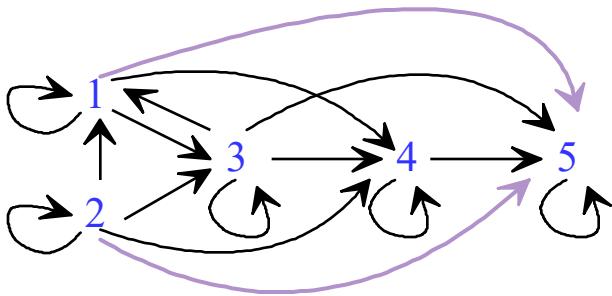
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



$$B_1 = I + A = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$B_2 = I + A + A^2 = I + B_1 \cdot A = \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$B_3 = I + A + A^2 + A^3 = I + B_2 \cdot A = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$B_4 = I + A + A^2 + A^3 + A^4 = I + B_3 \cdot A = B$$

3 matrix products overall

# Time complexity

$n - 1$  additions and  $n - 1$  products of boolean matrices  $n \times n$   
 $\Rightarrow O(n \cdot M(n))$

each product is done in  $O(n^3)$  operations  $\Rightarrow O(n^4)$

there exist matrix multiplication algorithms running in time  $o(n^3)$ :  
*Strassen* 1969:  $O(n^{2.8})$  (now improved to  $O(n^{2.37})$ )

*Four russians* (Арлазаров, Диниц, Кронрод, Фарадзев) 1970:  
 $O(n^3 / \log^2 n)$  (now improved to  $O(n^3 / \log^4 n)$ )



# Time complexity

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 $O(n^3 / \log^2 n)$  (now improved to  $O(n^3 / \log^4 n)$ )

**$O(n^4)$  is too much! can be done better with BFS/DFS:**

For each node  $i$ , run BFS with source node  $i$

$B[i, j] = 1$  iff  $j$  is reachable from  $i$

Running time  $O(n \cdot (n + m)) = O(n^3)$

# Speeding up

## Notation

$B_k$  = matrix of paths of length  $\leq k$  in  $G$   
( $B_k = A_0 + \dots + A_k$ )

$B_0 = I$  (identity matrix)

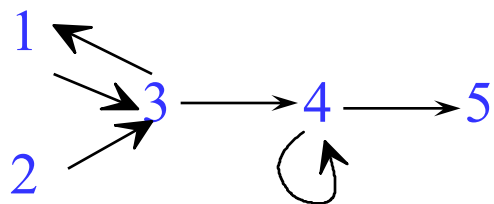
$B_1 =$  matrix of paths of length  $\leq 1$   $= I + A$

$B_{n-1} =$  matrix of simple paths  $= B$

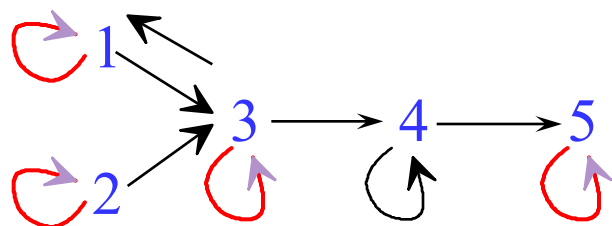
**Lemma:**  $B_k = B_{k-1} \cdot (I + A)$

$\Rightarrow$  For all  $k \geq 1$ ,  $B_k = (I + A)^k$  and then  $B_{2k} = B_k \cdot B_k$

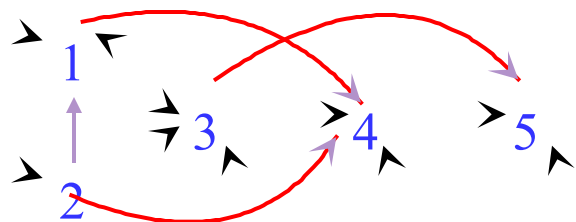
**Compute**  $B$  as an  $n - 1$  power in time  $O(\log(n) \cdot M(n)) = O(\log(n) \cdot n^3)$



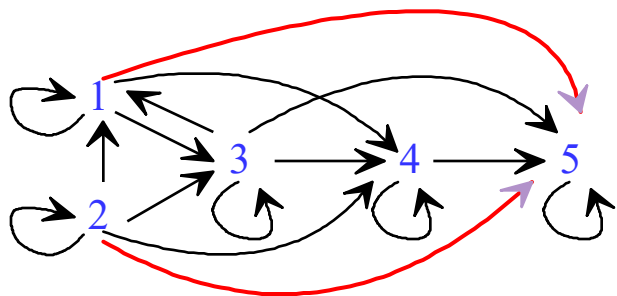
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



$$B_1 = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$B_2 = \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$B = B_4 = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

2 matrix products

# Warshall's (Roy-Warshall) algorithm (~1962)

$G = (V, E)$  with  $V = \{1, 2, \dots, n\}$

Paths in  $G$ :  $i \rightarrow s_1 \rightarrow s_2 \cdots \rightarrow s_l \rightarrow j$

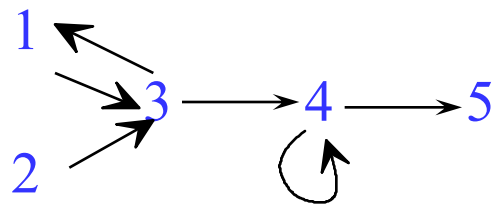
*Intermediate nodes*:  $s_1, s_2, \dots, s_l$

*Notation:*

$C_k$  = matrix of paths in  $G$  with  
*intermediate nodes  $\leq k$*

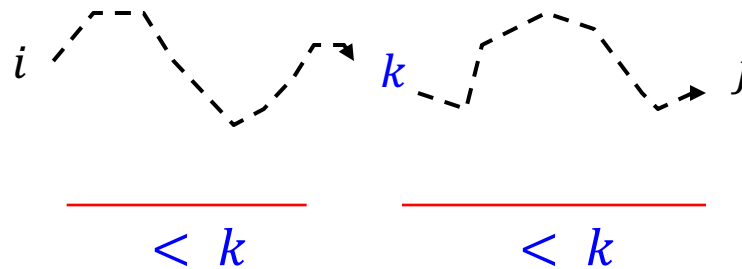
$C_0 = I + A$

$C_n$  = matrix of paths in  $G$  =  $B$



# Recurrence

Simple path



**Lemma** For all  $k \geq 1$ ,

$C_k[i, j] = 1$  iff  $C_{k-1}[i, j] = 1$  or ( $C_{k-1}[i, k] = 1$  and  $C_{k-1}[k, j] = 1$ )

**Computation**

of  $C_k$  from  $C_{k-1}$  in time  $O(n^2)$

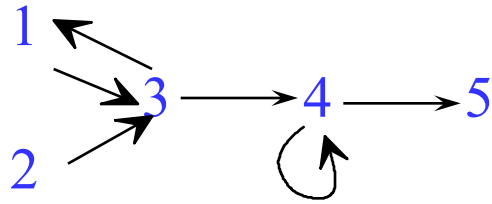
of  $B = C_n$  in time  $O(n^3)$

# Computing $C_k$ from $C_{k-1}$

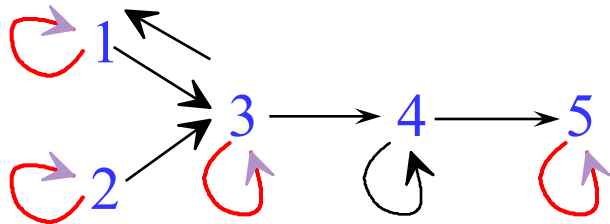
$$C_k = \begin{array}{c} j \\ | \\ i \text{ --- } * \end{array} \quad C_{k-1} = \begin{array}{c} k \quad j \\ | \quad | \\ \text{---} \text{---} * \\ | \quad | \\ i \text{ --- } * \end{array}$$

$$C_k[i, j] = C_{k-1}[i, j] \vee (C_{k-1}[i, k] \wedge C_{k-1}[k, j])$$

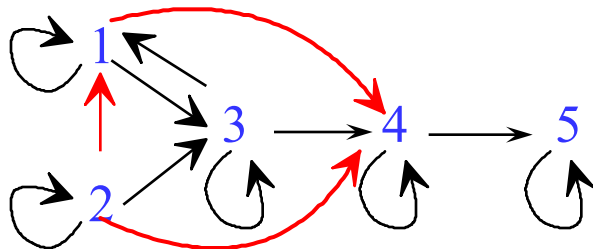
$$C_0 = I + A, \quad C_k[i, j] = 1 \text{ iff } C_{k-1}[i, j] = 1 \text{ or } (C_{k-1}[i, k] = 1 \text{ and } C_{k-1}[k, j] = 1)$$



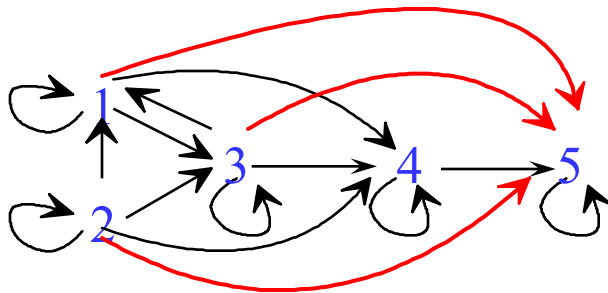
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



$$C_0 = C_1 = C_2 = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

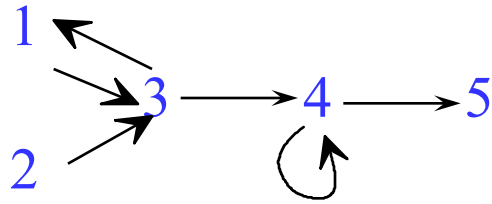


$$C_3 = \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

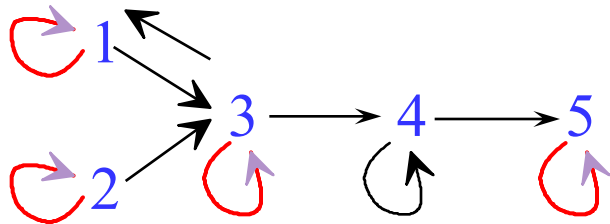


$$B = C_4 = C_5 = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

$$C_0 = I + A, \quad C_k[i, j] = 1 \text{ iff } C_{k-1}[i, j] = 1 \text{ or } (C_{k-1}[i, k] = 1 \text{ and } C_{k-1}[k, j] = 1)$$

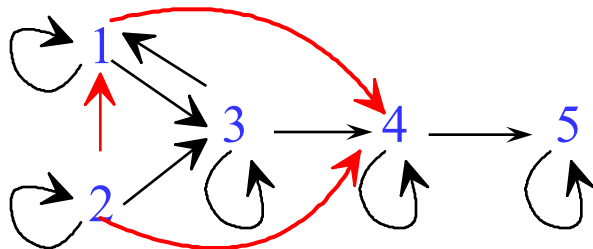


$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

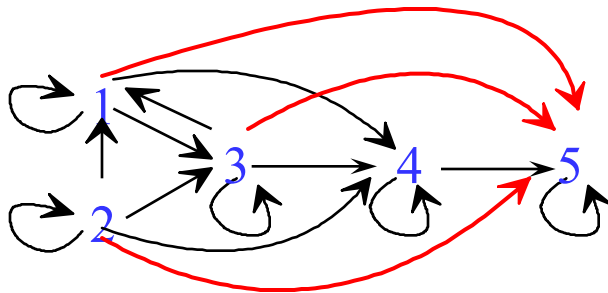


$$C_0 = C_1 = C_2 = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

computing  
 $C_1[4,3]$



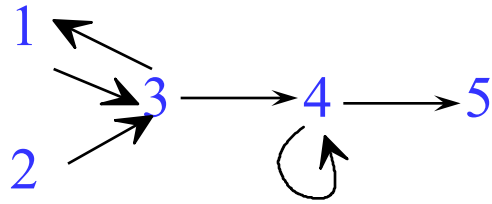
$$C_3 = \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



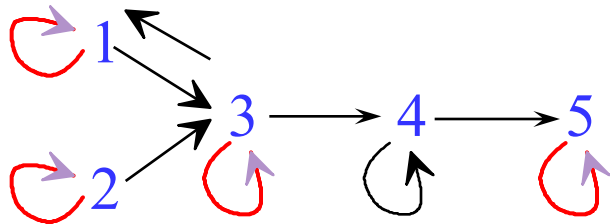
$$B = C_4 = C_5 = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



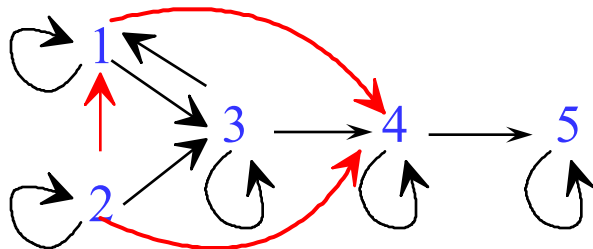
$$C_0 = I + A, \quad C_k[i, j] = 1 \text{ iff } C_{k-1}[i, j] = 1 \text{ or } (C_{k-1}[i, k] = 1 \text{ and } C_{k-1}[k, j] = 1)$$



$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

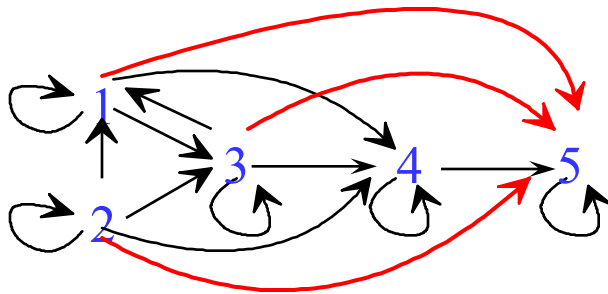


$$C_0 = C_1 = C_2 = \begin{pmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$C_3 = \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

computing  
 $C_3[2,4]$



$$B = C_4 = C_5 = \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

# Warshall's algorithm: code

WARSHALL( $G = (V, E)$ )

$n = |V|$

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

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

**if**  $i = j$  **or**  $A[i, j] = 1$  **then**

$C_0[i, j] = 1$

**else**

$C_0[i, j] = 0$

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

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

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

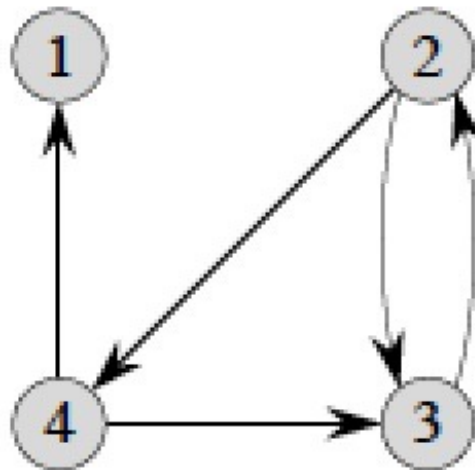
$C_k[i, j] = C_{k-1}[i, j] \vee (C_{k-1}[i, k] \wedge C_{k-1}[k, j])$

**return**  $C_n$

running time  $O(n^3)$

## Quiz 3.1.2

- Compute the transitive closure of the following graph using Warshall's algorithm



# What we have so far

*Three algorithms to compute the transitive closure:*

- matrix polynomial:  $O(n \cdot M(n)) = O(n^4)$
- matrix power:  $O(\log n \cdot M(n)) = O(\log n \cdot n^3)$
- Roy-Warshall algorithm :  $O(n^3)$

We now generalize these ideas to compute all-pairs shortest paths in a *weighted* graph

# What about weighted graphs?

$G = (V, E, w)$  weighted graph  $V = \{1, 2, \dots, n\}$ ,  $w: E \rightarrow \mathbb{R}$

We assume that there is no negative-cost cycle, but negative-cost edges may be present.

Weight matrix  $W$  defined by

$$W[i, j] = \begin{cases} 0 & \text{if } i = j \\ w(i, j) & \text{if } (i, j) \in E \\ \infty & \text{otherwise} \end{cases}$$

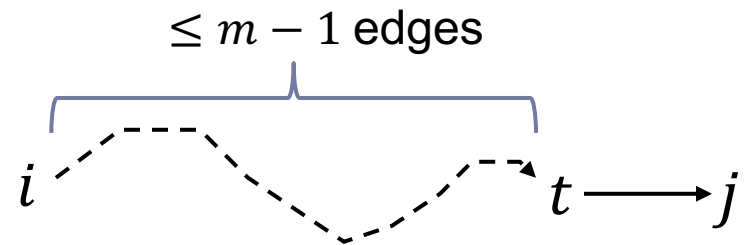
# First method: matrix product

Let  $d^{(m)}[i, j]$  be the minimum value of a path from  $i$  to  $j$  provided that this path contains **at most**  $m$  edges

We have to compute  $d[i, j] = d^{(n-1)}[i, j]$

*Idea* : proceed by induction on  $m$

$$d^{(0)}[i, j] = \begin{cases} 0 & \text{if } i = j \\ \infty & \text{otherwise} \end{cases}$$



For  $m \geq 1$ ,

$$d^{(m)}[i, j] = \min \left( d^{(m-1)}[i, j], \min_{1 \leq t \leq n} \{d^{(m-1)}[i, t] + W[t, j]\} \right) = \min_{1 \leq t \leq n} \{d^{(m-1)}[i, t] + W[t, j]\}$$

In terms of matrices, we have  $D^{(m)} = D^{(m-1)} \cdot W$ , where

$\min$  plays the role of **addition** and

$+$  plays the role of **multiplication**

Computing  $D = W^{n-1}$  by repeated squaring leads to the time complexity  $O(n^3 \cdot \log n)$

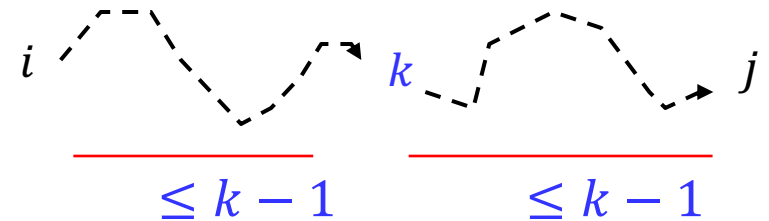
# Algorithm based on intermediate nodes: Floyd(-Warshall) algorithm

## Notation

$D_k = (D_k[i, j] \mid 1 \leq i, j \leq n)$  with  
 $D_k[i, j] = \min\{ w(c) \mid c \text{ path from } i \text{ to } j \text{ with}$   
 $\text{all intermediate nodes } \leq k \}$

$D_0 = W$

$D_n = \text{distance matrix of } G = D$



**Lemma** For all  $k \geq 1$ ,

$$D_k[i, j] = \min\{D_{k-1}[i, j], D_{k-1}[i, k] + D_{k-1}[k, j]\}$$

## Computation

of  $D_k$  from  $D_{k-1}$  in time  $O(n^2)$

of  $D = D_n$  in time  $O(n^3)$



FLOYD(G,w)

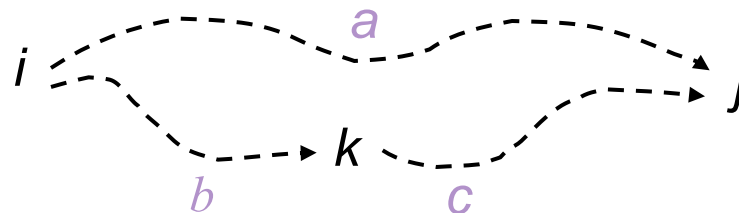
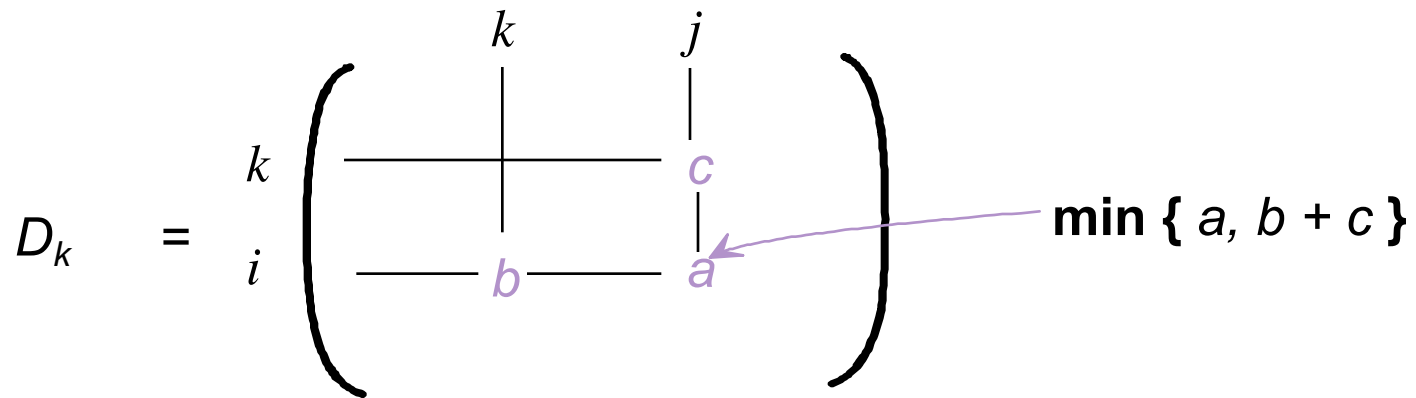
$D_0 = W$

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

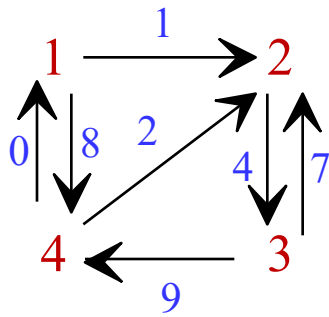
**for**  $i = 1$  **to**  $n$  **do**

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

$D_k[i, j] = \min \{ D_{k-1}[i, j], D_{k-1}[i, k] + D_{k-1}[k, j] \}$



$$D_k[i, j] = \min \{ D_{k-1}[i, j], D_{k-1}[i, k] + D_{k-1}[k, j] \}$$



$$D_0 = W = \begin{pmatrix} 0 & 1 & \infty & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & 2 & \infty & 0 \end{pmatrix}$$

$$D_1 = \begin{pmatrix} 0 & 1 & \infty & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & 1 & \infty & 0 \end{pmatrix}$$

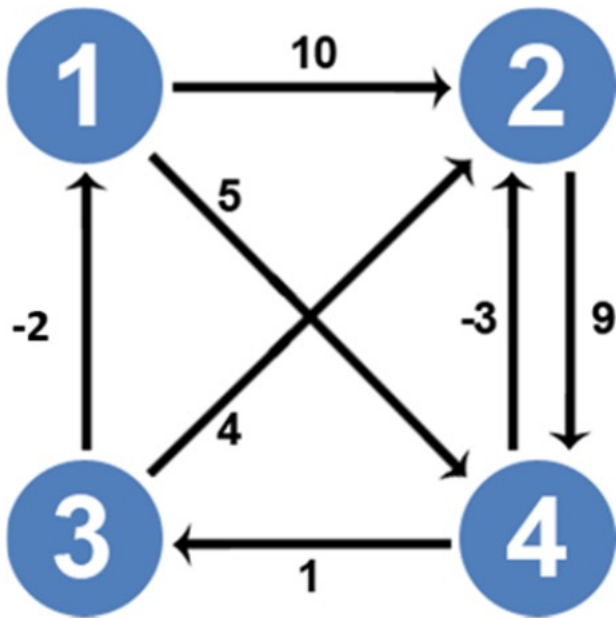
$$D_2 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$D_3 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ \infty & 0 & 4 & 13 \\ \infty & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$D_4 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ 13 & 0 & 4 & 13 \\ 9 & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

## Quiz 3.1.3

Run Floyd's algorithm to compute all-pairs shortest distances. Output the sum of the shortest distances between all pairs of vertices.



# Representing shortest paths

Explicitly storing shortest paths from  $i$  to  $j$ ,  $1 \leq i, j \leq n$   
 $n^2$  paths of maximum length  $n - 1$ : space  $O(n^3)$

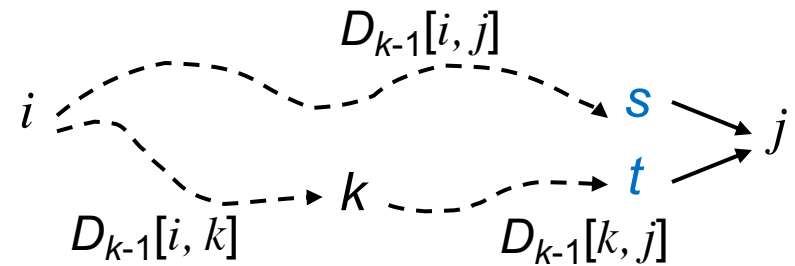
**Predecessor matrix:** space  $\Theta(n^2)$

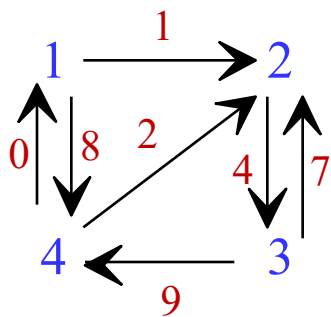
$\pi_k = (\pi_k[i, j] \mid 1 \leq i, j \leq n)$  where  
 $\pi_k[i, j]$  = predecessor of  $j$  on some shortest path from  $i$  to  $j$  with  
all intermediate nodes  $\leq k$

**Recurrence**

$$\pi_0[i, j] = \begin{cases} i, & \text{if } i \neq j \text{ and } (i, j) \in E \\ \text{nil} & \text{otherwise} \end{cases}$$

$$\pi_k[i, j] = \begin{cases} \pi_{k-1}[i, j], & \text{if } D_{k-1}[i, j] \leq D_{k-1}[i, k] + D_{k-1}[k, j] \\ \pi_{k-1}[k, j] & \text{otherwise} \end{cases}$$





$$D_0 = W = \begin{pmatrix} 0 & 1 & \infty & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & 2 & \infty & 0 \end{pmatrix}$$

$$\pi_0 = \begin{pmatrix} - & 1 & - & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & 4 & - & - \end{pmatrix}$$

$$D_1 = \begin{pmatrix} 0 & 1 & \infty & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & \mathbf{1} & \infty & 0 \end{pmatrix}$$

$$\pi_1 = \begin{pmatrix} - & 1 & - & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & \mathbf{1} & - & - \end{pmatrix}$$

$$D_2 = \begin{pmatrix} 0 & 1 & \mathbf{5} & 8 \\ \infty & 0 & 4 & \infty \\ \infty & 7 & 0 & 9 \\ 0 & 1 & \mathbf{5} & 0 \end{pmatrix}$$

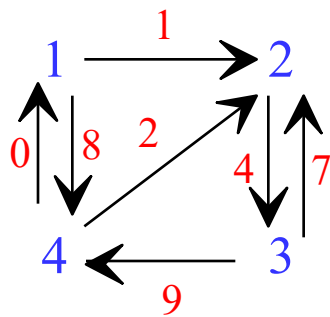
$$\pi_2 = \begin{pmatrix} - & 1 & \mathbf{2} & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & 1 & \mathbf{2} & - \end{pmatrix}$$

$$D_3 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ \infty & 0 & 4 & \mathbf{13} \\ \infty & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$\pi_3 = \begin{pmatrix} - & 1 & 2 & 1 \\ - & - & 2 & \mathbf{3} \\ - & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$

$$D_4 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ \mathbf{13} & 0 & 4 & 13 \\ \mathbf{9} & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$\pi_4 = \begin{pmatrix} - & 1 & 2 & 1 \\ \mathbf{4} & - & 2 & 3 \\ \mathbf{4} & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$



$$D_4 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ 13 & 0 & 4 & 13 \\ 9 & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$\pi_4 = \begin{pmatrix} - & 1 & 2 & 1 \\ 4 & - & 2 & 3 \\ 4 & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$

### Example of a path

distance from 2 to 1 =  $D_4[2,1] = 13$

$\pi_4[2,1] = 4$  ;  $\pi_4[2,4] = 3$  ;  $\pi_4[2,3] = 2$  ;



# Remarks

- ▶ For sparse graphs represented by adjacency lists there exists **Johnson's algorithm** that works in time  $O(n^2 \cdot \log n + nm)$ .
- ▶ Warshall's and Floyd-Warshall algorithms are examples of the ***dynamic programming*** technique that we will study later in more details

# Shortest paths: summary

## ***Unweighted single-source shortest paths***

*Breadth-first search*  $O(|V| + |E|)$

## ***Weighted single-source shortest paths***

*depending on assumptions:*

*Dijkstra's algorithm*  $O(|V|^2)$   
or  $O(|V| + |E| \cdot \log |V|)$   
*Bellman-Ford algorithm*  $O(|E| \cdot |V|)$

## ***All-pairs shortest paths***

*Floyd-Warshall algorithm*  $O(|V|^3)$