

**Transitive closure of graphs  
and  
all-pairs shortest paths**

## Transitive closure (accessibility)

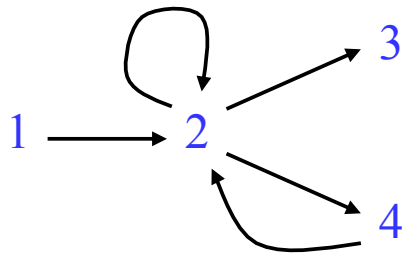
### Problem:

$G = (S, A)$  directed graph

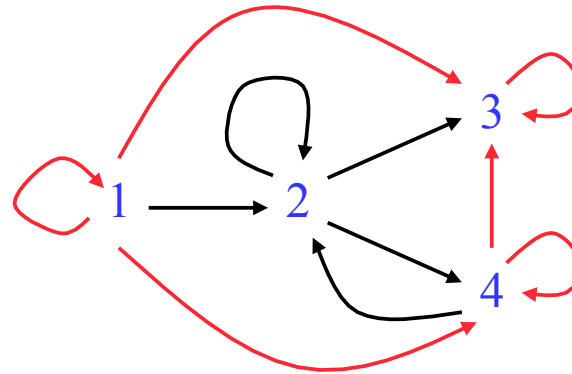
Compute  $H = (S, B)$  where  $B$  is the reflexive and transitive closure of  $A$ .

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

graph  $G$ :



graph  $H$ :



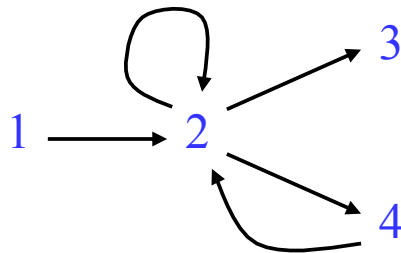
## Matrix representation

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

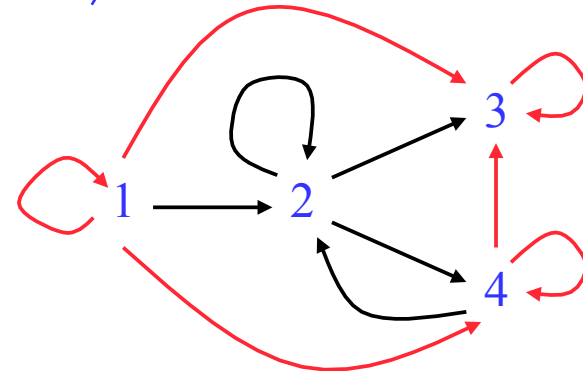
$A$  adjacency matrix of  $G$   
= matrix of paths of length 1

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

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



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



## Closure by matrix multiplication

### Notation

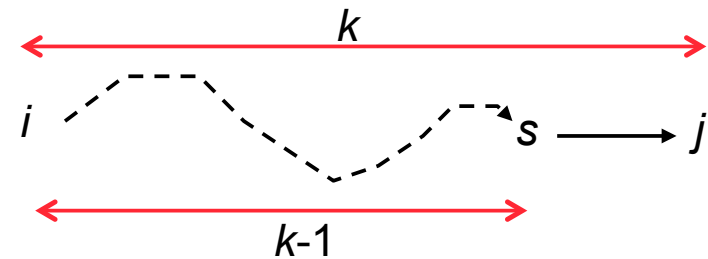
$A_k$  = matrix of paths of length  $k$  in  $G$

$A_0$  =  $I$  (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  $\exists s \in S$   $A_{k-1}[i, s] = 1$  and  $A[s, j] = 1$

let  $A_k[i, j] = \sum_s A_{k-1}[i, s] \cdot A[s, j]$  where  $\sum$  boolean sum (OR).

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

then  $A_k = A^k$

## Closure by matrix multiplication (cont)

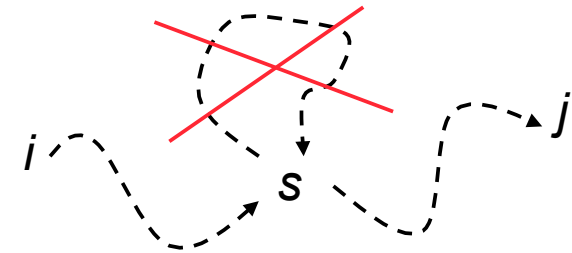
### Simple path:

path with all vertices distinct (=not containing a cycle)

### Lemma

$\exists$  path from  $i$  to  $j$  in  $G$  iff

$\exists$  simple path from  $i$  to  $j$  in  $G$



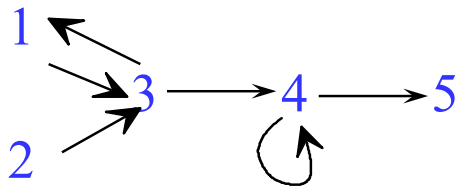
$B[i,j] = 1$       iff  $\exists$  path from  $i$  to  $j$  in  $G$   
                       iff  $\exists$  simple path from  $i$  to  $j$  in  $G$   
                       iff  $\exists k, 0 \leq k \leq |S| - 1, A_k[i,j] = 1$   
                       iff  $\exists k, 0 \leq k \leq |S| - 1 A^k[i,j] = 1$

therefore       $B = I + A + A^2 + \dots + A^{|S|-1}$

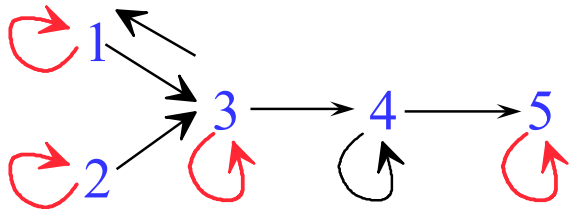
### Computation of $B$ using Horner's rule:

$$A_0 = I$$

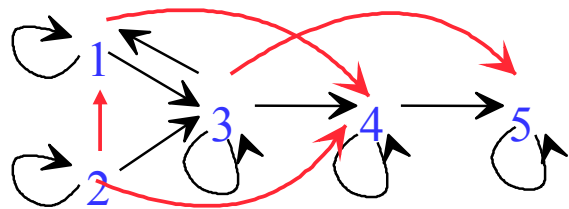
$$A_i = I + A_{i-1} A \quad \text{for } i=1..|S|-1$$



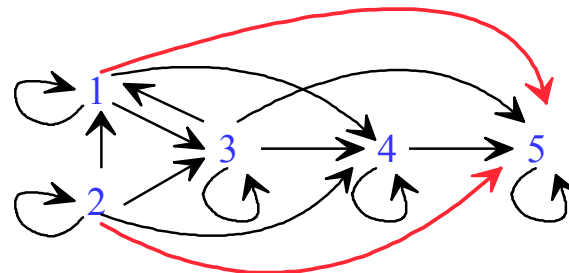
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



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



$$\begin{aligned} I + A + A^2 \\ = I + (I + A) \cdot A = \end{aligned} \begin{pmatrix} 1 & 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



$$\begin{aligned} I + A + A^2 + A^3 \\ = I + (I + A + A^2) \cdot A = \end{aligned} \begin{pmatrix} 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

3 matrix products

$$\begin{aligned} &= I + A + A^2 + A^3 + A^4 \\ &= I + (I + A + A^2 + A^3) \cdot A = B \end{aligned}$$

## Time complexity

$$n=|S|$$

$n-1$  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:  $O(n^{2.37})$  (now improved to  $O(n^{2.37})$ )

*Four russians* (Арпазаров, Диниц, Кронрод, Фарадзев) 1970:  $O(n^3/\log^2(n))$

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

Define  $v(i,j)=1$  for each  $(i,j) \in A$

For each node  $i$ , run Dijkstra's algorithm with source node  $i$

$B[i,j]=1$  ssi  $\delta(i,j) < \infty$

Running time  $O(n \cdot n^2) = O(n^3)$



## Speeding up

### Notation

$B_k$  = matrix of paths of length  $\leq k$  in  $G$

$B_0$  =  $I$  (identity matrix)

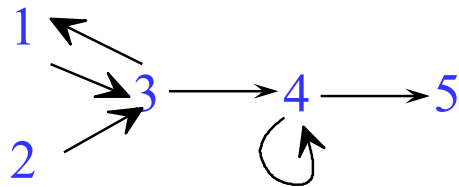
$B_1$  = matrix of paths of length  $\leq 1$  =  $I + A$

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

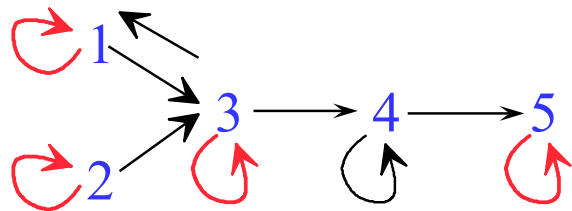
**Lemma:**  $B_k = B_{k-1} \cdot (I + A)$

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

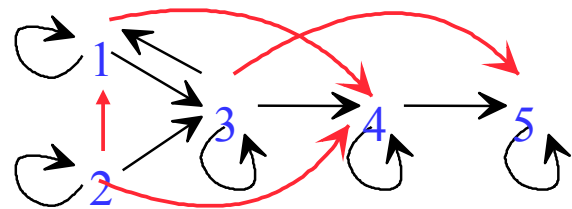
**Compute**  $B$  as an  $n-1$  power in time  $O(\log(n) \cdot M(n)) = O(\log(n) \cdot n^3)$



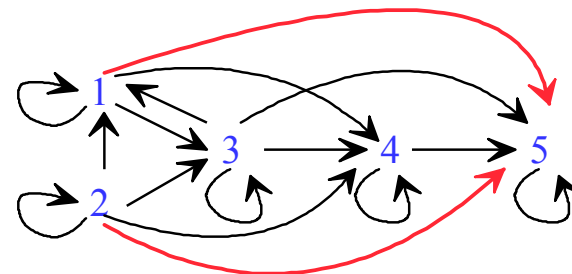
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



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



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



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

2 matrix products

## Warshall's algorithm (Roy-Warshall)

$G = (S, A)$  with  $S = \{1, 2, \dots, n\}$

Paths in  $G : i \rightarrow s_1 \rightarrow s_2 \dots s_m \rightarrow j$

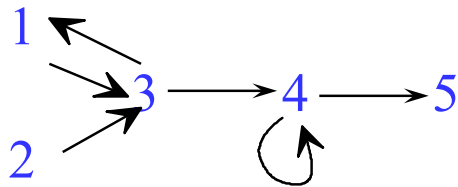
**Intermediate nodes** :  $s_1, s_2, \dots, s_m$

**Notation:**

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

$C_0 = I + A$

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

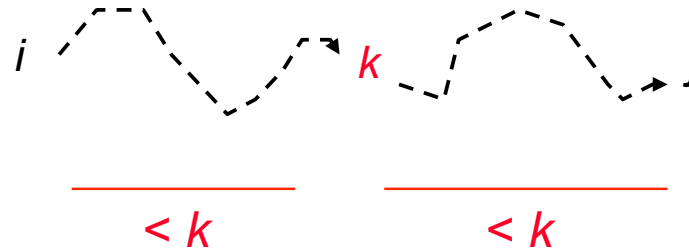


Paths from 2 to 4:  $(2,3), (3,1), (1,3), (3,4), (4,4)$

intermediate nodes: 1, 3, 4

## Recurrence

Simple path



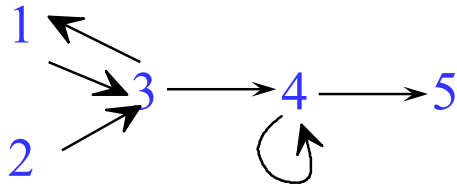
**Lemma** For all  $k \geq 1$ ,

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

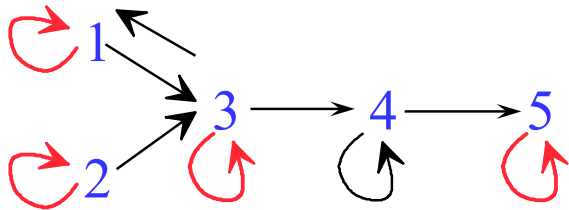
**Computation**

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

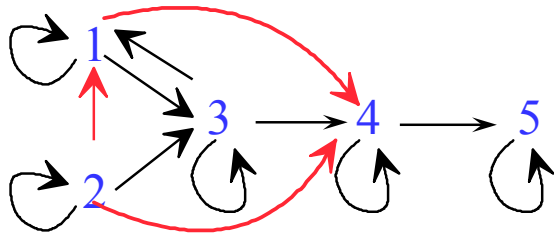
of  $B = C_n$  in time  $O(n^3)$



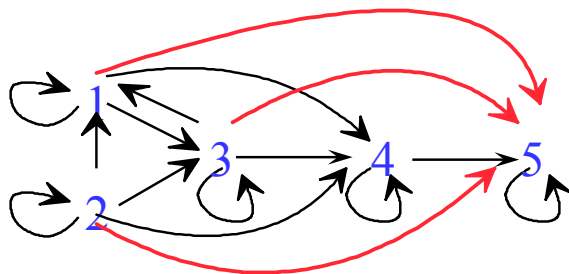
$$A = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$



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



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



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

~1 matrix product

```

function closure (graph  $G = (S, A)$ ) : matrix ;
begin
     $n \leftarrow |S|$  ;
    for  $i \leftarrow 1$  to  $n$  do
        for  $j \leftarrow 1$  to  $n$  do
            if  $i = j$  or  $A[i,j] = 1$  then
                 $B[i,j] \leftarrow 1$  ;
            else
                 $B[i,j] \leftarrow 0$  ;
        for  $k \leftarrow 1$  to  $n$  do
            for  $i \leftarrow 1$  to  $n$  do
                for  $j \leftarrow 1$  to  $n$  do
                     $B[i,j] \leftarrow B[i,j] + B[i,k] \cdot B[k,j]$  ;
    return  $B$  ;
end

```

+ is the boolean sum ; running time  $O(n^3)$

## 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

## Distances

$G = (S, A, v)$  weighted graph  $S = \{1, 2, \dots, n\}$   $v : A \rightarrow \mathbf{R}$ .

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

Weight matrix:  $W = (w[i, j])$  with

$$w[i, j] = \begin{cases} 0 & \text{if } i = j \\ v((i, j)) & \text{if } (i, j) \in A \\ \infty & \text{else} \end{cases}$$

**Weight of a sequence**  $c = ((s_0, s_1), (s_1, s_2), \dots, (s_{k-1}, s_k))$  where  $s_i \in S$

$$w(c) = \sum w[s_{i-1}, s_i]$$

**Distance** from  $s$  to  $t$

$$d(s, t) = \min\{w(c) \mid c \text{ sequence from } s \text{ to } t\}$$

**Shortest path** from  $s$  to  $t$  :

path  $c$ , if it exists, such that  $w(c) = d(s, t)$



## First method: matrix product

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

$$d(i,j) = d^{(n)}(i,j)$$

*Idea : proceed by induction on  $m$*

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

For  $m \geq 1$ ,

$$d^{(m)}(i,j) = \min (d^{(m-1)}(i,j), \min\{d^{(m-1)}(i,t)+w_{tj} \mid 1 \leq t \leq n\}) = \\ \min\{d^{(m-1)}(i,t)+w_{tj} \mid 1 \leq t \leq n\}$$

In terms of matrices, we have  $D^{(m)} = D^{(m-1)} \cdot W$  where  
 $\min$  plays the role of  $+$  and  
 $+$  plays the role of  $\cdot$ .

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

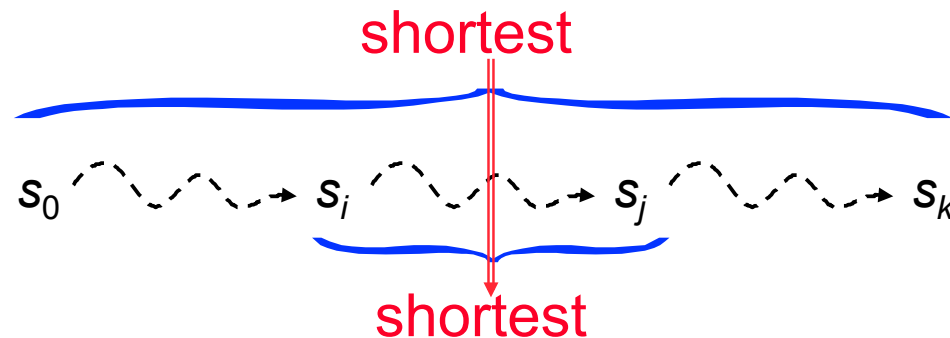
## Algorithm based on intermediate nodes

### Basic Lemma (reminder)

$((s_0, s_1), \dots, (s_i, s_{i+1}), \dots, (s_{j-1}, s_j), \dots, (s_{k-1}, s_k))$

a shortest path from  $s_0$  to  $s_k$  in  $G$

$\Rightarrow ((s_i, s_{i+1}), \dots, (s_{j-1}, s_j))$  a shortest path from  $s_i$  to  $s_j$  in  $G$



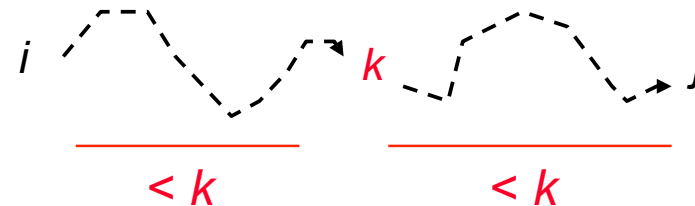
## Floyd(-Warshall) algorithm

### Notation

$D_k = (D_k[i, j] \mid 1 \leq i, j \leq n)$  with  
 $D_k[i, j] = \min\{ w(c) \mid c \text{ path from } i \text{ to } j \text{ with}$   
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  $O(n^2)$

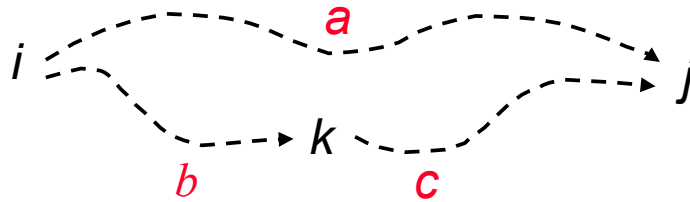
of  $D = D_n$  in time  $O(n^3)$

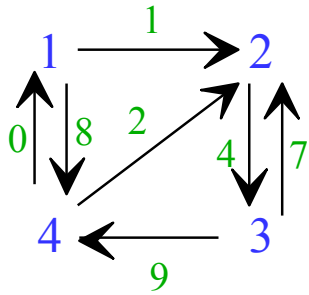
```

for  $k \leftarrow 1$  to  $n$  do
  for  $i \leftarrow 1$  to  $n$  do
    for  $j \leftarrow 1$  to  $n$  do
       $D[i, j] \leftarrow \min \{ D[i, j], D[i, k] + D[k, j] \};$ 

```

$$D_k = \begin{matrix} & k & j \\ \begin{matrix} k \\ i \end{matrix} & \left( \begin{array}{c|c} \hline & c \\ \hline b & a \end{array} \right) \end{matrix} \quad \min \{ a, b + c \}$$





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

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

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

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

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

## Representing shortest paths

Explicitly storing shortest paths from  $i$  to  $j$ ,  $1 \leq i, j \leq n$

$n^2$  paths of maximal length  $n-1$ : space  $O(n^3)$

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

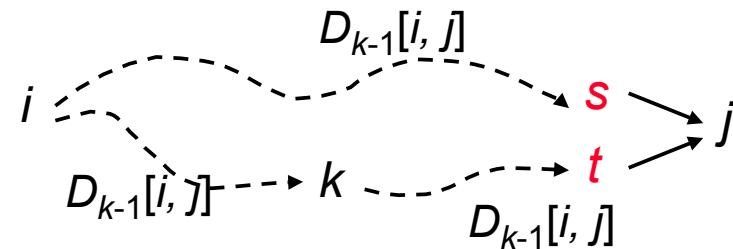
$\pi_k = (\pi_k[i, j] \mid 1 \leq i, j \leq n)$  where

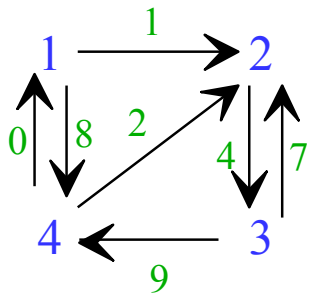
$\pi_k[i, j]$  = predecessor of  $j$  on some shortest path from  $i$  to  $j$  with  
all intermediate nodes  $\leq k$

**Recurrence**

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

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





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

$$P_0 = \begin{pmatrix} - & 1 & - & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & 4 & - & - \end{pmatrix}$$

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

$$P_1 = \begin{pmatrix} - & 1 & - & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & 1 & - & - \end{pmatrix}$$

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

$$P_2 = \begin{pmatrix} - & 1 & 2 & 1 \\ - & - & 2 & - \\ - & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$

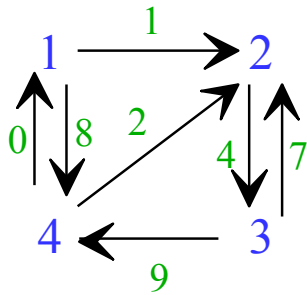
$$D_3 = \begin{pmatrix} 0 & 1 & 5 & 8 \\ \infty & 0 & 4 & 13 \\ \infty & 7 & 0 & 9 \\ 0 & 1 & 5 & 0 \end{pmatrix}$$

$$P_3 = \begin{pmatrix} - & 1 & 2 & 1 \\ - & - & 2 & 3 \\ - & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$

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

$$P_4 = \begin{pmatrix} - & 1 & 2 & 1 \\ 4 & - & 2 & 3 \\ 4 & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$





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

$$P_4 = \begin{pmatrix} - & 1 & 2 & 1 \\ \mathbf{4} & - & \mathbf{2} & \mathbf{3} \\ 4 & 3 & - & 3 \\ 4 & 1 & 2 & - \end{pmatrix}$$

### Example of a path

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

$P_4[2,1] = 4$  ;       $P_4[2,4] = 3$  ;     $P_4[2,3] = 2$  ;

