

Single-source shortest paths in DAGs.  
All Pairs Shortest Paths.  
The Floyd-Warshall algorithm

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# Single-source shortest paths in DAGs

A DAG (**directed acyclic graph**) is a digraph without cycles.

**ASSUMPTION:**  $G = (V, E)$  is a DAG with weight function  $w : E \rightarrow \mathbb{R}$ . We write  $w(u, v)$  for the weight of arc  $u \rightarrow v$ . We know that

- 1  $G$  has no cycles  $\Rightarrow G$  has no negative cycles  $\Rightarrow$  there is a minimal path between every pair of nodes.
- 2 The nodes of  $G$  can be sorted in topological order  $v_1, \dots, v_n$  such that  $(v_i \rightarrow v_j) \in E$  implies  $i < j$ .

# Single-source shortest paths in DAGs

DAGSHORTESTPATHS( $G, w, s$ )

1 topologically sort the nodes of  $G$

2 INITIALIZESINGLESOURCE( $G, s$ )

3 **for** each node  $u$  taken in topologically sorted order

4     **for** each node  $v \in \text{Adj}[u]$

5         RELAX( $u, v, w$ )

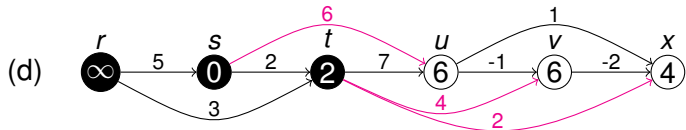
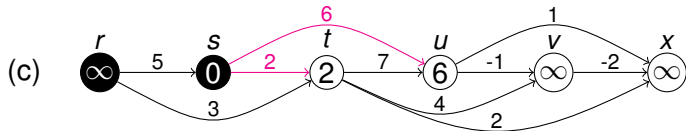
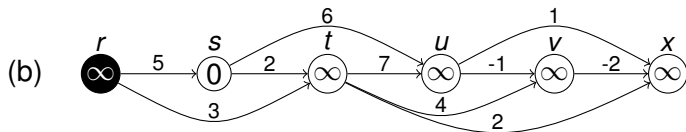
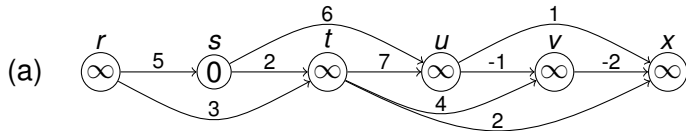
## Complexity analysis

- Line 1 (topological sort) can be done in  $\Theta(|V| + |E|)$  time.
- The outer **for** loop is done  $|V|$  times.
- For each node, the arcs that leave it are examined exactly once.
- Line 5 takes  $O(1)$  time.

$\Rightarrow \Theta(|V| + |E|)$  time.

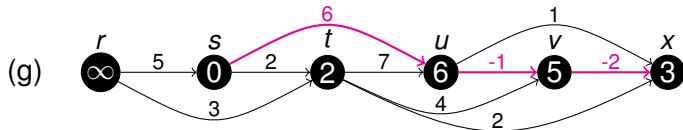
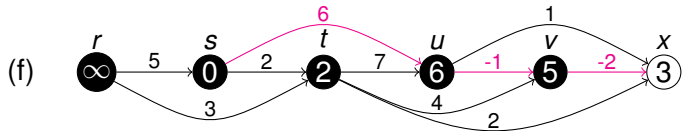
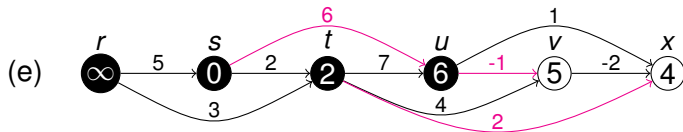
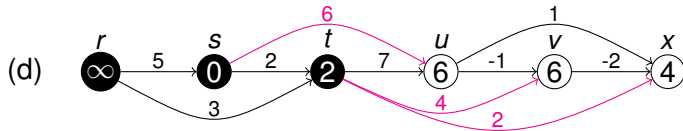
# Single-source shortest paths in DAGs

Running example



# Single-source shortest paths in DAGs

Running example



# Single-source shortest paths in DAGs

## Running example

- The **magenta** arrows are from the predecessor of a node to the node.
- The values inside circles represent the shortest path estimates after each execution of the **for** loop of the algorithm (lines 3-5).

# Single-source shortest paths in DAGs

Properties of the algorithm DAGSHORTESTPATHS

## Theorem

If a weighted directed graph  $G = (V, E)$  has source node  $s$  and no cycles, then at the termination of the DAGSHORTESTPATHS algorithm, we have

- 1  $d[v] = \delta(s, v)$  for all nodes  $v \in V$ , and
- 2 the predecessor subgraph  $G_\pi$  is a shortest paths tree.

PROOF. Cormen *et al*, *Introduction to Algorithms*, Section 25.4.

# All pairs shortest path: The problem

ASSUMPTION: Given a directed, weighted graph  $G = (V, E)$  with weight function  $w : E \rightarrow \mathbb{R}$ .

- Find shortest paths between all pairs of vertices in a graph.
- Example:
  - Make a table of distances between all pairs of cities for a road atlas.

What did we learn so far?

- Single-source shortest-path algorithms:
  - Dijkstra:  $O(|V|^2)$
  - Bellman-Ford:  $O(|V| |E|)$ .



# All pairs shortest path algorithms

Straightforward solution

- If  $w : E \rightarrow \mathbb{R}_+$ , we can run Dijkstra's single-path algorithm for every node  $v \in |V|$   
 $\Rightarrow$  overall complexity:  $O(|V|) \cdot O(|V|^2) = O(|V|^3)$ .
- $w : E \rightarrow \mathbb{R}$ , we can run Bellman-Ford single-path algorithm for every node  $v \in |V|$   
 $\Rightarrow$  overall complexity:  $O(|V|) \cdot O(|V| |E|) = O(|V|^4)$ .

**Q:** Can we do better?

# All pairs shortest path algorithms

- We assume given the adjacency matrix of  $G$  with  $n$  nodes
- For all  $i, j \in V$ , we know

$$w_{i,j} = \begin{cases} 0 & \text{if } i = j \\ \text{weight of directed edge } (i, j) & \text{if } i \neq j \text{ and } (i, j) \in E \\ \infty & \text{if } i \neq j \text{ and } (i, j) \notin E. \end{cases}$$

- We want to compute
  - The  $n \times n$  matrix  $D = (d_{i,j})$  where every  $d_{i,j}$  is the length of a shortest path from node  $i$  to node  $j$
  - The  $n \times n$  predecessor matrix  $\Pi = (\pi_{i,j})$  where

$$\pi_{i,j} = \begin{cases} \text{nil} & \text{if } i = j \text{ or there is no path from } i \text{ to } j \\ k & \text{if there exists a shortest path } i \rightsquigarrow k \rightarrow j \text{ from } i \text{ to } j. \end{cases}$$

# All pairs shortest path algorithms

- Suppose we know the predecessor matrix  $\Pi = (\pi_{i,j})$ .
- For every  $i \in V$ , define the **predecessor graph** of  $G$  for  $i$  as follows:  $G_{\pi,i} = (V_{\pi,i}, E_{\pi,i})$ , where
  - $V_{\pi,i} := \{j \in V \mid \pi_{i,j} \neq \text{nil}\} \cup \{i\}$
  - $E_{\pi,i} := \{(\pi_{i,j}, j) \mid j \in V_{\pi,i} \text{ and } \pi_{i,j} \neq \text{nil}\}$
- For all  $i \in V$ ,  $G_{\pi,i}$  is a shortest-path tree with root  $i$ .

# The structure of a shortest path

- All subpaths of a shortest path are shortest paths.
- Suppose  $i \overset{p}{\rightsquigarrow} j$  is a shortest path.
  - if  $i = j$  then  $p$  has weight 0 and no edges.
  - If  $i \neq j$  then  $p$  contains a finite number of edges, say  $\leq m$  edges, and we can write  $p = i \overset{p'}{\rightsquigarrow} k \rightarrow j$ , where  $p'$  is a path with  $\leq m - 1$  edges. Note that
    - $p'$  is a shortest path from  $i$  to  $k \Rightarrow \delta(i, j) = \delta(i, k) + w_{k, j}$

# All pairs shortest path algorithms

## A recursive solution

- Let  $d_{i,j}^{(m)} :=$  minimum weight of any path with length  $\leq m$  from  $i$  to  $j$ ,
- When  $m = 0$ , there is a shortest path from  $i$  to  $j$  with no edges if and only if  $i = j$ . Thus

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

- Recursive formula** for  $m > 0$ :

$$\begin{aligned} d_{i,j}^{(m)} &= \min_{1 \leq k \leq n} \left( d_{i,j}^{(m-1)}, \min_{1 \leq k \leq n} \{ d_{i,k}^{(m-1)} + w_{k,j} \} \right) \\ &= \min_{1 \leq k \leq n} \{ d_{i,k}^{(m-1)} + w_{k,j} \} \end{aligned}$$

NOTE. The last equality holds because  $w_{j,j} = 0$  for all  $j$ .

# Shortest path weights

**Q:** How to compute the shortest path weights  $\delta_{i,j}$  for all nodes  $i, j \in V$ ?

**A:** If the graph contains no negative-weight cycles then

- All shortest paths are simple and they contain at most  $n - 1$  edges
- A path from  $i$  to  $j$  with more than  $n - 1$  edges can not have less weight than a shortest path from  $i$  to  $j$
- $\Rightarrow \delta_{i,j} = \delta_{i,j}^{(n-1)} = \delta_{i,j}^{(n)} = \dots$

# Shortest path weights

The bottom-up approach

Given input matrix  $W = (w_{i,j})$

Compute the series of matrices  $D^{(1)}, \dots, D^{(n-1)}$  where for  $1 \leq m < n$ , we have  $D^{(m)} = (\delta_{i,j}^{(m)})$

- Since  $d_{i,j}^{(1)} = w_{i,j}$  for all  $i, j \in V$ , we have  $D^{(1)} = W$ .

- EXTEND-SHORTEST-PATHS( $D, W$ ) is supposed to take the following arguments:
  - $D = D^{(m-1)}$
  - The weight matrix  $W$  of the graphand will return  $D^{(m)}$

# Shortest path weights

The bottom-up approach

EXTEND-SHORTEST-PATHS( $D, W$ )

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

- Complexity =  $\Theta(n^3)$  – due to the three nested **for** loops.



# Shortest path weights

The relation between the bottom-up approach and matrix multiplication

- If  $A, B$  are  $n \times n$  matrices and  $C = (c_{i,j}) = A \cdot B$  then

$$c_{i,j} = \sum_{k=1}^n a_{i,k} \cdot b_{k,j}$$

- The matrix multiplication formula is equivalent with the recursive formula for  $d^{(m)}$ , if we do the following changes:

$$\begin{array}{lll} d^{(m-1)} & \leftrightarrow & a \\ w & \leftrightarrow & b \\ d^{(m)} & \leftrightarrow & c \\ \min & \leftrightarrow & + \\ + & \leftrightarrow & \cdot \end{array}$$

# Shortest path weights

The relation between the bottom-up approach and matrix multiplication

- Based on the previous observation, we get the following matrix multiplication algorithm by changing EXTEND-SHORTEST-PATHS:

**MATRIX-MULTIPLY**( $A, B$ )

```
1   $n \leftarrow \text{rows}[A]$ 
2  let  $C$  be an  $n \times n$  matrix
3  for  $i \leftarrow 1$  to  $n$ 
4      do for  $j \leftarrow 1$  to  $n$ 
5          do  $c_{ij} \leftarrow 0$ 
6              for  $k \leftarrow 1$  to  $n$ 
7                  do  $c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$ 
8  return  $C$ 
```

- Time complexity of MATRIX-MULTIPLY:  $\Theta(n^3)$ .

# Shortest path weights

The relation between the bottom-up approach and matrix multiplication

If we write  $A \cdot B$  instead of  $\text{EXTEND-SHORTEST-PATHS}(A, B)$ , we compute

Computation	Time complexity
$D^{(1)} = D^{(0)} \cdot W = W$	$\Theta(n^3)$
$D^{(2)} = D^{(1)} \cdot W = W^2$	$\Theta(n^3)$
$\vdots$	
$D^{(n-1)} = D^{(n-2)} \cdot W = W^{n-1}$	$\Theta(n^3)$

- In the end,  $W^{n-1}$  contains the shortest-path weights of all pairs of nodes.
- Total time complexity:  $(n - 1) \cdot \Theta(n^3) = \Theta(n^4)$  time.

# Shortest path weights

The bottom-up approach

## SLOW-ALL-PAIRS-SHORTEST-PATHS( $W$ )

```
1  $n \leftarrow \text{rows}[W]$ 
2  $D^{(1)} \leftarrow W$ 
3 for  $m \leftarrow 2$  to  $n - 1$ 
4     do  $D^{(m)} \leftarrow \text{EXTEND-SHORTEST-PATHS}(D^{(m-1)}, W)$ 
5 return  $D^{(n-1)}$ 
```

The computation of  $D^{(n-1)}$  can be improved if we use the method of **repeated squaring**:

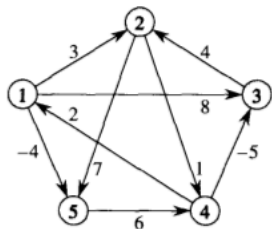
## FASTER-ALL-PAIRS-SHORTEST-PATHS( $W$ )

```
1  $n \leftarrow \text{rows}[W]$ 
2  $D^{(1)} \leftarrow W$ 
3  $m \leftarrow 1$ 
4 while  $n - 1 > m$ 
5     do  $D^{(2m)} \leftarrow \text{EXTEND-SHORTEST-PATHS}(D^{(m)}, D^{(m)})$ 
6      $m \leftarrow 2m$ 
7 return  $D^{(m)}$ 
```

Total time complexity:  $\lceil \log_2(n-1) \rceil \cdot \Theta(n^3) = \Theta(n^3 \cdot \lceil \log_2(n-1) \rceil)$ .

# The bottom-up approach without repeated squaring

## Example



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

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

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

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

# The Floyd-Warshall algorithm

Another characterization of the shortest path

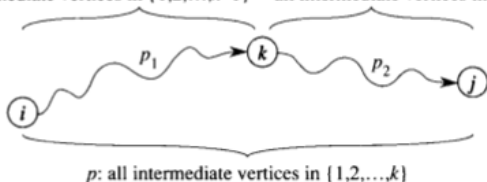
ASSUMPTION: there are no cycles with negative weight

- An **intermediate node** of a simple path  
 $p = (v_1, v_2, \dots, v_{\ell-1}, v_\ell)$  is any node  $v \in \{v_2, v_3, \dots, v_{\ell-1}\}$ .
- The Floyd-Warshall algorithm is based on the following observation:
  - Assume  $V = \{1, 2, \dots, n\}$ . For all  $i, j \in V$  consider all paths from  $i$  to  $j$  whose intermediate nodes are from  $\{1, \dots, k\}$ .  
Let  $i \overset{p}{\rightsquigarrow} j$  be such a minimum-weight path from  $i$  to  $j$ , and let  $d_{i,j}^{[k]}$  be its weight.
  - $d_{i,j}^{[k]}$  can be computed by recursion on  $k$ .

# The Floyd-Warshall algorithm

Another characterization of the shortest path (continued)

- ① If  $k$  is an intermediate node of path  $p$ , we have  $i \overset{p_1}{\rightsquigarrow} k \overset{p_2}{\rightsquigarrow} j$ :  
all intermediate vertices in  $\{1, 2, \dots, k-1\}$     all intermediate vertices in  $\{1, 2, \dots, k-1\}$



where  $p_1$  is minimum-weight path from  $i$  to  $k$  through nodes  $\{1, \dots, k-1\}$ , and  $p_2$  is a minimum-weight path from  $k$  to  $j$  through nodes  $\{1, \dots, k-1\}$ . Thus  $d_{i,j}^{[k]} = d_{i,k}^{[k-1]} + d_{k,j}^{[k-1]}$

- ② If  $j$  is not intermediate node of path  $p$  then all intermediate nodes of  $p$  are from  $\{1, \dots, k-1\}$ , thus  $d_{i,j}^{[k]} = d_{i,j}^{[k-1]}$ .

# The Floyd-Warshall algorithm

Another characterization of the shortest path (continued)

A recursive definition of  $d_{i,j}^{[k]}$ :

$$d_{i,j}^{[k]} := \begin{cases} w_{i,j} & \text{if } k = 0, \\ \min \left( d_{i,j}^{[k-1]}, d_{i,k}^{[k-1]} + d_{k,j}^{[k-1]} \right) & \text{if } k \geq 1. \end{cases}$$

- $d_{i,j}^{[n]}$  is the weight of a shortest path from  $i$  to  $j$  for all  $i, j \in V$ .  
 $\Rightarrow$  we wish to compute the  $n \times n$  matrix  $D^{[n]} = \left( d_{i,j}^{[n]} \right)$ .



# The Floyd-Warshall algorithm

- Computes  $d_{i,j}^{[k]}$  in order of increasing values of  $k$ .

FLOYD-WARSHALL( $W$ )

1  $n := \text{rows}[W]$

2  $D^{[0]} := W$

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

4     **for**  $i := 1$  to  $n$

5         **for**  $j := 1$  to  $n$

6              $d_{ij}^{[k]} := \min \left( d_{ij}^{[k-1]}, d_{ik}^{[k-1]} + d_{kj}^{[k-1]} \right)$

7 **return**  $D^{[n]}$

- Running time complexity:
  - There are 3 nested loops (lines 3-6)
  - The execution time of line 6 takes  $O(1)$  time $\Rightarrow \Theta(n^3)$  time complexity.

# The Floyd-Warshall algorithm

## Extensions

The predecessor matrix  $\Pi$  can be computed incrementally, via a sequence of matrices

$$\Pi^{[0]}, \Pi^{[1]}, \dots, \Pi^{[n]}$$

where  $\Pi = \Pi^{[n]}$  and  $\pi_{i,j}^{[k]}$  is the predecessor of vertex  $j$  on a shortest path from vertex  $i$  with all intermediate nodes in the set  $\{1, 2, \dots, k\}$ .

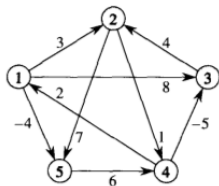
Formally, we have

$$\pi_{i,j}^{[0]} = \begin{cases} \text{nil} & \text{if } i = j \text{ or } w_{i,j} = \infty, \\ i & \text{if } i \neq j \text{ and } w_{i,j} < \infty \end{cases}$$

$$\pi_{i,j}^{[k]} = \begin{cases} \pi_{i,j}^{[k-1]} & \text{if } d_{i,j}^{[k-1]} \leq d_{i,k}^{[k-1]} + d_{k,j}^{[k-1]} \\ \pi_{k,j}^{[k-1]} & \text{if } d_{i,j}^{[k-1]} > d_{i,k}^{[k-1]} + d_{k,j}^{[k-1]} \end{cases}$$

# The Floyd-Warshall algorithm

## Running example



For the graph

we have

$$D^{[0]} = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & \infty & -5 & 0 & \infty \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

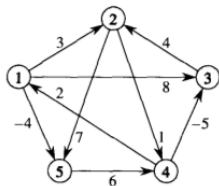
$$D^{[1]} = \begin{pmatrix} 0 & 3 & 8 & \infty & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & \infty & \infty \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

$$\Pi^{[0]} = \begin{pmatrix} nil & 1 & 1 & nil & 1 \\ nil & nil & nil & 2 & 2 \\ nil & 3 & nil & nil & nil \\ 4 & nil & 4 & nil & nil \\ nil & nil & nil & 5 & nil \end{pmatrix}$$

$$\Pi^{[1]} = \begin{pmatrix} nil & 1 & 1 & nil & 1 \\ nil & nil & nil & 2 & 2 \\ nil & 3 & nil & nil & nil \\ 4 & 1 & 4 & nil & 1 \\ nil & nil & nil & 5 & nil \end{pmatrix}$$

# The Floyd-Warshall algorithm

Running example (continued)



For the graph

we have

$$D^{[2]} = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & 5 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

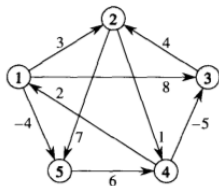
$$D^{[3]} = \begin{pmatrix} 0 & 3 & 8 & 4 & -4 \\ \infty & 0 & \infty & 1 & 7 \\ \infty & 4 & 0 & 5 & 11 \\ 2 & -1 & -5 & 0 & -2 \\ \infty & \infty & \infty & 6 & 0 \end{pmatrix}$$

$$\Pi^{[2]} = \begin{pmatrix} \text{nil} & 1 & 1 & 2 & 1 \\ \text{nil} & \text{nil} & \text{nil} & 2 & 2 \\ \text{nil} & 3 & \text{nil} & 2 & 2 \\ 4 & 1 & 4 & \text{nil} & 1 \\ \text{nil} & \text{nil} & \text{nil} & 5 & \text{nil} \end{pmatrix}$$

$$\Pi^{[3]} = \begin{pmatrix} \text{nil} & 1 & 1 & 2 & 1 \\ \text{nil} & \text{nil} & \text{nil} & 2 & 2 \\ \text{nil} & 3 & \text{nil} & 2 & 2 \\ 4 & 3 & 4 & \text{nil} & 1 \\ \text{nil} & \text{nil} & \text{nil} & 5 & \text{nil} \end{pmatrix}$$

# The Floyd-Warshall algorithm

Running example (continued)



For the graph

we have

$$D^{[4]} = \begin{pmatrix} 0 & 3 & -1 & 4 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix}$$

$$\Pi^{[4]} = \begin{pmatrix} nil & 1 & 4 & 2 & 1 \\ 4 & nil & 4 & 2 & 1 \\ 4 & 3 & nil & 2 & 1 \\ 4 & 3 & 4 & nil & 1 \\ 4 & 3 & 4 & 5 & nil \end{pmatrix}$$

$$D^{[5]} = \begin{pmatrix} 0 & 1 & -3 & 2 & -4 \\ 3 & 0 & -4 & 1 & -1 \\ 7 & 4 & 0 & 5 & 3 \\ 2 & -1 & -5 & 0 & -2 \\ 8 & 5 & 1 & 6 & 0 \end{pmatrix}$$

$$\Pi^{[5]} = \begin{pmatrix} nil & 3 & 4 & 5 & 1 \\ 4 & nil & 4 & 2 & 1 \\ 4 & 3 & nil & 2 & 1 \\ 4 & 3 & 4 & nil & 1 \\ 4 & 3 & 4 & 5 & nil \end{pmatrix}$$