

Dynamic Programming

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CPE 212 Algorithm Design (1/2018)

Topics

- Basic examples
- Knapsack problem
- Warshall's and Floyd's algorithms

Dynamic Programming

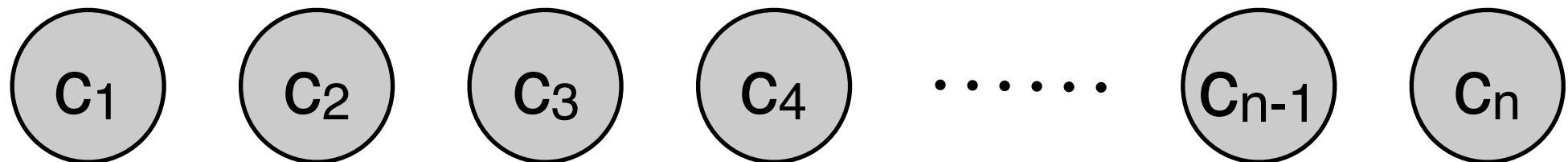
- Dynamic programming was invented by a prominent U.S. mathematician, Richard Bellman, in the 1950s as a general method for *optimizing multistage decision processes*.
- Thus, the word “*programming*” in the name of this technique stands for “*planning*” and does not refer to computer programming.

Dynamic Programming

- Dynamic programming is a technique for solving problems with overlapping subproblems.
- Typically, these subproblems arise from a recurrence relating a given problem's solution to solutions of its smaller subproblems.
- *Rather than solving overlapping subproblems again and again,* dynamic programming suggests solving each of the smaller subproblems only once and recording the results in a table from which a solution to the original problem can then be obtained.

Coin-Row Problem

- Row of 6 coins whose values are 5, 1, 2, 10, 6, 2
 - Pick up the maximum amount of money
 - No two adjacent coins can be picked up.



- In general, values are $c_1, c_2, c_3, \dots, c_n$ not necessarily distinct.

- $F(n)$ = Max amount that can be picked up from the row of n coins.
- Partition all the allowed coin selections into two groups
 - Those including the last coin $G1$
 - Those without the last coin $G2$
- In $G1$, the max amount is $c_n + F(n-2)$
- In $G2$, the max amount is $F(n-1)$
- Therefore

$$F(n) = \max\{c_n + F(n-2), F(n-1)\}, \quad n > 1$$

$$F(0) = 0, \quad F(1) = c_1$$

■ We can compute $F(n)$ by filling the one-row table left to right.

■ Consider the coin row 5, 1, 2, 10, 6, 2.

$$F(n) = \max\{c_n + F(n-2), F(n-1)\}, \quad n > 1$$

$$F(0) = 0, \quad F(1) = c_1$$

$$F[0] = 0, \quad F[1] = c_1 = 5$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5					

$$F[2] = \max\{1 + 0, 5\} = 5$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5				

$$F[3] = \max\{2 + 5, 5\} = 7$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7			

$$F(n) = \max\{c_n + F(n - 2), F(n - 1)\}, \quad n > 1$$

$$F(0) = 0, \quad F(1) = c_1$$

$$F[4] = \max\{10 + 5, 7\} = 15$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15		

$$F[5] = \max\{6 + 7, 15\} = 15$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15	15	

$$F[6] = \max\{2 + 15, 15\} = 17$$

index	0	1	2	3	4	5	6
C		5	1	2	10	6	2
F	0	5	5	7	15	15	17

ALGORITHM *CoinRow*($C[1..n]$)

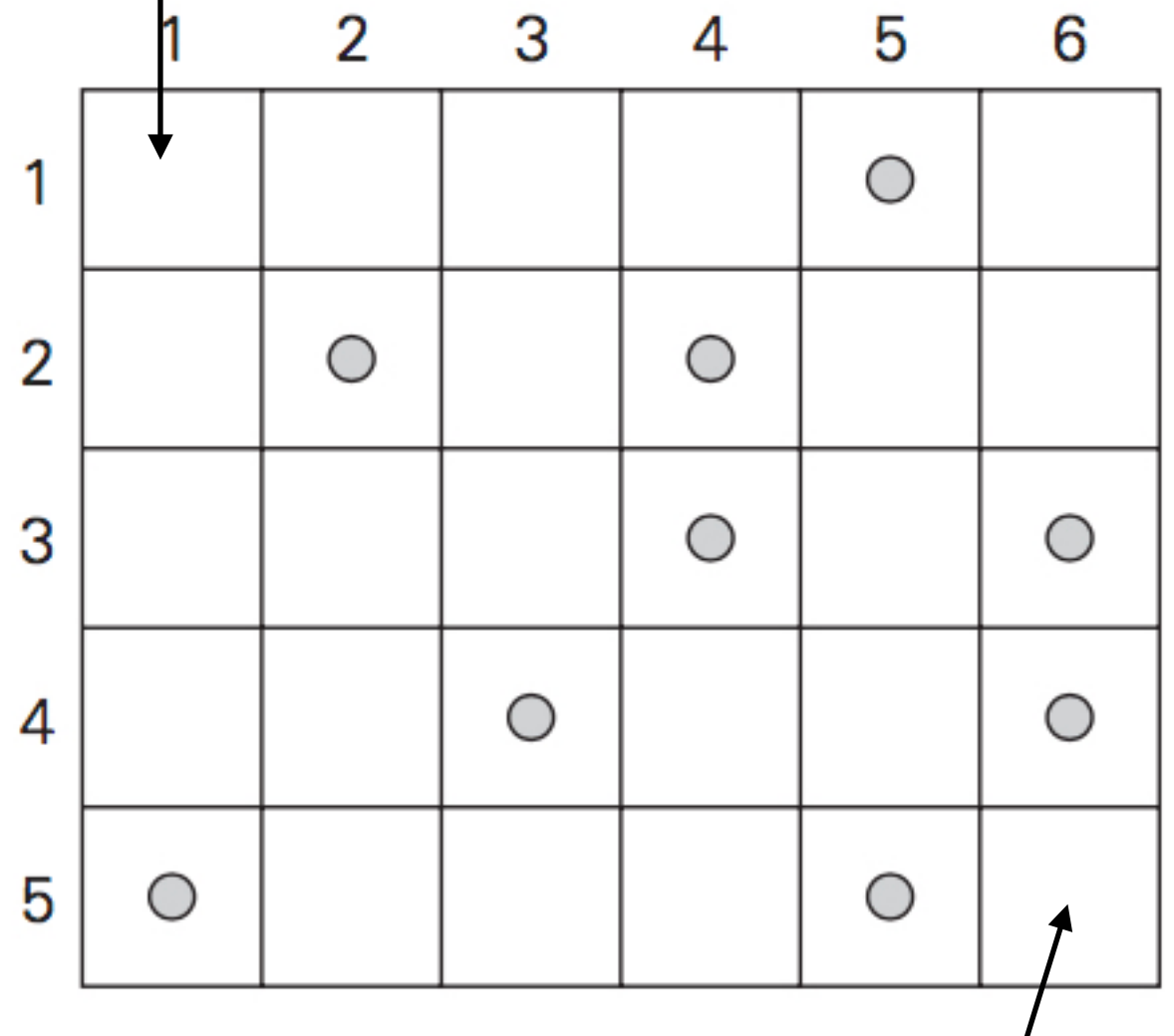
//Applies formula (8.3) bottom up to find the maximum amount of money
//that can be picked up from a coin row without picking two adjacent coins
//Input: Array $C[1..n]$ of positive integers indicating the coin values
//Output: The maximum amount of money that can be picked up
 $F[0] \leftarrow 0; \quad F[1] \leftarrow C[1]$
for $i \leftarrow 2$ **to** n **do**
 $F[i] \leftarrow \max(C[i] + F[i - 2], F[i - 1])$ Also keep track here of
return $F[n]$ which coins give higher value

Time complexity $\Theta(n)$
Space complexity $\Theta(n)$

Coin Collecting Problem

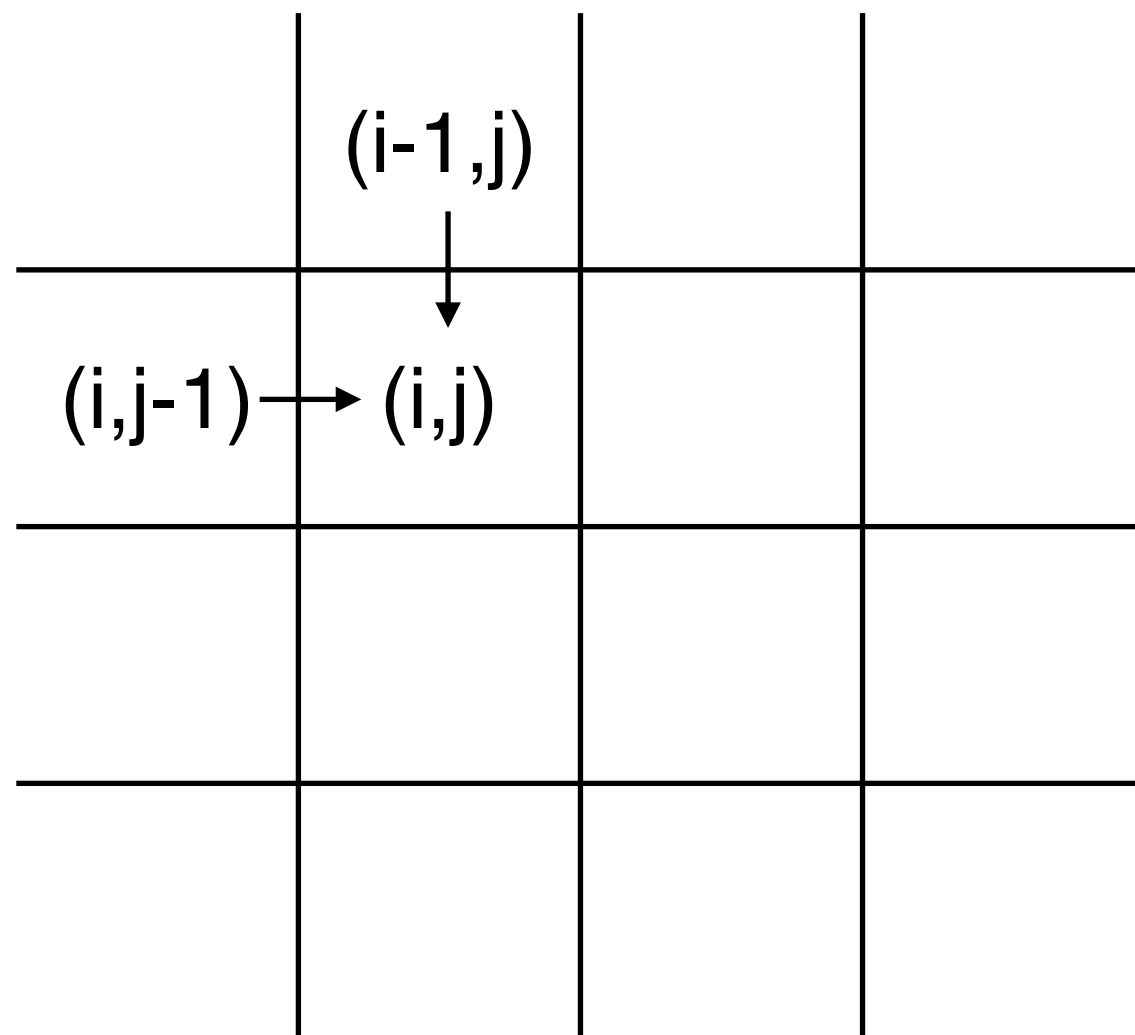
- Coins in cells of an $n \times m$ board.
- From an upper left cell, Robot is to collect as many coins as possible to bottom right.
 - Movement is either one cell down or right.
 - What is the path to follow?

Start here



Finish here

- Let $F(i,j)$ be the largest amount of coins that robot can collect and bring to cell (i,j)
 - Can be reached from adjacent cell on the left or above.
 - Largest # coins brought to these cells : $F(i,j-1)$ and $F(i-1,j)$



■ Therefore, $F(i,j)$ satisfies the following formula:

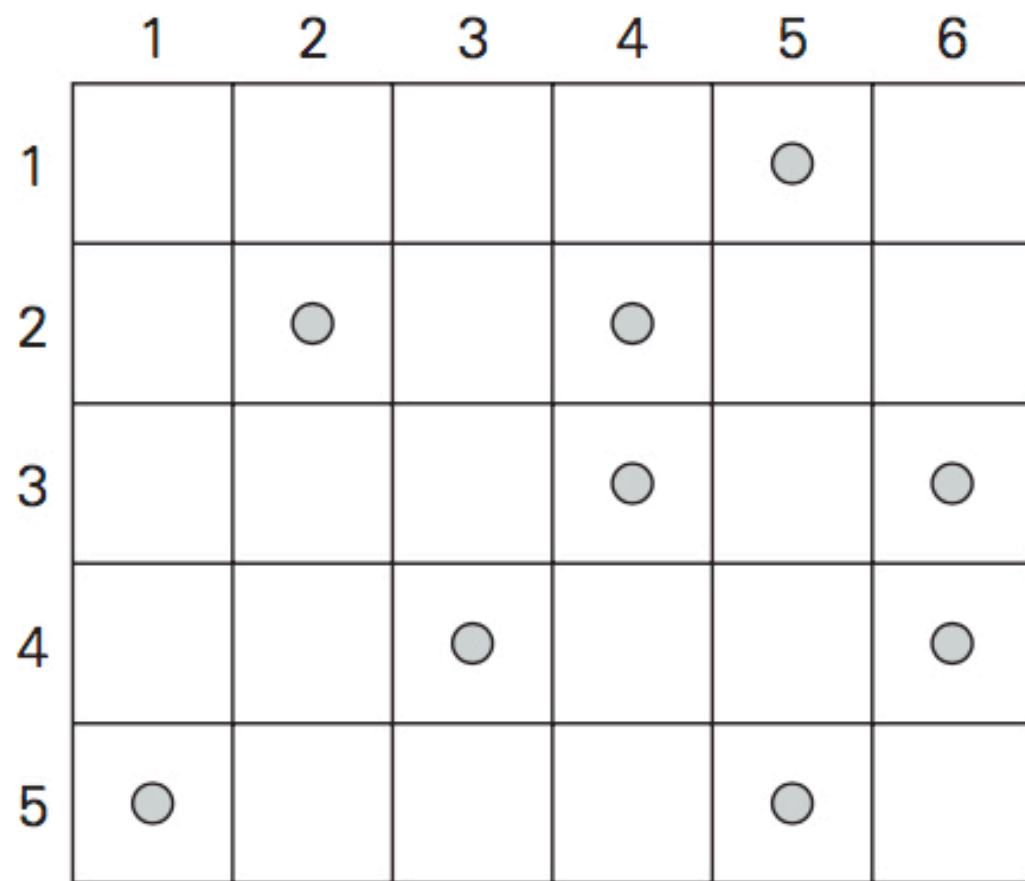
$$F(i, j) = \max\{F(i-1, j), F(i, j-1)\} + c_{ij}, \quad 1 \leq i \leq n, 1 \leq j \leq m$$

$$F(0, j) = 0, \quad 1 \leq j \leq m$$

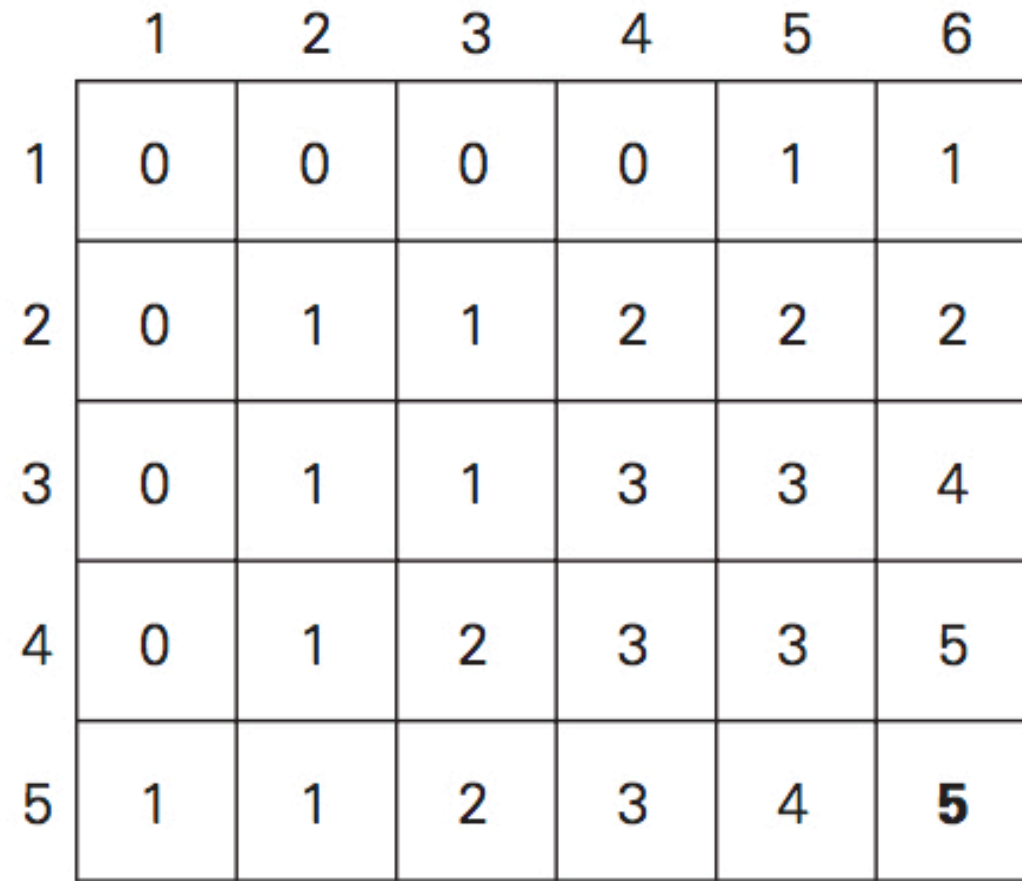
$$F(i, 0) = 0, \quad 1 \leq i \leq n$$

$$c_{ij} = \begin{cases} 1 & \text{coin in cell } (i, j) \\ 0 & \text{no coin in cell } (i, j) \end{cases}$$

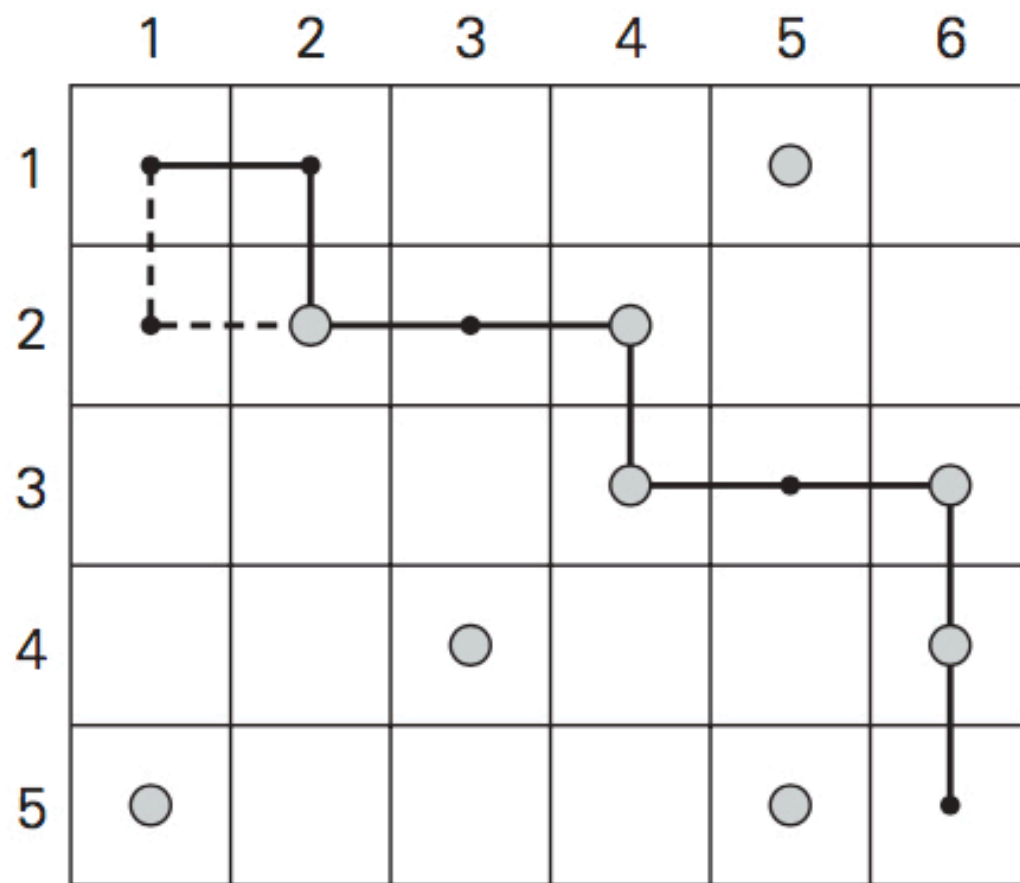
■ Fill in n by m table of $F(i,j)$ values either row-by-row or column-by-column.



(a)



(b)



(c)

General Concepts of Dynamic Programming

- Programming = Planning
- Technique to solve problems with overlapping subproblems.
 - Subproblems typically arise from recurrence relation of between the problem solution and subproblem solutions.
 - Subproblem could be solved only once and results recorded in a table similar to space-time trade-off design.

Knapsack Problem

item	weight	value
1	2	\$12
2	1	\$10
3	3	\$20
4	2	\$15

capacity $W = 5$.

- Select items whose sum of weight does not exceed W and sum of values is maximized.
- n items with known weights w_1, w_2, \dots, w_n , values v_1, v_2, \dots, v_n , and knapsack capacity W

- Let $F(i,j)$ be the *optimal value* obtained from a subset of first i items that fit into the knapsack of capacity j .
- All subsets of first i items can be divided into two cases
 - those not including i^{th} item
 - those including i^{th} item
- For those not including i^{th} item, $F(i,j) = F(i-1,j)$
- For those including i^{th} item ($j - w_i \geq 0$)
 - Optimal subset made up of i^{th} item and optimal subset of first $i - 1$ items in knapsack capacity $j - w_i$
 - Optimal value is therefore $v_i + F(i-1, j-w_i)$

ALGORITHM *MFKnapsack*(i, j)

//Implements the memory function method for the knapsack problem
//Input: A nonnegative integer i indicating the number of the first
// items being considered and a nonnegative integer j indicating
// the knapsack capacity

//Output: The value of an optimal feasible subset of the first i items

//Note: Uses as global variables input arrays $Weights[1..n]$, $Values[1..n]$,
//and table $F[0..n, 0..W]$ whose entries are initialized with -1 's except for
//row 0 and column 0 initialized with 0's

if $F[i, j] < 0$

if $j < Weights[i]$

$value \leftarrow MFKnapsack(i - 1, j)$

else

$value \leftarrow \max(MFKnapsack(i - 1, j),$
 $Values[i] + MFKnapsack(i - 1, j - Weights[i]))$

$F[i, j] \leftarrow value$

return $F[i, j]$

■ Therefore, among all feasible subsets of the first i items,

$$F(i, j) = \begin{cases} \max\{F(i-1, j), v_i + F(i-1, j-w_i)\} & \text{if } j-w_i \geq 0 \\ F(i-1, j) & \text{if } j-w_i < 0 \end{cases}$$

$$F(0, j) = 0, \quad j \geq 0$$

$$F(i, 0) = 0, \quad i \geq 0$$

■ Our goal is to find $F(n, W)$

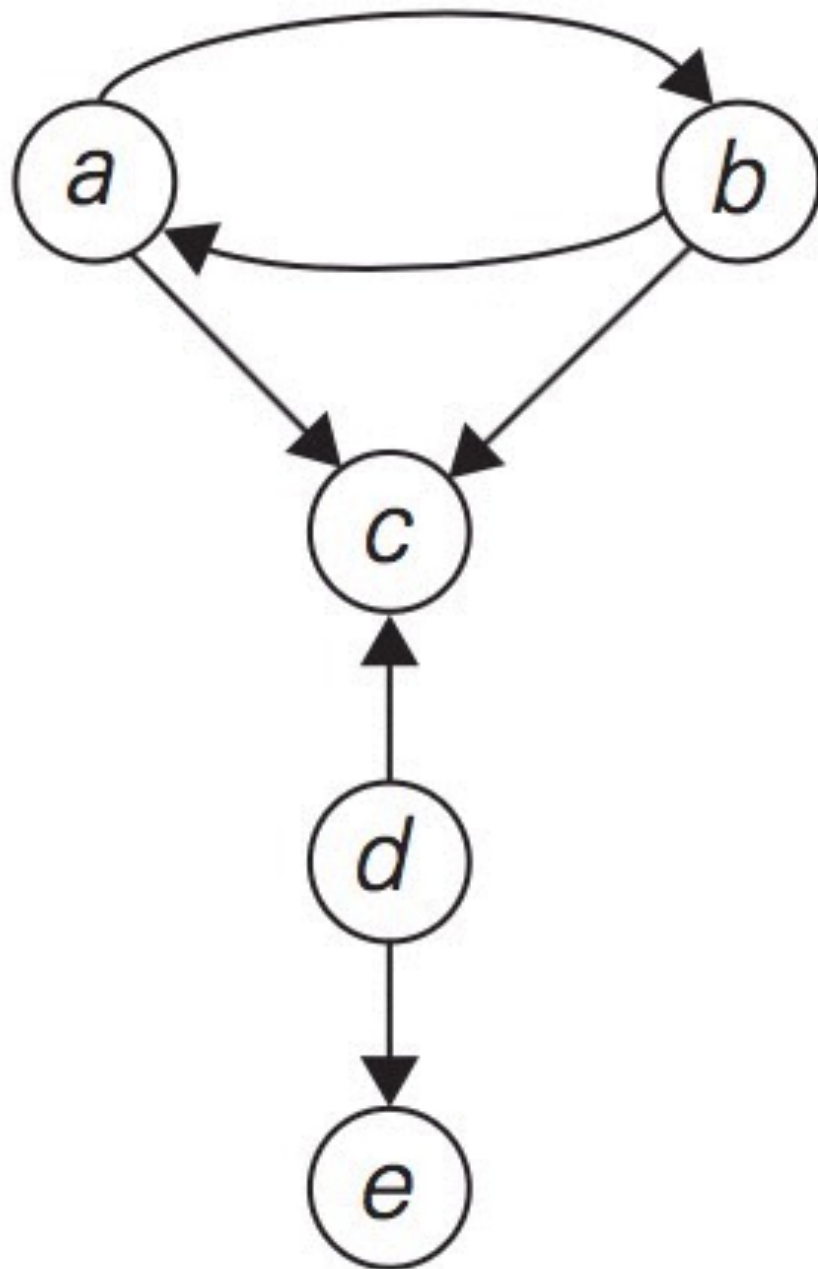
		0	$j-w_i$	j	W
w_i, v_i	0	0	0	0	0
	$i-1$	0	$F(i-1, j-w_i)$	$F(i-1, j)$	
	i	0		$F(i, j)$	
	n	0			goal

item	weight	value
1	2	\$12
2	1	\$10
3	3	\$20
4	2	\$15

capacity $W = 5$.

		capacity j						
		i	0	1	2	3	4	5
		0	0	0	0	0	0	0
$w_1 = 2, v_1 = 12$		1	0	0	12	12	12	12
$w_2 = 1, v_2 = 10$		2	0	10	12	22	22	22
$w_3 = 3, v_3 = 20$		3	0	10	12	22	30	32
$w_4 = 2, v_4 = 15$		4	0	10	15	25	30	37

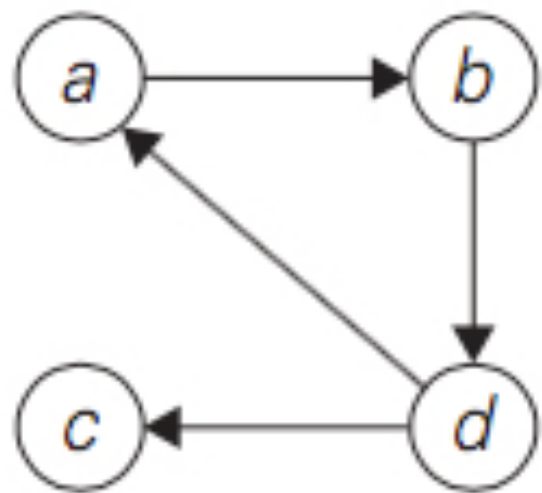
Digraph and Adjacency Matrix



What is the adjacency matrix ?

Transitive Closure of a Directed

- A matrix that tells (in a constant time) if j^{th} vertex is reachable from i^{th} vertex (there exists a directed path from i to j)
 - Represent by a boolean matrix $T = \{t_{ij}\}$



$$A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

$$T = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

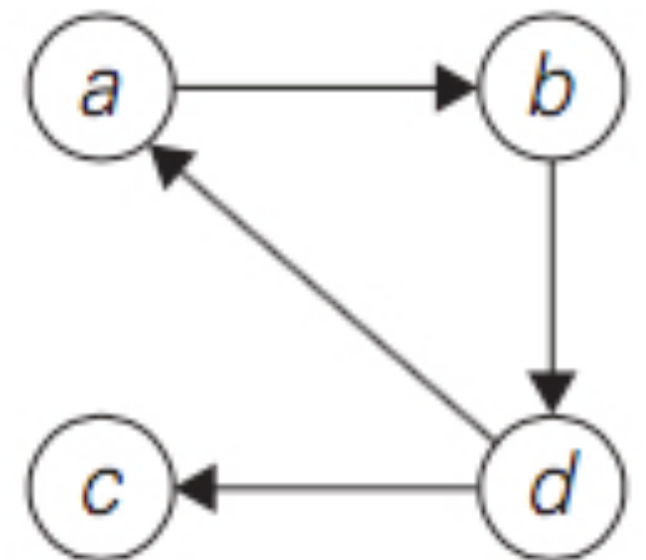
■ Application examples

- Dependency of spreadsheet cells
- Data flow in software design

■ Depth-first-search or Breadth-first-search can be used to generate a transitive closure of a digraph.

- Perform a traversal at i^{th} vertex and fill in columns in the i^{th} row.
- Ex: Try DFS starting at vertex a .

■ What wrong with DFS or BFS?



Warshall's Algorithm

- Label vertices from 1 to n
- Define matrix $R^{(k)}$ whose elements are

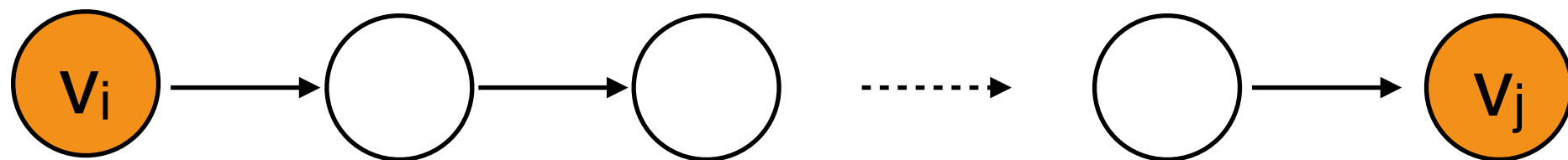
$$r_{ij}^{(k)} = \begin{cases} 1 & \text{there exists a directed path from } i \text{ to } j \text{ with} \\ & \text{intermediate vertices (if any) not higher than } k \\ 0 & \text{otherwise} \end{cases}$$

- Transitive closure is constructed through a series of $n \times n$ boolean matrices $R^{(0)}, R^{(1)}, \dots, R^{(k-1)}, R^{(k)}, \dots, R^{(n)}$.

■ Examples

- $R^{(0)}$ is just the adjacency matrix.
- $R^{(1)}$: Paths only contains the first vertex as intermediate
- $R^{(2)}$: Paths only contains the first two vertices as intermediate
- $R^{(n)}$: Paths contains any vertices as intermediate

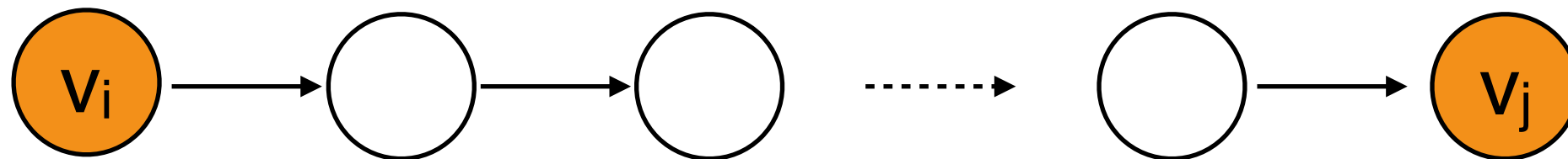
■ $R^{(k)}$ with entry $r_{ij}^{(k)} = 1$ means that there exists a path between i^{th} vertex v_i and j^{th} vertex v_j such that



intermediate vertices $\leq k$

■ Two situations regarding the path are possible.

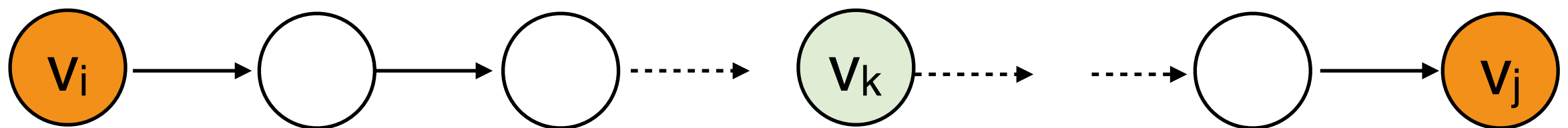
- list of intermediate vertices does not have k^{th} vertex.



intermediate vertices $\leq k-1$

$$r_{ij}^{(k-1)} = 1$$

- list of intermediate vertices have k^{th} vertex.



intermediate
vertices $\leq k-1$

intermediate
vertices $\leq k-1$

$$r_{ik}^{(k-1)} = r_{kj}^{(k-1)} = 1$$

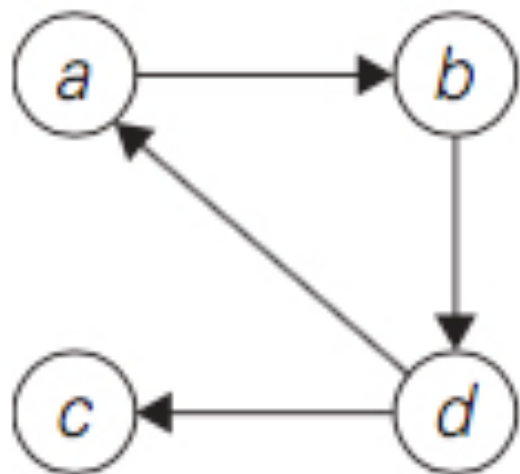
■ Therefore, if $r_{ij}^{(k)} = 1$, then either

$$r_{ij}^{(k)} = r_{ij}^{(k-1)} \quad \text{or} \quad \left(r_{ik}^{(k-1)} \text{ and } r_{kj}^{(k-1)} \right)$$

$$R^{(k-1)} = \begin{array}{cc} & \begin{array}{c} j \\ k \end{array} \\ \begin{array}{c} k \\ i \end{array} & \left[\begin{array}{cc} & \\ \boxed{1} & \\ \uparrow 0 \rightarrow & \boxed{1} \end{array} \right] \end{array} \implies R^{(k)} = \begin{array}{cc} & \begin{array}{c} j \\ k \end{array} \\ \begin{array}{c} k \\ i \end{array} & \left[\begin{array}{cc} & \\ 1 & \\ 1 & 1 \end{array} \right] \end{array}$$

■ Rules for applying Warshall's algorithm by hand:

- If r_{ij} is 1 in $R^{(k-1)}$, it remains 1 in $R^{(k)}$
- If r_{ij} is 0 in $R^{(k-1)}$, it has to be changed to 1 in $R^{(k)}$ if and only if r_{ik} and r_{kj} in $R^{(k-1)}$ are both 1.



$$R^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

$$R^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \end{bmatrix} \end{matrix}$$

$$R^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

Work out $R^{(3)}$ and $R^{(4)}$

ALGORITHM *Warshall*($A[1..n, 1..n]$)

//Implements Warshall's algorithm for computing the transitive closure

//Input: The adjacency matrix A of a digraph with n vertices

//Output: The transitive closure of the digraph

$R^{(0)} \leftarrow A$

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

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

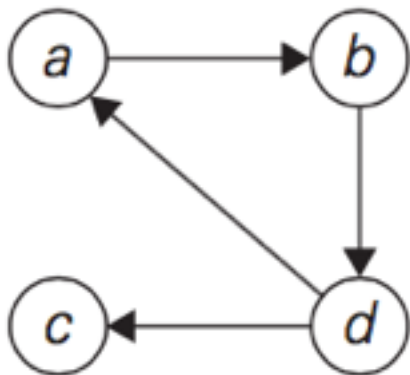
for $j \leftarrow 1$ **to** n **do**

$R^{(k)}[i, j] \leftarrow R^{(k-1)}[i, j] \text{ or } (R^{(k-1)}[i, k] \text{ and } R^{(k-1)}[k, j])$

return $R^{(n)}$

1. Apply Warshall's algorithm to find the transitive closure of the digraph defined by the following adjacency matrix:

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

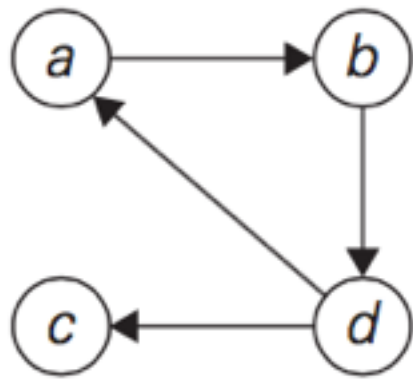


$$R^{(0)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{c|c|c|c} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{array} \right] \end{array}$$

1's reflect the existence of paths with no intermediate vertices ($R^{(0)}$ is just the adjacency matrix); boxed row and column are used for getting $R^{(1)}$.

$$R^{(1)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{c|c|c|c} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & \mathbf{1} & 1 & 0 \end{array} \right] \end{array}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 1, i.e., just vertex a (note a new path from d to b); boxed row and column are used for getting $R^{(2)}$.



$$R^{(0)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{cccc} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{array} \right] \end{array}$$

1's reflect the existence of paths with no intermediate vertices ($R^{(0)}$ is just the adjacency matrix); boxed row and column are used for getting $R^{(1)}$.

$$R^{(1)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{cccc} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 \end{array} \right] \end{array}$$

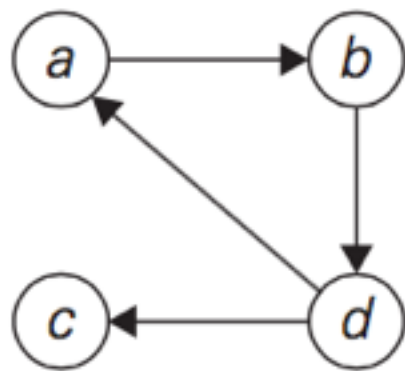
1's reflect the existence of paths with intermediate vertices numbered not higher than 1, i.e., just vertex a (note a new path from d to b); boxed row and column are used for getting $R^{(2)}$.

$$R^{(2)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{cccc} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{array} \right] \end{array}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 2, i.e., a and b (note two new paths); boxed row and column are used for getting $R^{(3)}$.

$$R^{(3)} = \begin{array}{c} a \\ b \\ c \\ d \end{array} \begin{array}{c} a \quad b \quad c \quad d \\ \left[\begin{array}{cccc} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{array} \right] \end{array}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (no new paths); boxed row and column are used for getting $R^{(4)}$.



$$R^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with no intermediate vertices ($R^{(0)}$ is just the adjacency matrix); boxed row and column are used for getting $R^{(1)}$.

$$R^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & \mathbf{1} & 1 & 0 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 1, i.e., just vertex a (note a new path from d to b); boxed row and column are used for getting $R^{(2)}$.

$$R^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & \mathbf{1} \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & \mathbf{1} \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 2, i.e., a and b (note two new paths); boxed row and column are used for getting $R^{(3)}$.

$$R^{(3)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

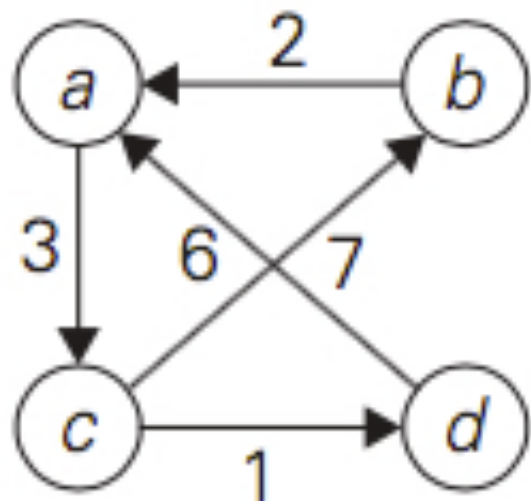
1's reflect the existence of paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (no new paths); boxed row and column are used for getting $R^{(4)}$.

$$R^{(4)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} \mathbf{1} & 1 & \mathbf{1} & 1 \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & 1 \\ 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

1's reflect the existence of paths with intermediate vertices numbered not higher than 4, i.e., a , b , c , and d (note five new paths).

All-Pairs Shortest-Paths Problem

- Consider a positive weighted connected graph W .
- Find the shortest paths from each vertex to all the others.
- Distance matrix D contains the shortest path length from i^{th} vertex to j^{th} vertex.



$$W = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

$$D = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 10 & 3 & 4 \\ 2 & 0 & 5 & 6 \\ 7 & 7 & 0 & 1 \\ 6 & 16 & 9 & 0 \end{bmatrix} \end{matrix}$$

Floyd's Algorithm

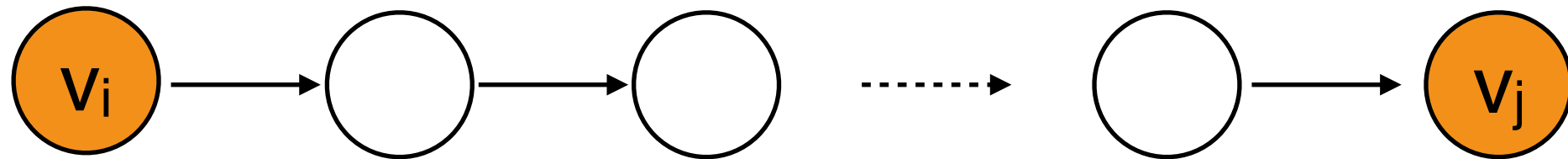
- Compute the distance matrix of a weighted graph with n vertices through a series of $n \times n$ matrices:

$$D^{(0)}, D^{(1)}, \dots, D^{(k-1)}, D^{(k)}, \dots, D^{(n)}$$

- d_{ij} in $D^{(k)}$ is the length of shortest path from i to j with intermediate vertices (if any) not higher than k .
 - $D^{(0)}$ is simply the weight matrix.
 - $D^{(k)}$: Length of shortest paths with only first k vertices as intermediates

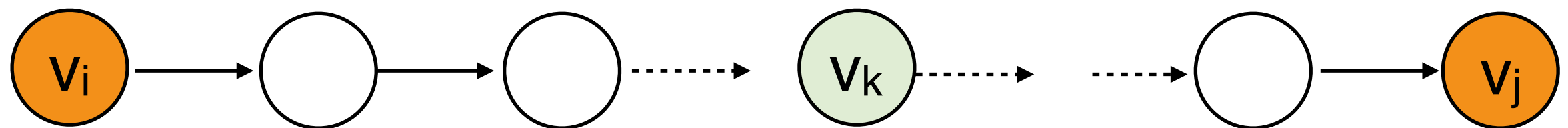
■ Two situations regarding the path for $d_{ij}^{(k)}$ are possible.

● I: List of intermediate vertices does not have k^{th} vertex.



intermediate vertices $\leq k-1$

● II: List of intermediate vertices have k^{th} vertex.



intermediate
vertices $\leq k-1$

intermediate
vertices $\leq k-1$

Floyd's Algorithm

ALGORITHM *Floyd*($W[1..n, 1..n]$)

//Implements Floyd's algorithm for the all-pairs shortest-paths problem

//Input: The weight matrix W of a graph with no negative-length cycle

//Output: The distance matrix of the shortest paths' lengths

$D \leftarrow W$ //is not necessary if W can be overwritten

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

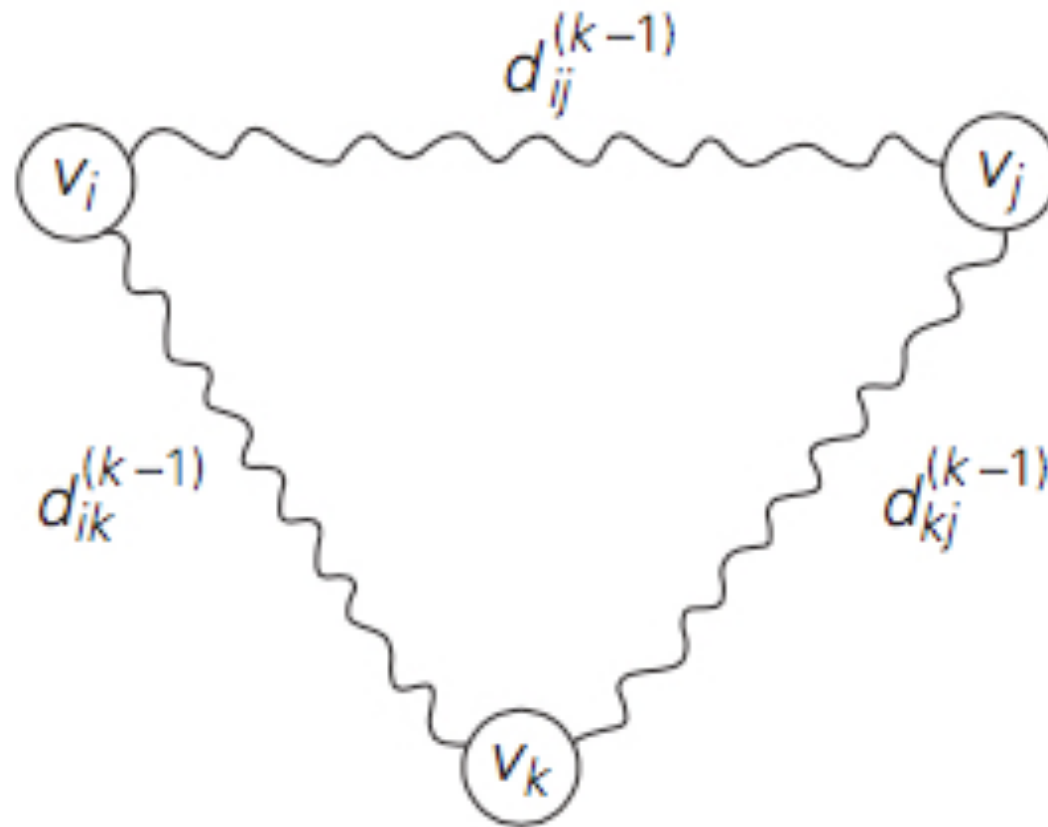
for $i \leftarrow 1$ **to** n **do**

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

$D[i, j] \leftarrow \min\{D[i, j], D[i, k] + D[k, j]\}$

return D

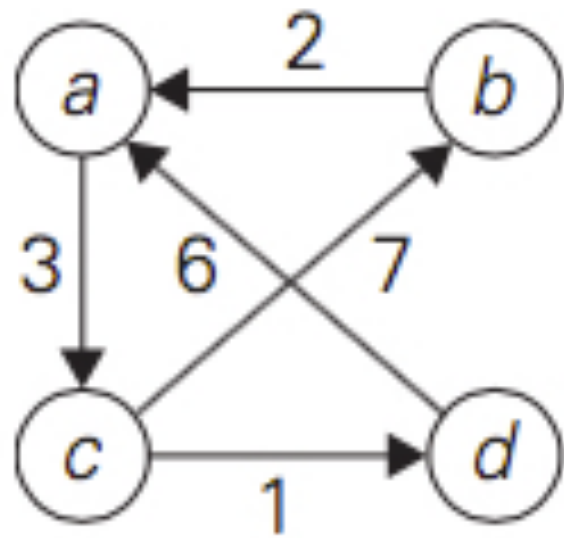
- For subset II, $d_{ij}^{(k)}$ will be different from $d_{ij}^{(k-1)}$



- Both subsets lead to the following recurrence:

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

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



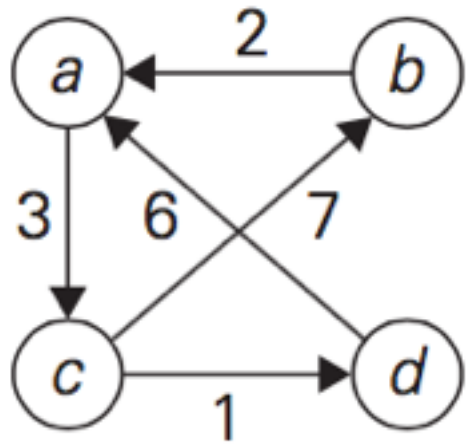
$$D^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

$$D^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \mathbf{5} & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \mathbf{9} & 0 \end{bmatrix} \end{matrix}$$

$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & 5 & \infty \\ \mathbf{9} & 7 & 0 & 1 \\ 6 & \infty & 9 & 0 \end{bmatrix} \end{matrix}$$

Work out $D^{(3)}$ and $D^{(4)}$

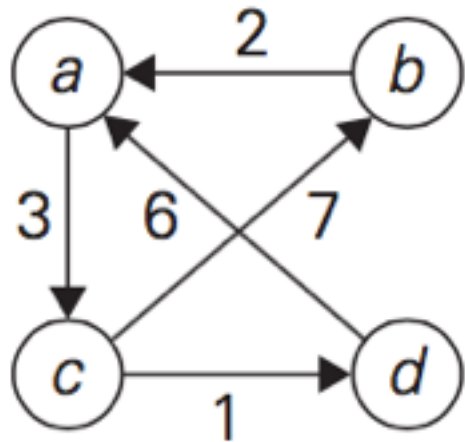
Floyd's Algorithm



$$D^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths
with no intermediate vertices
($D^{(0)}$ is simply the weight matrix).

Floyd's Algorithm



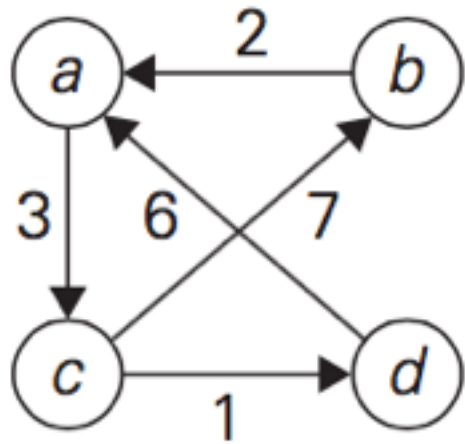
$$D^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with no intermediate vertices ($D^{(0)}$ is simply the weight matrix).

$$D^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \mathbf{5} & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \mathbf{9} & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 1, i.e., just a (note two new shortest paths from b to c and from d to c).

Floyd's Algorithm



$$D^{(0)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \infty & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with no intermediate vertices ($D^{(0)}$ is simply the weight matrix).

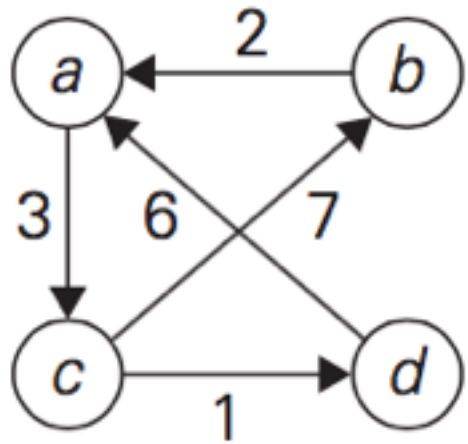
$$D^{(1)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & \mathbf{5} & \infty \\ \infty & 7 & 0 & 1 \\ 6 & \infty & \mathbf{9} & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 1, i.e., just a (note two new shortest paths from b to c and from d to c).

$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & 5 & \infty \\ \mathbf{9} & 7 & 0 & 1 \\ 6 & \infty & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 2, i.e., a and b (note a new shortest path from c to a).

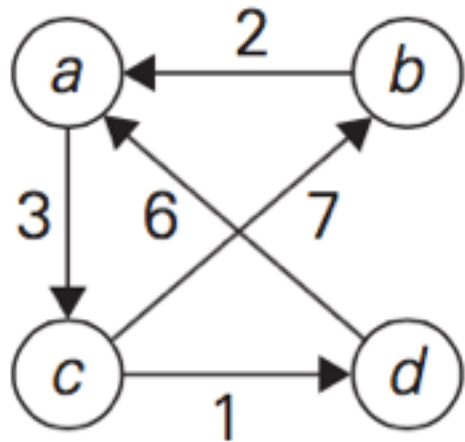
Floyd's Algorithm



$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & 5 & \infty \\ \mathbf{9} & 7 & 0 & 1 \\ 6 & \infty & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 2, i.e., a and b (note a new shortest path from c to a).

Floyd's Algorithm



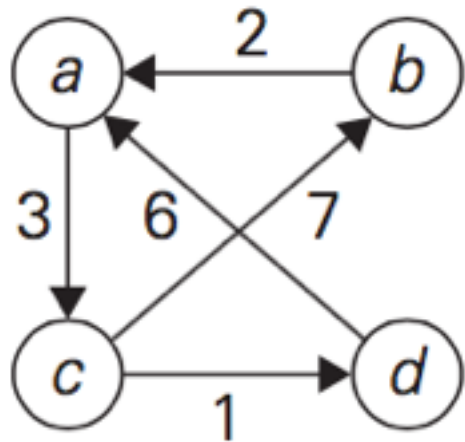
$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & 5 & \infty \\ \mathbf{9} & 7 & 0 & 1 \\ 6 & \infty & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 2, i.e., a and b (note a new shortest path from c to a).

$$D^{(3)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \mathbf{10} & 3 & \mathbf{4} \\ 2 & 0 & 5 & \mathbf{6} \\ 9 & 7 & 0 & 1 \\ \mathbf{6} & \mathbf{16} & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (note four new shortest paths from a to b , from a to d , from b to d , and from d to b).

Floyd's Algorithm



$$D^{(2)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \infty & 3 & \infty \\ 2 & 0 & 5 & \infty \\ \mathbf{9} & 7 & 0 & 1 \\ 6 & \infty & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 2, i.e., a and b (note a new shortest path from c to a).

$$D^{(3)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & \mathbf{10} & 3 & \mathbf{4} \\ 2 & 0 & 5 & \mathbf{6} \\ 9 & 7 & 0 & 1 \\ \mathbf{6} & \mathbf{16} & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 3, i.e., a , b , and c (note four new shortest paths from a to b , from a to d , from b to d , and from d to b).

$$D^{(4)} = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 10 & 3 & 4 \\ 2 & 0 & 5 & 6 \\ \mathbf{7} & 7 & 0 & 1 \\ 6 & 16 & 9 & 0 \end{bmatrix} \end{matrix}$$

Lengths of the shortest paths with intermediate vertices numbered not higher than 4, i.e., a , b , c , and d (note a new shortest path from c to a).