

DATA MINING ON SOCIAL NETWORKS

Laboratory Lectures Notes

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Data Files:

- authors.mat: [1155x3] cell array **authors** storing column-wise
 - author_id
 - author_surname
 - author_firstname
- ICMB_2002.mat ... ICM_2013.mat: [1155 × 1155] matrices
array_2002 ... array_2013

TEMPORAL ADJACENCY MATRICES

Each $\mathbf{W}_t \in M_{1155 \times 1155}$ with $t \in \{2002 \dots 2013\}$ is a symmetric adjacency matrix where the $\mathbf{W}_t(i, j)$ elements quantify the number of papers that have been co-authored between authors i and j .

Specifically, $\mathbf{W}_t = [\mathbf{W}_t(i, j)]$ such that:

$$W_t(i, j) = \begin{cases} \# \text{ papers co-authored between authors } i \text{ and } j \text{ at time } t, i \neq j; \\ \# \text{ of papers authored by author } i \text{ at time } t, i = j; \end{cases}$$

* Information along the diagonal elements will be discarded.

CO-AUTHORSHIP NETWORK: OVERALL ADJACENCY MATRIX

The final co-authorship network weight matrix $\mathbf{W}_0 \in M_{1155 \times 1155}$ to be constructed will be of the following form:

$\mathbf{W}_0 = [W_0(i, j)]$ where:

$$W_0(i, j) = \begin{cases} 1, & i \neq j \text{ when authors } i \text{ and } j \text{ have at least one paper in common;} \\ 0, & i = j. \end{cases}$$

The previous equation may be equivalently expressed as:

$$W_0(i, j) = \begin{cases} 1, & \sum_t W_t(i, j) > 1 \quad i \neq j; \\ 0, & i = j. \end{cases}$$

CO-AUTHORSHIP NETWORK: CODE REQUIREMENTS

Matlab Routines to be Implemented:

1. Load separate weight-matrices and construct overall network weight matrix.
2. Compute Degree Centrality Measure.
3. Construct the Degree Centrality distribution graph.
4. Report top N authors ranked by Degree Centrality (or any other centrality measure).
5. Implement algorithm for extracting connected components (**Breadth First Search Algorithm**).
6. Report top N connected components (ranked by size).
7. Implement Shortest Path extraction algorithm from predecessor matrix (**Floyd-Warshall Algorithm**).

COAUTHORSHIPNETWORKMANIPULATION.M I

```
1  clc
2  clear all
3  % Set the period of years.
4  Years = [2002:1:2013];
5  YearsNum = length(Years);
6  % Load weight matrices for each year.
7  for year = Years
8      filename = strcat(['ICMB-' num2str(year) '.mat'
9                          ]);
9      load(filename);
10 end;
11 % Load authors' names.
```

COAUTHORSHIPNETWORKMANIPULATION.M II

```
12 load('authors.mat');
13 % Set a container storing the weight matrices for
14 % all years.
15 ICMB = cell(1,numel(Years));
16 % Populate cell array
17 for y = 1:YearsNum
18     ICMB{y} = eval(genvarname(strcat(['array_'
19                                     num2str(Years(y))])));
19 end;
20 % Get the number of nodes N.
21 N = size(ICMB{1},1);
22 % Construct the overall graph weight matrix.
23 W = zeros(N,N);
```

COAUTHORSHIPNETWORKMANIPULATION.M III

```
24 for y = 1:1:YearsNum
25     W = W + ICMB{y};
26 end;
27 % Set up a vector of indices pointing to the
28 % diagonal elements of the weight matrix W.
29 Idiag = [1:N+1:N*N];
30
31 % Re-initialize the overall weight matrix W so that
32 % fundamental social network analysis tasks can be
33 % performed. W should be a binary adjacency matrix
34 % so that  $W[i,j] = 1$  indicates the presence of an
35 % edge between authors  $i$  and  $j$ . Moreover, the
```


COAUTHORSHIPNETWORKMANIPULATION.M IV

```
36 % diagonal elements of W should also be set to zero
    .
37 Wo = W;
38 Wo(Wo>1) = 1;
39 Wo(1diag) = 0;
40 % Extract Degree Centrality measure for each author
    .
41 Degrees = sum(Wo,2);
```

DEGREECENTRALITYDISTRIBUTION.M I

```
1 function [H] = DegreeCentralityDistribution(Degrees
    )
2 % This function computes and displays the Degree
3 % Centrality Distribution for a given vector of
4 % degree centralities.
5 min_degree = min(Degrees);
6 max_degree = max(Degrees);
7 degrees_range = [min_degree:max_degree];
8 H = hist(Degrees,degrees_range);
9 figure('Name','Degree Centrality Distribution');
10 bar(degrees_range,H);
11 axis([min_degree-1 max_degree+1 min(H) max(H)+5]);
```

```
12 xlabel('Degrees');  
13 ylabel('Absolute Frequency');  
14 grid on  
15 end
```

REPORTTOPNAUTHORS.M I

```
1 function ReportTopNAuthors(MeasureValues ,  
    MeasureName ,N, authors )  
2 % This function reports the top N authors ranked by  
3 % the measure identified by the input parameter  
4 % MeasureName. The corresponding measure values are  
5 % stored within the vector MeasureValues. The  
6 % number of N and the complete list of authors'  
7 % names are also given as input to the function.  
8 [SortedValues ,SortedIndices] = sort(MeasureValues ,'  
    descend ');  
9 TopNSortedValues = SortedValues(1:N);  
10 TopNSortedIndices = SortedIndices(1:N);
```

REPORTTOPNAUTHORS.M II

```
11 TopNAuthorsFirstNames = authors(TopNSortedIndices
    ,3);
12 TopNAuthorsSurNames = authors(TopNSortedIndices,2);
13 % Report Top N Authors' List.
14 fprintf('Top %d Authors according to %s\n',N,
    MeasureName);
15 for k = 1:1:N
16     fprintf('%s %s: %d\n',TopNAuthorsSurNames{k},
    TopNAuthorsFirstNames{k},TopNSortedValues(k)
    ));
17 end;
18 end
```

```
1 % Compute and display the degree centrality
2 % distribution.
3 H = DegreeCentralityDistribution(Degrees);
4 % Report top 10 authors according to
5 % Degree Centrality.
6 No = 10;
7 MeasureName = 'Degree Centrality';
8 MeasureValues = Degrees;
9 ReportTopNAuthors(MeasureValues, MeasureName, No,
    authors);
```

CONNECTED COMPONENTS I

In graph theory, a **connected component** (or just a component) of an undirected graph is a subgraph in which any two vertices are connected to each other by paths.

The connected components of a graph may be alternatively defined through the equivalence classes (induced subgraphs) of an equivalence relation.

Let $G = (V, E)$ where $V = \{v_1, v_2, \dots, v_n\}$ and $E \subset V \times V$.

Let $R \subset V \times V$, defined as:

$$(u,v) \in R \Leftrightarrow uRv, \text{ if vertex } (v) \text{ is reachable from } (u)$$

R is an equivalence relation because it has the following properties:

CONNECTED COMPONENTS II

1. reflexivity:

$$\forall u \in V, uRu.$$

(it holds since each vertex is reachable through the trivial path of zero length connecting each vertex to itself.)

2. symmetricity:

$$\forall (u, v) \in V^2, u \neq v : uRv \Rightarrow vRu.$$

(it holds since within an undirected graph the same path from (u) to (v) can be traversed backwards).

3. transitivity:

$$\forall (u, v, z) \in V^3, u \neq v \neq z : uRv \wedge vRz \Rightarrow uRz$$

(it holds since the path from (u) to (z) can be constructed through the concatenation of paths from (u) to (v) and from (v) to (z).

BREADTH FIRST SEARCH: ALGORITHM

Input:

- A graph $G = (V, E)$
- A vertex $v \in V$.
- A set **Visited** of already Visited nodes initialized to the empty set ($Visited = \{\emptyset\}$).

Output:

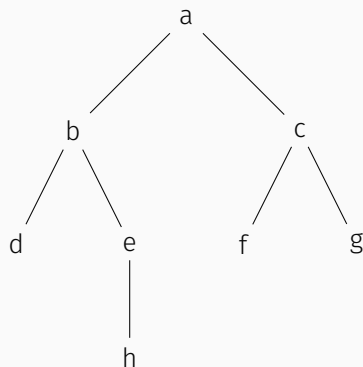
- A set **Reachable** of vertices that are reachable from vertex $v \in V$.

BREADTH FIRST SEARCH: ALGORITHM PSEUDO CODE

```
1 Procedure BFS(G,v,Visited):
2   Reachable = {};
3   Q = []; %Let Q be an empty queue.
4   Q.enqueue(v);
5   Visited = Visited U {v};
6   while Q is not empty:
7     v = Q.dequeue()
8     Reachable ← Reachable U {v}
9     % N(v) denotes the neighborhood of v.
10    for all w in N(v):
11      if w not in Visited:
12        Q.enqueue(w)
13        Visited = Visited U {w}
```

BREADTH FIRST SEARCH EXAMPLE I

Apply the BFS algorithm on the following graph:



NEIGHBOURS' LIST STRUCTURE:

NL(a)={b,c}

NL(b)={a,d,e}

NL(c)={a,f,g}

NL(d)={b}

NL(e)={b,h}

NL(f)={c}

NL(g)={c}

NL(h)={e}

BREADTH FIRST SEARCH EXAMPLE II

1. Calling Procedure BFS: $\text{BFS}(G, a, \{\emptyset\})$
2. Initialization:
 - $v = a$
 - $Visited = \{\emptyset\}$
 - $Reachable = \{\emptyset\}$
 - $Q = []$
3. Enqueue Operation: $(Q = [a], Visited = \{a\})$
4. While Loop Execution:
 - 4.1 $Q = [a] \neq []$:
 - Dequeue Operation: $(v = a, Q = [])$
 - $Reachable = \{a\}$
 - $NL(a) = \{b, c\}$
 - Enqueue Operation: $(Q = [b], Visited = \{a, b\})$
 - Enqueue Operation: $(Q = [b, c], Visited = \{a, b, c\})$

BREADTH FIRST SEARCH EXAMPLE III

4.2 $Q = [b, c] \neq []$:

- Dequeue Operation: ($v = b, Q = [c]$)
- $Reachable = \{a, b\}$
- $NL(b) = \{a, d, e\}$
- No Enqueue Operation: a is already visited
- Enqueue Operation: ($Q = [c, d], Visited = \{a, b, c, d\}$)
- Enqueue Operation: ($Q = [c, d, e], Visited = \{a, b, c, d, e\}$)

4.3 $Q = [c, d, e] \neq []$:

- Dequeue Operation: ($v = c, Q = [d, e]$)
- $Reachable = \{a, b, c\}$
- $NL(c) = \{a, f, g\}$
- No Enqueue Operation: a is already visited
- Enqueue Operation: ($Q = [d, e, f], Visited = \{a, b, c, d, e, f\}$)
- Enqueue Operation: ($Q = [d, e, f, g], Visited = \{a, b, c, d, e, f, g\}$)

4.4 $Q = [d, e, f, g] \neq []$:

- Dequeue Operation: ($v = d, Q = [e, f, g]$)

BREADTH FIRST SEARCH EXAMPLE IV

- $Reachable = \{a, b, c, d\}$
- $NL(d) = \{b\}$
- No Enqueue Operation: b is already visited

4.5 $Q = [e, f, g] \neq []$:

- Dequeue Operation: ($v = e, Q = [f, g]$)
- $Reachable = \{a, b, c, d, e\}$
- $NL(e) = \{b, h\}$
- No Enqueue Operation: b is already visited
- Enqueue Operation: ($Q = [f, g, h], Visited = \{a, b, c, d, e, f, g, h\}$)

4.6 $Q = [f, g, h] \neq []$:

- Dequeue Operation: ($v = f, Q = [g, h]$)
- $Reachable = \{a, b, c, d, e, f\}$
- $NL(f) = \{c\}$
- No Enqueue Operation: c is already visited

4.7 $Q = [g, h] \neq []$:

BREADTH FIRST SEARCH EXAMPLE V

- Dequeue Operation: ($v = g$, $Q = [h]$)
- $Reachable = \{a, b, c, d, e, f, g\}$
- $NL(g) = \{c\}$
- No Enqueue Operation: c is already visited

4.8 $Q = [h] \neq []$:

- Dequeue Operation: ($v = h$, $Q = []$)
- $Reachable = \{a, b, c, d, e, f, g, h\}$
- $NL(h) = \{e\}$
- No Enqueue Operation: e is already visited

4.9 $Q = []$ END OF WHILE LOOP

NEIGHBOURLIST.M I

```
1 function [NL] = NeighboursList(W)
2 % This function extracts the neighbors' list
3 % corresponding to the weight matrix W
4 % which is assumed to be the binary matrix
5 % indicating the presence or absence of an edge
6 % between a given pair of nodes. Diagonal
7 % elements of matrix W are also assumed to be zero.
8 % NL is a cell array of vectors such that
9 % the element NL{u} stores the indices of
10 % nodes that are reachable from node u.
11 nodes_num = size(W,1);
12 NL = cell(1,nodes_num);
```


NEIGHBOURLIST.M II

```
13 for v = 1:1:nodes_num
14     NL{v} = find(W(v,:) == 1);
15 end
16 end
```

CONNECTEDCOMPONENTS.M I

```
1 function [C] = ConnectedComponents(NL)
2 % This function extracts the connected components
3 % of a given undirected graph whose neighbors' list
4 % NL is given as input. C is a cell array of
5 % vectors so that each vector stores the indices of
6 % each connected component.
7
8 % Initialize the cell array C storing the connected
9 % components of the graph.
10 C = cell(1,0);
11 % Get the number of graph nodes.
12 nodes_num = length(NL);
```

CONNECTEDCOMPONENTS.M II

```
13 % Mark all nodes as unvisited.
14 visited = false * ones(1,nodes_num);
15 % Initialize the number of connected components
16 % found so far.
17 components_num = 0;
18 for v = 1:1:nodes_num
19     % If v is not visited yet, it's the start of a
20     % newly discovered component containing v.
21
22     % Process the component containing v.
23     if(~visited(v))
24         components_num = components_num + 1;
25         % Initialize component container.
```

CONNECTEDCOMPONENTS.M III

```
26     component = [];  
27     % Initialize queue for implementing  
28     % breadth-first search.  
29     Q = [];  
30     % Start the traversal from node v.  
31     Q = enqueue(Q,v);  
32     visited(v) = true;  
33     while(~isempty(Q))  
34         [Q,w] = dequeue(Q);  
35         % w is a node in this component.  
36         component = [component,w];  
37         % Get all nodes neighboring w.  
38         node_neighbours = NL{w};
```

CONNECTEDCOMPONENTS.M IV

```
39     % Traverse each unvisited node
40     % neighboring w.
41     for node_index = 1:1:length(
        node_neighbours)
42         node = node_neighbours(node_index);
43         if(~visited(node))
44             % Another node within the
45             % current component has been
46             % found.
47             visited(node) = true;
48             Q = enqueue(Q,node);
49         end
50     end
```

CONNECTEDCOMPONENTS.M V

```
51         end
52         C{components_num} = component;
53     end
54 end
55 function [Q] = enqueue(Q,element)
56 % This is a sub-function implementing the
57 % enqueue operation within a queue
58 % which is realized as a vector of elements
59 Q = [Q,element];
60 end
61 function [Q,element] = dequeue(Q)
62 % This is a sub-function implementing the
63 % dequeue operation within a queue
```

```
64     % which is realized as a vector of elements.  
65     element = Q(1);  
66     Q = Q(2:end);  
67     end  
68 end
```

REPORTTOPNCONNECTEDCOMPONENTS.M I

```
1 function [TopNComponentsSizes,TopNComponentsIndices
      ] = ReportTopNConnectedComponents(C,N,authors)
2
3 % This function reports the top N
4 % (measured by size) connected components
5 % of the co-authorship network that are stored in
6 % cell array C. The number of top N components and
7 % the initial authors' list are passed as
8 % input arguments to the function.
9
10 % Get the number of connected components.
11 components_num = length(C);
```


REPORTTOPNCONNECTEDCOMPONENTS.M II

```
12 % Get the size of each connected component.
13 components_sizes = zeros(1,components_num);
14 for k = 1:1:components_num
15     components_sizes(k) = length(C{k});
16 end
17 % Sort connected components sizes in descending
18 % order.
19 [SortedComponentsSizes,SortedComponentsIndices] =
20     sort(components_sizes, 'descend');
21 % Get the top N connected components sizes and
22 % corresponding indices.
23 TopNComponentsSizes = SortedComponentsSizes(1:N);
```

REPORTTOPNCONNECTEDCOMPONENTS.M III

```
23 TopNComponentsIndices = SortedComponentsIndices(1:N
    );
24
25 % Report Connected Components.
26 % Cycle through the top N connected components:
27 for n = 1:1:N
28     component_index = TopNComponentsIndices(n);
29     component_size = TopNComponentsSizes(n);
30     component = C{component_index};
31     fprintf('Component %d of size %d\n',
        component_index, component_size);
32 % Cycle through the authors of each connected
33 % component:
```

REPORTTOPNCONNECTEDCOMPONENTS.M IV

```
34     for m = 1:1:component_size
35         author_index = component(m);
36         author_firstname = authors(author_index,3);
37         author_lastname = authors(author_index,2);
38         fprintf( '%d: %s %s\n' ,m,cell2mat(
                author_lastname),cell2mat(
                author_firstname));
39     end
40 end
41
42 end
```

```
1 % Extract connected components of co-authorship
2 % network.
3
4 % Initially set the corresponding
5 % NeighboursList.
6 NL = NeighboursList(Wo);
7 C = ConnectedComponents(NL);
8
9 % Report top 6 connected components of the
10 % co-authorship network.
11 No = 6;
12 [TopNComponentsSizes, TopNComponentsIndices] =
    ReportTopNConnectedComponents(C, No, authors);
```

FLOYD-WARSHALL: PROBLEM DEFINITION

Find the shortest path between every pair (v_i, v_j) of vertices on a graph $G = (V, E)$ where $V = \{v_1, \dots, v_n\}$ and $E \subset V \times V$.

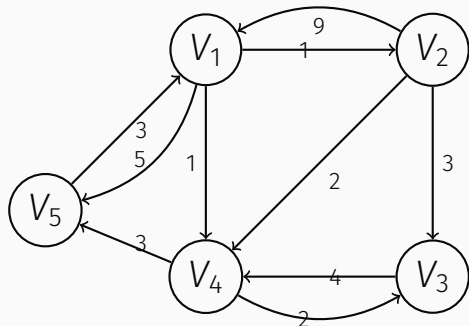
The graph may contain negative edges but not cycles with cumulative weight negative.

Weight-Matrix Representation:

- $W(i,j) = 0$, if $i=j$
- $W(i,j) = \infty$, if there is no edge between i and j with $i \neq j$.
- $W(i,j) =$ "actual weight" of the edge (i,j) with $i \neq j$.

FLOYD-WARSHALL: EXAMPLE GRAPH REPRESENTATION I

- Example Graph:



FLOYD-WARSHALL: EXAMPLE GRAPH REPRESENTATION II

· Weight Matrix:

$$W = \begin{bmatrix} 0 & 1 & \infty & 1 & 5 \\ 9 & 0 & 3 & 2 & \infty \\ \infty & \infty & 0 & 4 & \infty \\ \infty & \infty & 2 & 0 & 3 \\ 3 & \infty & \infty & \infty & 0 \end{bmatrix}$$

FLOYD-WARSHALL: SMALLER PROBLEMS

How can we define the shortest distance d_{ij} between nodes v_i and v_j in terms of “smaller” problems?

One way is to restrict the paths to include vertices exclusively from a restricted subset V^* .

Subset V^* is initially empty ($V_{(0)}^* = \emptyset$).

Finally, subset V^* will contain all possible intermediate nodes ($V_{(n)}^* = V$).

Let $D^{(k)}[i, j]$ denote the weight of the shortest path from v_i to v_j using only the vertices from the set $V_{(k)}^* = \{v_1, v_2, \dots, v_k\}$ as intermediate vertices in the path.

- $D^{(0)} = W$
- $D^{(n)} = D$ (which is the goal matrix)

FLOYD-WARSHALL: RECURSIVE DEFINITION I

How do we compute $D^{(k)}$ from $D^{(k-1)}$?

During the execution of the k -th step of the Floyd-Warshall algorithm, matrix $D^{(k-1)}$ has been computed based on the subset of intermediate nodes: $V_{(k-1)}^* = \{v_1, v_2, \dots, v_{k-1}\}$.

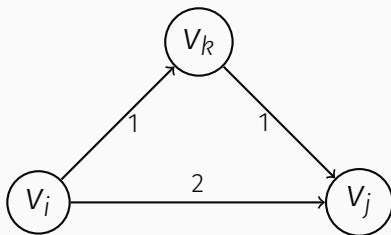
Case 1: The shortest path from v_i to v_j is composed by utilizing nodes from the set of intermediate vertices $V_{(k)}^*$ such that $v_k \notin V_{(k)}^*$. Then,

$$D^{(k)}[i, j] = D^{(k-1)}[i, j]$$

Case 2: The shortest path from v_i to v_j is composed by utilizing nodes from the set of intermediate vertices $V_{(k)}^*$ such that $v_k \in V_{(k)}^*$. Then,

$$D^{(k)}[i, j] = D^{(k-1)}[i, k] + D^{(k-1)}[k, j]$$

FLOYD-WARSHALL: GRAPHICAL REPRESENTATION



1: shortest path using intermediate vertices $\{v_1, v_2, \dots, v_k\}$

2: shortest path using intermediate vertices $\{v_1, v_2, \dots, v_{k-1}\}$

FLOYD-WARSHALL: RECURSIVE DEFINITION II

Since,

$$D^{(k)} = \begin{cases} D^{(k-1)}[i, j], & \text{if node } v_k \text{ is not included;} \\ D^{(k-1)}[i, k] + D^{(k-1)}[k, j], & \text{if node } v_k \text{ is included} \end{cases}$$

we may conclude that:

$$D^{(k)}[i, j] = \min\{D^{(k-1)}[i, j], D^{(k-1)}[i, k] + D^{(k-1)}[k, j]\}$$

FLOYD-WARSHALL: PREDECESSOR MATRIX \mathbf{P}

\mathbf{P} is an index matrix that can be used for extracting the full sequence of nodes that compose the shortest path between any given pair of vertices.

1. Matrix \mathbf{P} is initialized with zeros ($\mathbf{P} = \mathbf{0}_{n \times n}$).
2. Each time the shortest path between vertices v_i and v_j is being updated by including node v_k (i.e. when $D^{(k-1)}[i, k] + D^{(k-1)}[k, j] < D^{(k-1)}[i, j]$) the (i, j) -th element of \mathbf{P} is set to k (i.e. $\mathbf{P}[i, j] = k$).
3. Therefore, $\mathbf{P}[i, j] = k$ indicates that v_k is the last vertex that has to be traversed along the shortest path connecting nodes v_i and v_j .

FLOYD-WARSHALL: PSEUDO CODE I

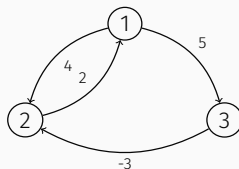
```
1 Floyd-Warshall(W):  
2 D = W;  
3 P = zeros(n,n);  
4 for k = 1:1:n  
5     for i = 1:1:n  
6         for j = 1:1:n  
7             if (D[i,j]>D[i,k]+D[k,j])  
8                 D[i,j] = D[i,k] + D[k,j];  
9                 P[i,j] = k;  
10            end  
11        end  
12    end  
13 end
```

FLOYD-WARSHALL: PSEUDO CODE II

```
1 Path(index q,r):
2 % Extract intermediate nodes within the shortest
3 % path from vertex index (q) to vertex index (r).
4 if(P[q,r] != 0)
5     Path(q,P[q,r]);
6     println("V"+P[q,r]);
7     Path(P[q,r],r);
8     return;
9 else
10    % No intermediate nodes
11    return;
12 end
```

FLOYD-WARSHALL EXAMPLE I

Apply Floyd-Warshall Algorithm on the following graph.



Initialize Distance and Predecessor matrices **D** and **P**.

$$D^{(0)} = \begin{bmatrix} 0 & 4 & 5 \\ 2 & 0 & \infty \\ \infty & -3 & 0 \end{bmatrix}$$

$$P = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

FLOYD-WARSHALL EXAMPLE II

- **Step 1:**

- Determine all possible pairs for which vertex $k = 1$ can act as an intermediate node: $\{(2, 3), (3, 2)\}$
- For the first pair $(2, 3)$, evaluate $D^{(1)}[2, 3]$ as:
 $D^{(1)}[2, 3] = \min\{D^{(0)}[2, 3], D^{(0)}[2, 1] + D^{(0)}[1, 3]\} = \min\{\infty, 2 + 5\} = 7$
- Assign $P[2, 3] = 1$
- For the second pair $(3, 2)$, evaluate $D^{(1)}[3, 2]$ as:
 $D^{(1)}[3, 2] = \min\{D^{(0)}[3, 2], D^{(0)}[3, 1] + D^{(0)}[1, 2]\} = \min\{-3, \infty + 4\} = -3$
- Thus, we have no change for the second pair $(3, 2)$

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

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

FLOYD-WARSHALL EXAMPLE III

- **Step 2:**

- Determine all possible pairs for which vertex $k = 2$ can act as an intermediate node: $\{(1, 3), (3, 1)\}$
- For the first pair $(1, 3)$, evaluate $D^{(2)}[1, 3]$ as:
 $D^{(2)}[1, 3] = \min\{D^{(1)}[1, 3], D^{(1)}[1, 2] + D^{(1)}[2, 3]\} = \min\{5, 4 + 7\} = 5$
- Thus, we have no change for the second pair $(1, 3)$
- For the second pair $(3, 1)$, evaluate $D^{(2)}[3, 1]$ as:
 $D^{(2)}[3, 1] = \min\{D^{(1)}[3, 1], D^{(1)}[3, 2] + D^{(1)}[2, 1]\} = \min\{\infty, -3 + 2\} = -1$
- Assign $P[3, 1] = 2$

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

$$P = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 2 & 0 & 0 \end{bmatrix}$$

FLOYD-WARSHALL EXAMPLE IV

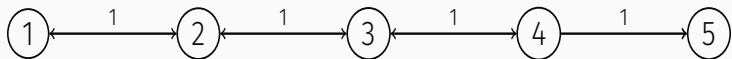
- **Step 3:**
 - Determine all possible pairs for which vertex $k = 3$ can act as an intermediate node: $\{(1, 2), (2, 1)\}$
 - For the first pair $(1, 2)$, evaluate $D^{(3)}[1, 2]$ as:
 $D^{(3)}[1, 2] = \min\{D^{(2)}[1, 2], D^{(2)}[1, 3] + D^{(2)}[3, 2]\} = \min\{4, 5 + (-3)\} = 2$
 - Assign $P[1, 2] = 3$
 - For the second pair $(2, 1)$, evaluate $D^{(3)}[2, 1]$ as:
 $D^{(3)}[2, 1] = \min\{D^{(2)}[2, 1], D^{(2)}[2, 3] + D^{(2)}[3, 1]\} = \min\{2, 7 + (-1)\} = 2$
 - Thus, we have no change for the second pair $(1, 2)$

$$D^{(3)} = \begin{bmatrix} 0 & 2 & 5 \\ 2 & 0 & 7 \\ -1 & -3 & 0 \end{bmatrix}$$

$$P = \begin{bmatrix} 0 & 3 & 0 \\ 0 & 0 & 1 \\ 2 & 0 & 0 \end{bmatrix}$$

RECURSIVE PATH RECONSTRUCTION EXAMPLE I

For the following linear graph:



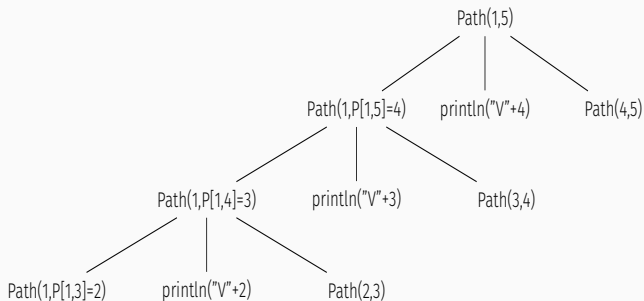
it is easy to deduce that:

$$D^{(5)} = \begin{bmatrix} 0 & 1 & 2 & 3 & 4 \\ 1 & 0 & 1 & 2 & 3 \\ 2 & 1 & 0 & 1 & 2 \\ 3 & 2 & 1 & 0 & 1 \\ 4 & 3 & 2 & 1 & 0 \end{bmatrix}$$

$$P = \begin{bmatrix} 0 & 0 & 2 & 3 & 4 \\ 0 & 0 & 0 & 3 & 4 \\ 2 & 0 & 0 & 0 & 4 \\ 2 & 3 & 0 & 0 & 0 \\ 2 & 3 & 4 & 0 & 0 \end{bmatrix}$$

RECURSIVE PATH RECONSTRUCTION EXAMPLE II

Determining the intermediate nodes on the shortest path between nodes 1 and 5 results in the following recursive calling sequence of the *Path()* function:



FLOYD-WARSHALL: ANALYSIS I

It is easy to deduce that the Runtime Complexity of the Floyd-Warshall algorithm is $\Theta(n^3)$

Its Space Complexity is $\Theta(n^2)$

However, it is not obvious why the Floyd-Warshall algorithm can be implemented through the utilization of a single $n \times n$ distance matrix \mathbf{D} .

- $\mathbf{D}[i, j]$ depends only on the elements in the k -th column and row.
- It is easy to show that the k -th row and the k -th column of the distance matrix remain unchanged when $\mathbf{D}^{(k)}$ is being computed.

FLOYD-WARSHALL: ANALYSIS II

Before showing that the k -th row and column of the distance matrix D remain unchanged, we will show that the elements along the main diagonal remain 0.

- $D^{(k)}[j, j] = \min\{D^{(k-1)}[j, j], D^{(k-1)}[j, k] + D^{(k-1)}[k, j]\} \Leftrightarrow$
 $D^{(k)}[j, j] = \min\{0, D^{(k-1)}[j, k] + D^{(k-1)}[k, j]\} \Leftrightarrow$
 $D^{(k)}[j, j] = 0$
- This is true since we have assumed that there may exist negative edges but not cycles with cumulative weight negative.

FLOYD-WARSHALL: ANALYSIS III

The k -th column of $\mathbf{D}^{(k)}$ is equal to the k -th column of $\mathbf{D}^{(k-1)}$.

Intuitively true since a path from v_i to v_k will not become shorter by adding v_k to the allowed subset of intermediate nodes.

$$\begin{aligned} \cdot \forall i, \mathbf{D}^{(k)}[i, k] &= \min\{D^{(k-1)}[i, k], D^{(k-1)}[i, k] + D^{(k-1)}[k, k]\} \Leftrightarrow \\ \mathbf{D}^{(k)}[i, k] &= \min\{D^{(k-1)}[i, k], D^{(k-1)}[i, k] + 0\} \Leftrightarrow \\ \mathbf{D}^{(k)}[i, k] &= D^{(k-1)}[i, k] \end{aligned}$$

The k -th row of $\mathbf{D}^{(k)}$ is equal to the k -th row of $\mathbf{D}^{(k-1)}$.

$$\begin{aligned} \cdot \forall j, \mathbf{D}^{(k)}[k, j] &= \min\{D^{(k-1)}[k, j], D^{(k-1)}[k, k] + D^{(k-1)}[k, j]\} \Leftrightarrow \\ \mathbf{D}^{(k)}[k, j] &= \min\{D^{(k-1)}[k, j], 0 + D^{(k-1)}[k, j]\} \Leftrightarrow \\ \mathbf{D}^{(k)}[k, j] &= D^{(k-1)}[k, j] \end{aligned}$$


```
1 function [D,P] = FloydWarshall(W)
2
3 % This function computes shortest paths' distances
4 % for each pair of nodes within the graph whose
5 % initial weight (adjacency) matrix is stored in
6 % matrix W. Matrix W is assumed to be properly
7 % initialized. Element D[i,j] of matrix D stores
8 % the shortest path distance from node i to node j.
9 % Matrix P is the corresponding predecessor matrix
10 % so that element P[i,j] stores the last vertex
11 % traversed within the shortest path connecting
12 % nodes i and j.
```

FLOYDWARSHALL.M II

```
13
14 % Get the number of nodes pertaining to the graph.
15 nodes_num = size(W,1);
16
17 % Initialize internal matrix D.
18 D = W;
19
20 % Initialize internal matrix P.
21 P = zeros(nodes_num,nodes_num);
22
23 % Main Algorithm.
24 for k = 1:1:nodes_num
25     for i = 1:1:nodes_num
```

FLOYDWARSHALL.M III

```
26         for j = 1:1:nodes_num
27             if( D(i , j) > D(i , k) + D(k , j))
28                 D(i , j) = D(i , k) + D(k , j);
29                 P(i , j) = k;
30             end
31         end
32     end
33 end
34
35 end
```

EXTRACTSHORTESTPATHS.M I

```
1 function [Dtop,Ptop] = ExtractShortestPaths(Wo,Ctop
   )
2
3 % This function extracts the pairwise shortest
4 % paths matrices and corresponding path
5 % reconstruction indices matrices for each one of
6 % the top No connected components of the
7 % co-authorship network.
8
9 % Wo: is the initial binary connectivity matrix.
10 % Ctop: is a cell array storing the indices of
11 % the top No connected components stored in
```

EXTRACTSHORTESTPATHS.M II

```
12 % decreasing order of magnitude.
13
14 % Get the number of top connected components
15 % stored in Ctop.
16 Ntop = length(Ctop);
17
18 % Initialize cell array containers for variables
19 % Dtop and Ptop.
20 % Each element of Dtop stores the matrix of
21 % pairwise shortest distances.
22 % Each element of Ptop stores the
23 % corresponding matrix of predecessor
24 % indices that can be utilized to reconstruct the
```

EXTRACTSHORTESTPATHS.M III

```
25 % sequence of nodes in each shortest path.
26
27 % Loop through the various connected components
28 % stored in Ctop.
29 for component_index = 1:1:Ntop
30     % Get the current component.
31     component = Ctop{component_index};
32     % Get the number of nodes pertaining to the
33     % current component.
34     Nc = length(component);
35     % Get the corresponding sub-weight matrix for
36     % the current component.
37     Wc = Wo(component, component);
```

EXTRACTSHORTESTPATHS.M IV

```
38 % Set the diagonal indices for the current
39 % sub-weight matrix.
40 Idiag = [1:Nc+1:Nc*Nc];
41 % Get the indices of all zero elements of the
42 % current sub-weight matrix.
43 Izero = find(Wc==0);
44 % Get the indices of all zero elements in the
45 % current sub-weight matrix
46 % that do not reside on its main diagonal.
47 Izero_non_diagonal = setdiff(Izero,Idiag);
48 % Set all non-diagonal zero entries of the
49 % current sub-weight matrix to Inf.
50 Wc(Izero_non_diagonal) = Inf;
```

```
51 % Run the Floyd–Warshall algorithm of the
52 % current connected component.
53 [Dc,Pc] = FloydWarshall(Wc);
54
55 % The following line of code should be
56 % uncommented if:
57 % the predecessor indices stored in Pc should
58 % be synchronized with the original author
59 % indices in Wo excluding the zero values of Pc
60
61 % Pc(Pc~=0) = component(Pc(Pc~=0));
```



```
62     % Set the corresponding entries of Dtop and Ptop
        .
63     Dtop{component_index} = Dc;
64     Ptop{component_index} = Pc;
65 end
66
67 end
```

RECONSTRUCTPATH.M I

```
1 function [Path] = ReconstructPath(P,source_node ,  
   target_node)  
2  
3 % This function reconstructs the path between the  
4 % given pair of (source_node,target_node)  
5 % based on the predecessor matrix P.  
6 % The predecessor matrix is assumed to be  
7 % associated with a connected component of an  
8 % underlying graph.  
9  
10 % Initially , construct the intermediate path  
11 % between the source_node and the target_node.
```

RECONSTRUCTPATH.M II

```
12 % (Mind that an intermediate path may not exist in
13 % case the source and target nodes are
14 % immediately connected).
15
16 % Initialize the intermediate path.
17 intermediate_path = [];
18
19 % Get the last intermediate node between the source
20 % and target nodes.
21 intermediate_node = P(source_node, target_node);
22
23 % While the intermediate node index is not (0)
24 % retrieve the full set of intermediate
```

RECONSTRUCTPATH.M III

```
25 % nodes in reverse order.
26 while(intermediate_node~=0)
27     intermediate_path = [intermediate_path,
28         intermediate_node];
28     intermediate_node = P(source_node,
29         intermediate_node);
29 end
30
31 % If the intermediate path is not empty reverse it.
32 if(~isempty(intermediate_path))
33     intermediate_path = intermediate_path(end:-1:1)
34     ;
34 end
```

RECONSTRUCTPATH.M IV

```
35  
36 % Construct the full path.  
37 Path = [source_node , intermediate_path , target_node ];  
38  
39 end
```

```
1 % Isolate the top No connected components.
2 Ctop = C(TopNComponentsIndices);
3
4 % Extract the pair-wise shortest paths and
5 % predecessor indices for each connected component.
6 [Dtop,Ptop] = ExtractShortestPaths(Wo,Ctop);
7
8 % Example: Report the shortest distance and
9 % corresponding path between nodes (1),(50) and
10 % (1),(97) in component 1.
11 component = Ctop{1};
12 P = Ptop{1};
```

```
13 D = Dtop{1};  
14 source_node = 1  
15 target_node = 50  
16 sortest_distance = D(source_node, target_node)  
17 Path = ReconstructPath(P, source_node, target_node)  
18 source_node = 1  
19 target_node = 97  
20 sortest_distance = D(source_node, target_node)  
21 Path = ReconstructPath(P, source_node, target_node)
```