

## Problem Set V

ECO 7427 - Econometric Theory II

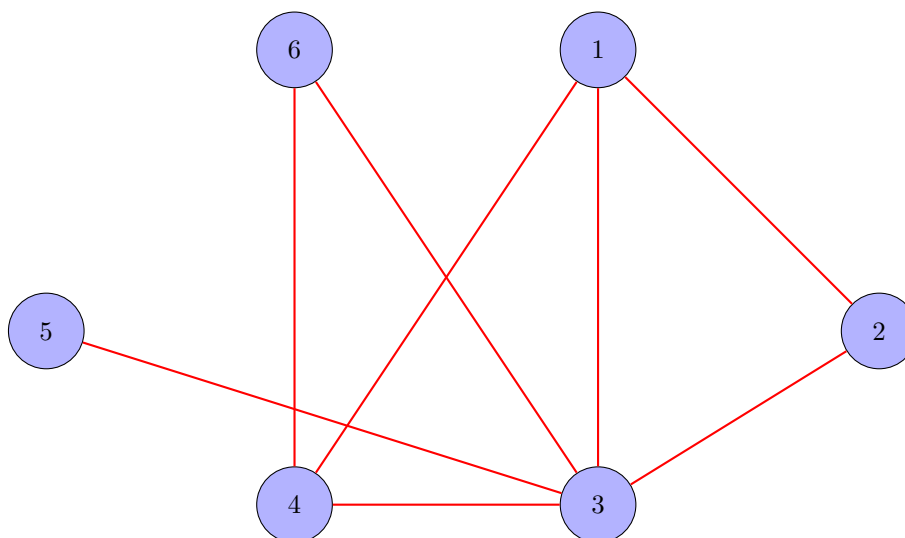
*Armand Kapplani*

Spring 2018

### Social Networks

Let  $N = \{1, 2, 3, 4, 5, 6\}$  and consider the following network  $(N, g)$ .

Figure 1: Undirected and Unweighted Network



**Construct the adjacency matrix:**

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

To compute the degree of each vertex we sum across each row

$$d_i(g) = \sum_j g_{ij} \quad (1)$$

**The neighborhood of each vertex:**

Table 1: Neighborhood of each vertex

Vertex	$N_i(g)$
N = 1	$\{2, 3, 4\}$
N = 2	$\{1, 3\}$
N = 3	$\{1, 2, 4, 5, 6\}$
N = 4	$\{1, 3, 6\}$
N = 5	$\{3\}$
N = 6	$\{3, 4\}$

**The degree of each node:**

Table 2: Degree of each vertex in the network

Vertex	Degree
N = 1	3
N = 2	2
N = 3	5
N = 4	3
N = 5	1
N = 6	2

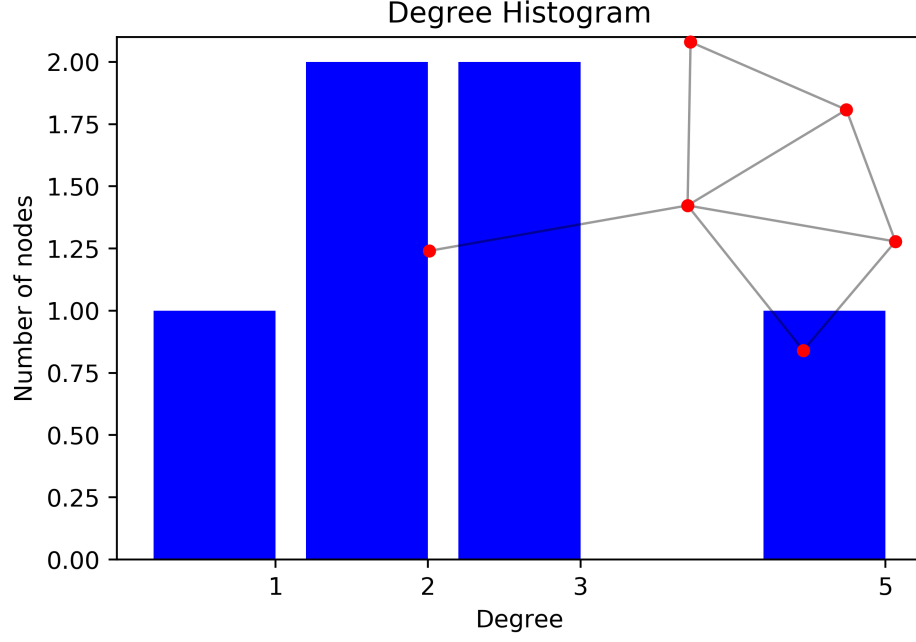
**Average degree of the network  $g \in G(N)$ :**

$$\frac{1}{N} \sum_{i \in N} d_i(g) = \frac{1}{6} [3 + 2 + 5 + 3 + 1 + 2] = \frac{8}{3} \quad (2)$$

**The density of the network  $g \in G(N)$ :**

$$\frac{\sum_{i \in N} d_i(g)}{n(n-1)} = \frac{\frac{1}{n} \sum_{i \in N} d_i(g)}{n-1} = \frac{8/3}{5} = \frac{8}{15} \quad (3)$$

Figure 2: Degree Histogram



The individual clustering of the six vertices:

$$CI_i(g) = \frac{\left| \{jk \in g \mid k \neq j, j \in N_i(g), k \in N_i(g)\} \right|}{d_i(g)[d_i(g) - 1]} \quad (4)$$

Table 3: Individual clustering of each vertex

Vertex	$CI_i(g)$
N = 1	2/3
N = 2	1
N = 3	3/10
N = 4	2/3
N = 5	0
N = 6	1

The average clustering coefficient:

$$CI^{Avg}(g) = \frac{1}{n} \sum_{i \in N} CI_i(g) = \frac{1}{6} \left[ \frac{2}{3} + 1 + \frac{3}{10} + \frac{2}{3} + 0 + 1 \right] = \frac{109}{180} \quad (5)$$

**Degree centrality for each vertex:**

$$CI_i(g) = \frac{\left\{jk \in g | k \neq j, j \in N_i(g), k \in N_i(g)\right\}}{d_i(g)[d_i(g) - 1]} \quad (6)$$

Table 4: Degree centrality for each vertex

Vertex	$CI_i(g)$
N = 1	3/5
N = 2	2/5
N = 3	1
N = 4	3/5
N = 5	1/5
N = 6	2/5

Hence from Table 4 we can see that vertex 3 is well connected in terms of direct connections as it has the highest degree of centrality. However using this measure alone we cannot tell if vertex 3 is well located in the network  $(N, g)$ .

**Closeness centrality fo each vertex:**

$$C_i(g) = \frac{1}{\frac{1}{n-1} \sum_{j \neq i} l(i, j)} \quad (7)$$

where  $l(i, j)$  denotes the number of links in the **shortest path** between  $i$  and  $j$ .

Table 5: Closeness centrality fo each vertex

Vertex	$C_i(g)$
N = 1	5/7
N = 2	5/8
N = 3	1
N = 4	5/7
N = 5	5/9
N = 6	5/8

**Decay centrality for each vertex:**

$$D_i^\delta(g) = \sum_{j \neq i} \delta^{l(i, j)} \quad (8)$$

where  $l(i, j)$  denotes the number of links in the **shortest path** between  $i$  and  $j$ .

Table 6: Decay centrality fo each vertex

Vertex	$D_i^\delta(g)$	$\delta = 1/2$
N = 1	$\delta(3 + 2\delta)$	2
N = 2	$\delta(2 + 3\delta)$	7/4
N = 3	$5\delta$	5/2
N = 4	$\delta(3 + 2\delta)$	2
N = 5	$\delta(1 + 4\delta)$	3/2
N = 6	$\delta(1 + 4\delta)$	3/2

**Betweenness centrality for each vertex:**

$$Ce_i^B(g) = \sum_{k \neq j: i \notin \{k, j\}} \frac{P_i(kj)/P(kj)}{\frac{1}{2}(n-1)(n-2)} \quad (9)$$

Table 7: Betweenness centrality of a vertex

Vertex	$P_i(k, j)$	$Ce_i^B(g)$
N = 1	{23, 24, 25, 26, 34, 35, 36, 45, 46, 56}	1/20
N = 2	{13, 14, 15, 16, 34, 35, 36, 45, 46, 56}	0
N = 3	{12, 14, 15, 16, 24, 25, 26, 45, 46, 56}	3/5
N = 4	{12, 13, 15, 16, 23, 25, 26, 35, 36, 56}	1/20
N = 5	{12, 13, 14, 16, 23, 24, 26, 34, 36, 46}	0
N = 6	{12, 13, 14, 15, 23, 24, 25, 34, 35, 45}	0

Notice that for the first vertex ( $N = 1$ ) the link 24 shown in red in the set  $P_i(k, j)$  can either go through vertex 1 or through vertex 3 since both are geodesic.

**Eigenvector centrality for each node:**

We solve the following problem numerically to compute the eigencentality of each vertex in the graph.

$$\lambda C^e(g) = g C^e(g) \quad (10)$$

where  $C^e(g)$  is left-hand eigenvector of the adjacency matrix  $g$  and  $\lambda$  is the corresponding eigenvalue.

where  $|\lambda_i| \leq \lambda_1 = 3.04$  ( $\forall i \neq 1$ ) is the **Perron-Frobenius eigenvalue** of the adjacency matrix  $g$  also known as the **spectral radius**<sup>1</sup>  $\rho(g)$ . Eigencentality is a measure of the influence of a node in a network.

<sup>1</sup>Theorem: Let  $g \in C^{n \times n}$  with spectral radius  $\rho(g)$ , then  $\rho_A < 1$  if and only if  $\lim_{g \rightarrow \infty} g^k = 0$  and if  $\rho(g) > 1$  then  $\lim_{k \rightarrow \infty} \|g^k\| = \infty$

Table 8: Largest eigenvalue and its eigenvector

Vertices	$\lambda_{max} = \rho(g) = \mathbf{3.04}$
N = 1	0.45
N = 2	0.34
N = 3	<b>0.58</b>
N = 4	0.45
N = 5	0.19
N = 6	0.34

### Katz-Bonacich Centrality:

$$C^B(g, a, b) = (I_n - bg)^{-1} ag \mathbf{1}_n \quad (11)$$

Table 9: Number of walks emanating from each vertex

Vertices	k = 1	...	k = $\infty$
N = 1	3	...	$\infty$
N = 2	2	...	$\infty$
N = 3	5	...	$\infty$
N = 4	3	...	$\infty$
N = 5	1	...	$\infty$
N = 6	2	...	$\infty$

where we would expect that the number of walks goes to infinity as  $\lim_{k \rightarrow \infty} \|A^k\| = \infty$

We compute Katz-Bonacich centrality using an algorithmic approach in Python

$$x_i = \beta \sum_j A_{ij} x_j + \alpha \quad (12)$$

where the parameter  $\alpha$  controls the initial centrality and  $\beta < \frac{1}{\lambda_{max}}$ .

Table 10: Katz-Bonacich centrality

Vertices	$\alpha = 1, \beta = 0.5$	$\alpha = 1, \beta = 1/3$
N = 1	0.47	0.45
N = 2	0.31	0.34
<b>N = 3</b>	<b>0.60</b>	<b>0.58</b>
N = 4	0.47	0.45
N = 5	0.08	0.19
N = 6	0.31	0.34

## Additional scatter plots

Figure 3: Scatter plot

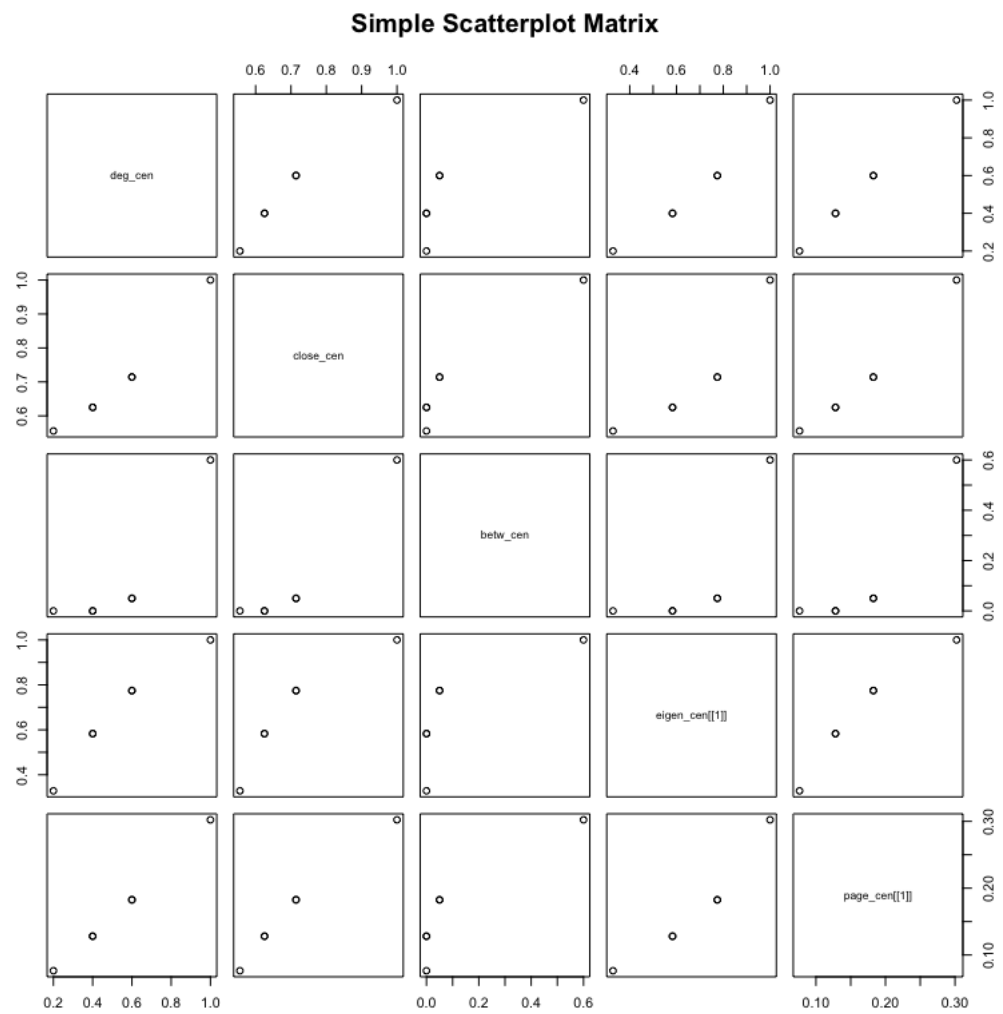


Figure 4: Scatter plot correlations

