

# Social and Economic Networks

## WSU Python Working Group

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# Measures of Connectivity

## Betweenness Centrality:

The betweenness centrality of a node  $v$  is the sum of the fraction of all-pairs shortest paths that pass through  $v$ :

$$c_B(v) = \sum_{s,t \in V} \frac{\sigma(s,t|v)}{\sigma(s,t)}$$

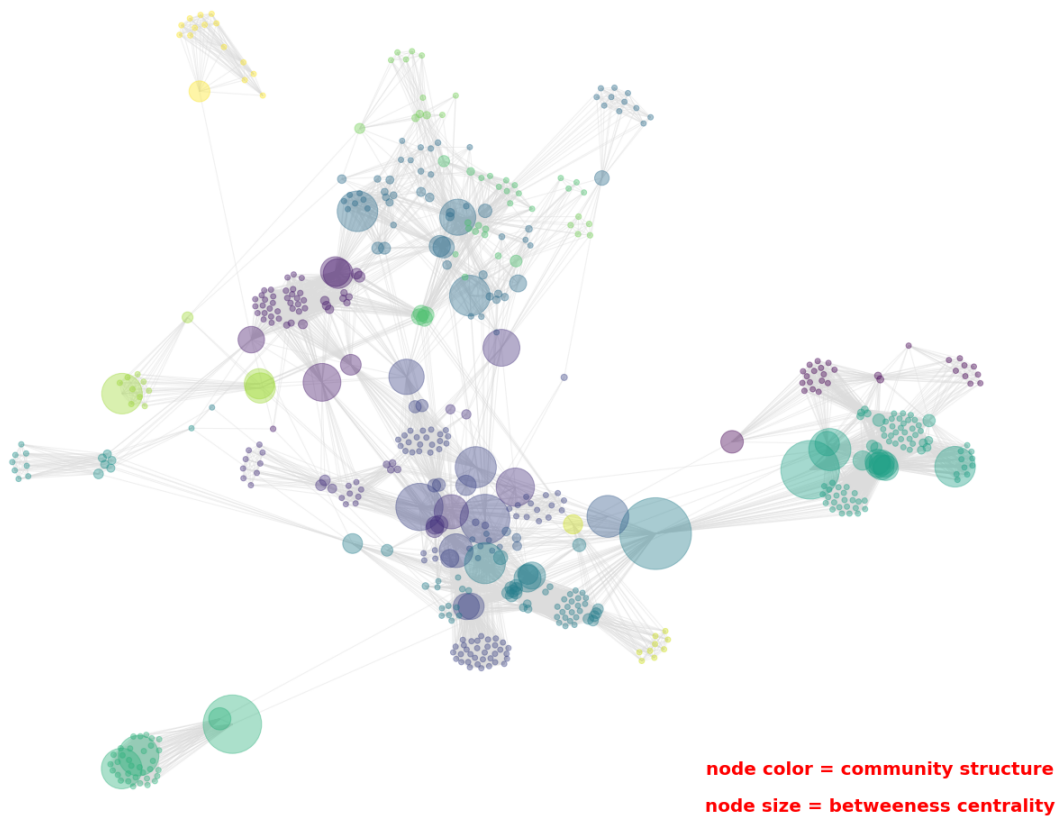
where  $V$  is the set of nodes,  $\sigma(s,t)$  is the number of shortest  $(s,t)$ -paths, and  $\sigma(s,t|v)$  is the number of those paths passing through some node  $v$  other than  $s,t$ . If  $s = t$ ,  $\sigma(s,t) = 1$ , and if  $v \in \{s,t\}$ ,  $\sigma(s,t|v) = 0$

<https://networkx.org/documentation/stable/>

# Measures of Connectivity

## Betweenness Centrality of Positive Gene Functional Associations

Gene functional association network (C. elegans)



# Measures of Connectivity

## In-degree Centrality:

The in-degree centrality  $x_i$  of node  $i$  is given by:  $x_i = \sum_k a_{k,i}$  or in matrix form ( $\mathbf{1}$  is a vector with all components equal to unity):  $x = \mathbf{1}A$

The out-degree centrality  $y_i$  of node  $i$  is given by:  $y_i = \sum_k a_{i,k}$  or in matrix form:  $y = A\mathbf{1}$ .

<https://networkx.org/documentation/stable/>

# Measures of Connectivity

## Closeness Centrality:

Closeness centrality of a node  $u$  is the reciprocal of the average shortest path distance to  $u$  over all  $n - 1$  reachable nodes

$$C(u) = \frac{n - 1}{\sum_{v=1}^{n-1} d(v, u)}$$

where  $d(v, u)$  is the shortest-path distance between  $v$  and  $u$ , and  $n$  is the number of nodes that can reach  $u$ . Notice that the closeness distance function computes the incoming distance to  $u$  for directed graphs.

<https://networkx.org/documentation/stable/>

# Preliminaries for Social and Economic Networks

## Matrix of Beliefs $T_{n \times n}$

- Let there be a finite set of  $N = \{1, \dots, n\}$  *agents* or *nodes* that interact in a social network.
- The interactions between agents are defined by an  $n \times n$  nonnegative row stochastic matrix  $\mathbf{T}$ , where each element of the  $\mathbf{T}$  matrix  $T_{ik}$  represents the amount of attention agent  $i$  gives to agent  $k$ .
- The  $\mathbf{T}$  matrix is called an interaction matrix, and  $T_{ik}$  can be seen as the weight or trust agent  $i$  puts on the current belief of agent  $k$  in constructing their belief for next period.

<https://web.stanford.edu/~jacksonm/books.html>

# Preliminaries for Social and Economic Networks

## The DeGroot (1974) Model

Each agent has belief  $p_i^{(t)} \in \mathbb{R}$ , where for simplicity it is assumed to be between  $[0, 1]$ , and the vector of all agent's beliefs is represented by an  $n \times 1$  array  $\mathbf{p}^{(t)}$ . The belief updating rule is

$$\mathbf{p}^{(t)} = \mathbf{T}\mathbf{p}^{(t-1)}$$

which implies

$$\mathbf{p}^{(t)} = \mathbf{T}^t \mathbf{p}^{(0)} \tag{1}$$

where,

$$\mathbf{T}_{n \times n} = \begin{bmatrix} T_{11} & T_{12} & \dots & T_{1n} \\ T_{21} & T_{22} & \dots & T_{2n} \\ \vdots & \vdots & \dots & \vdots \\ T_{n1} & T_{n2} & \dots & T_{nn} \end{bmatrix} \quad \text{and} \quad \mathbf{p}_{n \times 1}^{(t)} = \begin{bmatrix} p_1^{(t)} \\ p_2^{(t)} \\ \vdots \\ p_n^{(t)} \end{bmatrix}.$$

<https://web.stanford.edu/~jacksonm/books.html>

# Preliminaries for Social and Economic Networks

## Consensus

A consensus is reached in the DeGroot model if and only if there is exactly one strongly connected and closed group of agents,  $\mathbf{T}$  is aperiodic on that group, and this convergence happens for any initial belief vector  $\mathbf{p}^{(0)}$ .

## Social Influence

Assuming that a consensus can be reached, the social influence of agent  $i$  in a society is the  $i^{th}$  entry in the converged vector of beliefs if and only if that belief vector sums to 1. More formally, and skipping a lot of theory, we seek a limiting vector  $s \in [0, 1]^n$  such that

$$s\mathbf{T} = s$$

where,  $\sum_{i=1}^n s_i = 1$ , and  $s$  is the left-hand eigenvector of  $\mathbf{T}$ .

<https://web.stanford.edu/~jacksonm/books.html>



# Finding the Influence Vector

## An Example using The Left-hand Eigenvector Theory

Let,

$$\mathbf{T}_{3 \times 3} = \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

this implies the  $s$  that solves  $s\mathbf{T} = s$  is

$$\begin{bmatrix} 2/5 & 2/5 & 1/5 \end{bmatrix} \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 2/5 & 2/5 & 1/5 \end{bmatrix}.$$

<https://web.stanford.edu/~jacksonm/books.html>

# Another Way of Finding the Influence Vector

## An Example Using the Power Iteration

Consider the same example from above where,

$$\mathbf{T}_{3 \times 3} = \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad \text{and} \quad \mathbf{p}_{3 \times 1}^{(0)} = \begin{bmatrix} p_1^{(0)} \\ p_2^{(0)} \\ p_3^{(0)} \end{bmatrix}.$$

We know that any matrix  $\mathbf{T}$  should converge to a consensus  $\mathbf{p}_{n \times 1}^{(\infty)}$  for any initial belief vector  $\mathbf{p}_{n \times 1}^{(0)}$  such that

$$\mathbf{p}_{n \times 1}^{(\infty)} = \mathbf{T}^{\infty} \mathbf{p}_{n \times 1}^{(0)}$$

implying that if we multiply  $\mathbf{T}$  to itself enough times, we should find a convergent matrix. This is apparent in this example since

$$\mathbf{T}_{3 \times 3}^{\infty} = \begin{bmatrix} 2/5 & 2/5 & 1/5 \\ 2/5 & 2/5 & 1/5 \\ 2/5 & 2/5 & 1/5 \end{bmatrix} \quad \text{where} \quad s = \begin{bmatrix} 2/5 & 2/5 & 1/5 \end{bmatrix}.$$

<https://web.stanford.edu/~jacksonm/books.html>

# References

- Aric A. Hagberg, Daniel A. Schult and Pieter J. Swart, “Exploring network structure, dynamics, and function using NetworkX”, in Proceedings of the 7th Python in Science Conference (SciPy2008), Gäel Varoquaux, Travis Vaught, and Jarrod Millman (Eds), (Pasadena, CA USA), pp. 11–15, Aug 2008
- Ara Cho, Junha Shin, Sohyun Hwang, Chanyung Kim, Hong Suk Shim, Hyojin Kim, Hanhae Kim, Insuk Lee, WormNet v3: a network-assisted hypothesis-generating server for *Caenorhabditis elegans*. Nucl. Acids Res. published online May 9, 2014
- DeGroot, M.H. (1974). “Reaching a Consensus” *Journal of the American Statistical Association* 69:118-121
- Golub and Jackson (2010), “Naïve Learning in Social Networks and the Wisdom of Crowds” *American Economic Journal: Microeconomics* 2:1, 112–149
- Jackson M. O. (2011). *Social and Economic Networks*. Princeton University Press.

**Note:** If any citations need to be added, please feel free to e-mail.