node *centrality*

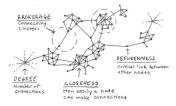
introduction to network science in Python (NetPy)

Lovro Šubelj University of Ljubljana 14th Dec 2021

centrality *measures*

which *nodes* are most *important*?

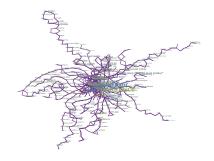
- node centrality measures for (un)directed networks
 - clustering coefficients [WS98, SV05, dNMB05]
 - distance-based centrality [Fre77, FBW91, New05]
 - spectral analysis centrality [Kat53, Bon87, BP98]



— link analysis algorithms for directed networks

networkology LPP

- partial LPP public bus transport network*
- n = 416 bus stops with $\langle k \rangle = 5.62$ connections
- giant component 95.4% nodes (6 components)
- "small-world" with $\langle C \rangle = 0.09$ and $\langle d \rangle = 14.26$
- "scale-free" with $\gamma = 2.62$ for cutoff $k_{min} = 5$



^{*} reduced to largest connected component

centrality *clustering*

important *nodes* are *strongly embedded*

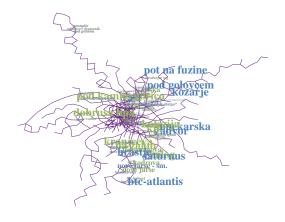
- for undirected G clustering coefficient C [WS98] of i is
 - $-t_i$ is number of *linked neighbors* or *triangles* of i

$$C_i = \frac{2t_i}{k_i(k_i-1)}$$
 $C_i = 0$ for $k_i \le 1$

— C fails for hub nodes in scale-free networks [dNMB05]

networkology clustering

- clustering coefficient C in partial LPP network[†]
- highest $C_i = 1.0$ nodes are Na Žalah etc. with $k_i = 2$



reduced to simple undirected graph

centrality *closeness*

important *nodes* are *close to other* nodes

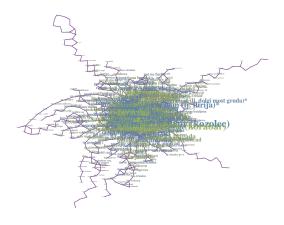
- for (un)directed G closeness centrality ℓ^{-1} [New10] of i is
 - d_{ij} is (un)directed distance between i and j
 - $-d_{ij} = \infty$ for nodes in different components

$$\ell_i^{-1} = \frac{1}{n-1} \sum_{j \neq i} \frac{1}{d_{ij}}$$

— ℓ^{-1} spans *small range* in *small-world* networks

networkology *closeness*

- closeness centrality ℓ^{-1} in partial LPP network[‡]
- highest $\ell_i^{-1} = 0.208$ node is Gosposvetska with $k_i = 14$



[‡]reduced to simple undirected graph

centrality betweenness

important *nodes* are *bridges btw other* nodes

- for (un)directed G betweenness centrality σ [Fre77] of i is
 - g_{st} is number of shortest paths between s and t
 - $-g_{st}^{i}$ is number of such shortest paths through i

$$\sigma_i = \frac{1}{n^2} \sum_{st} \frac{g_{st}^i}{g_{st}}$$

— σ considers *only shortest paths* [FBW91, New05]

networkology betweenness

- betweenness centrality σ in partial LPP network \S
- highest $\sigma_i = 0.235$ node is Razstavišče with $k_i = 11$



[§] reduced to simple undirected graph

centrality degrees

important nodes are linked by many nodes

- for undirected G degree centrality d of i is $d_i = \frac{1}{n-1} \sum_{j \neq i} A_{ij} = \frac{k_i}{n-1}$
- in directed G in-degree centrality d^{in} of i is

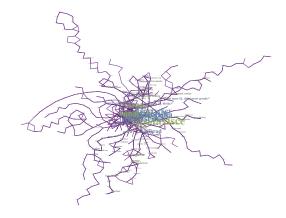
$$d_i^{in} = \frac{1}{n-1} \sum_{j \neq i} A_{ij} = \frac{k_i^{in}}{n-1}$$

— in directed G out-degree centrality d^{out} of i is

$$d_i^{out} = \frac{1}{n-1} \sum_{j \neq i} A_{ji} = \frac{k_i^{out}}{n-1}$$

networkology *degrees*

- degree centrality d in partial LPP network
- highest $d_i = 0.099$ node is Razstavišče with $k_i = 41$
- highest d_i node is Razstavišče with $k_i^{in} = 20$ and $k_i^{out} = 21$



centrality eigenvector

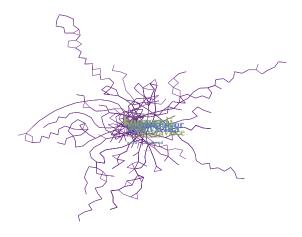
important nodes are linked by important nodes

- for (un)directed G eigenvector centrality e [Bon87] of i is

 e is leading eigenvector v_1 of A with eigenvalue λ_1^{-1} $e_i = \lambda_1^{-1} \sum_j A_{ij} e_j$
- in directed G = 0 for $k^{in} = 0$ nodes etc.

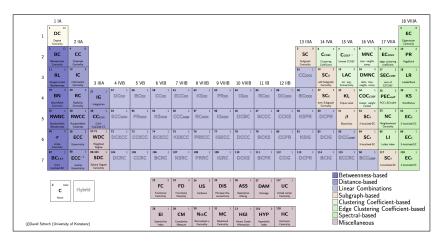
networkology eigenvector

- eigenvector centrality e in partial LPP network
- highest $e_i = 0.082$ node is Konzorcij with $k_i = 30$



centrality overview

which *nodes* are most *important*?



centrality references



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