### node *centrality*

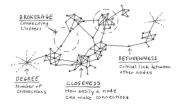
introduction to network science in Python (NetPy)

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### centrality *measures*

### which *nodes* are most *important*?

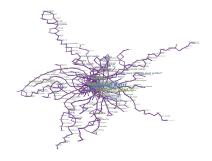
- node centrality measures for (un)directed networks
  - clustering coefficients [WS98, SV05, dNMB05]
  - geodesic-based measures [Fre77, FBW91, New05]
  - spectral analysis measures [Kat53, Bon87, BP98]



— link analysis algorithms primarily 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  $k_{min} = 5$



<sup>\*</sup> reduced to largest connected component

## centrality clustering

#### important nodes are strongly embedded

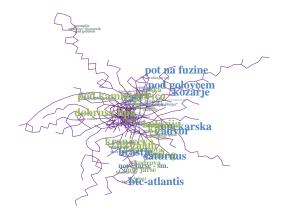
- for undirected G clustering coefficient C [WS98] of i is
  - $t_i$  is number of *linked neighbors* or *triangles* of i

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

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

# networkology *clustering*

- clustering coefficient C in partial LPP network<sup>†</sup>
- highest  $C_i = 1.0$  nodes are Na Žalah etc. with  $k_i = 2$



<sup>†</sup>reduced to simple undirected graph

## centrality *closeness*

#### important *nodes* are *close to other* nodes

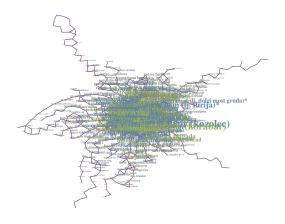
- for (un) directed G closeness centrality  $\ell^{-1}$  [New10] of i is
  - $d_{ij}$  is (un) directed distance between i and j
  - $d_{ij} = \infty$  for nodes in different components

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

—  $\ell^{-1}$  spans *small range* in *small-world* networks

## networkology *closeness*

- closeness centrality  $\ell^{-1}$  in partial LPP network<sup>‡</sup>
- highest  $\ell_i^{-1} = 0.208$  node is Gosposvetska with  $k_i = 14$



<sup>&</sup>lt;sup>‡</sup>reduced to simple undirected graph

### centrality betweenness

#### important *nodes* are *bridges between other* nodes

- for (un)directed G betweenness centrality  $\sigma$  [Fre77] of i is
  - g<sub>st</sub> is number of shortest paths between s and t
  - $-g_{st}^{i}$  is number of such shortest paths through i

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

—  $\sigma$  considers *only shortest paths* [FBW91, New05]

# networkology betweenness

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



<sup>§</sup> 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

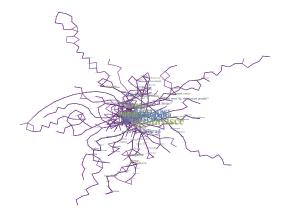
$$d_i^{in} = \frac{1}{n-1} \sum_{j \neq i} A_{ij} = \frac{k_i^{in}}{n-1}$$

— in directed G out-degree centrality dout 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  $k_i^{out} = 21$



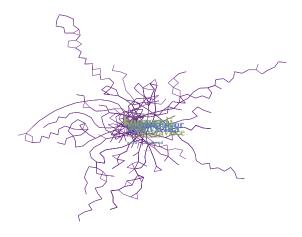
## centrality eigenvector

#### 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  $\lambda_1^{-1}$   $e_i = \lambda_1^{-1} \sum_j A_{ij} e_j$
- 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 Katz

#### nodes get small amount of importance for free

- for (un) directed G Katz centrality z [Kat53] of i is
  - $\alpha$  and  $\beta$  are appropriate positive constants

$$\mathbf{z}_i = \alpha \sum_j A_{ij} \mathbf{z}_j + \beta$$

- for *convenience*  $\beta = 1$  whereas  $\alpha < \lambda_1^{-1}$ 
  - $-\lambda_1$  is leading eigenvalue of A for eigenvector  $v_1$

# centrality PageRank

#### nodes distribute equal amount of importance

— for (un)directed G PageRank centrality p [BP98] of i is  $-\alpha$  and  $\beta$  are appropriate positive constants

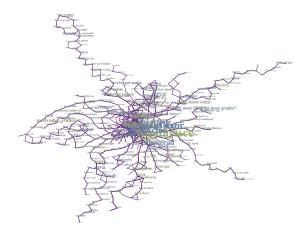
$$p_i = \alpha \sum_j A_{ij} \frac{p_j}{k_j} + \beta$$

- for *convenience*  $\beta = \frac{1-\alpha}{n}$  whereas  $\alpha = 0.85$
- p probability of random surfer with teleports

see PageRank algorithm NetLogo demo

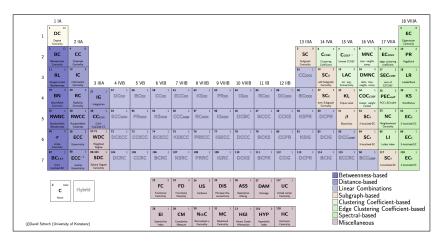
# networkology PageRank

- PageRank centrality p in partial LPP network
- highest  $p_i = 0.011$  node is Razstavišče with  $k_i = 41$



### centrality overview

#### which *nodes* are most *important*?



### centrality references



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