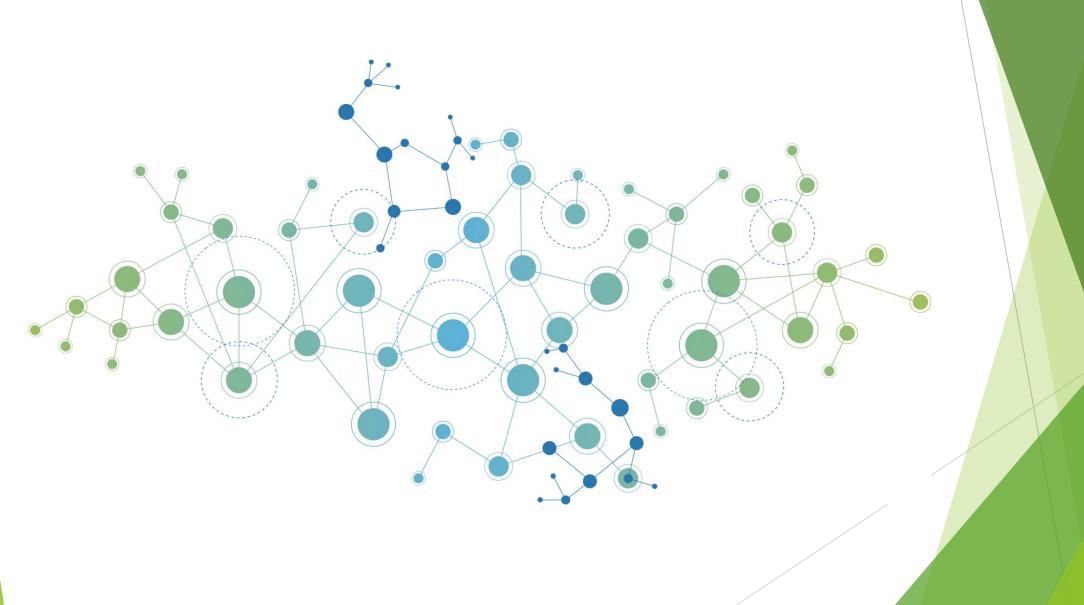
Spectral Centrality Measures in Complex Networks

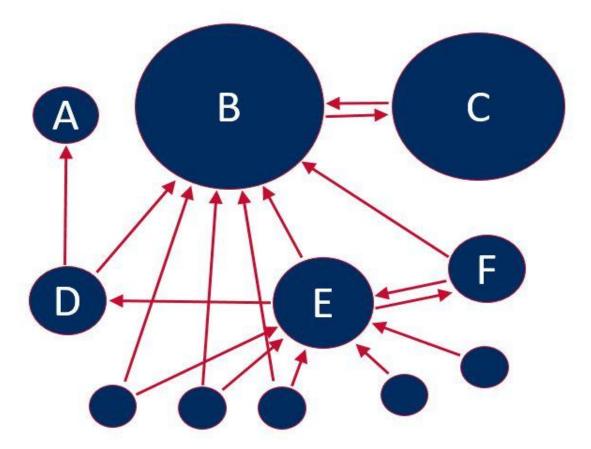
Review of the paper released by Nicola Perra and Santo Fortunato

Ilija Doknić 53m/22 Stefan Komarica 236m/22 Andrija Velimirović 45m/22

Introduction



PageRank



$$p(i) = \frac{q}{n} + (1 - q) \sum_{j:j \to i} \frac{p(j)}{k_{out}(j)}, \quad i = 1, 2, ..., n$$

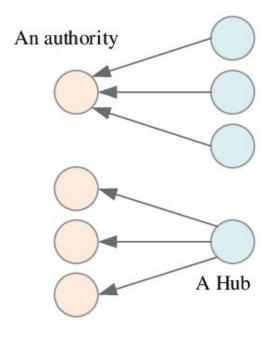
Eigenvector Centrality

$$\lambda x_i = \sum_{j:j\to i} x_j = \sum_j A_{ji} x_j = (\mathbf{A}^t \mathbf{x})_i$$

Limitations

$$x_i = \alpha(\mathbf{A}^t \mathbf{x})_i + \epsilon$$

HITS Scores



$$\lambda y_i = \sum_{j:j\to i} x_j = \sum_j A_{ji} x_j = (\mathbf{A}^t \mathbf{x})_i$$

$$\mu x_i = \sum_{j:i\to j} y_j = \sum_j A_{ij} y_j = (\mathbf{A}\mathbf{y})_i$$

$$\lambda \mu x_i = (\mathbf{A} \mathbf{A}^t \mathbf{x})_i$$

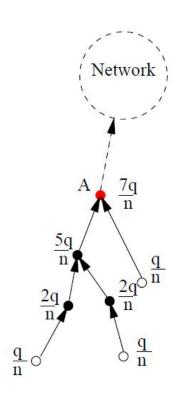
$$\lambda \mu y_i = (\mathbf{A}^t \mathbf{A} \mathbf{y})_i$$

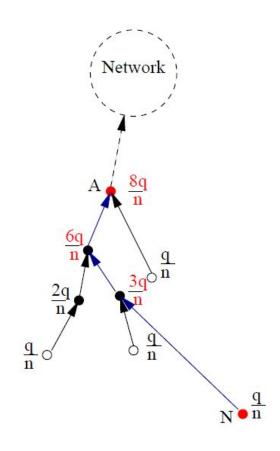
Results for PageRank

- ► Two main limits, $q \rightarrow 0$ and $q \rightarrow 1$
- Focus od DMS graph
- Expand to undirected graphs

$$\Pi(k_j, a) = \frac{a + k_j}{\sum_{l=1}^{i-1} (a + k_l)}$$

Attaching new node





$$p(i) \sim \frac{q}{n} + \sum_{j:j \to i} p(j)$$
 $i = 1, 2, \dots, n$

 $q \rightarrow 0$

- All PR values are multiples of the elementary unit q/n. When we say PR is increased by 1, that means it is increased by q/n
- PR increases if one moves from a node to another node following a link
- PR at each node is equal to the number of predecessors of that node

q -> 1

- Trivial since all nodes end up having PR value 1/n
- Defining the reduced PageRank

$$p_r(i) = p(i) - \frac{q}{n}$$
 $i = 1, 2, \dots, n$

$$p_r(i) = \frac{q(1-q)}{mn} k_{in}(i), \qquad i = 1, 2, \dots, n$$

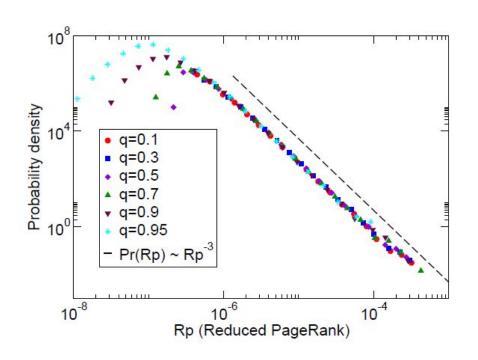
Extension to undirected graphs

Primary idea: extension to random walk, undirected links can be crossed in both directions

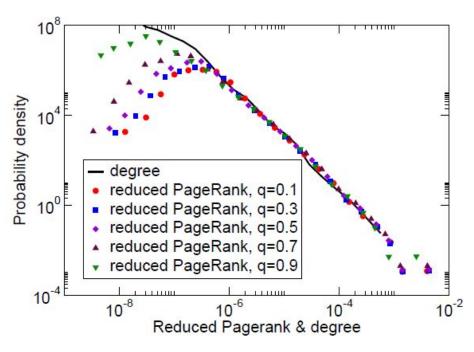
$$p(i) = \frac{q}{n} + (1 - q) \sum_{j:j \leftrightarrow i} \frac{p(j)}{k_j}.$$
 $i = 1, 2, ..., n$

 Contribution of random jumping is present and this mixed process is still hard to solve

Reduced PageRank on undirected graphs



Results on DMS graph



Results on .gov domain

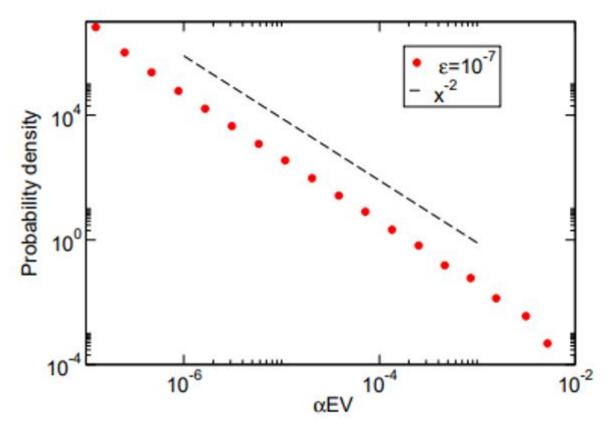
Eigenvector centrality. Directed Graphs

Similarity between Eigenvector centrality and Page Rank

$$x_i = \alpha(\mathbf{A}^t \mathbf{x})_i + \epsilon.$$

$$p(i) \sim \frac{q}{n} + \sum_{j:j \to i} p(j)$$
 $i = 1, 2, \dots, n$

Eigenvector centrality. Directed Graphs



Distribution on DMS graph

Eigenvector centrality. Undirected graphs

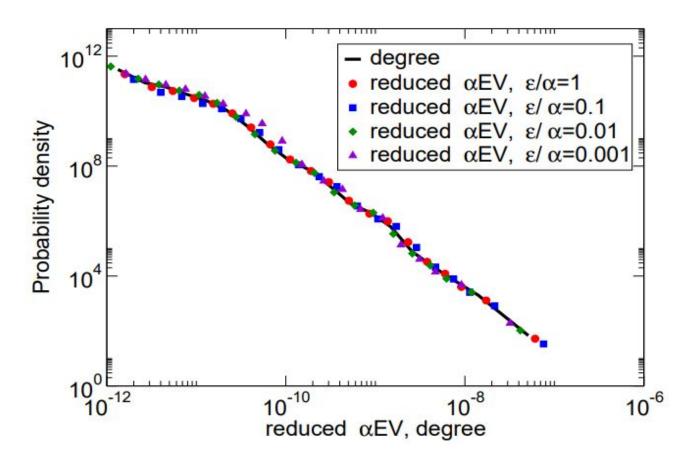
Formula for reduced Eigenvector centrality on Undirected Graphs

$$x_i^r = \alpha(\mathbf{A}\mathbf{x}^r)_i + k_i \alpha \epsilon$$

Approximated sum of neighbors alpha centralities

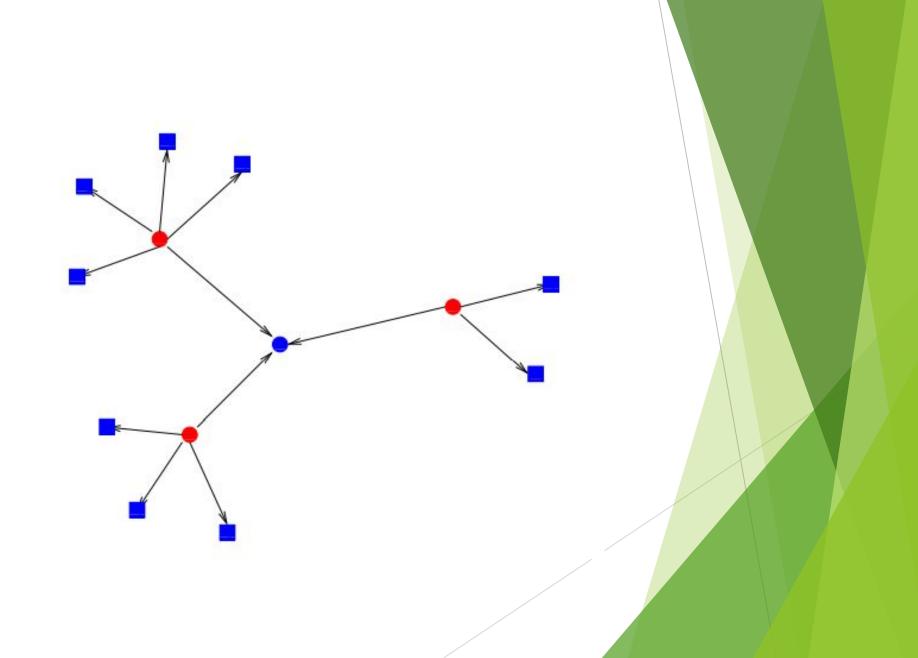
$$k_i\langle x^r\rangle$$

Eigenvector centrality. Undirected graphs



Distribution on .gov domain

HITS scores



HITS scores

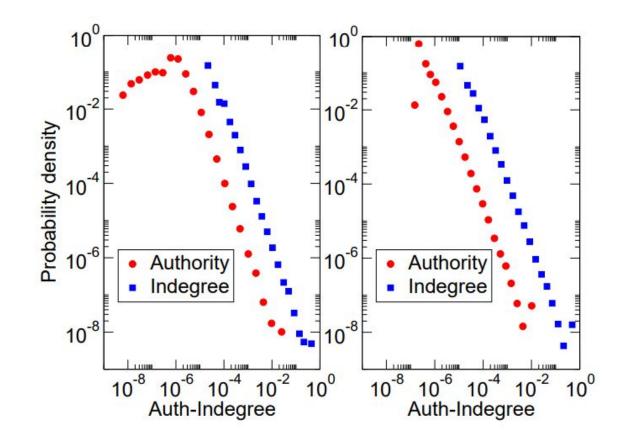
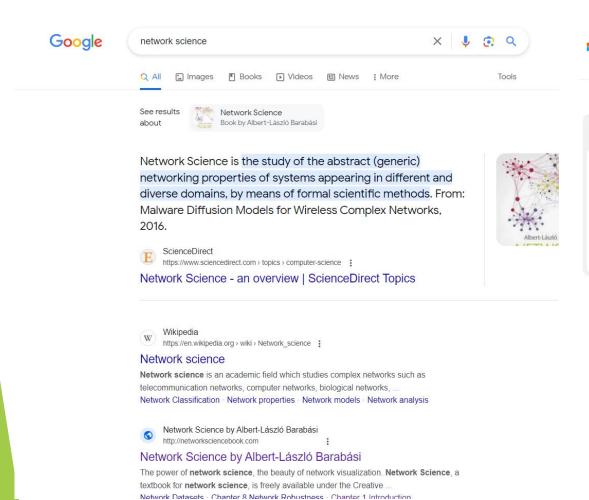
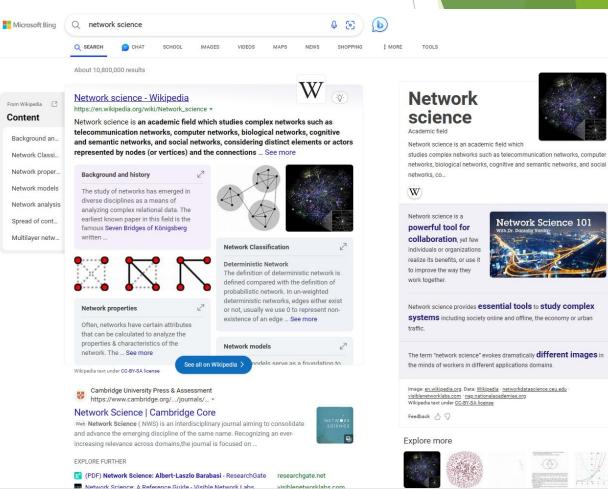


FIG. 10: Distribution of the authority scores versus indegree distribution. (Left) DMS graph with 10^5 nodes, m = 10 and a = 1. (Right) DMS graph with 10^5 nodes, m = 50 and a = 1.

Rankings





Rankings

Measures	au
$PR-\alpha EV$	0.8192
PR-AUTH	0.5774
PR-HUBS	0.1213
PR-IN	0.6444
PR-OUT	-0.3012
α EV-AUTH	0.5788
$lpha ext{EV-IN}$	0.6487
α EV-HUBS	0.1220
$\alpha \text{EV-OUT}$	0.5788
AUTH-IN	0.5458
AUTH-HUBS	0.1076
AUTH-OUT	-0.2611
HUBS-IN	0.1142
HUBS-OUT	-0.2126
IN-OUT	-0.2507

TABLE I: Kendall's τ for each pair of centrality measures computed for a DMS directed graph, with $n=10^6,\ m=3$ and a=3.

Rankings

3.5	
Measures	au
$PR-\alpha EV$	0.09
PR-AUTH	0.14
PR-HUBS	0.04
PR-IN	0.14
PR-OUT	0.02
α EV-AUTH	0.12
α EV-IN	0.07
α EV-HUBS	0.08
$\alpha \text{EV-OUT}$	0.01
AUTH-IN	0.12
AUTH-HUBS	0.07
AUTH-OUT	0.01
HUBS-IN	0.02
HUBS-OUT	0.07
IN-OUT	0.07

TABLE II: Kendall's τ for each pairs of centrality measures for the network of political blogs studied by Adamic and Glance.

Conclusion

Thank you for your attention

Questions?