

# Link Analysis

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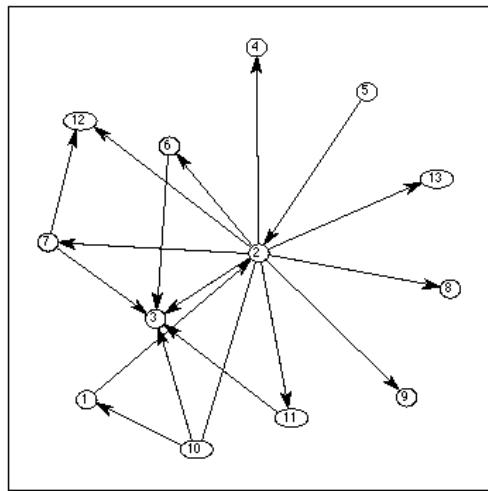
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НАЦИОНАЛЬНЫЙ ИССЛЕДОВАТЕЛЬСКИЙ  
УНИВЕРСИТЕТ

# Link Analysis

Graph  $G(n, m)$  and adjacency matrix  $A_{ij}$ , edge  $i \rightarrow j$



	1	2	3	4	5	6	7	8	9	10	11	12	13
1	*												
2		*	*										
3													
4													
5					*								
6						*							
7							*						
8								*					
9									*				
10										*	*	*	
11											*		
12													*
13													

nz = 19

# Status or Rank Prestige

Leo Katz, 1953.

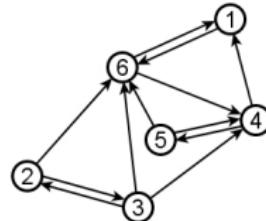
Take into account status (prestige) of directly connected actors

$$p_i \leftarrow \sum_{j \in N(i)} p_j = \sum_j A_{ji} p_j$$

$$p_i = \sum_j A_{ji} p_j$$

$$\mathbf{p} = \mathbf{A}^T \mathbf{p}$$

$$(\mathbf{I} - \mathbf{A}^T) \mathbf{p} = 0$$



Nontrivial solution only if  $\det(\mathbf{I} - \mathbf{A}^T) = 0$ . Need to constraint matrix

# Status or Rank Prestige

Leo Katz, 1953

An actor gives equal parts of its prestige to all nearest neighbours

$$p_i \leftarrow \sum_j A_{ji} \frac{p_j}{k_{out}(j)} = \sum_j \frac{A_{ji}}{k_{out}(j)} p_j$$

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

where  $\mathbf{D}_{ii} = \max(k_i, 1)$

$\mathbf{D}^{-1} \mathbf{A}$  - stochastic matrix,  $\sum_j (\mathbf{D}^{-1} \mathbf{A})_{ij} = 1$ , guaranteed  $\lambda_{max} = 1$

$$(\mathbf{D}^{-1} \mathbf{A})^T \mathbf{p} = \lambda \mathbf{p}$$

# Eigenvector Centrality

Phillip Bonacich, 1972.

$$c_i \leftarrow \sum_j A_{ji} c_j$$

$$c_i = \frac{1}{\kappa} \sum_j A_{ji} c_j$$

$$\kappa \mathbf{c} = \mathbf{A}^T \mathbf{c}$$

$$(\kappa \mathbf{I} - \mathbf{A}^T) \mathbf{p} = 0$$

Nontrivial solution only when  $\det(\kappa \mathbf{I} - \mathbf{A}^T) = 0$ . Eigenvalue problem.

Choose eigenvector corresponding do maximum eigenvalue:

$$\kappa_{max} = \kappa_1, \mathbf{c} = \mathbf{c}_1$$

# Status or Rank Prestige. Eigenvector Centrality

Phillip Bonacich, 1987.

Parametrized centrality measure  $c(\alpha, \beta)$

$$c_i = \sum_j (\alpha + \beta c_j) A_{ji}$$

$$\mathbf{c} = \alpha \mathbf{A}^T \mathbf{e} + \beta \mathbf{A}^T \mathbf{c}$$

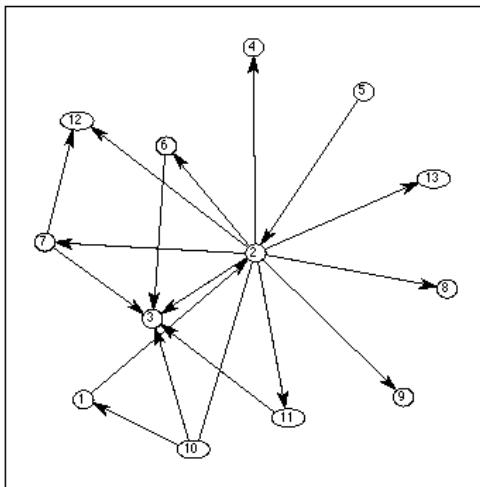
$$\mathbf{c} = \alpha (\mathbf{I} - \beta \mathbf{A}^T)^{-1} \mathbf{A}^T \mathbf{e}$$

$\alpha$  - found from normalization  $\|\mathbf{c}\|_2 = \sum c_i^2 = 1$

$\beta$  - parameter, degree and direction of dependence on others

# Link Analysis

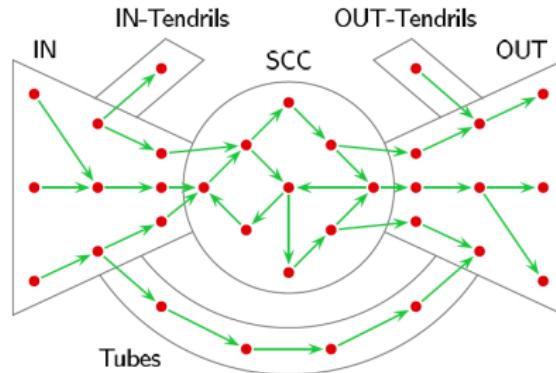
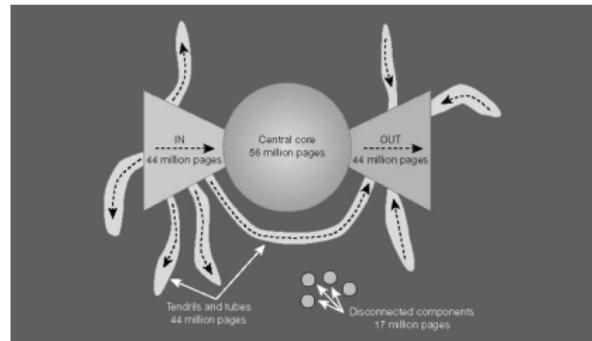
Graph  $G(n, m)$



- zero out degree nodes,  $k_{out}(i) = 0$
- zero in degree nodes,  $k_{in}(i) = 0$

# Real World

## Bow tie structure of the web



# Perron-Frobenius Theorem

Oscar Perron, 1907, Georg Frobenius, 1912.

Eigenvalue problem:

$$\mathbf{P}\mathbf{p} = \lambda\mathbf{p}$$

Perron-Frobenius theorem: Real square matrix with positive entries

- stochastic (non-negative and rows sum up to one)
- irreducible (strongly connected graph)
- aperiodic

then unique largest eigenvalue  $\lambda_{max} = 1$ , with positive left eigenvector and power iterations converges to it. Solution satisfies  $|\mathbf{p}|_1 = 1$

Stationary distribution of Markov chain

# PageRank

Sergey Brin and Larry Page, 1998

Transition matrix:

$$\mathbf{P} = \mathbf{D}^{-1}\mathbf{A}$$

Stochastic matrix:

$$\mathbf{P}' = \mathbf{P} + \frac{\mathbf{d}\mathbf{e}^T}{n}$$

PageRank matrix:

$$\mathbf{P}'' = \alpha\mathbf{P}' + (1 - \alpha)\frac{\mathbf{e}\mathbf{e}^T}{n}$$

Eigenvalue problem (choose solution with  $\lambda = 1$ ):

$$\mathbf{P}''^T \mathbf{p} = \lambda \mathbf{p}$$

Notations:

**e** - unit column vector, **d** - absorbing nodes indicator vector (column)

# PageRank computations

- Eigenvalue problem

$$\left[ \alpha \left( (\mathbf{D}^{-1} \mathbf{A})^T + \frac{\mathbf{e} \mathbf{d}^T}{n} \right) + (1 - \alpha) \frac{\mathbf{e} \mathbf{e}^T}{n} \right] \mathbf{p} = \lambda \mathbf{p}$$

- Power iterations

$$\mathbf{p} \leftarrow \alpha (\mathbf{D}^{-1} \mathbf{A})^T \mathbf{p} + \alpha \frac{\mathbf{e}}{n} (\mathbf{d}^T \mathbf{p}) + (1 - \alpha) \frac{\mathbf{e}}{n} (\mathbf{e}^T \mathbf{p})$$

$$\mathbf{p} \leftarrow \frac{\mathbf{p}}{\|\mathbf{p}\|}$$

- Sparse linear system ( $\lambda = 1$ ,  $\|\mathbf{p}\|_1 = 1$ )

$$\left[ \mathbf{I} - \alpha \left( (\mathbf{D}^{-1} \mathbf{A})^T + \frac{\mathbf{e} \mathbf{d}^T}{n} \right) \right] \mathbf{p} = (1 - \alpha) \frac{\mathbf{e}}{n}$$

# Hubs and Authorities

HITS, Jon Kleinberg, 1999

Citation networks. Reviews vs original research (authoritative) papers

- authorities, contain useful information,  $a_i$ ;
- hubs, contains links to authorities,  $h_i$ ;

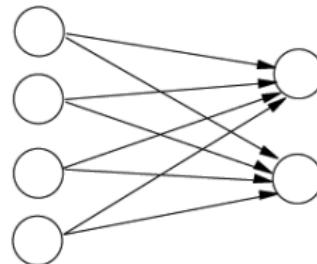
Mutual recursion

- Good authorities referred by good hubs

$$a_i \leftarrow \sum_j A_{ji} h_j$$

- Good hubs point to good authorities

$$h_i \leftarrow \sum_j A_{ij} a_j$$



## System of linear equations

$$\begin{aligned}\mathbf{a} &= \alpha \mathbf{A}^T \mathbf{h} \\ \mathbf{h} &= \beta \mathbf{A} \mathbf{a}\end{aligned}$$

## Symmetric eigenvalue problem

$$\begin{aligned}(\mathbf{A}^T \mathbf{A})\mathbf{a} &= \lambda \mathbf{a} \\ (\mathbf{A} \mathbf{A}^T)\mathbf{h} &= \lambda \mathbf{h}\end{aligned}$$

where eigenvalue  $\lambda = (\alpha\beta)^{-1}$

## References

- A new status index derived from sociometric analysis, L. Katz, *Psychometrika*, 19, 39-43, 1953.
- Power and centrality: A family of measures, P. Bonacich, *American Journal of Sociology*, 92, 1170-1182, 1987.
- Graph structure in the Web, Andrei Broder et al., *Procs of the 9th international World Wide Web conference on Computer networks*, 2000.
- The PageRank Citation Ranking: Bringing Order to the Web, S. Brin, L. Page, R. Motwani, T. Winograd, *Stanford Digital Library Technologies Project*, 1998
- Authoritative Sources in a Hyperlinked Environment, Jon M. Kleinberg, *Proc. 9th ACM-SIAM Symposium on Discrete Algorithms*,