

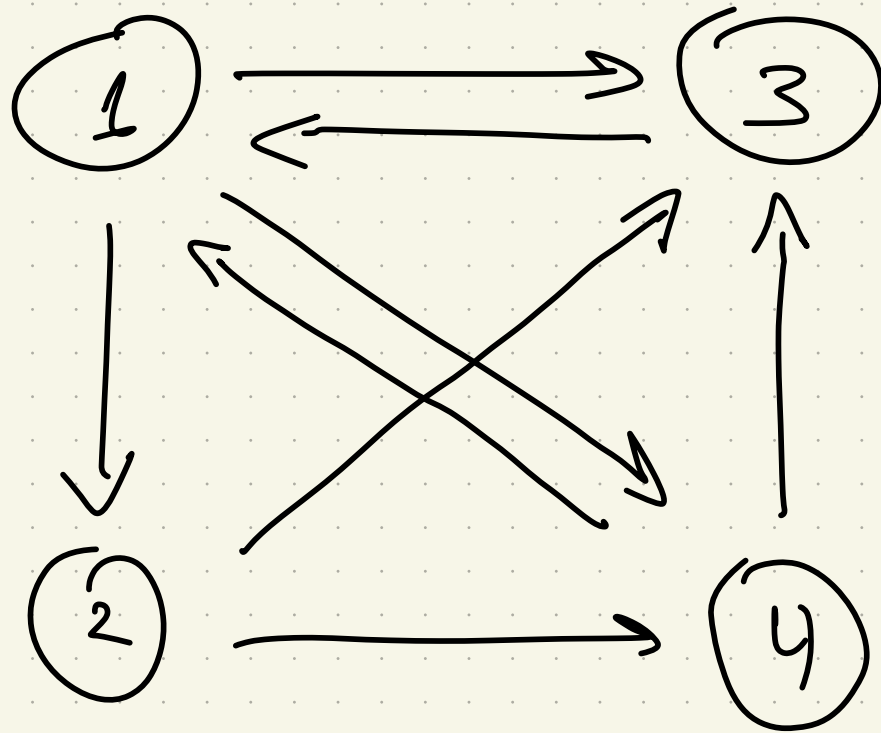
CS Lens

Page Rank

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(see more in "Mathematics of Data Science"
book draft on my page)



$x_k \in \mathbb{R}$ ← "importance of page k ".

Idea #1: $x_k = \#$ back links to k .

Some backlinks matter more than others.

Idea #2:

$$X_k = \sum_{j \in E_k} x_j.$$

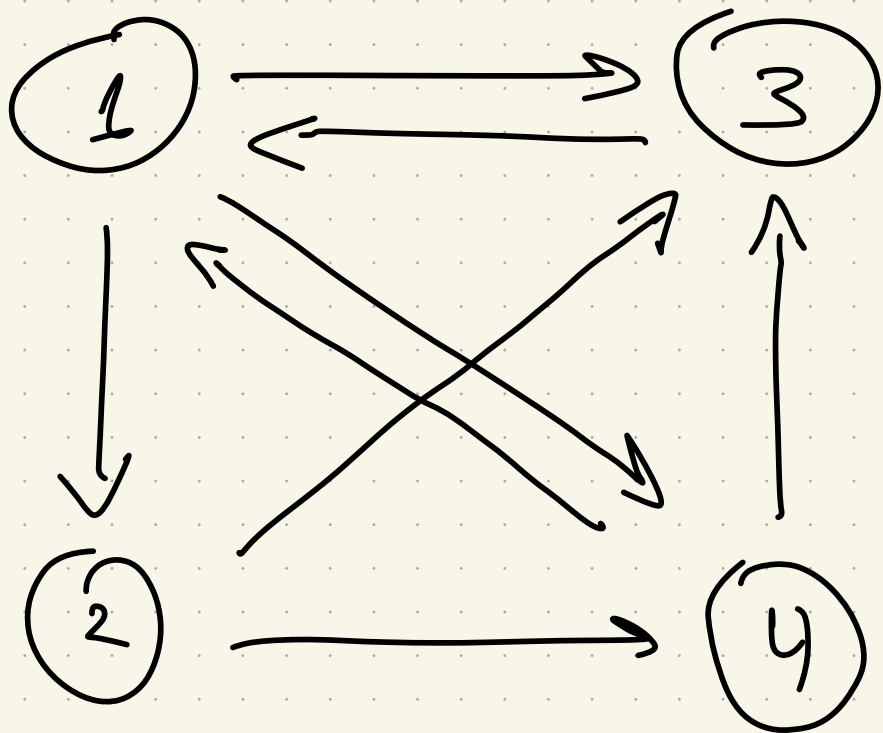
E_k set of
pages that
link to k

Idea #3:

normalize
by # of out-links

$$X_k = \sum_{j \in E_k} \frac{x_j}{n_j}$$

$n_j = \#$ of outlink in page (j) .



$$x_1 = 1x_3 + \frac{1}{2}x_4$$

$$x_2 = \frac{1}{3}x_1$$

$$x_3 = \frac{1}{3}x_1 + \frac{1}{2}x_2 + \frac{1}{2}x_4$$

$$x_4 = \frac{1}{3}x_1 + \frac{1}{2}x_2$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & \frac{1}{2} \\ \frac{1}{3} & 0 & 0 & 0 \\ \frac{1}{3} & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & \frac{1}{2} & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

A

$$x = Ax$$

x is an eigenvector of A corresponding to eigenvalue 1.

Prop: if λ is an e-value of M
then it is also an e-value of
 M^T .

$$A^T = \begin{bmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & \lambda_2^{-1} & 0 \\ 0 & 0 & 0 & \lambda_3^{-1} \end{bmatrix}$$

$$A^T \begin{matrix} \mathbf{1} \\ \uparrow \\ \begin{bmatrix} 1 \\ \vdots \end{bmatrix} \end{matrix} = \mathbf{1}.$$

see: Perron - Frobenius.