Activity: PageRank and networks



Eigen decomposition

$$oldsymbol{B}oldsymbol{e}_i=\lambda_ioldsymbol{e}_i$$

eigenvector eigenvalue

 $oldsymbol{B}$ (square) symmetric matrix:

metric matrix: diagonal $oldsymbol{B} = oldsymbol{E} oldsymbol{\Lambda} oldsymbol{E}^T$ orthonormal rows, cols

Connection between eigenvecs and SVD

$$oldsymbol{A}oldsymbol{A}^T = oldsymbol{U}oldsymbol{\Sigma}oldsymbol{V}^Toldsymbol{V}oldsymbol{\Sigma}^Toldsymbol{U}^T = oldsymbol{U}oldsymbol{\Sigma}^2oldsymbol{U}^T$$

 \longrightarrow Eigenvectors of AA^T are left singular vectors of A

$$\boldsymbol{A}^T \! \boldsymbol{A} = \boldsymbol{V} \boldsymbol{\Sigma}^T \boldsymbol{U}^T \boldsymbol{U} \boldsymbol{\Sigma} \boldsymbol{V}^T = \boldsymbol{V} \boldsymbol{\Sigma}^2 \boldsymbol{V}^T$$

 \longrightarrow Eigenvectors of $\mathbf{A}^T \mathbf{A}$ are right singular vectors of \mathbf{A}

Eigenvalues: $\lambda_i = \sigma_i^2$

Power iteration (main idea)

$$(oldsymbol{A}oldsymbol{A}^T)^k = oldsymbol{U}\Sigma^2oldsymbol{U}^Toldsymbol{U}\Sigma^2oldsymbol{U}^T\dotsoldsymbol{U}\Sigma^2oldsymbol{U}^T \ = oldsymbol{U}\Sigma^{2k}oldsymbol{U}^T \ oldsymbol{oldsymbol{U}} oldsymbol{U} oldsymbol{$$

Adjacency matrix and PageRank

- Graph: nodes with edges between them
- Adjacency matrix: non-zero entry $ilde{m{A}}_{ij}$ if edge from j to i
- Transition probability matrix: normalize columns of $ilde{A}$ to 1

$$\tilde{A} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{bmatrix} \rightarrow Q = \begin{bmatrix} 0 & 1/3 & 0 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 0 \\ 0 & 1/3 & 0 & 1/2 & 0 \\ 0 & 0 & 1/2 & 0 & 1 \\ 1/2 & 1/3 & 1/2 & 1/2 & 0 \end{bmatrix}$$
Graph Adjacency matrix Transition probability matrix (Markov Chain)

 $Q^k b \to \text{direction of first eigenvector of } Q.$

The first eigenvector is the steady-state probability distribution