



Activity: PageRank and networks

Eigen decomposition

$$B e_i = \lambda_i e_i$$

eigenvector eigenvalue

B (square) symmetric matrix:

$$B = E \Lambda E^T$$

diagonal
orthonormal rows, cols

Connection between eigenvcs and SVD

$$A A^T = U \Sigma V^T V \Sigma^T U^T = U \Sigma^2 U^T$$

→ Eigenvectors of $A A^T$ are left singular vectors of A

$$A^T A = V \Sigma^T U^T U \Sigma V^T = V \Sigma^2 V^T$$

→ Eigenvectors of $A^T A$ are right singular vectors of A

Eigenvalues: $\lambda_i = \sigma_i^2$

Power iteration (main idea)

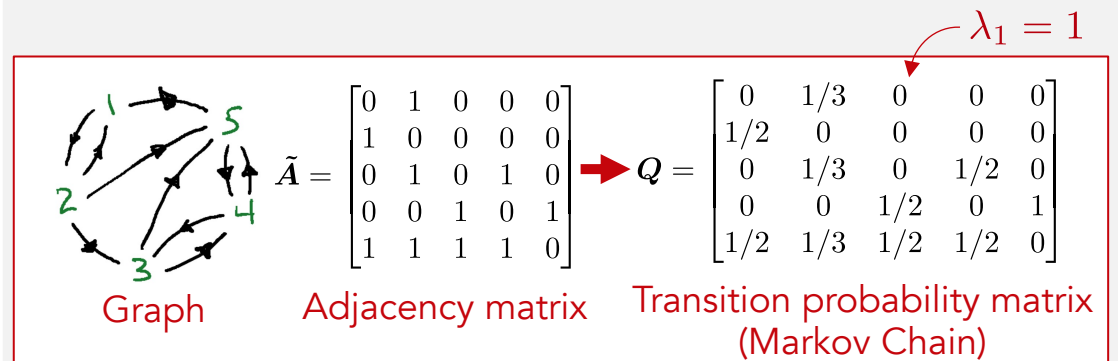
$$\begin{aligned} (A A^T)^k &= U \Sigma^2 U^T U \Sigma^2 U^T \dots U \Sigma^2 U^T \\ &= U \Sigma^{2k} U^T \\ \lim_{k \rightarrow \infty} \Sigma^{2k} &= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 0 & & \vdots \\ \vdots & & \ddots & \\ 0 & \dots & & 0 \end{bmatrix} \end{aligned}$$

→ imagine that $\sigma_1 = 1$, others less than 1

$$= U \Sigma^{2k} U^T \rightarrow u_1 u_1^T$$

Adjacency matrix and PageRank

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



$Q^k b \rightarrow$ direction of first eigenvector of Q .

The first eigenvector is the steady-state probability distribution