

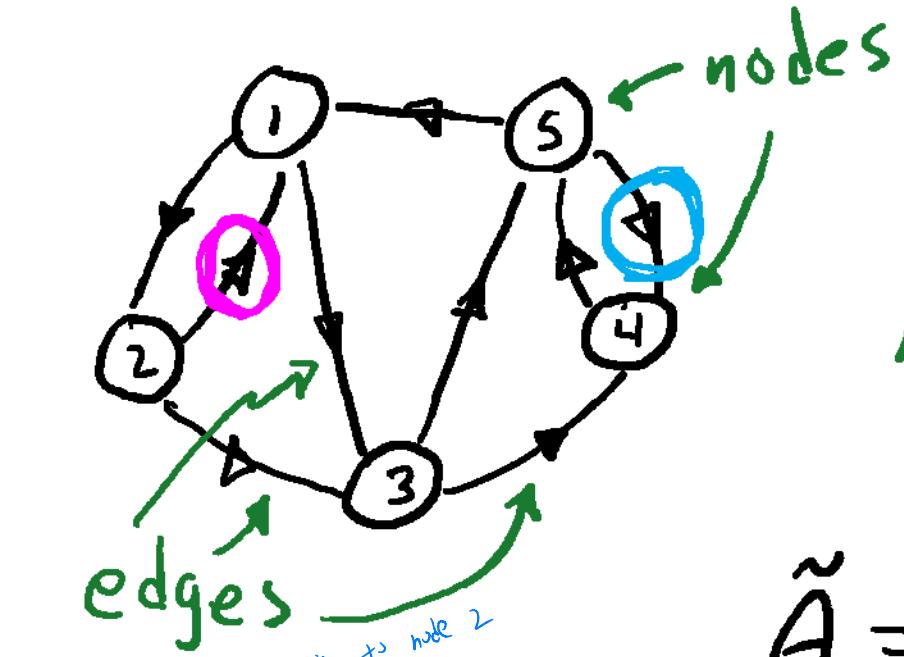
Network Graphs and the PageRank Algorithm

Objectives

- Introduce matrix representations for network graphs
- Define transition probability matrix and paths on graph
- Illustrate PageRank algorithm concepts

Matrices represent network graphs

2



Examples: webpages / links,
cities / roads, routers / wires

Adjacency matrix; connection topology

edges

at time t , it will transition to node 2

$$\tilde{A} = \begin{bmatrix} 0 & 1/2 & 0 & 0 & 1/2 \\ 1/4 & 0 & 0 & 0 & 0 \\ 3/4 & 1/2 & 0 & 0 & 0 \\ 0 & 0 & 1/8 & 0 & 1/2 \\ 0 & 0 & 1/8 & 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 & 0 \end{bmatrix} +_6$$

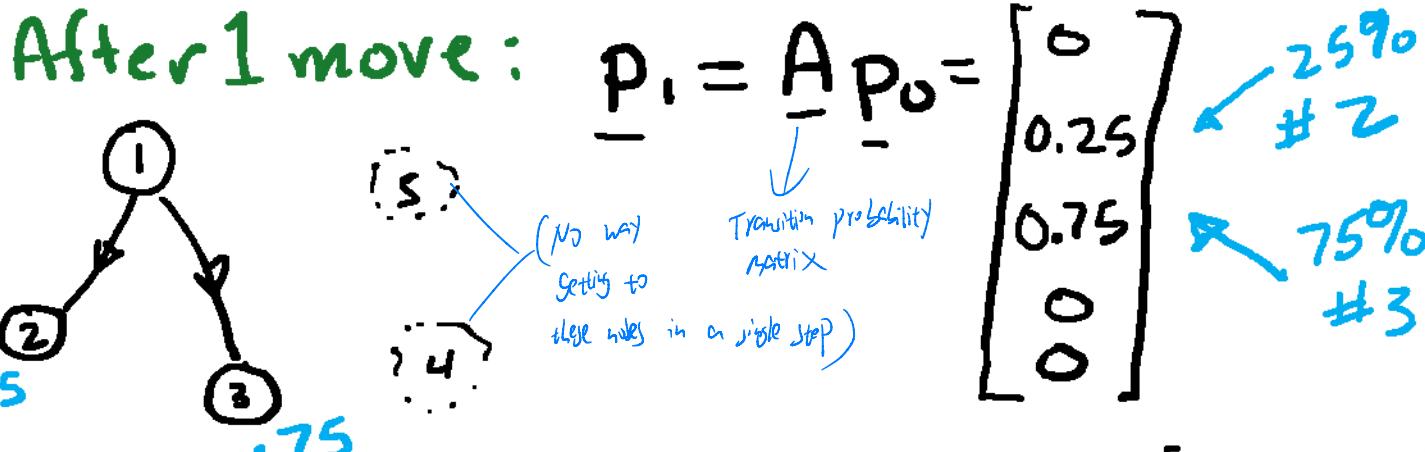
transition probability matrix -
columns sum to 1

Transition probability matrix predicts "paths"

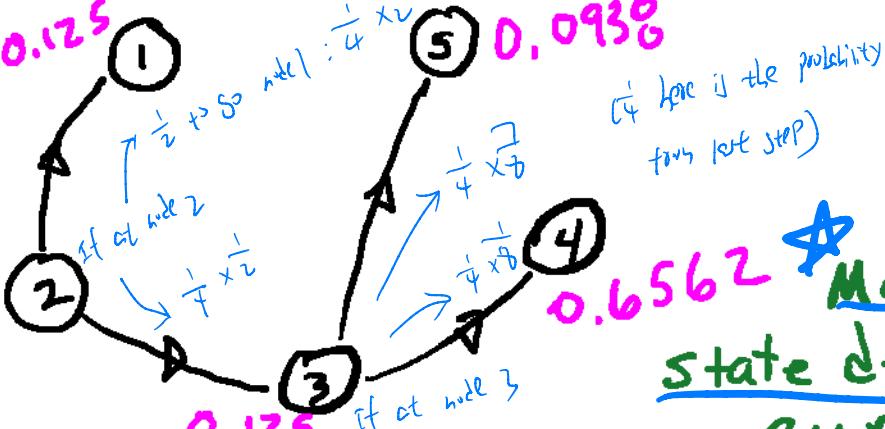
posted at node 1 3

$$A = \begin{bmatrix} 0 & \frac{1}{2} & 0 & 0 & \frac{1}{2} \\ \frac{1}{4} & 0 & 0 & 0 & 0 \\ \frac{3}{4} & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & \frac{7}{18} & 0 & \frac{4}{9} \\ 0 & 0 & \frac{4}{9} & 1 & 0 \end{bmatrix}$$

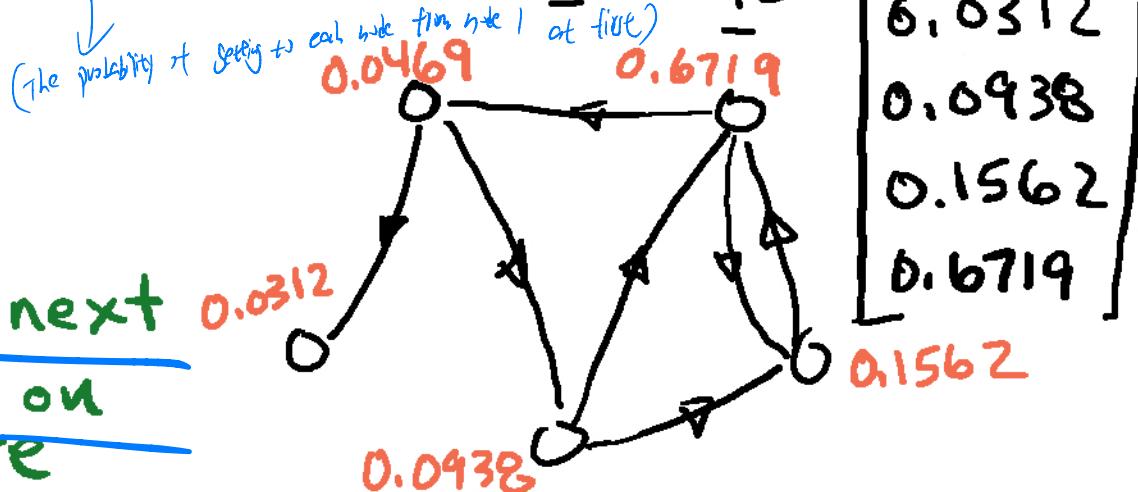
Start at node 1: $P_0 = [1 \ 0 \ 0 \ 0 \ 0]^T$



After 2 moves: $P_2 = AP_1 =$



After 3 moves: $P_3 = AP_2 =$

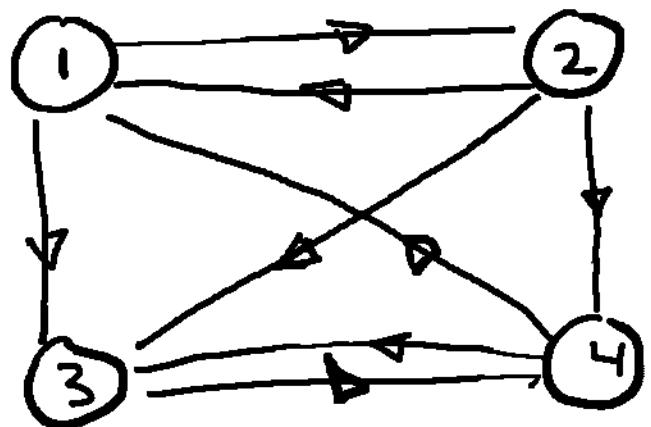


Markov chain: next state depends only on current state

PageRank algorithm ranks web pages

4

where will I visit most?



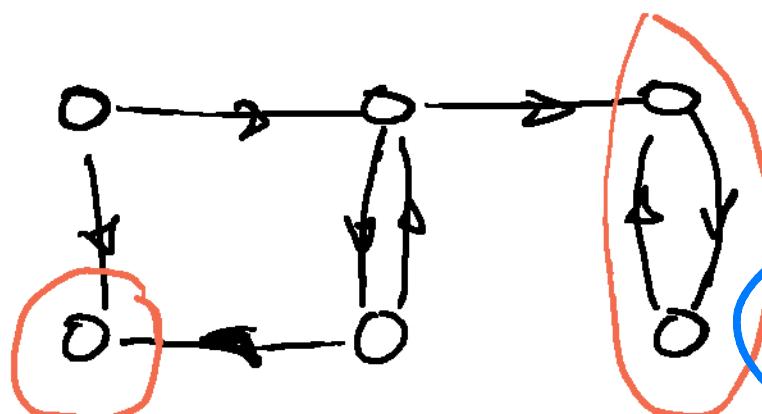
1) Adjacency matrix

$$\tilde{A} = \begin{bmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$

2) Normalize columns
(equal prob. outlinks)

$$A = \begin{bmatrix} 0 & \frac{1}{3} & 0 & \frac{1}{2} \\ \frac{1}{2} & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{3} & 0 & \frac{1}{2} \\ 0 & \frac{1}{3} & 1 & 0 \end{bmatrix}$$

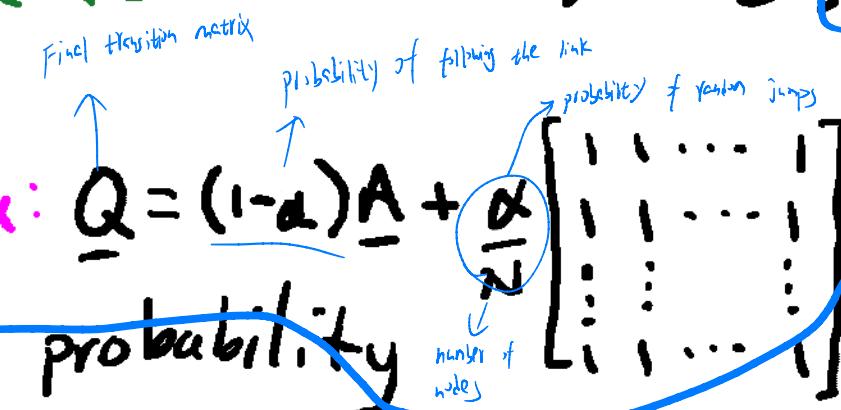
3) Eliminate traps



introduce small probability
to go from any node to
any other node

\star Transition matrix: $Q = (1-\alpha)\underline{A} + \frac{\alpha}{N} \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$

$\frac{\alpha}{N}$: random jump probability



Eigenvector of \underline{Q} , ranks pages

\underline{Q} is irreducible (no traps) column stochastic (cols sum to 1), with non negative entries \Rightarrow

(Perron - Frobenius) Largest eval is 1, evect $\underline{P} = [\underline{p}_1 \cdots \underline{p}_N]^T$
 satisfies $p_i > 0$, $\sum_i p_i = 1$

\downarrow
eigenvalue

\downarrow
eigenvector

$$\underline{u} = \frac{1}{N} [1 \ 1 \ \dots \ 1]^T$$

$$\lim_{k \rightarrow \infty} \underline{Q}^k \underline{u} = \underline{P}$$

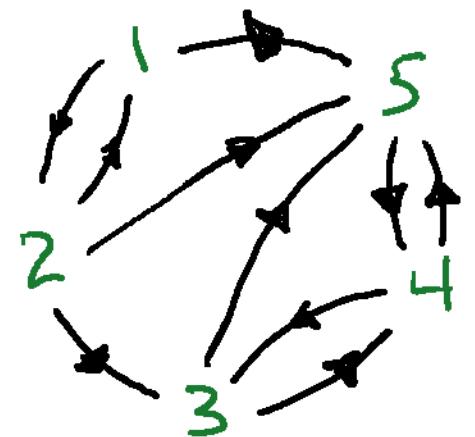
Steady-state
Distribution

$$\underline{Q} \underline{P} = \underline{P}$$

The scale of the previous power iteration
algorithm

\underline{P} ranks importance of pages

Example:



$$\tilde{A} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{bmatrix}$$

Normalize and $\alpha = 0.01$ give us \underline{Q} (transition prob matrix)

$$\underline{Q} = (\alpha = 0.01)$$

$$\underline{Q} = \begin{bmatrix} .002 & .332 & .002 & .002 & .002 \\ .497 & .002 & .002 & .002 & .002 \\ .002 & .932 & .002 & .497 & .002 \\ .002 & .002 & .497 & .062 & .992 \\ .497 & .332 & .497 & .497 & .002 \end{bmatrix}$$

$$\underline{u} = \begin{bmatrix} 0.2 \\ 0.2 \\ 0.2 \\ 0.2 \\ 0.2 \end{bmatrix}, \quad \underline{Q}\underline{u} = \begin{bmatrix} .07 \\ .10 \\ .17 \\ .30 \\ .36 \end{bmatrix}, \quad \underline{Q}^2\underline{u} = \begin{bmatrix} .003 \\ .004 \\ -.18 \\ -.45 \\ -.30 \end{bmatrix}, \quad \dots \quad \underline{Q}^{10}\underline{u} = \begin{bmatrix} .003 \\ .004 \\ .22 \\ -.44 \\ -.33 \end{bmatrix}$$

Converged

Initial guess, assume the probability are the same

$$\underline{P} = \begin{bmatrix} 0.0032 & 0.0036 & 0.2211 & 0.4401 & 0.3320 \end{bmatrix}$$

Main eigenvector

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