

PageRank: Matrix Formulation

- **Stochastic adjacency matrix M**

- Let page i has d_i out-links

- If $i \rightarrow j$, then $M_{ji} = \frac{1}{d_i}$ else $M_{ji} = 0$

- M is a **column stochastic matrix**

- Columns sum to 1

- **Rank vector r :** vector with an entry per page

- r_i is the importance score of page i

- $\sum_i r_i = 1$

- **The flow equations can be written**

$$r = M \cdot r$$

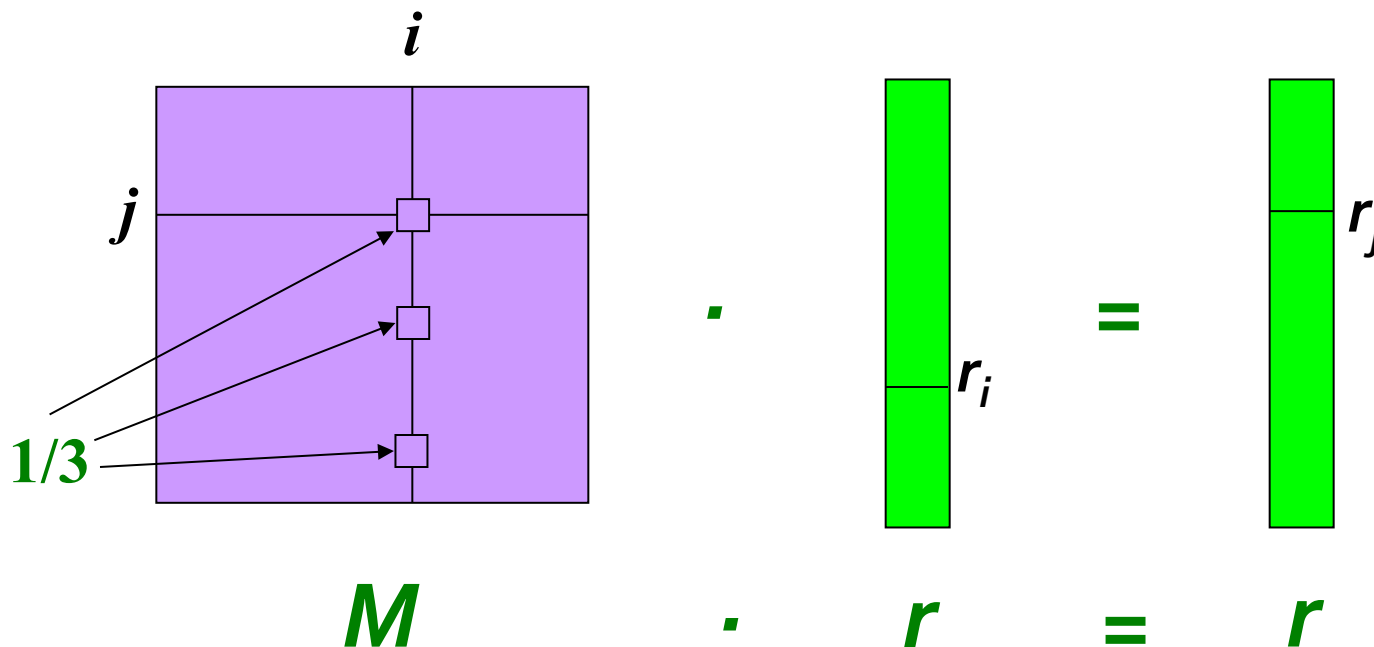
$$r_j = \sum_{i \rightarrow j} \frac{r_i}{d_i}$$

Example

- Remember the flow equation: $r_j = \sum_{i \rightarrow j} \frac{r_i}{d_i}$
- Flow equation in the matrix form

$$M \cdot r = r$$

- Suppose page i links to 3 pages, including j



Eigenvector Formulation

- The flow equations can be written

$$\mathbf{r} = \mathbf{M} \cdot \mathbf{r}$$

- So the rank vector \mathbf{r} is an eigenvector of the stochastic web matrix \mathbf{M}

- In fact, its first or principal eigenvector, with corresponding eigenvalue 1

- Largest eigenvalue of \mathbf{M} is 1 since \mathbf{M} is column stochastic

- *Why? We know \mathbf{r} is unit length and each column of \mathbf{M} sums to one, so $\mathbf{M}\mathbf{r} \leq 1$*

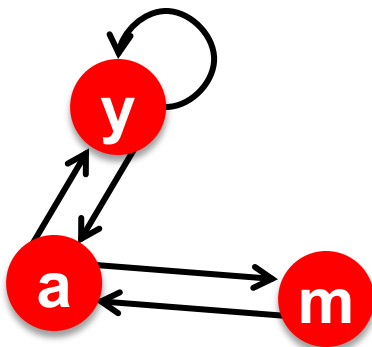
- We can now efficiently solve for \mathbf{r} !

The method is called Power iteration

NOTE: \mathbf{x} is an eigenvector with the corresponding eigenvalue λ if:

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x}$$

Example: Flow Equations & M



	y	a	m
y	$\frac{1}{2}$	$\frac{1}{2}$	0
a	$\frac{1}{2}$	0	1
m	0	$\frac{1}{2}$	0

$$r = M \cdot r$$

$$r_y = r_y/2 + r_a/2$$

$$r_a = r_y/2 + r_m$$

$$r_m = r_a/2$$

$$\begin{bmatrix} y \\ a \\ m \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & 0 & 1 \\ 0 & \frac{1}{2} & 0 \end{bmatrix} \begin{bmatrix} y \\ a \\ m \end{bmatrix}$$