

GENERALIZING GOOGLE'S PAGERANK TO RANK NATIONAL FOOTBALL LEAGUE TEAMS

Anjela Govan, Carl D. Meyer, and Russell Albright

North Carolina State University,
SAS Institute Inc.

March 2008

Definition

A *ranking* is a technique of assigning a positive integer (rank) to each object in a finite set.

Ranks are often based on a rating score denoting the degree of importance of each object in the set.

PageRank Model

Use link structure of the web to rank web pages.

Random web surfer:

- Choose a link in the current web page at random
- The proportion of time spent at web page i is the rating of web page i
- Random surfer will spend the larger proportion of time on important web sites

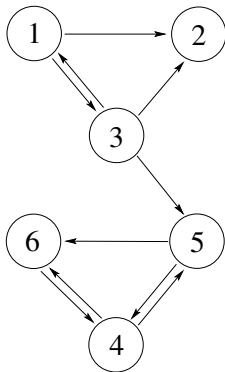
A web page is important if it is pointed to by other important web page(s).

Problem with random web surfer model: “rank sinks”

PageRank Algorithm, hyperlink matrix

1. Web with n web pages \rightarrow directed graph, with n nodes.
2. Form hyperlink matrix \mathbf{H} .

$$H_{ij} = \begin{cases} 1/\text{deg}^-(P_i) & P_i \rightarrow P_j \\ 0 & P_i \not\rightarrow P_j \end{cases}$$



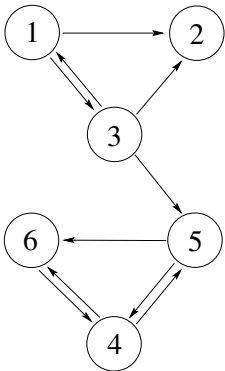
$$\mathbf{H} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & 0 & \frac{1}{3} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

PageRank Algorithm, stochastic matrix

3. Form stochastic matrix \mathbf{S} (fix dangling nodes)

$$\mathbf{S} = \mathbf{H} + \mathbf{a}\mathbf{u}^T \quad a_i = \begin{cases} 1 & \text{if } \mathbf{H}_i = 0 \\ 0 & \text{otherwise} \end{cases}$$

\mathbf{u} is a probability distribution vector.



$$\mathbf{H} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & 0 & \frac{1}{3} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

PageRank Outline, Google matrix

4. Form Google matrix \mathbf{G}

$$\mathbf{G} = \alpha \mathbf{S} + (1 - \alpha) \mathbf{e} \mathbf{v}^T$$

$0 < \alpha < 1$, and $\mathbf{v} > 0$ is a probability distribution vector.

5. The vector containing the ratings of each web page is π such that

$$\pi^T = \pi^T \mathbf{G}$$

Use rating scores in π to rank web pages.

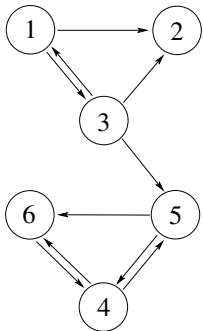
Small Example, Google matrix

$$\mathbf{G} = 0.85\mathbf{S} + (1 - 0.85)(1/6)\mathbf{e}\mathbf{e}^T =$$

$$\begin{pmatrix} \frac{1}{40} & \frac{9}{20} & \frac{9}{20} & \frac{1}{40} & \frac{1}{40} & \frac{1}{40} \\ \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \\ \frac{6}{37} & \frac{6}{37} & \frac{6}{1} & \frac{6}{1} & \frac{6}{37} & \frac{6}{1} \\ \frac{120}{1} & \frac{120}{1} & \frac{40}{1} & \frac{40}{1} & \frac{120}{9} & \frac{40}{9} \\ \frac{40}{1} & \frac{40}{1} & \frac{40}{1} & \frac{40}{1} & \frac{20}{9} & \frac{20}{9} \\ \frac{40}{1} & \frac{40}{1} & \frac{40}{1} & \frac{20}{7} & \frac{40}{1} & \frac{20}{1} \\ \frac{1}{40} & \frac{1}{40} & \frac{1}{40} & \frac{1}{8} & \frac{1}{40} & \frac{1}{40} \end{pmatrix}$$

Small Example, rating vector π

$$\pi^T \cong (0.05170 \quad 0.0737 \quad 0.0574 \quad 0.3487 \quad 0.1999 \quad 0.2686)$$



The list of the web pages in the order of rating scores (highest to lowest) is

4 6 5 2 3 1

GeM (Generalized Markov Method)

1. Sport season with n teams \rightarrow weighted directed graph with n nodes.

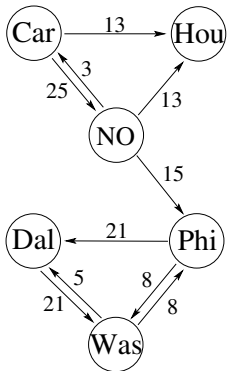
Game between i and $j \rightarrow$ edge from loser to winner with weight equal to the positive difference of the game statistic (e.g. scores, yards, fumbles, etc.).

If i lost to team j more than once during a season w_{ij} is the sum of the positive differences of the statistic of the games team i lost to j .

GeM Method, Gamelink matrix

2. Form matrix H .

$$H_{ij} = \begin{cases} 1 / \sum_{k=1}^n w_{ik} & \text{team } i \text{ lost to } j \\ 0 & \text{otherwise} \end{cases}$$



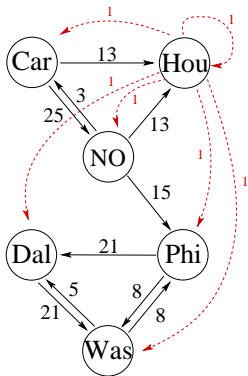
	Car	Dal	Hou	NO	Phi	Was
Car	0	0	$\frac{13}{38}$	$\frac{25}{38}$	0	0
Dal	0	0	0	0	0	1
Hou	0	0	0	0	0	0
NO	$\frac{3}{31}$	0	$\frac{13}{31}$	0	$\frac{15}{31}$	0
Phi	0	$\frac{21}{29}$	0	0	0	$\frac{8}{29}$
Was	0	$\frac{5}{13}$	0	0	$\frac{8}{13}$	0

GeM Method, Stochastic matrix

3. Form matrix S .

$$S = H + \mathbf{a}\mathbf{u}^T, \quad a_i = \begin{cases} 1 & \text{if } \mathbf{H}_i = 0 \\ 0 & \text{otherwise} \end{cases}$$

\mathbf{u} is a probability distribution vector.



	Car	Dal	Hou	NO	Phi	Was
Car	0	0	$\frac{13}{38}$	$\frac{25}{38}$	0	0
Dal	0	0	0	0	0	1
Hou	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$
NO	$\frac{3}{31}$	0	$\frac{13}{31}$	0	$\frac{15}{31}$	0
Phi	0	$\frac{21}{29}$	0	0	0	$\frac{8}{29}$
Was	0	$\frac{5}{13}$	0	0	$\frac{8}{13}$	0

GeM, GeM matrix

4. Form GeM matrix \mathbf{G} .

$$\mathbf{G} = \alpha_0 \mathbf{S}_0 + \alpha_1 \mathbf{S}_1 + \dots + \alpha_p \mathbf{S}_p$$

where $0 \leq \alpha_i < 1$, $\sum \alpha_i = 1$ and \mathbf{S}_i is a stochastic matrix $0 \leq i \leq p$, we will call it an i th feature matrix.

5. The vector containing the rating scores of each team is π such that

$$\pi^T = \pi^T \mathbf{G}$$

Use rating scores in π to rank teams.

Feature Matrices GeM (FMGeM).

$$\mathbf{G} = \alpha_0 \mathbf{S}_0 + \alpha_1 \mathbf{S}_1 + \dots + \alpha_p \mathbf{S}_p$$

where $0 \leq \alpha_i < 1$, $\sum \alpha_i = 1$ and \mathbf{S}_i is the i th feature matrix
 $0 \leq i \leq p$.

$\mathbf{S}_i = \mathbf{H}_i + \mathbf{a}_i \mathbf{u}_i^T$, where \mathbf{H}_i is a gamelink matrix formed using i th statistic (e.g. game scores, total yards, rushing yards)

Rank One update GeM (ROGeM).

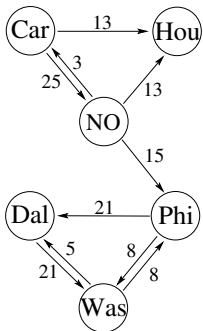
$$\mathbf{G} = \alpha[\mathbf{H} + \mathbf{a}\mathbf{u}^T] + (1 - \alpha)\mathbf{e}\mathbf{v}^T$$

where $0 < \alpha < 1$, $\mathbf{v} > 0$ and \mathbf{u} are probability distribution vectors and

$$a_i = \begin{cases} 1 & \text{if } \mathbf{H}_i = 0 \\ 0 & \text{otherwise} \end{cases} .$$

Small Example, rating vector π

$$\pi^T \approx (0.0389 \quad 0.2824 \quad 0.0656 \quad 0.056 \quad 0.2289 \quad 0.3281)$$



The list of the teams in the order of rating scores (highest to lowest) is

Was Dal Phi Hou NO Car