Data Sci Discover Project

Ryan Greenup

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The Random Surfer model is:

$$S = \alpha T + F$$

where:

- T
- is an $i \times j$ matrix that describes the probability of travelling from vertex j to i
 - * This is transpose from the way that igraph produces an adjacency matrix.

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$$\mathbf{F} = \begin{bmatrix} \frac{1}{n} \\ \frac{1}{n} \\ \frac{1}{n} \end{bmatrix}$$

Interpreting the transition probability matrix in this way is such that $\mathbf{T} = \mathbf{A}\mathbf{D}_A^{-1}$ under the following conditions:

- No column of A sums to zero
 - If this does happen the question arises how to deal with $\mathbf{D}_{\mathbf{A}}^{-1}$
 - * I've been doing $D'_{\mathbf{A},i,j} := \operatorname{diag}\left(\frac{1}{\operatorname{colsums}(\mathbf{A})}\right)$ and then replacing any 0 on the diagonal with 1.

- What is done in the paper is to make another matrix **Z** that is filled with 0, if a column sum of **A** adds to zero then that column in **Z** becomes $\frac{1}{n}$
 - * This has the effect of making each row identical
 - * The probability of going from an orphaned vertex to any other vertex would hence be $\frac{1}{n}$
 - * The idea with this method is then to use $D_{(\mathbf{A}+\mathbf{Z})}^{-1}$ this will be consistent with the *Random Surfer* the method using \mathbf{F} in]

where each row is identical that is a 0

The way to deal with the *Power Walk* is more or less the same.

observe that:

$$\mathbf{B} = \beta^{\mathbf{A}} \wedge \mathbf{A}_{i,j} \in \mathbb{R} \implies |\mathbf{B}_{i,j}| > 0 \quad \forall i, j > n \in \mathbb{Z}^+$$

Be mindful that the use of exponentiation in] is not an element wise exponentiation and not an actual matrix exponential (which would be defined by using power series and logs but is defined)

So if I have:

- $\mathbf{O}_{i,j} := 0, \quad \forall i, j \leq n \in \mathbb{Z}^+$
- $\vec{p_i}$ as the state distribution, being a vector of length n

Then It can be shown (see (2)):

$$\mathbf{O}\mathbf{D}_{\mathbf{B}}^{-1}\vec{p_i} = \vec{\delta}^{\mathbf{T}}\vec{p_i}$$

where:

•
$$\vec{\delta_i} = \frac{1}{\text{colsums}(\mathbf{R})}$$

This means we can do:

$$\begin{split} \vec{p_{i+1}} &= \mathbf{B} \mathbf{D}_{\mathbf{B}}^{-1} \vec{p_i} \\ &= \left(\mathbf{B} - \mathbf{O} + \mathbf{O} \right) \mathbf{D}_{\mathbf{B}}^{-1} \vec{p_i} \\ &= \left(\left(\mathbf{B} - \mathbf{O} \right) \mathbf{D}_{\mathbf{B}}^{-1} + \mathbf{O} \mathbf{D}_{\mathbf{B}}^{-1} \right) \vec{p_i} \\ &= \left(\mathbf{B} - \mathbf{O} \right) \mathbf{D}_{\mathbf{B}}^{-1} \vec{p_i} + \mathbf{O} \mathbf{D}_{\mathbf{B}}^{-1} \vec{p_i} \\ &= \left(\mathbf{B} - \mathbf{O} \right) \mathbf{D}_{\mathbf{B}}^{-1} \vec{p_i} + \vec{\delta'} \vec{p_i} \vec{1} \end{split}$$

$$\mathbf{OD_{B}^{-1}} \overrightarrow{p_{i}} = \begin{pmatrix}
1 & 1 & 1 & & \\
1 & 1 & 1 & & \\
1 & 1 & 1 & & \\
1 & 1 & 1 & & \\
\vdots & \ddots & & \\
\end{pmatrix} \begin{pmatrix}
\sum_{i=1}^{n} [A_{i,1}] & 0 & 0 & & \\
0 & \sum_{i=1}^{n} [A_{i,2}] & 0 & & \dots \\
0 & 0 & \sum_{i=1}^{n} [A_{i,3}] & & \\
0 & \vdots & 0 & \ddots & \\
\end{pmatrix} \begin{pmatrix}
p_{1} & & & \\
p_{2} & & & \\
p_{3} & & & \\
\vdots & & & \ddots & \\
\end{pmatrix} \\
= \begin{pmatrix}
\frac{p_{1}}{\sum_{i=1}^{n} [A_{i,1}]} & \frac{p_{2}}{\sum_{i=1}^{n} [A_{2,1}]} & \frac{p_{3}}{\sum_{i=1}^{n} [A_{3,1}]} & & \\
\sum_{i=1}^{n} [A_{i,1}] & \frac{p_{2}}{\sum_{i=1}^{n} [A_{2,1}]} & \frac{p_{3}}{\sum_{i=1}^{n} [A_{3,1}]} & & \\
\vdots & \vdots & \vdots & \vdots & \vdots & \\
= \begin{pmatrix}
\frac{1}{\sum_{i=1}^{n} [A_{i,1}]} & \frac{1}{\sum_{i=1}^{n} [A_{i,2}]} & \frac{1}{\sum_{i=1}^{n} [A_{i,3}]} & \end{pmatrix} \begin{pmatrix} p_{1} \\ p_{2} \\ p_{3} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\
= \delta^{T} \overrightarrow{p} \overrightarrow{1} & (2)$$