Linear Algebra hw4 PageRank

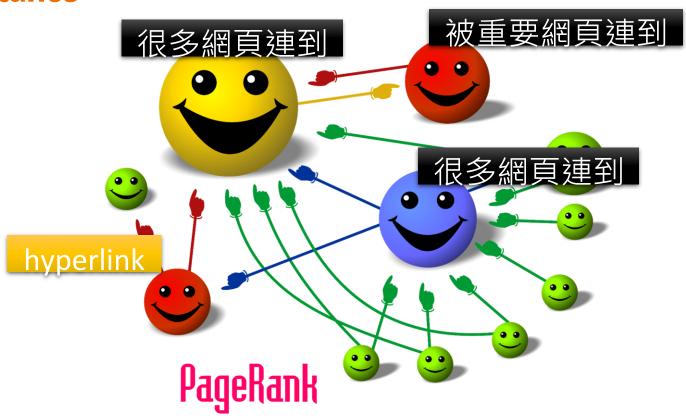
呂元翔

Introduction - PageRank

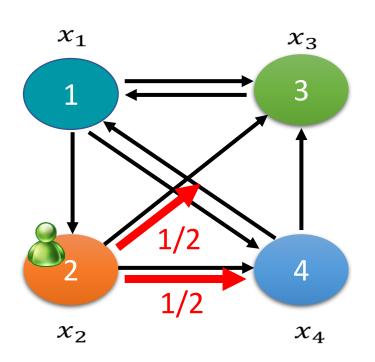
- PageRank is an algorithm to measure the importance of a website.
- It was developed by Larry Page and Sergey Brin in 1996.
- The more links a website received from other websites, the more important it is.
- The algorithm iteratively calculate the importance of a website according to the websites that have links to it and their importance.

教授上課影片: Page Rank

Importance



Importance - Formulas



$$x_{1} = x_{3} + \frac{1}{2}x_{4}$$

$$x_{2} = \frac{1}{3}x_{1}$$

$$x_{3} = \frac{1}{3}x_{1} + \frac{1}{2}x_{2} + \frac{1}{2}x_{4}$$

$$x_{4} = \frac{1}{3}x_{1} + \frac{1}{2}x_{2}$$

Importance - Formulas

$$\mathbf{A} = \begin{bmatrix} 0 & 0 & 1 & \frac{1}{2} \\ \frac{1}{3} & 0 & 0 & 0 \\ \frac{1}{3} & \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{3} & \frac{1}{2} & 0 & 0 \end{bmatrix} \qquad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} \qquad \mathbf{x}_1 = x_3 + \frac{1}{2}x_4$$
$$\mathbf{x}_2 = \frac{1}{3}x_1$$

$$Ax = x$$

The solution x is in the eigenspace of eigenvalue $\lambda = 1$

$$x_{1} = x_{3} + \frac{1}{2}x_{4}$$

$$x_{2} = \frac{1}{3}x_{1}$$

$$x_{3} = \frac{1}{3}x_{1} + \frac{1}{2}x_{2} + \frac{1}{2}x_{4}$$

$$x_{4} = \frac{1}{3}x_{1} + \frac{1}{2}x_{2}$$

Does it has a unique ranking solution?

1. There doesn't exist a eigenvalue λ of the matrix that equals 1.

Q: How to ensure there exists a $\lambda = 1$?

Sol: Use Column-Stochastic Matrix

An example of column-stochastic matrix

1. The dimension of eigenspace is > 1

Q: How to ensure the dimension of the eigenspace = 1?

Sol: Use Column-Stochastic matrix and "Positive"

$$\mathbf{M} = (1 - m)\mathbf{A} + m\mathbf{S}$$

ref: https://en.wikipedia.org/wiki/Stochastic matrix

Power method

Find x^* , such that $x^* = Mx^*$

M is very large

$$x^{1} = Mx^{0}$$

$$x^{2} = Mx^{1}$$

$$\vdots$$

$$x^{k} = Mx^{k-1}$$

Start from x^0

$$x^0 = \begin{bmatrix} 1/n \\ \vdots \\ 1/n \end{bmatrix}$$

all positive, sum to 1

If
$$k \to \infty$$

$$x_k = x^*$$

HW4

$$\mathbf{R} = egin{bmatrix} ext{PageRank}(p_1) \ ext{PageRank}(p_2) \ ext{} dots \ ext{PageRank}(p_N) \end{bmatrix}$$

 $\ell(p_i,p_j)$ 1/#ofOutGoingLinksfromPj, if $\mathsf{p_j}$ has a link to $\mathsf{p_i}$ 0. else

$$\mathbf{R} = \left[egin{array}{ccccc} 1/N & & 1/N \ 1/N & & 1/N \ & \ddots & & \ 1/N & & 1/N \ \end{array}
ight] + d \left[egin{array}{ccccc} \ell(p_1,p_1) & \ell(p_1,p_2) & \cdots & \ell(p_1,p_N) \ \ell(p_2,p_1) & \ddots & & \ dots & & \ell(p_i,p_j) \ \ell(p_N,p_1) & & & \ell(p_N,p_N) \ \end{array}
ight] \mathbf{R}$$

uniform

Original transition matrix

nitialize R with
$$egin{bmatrix} 1/N \ 1/N \ dots \ 0 \ \end{matrix}$$

Original transition matrix

Initialize R with
$$\begin{bmatrix} 1/N \\ 1/N \\ \vdots \\ 1/N \end{bmatrix}$$
 . In hw4 we set d = 0.85 and repeat the multiplication for 50 times
$$\mathbf{R}^t = \begin{bmatrix} 1/N & 1/N \\ 1/N & 1/N \\ 1/N & 1/N \end{bmatrix} + d \begin{bmatrix} \ell(p_1, p_1) & \ell(p_1, p_2) & \cdots & \ell(p_1, p_N) \\ \ell(p_2, p_1) & \ddots & \vdots & \ell(p_i, p_j) \\ \vdots & & \ell(p_N, p_N) \end{bmatrix} \mathbf{R}$$

uniform

Until it R converge.



colab link: https://colab.research.google.com/drive/1mRRYa-2BB9T36g0apvveusbDPZ3m4hD_?usp=sharing

Implement the PageRank algorithm and turn a 2-dimensional list into the ordering of web pages.

Download a repository called "Linear_Algebra_2023_fall_hw4" from the github link in colab.

```
1. ) ls -R
PageRank.py inputs outputs

./inputs:
summary1.json summary2.json

./outputs:
answer1.json answer2.json
```

- 1. You will get a 2-dimensional list, describing the topology of a network.
 - E.g. [[0,3],[1,2,3],[2],[0,1]]. The first list [0,3] in this list means the webpage with id 0 has incoming links from 0 and 3. The second list [1,2,3] is the websites webpage 1 has links from, etc.
 - Our list have a length of 300. Each sublist has at least a number from 0~299.

HW4

3. You will also get a PageRank.py python file that helps you start. Try

\$python PageRank.py <input_file> <output_file>

Your can name your <output_file> whatever you like, but need to be a .json file.

Modify the PageRank.py file so that it outputs the correct ordering.

Your <output_file> contains a list, like [2,0,3,1], which should be the ordering of the web pages based on pagerank. The web page with a higher pagerank score comes first.

Float will be acceptable [2.0, 0.0, 3.0,1.0].

Grading

PageRankpy (6%)

- 不要抄作業,不要交別人的答案,作弊一律0分計算
- 僅需要上傳一個檔案到 NTU COOL
- PageRank.py (This is your modified PageRank.py file.)
- DEADLINE: 2023/12/01(五) 23:59 (GMT+8:00)
- 遲交、改格式、改檔名、改檔案、各種奇怪的錯誤無法改作業:每次分數×0.7,每次修改請完整按照格式繳交。

Q & A

- 若有作業相關問題請到NTU COOL作業討論版發問
- 你的問題很可能也是其他同學的問題:)