```
PageRank
Import libs
 import numpy as np
 import pandas as pd
Using transation matrix
 def PageRank1 ( M, dmp = 0.85, eps = 1.0e-5, max iter = 100):
                  : transation matrix
                  : Damping factor ; usually set to 0.85
                  : Pre-specified threshold (desired precision); //used in Stopping condition
         max iter : Maximum number of iterations
     nb site = len (M)
     R_old = [1/nb_site for _ in range(nb_site)]
     cpp = 0
     while True:
         R = M.dot(R_old)
         R = [ (1-dmp)/nb \text{ site } + dmp*r \text{ for } r \text{ in } R]
         flag = True
         for r1, r2 in zip(R, R old):
             if abs(r1 - r2) > eps:
                 flag = False
                 break
         if flag or cpp >= max iter:
             break
         R \text{ old} = R
         cpp+=1
     return cpp, R
 A = np.array([[ 0, 0, 1, 1/2],
                [ 1/3, 0, 0, 0],
               [ 1/3, 1/2, 0, 1/2],
                [ 1/3, 1/2, 0, 0]])
 eps = 0.00001
 nb etr = 100
 Steps = PageRank1(A, eps = eps, max_iter = nb_etr)[0]
 PRvector = PageRank1(A, eps = eps, max_iter = nb_etr)[1]
 print('Steps :', Steps)
 print('PageRank vector :', PRvector)
Steps: 12
PageRank vector: [0.3681575511754208, 0.1418090346209795, 0.28795884393472515, 0.2020745702688743]
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In [4]: B = np.array([[1/11, 1/11, 1/11, 1/11, 1/11, 1/11, 1/11, 1/11, 1/11, 1/11, 1/11], [0.15/11, 0.15/11, (0.15/11) + 0.85, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11][0.15/11, (0.15/11) + 0.85, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11, 0.15/11][(0.15/11)+(0.85/2), (0.15/11)+(0.85/2), 0.15/11, 0.15/[0.15/11, (0.15/11) + (0.85/3), (0.15/11) + (0.85/3), (0.85/11) + (0.85/3), (0.85/11) + (0.8[0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0.15/11, 0.15/11, 0.15/11,

> [0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0[0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0[0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0.15/11, (0.15/11) + (0.85/2), 0.15/11, 0 $[0.15/11,\ 0.15/11,\ 0.15/11,\ 0.15/11,\ (0.15/11) + (0.85/1),\ 0.15/11,\$ [0.15/11, 0.15/11, 0.15/11, 0.15/11, (0.15/11) + (0.85/1), 0.15/11, 0.15/

[0.01363636 0.01363636 0.86363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.01363636\ 0.86363636\ 0.01363636\ 0.01363636\ 0.01363636\ 0.01363636$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.43863636\ 0.43863636\ 0.01363636\ 0.01363636\ 0.01363636\ 0.01363636$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.01363636 \ 0.2969697 \ 0.01363636 \ 0.2969697 \ 0.01363636 \ 0.2969697$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.01363636\ 0.43863636\ 0.01363636\ 0.01363636\ 0.43863636\ 0.01363636$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] [0.01363636 0.43863636 0.01363636 0.01363636 0.43863636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.01363636\ 0.43863636\ 0.01363636\ 0.01363636\ 0.43863636\ 0.01363636$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] $[0.01363636\ 0.43863636\ 0.01363636\ 0.01363636\ 0.43863636\ 0.01363636$ 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] [0.01363636 0.01363636 0.01363636 0.01363636 0.86363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636] [0.01363636 0.01363636 0.01363636 0.01363636 0.86363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636 0.01363636]] 0.09090909090909091]) T = np.array([[0,0,0,0,0,0,0,0,0,0,0][0,0,0.85/1,0,0,0,0,0,0,0,0],

 $[[0.09090909 \ 0.09090909 \ 0.09090909 \ 0.09090909 \ 0.09090909 \ 0.09090909 \ 0.09090909$

[0, 0.85/1, 0, 0, 0, 0, 0, 0, 0, 0][0.85/2, 0.85/2, 0, 0, 0, 0, 0, 0, 0, 0, 0],

[0,0,0,0,0.85/1,0,0,0,0,0,0], [0,0,0,0,0.85/1,0,0,0,0,0]

0, 0.85/3, 0, 0.85/3, 0, 0.85/3, 0, 0, 0, 0, 0], 0, 0.85/2, 0, 0, 0.85/2, 0, 0, 0, 0, 0, 0], 0, 0.85/2, 0, 0, 0.85/2, 0, 0, 0, 0, 0, 0], 0, 0.85/2, 0, 0, 0.85/2, 0, 0, 0, 0, 0, 0], 0, 0.85/2, 0, 0, 0.85/2, 0, 0, 0, 0, 0, 0],

print('Steps :', PageRank1(T)[0]) print('PageRank vector :', PageRank1(T)[1]) Steps: 22 PageRank vector: [0.0136363636363636364, 0.04916371515284579, 0.04916371515284579, 0.036326175758241076, 0.04 577578244757678, 0.0479391677312744, 0.0479391677312744, 0.0479391677312744, 0.0479391677312744, 0.046714620 30970301, 0.04671462030970301] pages = ['A' , 'B' , 'C' , 'D' , 'E' , 'F' , 'G' , 'H' , 'I' , 'L' , 'M'] frame = pd.DataFrame(list(zip(pages , PageRank1(T)[1])),columns=['pages','Vector']) frame

5 F 0.047939 6 G 0.047939 H 0.047939 7 8 I 0.047939

the structure of the incoming links. It was originally designed as an algorithm to rank web pages.

1 0.333333 0.0 0.0 0.0 2 0.333333 0.5 0.0 0.5 3 0.333333 0.5 0.0 0.0

Steps: 12

link

0 0 page1 0.368158

1 2 page3 0.287959

2 3 page4 0.202075

3 1 page2 0.141809

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rank

])

print(B)

])

pages

0

2

3

4

10

In [9]:

Vector

A 0.013636

B 0.049164

C 0.049164

D 0.036326

E 0.045776

L 0.046715

Using xml graph

import xml.etree.ElementTree as ET

def pagerank2(G, dmp = 0.85, eps = 1.0e-5, max iter = 100):

PageRank computes a ranking of the nodes in the graph G based on

M 0.046715

Out[7]:

print(PageRank1(B))

```
Parameters
                  : xml path of graph
                  : transation matrix; The adjacency matrix of the web graph
                  : damping factor ; usually set to 0.85
                  : Pre-specified threshold (desired precision); used in Stopping condition
        max iter : Maximum number of iterations ; used in Stopping condition
        Returns
        pagerank: dictionary; Dictionary of nodes with PageRank as value
    graph = ET.parse(G).getroot()
    nb site = len(graph)
    # Transform graph from xml into dictionary
    link dict ={}
    for i, node in enumerate(graph):
        link dict[node.get("link")] =[ i, [link.get('value') for link in node] ]
    print(link dict)
    print('################################")
    # dictionary 2 translation matrix
    M = [[0 for _ in range(nb_site)] for _ in range(nb_site)] # transation matrix
    for i, node in enumerate(link dict):
        for link in link dict[node][1]:
            M[link dict[link][0]][link dict[node][0]] = 1/len(link dict[node][1])
    M = np.array(M)
    print('M:')
    print(pd.DataFrame(M))
    print('#############################")
    R old = [1/nb site for in range(nb site)]
    cpp = 0
    while True:
        R = M.dot(R old)
        R = [ (1-dmp)/nb \text{ site } + dmp*r \text{ for } r \text{ in } R]
        flag = True
        for r1, r2 in zip(R, R old):
            if abs(r1 - r2) > eps:
               flag = False
               break
        if flag or cpp >= max iter: # Stopping condition
           break
        R old = R # update
        cpp+=1
    print('Steps :',cpp)
                           return {'id':list(range(nb site)), 'link': list(link dict), 'rank': R}
dict1 = pagerank2('graph1.xml')
df = pd.DataFrame(dict1)
df.sort values(by=['rank'], ascending=False, inplace=True)
df.reset index(drop=True,inplace=True)
{'page1': [0, ['page2', 'page3', 'page4']], 'page2': [1, ['page3', 'page4']], 'page3': [2, ['page1']], 'page
4': [3, ['page3', 'page1']]}
0 0.000000 0.0 1.0 0.5
```