

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
```

1a)

```
In [3]: # Circle topology
# Unweighted adjacency matrix

# Option 1: Manually enter the entries
Atilde = np.array(
    [[0,1,0,0,0,0,0,1],
     [1,0,1,0,0,0,0,0],
     [1,1,0,1,1,0,0,0],
     [0,0,1,0,1,0,0,0],
     [0,0,0,1,0,1,0,0],
     [0,0,0,0,1,0,1,0],
     [0,0,0,0,0,1,0,1],
     [1,0,0,0,0,0,1,0]])

# Option 2: or you can exploit the patterns
# Atilde = np.zeros((8,8))
# for i in range(8): #
#     Atilde[i,(i+1)%8] = 1
#     Atilde[i,(i-1)%8] = 1
# Atilde[2,0] = 1
# Atilde[2,4] = 1

print('Unweighted adjacency matrix')
print(Atilde)
print(' ')
```

Unweighted adjacency matrix

```
[[0 1 0 0 0 0 0 1]
 [1 0 1 0 0 0 0 0]
 [1 1 0 1 1 0 0 0]
 [0 0 1 0 1 0 0 0]
 [0 0 0 1 0 1 0 0]
 [0 0 0 0 1 0 1 0]
 [0 0 0 0 0 1 0 1]
 [1 0 0 0 0 0 1 0]]
```

1b)

```
In [6]: # Find weighted adjacency matrix
# option 1: normalize columns with a for loop
A = np.zeros((8,8), dtype=float)
for k in range(8):
    norm = np.sum(Atilde[:,k])
    if norm != 0:
        A[:,k] = np.round(Atilde[:,k] / norm,2)
    else:
        A[:,k] = Atilde[:,k]
```

```

    # option 2: normalize using numpy.sum() and broadcasting, in a single line
    # A = ???

    print('Weighted adjacency matrix')
    print(A)

```

```

Weighted adjacency matrix
[[0.  0.5  0.  0.  0.  0.  0.  0.5 ]
 [0.33 0.  0.5  0.  0.  0.  0.  0. ]
 [0.33 0.5  0.  0.5  0.33 0.  0.  0. ]
 [0.  0.  0.5  0.  0.33 0.  0.  0. ]
 [0.  0.  0.  0.5  0.  0.5  0.  0. ]
 [0.  0.  0.  0.  0.33 0.  0.5  0. ]
 [0.  0.  0.  0.  0.  0.5  0.  0.5 ]
 [0.33 0.  0.  0.  0.  0.  0.5  0. ]]

```

In [ ]:

1c) and 1d)

```

In [15]: # Power method

b0 = 0.125*np.ones((8,1))
print('b0 = ', b0)
print(' ')

b1 = np.dot(A, b0)
print('b1 = ', b1)
print(' ')

b = b0.copy()
for k in range(1000):
    b = np.dot(A, b)

print('1000 iterations')
print('b = ',b)

```

```

b0 = [[0.125]
      [0.125]
      [0.125]
      [0.125]
      [0.125]
      [0.125]
      [0.125]]

b1 = [[0.125  ]
      [0.10375]
      [0.2075 ]
      [0.10375]
      [0.125  ]
      [0.10375]
      [0.125  ]
      [0.10375]]

1000 iterations
b = [[0.00112892]
     [0.00150232]
     [0.00225262]
     [0.00150232]
     [0.00112892]
     [0.00075029]
     [0.00075203]
     [0.00075029]]

```

**1e) Explanation goes here.**

```
In [ ]: # Nodes 3, 2, 4 (in that order) are the most important.
```

**2a)**

```
In [16]: # Hub topology

Atildehub = np.array(
    [[0,0,0,0,0,0,0,0,1],
     [1,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [0,0,0,0,0,0,0,0,1],
     [1,1,1,1,1,1,1,1,0]])

print('Unweighted adjacency matrix')
print(Atildehub)
print(' ')
```

Unweighted adjacency matrix

```
[[0 0 0 0 0 0 0 0 1]
 [1 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [0 0 0 0 0 0 0 0 1]
 [1 1 1 1 1 1 1 1 0]]
```

2b)

```
In [17]: # find weighted adjacency matrix

Ahub = np.zeros((9,9), dtype=float)
for k in range(9):
    norm = np.sum(Atildehub[:,k])
    Ahub[:,k] = Atildehub[:,k] / norm

print('Weighted adjacency matrix')
print(Ahub)
```

Weighted adjacency matrix

```
[[0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.5   0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.    0.    0.    0.    0.    0.    0.    0.    0.125]
 [0.5   1.    1.    1.    1.    1.    1.    1.    0.   ]]
```

2c) and 2d)

```
In [18]: b0 = (1/9)*np.ones((9,1))
print('b0 = ', b0)
print(' ')

bhub1 = Ahub @ b0
print('bhub1 = ', bhub1)
print(' ')

bhub = b0.copy()
for k in range(1000):
    bhub = Ahub @ bhub

print('1000 iterations')
print('bhub = ', bhub)
print(' ')

bhubr = b0.copy()
for k in range(100):
```

```
bhubr = Ahub @ bhubr

print('100 iterations')
print('bhubr = ',bhubr)
print(' ')

bhub3 = b0.copy()
for k in range(101):
    bhub3 = Ahub @ bhub3

print('101 iterations')
print('bhub3 = ',bhub3)
```

```
b0 = [[0.11111111]
      [0.11111111]
      [0.11111111]
      [0.11111111]
      [0.11111111]
      [0.11111111]
      [0.11111111]
      [0.11111111]]
```

```
bhub1 = [[0.01388889]
          [0.06944444]
          [0.01388889]
          [0.01388889]
          [0.01388889]
          [0.01388889]
          [0.01388889]
          [0.01388889]
          [0.83333333]]
```

1000 iterations

```
bhub = [[0.06060606]
         [0.09090909]
         [0.06060606]
         [0.06060606]
         [0.06060606]
         [0.06060606]
         [0.06060606]
         [0.06060606]
         [0.06060606]
         [0.48484848]]
```

100 iterations

```
bhubr = [[0.06065482]
          [0.09093172]
          [0.06065482]
          [0.06065482]
          [0.06065482]
          [0.06065482]
          [0.06065482]
          [0.06065482]
          [0.48448454]]
```

101 iterations

```
bhub3 = [[0.06056057]
          [0.09088798]
          [0.06056057]
          [0.06056057]
          [0.06056057]
          [0.06056057]
          [0.06056057]
          [0.06056057]
          [0.48518805]]
```

**Complete 2e and 2f below.**

```
In [19]: # 2e comment:
# Node 9 is clearly the most important as every other node points to it.
# Node 2 is slightly more important than the rest of the branch nodes because node

# 2f comment:
# After 1000 iterations score9 is 0.4848 which rounds to 0.485 (I assume this to be
# After 101 iterations, score9 is 0.4852 which rounds to 0.485.
# So 101 iterations is required for 3 decimal point accuracy.
```

```
In [21]: # Problem 3 comment:
# The sign of the singular vectors is not unique.
# If you consider the case where  $u_i = -u_i$  and  $v_i = -v_i$ ,
# then  $\sigma_i * -u_i * -v_i^T = \sigma_i * u_i * v_i^T$ 
# Which is the valid singular value decomposition for  $X$ .
```

```
In [ ]:
```