

# Lab 3

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MECH 6317.001: Dynaics of Complex Networks & Systems

import all important libraries

```
In [302...
import networkx as nx
import numpy
import numpy.linalg as la
import matplotlib.pyplot as plt
```

## Stub code functions

This function prints the top five (or `num`) nodes according to the centrality vector `v`, where `v` takes the form: `v[nidx]` is the centrality of the node that is the `nidx`-th element of `G.nodes()`

```
In [303...
def print_top_5(G,v, num=5):
    thenodes = list(G.nodes())
    idx_list = [(i,v[i]) for i in range(len(v))]
    idx_list = sorted(idx_list, key = lambda x: x[1], reverse=True)
    for i in range(min(num,len(idx_list))):
        nidx, score = idx_list[i]
        print(' %i. %s (%1.4f)' % (i+1,thenodes[nidx],score))
        #print ' %i. %s' % (i+1,G.node_object(idx))
```

This function returns the index of the maximum of the array. If two or more indices have the same max value, the first index is returned.

```
In [304...
def index_of_max(v):
    return numpy.where(v == max(v))[0]
```

This function accepts a dictionary of nodes with centrality values and returns a centrality vector

```
In [305... def centrality_vector(G,d):
    thenodes = list(G.nodes())
    v = numpy.zeros((G.number_of_nodes(),))
    for i,u in enumerate(thenodes):
        v[i] = d[u]
    return v
```

This function provides the index of a node based on its order in `G.nodes()`

```
In [306... def node_index(G,n):
    thenodes = list(G.nodes())
    return thenodes.index(n)
```

Now we read in the edgelist file that contains the coappearance network we will analyze. We will look at two different networks, corresponding to only the *Lord of the Rings* series and the *Lord of the Rings* series plus the prequel, *The Hobbit*. The `unweighted` boolean, if set to `True` will set all the edge weights to one. Recall that setting all weights to 1 is different (in NetworkX) from having no weights assigned, which could be accomplished instead by: `G = nx.read_edgelist('LoTR_characters.edgelist',data=False)`.

```
In [307... unweighted = False
#G = nx.read_weighted_edgelist('LotR_characters.edgelist') # just Lord of the Rings
G = nx.read_weighted_edgelist('hobbit_LoTR_characters.edgelist') # with the Hobbit
if unweighted:
    for u,v in G.edges():
        G[u][v]['weight'] = 1
A = nx.to_numpy_array(G)
N = G.number_of_nodes()
```

## Section 7.1: Degree Centrality

use the defined function from the stub code to print out top 5 characters with highest degree centrality

```
In [308... d = dict.fromkeys(G.nodes(),0)           # predefine d as a dictionary for all nodes in G with degree 0
for i in G.nodes():                       # loop over all nodes
    idx = node_index(G,i)                 # get index for node i
    d[i] = sum(A[idx])                    # sum all weights in row of adjacency matrix
v = centrality_vector(G,d)                # use function to get centrality vector
```

```
print('top 5 degree centrality characters') # Label output
print_top_5(G,v)                          # use function to print top five characters in terms of degree centrality (
```

top 5 degree centrality characters

1. gandalf (901.0000)
2. frodo (661.0000)
3. aragorn (632.0000)
4. pippin (606.0000)
5. bilbo (602.0000)

## Section 7.2: Eigenvector Centrality

print out the eigenvector centrality by using the built in function `eigenvector_centrality`

In [309...

```
print('Eigenvector Centrality (by NetworkX):') # display the method via networkx
d = nx.eigenvector_centrality(G,weight='weight') # dictionary output of networkx function
v = centrality_vector(G,d)                       # get eigenvector centrality vector
print_top_5(G,v)                                # use function to output top 5 characters
```

Eigenvector Centrality (by NetworkX):

1. gandalf (0.3570)
2. frodo (0.2827)
3. aragorn (0.2783)
4. pippin (0.2723)
5. bilbo (0.2287)

print out the eigenvector centrality by using linear algebra

In [310...

```
print('Eigenvector Centrality (by linear algebra):') # display the method via linear algebra
k, v = la.eig(A)                                     # find eigenvalues k and eigenvectors v
k1_idx = index_of_max(k)                             # find the index of the largest eigenvalue
v_R = numpy.abs(v[:,k1_idx])                         # centrality vector of function (assumes no complex values)
print_top_5(G,v_R)                                   # use function to output top 5 characters
```

Eigenvector Centrality (by linear algebra):

1. gandalf (0.3570)
2. frodo (0.2827)
3. aragorn (0.2783)
4. pippin (0.2723)
5. bilbo (0.2287)

print out both the centrality and the weighted sum of the centralities(normalized by  $k_1$ ) of the neighbors of any character

```
In [311... noi = 'aragorn'           # choose any character
noi_idx = node_index(G,noi) # use function to find index value for given character
# label output
print('Confirming that eigenvector centrality is a steady-state of sorts for node %s:' % noi)

centrality = numpy.abs(v[noi_idx,k1_idx]) # find eigenvector centrality from centra
print('Eigenvector centrality for node %s: %1.4f' % (noi,centrality)) # output eigenvector centrality
centrality_sum = 0 # initialize a value at 0
noi_neighb_list = list(G.neighbors(noi)) # find all neighbors
for neighb in noi_neighb_list: # add each neighbor
    neighb_idx = node_index(G,neighb) # index of neighbors
    neighb_centrality = numpy.abs(v[neighb_idx,k1_idx]) # centrality value of neighbors
    centrality_sum = centrality_sum + A[noi_idx,neighb_idx]*neighb_centrality # add weighted centrality of neighbors
centrality_sum = centrality_sum / numpy.abs(max(k)) # normalize by largest eigenvector
# output normalized centrality sum
print('Sum of the centralities of neighbors of %s normalized by the larges eigenvalue: %1.4f' % (noi,centrality_sum))
```

Confirming that eigenvector centrality is a steady-state of sorts for node aragorn:

Eigenvector centrality for node aragorn: 0.2783

Sum of the centralities of neighbors of aragorn normalized by the larges eigenvalue: 0.2783

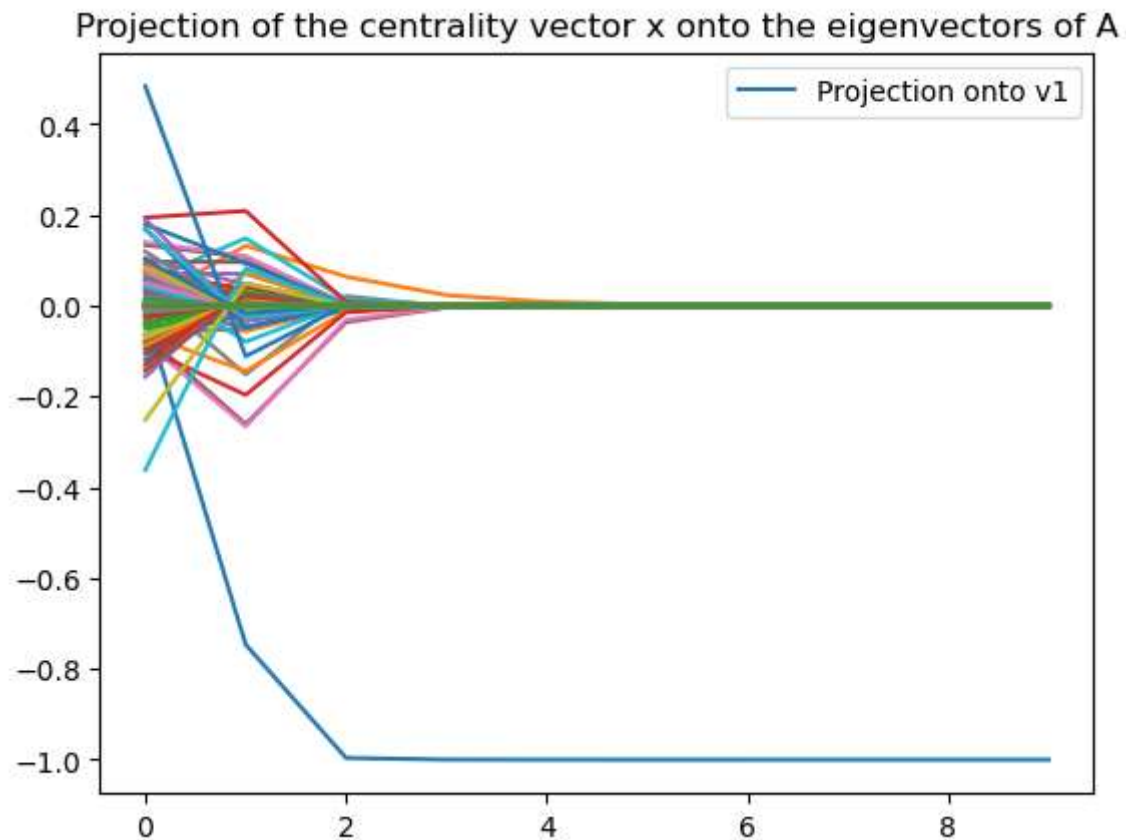
The following section should run and produce a plot that you need to interpret.

```
In [312... print('Showing the convergece of eigenvector centrality...')
num_steps = 10
x = numpy.zeros((N,1)) # initial centrality vector
x[76] = 1
cs = numpy.zeros((N,num_steps))
for i in range(num_steps):
    x = x/la.norm(x) # at each step we need to normalize the centrality vector
    for j in range(G.number_of_nodes()):
        cs[j,i] = numpy.real(numpy.dot( x.T , v[:,j] ))[0] # project x onto each of the eigenvectors
    x = numpy.dot(A,x) # "pass" the centrality one step forward

plt.figure() # this creates a figure to plot in
for i in range(G.number_of_nodes()): # for each eigenvector plot the projection of x onto it over the steps
    if i == k1_idx:
        plt.plot(range(num_steps),cs[i,:],label='Projection onto v1') # only label the eigenvector v1
    else:
        plt.plot(range(num_steps),cs[i,:])
```

```
#plt.ylim([-0.2,1.1]) # this sets the limits for the y axis
plt.legend(loc='best') # this attaches a legend
plt.title('Projection of the centrality vector x onto the eigenvectors of A') # this adds a title
plt.show() # this makes the figure appear
```

Showing the converge of eigenvector centrality...



In this graph, we are seeing the evolution of the eigenvector centrality from an arbitrary starting point. The blue line that quickly approaches 1 is the eigenvector that relates to the largest eigenvalue. This eigenvector is equivalent to the centrality vector while the other eigenvectors are ignored. The graph proves that this method is valid as the ignored eigenvectors converge to zero in steady state, and the eigenvector centrality converges to 1 in steady state.

## Section 7.3: Katz Centrality

## print the top 5 central characters in terms of Katz centrality for varying values of $\alpha$ using linear algebra

In [313...

```
print('Katz Centrality:')
inv_max_eig = 1/max(numpy.abs(k))
alpha_list = [inv_max_eig, inv_max_eig-0.0001, inv_max_eig/2, inv_max_eig/3, inv_max_eig/4]
for alpha in alpha_list:
    v = numpy.dot( la.inv(numpy.eye(N) - alpha*A) , numpy.ones((N,1)) )
    print('Top 5 central characters using Katz centrality (alpha = %1.4f):' % alpha)
    print_top_5(G,v)
```

```
# Label section
# inverse of largest eigenvalue
# various alpha values
# repeat for each alpha value
# linear algebraic equation for
# Label output with alpha value
# output top characters
```

Katz Centrality:

Top 5 central characters using Katz centrality (alpha = 0.0032):

1. gandalf (3506233965768915.0000)
2. frodo (2776134896343451.5000)
3. aragorn (2733443552956267.5000)
4. pippin (2674606740861643.0000)
5. bilbo (2246654230718593.0000)

Top 5 central characters using Katz centrality (alpha = 0.0031):

1. gandalf (85.9849)
2. frodo (68.0973)
3. aragorn (66.9911)
4. pippin (65.5381)
5. bilbo (55.6171)

Top 5 central characters using Katz centrality (alpha = 0.0016):

1. gandalf (3.8101)
2. frodo (3.1354)
3. aragorn (3.0695)
4. pippin (3.0064)
5. bilbo (2.8592)

Top 5 central characters using Katz centrality (alpha = 0.0011):

1. gandalf (2.4204)
2. frodo (2.0663)
3. aragorn (2.0286)
4. pippin (1.9938)
5. bilbo (1.9446)

Top 5 central characters using Katz centrality (alpha = 0.0008):

1. gandalf (1.9521)
2. frodo (1.7105)
3. aragorn (1.6839)
4. pippin (1.6595)
5. bilbo (1.6343)

## Section 7.4: PageRank

print the top 5 central characters in terms of PageRank using linear algebra

```
In [314... print('PageRank')           # Label section
D = numpy.zeros((N,N))      # create fully zero matrix of size NxN
for i in G.nodes():         # Loop to find nodes that have zero out-degree
    idx = node_index(G,i)    # find index value of node
    k_out = sum(A[:,idx])     # calculate out-degree of the node
    D[idx,idx] = max(k_out,1) # definition of D
alpha = 0.95                # guess below 1
v = numpy.dot( la.inv( numpy.eye(N) - alpha*numpy.dot(A,la.inv(D)) ), numpy.ones((N,1)) ) # calculate PageRank central
print_top_5(G,v) # output top characters
```

```
PageRank
1. gandalf (160.2037)
2. frodo (118.7663)
3. aragorn (111.7720)
4. pippin (108.4245)
5. bilbo (107.6881)
```

The proof that  $v_1 = (k_1, k_2, \dots, k_n)$  where  $k_i$  is the degree of node  $i$ , is an eigenvector of  $AD^{-1}$  is shown in the lab3\_tjk19000.pdf file submitted with this lab

## Section 7.5: Hubs & Authorites

the expression of hub eigenvectors in terms of the authorities eigenvectors is shown in the lab3\_tjk190000.odf file submitted with this lab

## Section 7.7: Betweenness Centrality

Using the networkx function for betweenness centrality print the top 5 characters with the highest betweenness centrality

```
In [315... print('Betweenness Centrality')           # Label section
d = nx.betweenness centrality(G,weight='weight') # use function to recieve dictionary
```

```
v = centrality_vector(G,d)      # find betweenness centrality vector from dictionary  
print_top_5(G,v)               # output top characters
```

Betweenness Centrality

1. gandalf (0.0498)
2. pippin (0.0370)
3. aragorn (0.0339)
4. frodo (0.0296)
5. bilbo (0.0292)