DNA_06

September 4, 2016

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2 Introduction

Dynamic Network Analysis of Enron Email Network Data.

I use the Enron email network data from John Hopkins which has time, sender and receiver pair format data.

Here I consolidate earlier analysis to make presentation of the final results easier and more readable

From the JHU data, I have done the following in Excel: - The first column represents seconds elapsed since 1 January 1970, so I convert this in to days - I then add these days to the date to

get time stamps for all nodes - From the timestamps, I extract the year field - The network can be partitioned by the required level of granularity

```
In [1]: import pandas as pd
        import numpy as np
        import networkx as nx
        import seaborn as sns
        import scipy as sc
        import random
        from scipy.signal import *
        from numpy.linalg import *
        from sklearn.decomposition import *
        from sklearn.preprocessing import *
        import librosa
        import abel
        #plotting parameters
        %matplotlib inline
        sns.set(style="whitegrid", color_codes=True, context='paper')
In [221]: from matplotlib import rcParams
          rcParams['font.family'] = 'serif'
          rcParams['font.sans-serif'] = ['CMU Serif']
In [3]: import matplotlib.pyplot as plt
In [222]: plt.rc('axes', grid=False, titlesize='large', labelsize='medium', labelwes
          plt.rc('lines', linewidth=4)
          plt.rc('figure', figsize = (8,6),titlesize='large',titleweight='black')
          plt.rc('font', weight='medium', size=9)
          plt.rc('grid',linewidth=3)
In [5]: sns.palplot(sns.cubehelix_palette(10, hue=0.3, reverse=True, rot=-0.55, dark
```

```
In [9]: months = ['Nov98', 'Dec98', 'jan_99', 'feb_99', 'mar_99', 'apr_99', 'may_99'
                  'nov_99', 'dec_99', 'jan_2k', 'feb_2k', 'mar_2k', 'apr_2k', 'may_
                  'nov_2k', 'dec_2k', 'jan_2k1', 'feb_2k1', 'mar_2k1', 'apr_2k1', '
                  'oct_2k1', 'nov_2k1', 'dec_2k1', 'jan_2k2', 'feb_2k2', 'mar_2k2',
In [10]: df_98 = data[data.year==years[0]]
         df_99 = data[data.year==years[1]]
         df_2k = data[data.year==years[2]]
         df_2k1 = data[data.year==years[3]]
         df_2k2 = data[data.year==years[4]]
In [11]: def create_graph(df):
             tmp = df.values[:, 1:3]
             G= nx.Graph()
             G = nx.from\_edgelist(tmp)
             return G
In [12]: Gt0 = create_graph(df_98)
         Gt1 = create_graph(df_99)
         Gt2 = create_graph(df_2k)
         Gt3 = create_graph(df_2k1)
         Gt4 = create\_graph(df\_2k2)
In [13]: nov_98 = df_98[df_98.month==11]
         dec_98 = df_98[df_98.month == 12]
         G_nov98 = create_graph(nov_98)
         G_dec98 = create_graph(dec_98)
In [14]: jan_99=df_99[df_99.month==1]
         feb_99=df_99[df_99.month==2]
         mar_99=df_99[df_99.month==3]
         apr_99=df_99[df_99.month==4]
         may_99=df_99[df_99.month==5]
         jun_99=df_99[df_99.month==6]
         jul_99=df_99[df_99.month==7]
         aug_99=df_99[df_99.month==8]
         sep_{99}=df_{99}[df_{99.month}=9]
         oct_99=df_99[df_99.month==10]
         nov_99=df_99[df_99.month==11]
         dec_99=df_99[df_99.month==12]
         G_jan_99=create_graph(jan_99)
         G_feb_99=create_graph(feb_99)
         G_mar_99=create_graph(mar_99)
         G_apr_99=create_graph(apr_99)
         G_may_99=create_graph (may_99)
```

```
G_jul_99=create_graph(jul_99)
         G_aug_99=create_graph(aug_99)
         G_sep_99=create_graph(sep_99)
         G oct 99=create graph(oct 99)
         G_nov_99=create_graph(nov_99)
         G dec 99=create graph (dec 99)
In [15]: jan_2k=df_2k[df_2k.month==1]
         feb_2k=df_2k[df_2k.month==2]
         mar_2k=df_2k[df_2k.month==3]
         apr_2k=df_2k[df_2k.month==4]
         may_2k=df_2k[df_2k.month==5]
         jun_2k=df_2k[df_2k.month==6]
         jul_2k=df_2k[df_2k.month==7]
         aug_2k=df_2k[df_2k.month==8]
         sep_2k=df_2k[df_2k.month==9]
         oct_2k=df_2k[df_2k.month==10]
         nov_2k=df_2k[df_2k.month==11]
         dec_2k=df_2k[df_2k.month==12]
         G_jan_2k=create_graph(jan_2k)
         G_feb_2k=create_graph(feb_2k)
         G_mar_2k=create_graph(mar_2k)
         G_apr_2k=create_graph(apr_2k)
         G_may_2k=create_graph(may_2k)
         G_jun_2k=create_graph(jun_2k)
         G_jul_2k=create_graph(jul_2k)
         G_aug_2k=create_graph(aug_2k)
         G_sep_2k=create_graph(sep_2k)
         G_oct_2k=create_graph(oct_2k)
         G_nov_2k=create_graph(nov_2k)
         G_dec_2k=create_graph(dec_2k)
In [16]: jan_2k1=df_2k1[df_2k1.month==1]
         feb_2k1=df_2k1[df_2k1.month==2]
         mar_2k1=df_2k1[df_2k1.month==3]
         apr_2k1=df_2k1[df_2k1.month==4]
         may_2k1=df_2k1[df_2k1.month==5]
         jun_2k1=df_2k1[df_2k1.month==6]
         jul_2k1=df_2k1[df_2k1.month==7]
         aug_2k1=df_2k1[df_2k1.month==8]
         sep_2k1=df_2k1[df_2k1.month==9]
         oct 2k1=df 2k1[df 2k1.month==10]
         nov_2k1=df_2k1[df_2k1.month==11]
         dec_2k1=df_2k1[df_2k1.month==12]
```

G_jun_99=create_graph(jun_99)

```
G_feb_2k1=create_graph(feb_2k1)
         G_mar_2k1=create_graph(mar_2k1)
         G_apr_2k1=create_graph(apr_2k1)
         G_may_2k1=create_graph(may_2k1)
         G_jun_2k1=create_graph(jun_2k1)
         G_jul_2k1=create_graph(jul_2k1)
         G_aug_2k1=create_graph(aug_2k1)
         G_sep_2k1=create_graph(sep_2k1)
         G_oct_2k1=create_graph (oct_2k1)
         G_nov_2k1=create_graph (nov_2k1)
         G_dec_2k1=create_graph(dec_2k1)
In [17]: jan_2k2=df_2k2[df_2k2.month==1]
         feb_2k2=df_2k2[df_2k2.month==2]
         mar_2k2=df_2k2[df_2k2.month==3]
         apr_2k2=df_2k2[df_2k2.month==4]
         may_2k2=df_2k2[df_2k2.month==5]
         jun_2k2=df_2k2[df_2k2.month==6]
         jul_2k2=df_2k2[df_2k2.month==7]
         aug_2k2=df_2k2[df_2k2.month==8]
         sep_2k2=df_2k2[df_2k2.month==9]
         oct_2k2=df_2k2[df_2k2.month==10]
         nov_2k2=df_2k2[df_2k2.month==11]
         dec_2k2=df_2k2[df_2k2.month==12]
         G_jan_2k2=create_graph(jan_2k2)
         G_feb_2k2=create_graph(feb_2k2)
         G_mar_2k2=create_graph (mar_2k2)
         G_apr_2k2=create_graph(apr_2k2)
         G_may_2k2=create_graph (may_2k2)
         G_jun_2k2=create_graph(jun_2k2)
         G_jul_2k2=create_graph(jul_2k2)
         G_aug_2k2=create_graph(aug_2k2)
         G_sep_2k2=create_graph(sep_2k2)
         G_oct_2k2=create_graph (oct_2k2)
         G_nov_2k2=create_graph(nov_2k2)
         G_dec_2k2=create_graph (dec_2k2)
In [18]: all_year_G =tuple([Gt0,Gt1,Gt2,Gt3,Gt4])
In [19]: all_month_G = tuple([G_nov98,G_dec98,G_jan_99,G_feb_99,G_mar_99,G_apr_99,G_
                        G_nov_99, G_dec_99, G_jan_2k, G_feb_2k, G_mar_2k, G_apr_2k, G_may_
                        G_oct_2k, G_nov_2k, G_dec_2k, G_jan_2k1, G_feb_2k1, G_mar_2k1, G_a
                       G_aug_2k1, G_sep_2k1, G_oct_2k1, G_nov_2k1, G_dec_2k1, G_jan_2k2,
```

G_jan_2k1=create_graph(jan_2k1)

3 Network Visualisation

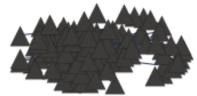
- Node Link Diagram
- Audio Waveform
- Matrix Visualisation

3.1 Yearly Networks

Network in 1998



Network in 2000



Network in 2002



Network in 1999

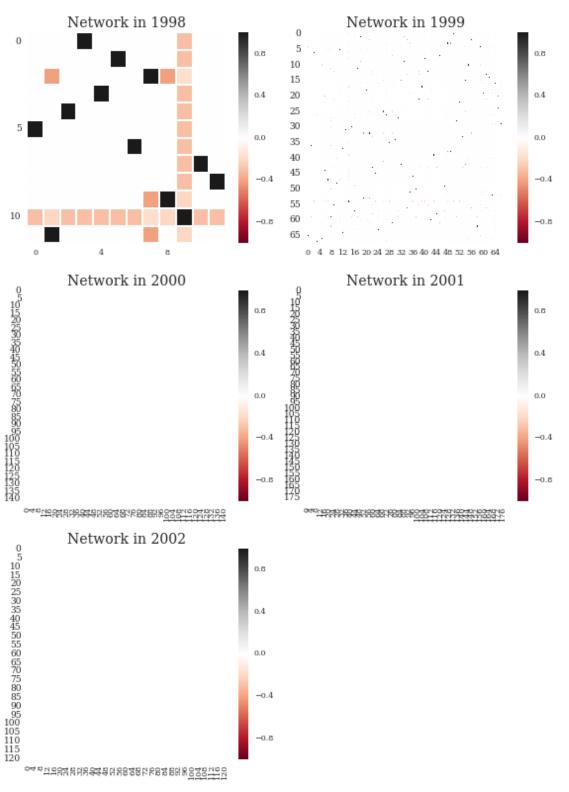


Network in 2001



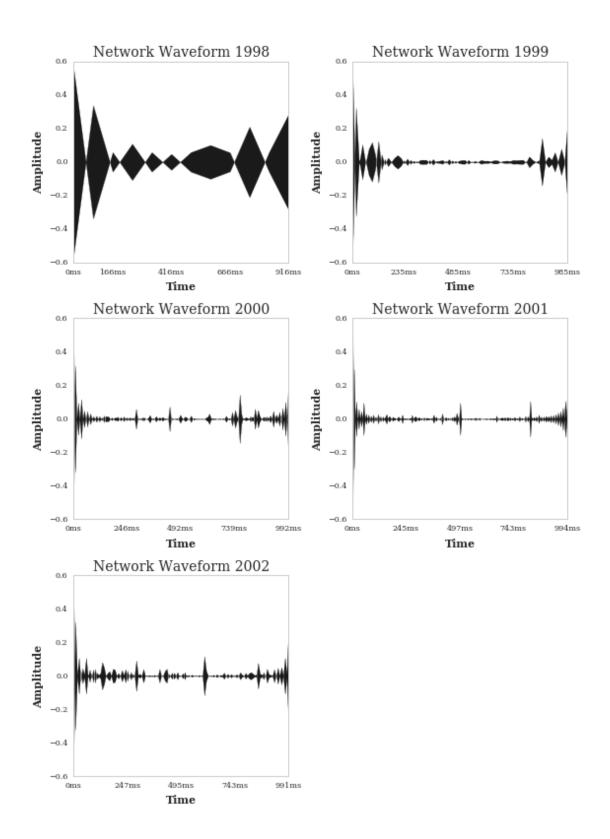
```
In [112]: plt.figure(figsize=(8,11))
    for i in range(len(all_year_G)):
        m = nx.normalized_laplacian_matrix(all_year_G[i]).todense()
        g = sns.clustermap(m)
        plt.close()
        ind = g.dendrogram_row.reordered_ind
        plt.subplot(3,2,i+1)
        sns.heatmap(m[ind][ind], cmap='RdGy', linewidths=1, xticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,yticklabels=4,ytickla
```

```
plt.title("Network in " + str(years[i]), fontsize=14)
plt.yticks(fontsize=9, rotation=360)
plt.tight_layout()
plt.savefig('images/yearly_net_mat.png')
```

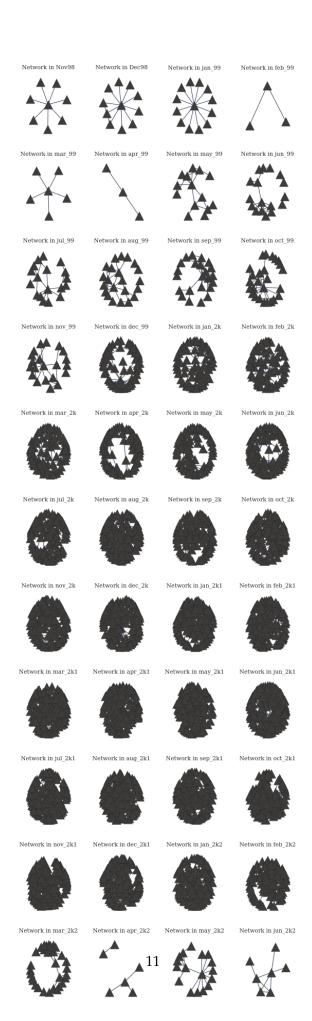


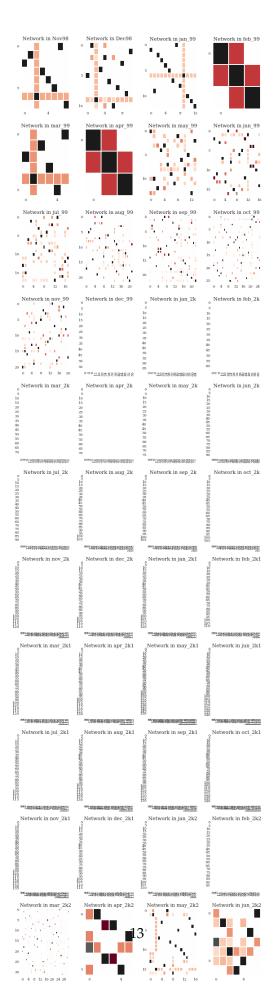
```
In [122]: plt.figure(figsize=(8,11))
    for i in range(len(all_year_G)):
        f = sc.fftpack.rfft(nx.normalized_laplacian_matrix(all_year_G[i]).too
        fs = f.shape[0]

        plt.subplot(3,2,i+1)
        librosa.display.waveplot(f, fs,color='k')
        plt.title("Network Waveform " + str(years[i]), fontsize=14)
        plt.xlabel("Time", fontsize=11)
        plt.ylabel("Amplitude", fontsize=11)
        plt.grid(False)
        plt.tight_layout()
        plt.savefig('images/yearly_net_audio.png')
```



3.2 Monthly networks



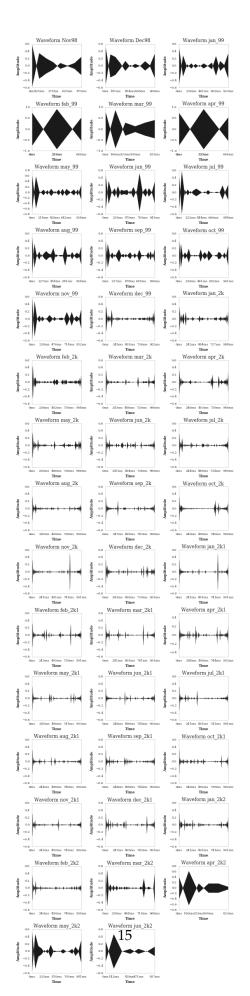


```
In [135]: plt.figure(figsize=(8,36))

for i in range(len(all_month_G)):
    f = sc.fftpack.rfft(nx.normalized_laplacian_matrix(all_month_G[i]).to
    fs = f.shape[0]

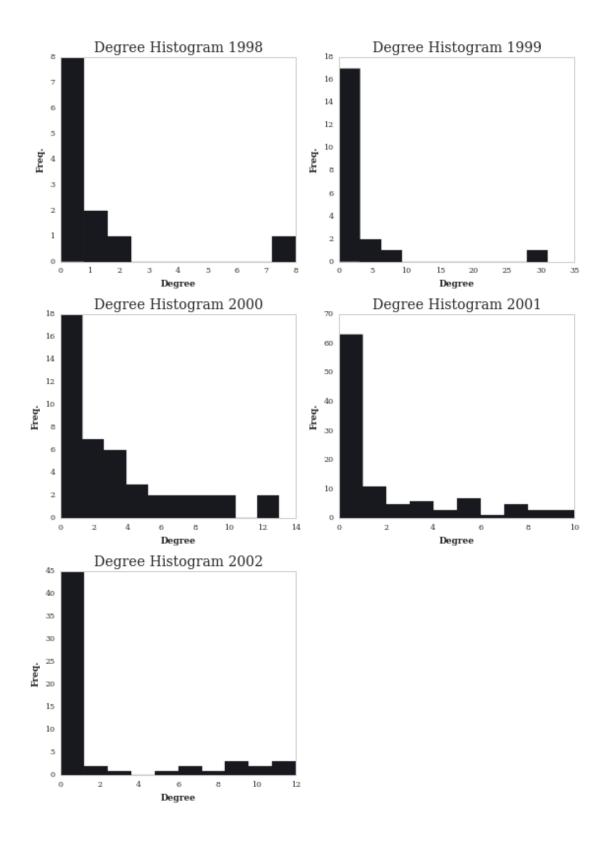
    plt.subplot(16, 3, i+1)
        librosa.display.waveplot(f, fs,color='k')
        plt.title("Waveform " + str(months[i]), fontsize=12)

        plt.xlabel("Time")
        plt.ylabel("Amplitude")
        plt.grid(False)
        plt.tight_layout()
        plt.savefig('images/mth_net_audio.png')
```

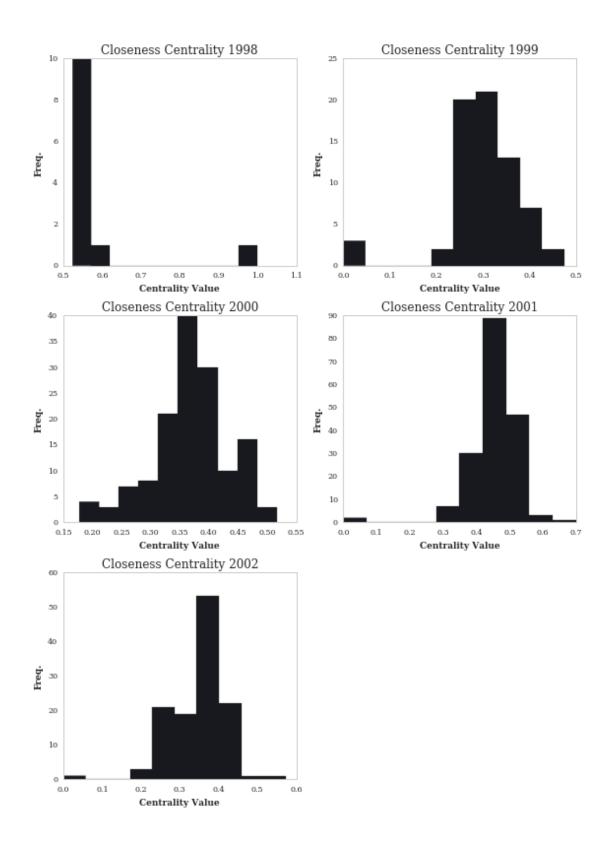


4 Exploratory Analysis

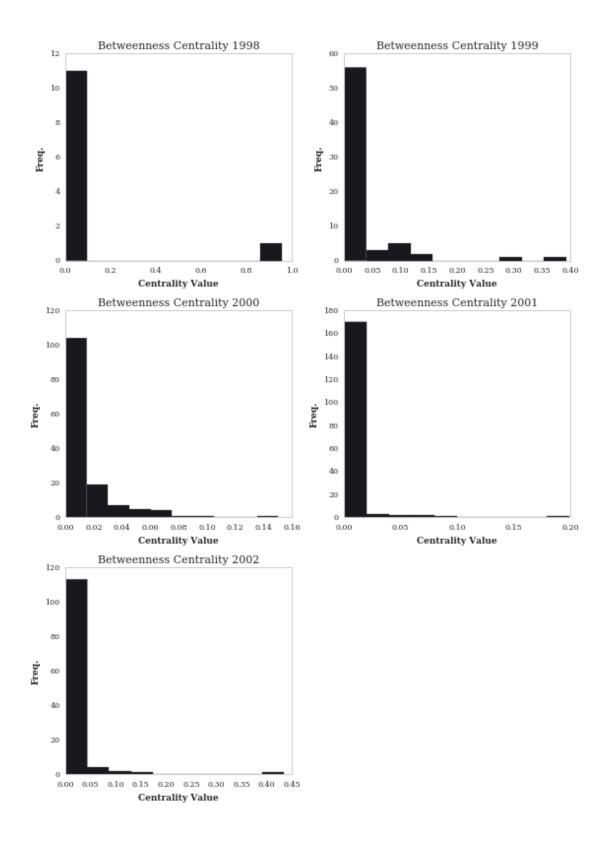
4.1 Yearly Networks



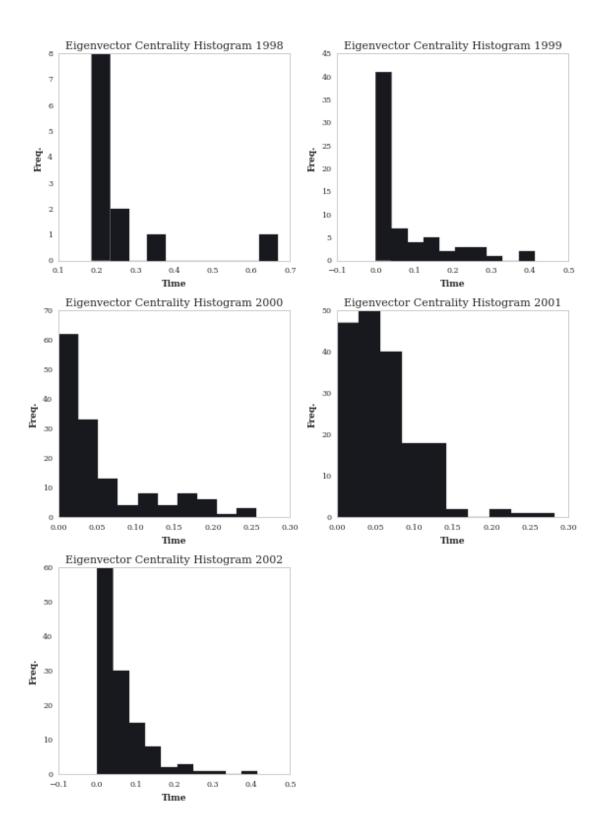
```
deg = nx.closeness_centrality(all_year_G[i]).values()
deg = sorted(deg)
plt.subplot(3, 2, i+1)
plt.hist(deg)
plt.title("Closeness Centrality " + str(years[i]), fontsize=12)
plt.xlabel("Centrality Value")
plt.ylabel("Freq.")
plt.grid(False)
plt.tight_layout()
plt.savefig('images/year_clohist.png')
```



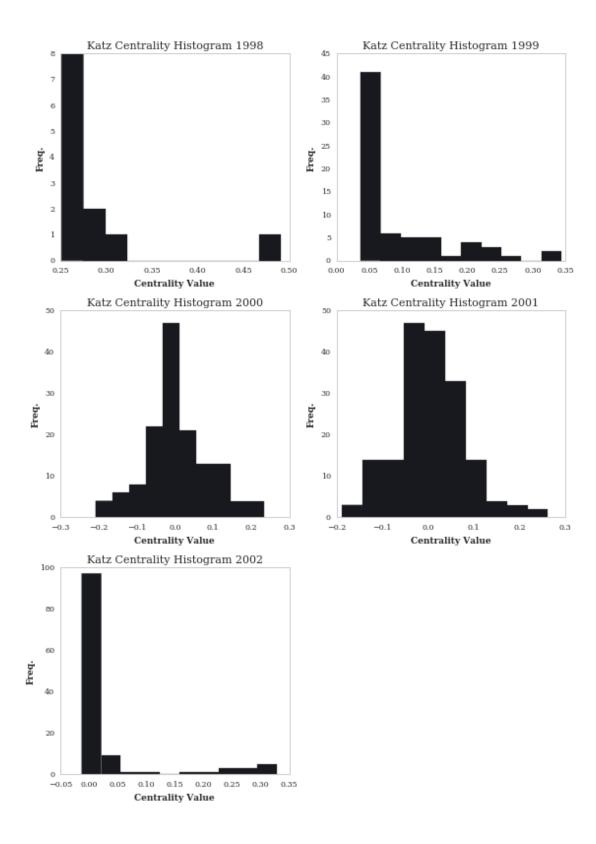
```
deg = nx.betweenness_centrality(all_year_G[i]).values()
  deg = sorted(deg)
  plt.subplot(3, 2, i+1)
  plt.hist(deg)
  plt.title("Betweenness Centrality " + str(years[i]), fontsize=11)
  plt.xlabel("Centrality Value")
  plt.ylabel("Freq.")
  plt.grid(False)
  plt.tight_layout()
plt.savefig('images/year_bethist.png')
```



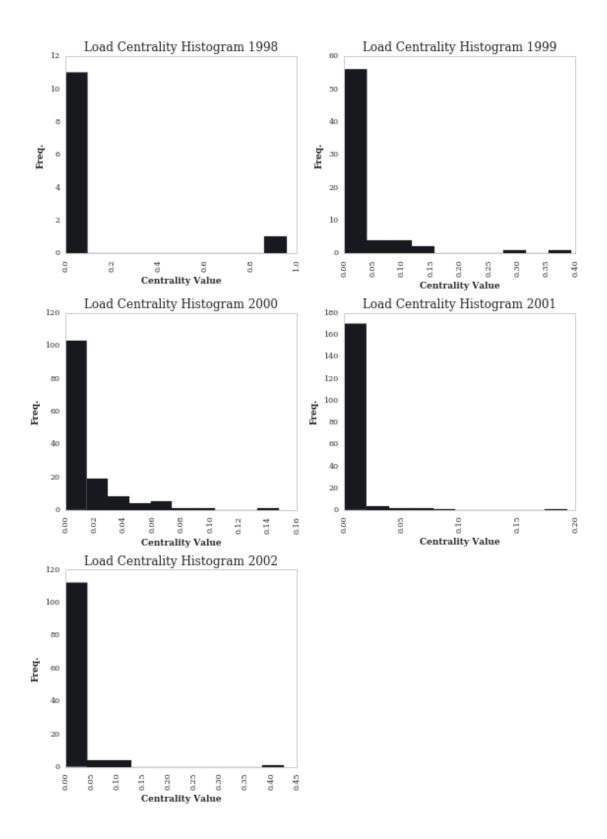
```
deg = nx.eigenvector_centrality_numpy(all_year_G[i]).values()
deg = sorted(deg)
plt.subplot(3, 2, i+1)
plt.hist(deg)
plt.title("Eigenvector Centrality Histogram " + str(years[i]), fonts:
plt.xlabel("Time")
plt.ylabel("Freq.")
plt.grid(False)
plt.tight_layout()
plt.savefig('images/year_eighist.png')
```



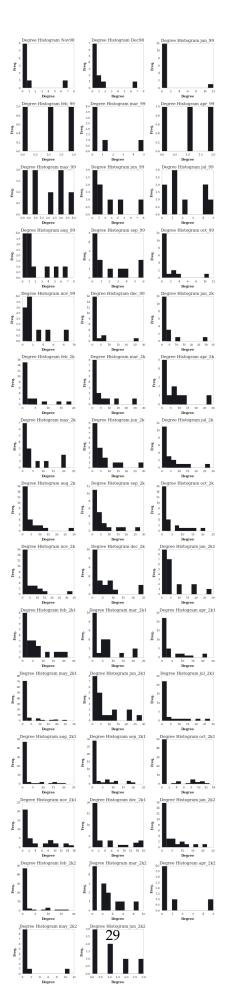
```
deg = nx.katz_centrality_numpy(all_year_G[i]).values()
deg = sorted(deg)
plt.subplot(3, 2, i+1)
plt.hist(deg)
plt.title("Katz Centrality Histogram " + str(years[i]), fontsize=11)
plt.xlabel("Centrality Value")
plt.ylabel("Freq.")
plt.grid(False)
plt.tight_layout()
plt.savefig('images/year_katzhist.png')
```

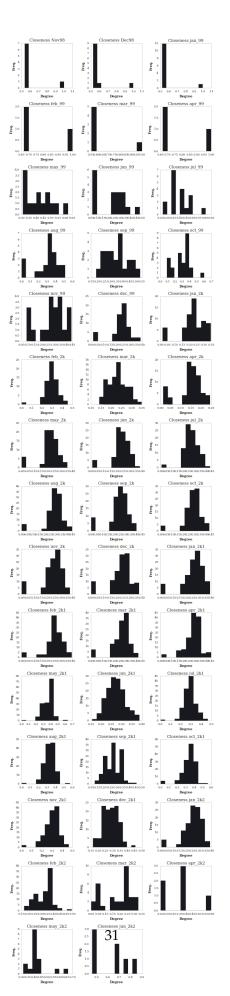


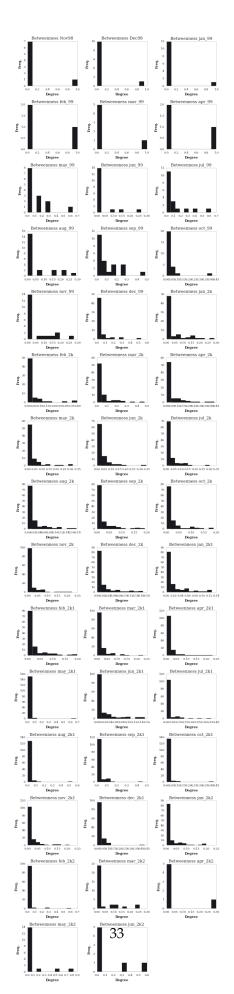
```
deg = nx.load_centrality(all_year_G[i]).values()
deg = sorted(deg)
plt.subplot(3, 2, i+1)
plt.hist(deg)
plt.title("Load Centrality Histogram " + str(years[i]), fontsize=12)
plt.xticks(rotation=90)
plt.xlabel("Centrality Value")
plt.ylabel("Freq.")
plt.grid(False)
plt.tight_layout()
plt.savefig('images/year_loadhist.png')
```



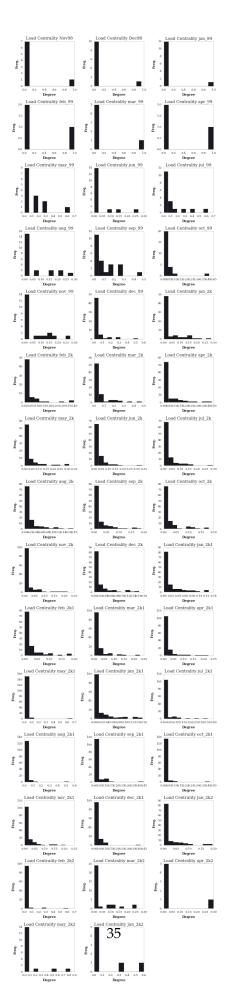
4.2 Monthly Networks



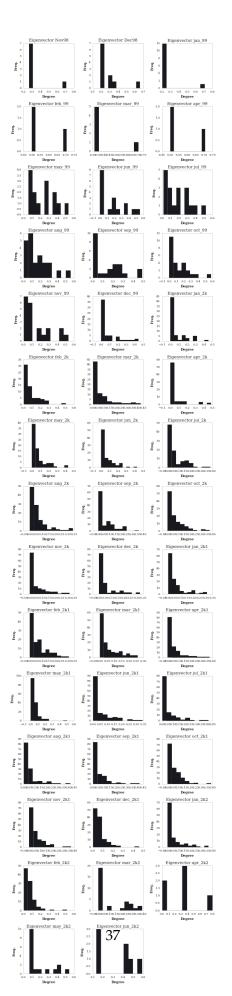


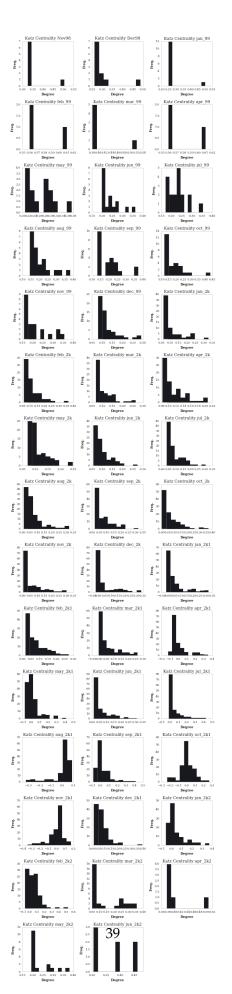


/home/arshad/anaconda3/lib/python3.5/site-packages/matplotlib/figure.py:1742: UserWarnings.warn("This figure includes Axes that are not "



/home/arshad/anaconda3/lib/python3.5/site-packages/matplotlib/figure.py:1742: UserWarnings.warn("This figure includes Axes that are not "





5 Attribute Analysis

5.1 Traditional Measures

5.1.1 Centrality

- Degree
- Betweeness
- Closeness
- Katz
- Load

5.1.2 Assortativity & Linear Algebra

- Density
- Average Clustering Coefficient
- Algebraic Connectivity

5.2 Complex Trace Attributes

- Instantaneous Amplitude
- Instantaneous Frequency
- Instantaneous Phase
- Derivative of Amplitude
- Second Derivative of Amplitude
- Power
- Cosine of Instantaneous Phase
- Amplitude weighted Frequency
- Amplitude weighted Phase
- Power Spectral Density

5.3 Matrix

- Resistance Distance
- Stationarity Ratio
- Subgraph Stationarity
- Curvature
- Norm of Abel Transform

5.4 Matrix Decomposition

- KLPCA Ratio Change
- Norm NMF Ratio Change

5.5 Music Attributes

- Zero Crossing Rate
- Spectral Centroid

5.6 Aggregation Measures

- Persistence
- Emergence
- NRMS of Attributes

```
In [38]: def calc_seisatt(net):
             M = nx.normalized laplacian matrix(net).todense()
             Ht = hilbert(M)
             rHt = np.real(Ht)
             iHt = np.imag(Ht)
             #Basic attributes IA, IP, IF
             IA = np.real(np.nan_to_num(np.sqrt(np.dot(rHt,rHt)+np.dot(iHt,iHt))))
             IP = np.real(np.nan_to_num(np.arctan(iHt/rHt)))
             IF,_ = np.real(np.nan_to_num(np.asarray(np.gradient(IP))))
             P = np.square(IA)
             #Derivatives
             dIA,_ = np.nan_to_num(np.asarray(np.gradient(IA)))
             d2IA,_ = np.nan_to_num(np.asarray(np.gradient(dIA)))
             IAcc,_ =np.nan_to_num(np.asarray(np.gradient(IF)))
             #Derived Attributes
             cosIP = np.cos(IP)
             IA_weit_IF = IA * IF
             IA\_weit\_IP = IA*IP
             return [IA, IP, IF, P, dIA, d2IA, IAcc, cosIP, IA_weit_IF, IA_weit_IP]
In [39]: def stationarity_ratio(G):
             #stationarity ratio with laplian
             L = nx.normalized_laplacian_matrix(G).todense()
             U = eigvals(L)
             C = np.cov(L)
             CF = np.dot(L, np.dot(np.dot(U.T, C), U))
             r = np.linalg.norm(np.diag(CF))/np.linalg.norm(CF)
             return r
In [40]: #cite:`klein1993resistance`
         def resistance_distance(net):
             M = nx.normalized_laplacian_matrix(net).todense()
             pseudo = pinv(M)
```

```
N = M.shape[0]
             d = np.diag(pseudo)
             rd = np.kron(d, np.ones((N, 1))).T+np.kron(d, np.ones((N, 1))).T - pseudo
             return rd.mean()
In [41]: def curvature(net):
             from skimage.feature import hessian_matrix, hessian_matrix_det, hessian_matrix_det,
             M = nx.normalized_laplacian_matrix(net).todense()
             M = np.float64(M)
             fx, fy = np.gradient(M)
             Hxx, Hxy, Hyy = hessian_matrix(M)
             K = np.divide((np.dot(Hxx, Hxy) -np.dot(Hxy, Hxy)), \
                            (1+np.dot(fx,fx)+np.dot(fy,fy))
             He1,_ = hessian_matrix_eigvals(Hxx, Hxy, Hyy)
             mean_curv = np.trace(He1)
             return mean_curv
In [42]: def kpca_ratio(net):
             from sklearn.decomposition import KernelPCA
             kpca = KernelPCA(n_components=3, kernel='rbf')
             m = nx.normalized_laplacian_matrix(net).todense()
             X_kpca = kpca.fit_transform(m)
             pc1 = X_kpca[:, 0]
             pc2 = X_kpca[:,1]
             pc3 = X_kpca[:,2]
             kpca\_rat = norm(pc1-pc3/(pc1-pc2))
             return kpca_rat
In [43]: def kpca_att(net):
             kpca\_chg = []
             for i in range(len(net)-1):
                 x = int(i)
                 y = 1+x
                 kpcar1= kpca_ratio(net[x])
                 kpcar2= kpca_ratio(net[y])
                 kpca_chg.append(np.divide(kpcar2,kpcar1))
             kpca_chgpad = np.zeros(len(kpca_chg)+1);
             kpca_chgpad[1:] = kpca_chg
             return kpca_chgpad
In [44]: def nmf_att(net):
             nmf_chg = []
             nmf = NMF(init='nndsvda', solver='cd', random_state=0, l1_ratio=1)
             for i in range(len(net)-1):
                 x = int(i)
```

```
y = 1+x
                 m1= abs(nx.normalized_laplacian_matrix(net[x]).todense())
                 m2= abs(nx.normalized_laplacian_matrix(net[y]).todense())
                 nmf1 = norm(nmf.fit_transform(m1))
                 nmf2 = norm(nmf.fit transform(m2))
                 nmf_chg.append(np.divide(nmf2,nmf1))
             nmf_chgpad = np.zeros(len(nmf_chg)+1);
             nmf_chgpad[1:] = nmf_chg
             return nmf_chgpad
In [45]: def pad_shape(x,ref,offset=0):
             result = np.zeros_like(ref)
             result[0:x.shape[0]+0,0:x.shape[1]+0] = x
             return result
         def subgraph_stat(net1, net2):
             net1_int_net2 = net1.copy()
             net1_int_net2.remove_nodes_from(n for n in net1 if n not in net2)
             net1_u_net2 = nx.disjoint_union(net1, net2)
             int_adjmat = nx.normalized_laplacian_matrix(net1_int_net2).todense()
             uni_adjmat = nx.normalized_laplacian_matrix(net1_u_net2).todense()
             int_adjmat_pad = pad_shape(int_adjmat,uni_adjmat)
             Ct = np.divide(norm(int_adjmat_pad), norm(uni_adjmat))
             return Ct
         def zeta(n):
             Ct_m = []
             for i in range(len(n)-1):
                 x = int(i)
                 y = x + 1
                 Ct_m.append(subgraph_stat(n[x],n[y]))
             Ct_m_pad = np.zeros(len(Ct_m)+1);
             Ct_m_pad[1:] = Ct_m
             return Ct_m_pad
In [46]: def music_att(n):
             #music attributes
             f = sc.fftpack.rfft(nx.normalized_laplacian_matrix(n).todense()).mean
             zcr = librosa.feature.zero_crossing_rate(f)[0,0]
             spc = librosa.feature.spectral_centroid(f)[0,0]
             return [zcr, spc]
In [47]: def norm_fabel(x):
```

```
import abel
             fabel_att =[]
             for n in x:
                 tmp = nx.normalized_laplacian_matrix(n).todense()
                 tmp2 = sc.fftpack.fft2(tmp)
                 fabel2 = abel.Transform(tmp2, direction='forward', method='direct')
                 mag = np.sqrt(np.square(np.real(fabel2)) + np.square(np.imag(fabel
                 fabel_att.append(norm(mag))
             return np.log10(fabel_att)
In [48]: def cal_avg_atts(x):
             #define attributes placeholders here
             deg_y = []
             bet_y = []
             clo_y = []
             load_y = []
             eig_y = []
             katz_y = []
             den_y = []
             algc_y = []
             cluscof_y = []
             IA_y = []
             IP_y = []
             IF_y = []
             P_y = []
             dIA_y = []
             d2IA_y = []
             IAcc_y = []
             cosIP_y = []
             IA\_weit\_IF\_y = []
             IA\_weit\_IP\_y = []
             psd_y = []
             rd_y = []
             zcr_y = []
             spc_y = []
             statr_y = []
             meank_y = []
             #matrix decompostion attributes
             zeta_y = zeta(x)
             nmf_ratio_y = nmf_att(x)
             kpca_chg_y= kpca_att(x)
             prop_members_chg_1_zeta = 1-zeta_y
             fabel = norm_fabel(x)
             for n in x:
```

```
deg = np.mean(sorted(set(nx.degree_centrality(n).values())))
   bet = np.mean(sorted(set(nx.betweenness_centrality(n).values())))
    clo = np.mean(sorted(set(nx.closeness_centrality(n).values())))
    katz = np.mean(sorted(set(nx.katz_centrality_numpy(n).values())))
    eig = np.mean(sorted(set(nx.eigenvector_centrality_numpy(n).values
    load = np.mean(sorted(set(nx.degree_centrality(n).values())))
    den = nx.density(n)
    algc = nx.algebraic_connectivity(n)
    clustcof = nx.average_clustering(n)
    #all network metrics
    deg_y.append(deg), bet_y.append(bet), clo_y.append(clo), load_y.ap
    den_y.append(den), algc_y.append(algc), cluscof_y.append(clustcome)
    #complex trace attributes
    IA, IP, IF, P, dIA, d2IA, IAcc, cosIP, IA_weit_IF, IA_weit_IP = calc_seis
    IA_y.append(IA.mean())
    IP_y.append(IP.mean())
    IF_y.append(IF.mean())
    P_y.append(P.mean())
    dIA_y.append(dIA.mean())
    d2IA_y.append(d2IA.mean())
    IAcc_y.append(IAcc.mean())
    cosIP_y.append(cosIP.mean())
    IA_weit_IF_y.append(IA_weit_IF.mean())
    IA_weit_IP_y.append(IA_weit_IP.mean())
   psd, _ = plt.psd(nx.laplacian_matrix(n).todense());
    plt.close()
    psd_y.append(psd.mean())
    zcr, spc = music_att(n)
    zcr_y.append(zcr)
    spc_y.append(np.log10(spc))
    #matrix attributes
    rdm = resistance_distance(n)
    rd_y.append(rdm)
    statrat = stationarity_ratio(n)
    statr_y.append(statrat)
    meank = curvature(n)
    meank_y.append(meank)
colnames = ['AvgDeg','AvgBet','AvgClo','AvgLoad','AvgKatz','AvgDensity
```

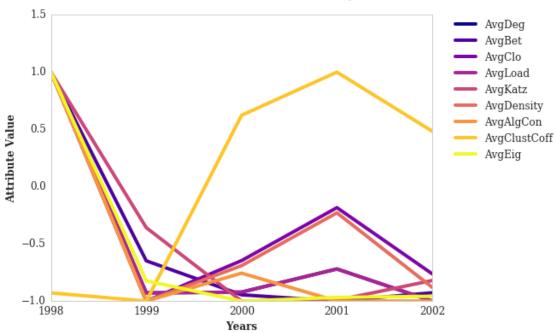
```
'AvgEig' ,'InstAmp','InstPhase','InstFreq','Power','dInstAr
                        'A_wt_IF','A_wt_IP','PowerSpecDen','ResDist','ZeroCrossRate
             attvol_y = pd.DataFrame([deg_y,bet_y, clo_y ,load_y ,katz_y ,den_y, al
                          P_y ,dIA_y ,d2IA_y,IAcc_y ,cosIP_y ,IA_weit_IF_y ,IA_weit
                          statr_y , meank_y]).T
             attvol_y.columns = colnames
             attvol_y['SubgraphStat']=zeta_y
             attvol_y['1-Zeta'] = prop_members_chg_1_zeta
             attvol_y['LogKPCARatioChg'] = np.log10(kpca_chg_y)
             attvol_y.LogKPCARatioChg[0]=0
             attvol_y['NormNMFRatioChg'] = nmf_ratio_y
             attvol_y['NormFAbel'] = fabel
             attvol_y_sc = attvol_y.apply(lambda x: minmax_scale(x, feature_range=
             return attvol_y_sc
In [49]: attvol_y = cal_avq_atts(all_year_G)
/home/arshad/anaconda3/lib/python3.5/site-packages/abel/transform.py:341: ComplexWa
  self.IM = self.IM.astype('float64')
In [50]: attvol_m = cal_avg_atts(all_month_G)
/home/arshad/anaconda3/lib/python3.5/site-packages/abel/transform.py:341: ComplexWa
  self.IM = self.IM.astype('float64')
```

6 Overview Plots

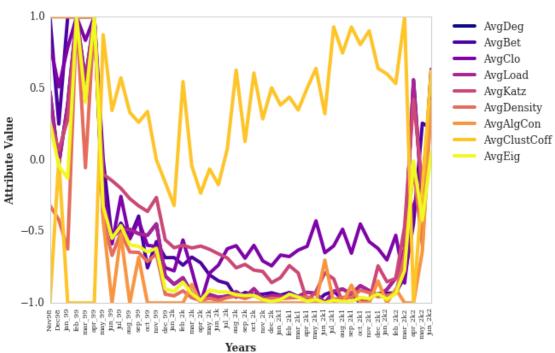
```
In [153]: attvol_y.iloc[:,:9].plot.line(use_index=True,cmap='plasma')
    plt.xlabel("Years", fontsize=12)
    plt.ylabel("Attribute Value", fontsize=12)
    plt.legend(fontsize=12, bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0
    plt.xticks(np.arange(len(years)), years,fontsize=12)
    plt.yticks(fontsize=12)
    plt.suptitle("Plot of Benchmark Measures over years", fontsize=16)

    attvol_m.iloc[:,:9].plot.line(use_index=True,cmap='plasma', rot=90)
    plt.xlabel("Years", fontsize=12)
    plt.ylabel("Attribute Value", fontsize=12)
    plt.legend(fontsize=12, bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0
    plt.xticks(np.arange(len(months)), months)
    plt.yticks(fontsize=12)
    plt.suptitle("Plot of Benchmark Measures over months", fontsize=16)
Out [153]: <matplotlib.text.Text at 0x7f625cd4bfd0>
```

Plot of Benchmark Measures over years

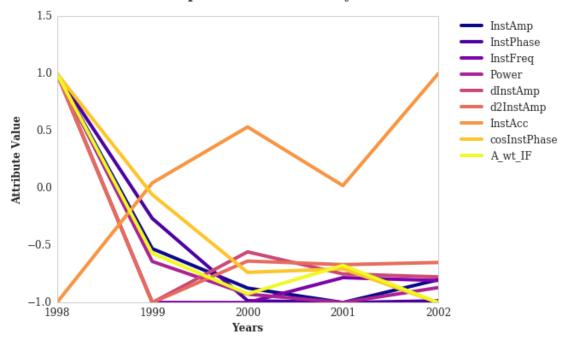


Plot of Benchmark Measures over months

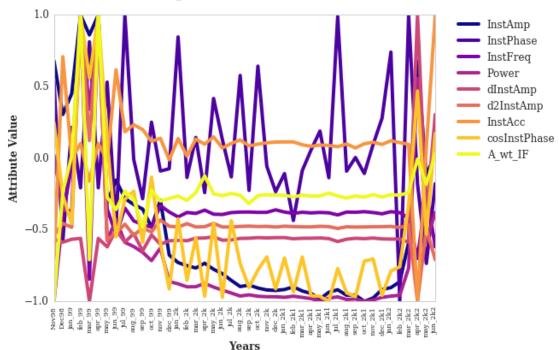


```
In [155]: attvol_y.iloc[:,9:18].plot.line(use_index=True,cmap='plasma')
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Attribute Value", fontsize=12)
          plt.legend(fontsize=12, bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0
          plt.xticks(np.arange(len(years)), years, fontsize=12)
          plt.yticks(fontsize=12)
          plt.suptitle("Plot of Complex Attributes over years", fontsize=16)
          plt.autoscale()
          attvol_m.iloc[:,9:18].plot.line(use_index=True,cmap='plasma', rot=90)
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Attribute Value", fontsize=12)
          plt.legend(fontsize=12, bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.
          plt.xticks(np.arange(len(months)), months)
          plt.yticks(fontsize=12)
          plt.suptitle("Plot of Complex Attributes over months", fontsize=16)
          plt.autoscale()
```

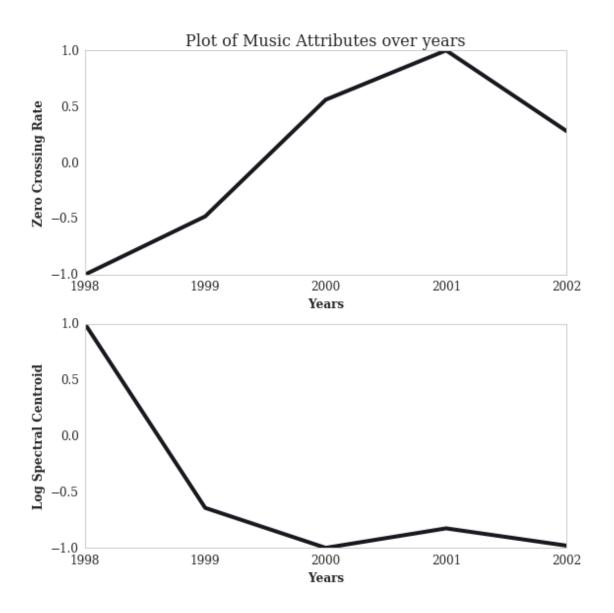
Plot of Complex Attributes over years



Plot of Complex Attributes over months



```
In [156]: plt.figure(figsize=(8,8))
          plt.subplot(2,1,1)
          plt.title("Plot of Music Attributes over years", fontsize=16)
          plt.plot(attvol_y.ZeroCrossRate)
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Zero Crossing Rate", fontsize=12)
          plt.xticks(np.arange(len(years)), years, fontsize=12)
          plt.yticks(fontsize=12)
          plt.autoscale()
          plt.subplot (2,1,2)
          plt.plot(attvol_y.LogSpecCentroid)
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Log Spectral Centroid", fontsize=12)
          plt.xticks(np.arange(len(years)), years, fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/musicatt_yrs.png')
```

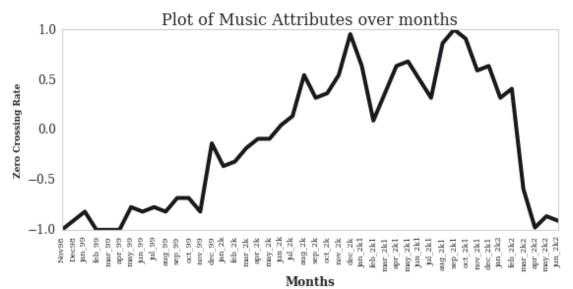


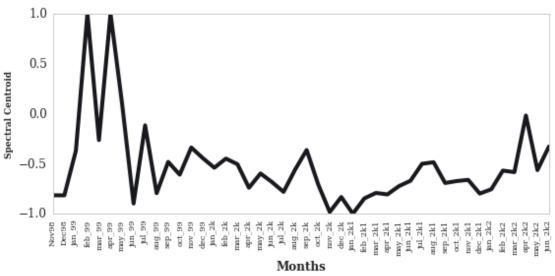
```
In [157]: plt.figure(figsize=(8,8))

    plt.subplot(2,1,1)
    plt.title("Plot of Music Attributes over months", fontsize=16)
    plt.plot(attvol_m.ZeroCrossRate.values)
    plt.xlabel("Months", fontsize=12)
    plt.ylabel("Zero Crossing Rate")
    plt.xticks(np.arange(len(months)), months, rotation=90)
    plt.yticks(fontsize=12)
    plt.autoscale()

plt.subplot(2,1,2)
    plt.plot(attvol_m.LogSpecCentroid.values)
```

```
plt.xlabel("Months", fontsize=12)
plt.ylabel("Spectral Centroid")
plt.xticks(np.arange(len(months)), months, rotation=90)
plt.yticks(fontsize=12)
plt.autoscale()
plt.tight_layout()
plt.savefig('images/musicatt_mth.png')
```

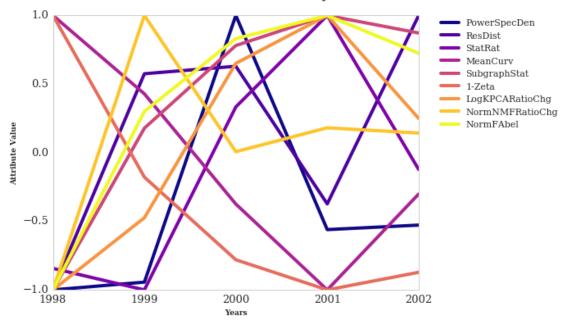


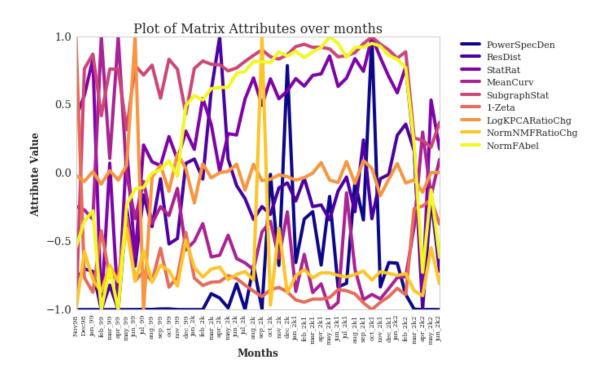


```
plt.yticks(fontsize=13)
plt.suptitle("Plot of Matrix Attributes over years", fontsize=16)
plt.autoscale()

attvol_m.drop(['ZeroCrossRate','LogSpecCentroid'], axis=1).iloc[:,19:].plt.xlabel("Months", fontsize=12)
plt.ylabel("Attribute Value", fontsize=12)
plt.legend(fontsize=11, bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0;
plt.xticks(np.arange(len(months)), months)
plt.yticks(fontsize=13)
plt.title("Plot of Matrix Attributes over months", fontsize=16)
plt.autoscale()
```

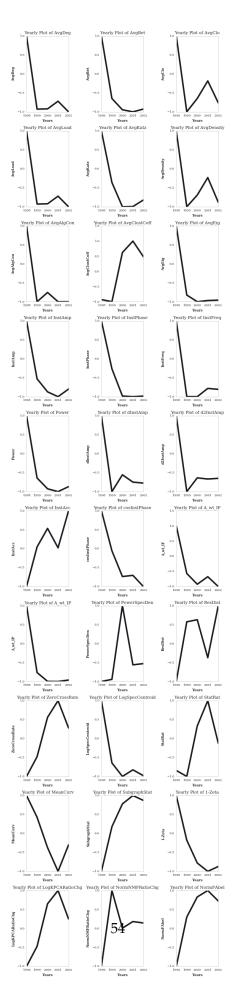
Plot of Matrix Attributes over years





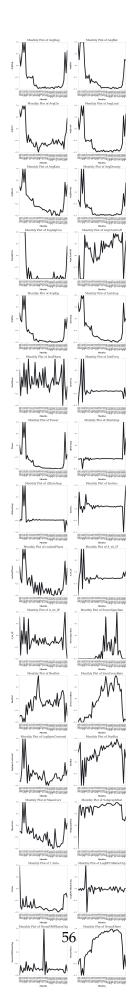
```
In [56]: all_cols = list(attvol_m.columns)
In [57]: attvol_y.shape
Out[57]: (5, 30)
In [162]: plt.figure(figsize=(8,38))

for i in range(len(all_cols)):
        plt.subplot(11,3,i+1)
            attvol_y.iloc[:,i].plot.line(use_index=True)
        plt.xlabel("Years")
        plt.ylabel(all_cols[i])
        plt.xticks(np.arange(len(years)), years)
        plt.title("Yearly Plot of " +all_cols[i], fontsize=11)
        plt.tight_layout()
        plt.savefig('images/avg_allatt_yrs.png')
```



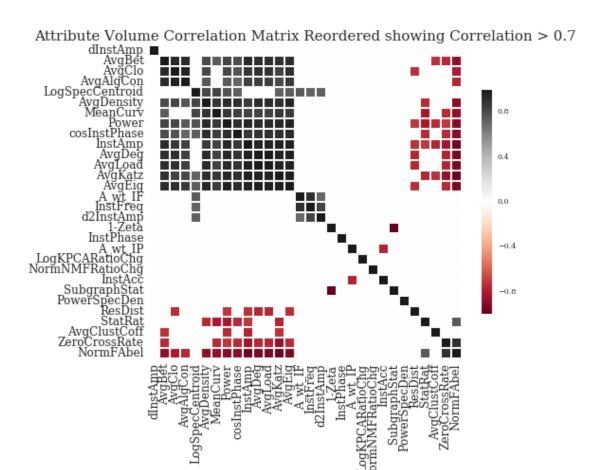
```
In [163]: plt.figure(figsize=(8,64))

for i in range(len(all_cols)):
    plt.subplot(15,2,i+1)
    attvol_m.iloc[:,i].plot.line(rot=90)
    plt.xlabel("Months")
    plt.ylabel(all_cols[i])
    plt.xticks(np.arange(len(months)), months)
    plt.title("Monthly Plot of " +all_cols[i], fontsize=13)
    plt.tight_layout()
    plt.savefig('images/avg_allatt_mth.png')
```



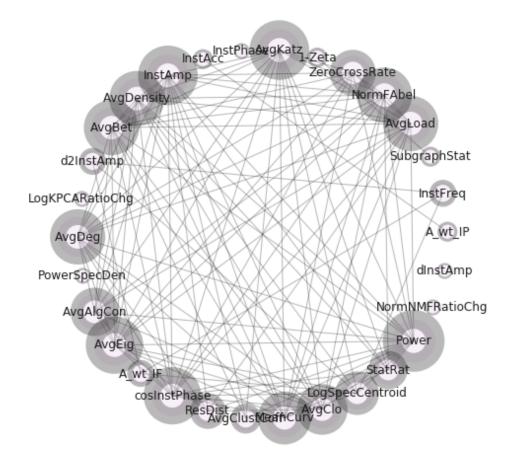
7 Correlation Analysis

7.1 Correlation > 0.7

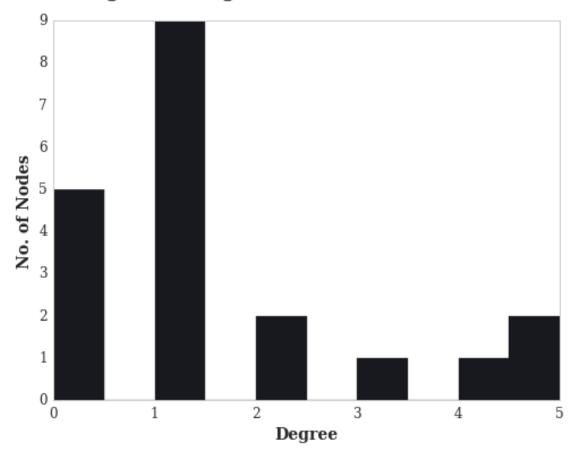


```
In [169]: names = corr m.index.values
          G_corr = nx.Graph(corr_m.values)
          pos=nx.fruchterman_reingold_layout(G_corr, iterations=1000, k=200)
In [172]: #ref: https://github.com/traims/correlation-based-networks/blob/master/ci
          components = nx.connected_components(G_corr)
          plt.figure(figsize=(8,8))
          plt.title("Correlation Network of Attributes", fontsize=16)
          plt.axis('off')
          for i in components:
              component = G_corr.subgraph(i)
              nx.draw_networkx(component, with_labels = True, node_size = [component
                  labels = dict([(x, names[x]) for x in component.nodes()]), pos=po
                              edge_color='k', node_color='#E9CFEC', linewidths=[cor
                               fontsize=11)
          plt.tight_layout()
          plt.savefig('images/corrnet.png')
```

Correlation Network of Attributes



Degree Historgram of Correlation Network



8 Regression Analysis for Feature Ranking

```
In [66]: from sklearn.metrics import mean_squared_error
         from sklearn import ensemble
In [67]: def RMS(x, axis=None):
             rms = np.sqrt(np.mean(np.square(x), axis=axis))
             return rms
In [68]: X= attvol_m.drop(['AvgDeg'],axis=1)
         y = attvol_m.AvgDeg
In [70]: X[:5]
Out [70]:
              AvgBet
                        AvgClo
                                  AvgLoad
                                          AvgKatz AvgDensity AvgAlgCon \
         0 \quad 1.000000 \quad 0.810523 \quad 0.475524 \quad 0.267944 \quad -0.318532
                                                                        1.0
            0.250903 0.509903 -0.027972 0.069435
                                                     -0.419219
                                                                        1.0
           1.000000 0.785809 0.388112 0.282044 -0.622808
                                                                        1.0
```

```
4 1.000000 0.835787 0.559441 0.423789
                                                    -0.054825
                                                                       1.0
            AvgClustCoff
                            AvgEig
                                     InstAmp InstPhase
                                                                      ResDist
         0
               -1.000000 0.262262 0.679834
                                               0.535455
                                                                    -0.774637
                                                            . . .
         1
               -0.000165 -0.044277 0.305093 -0.578269
                                                                    -0.701077
                                                            . . .
         2
               -1.000000 -0.127135 0.450396 0.214700
                                                                    -0.720470
                                                            . . .
               -1.000000 1.000000 1.000000 -0.207114
                                                                    -1.000000
                                                            . . .
               -1.000000 0.403511 0.860240
                                                                    -0.820587
                                               0.812100
                                                            . . .
                                                                                 1 - 2
            ZeroCrossRate LogSpecCentroid
                                            StatRat MeanCurv
                                                                SubgraphStat
                                                                   -1.000000 1.000
         0
                -1.000000
                                 -0.815713 0.407864 -0.242635
         1
                -0.909091
                                 -0.816057 0.570795 -0.269438
                                                                     0.766117 - 0.766
         2
                                                                     0.873376 - 0.873
                -0.818182
                                 -0.371199 0.836955 -0.337965
                -1.000000
                                  1.000000 -1.000000 1.000000
                                                                     0.421859 - 0.421
                -1.000000
                                 -0.264125 0.074177 0.108395
                                                                     0.763599 - 0.763
            LogKPCARatioChg NormNMFRatioChg NormFAbel
         0
                  -0.015620
                                   -1.000000 -0.543745
         1
                  -0.059551
                                   -0.573875 -0.352246
         2
                   0.012093
                                   -0.780922 -0.276008
         3
                  -0.079774
                                   -0.869723 \quad -1.000000
                   0.019315
                                   -0.673657 -0.702049
         [5 rows x 29 columns]
In [71]: X = X.astype(np.float32)
         offset = int(X.shape[0] * 0.5)
         X_train, y_train = X[:offset], y[:offset]
         X_test, y_test = X[offset:], y[offset:]
In [72]: params = {'n_estimators': 10, 'max_depth': 10, 'min_samples_split': 10,
                   'learning_rate': 0.1, 'loss': 'ls'}
         clf = ensemble.GradientBoostingRegressor(**params)
         clf.fit(X_train, y_train)
Out[72]: GradientBoostingRegressor(alpha=0.9, init=None, learning_rate=0.1, loss='1
                      max_depth=10, max_features=None, max_leaf_nodes=None,
                      min_samples_leaf=1, min_samples_split=10,
                      min_weight_fraction_leaf=0.0, n_estimators=10, presort='auto
                      random state=None, subsample=1.0, verbose=0, warm start=False
In [174]: feature_importance = clf.feature_importances_
          # make importances relative to max importance
          plt.figure(figsize=(8,11))
          feature_importance = 100.0 * (feature_importance / feature_importance.max
          sorted_idx = np.argsort(feature_importance)
          feature names = list(X)
```

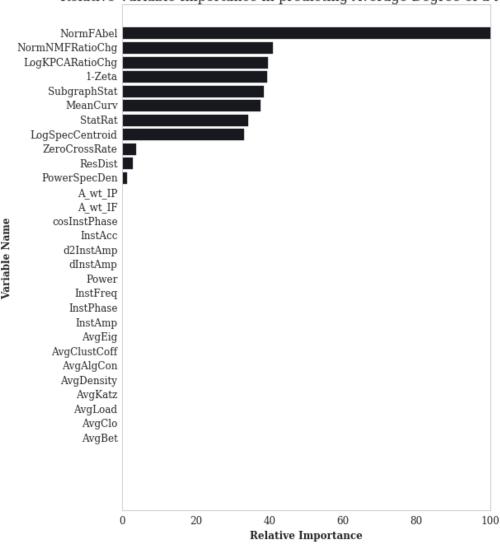
1.000000

1.0

3 1.000000 1.000000 1.000000 1.000000

```
pos = np.arange(sorted_idx.shape[0])
plt.barh(pos, feature_importance[sorted_idx], align='center')
plt.yticks(pos, feature_names, fontsize=12)
plt.ylabel("Variable Name", fontsize=12)
plt.xlabel('Relative Importance', fontsize=12)
plt.title('Relative Variable Importance in predicting Average Degree of a
plt.xticks(fontsize=12)
plt.savefig('images/feature_ranking.png')
```

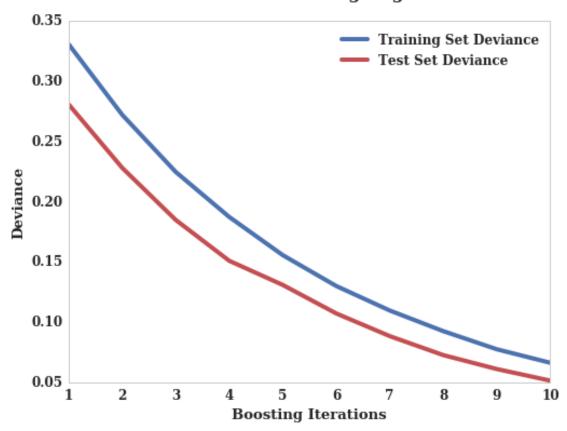
Relative Variable Importance in predicting Average Degree of a Network



```
In [74]: test_score = np.zeros((params['n_estimators'],), dtype=np.float64)

for i, y_pred in enumerate(clf.staged_predict(X_test)):
    test_score[i] = clf.loss_(y_test, y_pred)
```

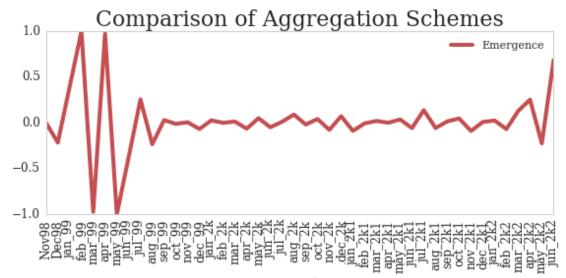
Deviance of Gradient Boosting Regression Model



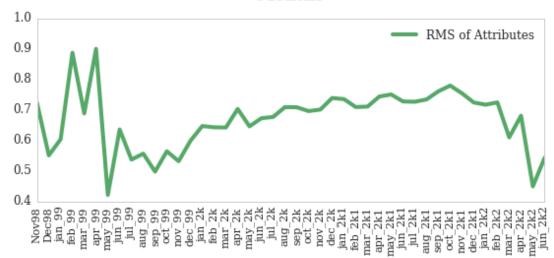
9 Aggregation Measures

```
In [77]: persistence_m = attvol_m.mean(axis=1)/(attvol_m.shape[0]-1)
         persistence_y = attvol_y.mean(axis=1)/(attvol_y.shape[0]-1)
In [78]: def emergence(per):
             tmp = np.asarray(per)
             emerg = []
             for i in range (len (tmp) - 1):
                 x = int(i)
                 y = x + 1
                  \#print(tmp[y], tmp[x])
                 if tmp[y] == tmp[x]:
                      emerg.append(0)
                 elif tmp[y] < (0) or tmp[x] < 0:
                      res = (tmp[y]-tmp[x])/(abs(tmp[y])+abs(tmp[x]))
                      emerg.append(res)
                 else:
                      res = (tmp[y]-tmp[x])/max([tmp[y],tmp[x]])
                      emerg.append(res)
             tmp2 = np.zeros(len(emerg)+1)
             tmp2[1:] = emerg
             return tmp2
         def NRMS(n):
             nrms = []
             for i in range (len (n) - 1):
                 x = int(i)
                 y = x + 1
                 a = n[x]
                 b = n[y]
                 nrms_{-} = np.divide((RMS(a-b)), (RMS(a)+RMS(b)))
                 nrms.append(nrms_)
             tmp2 = np.zeros(len(nrms)+1)
             tmp2[1:] = nrms
             return tmp2
In [79]: rms_m = attvol_m.apply(lambda x: RMS(x), axis=1)
         nrms_m = NRMS(rms_m)
         emerg_m = emergence(persistence_m)
In [80]: emerg_y = emergence(persistence_y)
In [181]: plt.figure(figsize=(8,11))
          plt.subplot(3,1,1)
```

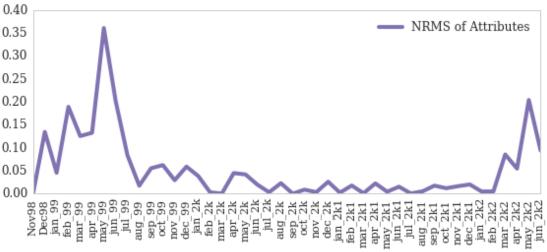
```
plt.plot(emerg_m,'r', label='Emergence')
plt.xticks(np.arange(len(months)), months,fontsize=12, rotation=90)
plt.yticks(fontsize=12)
plt.xlabel('Months', fontsize=16)
plt.legend(fontsize=11, loc=1)
plt.tight_layout()
plt.title("Comparison of Aggregation Schemes", fontsize=22)
plt.autoscale()
plt.subplot (3, 1, 2)
plt.plot(rms_m,'g',label='RMS of Attributes',)
plt.yticks(fontsize=12)
plt.xticks(np.arange(len(months)), months, fontsize=11, rotation=90)
plt.xlabel('Months', fontsize=16)
plt.legend(fontsize=12, loc=1)
plt.tight_layout()
plt.autoscale()
plt.subplot(3,1,3)
plt.plot(nrms_m,'m',label='NRMS of Attributes')
plt.xticks(np.arange(len(months)), months, fontsize=11, rotation=90)
plt.yticks(fontsize=12)
plt.xlabel('Months', fontsize=16)
plt.legend(fontsize=12, loc=1)
plt.autoscale()
plt.savefig('images/agg_comp.png')
```



Months



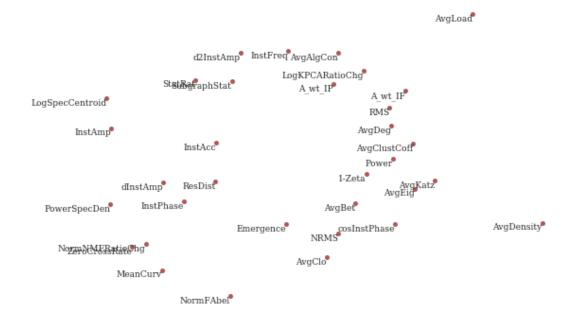
Months



Months

10 MDS and TSNE

Non-Metric Multi Dimensional Scaling of Attribute Volume



```
In [185]: tsne = TSNE(metric='euclidean', random_state=0)

y_tsne = tsne.fit_transform(final_attvol_m)

plt.axis('off')
plt.suptitle("TSNE Plot of Attribute Volume with Euclidean Distance", for
    for i, txt in enumerate(final_attvol_m.columns):
        plt.scatter(y_tsne[:, 1][i],y_tsne[:, 0][i], c='r')
        plt.annotate(txt, (y_tsne[:, 1][i],y_tsne[:, 0][i]), horizontalalignr

plt.autoscale()
plt.tight_layout()
plt.savefig('images/tsne_plot_euc.png')
```

TSNE Plot of Attribute Volume with Euclidean Distance

NormFAbel*

```
AvgKatz AvgClo

InstPhase AvgDeg
AvgEig wgDensity Emergence

Instelled InstAmp Emergence

InstAmp Power AlgCon A_wt_IF

LogSpecCentroid InstFreq RMS A_wt_IF

StatRat ResDist
GlnstAmp MeanGushustPhase 1-ZetavgClustCoff
NRMS

AvgBet

NormNMFRatioWhrspecDen AvgLoad
```

SubgraphStat

```
In [186]: tsne = TSNE(metric='canberra', random_state=0)

y_tsne = tsne.fit_transform(final_attvol_m)

plt.axis('off')
```

```
plt.suptitle("TSNE Plot of Attribute Volume with Canberra Distance", font
for i, txt in enumerate(final_attvol_m.columns):
    plt.scatter(y_tsne[:, 1][i],y_tsne[:, 0][i], c='r')
    plt.annotate(txt, (y_tsne[:, 1][i],y_tsne[:, 0][i]), horizontalalignr

plt.autoscale()
plt.tight_layout()
plt.savefig('images/tsne_plot.png')
```

TSNE Plot of Attribute Volume with Canberra Distance

AvgClo*

```
AvgClustCoff AvgEig

AvgDensity A wt IF

Power AvgKatz SubgraphStat InstFreeIrmFAbel

dInstAmpeanCurvAvgAlgCon

LogSpecedKFCoff AttioChg RMS InstAmpeanCurvAvgAlgCon

LogSpecedKFCoff AttioChg RMS InstAmpeanCurvAvgAlgCon

A wt IFmergence StatRat InstPhase

dZInstAmp NRMSInstPhase

ResDist InstAmp

AvgBet
```

AvgLoad

```
In [187]: tsne = TSNE(metric='correlation', random_state=0)

y_tsne = tsne.fit_transform(final_attvol_m)

plt.axis('off')

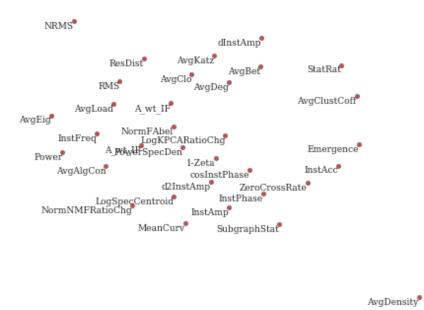
plt.suptitle("TSNE Plot of Attribute Volume with Correlation Distance", for i, txt in enumerate(final_attvol_m.columns):
    plt.scatter(y_tsne[:, 1][i],y_tsne[:, 0][i], c='r')
    plt.annotate(txt, (y_tsne[:, 1][i],y_tsne[:, 0][i]), horizontalalign

plt.autoscale()

plt.tight_layout()

plt.savefig('images/tsne_plot_corr.png')
```

TSNE Plot of Attribute Volume with Correlation Distance

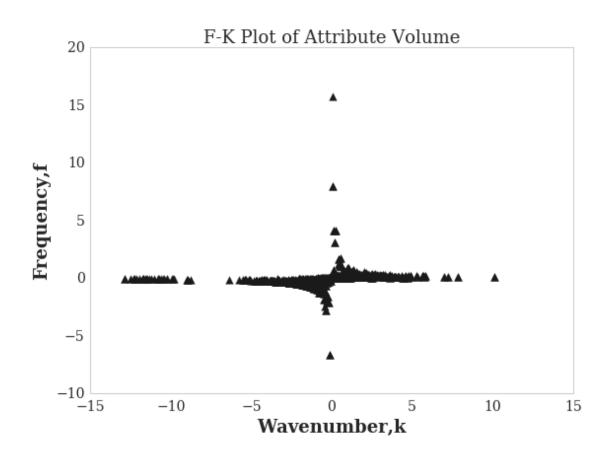


11 FK and Radon Plot

```
plt.ylabel("Frequency,f", fontsize=18)
plt.savefig('images/fkplot.png')
```

/home/arshad/anaconda3/lib/python3.5/site-packages/numpy/core/numeric.py:533: Complete return array(a, dtype, copy=False, order=order, subok=True)

/home/arshad/anaconda3/lib/python3.5/site-packages/matplotlib/figure.py:1742: UserWarnings.warn("This figure includes Axes that are not "



In [192]: sgram = radon(final_attvol_m.values)

/home/arshad/anaconda3/lib/python3.5/site-packages/skimage/transform/radon_transform warn('Radon transform: image must be zero outside the '

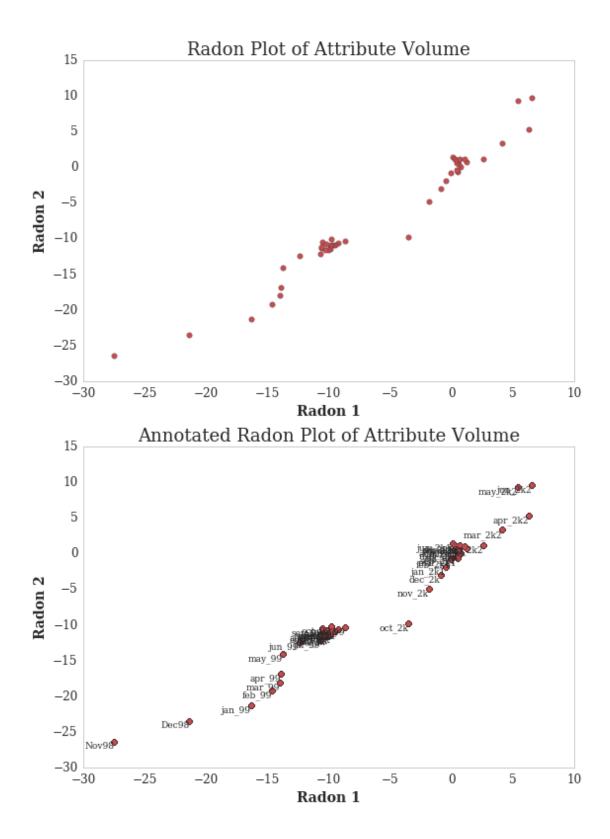
```
In [190]: plt.figure(figsize=(8,11))

    plt.subplot(211)
    plt.scatter(sgram[0],sgram[1], c='r', s=30)
    plt.xticks(fontsize=12)
    plt.yticks(fontsize=12)
    plt.xlabel("Radon 1", fontsize=14)
```

```
plt.ylabel("Radon 2", fontsize=14)
plt.title("Radon Plot of Attribute Volume", fontsize=18)
plt.tight_layout()

plt.subplot(212)
plt.title("Annotated Radon Plot of Attribute Volume", fontsize=18)

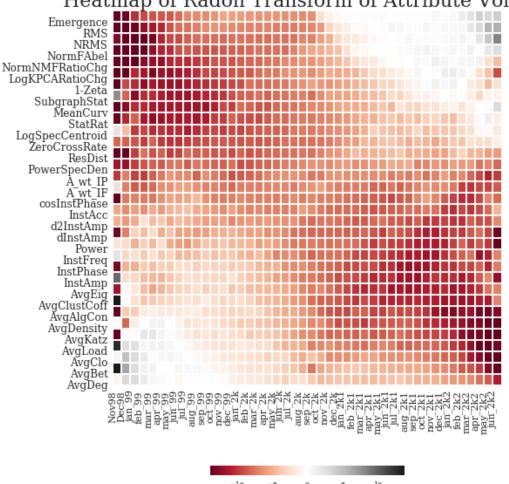
for i, txt in enumerate(months):
    plt.scatter(sgram[0], sgram[1], c='r', s=30)
    plt.annotate(txt, (sgram[0][i], sgram[1][i]), horizontalalignment='rig
plt.xticks(fontsize=12)
plt.yticks(fontsize=12)
plt.xlabel("Radon 1", fontsize=14)
plt.ylabel("Radon 2", fontsize=14)
plt.tight_layout()
plt.savefig('images/radonplot.png')
```

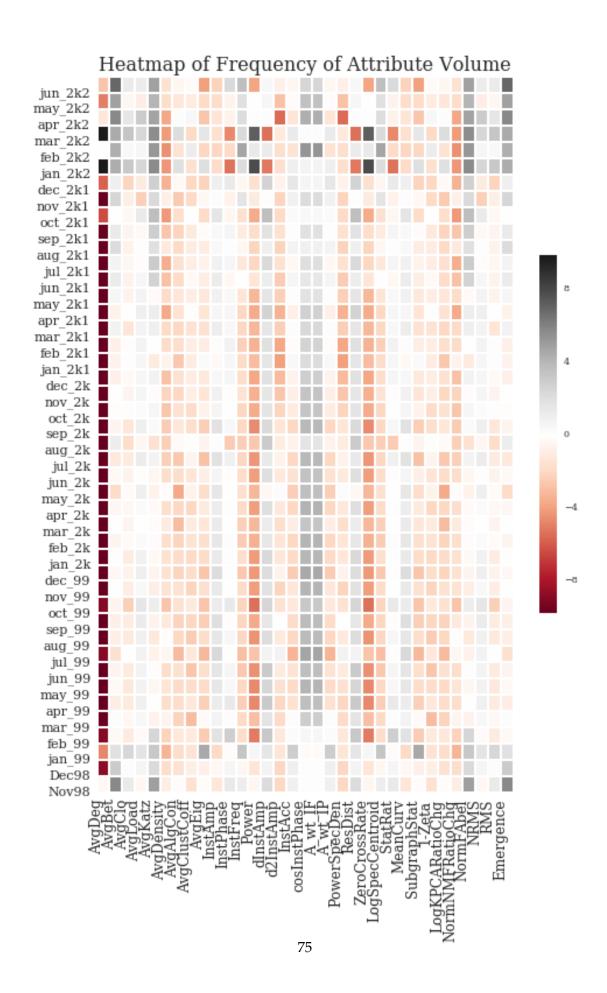


In [194]: plt.figure(figsize=(8,11))

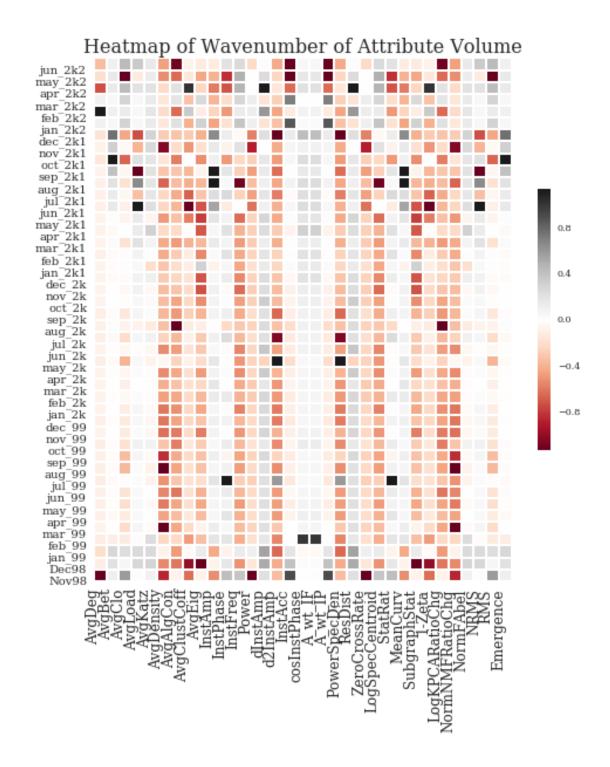
```
sns.heatmap(sgram, cmap='RdGy', robust=True,cbar_kws={'orientation':'hors
plt.xticks(np.arange(len(months)), months,fontsize=11, rotation=90);
plt.yticks(np.arange(len(final_attvol_m.columns)), final_attvol_m.columns
plt.title("Heatmap of Radon Transform of Attribute Volume", fontsize=16)
plt.savefig('images/radon_heat.png')
```







/home/arshad/anaconda3/lib/python3.5/site-packages/numpy/ma/core.py:3095: ComplexWa output = self._data.astype(newtype).view(type(self))

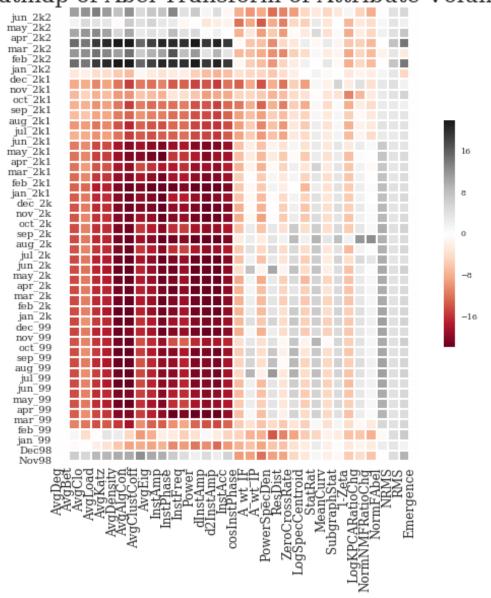


12 Exotic Transforms

In [96]: fabel = abel.Transform(final_attvol_m.values, direction='forward', method=

Out[199]: <matplotlib.text.Text at 0x7f625b656fd0>

Heatmap of Abel Transform of Attribute Volume



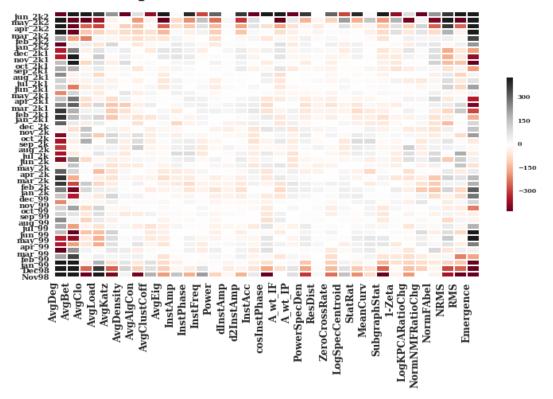
In [98]: H = sc.fftpack.fft2(fabel)

```
In [99]: sns.heatmap(H, cmap='RdGy', robust=True,cbar_kws={'orientation':'vertical'
    plt.yticks(np.arange(len(months)), months,fontsize=10, rotation=360);
    plt.xticks(np.arange(len(final_attvol_m.columns)), final_attvol_m.columns,
    plt.suptitle("Heatmap of Hankel Transform of Attribute Volume", fontsize=2
```

/home/arshad/anaconda3/lib/python3.5/site-packages/numpy/ma/core.py:3095: ComplexWa output = self._data.astype(newtype).view(type(self))

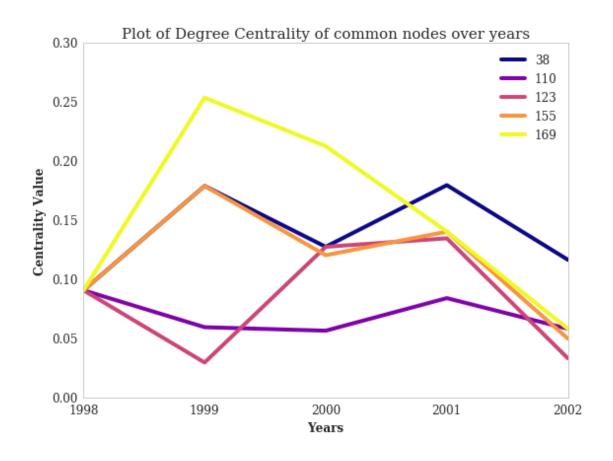
Out[99]: <matplotlib.text.Text at 0x7f625b115f98>

Heatmap of Hankel Transform of Attribute Volume

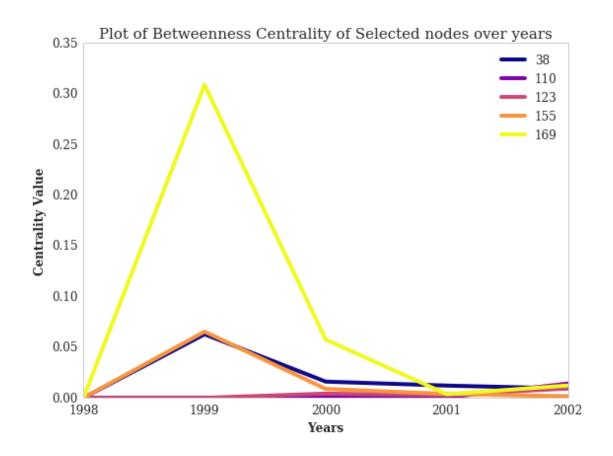


13 Which nodes are common to all years?

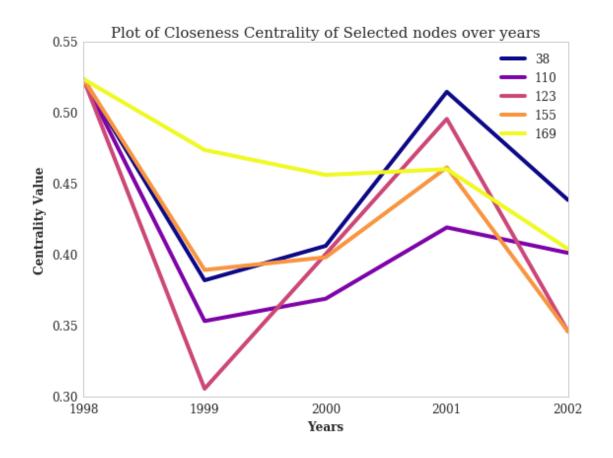
```
In [101]: def get_cent(net):
              degC = nx.degree_centrality(net)
              cloC = nx.closeness_centrality(net)
              betC = nx.betweenness_centrality(net)
              eigC = nx.eigenvector_centrality_numpy(net)
              katzC = nx.katz_centrality_numpy(net)
              loadC = nx.load centrality(net)
              return [degC,cloC,betC,eigC,katzC, loadC]
In [102]: degC0, cloC0, betC0, eigC0, katzC0, loadC0 = get_cent(Gt0)
          degC1, cloC1, betC1, eigC1, katzC1, loadC1 = get_cent(Gt1)
          degC2, cloC2, betC2, eigC2, katzC2, loadC2 = get_cent(Gt2)
          degC3, cloC3, betC3, eigC3, katzC3, loadC3 = get_cent(Gt3)
          degC4, cloC4, betC4, eigC4, katzC4, loadC4 = get_cent(Gt4)
In [236]: node_deg = pd.DataFrame.from_dict(degC0, orient='index').\
          join(pd.DataFrame.from_dict(degC1,orient='index'), rsuffix = 'Gt1', lsuff
          join(pd.DataFrame.from_dict(degC2,orient='index'), rsuffix = 'Gt2', lsuff
          join(pd.DataFrame.from_dict(degC3,orient='index'), rsuffix = 'Gt3', lsuff
          join(pd.DataFrame.from_dict(deqC4,orient='index'), rsuffix = 'Gt4', lsuff
          node_deg = node_deg.T
          node_deg.columns = years
          node_deg = node_deg.T
          node_deg.plot.line(figsize=(8,6), cmap='plasma')
          plt.title("Plot of Degree Centrality of common nodes over years", fontsiz
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_deg.png')
```



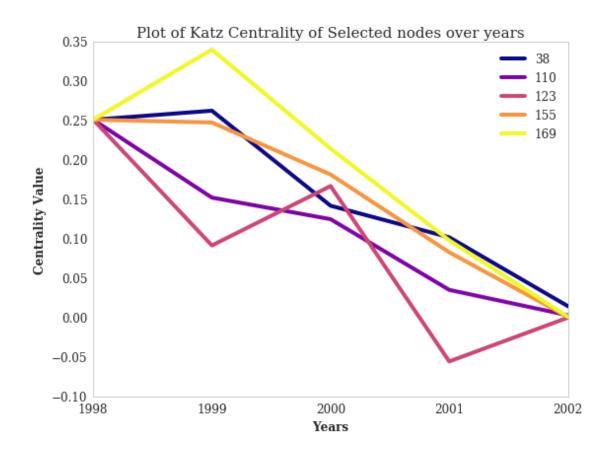
```
In [237]: node_bet = pd.DataFrame.from_dict(betC0, orient='index').\
          join(pd.DataFrame.from_dict(betC1,orient='index'), rsuffix = 'Gt1', lsuff
          join(pd.DataFrame.from_dict(betC2, orient='index'), rsuffix = 'Gt2', lsuff
          join(pd.DataFrame.from_dict(betC3,orient='index'), rsuffix = 'Gt3', lsuff
          join(pd.DataFrame.from_dict(betC4,orient='index'), rsuffix = 'Gt4', lsuff
          node_bet = node_bet.T
          node_bet.columns = years
          node_bet = node_bet.T
          node_bet.plot(cmap='plasma')
          plt.title("Plot of Betweenness Centrality of Selected nodes over years",
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_bet.png')
```



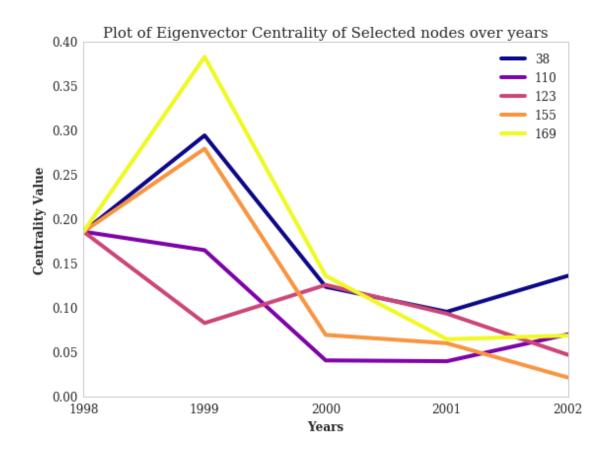
```
In [238]: node_clo = pd.DataFrame.from_dict(cloC0, orient='index').\
          join(pd.DataFrame.from_dict(cloC1,orient='index'), rsuffix = 'Gt1', lsuff
          join(pd.DataFrame.from_dict(cloC2,orient='index'), rsuffix = 'Gt2', lsuff
          join(pd.DataFrame.from_dict(cloC3,orient='index'), rsuffix = 'Gt3', lsuff
          join(pd.DataFrame.from_dict(cloC4,orient='index'), rsuffix = 'Gt4', lsuff
          node_clo = node_clo.T
          node_clo.columns = years
          node_clo = node_clo.T
          node_clo.plot(figsize=(8,6), cmap='plasma')
          plt.title("Plot of Closeness Centrality of Selected nodes over years", fo
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_clo.png')
```



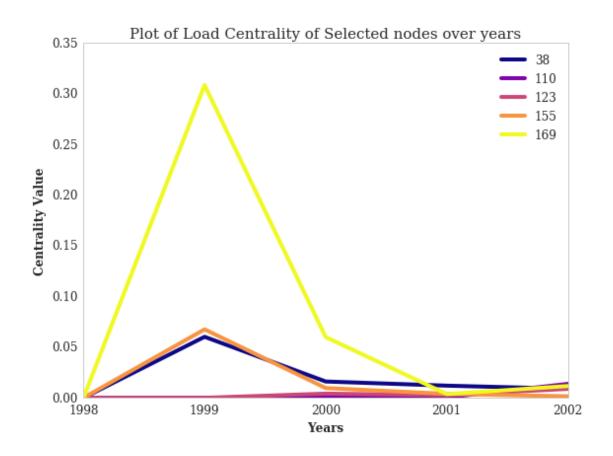
```
In [239]: node_katz = pd.DataFrame.from_dict(katzC0, orient='index').\
          join(pd.DataFrame.from_dict(katzC1,orient='index'), rsuffix = 'Gt1', lsuffix
          join(pd.DataFrame.from_dict(katzC2,orient='index'), rsuffix = 'Gt2', lsuffix
          join(pd.DataFrame.from_dict(katzC3, orient='index'), rsuffix = 'Gt3', lsuf
          join(pd.DataFrame.from_dict(katzC4, orient='index'), rsuffix = 'Gt4', lsuffix
          node_katz = node_katz.T
          node_katz.columns = years
          node_katz = node_katz.T
          node_katz.plot(figsize=(8,6), cmap='plasma')
          plt.title("Plot of Katz Centrality of Selected nodes over years", fontsiz
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_katz.png')
```



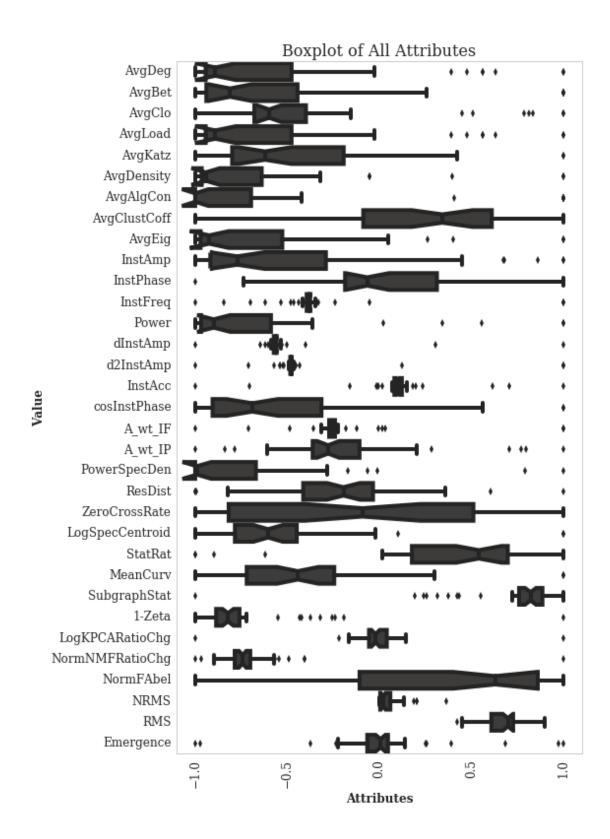
```
In [240]: node_eig = pd.DataFrame.from_dict(eigC0, orient='index').\
          join(pd.DataFrame.from_dict(eigC1,orient='index'), rsuffix = 'Gt1', lsuff
          join(pd.DataFrame.from_dict(eigC2,orient='index'), rsuffix = 'Gt2', lsuff
          join(pd.DataFrame.from_dict(eigC3,orient='index'), rsuffix = 'Gt3', lsuff
          join(pd.DataFrame.from_dict(eigC4,orient='index'), rsuffix = 'Gt4', lsuff
          node_eig = node_eig.T
          node_eig.columns = years
          node_eig = node_eig.T
          node_eig.plot(figsize=(8,6), cmap='plasma')
          plt.title("Plot of Eigenvector Centrality of Selected nodes over years",
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_eig.png')
```



```
In [241]: node_load = pd.DataFrame.from_dict(loadC0, orient='index').\
          join(pd.DataFrame.from_dict(loadC1, orient='index'), rsuffix = 'Gt1', lsuf
          join(pd.DataFrame.from_dict(loadC2, orient='index'), rsuffix = 'Gt2', lsuf
          join(pd.DataFrame.from_dict(loadC3, orient='index'), rsuffix = 'Gt3', lsuf
          join(pd.DataFrame.from_dict(loadC4, orient='index'), rsuffix = 'Gt4', lsuf
          node_load = node_load.T
          node_load.columns = years
          node_load = node_load.T
          node_load.plot(figsize=(8,6),cmap='plasma')
          plt.title("Plot of Load Centrality of Selected nodes over years", fontsiz
          plt.xlabel("Years", fontsize=12)
          plt.ylabel("Centrality Value", fontsize=12)
          plt.legend(fontsize=12, loc=1)
          plt.xticks(years,[i for i in years],fontsize=12)
          plt.yticks(fontsize=12)
          plt.tight_layout()
          plt.savefig('images/node_load.png')
```



/home/arshad/anaconda3/lib/python3.5/site-packages/seaborn/categorical.py:2171: Use warnings.warn(msg, UserWarning)



In []:

In []: