

```
1 import sys
2 from datetime import datetime
3
4 import numpy as np
5 import matplotlib.pyplot as plt
6 from matplotlib.collections import LineCollection
7 from six.moves.urllib.request import urlopen
8 from six.moves.urllib.parse import urlencode
9 from sklearn import cluster, covariance, manifold
10 import quandl
11 import pandas as pd
12 import datetime
13 from pandas_datareader import data as pdr
14 import fix_yahoo_finance as yf
15 from itertools import compress
16 from helper import read_txt
17
18 print(__doc__)
19
20 def retry(f, n_attempts=3):
21     "Wrapper function to retry function calls in case of
exceptions"
22     def wrapper(*args, **kwargs):
23         for i in range(n_attempts):
24             try:
25                 return f(*args, **kwargs)
26             except Exception:
27                 if i == n_attempts - 1:
28                     raise
29     return wrapper
30
31 def quotes_historical_google(symbol, start_date, end_date):
32
33     format = '%Y-%m-%d' # Formatting directives
34     start = start_date.strftime(format)
35     end = end_date.strftime(format)
36
37     yf.pdr_override() # <== that's all it takes :-)
38     data = pdr.get_data_yahoo(symbol, start=start, end=end)
39
40     return data
41
42
43
44 symbol_dict = {
45     'NYSE:TOT': 'Total',
46     'NYSE:XOM': 'Exxon',
47     'NYSE:CVX': 'Chevron',
48     'NYSE:COP': 'ConocoPhillips',
49     'NYSE:VLO': 'Valero Energy',
```

```
50 'NASDAQ:MSFT': 'Microsoft',
51 'NYSE:IBM': 'IBM',
52 'NYSE:TWX': 'Time Warner',
53 'NASDAQ:CMCSA': 'Comcast',
54 'NYSE:CVC': 'Cablevision',
55 'NASDAQ:YHOO': 'Yahoo',
56 'NASDAQ:DELL': 'Dell',
57 'NYSE:HPQ': 'HP',
58 'NASDAQ:AMZN': 'Amazon',
59 'NYSE:TM': 'Toyota',
60 'NYSE:CAJ': 'Canon',
61 'NYSE:SNE': 'Sony',
62 'NYSE:F': 'Ford',
63 'NYSE:HMC': 'Honda',
64 'NYSE:NAV': 'Navistar',
65 'NYSE:NOC': 'Northrop Grumman',
66 'NYSE:BA': 'Boeing',
67 'NYSE:KO': 'Coca Cola',
68 'NYSE:MMM': '3M',
69 'NYSE:MCD': 'McDonald\'s',
70 'NYSE:PEP': 'Pepsi',
71 'NYSE:K': 'Kellogg',
72 'NYSE:UN': 'Unilever',
73 'NASDAQ:MAR': 'Marriott',
74 'NYSE:PG': 'Procter Gamble',
75 'NYSE:CL': 'Colgate-Palmolive',
76 'NYSE:GE': 'General Electrics',
77 'NYSE:WFC': 'Wells Fargo',
78 'NYSE:JPM': 'JPMorgan Chase',
79 'NYSE:AIG': 'AIG',
80 'NYSE:AXP': 'American express',
81 'NYSE:BAC': 'Bank of America',
82 'NYSE:GS': 'Goldman Sachs',
83 'NASDAQ:AAPL': 'Apple',
84 'NYSE:SAP': 'SAP',
85 'NASDAQ:CSCO': 'Cisco',
86 'NASDAQ:TXN': 'Texas Instruments',
87 'NYSE:XRX': 'Xerox',
88 'NYSE:WMT': 'Wal-Mart',
89 'NYSE:HD': 'Home Depot',
90 'NYSE:GSK': 'GlaxoSmithKline',
91 'NYSE:PFE': 'Pfizer',
92 'NYSE:SNY': 'Sanofi-Aventis',
93 'NYSE:NVS': 'Novartis',
94 'NYSE:KMB': 'Kimberly-Clark',
95 'NYSE:R': 'Ryder',
96 'NYSE:GD': 'General Dynamics',
97 'NYSE:RTN': 'Raytheon',
98 'NYSE:CVS': 'CVS',
99 'NYSE:CAT': 'Caterpillar',
```

```
100     'NYSE:DD': 'DuPont de Nemours'
101     }
102
103     # #####
104     symbols, names = np.array(sorted(symbol_dict.items())).T
105
106     symbols_new = [] #symbols without "NYSE"
107     for n in range(len(symbols)):
108         symbols_new.append(symbols[n].split(":")[1])
109     print symbols_new
110
111
112     start = datetime.datetime(2017, 1, 1)
113     end = datetime.datetime(2018, 4, 1)
114
115     quotes = []
116
117     for symbol in symbols_new:
118         print('Fetching quote history for %r' % symbol)
119         quotes.append(quotes_historical_google(
120             symbol, start, end))
121     print type(quotes[0])
122
123     empty_dataframe = []
124     for q in quotes:
125         empty_dataframe.append(q.empty)
126
127     symbols_valid = []
128     name_valid = []
129
130     for i,j in enumerate(empty_dataframe):
131         if j == False:
132             symbols_valid.append(symbols_new[i])
133             name_valid.append(names[i])
134
135     print "drop the empty dataframe and then download the
136           stock price data...."
137     # #####
138     quotes = []
139
140     for symbol in symbols_valid:
141         print('Fetching quote history for %r' % symbol)
142         quotes.append(quotes_historical_google(
143             symbol, start, end))
144
145     print "complete fetching the stock price data ...."
146
```

```
147
148 close_prices = np.vstack([q.iloc[:,4] for q in quotes]) #4
    stands for Adj Close
149 open_prices = np.vstack([q.iloc[:,0] for q in quotes]) #0
    stands for Open
150
151 # The daily variations of the quotes are what carry most
    information
152 variation = close_prices - open_prices
153
154
155 # #####
    #####
156
157 # Learn a graphical structure from the correlations
158 edge_model = covariance.GraphLassoCV()
159
160 # standardize the time series: using correlations rather
    than covariance
161 # is more efficient for structure recovery
162 X = variation.copy().T #copy and then transpose
163 X /= X.std(axis=0) #the std of each column(stock), thus if
    there two stocks, then the X.std(axis=0) only have two
    values.
164 edge_model.fit(X)
165
166 # #####
    #####
167 # Cluster using affinity propagation
168 #
169 _, labels = cluster.affinity_propagation(edge_model.
    covariance_)
170 n_labels = labels.max()
171 #
172
173 for i in range(n_labels + 1):
174     print 'Cluster %i: %s' % ((i + 1), ', '.join(list(
        compress(name_valid, labels == i))))
175
176
177
178 # #####
    #####
179 # Find a low-dimension embedding for visualization: find
    the best position of
180 # the nodes (the stocks) on a 2D plane
181
182 # We use a dense eigen_solver to achieve reproducibility (
    arpack is
183 # initiated with random vectors that we don't control). In
```

```
183 addition, we
184 # use a large number of neighbors to capture the large-
scale structure.
185 node_position_model = manifold.LocallyLinearEmbedding(
186     n_components=3, eigen_solver='dense', n_neighbors=6)
187
188 embedding = node_position_model.fit_transform(X.T).T
189
190 # #####
#####
191 # Visualization
192 plt.figure(1, facecolor='w', figsize=(10, 8))
193 plt.clf()
194 ax = plt.axes([0., 0., 1., 1.])
195 plt.axis('off')
196
197 # Display a graph of the partial correlations
198 partial_correlations = edge_model.precision_.copy()
199 d = 1 / np.sqrt(np.diag(partial_correlations))
200 partial_correlations *= d
201 partial_correlations *= d[:, np.newaxis]
202 non_zero = (np.abs(np.triu(partial_correlations, k=1)) > 0
203             .02)
204
205 # Plot the nodes using the coordinates of our embedding
206 plt.scatter(embedding[0], embedding[1], s=100 * d ** 2, c=
207             labels,
208             cmap=plt.cm.spectral)
209
210 # Plot the edges
211 start_idx, end_idx = np.where(non_zero)
212 # a sequence of (*line0*, *line1*, *line2*), where::
#         linen = (x0, y0), (x1, y1), ... (xm, ym)
213 segments = [[embedding[:, start], embedding[:, stop]]
214             for start, stop in zip(start_idx, end_idx)]
215 values = np.abs(partial_correlations[non_zero])
216 lc = LineCollection(segments,
217                   zorder=0, cmap=plt.cm.hot_r,
218                   norm=plt.Normalize(0, .7 * values.max(
219 )))
220 lc.set_array(values)
221 lc.set_linewidths(15 * values)
222 ax.add_collection(lc)
223
224 # Add a label to each node. The challenge here is that we
want to
225 # position the labels to avoid overlap with other labels
226 for index, (name, label, (x, y)) in enumerate(
227     zip(names, labels, embedding.T)):
```

```
227     dx = x - embedding[0]
228     dx[index] = 1
229     dy = y - embedding[1]
230     dy[index] = 1
231     this_dx = dx[np.argmin(np.abs(dy))]
232     this_dy = dy[np.argmin(np.abs(dx))]
233     if this_dx > 0:
234         horizontalalignment = 'left'
235         x = x + .002
236     else:
237         horizontalalignment = 'right'
238         x = x - .002
239     if this_dy > 0:
240         verticalalignment = 'bottom'
241         y = y + .002
242     else:
243         verticalalignment = 'top'
244         y = y - .002
245     plt.text(x, y, name, size=10,
246             horizontalalignment=horizontalalignment,
247             verticalalignment=verticalalignment,
248             bbox=dict(facecolor='w',
249                     edgecolor=plt.cm.spectral(label /
250 float(n_labels))),
251             alpha=.6))
252 plt.xlim(embedding[0].min() - .15 * embedding[0].ptp(),
253         embedding[0].max() + .10 * embedding[0].ptp(),)
254 plt.ylim(embedding[1].min() - .03 * embedding[1].ptp(),
255         embedding[1].max() + .03 * embedding[1].ptp())
256
257 plt.show()
258
```