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Prediction of Stock Price for United Airlines via Sentiment Analytics
 In [1]: import pandas as pd
         import numpy as np
         import datetime
         import pandas datareader.data as web
         import re
         Get Stock Price
 In [2]: start = datetime.datetime(2017,3,1)
         end = datetime.datetime(2017,4,28)
         UAL = web.DataReader("UAL", "yahoo", start, end)
 In [3]: UAL price=UAL.loc[:,['Open','Close']]
         UAL price['price'] = list(map(lambda x,y: x-y, UAL price.loc[:,'Open'], UAL price.loc[:,'Close']))
 In [4]: UAL price['trend']=1
         for i in np.arange(len(UAL_price.loc[:,'price'])):
             if UAL price.iloc[i,2] > 0:
                 UAL_price.iloc[i,3]='up'
             elif UAL price.iloc[i,2] == 0:
                 UAL price.iloc[i,3]='-'
             else:
                 UAL price.iloc[i,3]='dwon'
 In [5]: UAL_price.head()
 Out[5]:
                    Open
                              Close
                                        price
                                                 trend
          Date
                              75.389999
                                        -0.750000
          2017-03-01
                    74.639999
                                                 dwon
                                        1.069999
          2017-03-02 | 74.349998 | 73.279999
          2017-03-03 73.699997
                              75.589996
                                        -1.889999
                                                 dwon
          2017-03-06 | 74.849998 | 73.139999
                                        1.709999
          2017-03-07 | 72.690002 | 72.639999 | 0.050003
 In [6]: UAL price.trend.value counts()
 Out[6]: dwon
                 22
                 20
         Name: trend, dtype: int64
         Import Tweets
 In [7]: #import the tweets file
         tweets_table_428 = pd.read_csv(r'C://Users//adminstrator//Google Drive//660//chromedriver_win32//tw
         eets_try_ua_0428_formal.txt',encoding='utf-8', error_bad_lines=False,sep='\t',header=None)
         tweets table 420 = pd.read csv(r'C://Users//adminstrator//Google Drive//660//chromedriver win32//tw
         eets_try_ua.txt',encoding='utf-8', error_bad_lines=False,sep='\t',header=None)
 In [8]: def dataclean(tweets table):
             tweets table=tweets table.iloc[:,[0,2,4,6,8]]
             tweets_table=tweets_table.rename(columns={0:'tweets',2:'retweet',4:'favorites',6:'replies',8:'da
         te'})
             q = lambda x: str(int(x))
             tweets table=tweets table.dropna(subset=['date'])
             tweets table.replies = tweets table.replies.fillna("0").apply(g)
             f = lambda x: float(x.replace("K",""))*1000 if "K" in x else int(x)
             tweets table.retweet = tweets table.retweet.fillna("0").apply(f)
             tweets_table.favorites = tweets_table.favorites.fillna("0").apply(f)
             tweets table.replies = tweets_table.replies.fillna("0").apply(f)
             print(tweets table.isnull().sum())
             return tweets table
 In [9]: tweets_table_428 = dataclean(tweets_table_428)
                      0
         tweets
         retweet
         favorites 0
         replies
         date
                      0
         dtype: int64
In [10]: tweets table 420 = dataclean(tweets table 420)
         tweets
         retweet
                      0
                      0
         favorites
         replies
         date
         dtype: int64
In [11]: import codecs
         #function that loads a lexicon of positive words to a set and returns the set
         def loadLexicon(fname):
             lex conn=codecs.open(fname, encoding='utf-8')
             #add every word in the file to the set
             for line in lex conn:
                 newLex.add(line.strip())# remember to strip to remove the lin-change character
             lex_conn.close()
             return newLex
In [12]: def sentiment(line):
             #load the positive and negative lexicons
             posLex=loadLexicon('positive-words.txt')
             negLex=loadLexicon('negative-words.txt')
             posList=[] #list of positive words in the review
             negList=[] #list of negative words in the review
             line=line.lower().strip()
             words=line.split(' ') # slit on the space to get list of words
             for word in words: #for every word in the review
                 if word in posLex: # if the word is in the positive lexicon
                         posList.append(word) #update the positive list for this review
                 if word in negLex: # if the word is in the negative lexicon
                         negList.append(word) #update the negative list for this review
                  # 0 for neutral
             if len(posList)>len(negList): # more pos words than neg
                     decision='positive' # 1 for positiv
             elif len(negList)>len(posList): # more neg than pos
                     decision='negative' # -1 for negative
             else:
                     decision = 'neutral'
             return decision
In [13]: #Apply sentiment function for all tweets
         tweets table 420['sentiment'] = list(map(sentiment, tweets table 420['tweets']))
         tweets_table_428['sentiment'] = list(map(sentiment, tweets_table_428['tweets']))
In [14]: from dateutil.parser import parse
         import datetime
In [15]: def Date Formate(tweets table):
         #Change the format of date
             tweets_table=tweets_table.reset_index()
             for i in np.arange(len(tweets table.loc[:,'date'])):
                 tweets_table.loc[i,'date'] = datetime.datetime.strptime(str(tweets_table.loc[i,'date']).lstrip
          (), '%d %b %Y').strftime('%Y-%m-%d')
                 i += 1
             #Remove the the tweets before 2017-03-01 which are not target data
             tweets table=tweets table[tweets table['date']>='2017-03-01']
             #tweets_table=tweets_table[tweets_table['date']<'2017-04-20']</pre>
             tweets table 1=tweets table.set index(['date'])
             tweets table 1=tweets table 1.rename({'2017-03-04':'2017-03-06','2017-03-05':'2017-03-06',
                                  '2017-03-11':'2017-03-13','2017-03-12':'2017-03-13',
                                 '2017-03-18': '2017-03-20', '2017-03-19': '2017-03-20',
                                 '2017-03-25': '2017-03-27', '2017-03-26': '2017-03-27',
                                '2017-04-01':'2017-04-03','2017-04-02':'2017-04-03',
                                '2017-04-08': '2017-04-10', '2017-04-09': '2017-04-10',
                                '2017-04-14':'2017-04-17','2017-04-15':'2017-04-17','2017-04-16':'2017-04-17',
                                 '2017-04-22':'2017-04-24','2017-04-23':'2017-04-24'})
             #Join the table of tweets and stock price based on the date index
             result=tweets_table_1.join(UAL_price, how='outer')
             #Remove unnecessary
             result_df=result.drop('index',1).drop('Open',1).drop('Close',1).drop('price',1).drop('tweets',1)
             result_df_1=result_df.dropna(subset=['sentiment'])
             print(result df 1.isnull().sum())
             print(result.sentiment.value_counts())
             return result df 1
In [16]: result_df_420 = Date_Formate(tweets_table_420)
         retweet
         favorites 0
         replies
                      0
         sentiment 0
         trend
                      0
         dtype: int64
         neutral
         negative
                     367
         positive
         Name: sentiment, dtype: int64
In [17]: result df 428 = Date_Formate(tweets_table_428)
         retweet
         favorites
                      0
                      0
         replies
                      0
         sentiment
         trend
         dtype: int64
         neutral
                   1287
         negative
         positive
                    637
         Name: sentiment, dtype: int64
         Modeling and Predicting
In [34]: from sklearn import preprocessing
         def encode_variable(result_df_1):
             label encoder = preprocessing.LabelEncoder()
             result_df_1['sentiment'] = label_encoder.fit_transform(result_df_1['sentiment'])
             result_df_1['trend'] = label_encoder.fit_transform(result_df_1['trend'])
             return result_df_1
In [35]: df_420 = encode_variable(result_df_420)
In [37]: df 428 = encode variable(result df 428)
In [38]: from sklearn.model_selection import train_test_split
         from sklearn.linear model import LogisticRegression
         from sklearn.grid_search import GridSearchCV
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier, VotingClassifier
         from sklearn.metrics import accuracy score
         from xgboost import XGBClassifier
         C:\Users\amyhu\Anaconda3\lib\site-packages\sklearn\cross_validation.py:44: DeprecationWarning: T
         his module was deprecated in version 0.18 in favor of the model_selection module into which all
         the refactored classes and functions are moved. Also note that the interface of the new CV itera
         tors are different from that of this module. This module will be removed in 0.20.
           "This module will be removed in 0.20.", DeprecationWarning)
         C:\Users\amyhu\Anaconda3\lib\site-packages\sklearn\grid_search.py:43: DeprecationWarning: This m
         odule was deprecated in version 0.18 in favor of the model_selection module into which all the r
         efactored classes and functions are moved. This module will be removed in 0.20.
           DeprecationWarning)
In [39]: random state = 367
In [40]: def model accarucy(x):
             train, test = train_test_split(x, train_size = 0.6)
             x train = train.iloc[:,0:4]
             y train = train.iloc[:,-1]
             x test = test.iloc[:,0:4]
             y test = test.iloc[:,-1]
             LREG grid = [ {'C':[0.5,1,1.5,2],'penalty':['11','12']}]
             LREG classifier=LogisticRegression()
             gridsearchLREG = GridSearchCV(LREG classifier, LREG grid, cv=5)
             gridsearchLREG.fit(x_train,y_train)
             KNN grid = [{'n neighbors': [1,3,5,7,9,11,13,15,17], 'weights':['uniform','distance']}]
             KNN=KNeighborsClassifier()
             gridsearchKNN = GridSearchCV(KNN, KNN grid, cv=5)
             gridsearchKNN.fit(x_train, y_train)
             RF grid = [{'max depth': [3,4,5,6,7,8,9,10,11,12],'criterion':['gini','entropy']}]
             RF = RandomForestClassifier(n_estimators=250, criterion='entropy', n_jobs = -1, random_state=r
         andom state)
             gridsearchRF = GridSearchCV(RF, RF_grid, cv=5)
             gridsearchRF.fit(x train, y train)
             GBC_grid = [{'n_estimators':[1,2,3,4,5,6,7],'max_depth':[10,20,30,40,50], 'min_samples_split':[2
         00,300,\overline{400,5001}
             GBC = GradientBoostingClassifier()
             gridsearchGBC = GridSearchCV(GBC, GBC_grid, cv=5,scoring='roc_auc',n_jobs=4,iid=False,)
             gridsearchGBC.fit(x train, y train)
             XGB_grid = [{'max_depth':[10,20,30,40,50],'min_child_weight':[1,2,3,4,5,6,7]}]
             XGB = XGBClassifier(seed=random_state)
             gridsearchXGB = GridSearchCV(XGB, XGB grid, cv=5)
             gridsearchXGB.fit(x_train, y_train)
             eclf 1 = VotingClassifier(estimators=[
             ('KNN', gridsearchKNN), ('RF', gridsearchRF), ('GBC', gridsearchGBC), ('XGB', gridsearchXGB), ('LR
         EG', gridsearchLREG)], voting='soft')
             eclf_1.fit(x_train, y_train)
             y pred 1 = eclf 1.predict(x test)
             return accuracy score(y test, y pred 1)
In [41]: model_accarucy(df_420)
Out[41]: 0.92343387470997684
         Keywords visualization with WordCloud
In [18]: from os import path
         from PIL import Image
         import matplotlib.pyplot as plt
         from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
In [89]: def wordcloud(tweets_table):
             text=[]
             for i in tweets table.tweets:
                 text += [i]
             text = str(text).lower()
             text = re.sub('[^a-z]',' ',text)
             text = re.sub('united airlines',' ',text)
             text = re.sub('united airline',' ',text)
             text = re.sub('unitedairline',' ',text)
             text = re.sub('united',' ',text)
             text = re.sub('flight',' ',text)
             text = re.sub('airline',' ',text)
             return text
In [90]: text 1 = wordcloud(tweets table 420)
         text 2 = wordcloud(tweets table 428)
In [93]: def wordscloud(text):
             d = path.dirname("C://Users//adminstrator//Google Drive//660//chromedriver win32")
             # read the mask image
             mask = np.array(Image.open(path.join(d, "C://Users//adminstrator//Google Drive//660//chromedrive
         r win32//airplane 2.jpg")))
             stopwords = set(STOPWORDS)
             stopwords.add("said")
             wc = WordCloud(background color="white", max words=2000, mask=mask,
                        stopwords=stopwords)
             # generate word cloud
             wc.generate(text)
             # store to file
             wc.to file(path.join(d, "C://Users//adminstrator//Google Drive//660//chromedriver win32//UA 2.pn
         g"))
             plt.imshow(wc, interpolation='bilinear')
             plt.axis("off")
             plt.figure()
             plt.imshow(mask, cmap=plt.cm.gray, interpolation='bilinear')
             plt.axis("off")
             return plt.show()
In [92]: wordscloud(text 1)
                   employee
              video
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trave passenger flying trave passenger flyseat need need than ticket passenger flyseat flying the flying passenger flyseat flying passenger flyseat flying passenger flying flying flying passenger flying flying flying passenger flying flying passenger flying fly

guy man

In [94]: wordscloud(text 2)

