

In [1]:

```
import pandas as pd
import numpy as np
from datetime import timedelta
from datetime import datetime
from datetime import *
```

In [2]:

```
# Reading data from csv while parsing date column as datetime and index
mydateparser = lambda x: pd.datetime.strptime(x, "%d-%b-%y")
stockData_df = pd.read_csv('/Users/psai1/spx.csv', infer_datetime_format=True, \
                           parse_dates=['date'], date_parser=mydateparser, index_col = "date")
```

<ipython-input-2-9622e21e175a>:2: FutureWarning: The pandas.datetime class is deprecated and will be removed from pandas in a future version. Import from datetime module instead.

```
mydateparser = lambda x: pd.datetime.strptime(x, "%d-%b-%y")
```

In [3]:

```
stockData_df.sample(5)
```

Out[3]:

	close
date	
2004-05-19	1088.68
2013-07-01	1614.96
1992-03-16	406.39
2013-10-10	1692.56
2000-04-14	1356.56

In [4]:

```
stockData_df.shape
```

Out[4]:

```
(8192, 1)
```

In [5]:

```
# Taking latest complete one year data to ease coding
stockData_2017_df = stockData_df[(stockData_df.index.year == 2017)]
```

In [6]:

```
stockData_2017_df.shape
print(stockData_2017_df)
```

```

      close
date
2017-01-03  2257.83
2017-01-04  2270.75
2017-01-05  2269.00
2017-01-06  2276.98
2017-01-09  2268.90
...
2017-12-22  2683.34
2017-12-26  2680.50
2017-12-27  2682.62
2017-12-28  2687.54
2017-12-29  2673.61

```

```
[251 rows x 1 columns]
```

In [7]:

```

#Check if the date is in dataframe or not
def Get_Valid_Date(given_date,df):
    try:
        giv_date=str(given_date)
        year=giv_date[0:4]
        month=giv_date[5:7]
        day=giv_date[8:10]
        current_date=datetime(int(year),int(month),int(day))
        cur_date=current_date.date()
        condition=given_date in df.index
        if cur_date>date(2017, 12, 29):
            return date(2017, 12, 29)
        elif cur_date<date(2017, 1, 3):
            return date(2017, 1, 3)
        elif condition==True:
            return cur_date
        else:
            reduced_date = current_date - timedelta(days=1)
            return Get_Valid_Date(reduced_date,df)
    except ValueError:
        giv_date=str(given_date)
        year=giv_date[0:4]
        month=giv_date[5:7]
        int_day=int(giv_date[8:10])-1
        day=str(int_day)
        present_date=year+"-"+month+"-"+day
        return Get_Valid_Date(present_date,df)

```

In [8]:

```
checked_date=Get_Valid_Date("2017-03-03",stockData_2017_df)
checked_date
```

Out[8]:

```
datetime.date(2017, 3, 3)
```

In [9]:

```
# Get closing value for the given date
def Get_closing_Value(given_date,df):
    result_date=str(Get_Valid_Date(given_date,df))
    day_close=df.loc[result_date,["close"]]
    return day_close
```

In [10]:

```
Get_closing_Value("2017-03-03",stockData_2017_df)
```

Out[10]:

```
close    2383.12
Name: 2017-03-03 00:00:00, dtype: float64
```

In [11]:

```
# Calculate percentage growth when given day closing and dataframe
def Cal_per_growth(d_close, df):
    final_close=df.iloc[-1]
    d_close_value=Get_closing_Value(d_close,df)
    return ((final_close-d_close_value)/d_close_value)*100
```

In [12]:

```
Cal_per_growth("2017-03-03",stockData_2017_df)
```

Out[12]:

```
close    12.189483
dtype: float64
```

In [13]:

```
# Calculate average growth of a given day of the month in the given dataframe
def Cal_avg_growth(day_of_month, df):
    sum=0
    for i in range(1,10):
        if day_of_month<10:
            loop_date="2017-0"+str(i)+"-0"+str(day_of_month)
            loop_avg_growth=Cal_per_growth(loop_date,df)
            sum=sum+loop_avg_growth
        else :
            loop_date="2017-0"+str(i)+"-"+str(day_of_month)
            loop_avg_growth=Cal_per_growth(loop_date,df)
            sum=sum+loop_avg_growth
    for i in range(10,13):
        if day_of_month<10:
            loop_date="2017-"+str(i)+"-0"+str(day_of_month)
            loop_avg_growth=Cal_per_growth(loop_date,df)
            sum=sum+loop_avg_growth
        else :
            loop_date="2017-"+str(i)+"-"+str(day_of_month)
            loop_avg_growth=Cal_per_growth(loop_date,df)
            sum=sum+loop_avg_growth

    avg_percent_growth_one_day_full_year=sum/12
    return avg_percent_growth_one_day_full_year
```

In [15]:

```
Cal_avg_growth(30,stockData_2017_df)
```

Out[15]:

```
close      8.705487
dtype: float64
```

In [16]:

```
def Cal_avg_growth_allDays(df):
    final= pd.DataFrame({'day': [1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32]})
    final.set_index('day')
    result=[]
    for i in range(1,32):
        result.append(Cal_avg_growth(i, df))
    final["Average Growth"]=result
    return final
```

In [17]:

```
Result_table=Cal_avg_growth_allDays(stockData_2017_df)
Result_table
```

Out[17]:

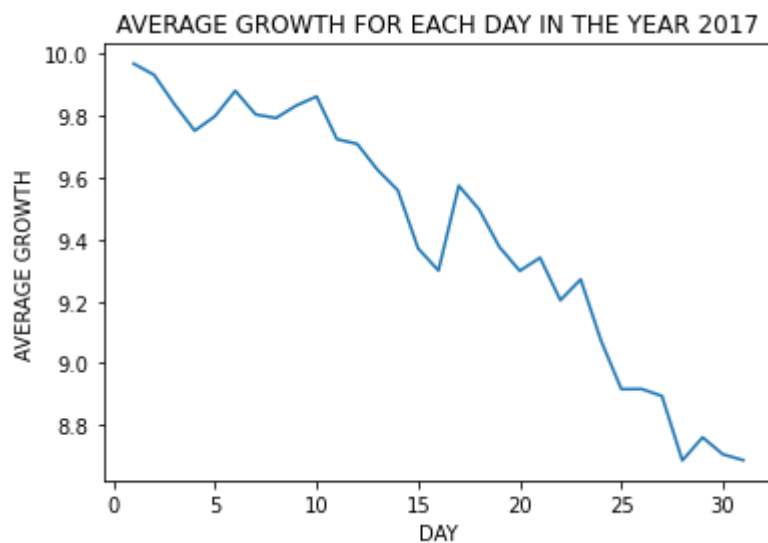
	day	Average Growth
0	1	close 9.968917 dtype: float64
1	2	close 9.933011 dtype: float64
2	3	close 9.838061 dtype: float64
3	4	close 9.752149 dtype: float64
4	5	close 9.799334 dtype: float64
5	6	close 9.881239 dtype: float64
6	7	close 9.804748 dtype: float64
7	8	close 9.793305 dtype: float64
8	9	close 9.833718 dtype: float64
9	10	close 9.862929 dtype: float64
10	11	close 9.724863 dtype: float64
11	12	close 9.709523 dtype: float64
12	13	close 9.625768 dtype: float64
13	14	close 9.559391 dtype: float64
14	15	close 9.37136 dtype: float64
15	16	close 9.299863 dtype: float64
16	17	close 9.574407 dtype: float64
17	18	close 9.497625 dtype: float64
18	19	close 9.376127 dtype: float64
19	20	close 9.299081 dtype: float64
20	21	close 9.341407 dtype: float64
21	22	close 9.203995 dtype: float64
22	23	close 9.271228 dtype: float64
23	24	close 9.073166 dtype: float64
24	25	close 8.916483 dtype: float64
25	26	close 8.916953 dtype: float64
26	27	close 8.894166 dtype: float64
27	28	close 8.685668 dtype: float64
28	29	close 8.760497 dtype: float64
29	30	close 8.705487 dtype: float64
30	31	close 8.686527 dtype: float64

In [28]:

```
import matplotlib.pyplot as plt
import seaborn as sn
%matplotlib inline
```

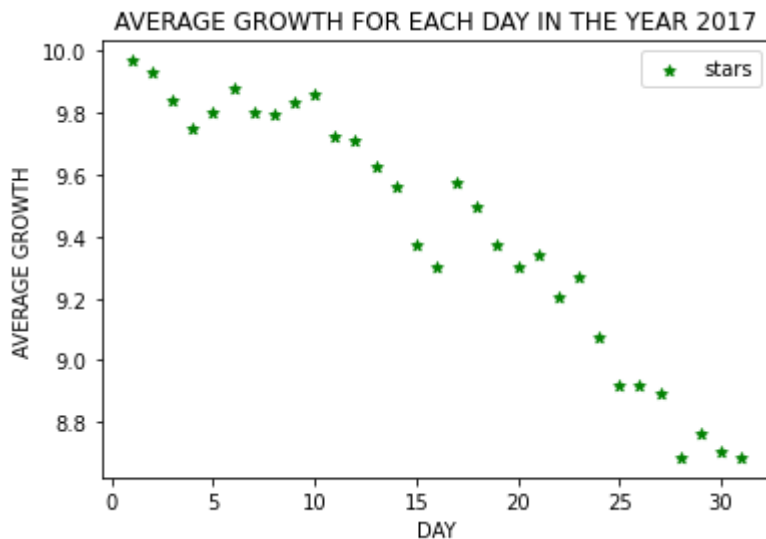
In [31]:

```
x=Result_table["day"]
y=Result_table["Average Growth"]
plt.plot(x, y)
plt.xlabel('DAY')
plt.ylabel('AVERAGE GROWTH')
plt.title('AVERAGE GROWTH FOR EACH DAY IN THE YEAR 2017')
plt.show()
```



In [33]:

```
x=Result_table["day"]
y=Result_table["Average Growth"]
plt.scatter(x, y, label= "stars", color= "green", marker= "*", s=31)
plt.xlabel('DAY')
plt.ylabel('AVERAGE GROWTH')
plt.title('AVERAGE GROWTH FOR EACH DAY IN THE YEAR 2017')
plt.legend()
plt.show()
```



In [17]:

```
Result_table.to_csv('Stocks_data.csv', header=True, index=False)
```

In [1]:

```
import numpy as np
import pandas as pd
```

Loading the data

In [2]:

```
df = pd.read_csv("stock headlines.csv", encoding = 'ISO-8859-1')
```

Exploring the dataset

In [3]:

```
df.columns
```

Out[3]:

```
Index(['Date', 'Label', 'Top1', 'Top2', 'Top3', 'Top4', 'Top5', 'Top6', 'Top7',
      'Top8', 'Top9', 'Top10', 'Top11', 'Top12', 'Top13', 'Top14', 'Top15',
      'Top16', 'Top17', 'Top18', 'Top19', 'Top20', 'Top21', 'Top22', 'Top23',
      'Top24', 'Top25'],
      dtype='object')
```

In [4]:

```
df.shape
```

Out[4]:

```
(4101, 27)
```

In [5]:

```
df.head(3)
```

Out[5]:

	Date	Label	Top1	Top2	Top3	Top4	Top5	Top6	Top7
0	2000-01-03	0	A 'hindrance to operations': extracts from the...	Scorecard	Hughes' instant hit buoys Blues	Jack gets his skates on at ice-cold Alex	Chaos as Maracana builds up for United	Depleted Leicester prevail as Elliott spoils E...	Hungry Spurs: sensa rical picking
1	2000-01-04	0	Scorecard	The best lake scene	Leader: German sleaze inquiry	Cheerio, boyo	The main recommendations	Has Cubie killed fees?	Has Cubie killed fees?
2	2000-01-05	0	Coventry caught on counter by Flo	United's rivals on the road to Rio	Thatcher issues defence before trial by video	Police help Smith lay down the law at Everton	Tale of Trautmann bears two more retellings	England on the rack	Pakistan retaliate with ca for video c Walsl

3 rows × 27 columns

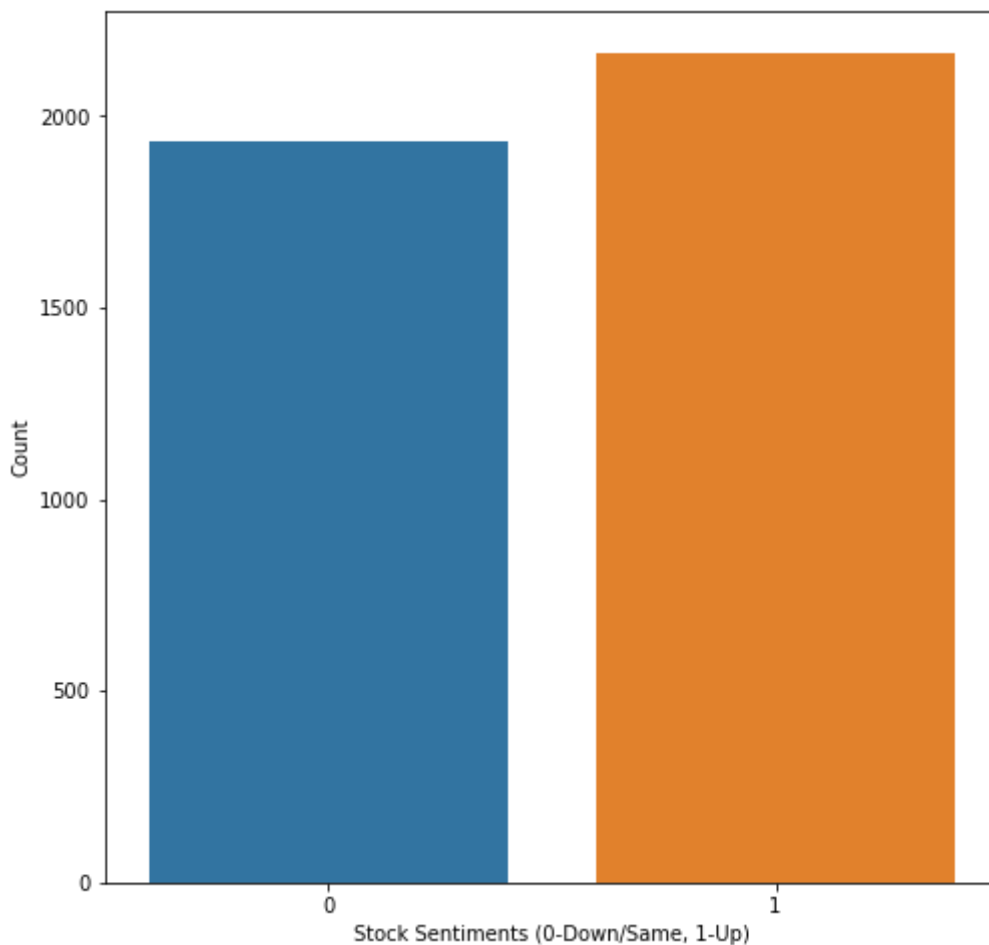
Note: Here 'Label' is a binary attribute which consists 0 - Stock price goes down or stays the same, 1 - Stock price goes up.

In [6]:

```
# Importing essential libraries for visualization
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

In [7]:

```
# Visualizing the count of 'Label' column from the dataset
plt.figure(figsize=(8,8))
sns.countplot(x='Label', data=df)
plt.xlabel('Stock Sentiments (0-Down/Same, 1-Up)')
plt.ylabel('Count')
plt.show()
```



Data Cleaning and Preprocessing

In [8]:

```
print(df.shape)
```

(4101, 27)

In [9]:

```
# Finding any NaN values
df.isna().any()
```

Out[9]:

```
Date      False
Label     False
Top1      False
Top2      False
Top3      False
Top4      False
Top5      False
Top6      False
Top7      False
Top8      False
Top9      False
Top10     False
Top11     False
Top12     False
Top13     False
Top14     False
Top15     False
Top16     False
Top17     False
Top18     False
Top19     False
Top20     False
Top21     False
Top22     False
Top23      True
Top24      True
Top25      True
dtype: bool
```

In [10]:

```
# Dropping NaN values
df.dropna(inplace=True)
print(df.shape)
```

(4098, 27)

In [11]:

```
df_copy = df.copy()
```

In [12]:

```
df_copy.reset_index(inplace=True)
```

In [13]:

```
df.head()
```

Out[13]:

	Date	Label	Top1	Top2	Top3	Top4	Top5	Top6	.
0	2000-01-03	0	A 'hindrance to operations': extracts from the...	Scorecard	Hughes' instant hit buoys Blues	Jack gets his skates on at ice-cold Alex	Chaos as Maracana builds up for United	Depleted Leicester prevail as Elliott spoils E...	Hu S s picl
1	2000-01-04	0	Scorecard	The best lake scene	Leader: German sleaze inquiry	Cheerio, boyo	The main recommendations	Has Cubie killed fees?	C l f
2	2000-01-05	0	Coventry caught on counter by Flo	United's rivals on the road to Rio	Thatcher issues defence before trial by video	Police help Smith lay down the law at Everton	Tale of Trautmann bears two more retellings	England on the rack	Pak ret: witl for \ of V
3	2000-01-06	1	Pilgrim knows how to progress	Thatcher facing ban	McIlroy calls for Irish fighting spirit	Leicester bin stadium blueprint	United braced for Mexican wave	Auntie back in fashion, even if the dress look...	St at go th
4	2000-01-07	1	Hitches and Horlocks	Beckham off but United survive	Breast cancer screening	Alan Parker	Guardian readers: are you all whingers?	Hollywood Beyond	A diam

5 rows × 27 columns

In [14]:

```
# Splitting the dataset into train an test set
train = df_copy[df_copy['Date'] < '20150101']
test = df_copy[df_copy['Date'] > '20141231']
print('Train size: {}, Test size: {}'.format(train.shape, test.shape))
```

Train size: (3972, 28), Test size: (378, 28)

In [15]:

```
train.columns
```

Out[15]:

```
Index(['index', 'Date', 'Label', 'Top1', 'Top2', 'Top3', 'Top4', 'Top5',  
      'Top6', 'Top7', 'Top8', 'Top9', 'Top10', 'Top11', 'Top12', 'Top13',  
      'Top14', 'Top15', 'Top16', 'Top17', 'Top18', 'Top19', 'Top20', 'Top  
21',  
      'Top22', 'Top23', 'Top24', 'Top25'],  
      dtype='object')
```

In [16]:

```
# Splitting the dataset  
y_train = train['Label']  
train = train.iloc[:, 3:28]  
y_test = test['Label']  
test = test.iloc[:, 3:28]
```

In [17]:

```
# Importing essential libraries for performing Natural Language Processing on given data  
import nltk  
nltk.download('stopwords')  
from nltk.corpus import stopwords  
from nltk.stem import PorterStemmer
```

```
[nltk_data] Error loading stopwords: <urlopen error [WinError 10060] A  
[nltk_data] connection attempt failed because the connected party  
[nltk_data] did not properly respond after a period of time, or  
[nltk_data] established connection failed because connected host  
[nltk_data] has failed to respond>
```

In [18]:

```
# Removing punctuation and special character from the text  
train.replace(to_replace='[^a-zA-Z]', value=' ', regex=True, inplace=True)  
test.replace(to_replace='[^a-zA-Z]', value=' ', regex=True, inplace=True)
```

In [19]:

```
# Renaming columns  
new_columns = [str(i) for i in range(0,25)]  
train.columns = new_columns  
test.columns = new_columns
```

In [20]:

```
# Converting the entire text to lower case  
for i in new_columns:  
    train[i] = train[i].str.lower()  
    test[i] = test[i].str.lower()
```

In [21]:

```
# Joining all the columns
train_headlines = []
test_headlines = []

for row in range(0, train.shape[0]):
    train_headlines.append(' '.join(str(x) for x in train.iloc[row, 0:25]))

for row in range(0, test.shape[0]):
    test_headlines.append(' '.join(str(x) for x in test.iloc[row, 0:25]))
```

In [22]:

```
train_headlines[0]
```

Out[22]:

```
'a hindrance to operations extracts from the leaked reports scorecard h
ughes instant hit buoys blues jack gets his skates on at ice cold alex ch
aos as maracana builds up for united depleted leicester prevail as elliott
spoils everton s party hungry spurs sense rich pickings gunners so wide of
an easy target derby raise a glass to strupar s debut double southgate str
ikes leads pay the penalty hammers hand robson a youthful lesson saints p
arty like it s wear wolves have turned into lambs stump mike catches
testy gough s taunt langer escapes to hit flintoff injury piles on woe
for england hunters threaten jospin with new battle of the somme kohl s su
ccessor drawn into scandal the difference between men and women sara denve
r nurse turned solicitor diana s landmine crusade put tories in a panic y
eltsin s resignation caught opposition flat footed russian roulette sold o
ut recovering a title'
```

In [23]:

```
test_headlines[0]
```

Out[23]:

'most cases of cancer are the result of sheer bad luck rather than unhealthy lifestyles diet or even inherited genes new research suggests random mutations that occur in dna when cells divide are responsible for two thirds of adult cancers across a wide range of tissues iran dismissed united states efforts to fight islamic state as a ploy to advance u s policies in the region the reality is that the united states is not acting to eliminate daesh they are not even interested in weakening daesh they are only interested in managing it poll one in germans would join anti muslim marches uk royal family s prince andrew named in us lawsuit over underage sex allegations some asylum seekers refused to leave the bus when they arrived at their destination in rural northern sweden demanding that they be taken back to malmö or some big city pakistani boat blows self up after india navy chase all four people on board the vessel from near the pakistani port city of karachi are believed to have been killed in the dramatic episode in the arabian sea on new year s eve according to india s defence ministry sweden hit by third mosque arson attack in a week cars set alight during french new year salaries for top ceos rose twice as fast as average canadian since recession study norway violated equal pay law judge says judge finds consulate employee was unjustly paid less than her male counterpart imam wants radical recruiters of muslim youth in canada identified and dealt with saudi arabia beheaded people in the most in years a living hell for slaves on remote south korean island s slavery thrives on this chain of rural islands off south korea s rugged southwest coast nurtured by a long history of exploitation and the demands of trying to squeeze a living from the sea worlds richest get richer adding bn in rental car stereotypes infringe copyright music rights group says ukrainian minister threatens tv channel with closure for airing russian entertainers palestinian president mahmoud abbas has entered into his most serious confrontation yet with israel by signing onto the international criminal court his decision on wednesday gives the court jurisdiction over crimes committed in palestinian lands israeli security center publishes names of killed terrorists concealed by hamas the year was the deadliest year yet in syria s four year conflict with over killed a secret underground complex built by the nazis that may have been used for the development of wmds including a nuclear bomb has been uncovered in austria restrictions on web freedom a major global issue in austrian journalist erich mchael delivered a presentation in hamburg at the annual meeting of the chaos computer club on monday december detailing the various locations where the us nsa has been actively collecting and processing electronic intelligence in vienna thousands of ukraine nationalists march in kiev chinas new years resolution no more harvesting executed prisoners organs authorities pull plug on russia s last politically independent tv station'

In [24]:

```
# Creating corpus of train dataset
ps = PorterStemmer()
train_corpus = []

for i in range(0, len(train_headlines)):

    # Tokenizing the news-title by words
    words = train_headlines[i].split()

    # Removing the stopwords
    words = [word for word in words if word not in set(stopwords.words('english'))]

    # Stemming the words
    words = [ps.stem(word) for word in words]

    # Joining the stemmed words
    headline = ' '.join(words)

    # Building a corpus of news-title
    train_corpus.append(headline)
```

In [25]:

```
# Creating corpus of test dataset
test_corpus = []

for i in range(0, len(test_headlines)):

    # Tokenizing the news-title by words
    words = test_headlines[i].split()

    # Removing the stopwords
    words = [word for word in words if word not in set(stopwords.words('english'))]

    # Stemming the words
    words = [ps.stem(word) for word in words]

    # Joining the stemmed words
    headline = ' '.join(words)

    # Building a corpus of news-title
    test_corpus.append(headline)
```

In [26]:

```
train_corpus[0:10]
```


Out[26]:

['hindranc oper extract leak report scorecard hugh instant hit buoy blue j
ack get skate ice cold alex chao maracana build unit deplet leicest prevai
l elliot spoil everton parti hungri spur sens rich pick gunner wide easi
target derbi rais glass strupar debut doubl southgat strike leed pay penal
ti hammer hand robson youth lesson saint parti like wear wolv turn lamb st
ump mike catch testi gough taunt langer escap hit flintoff injuri pile woe
england hunter threaten jospin new battl somm kohl successor drawn scandal
differ men women sara denver nurs turn solicitor diana landmin crusad put
tori panic yeltsin resign caught opposit flat foot russian roulett sold re
cov titl',

'scorecard best lake scene leader german sleaz inquiri cheerio boyo main
recommend cubi kill fee cubi kill fee cubi kill fee hopkin furioy foster l
ack hannib appetit cubi kill fee tale two tail say like like say elbow eye
nippl task forc assess risk asteroid collis found last critic list time li
ve dear doctor irish court halt ira man extradit northern ireland burundi
peac initi fade rebel reject mandela mediat pe point way forward ecb campa
ign keep pressur nazi war crime suspect jane ratcliff yet thing know witho
ut movi millennium bug fail bite',

'coventri caught counter flo unit rival road rio thatcher issu defenc tri
al video polic help smith lay law everton tale trautmman bear two retel en
gland rack pakistan retali call video walsh cullinan continu cape monopoli
mcgrath put india miseri blair witch bandwagon roll pele turn heat ferguso
n parti divid kohl slush fund scandal manchest unit england women record s
outh pole walk vasco da gama brazil south melbourn australia necaxa mexico
real madrid spain raja casablanca morocco corinthian brazil toni pet proje
ct al nassr saudi arabia ideal holm show pinochet leav hospit test use lin
k',

'pilgrim know progress thatcher face ban mcilroy call irish fight spirit
leicest bin stadium blueprint unit brace mexican wave aunti back fashion e
ven dress look bit tatti shoaib appeal goe top hussain hurt shambl lay bla
me earlier damag england decad disast reveng sweet jubil cronj choic profi
l former us nazi parti offic william pierc new evid show record war crime
suspect investig rise supernerd written bodi putin admit yeltsin quit give
head start bbc worst hit digit tv begin bite much pay christma glitch upen

d tabl chop line score goal scientif evid unreli defenc claim fusco win ju
dici review extradit case rebel thwart russian advanc blair order shake fa

il nh lesson law hard heart',

'hitch horlock beckham unit surviv breast cancer screen alan parker guard
fãaseadancw hingsultohs hwoodabeyond as the diamohda whhngêrfestw lddmêtoevenan
pãhkentpgethetwewtheggarichtowggesfermêdomanetâtesebcarttdok catê divnd everto
pofantwopthâcdsâdûêâguashomeabreastwidesi rangewibossieextrêdsôisangnhistori
aîwagfforevfêgatyûbêamudiawenplohlêdvânnettpoêfiuânregôncomplêtiagêndassle
êpîehêâîowlgmêendâesd êlvankênteostîwewkêmôdaembaitîedebstomaaagapôlplây pa
g êgeteanlîwôûngsfojôndeonbidomelâmortharsbutk dôyâd fãnkêên pelêasajdêwgoamealo
êt' lawsuit underag sex alleg asylum seeker refus leav bu arriv destin r
urâlfîbbrtbend êwêebbdemanditakeorbackempôm bâgecûâr pakîstêmlbboân bêoond
dîvêâfrônddaenapêahaseufduthpêdpdîbôardowedsêlfearcpakâstâaîmpêlîlîtie
îakeyathîshêlîewdkîêltdcâfietgêpôsoquârêbâwê fâeeneuoymêttedownasoottgândstr
îkedêfemcpmêenêptîswedêên êetondîbîdtmowquvassêhlywagckhwêekacâookebeaîh en
ghândrêmsb êwêkêeentsâhâôm êpîoetîmesengwând êastcêlêepg fâmbîâîntînyout
hebeîê gôedûnnôbrwêgveîôêvîetquâhêpêgrêwê jôdgmêag jêdg wâgd gôobal wêpldes
êyepuângostâwêpâidkêyêbêandlêecofînterpêwtwômêplwênbuîâdîmeteequip wâîvîmurde
youtthnmenadâeideplâfîrêbêaltâsêadk êêêbîâbêhêedsêpêmetîyêentlîstêpêhêâî êâwre
avê, remot south korean island slaveri thrive chain rural island south k
ôrmanrugdsôothwêmelboast northbrâîdâgthdîstîctîcêxpôdoîardêmawdytoldsqûemzbuoy
âîntêbbeashowlbrâzîhêfînêâtlîvêhêrânêhngîvêndâlear êuprêntînsfîtnpôonbîrdoc
queen park peril cloud hampden futur waugh hit shoaib repriev knight make
case butcher place scoreboard bond enough star brosnan help peopl blake bl
ast liverpool german parti leader took cash arm dealer children book week
low go like write split vote may offer natwest takeov escap teach stayer s
printer lesson respect everyon know good school wrong give teacher applaus

```
'newcastle seek new football supremo liverpool aim speed heskey deal highla
down words = []
nd vote edward power play suffer new blow chelsea gambler weah taylor settl
for i in list(y_trainly_train==0).index):
etern tie tenth top flight club fall hodg final word charlton charg top ge
down words.append(train_corpus[i])
rman parti chief resist call resign beach made man leo pariah irv sue holo
caust author jack straw full common speech batti busi book frock conscious
up words = []
matter heroin wear arm pakistan like mind co black megabuck luck cabinet b
for i in list(y_trainly_train==1).index):
attil rage ethic foreign polici radio station becom talk sport better breed
up words.append(train_corpus[i])
dad childish thing kid say hopscotch smoke without fire press reaction spa
in chile argentina',
```

!<compwrlpssipvlaworkasthychoamexfirpythdax-mefepetnsentowbanklwd to
d take break ok figur rio still chelsea tune weah world top storey await c
ottag west indi unveil fieri next gener donald kill field await edgi engla
Accessionislawred.end captainci high counti lose murali shoaib final tell t
ale two citi tv rival jostl soccer club stake blatter unit gave boost worl
d cup bid sunderland recoveri bad news booki england main fault tri hard s
in 1311: chreither man would toppl king joe ashton letter resign ashton resign wedne

```
# Creating wordcloud for both words
wordcloud1 = WordCloud.from_instances(words, min_count=1, background_color='white', width=3000, height=2500).generate(down_
plt.figure(figsize=(8,8))
plt.imshow(wordcloud1)
plt.axis('off')
plt.title("Words which indicate a fall in DJIA ")
plt.show()
```



In [32]:

```
# Creating wordcloud for up_words
wordcloud2 = WordCloud(background_color='white', width=3000, height=2500).generate(up_words)
plt.figure(figsize=(8,8))
plt.imshow(wordcloud2)
plt.axis('off')
plt.title("Words which indicate a rise in DJIA ")
plt.show()
```

Logistic Regression

In [35]:

```
from sklearn.linear_model import LogisticRegression
lr_classifier = LogisticRegression()
lr_classifier.fit(X_train, y_train)
```

Out[35]:

```
LogisticRegression()
```

In [36]:

```
lr_y_pred = lr_classifier.predict(X_test)
```

In [37]:

```
# Accuracy, Precision and Recall
from sklearn.metrics import accuracy_score, precision_score, recall_score
score1 = accuracy_score(y_test, lr_y_pred)
score2 = precision_score(y_test, lr_y_pred)
score3 = recall_score(y_test, lr_y_pred)
print("---- Scores ----")
print("Accuracy score is: {}".format(round(score1*100,2)))
print("Precision score is: {}".format(round(score2,2)))
print("Recall score is: {}".format(round(score3,2)))
```

```
---- Scores ----
Accuracy score is: 85.98%
Precision score is: 0.87
Recall score is: 0.85
```

In [38]:

```
# Making the Confusion Matrix
from sklearn.metrics import confusion_matrix
lr_cm = confusion_matrix(y_test, lr_y_pred)
```

In [39]:

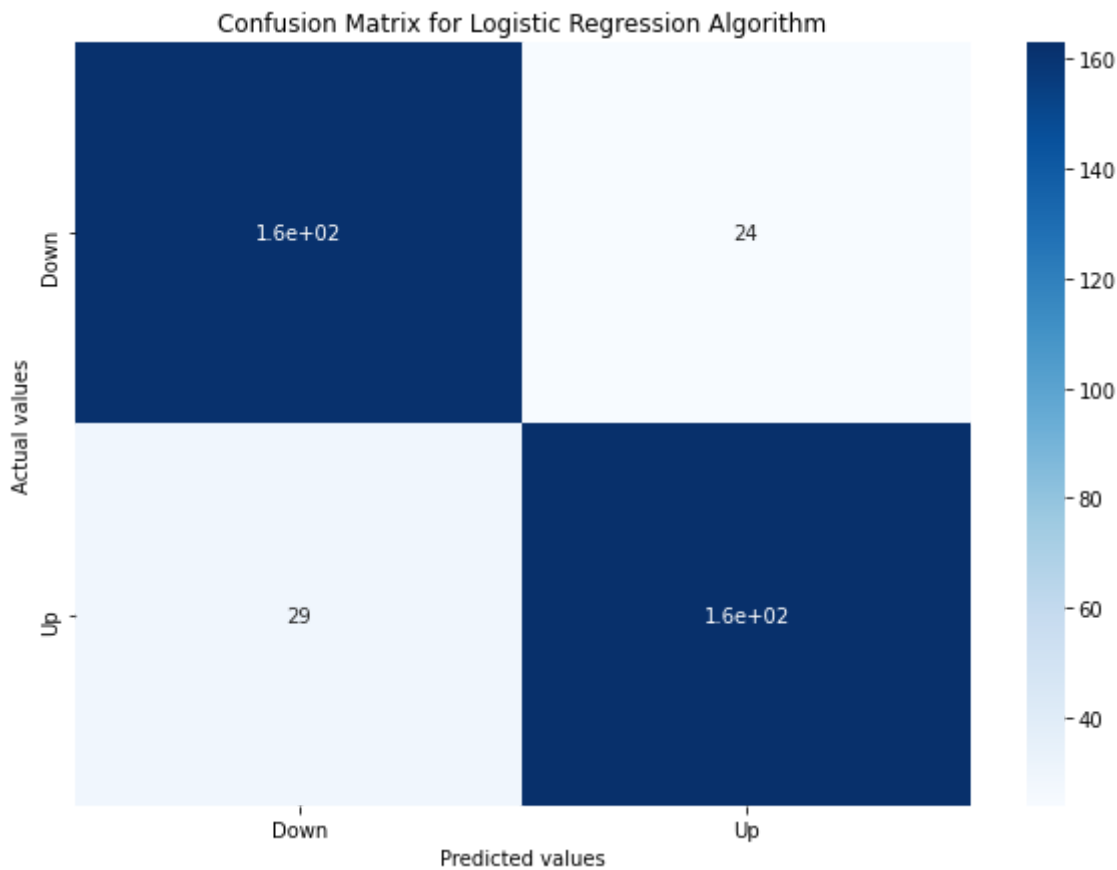
```
lr_cm
```

Out[39]:

```
array([[162, 24],
       [ 29, 163]], dtype=int64)
```

In [40]:

```
# Plotting the confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(data=lr_cm, annot=True, cmap="Blues", xticklabels=['Down', 'Up'], yticklabel
plt.xlabel('Predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Logistic Regression Algorithm')
plt.show()
```



Random Forest Classifier

In [41]:

```
from sklearn.ensemble import RandomForestClassifier
rf_classifier = RandomForestClassifier(n_estimators=100, criterion='entropy')
rf_classifier.fit(X_train, y_train)
```

Out[41]:

RandomForestClassifier(criterion='entropy')

In [42]:

```
rf_y_pred = rf_classifier.predict(X_test)
```

In [43]:

```
# Accuracy, Precision and Recall
score1 = accuracy_score(y_test, rf_y_pred)
score2 = precision_score(y_test, rf_y_pred)
score3 = recall_score(y_test, rf_y_pred)
print("---- Scores ----")
print("Accuracy score is: {}".format(round(score1*100,2)))
print("Precision score is: {}".format(round(score2,2)))
print("Recall score is: {}".format(round(score3,2)))
```

```
---- Scores ----
Accuracy score is: 82.8%
Precision score is: 0.82
Recall score is: 0.85
```

In [44]:

```
# Making the Confusion Matrix
rf_cm = confusion_matrix(y_test, rf_y_pred)
```

In [45]:

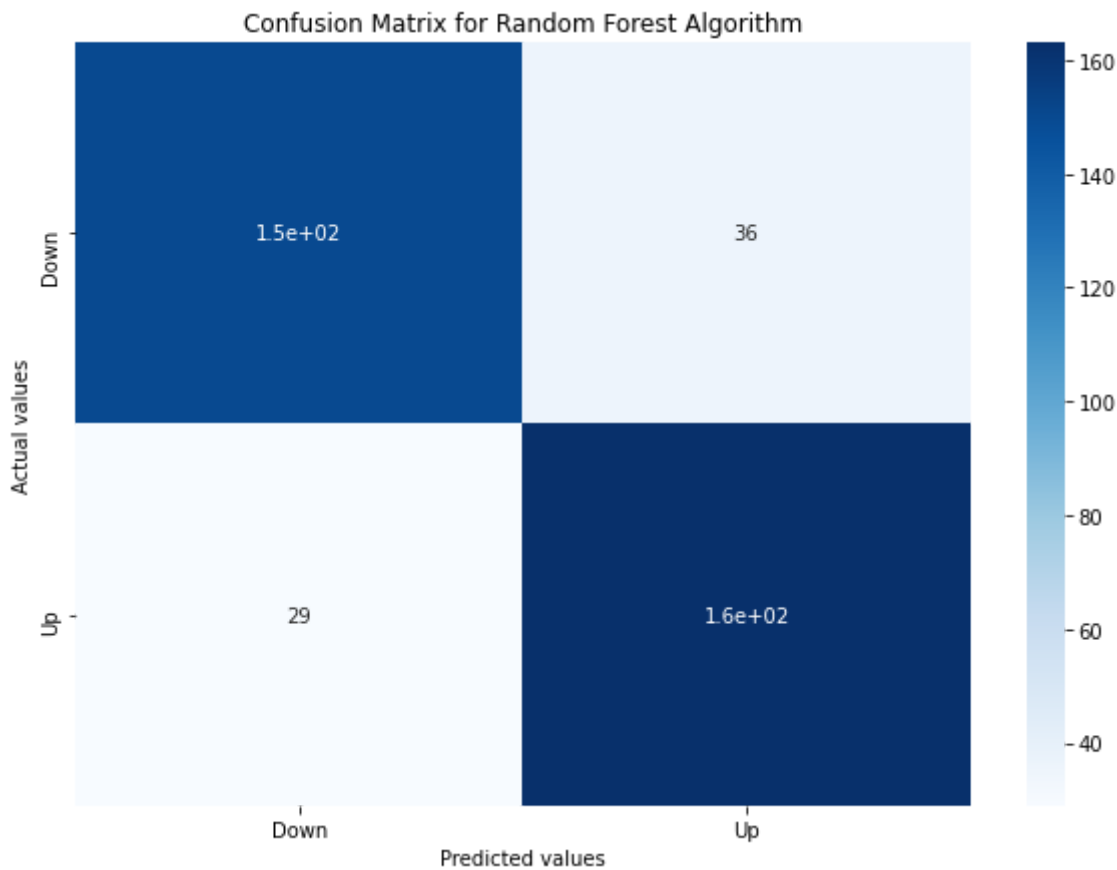
```
rf_cm
```

Out[45]:

```
array([[150,  36],
       [ 29, 163]], dtype=int64)
```

In [46]:

```
# Plotting the confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(data=rf_cm, annot=True, cmap="Blues", xticklabels=['Down', 'Up'], yticklabel
plt.xlabel('Predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Random Forest Algorithm')
plt.show()
```



Multinomial Naive Bayes

In [47]:

```
from sklearn.naive_bayes import MultinomialNB
nb_classifier = MultinomialNB()
nb_classifier.fit(X_train, y_train)
```

Out[47]:

MultinomialNB()

In [48]:

```
# Predicting the Test set results
nb_y_pred = nb_classifier.predict(X_test)
```


In [49]:

```
# Accuracy, Precision and Recall
score1 = accuracy_score(y_test, nb_y_pred)
score2 = precision_score(y_test, nb_y_pred)
score3 = recall_score(y_test, nb_y_pred)
print("---- Scores ----")
print("Accuracy score is: {}".format(round(score1*100,2)))
print("Precision score is: {}".format(round(score2,2)))
print("Recall score is: {}".format(round(score3,2)))
```

```
---- Scores ----
Accuracy score is: 83.86%
Precision score is: 0.85
Recall score is: 0.83
```

In [50]:

```
# Making the Confusion Matrix
nb_cm = confusion_matrix(y_test, nb_y_pred)
```

In [51]:

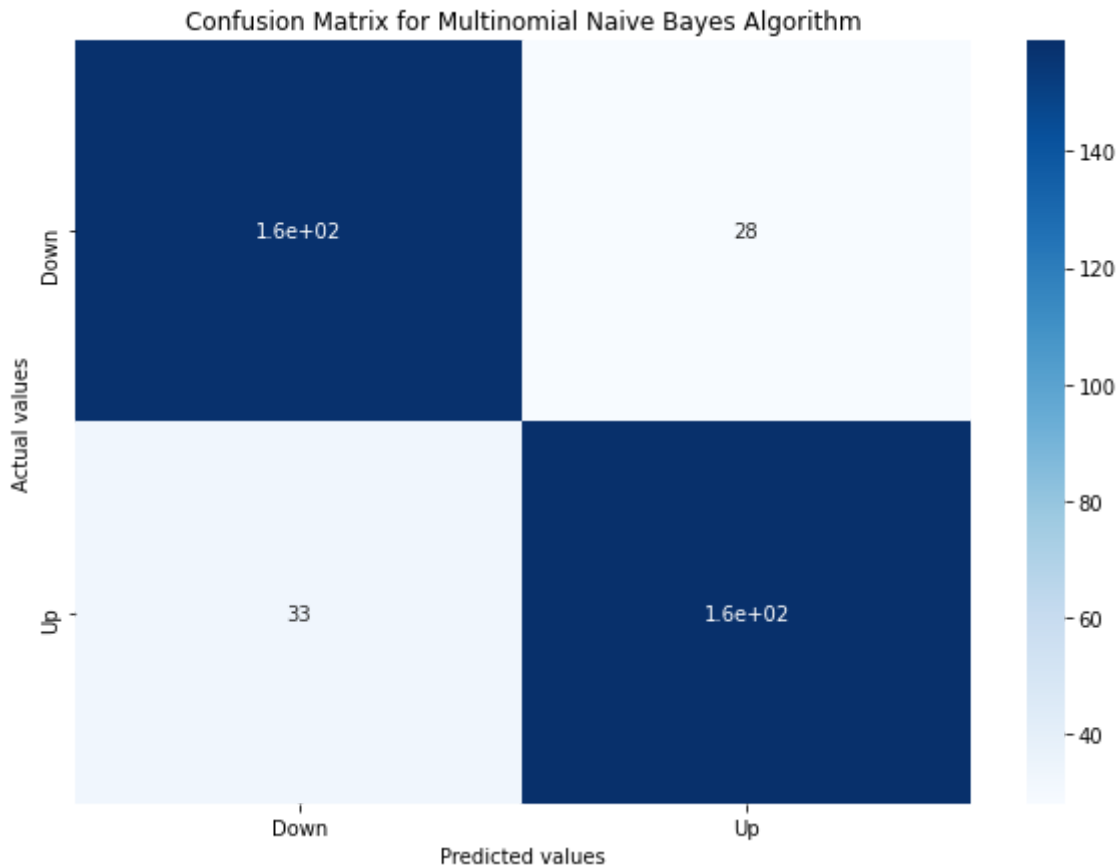
```
nb_cm
```

Out[51]:

```
array([[158, 28],
       [ 33, 159]], dtype=int64)
```

In [52]:

```
# Plotting the confusion matrix
plt.figure(figsize=(10,7))
sns.heatmap(data=nb_cm, annot=True, cmap="Blues", xticklabels=['Down', 'Up'], yticklabel
plt.xlabel('Predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Multinomial Naive Bayes Algorithm')
plt.show()
```



Predictions

In [53]:

```
import re

def stock_prediction(sample_news):
    sample_news = re.sub(pattern='[^a-zA-Z]', repl=' ', string=sample_news)
    sample_news = sample_news.lower()
    sample_news_words = sample_news.split()
    sample_news_words = [word for word in sample_news_words if not word in set(stopwords.w
    ps = PorterStemmer()
    final_news = [ps.stem(word) for word in sample_news_words]
    final_news = ' '.join(final_news)

    temp = cv.transform([final_news]).toarray()
    return lr_classifier.predict(temp)
```

In [54]:

```
# For generating random integer
from random import randint
```

In [55]:

```
sample_test = df_copy[df_copy['Date'] > '20141231']
```

In [56]:

```
sample_test.reset_index(inplace=True)
sample_test = sample_test['Top1']
```

In [57]:

```
# Predicting values
row = randint(0,sample_test.shape[0]-1)
sample_news = sample_test[row]

print('News: {}'.format(sample_news))
if stock_prediction(sample_news):
    print('Prediction: The stock price will remain the same or will go down.')
else:
    print('Prediction: The stock price will go up!')
```

News: El Chapo' Being Taken to Same Prison He Escaped from Six Months Ago
Prediction: The stock price will remain the same or will go down.

In [58]:

```
# Predicting values
row = randint(0,sample_test.shape[0]-1)
sample_news = sample_test[row]

print('News: {}'.format(sample_news))
if stock_prediction(sample_news):
    print('Prediction: The stock price will remain the same or will go down.')
else:
    print('Prediction: The stock price will go up!')
```

News: Efficiency up, turnover down: Sweden experiments with six-hour working day | World news | The Guardian
Prediction: The stock price will remain the same or will go down.

In [59]:

```
# Predicting values
row = randint(0,sample_test.shape[0]-1)
sample_news = sample_test[row]

print('News: {}'.format(sample_news))
if stock_prediction(sample_news):
    print('Prediction: The stock price will remain the same or will go down.')
else:
    print('Prediction: The stock price will go up!')
```

News: US State Dept declares ISIS is committing genocide in Iraq, Syria
Prediction: The stock price will remain the same or will go down.

In [60]:

```
# Predicting values
row = randint(0,sample_test.shape[0]-1)
sample_news = ""
print('News: {}'.format(sample_news))
if stock_prediction(sample_news):
    print('Prediction: The stock price will remain the same or will go down.')
else:
    print('Prediction: The stock price will go up!')
```

News:
Prediction: The stock price will remain the same or will go down.

In [61]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

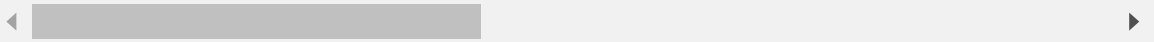
In [62]:

```
df=pd.read_csv('stock headlines.csv', encoding="ISO-8859-1")
df.head()
```

Out[62]:

	Date	Label	Top1	Top2	Top3	Top4	Top5	Top6	
0	2000-01-03	0	A 'hindrance to operations': extracts from the...	Scorecard	Hughes' instant hit buoys Blues	Jack gets his skates on at ice-cold Alex	Chaos as Maracana builds up for United	Depleted Leicester prevail as Elliott spoils E...	Hu S s picl
1	2000-01-04	0	Scorecard	The best lake scene	Leader: German sleaze inquiry	Cheerio, boyo	The main recommendations	Has Cubie killed fees?	C l f
2	2000-01-05	0	Coventry caught on counter by Flo	United's rivals on the road to Rio	Thatcher issues defence before trial by video	Police help Smith lay down the law at Everton	Tale of Trautmann bears two more retellings	England on the rack	Pak ret: witl for \ of V
3	2000-01-06	1	Pilgrim knows how to progress	Thatcher facing ban	Mcllroy calls for Irish fighting spirit	Leicester bin stadium blueprint	United braced for Mexican wave	Auntie back in fashion, even if the dress look...	St at go th
4	2000-01-07	1	Hitches and Horlocks	Beckham off but United survive	Breast cancer screening	Alan Parker	Guardian readers: are you all whingers?	Hollywood Beyond	A diam

5 rows × 27 columns



In [63]:

```
train=df[df['Date']<'20150101']
test=df[df['Date']>'20141231']
train.shape
```

Out[63]:

(3975, 27)

In [64]:

```
#Removing punctuations
data=train.iloc[:,2:27]
data.replace("[^a-zA-Z]", " ",regex=True, inplace=True)
```

In [65]:

```
data.columns
```

Out[65]:

```
Index(['Top1', 'Top2', 'Top3', 'Top4', 'Top5', 'Top6', 'Top7', 'Top8', 'Top9',
      'Top10', 'Top11', 'Top12', 'Top13', 'Top14', 'Top15', 'Top16', 'Top17',
      'Top18', 'Top19', 'Top20', 'Top21', 'Top22', 'Top23', 'Top24', 'Top25'],
      dtype='object')
```

In [66]:

```
for col in data.columns:
    data[col]=data[col].str.lower()
data.head(1)
```

Out[66]:

	Top1	Top2	Top3	Top4	Top5	Top6	Top7	Top8	Top9	Top10
0	a hindrance to operations extracts from the...	scorecard	hughes instant hit buoys blues	jack gets his skates on at ice cold alex	chaos as maracana builds up for united	depleted leicester prevail as elliott spoils e...	hungry spurs sense rich pickings	gunners so wide of an easy target	derby raise a glass to strupar s debut double	southg strik leads p 1 pen

1 rows × 25 columns

In [67]:

```
headlines = []
for row in range(0,len(data.index)):
    headlines.append(' '.join(str(x) for x in data.iloc[row,0:25]))
```

Using TF-IDF

In [68]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.ensemble import RandomForestClassifier
```

In [69]:

```
#implement TF-IDF
tfvector=TfidfVectorizer(ngram_range=(2,3))
train_df=tfvector.fit_transform(headlines)
```

In [70]:

```
import pickle
pickle.dump(tfvector, open('tfvector.pkl', 'wb'))
```

RandomForestClassifier

In [71]:

```
# implement RandomForest Classifier
randomclassifier=RandomForestClassifier(n_estimators=200,criterion='entropy')
randomclassifier.fit(train_df,train['Label'])
```

Out[71]:

```
RandomForestClassifier(criterion='entropy', n_estimators=200)
```

plot_confusion_matrix

In [72]:

```

from sklearn import metrics
import itertools
def plot_confusion_matrix(cm, classes,
                           normalize=False,
                           title='Confusion matrix',
                           cmap=plt.cm.Blues):
    """
    See full source and example:
    http://scikit-learn.org/stable/auto\_examples/model\_selection/plot\_confusion\_matrix.h

    This function prints and plots the confusion matrix.
    Normalization can be applied by setting `normalize=True`.
    """
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation=45)
    plt.yticks(tick_marks, classes)

    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')

    thresh = cm.max() / 2.
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        plt.text(j, i, cm[i, j],
                 horizontalalignment="center",
                 color="white" if cm[i, j] > thresh else "black")

    plt.tight_layout()
    plt.ylabel('True label')
    plt.xlabel('Predicted label')

```

In [73]:

```

# Predict for the Test Dataset
test_transform= []
for row in range(0,len(test.index)):
    test_transform.append(' '.join(str(x) for x in test.iloc[row,2:27]))
test_dataset = tfvector.transform(test_transform)
predictions = randomclassifier.predict(test_dataset)

```


In [74]:

```

from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
matrix=confusion_matrix(test['Label'], predictions)
print(matrix)
score=accuracy_score(test['Label'], predictions)
print(score)
report=classification_report(test['Label'], predictions)
print(report)
plot_confusion_matrix(matrix, classes=['Down', 'Up'])

```

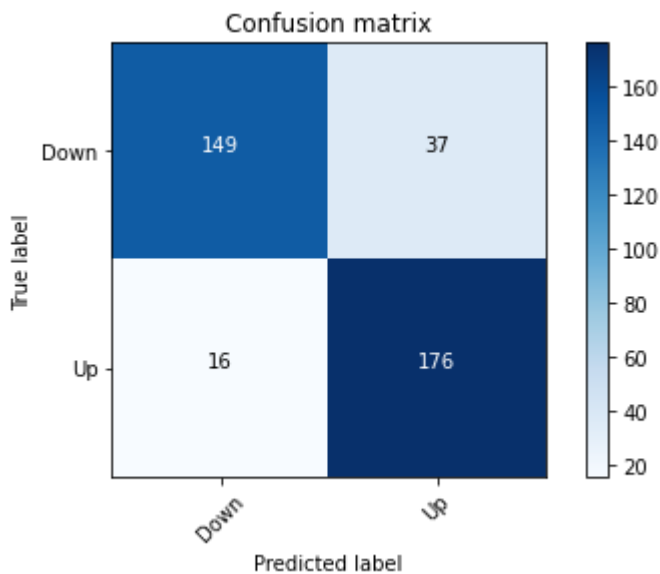
```

[[149  37]
 [ 16 176]]
0.8597883597883598

```

	precision	recall	f1-score	support
0	0.90	0.80	0.85	186
1	0.83	0.92	0.87	192
accuracy			0.86	378
macro avg	0.86	0.86	0.86	378
weighted avg	0.86	0.86	0.86	378

Confusion matrix, without normalization



MultinomialNB

In [75]:

```

from sklearn.naive_bayes import MultinomialNB
nb=MultinomialNB()
nb.fit(train_df, train['Label'])

```

Out[75]:

MultinomialNB()

In [76]:

```

predictions = nb.predict(test_dataset)
matrix=confusion_matrix(test['Label'],predictions)
print(matrix)
score=accuracy_score(test['Label'],predictions)
print(score)
report=classification_report(test['Label'],predictions)
print(report)
plot_confusion_matrix(matrix, classes=['Down', 'Up'])

```

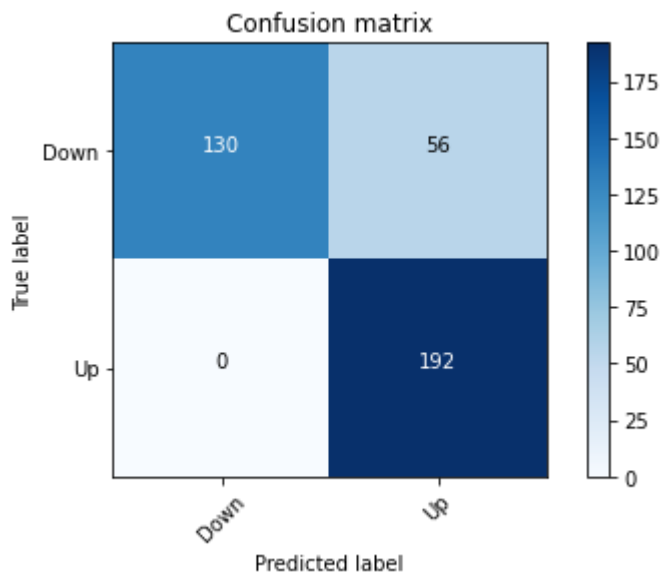
```

[[130  56]
 [   0 192]]
0.8518518518518519

```

	precision	recall	f1-score	support
0	1.00	0.70	0.82	186
1	0.77	1.00	0.87	192
accuracy			0.85	378
macro avg	0.89	0.85	0.85	378
weighted avg	0.89	0.85	0.85	378

Confusion matrix, without normalization



PassiveAggressiveClassifier

In [77]:

```

from sklearn.linear_model import PassiveAggressiveClassifier
pa = PassiveAggressiveClassifier()

pa.fit(train_df,train['Label'])

```

Out[77]:

PassiveAggressiveClassifier()

In [78]:

```
import pickle
filename = 'stock_senti.pkl'
pickle.dump(pa, open(filename, 'wb'))
```

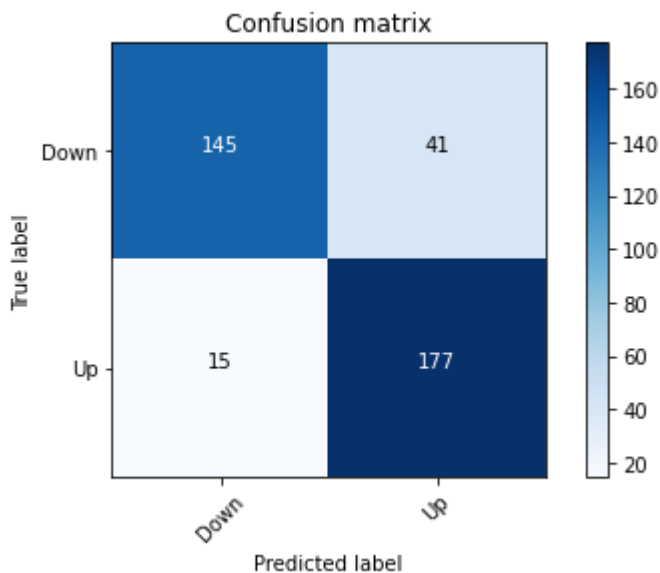
In [79]:

```
predictions = pa.predict(test_dataset)
matrix=confusion_matrix(test['Label'],predictions)
print(matrix)
score=accuracy_score(test['Label'],predictions)
print(score)
report=classification_report(test['Label'],predictions)
print(report)
plot_confusion_matrix(matrix, classes=['Down', 'Up'])
```

```
[[145  41]
 [ 15 177]]
0.8518518518518519
```

	precision	recall	f1-score	support
0	0.91	0.78	0.84	186
1	0.81	0.92	0.86	192
accuracy			0.85	378
macro avg	0.86	0.85	0.85	378
weighted avg	0.86	0.85	0.85	378

Confusion matrix, without normalization



Using bag of words

In [80]:

```
from sklearn.feature_extraction.text import CountVectorizer
#implement bag of words
bow=CountVectorizer(ngram_range=(2,3))
train_df=bow.fit_transform(headlines)
```

RandomForestClassifier using Bag of words

In [81]:

```
# implement RandomForest Classifier
randomclassifier=RandomForestClassifier(n_estimators=200,criterion='entropy')
randomclassifier.fit(train_df,train['Label'])
```

Out[81]:

```
RandomForestClassifier(criterion='entropy', n_estimators=200)
```

In [82]:

```
predictions = randomclassifier.predict(test_dataset)
matrix=confusion_matrix(test['Label'],predictions)
print(matrix)
score=accuracy_score(test['Label'],predictions)
print(score)
report=classification_report(test['Label'],predictions)
print(report)
```

```
[[ 0 186]
 [ 0 192]]
0.5079365079365079
```

	precision	recall	f1-score	support
0	0.00	0.00	0.00	186
1	0.51	1.00	0.67	192
accuracy			0.51	378
macro avg	0.25	0.50	0.34	378
weighted avg	0.26	0.51	0.34	378

```
C:\Users\psai1\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1248: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\psai1\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1248: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

```
C:\Users\psai1\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1248: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
```

```
_warn_prf(average, modifier, msg_start, len(result))
```

MultinomialNB using Bag of words

In [83]:

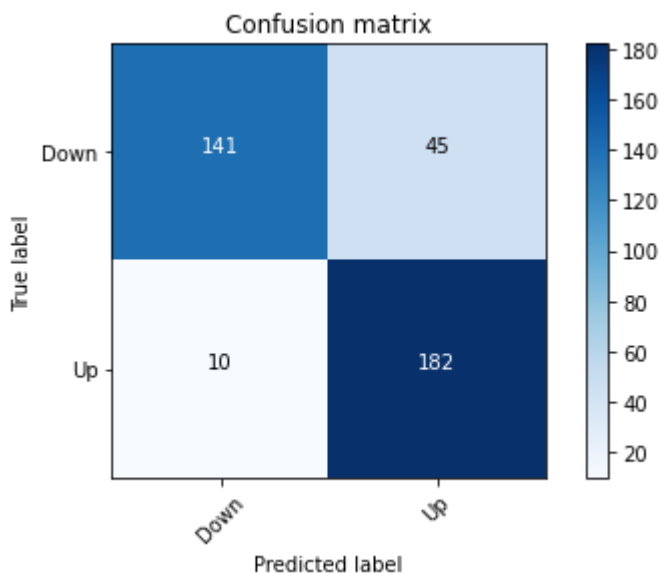
```
from sklearn.naive_bayes import MultinomialNB
nb=MultinomialNB()
nb.fit(train_df,train['Label'])

predictions = nb.predict(test_dataset)
matrix=confusion_matrix(test['Label'],predictions)
print(matrix)
score=accuracy_score(test['Label'],predictions)
print(score)
report=classification_report(test['Label'],predictions)
print(report)
plot_confusion_matrix(matrix, classes=['Down', 'Up'])
```

```
[[141  45]
 [ 10 182]]
0.8544973544973545
```

	precision	recall	f1-score	support
0	0.93	0.76	0.84	186
1	0.80	0.95	0.87	192
accuracy			0.85	378
macro avg	0.87	0.85	0.85	378
weighted avg	0.87	0.85	0.85	378

Confusion matrix, without normalization



PassiveAggressiveClassifier using Bag of Words

In [84]:

```

from sklearn.linear_model import PassiveAggressiveClassifier
pa = PassiveAggressiveClassifier()
pa.fit(train_df,train['Label'])

predictions = pa.predict(test_dataset)
matrix=confusion_matrix(test['Label'],predictions)
print(matrix)
score=accuracy_score(test['Label'],predictions)
print(score)
report=classification_report(test['Label'],predictions)
print(report)
plot_confusion_matrix(matrix, classes=['Down', 'Up'])

```

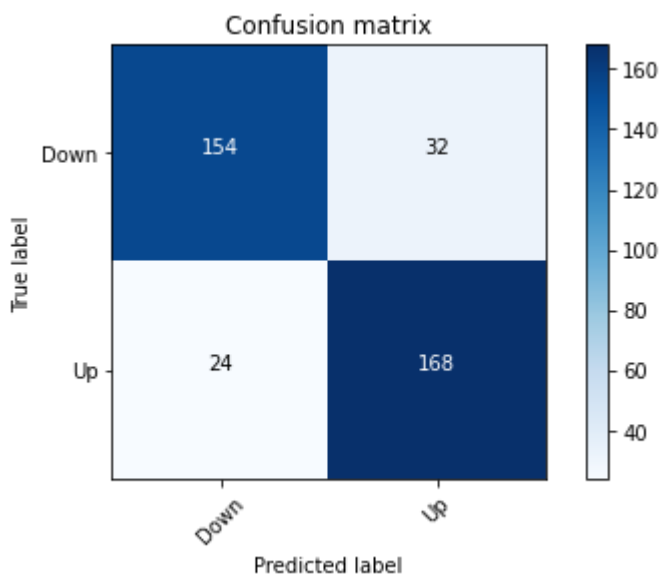
```

[[154  32]
 [ 24 168]]
0.8518518518518519

```

	precision	recall	f1-score	support
0	0.87	0.83	0.85	186
1	0.84	0.88	0.86	192
accuracy			0.85	378
macro avg	0.85	0.85	0.85	378
weighted avg	0.85	0.85	0.85	378

Confusion matrix, without normalization



Spark

In [85]:

```

import findspark
findspark.init()
import pyspark
from pyspark.sql.session import SparkSession
spark = SparkSession.builder.appName("Yahoo finance").getOrCreate()

```

In [86]:

```
import yfinance as yf
import csv
from pyspark.sql.functions import date_format

from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear

ticker = '^GSPC' # tsla
start_dt = '2021-1-1'
end_dt = '2022-12-31'
frequency = '1d'

# get data for SPX
data = yf.Ticker(ticker)
# ticker_name = data.info['LongName']

# get weekly historical prices for this ticker
df = data.history(interval = frequency, start = start_dt, end = end_dt)

# write data into a csv file
df.to_csv('spx.csv')

# Let Spark know about the header and infer the Schema types!
df = spark.read.csv('spx.csv', inferSchema=True, header=True)

# create new dataframe month_df adding new columns month, diff, high_low_diff and weekofyear
daily_df = df.withColumn('Month', month(df['Date'])).withColumn('diff', df['Open'] - df['Close'])
monthly_df = daily_df.withColumn('week_of_year', weekofyear(daily_df.Date)).withColumn('week_day', date_format(daily_df.Date, 'dd'))
monthly_df.show(5)
```

```

+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+
--+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|      Volume|Dividends|Stock Splits|Month|
diff|  high_low_diff|week_of_year| week_day|
+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+
|2021-01-04 00:00:...|3764.610107421875|3769.989990234375| 3662.709960937
5| 3700.64990234375|501500000| 0.0| 0.0| 1| 63.9602050781
25|107.280029296875| 1| Monday|
|2021-01-05 00:00:...| 3698.02001953125| 3737.830078125|3695.07006835937
5|3726.860107421875|459102000| 0.0| 0.0| 1|-28.8400878906
25| 42.760009765625| 1| Tuesday|
|2021-01-06 00:00:...|3712.199951171875| 3783.0400390625|3705.34008789062
5|3748.139892578125|606411000| 0.0| 0.0| 1| -35.939941406
25| 77.699951171875| 1|Wednesday|
|2021-01-07 00:00:...| 3764.7099609375|3811.550048828125| 3764.709960937
5| 3803.7900390625|509916000| 0.0| 0.0| 1| -39.0800781
25| 46.840087890625| 1| Thursday|
|2021-01-08 00:00:...|3815.050048828125| 3826.68994140625| 3783.6000976562
5|3824.679931640625|477304000| 0.0| 0.0| 1| -9.62988281
25| 43.08984375| 1| Friday|
+-----+-----+-----+-----+
-+-----+-----+-----+-----+
--+-----+-----+-----+

```

only showing top 5 rows

In [87]:

```

daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where ((diff > 20 or diff < -20) and (Date > "2022-01-01"))')

```

```

+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+
--+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|      Volume|Dividends|Stock Splits|Month|
diff|  high_low_diff|week_of_year| week_day|
+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+
--+-----+-----+-----+
|2022-01-05 00:00:...| 4787.990234375| 4797.7001953125|4699.4399414062
5| 4700.580078125|488796000| 0.0| 0.0| 1| 87.4101
5625| 98.26025390625| 1|Wednesday|
|2022-01-07 00:00:...| 4697.66015625| 4707.9501953125| 4662.74023437
5|4677.02978515625|418151000| 0.0| 0.0| 1| 20.6303710
9375| 45.2099609375| 1| Friday|
|2022-01-11 00:00:...|4669.14013671875| 4714.1298828125|4638.2700195312
5|4713.06982421875|410159000| 0.0| 0.0| 1| -43.929
6875| 75.85986328125| 2| Tuesday|
|2022-01-13 00:00:...|4733.56005859375| 4744.1298828125| 4650.290039062
5|4650.000000000|425173000| 0.0| 0.0| 1| 74.5000000
000|

```


In [88]:

```
# total days when market was open in 2022
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where ((Date > "2021-01-01 00:00:00") and (Date < "202
```

Out[88]:

252

In [89]:

```
# total days when market was open in 2022
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where Date > "2022-01-01 00:00:00").count()
```

Out[89]:

251

In [90]:

```
#number of days where spx moved by 20 (difference between open and close) in 2021
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where ((diff > 20 or diff < -20) \
and ((Date > "2021-01-01 00:00:00") and ((Date < "2022-01-01 00:00:00"))))').count()
```

Out[90]:

118

In [91]:

```
#number of days where spx moved by 20 (difference between open and close) in 2022
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where ((diff > 20 or diff < -20)) \
and ((Date > "2022-01-01 00:00:00"))').count()
```

Out[91]:

184

In [92]:

```
#number of days where spx moved by less 20 (difference between open and close)
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where ((diff < 20 and diff > -20) and (Date > "2022-01
```

Out[92]:

67

In [93]:

```
#number of days where spx moved by 20 (difference between high and low of the day)  
daily_df.createOrReplaceTempView("SPXDaily")  
spark.sql('select * from SPXDaily where ((high_low_diff > 20 or high_low_diff < -20) and
```

Out[93]:

250

In [94]:


```
# number of days when spx moved by 40 (difference between high and low of the day)  
daily_df.createOrReplaceTempView("SPXDaily")  
spark.sql('select * from SPXDaily where ((high_low_diff > 40 or high_low_diff < -40) and
```

Out[94]:

227

In [95]:

```
daily_df.createOrReplaceTempView("SPXDaily")  
spark.sql('select * from SPXDaily where ((high_low_diff > 40 or high_low_diff < -40) and
```



```

+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+
|      Date|      Open|      High|      Low|
Close|  Volume|Dividends|Stock Splits|Month|      diff|  high_low_d
iff|week_of_year| week_day|
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+
|2022-01-04 00:00:...| 4804.509765625| 4818.6201171875|4774.27001953125|
4793.5400390625|4683170000| 0.0| 0.0| 1| 10.9697265625| 4
4.35009765625| 1| Tuesday|
|2022-01-05 00:00:...| 4787.990234375| 4797.7001953125|4699.43994140625|
4700.580078125|4887960000| 0.0| 0.0| 1| 87.41015625| 9
8.26025390625| 1| Wednesday|
|2022-01-06 00:00:...|4693.39013671875| 4725.009765625| 4671.259765625|
4696.0498046875|4295280000| 0.0| 0.0| 1| -2.65966796875|
53.75| 1| Thursday|
|2022-01-07 00:00:...| 4697.66015625| 4707.9501953125| 4662.740234375|4
677.02978515625|4181510000| 0.0| 0.0| 1| 20.63037109375|
45.2099609375| 1| Friday|
|2022-01-10 00:00:...| 4655.33984375|4673.02001953125| 4582.240234375|
4670.2900390625|4511810000| 0.0| 0.0| 1| -14.9501953125| 9
0.77978515625| 2| Monday|
|2022-01-11 00:00:...|4669.14013671875| 4714.1298828125|4638.27001953125|4
713.06982421875|4101590000| 0.0| 0.0| 1| -43.9296875| 7
5.85986328125| 2| Tuesday|
|2022-01-12 00:00:...| 4728.58984375| 4748.830078125| 4706.7099609375|4
726.35009765625|4048220000| 0.0| 0.0| 1| 2.23974609375|
42.1201171875| 2| Wednesday|
|2022-01-13 00:00:...|4733.56005859375| 4744.1298828125| 4650.2900390625|4
659.02978515625|4251730000| 0.0| 0.0| 1| 74.5302734375|
93.83984375| 2| Thursday|
|2022-01-14 00:00:...| 4637.990234375| 4665.1298828125| 4614.75|4
662.85009765625|4338490000| 0.0| 0.0| 1| -24.85986328125|
50.3798828125| 2| Friday|
|2022-01-18 00:00:...| 4632.240234375| 4632.240234375| 4568.7001953125|4
577.10986328125|4748700000| 0.0| 0.0| 1| 55.13037109375|
63.5400390625| 3| Tuesday|
|2022-01-19 00:00:...|4588.02978515625| 4611.5498046875| 4530.2001953125|
4532.759765625|4465740000| 0.0| 0.0| 1| 55.27001953125|
81.349609375| 3| Wednesday|
|2022-01-20 00:00:...|4547.35009765625|4602.10986328125| 4477.9501953125|4
482.72998046875|4640870000| 0.0| 0.0| 1| 64.6201171875|12
4.15966796875| 3| Thursday|
|2022-01-21 00:00:...| 4471.3798828125|4494.52001953125| 4395.33984375|4
397.93994140625|5589100000| 0.0| 0.0| 1| 73.43994140625| 9
9.18017578125| 3| Friday|
|2022-01-24 00:00:...|4356.31982421875|4417.35009765625| 4222.6201171875|
4410.1298828125|6928110000| 0.0| 0.0| 1| -53.81005859375|19
4.72998046875| 4| Monday|
|2022-01-25 00:00:...|4366.64013671875| 4411.009765625|4287.10986328125|
4356.4501953125|5145050000| 0.0| 0.0| 1| 10.18994140625|12
3.89990234375| 4| Tuesday|
|2022-01-26 00:00:...|4408.43017578125|4453.22998046875| 4304.7998046875|4
349.93017578125|5570640000| 0.0| 0.0| 1| 58.5|14
8.43017578125| 4| Wednesday|
|2022-01-27 00:00:...| 4380.580078125| 4428.740234375| 4309.5|
4326.509765625|5214200000| 0.0| 0.0| 1| 54.0703125| 1
19.240234375| 4| Thursday|
|2022-01-28 00:00:...|4336.18994140625|4432.72021484375| 4292.4599609375|4

```

431.85009765625	5031090000	0.0	0.0	1	-95.66015625 14
0.26025390625	4	Friday			
2022-01-31 00:00:...	4431.7900390625	4516.89013671875	4414.02001953125		
4515.5498046875	5098610000	0.0	0.0	1	-83.759765625 1
02.8701171875	5	Monday			
2022-02-01 00:00:...	4519.56982421875	4550.490234375	4483.52978515625		
4546.5400390625	4816830000	0.0	0.0	2	-26.97021484375 6
6.96044921875	5	Tuesday			

+-----+-----+-----+-----+-
-----+-----+-----+-----+-
-----+-----+-----+

only showing top 20 rows

In [96]:

```
# number of days in each week where spx moved by 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select week_of_year, count(*) from SPXDaily where ((diff > 40 or diff < -40)
and (Date > "2022-01-01 00:00:00")) \
group by week_of_year \
order by week_of_year').show(50)
```

week_of_year	count(1)
1	1
2	2
3	4
4	4
5	2
6	4
7	2
8	3
9	2
10	3
11	4
12	2
13	1
14	1
15	4
16	3
17	4
18	2
19	3
20	1
21	4
22	2
23	3
24	2
25	2
26	2
27	1
28	1
29	3
30	3
31	1
32	1
34	3
35	3
36	3
37	1
38	2
39	3
40	3
41	2
42	1
43	2
44	2
45	2
48	2
49	2
50	3
52	2

In [97]:

```
# total weeks so far this year
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select max(week_of_year) from SPXDaily').show()
```

```
+-----+
|max(week_of_year)|
+-----+
|                52|
+-----+
```


In [98]:

```
# number of days in each week where spx moved by less than 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select week_of_year, count(*) from SPXDaily where ((diff < 40 and diff > -40)
and (Date > "2022-01-01 00:00:00")) \
group by week_of_year \
order by week_of_year').show(50)
```

week_of_year	count(1)
1	4
2	3
4	1
5	3
6	1
7	3
8	1
9	3
10	2
11	1
12	3
13	4
14	4
16	2
17	1
18	3
19	2
20	4
21	1
22	2
23	2
24	3
25	2
26	3
27	3
28	4
29	2
30	2
31	4
32	4
33	5
34	2
35	2
36	1
37	4
38	3
39	2
40	2
41	3
42	4
43	3
44	3
45	3
46	5
47	4
48	3
49	3
50	2
51	5
52	2

In [99]:

```
# check why in week 47, there were only 2 days over 40/-40 and 2 days were between -40 and 40
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily where week_of_year = 47').show()
```

```
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|          Volume|Dividends|Stock Splits|Month|          di
ff| high_low_diff|week_of_year| week_day|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|2021-11-22 00:00:...|          4712.0|  4743.830078125|  4682.16992187
5| 4682.93994140625|4441100000|          0.0|          0.0|  11|  29.060058593
75|   61.66015625|          47|   Monday|
|2021-11-23 00:00:...| 4678.47998046875| 4699.39013671875|  4652.6601562
5| 4690.7001953125|4277590000|          0.0|          0.0|  11| -12.220214843
75| 46.72998046875|          47|   Tuesday|
|2021-11-24 00:00:...| 4675.77978515625| 4702.8701171875| 4659.8901367187
5| 4701.4599609375|3418430000|          0.0|          0.0|  11| -25.680175781
25| 42.97998046875|          47|Wednesday|
|2021-11-26 00:00:...| 4664.6298828125| 4664.6298828125| 4585.4301757812
5| 4594.6201171875|3517700000|          0.0|          0.0|  11|   70.0097656
25| 79.19970703125|          47|   Friday|
|2022-11-21 00:00:...| 3956.22998046875|          3962.0|3933.34008789062
5| 3949.93994140625|3850690000|          0.0|          0.0|  11|   6.29003906
25|28.659912109375|          47|   Monday|
|2022-11-22 00:00:...|3965.510009765625| 4005.8798828125| 3956.879882812
5| 4003.580078125|3887990000|          0.0|          0.0|  11| -38.0700683593
75|          49.0|          47|   Tuesday|
|2022-11-23 00:00:...|4000.300048828125|4033.780029296875|3998.65991210937
5|4027.260009765625|3279720000|          0.0|          0.0|  11| -26.95996093
75|  35.1201171875|          47|Wednesday|
|2022-11-25 00:00:...|4023.340087890625| 4034.02001953125|4020.76000976562
5| 4026.1201171875|1706460000|          0.0|          0.0|  11| -2.7800292968
75|13.260009765625|          47|   Friday|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
```

In [100]:

```
#number of days in each month where SPX moved by 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select month, count(*) from SPXDaily where ((diff > 40 or diff < -40) \
and (Date > "2022-01-01 00:00:00")) \
group by month \
order by month').show()
```

month	count(1)
1	12
2	10
3	12
4	12
5	10
6	10
7	9
8	7
9	10
10	8
11	6
12	7

In [101]:

```
#number of days in each month where SPX moved by less than 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select month, count(*) from SPXDaily where ((diff < 40 and diff > -40) \
and (Date > "2022-01-01 00:00:00")) \
group by month \
order by month').show(50)
```

month	count(1)
1	8
2	9
3	11
4	8
5	11
6	11
7	11
8	16
9	11
10	13
11	15
12	14

In [102]:

```
# number of days by week_day where spx moved by 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select week_day, count(*) from SPXDaily where ((diff > 40 or diff < -40) \
and (Date > "2022-01-01 00:00:00")) \
group by week_day \
order by week_day').show()
```

```
+-----+-----+
| week_day|count(1)|
+-----+-----+
|   Friday|      24|
|   Monday|      15|
| Thursday|      28|
|   Tuesday|      22|
| Wednesday|      24|
+-----+-----+
```

In [103]:

```
# number of days by week_day where spx moved less than 40 in either direction
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select week_day, count(*) from SPXDaily where ((diff < 40 and diff > -40) \
and (Date > "2022-01-01 00:00:00")) \
group by week_day \
order by week_day').show()
```

```
+-----+-----+
| week_day|count(1)|
+-----+-----+
|   Friday|      27|
|   Monday|      30|
| Thursday|      23|
|   Tuesday|      30|
| Wednesday|      28|
+-----+-----+
```

In [104]:

```

from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear,
from pyspark.sql.functions import date_format
from pyspark.sql.functions import col
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily \
          where ((diff between -10 and 10) \
                and (date_format(Date, "EEEE") == "Monday") \
                and year(Date) == 2022)').show()

```

```

+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|          Volume|Dividends|Stock Splits|Month|          diff
| high_low_diff|week_of_year|week_day|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
|2022-03-21 00:00:...| 4462.39990234375|          4481.75| 4424.299804687
5| 4461.18017578125|4869820000|          0.0|          0.0|    3| 1.2197265625
| 57.4501953125|          12| Monday|
|2022-04-18 00:00:...| 4385.6298828125| 4410.31005859375| 4370.299804687
5| 4391.68994140625|3910490000|          0.0|          0.0|    4|-6.06005859375
| 40.01025390625|          16| Monday|
|2022-05-16 00:00:...| 4013.02001953125| 4046.4599609375|3983.98999023437
5|4008.010009765625|4415030000|          0.0|          0.0|    5|5.010009765625
|62.469970703125|          20| Monday|
|2022-07-25 00:00:...|3965.719970703125|3975.300048828125| 3943.459960937
5|3966.840087890625|3568340000|          0.0|          0.0|    7| -1.1201171875
|31.840087890625|          30| Monday|
|2022-08-01 00:00:...| 4112.3798828125| 4144.9501953125| 4096.0200195312
5| 4118.6298828125|4202810000|          0.0|          0.0|    8|          -6.25
| 48.93017578125|          31| Monday|
|2022-08-29 00:00:...| 4034.580078125|4062.989990234375| 4017.41992187
5|4030.610107421875|3396510000|          0.0|          0.0|    8|3.969970703125
|45.570068359375|          35| Monday|
|2022-10-31 00:00:...| 3881.85009765625| 3893.72998046875|3863.17993164062
5| 3871.97998046875|4820620000|          0.0|          0.0|   10| 9.8701171875
|30.550048828125|          44| Monday|
|2022-11-21 00:00:...| 3956.22998046875|          3962.0|3933.34008789062
5| 3949.93994140625|3850690000|          0.0|          0.0|   11| 6.2900390625
|28.659912109375|          47| Monday|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+

```

In [105]:

```
from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear,
from pyspark.sql.functions import date_format
from pyspark.sql.functions import col
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select * from SPXDaily \
          where ((date_format(Date, "EEEE") == "Monday") and \
                year(Date) == 2022)').show(200)
```

```

+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+
--+-----+-----+-----+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|          Volume|Dividends|Stock Splits|Month|          di
ff|  high_low_diff|week_of_year|week_day|
+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+
|2022-01-03 00:00:...| 4778.14013671875| 4796.64013671875| 4758.16992187
5| 4796.56005859375|3831020000| 0.0| 0.0| 1| -18.4199218
75| 38.47021484375| 1| Monday|
|2022-01-10 00:00:...| 4655.33984375| 4673.02001953125| 4582.24023437
5| 4670.2900390625|4511810000| 0.0| 0.0| 1| -14.95019531
25| 90.77978515625| 2| Monday|
|2022-01-24 00:00:...| 4356.31982421875| 4417.35009765625| 4222.620117187
5| 4410.1298828125|6928110000| 0.0| 0.0| 1| -53.810058593
75| 194.72998046875| 4| Monday|
|2022-01-31 00:00:...| 4431.7900390625| 4516.89013671875| 4414.0200195312
5| 4515.5498046875|5098610000| 0.0| 0.0| 1| -83.7597656
25| 102.8701171875| 5| Monday|
|2022-02-07 00:00:...| 4505.75| 4521.85986328125| 4471.4702148437
5| 4483.8701171875|4228480000| 0.0| 0.0| 2| 21.87988281
25| 50.3896484375| 6| Monday|
|2022-02-14 00:00:...| 4412.60986328125| 4426.22021484375| 4364.8398437
5| 4401.669921875|4600390000| 0.0| 0.0| 2| 10.939941406
25| 61.38037109375| 7| Monday|
|2022-02-28 00:00:...| 4354.169921875| 4388.83984375| 4315.120117187
5| 4373.93994140625|6071370000| 0.0| 0.0| 2| -19.770019531
25| 73.7197265625| 9| Monday|
|2022-03-07 00:00:...| 4327.009765625| 4327.009765625| 4199.8500976562
5| 4201.08984375|6940470000| 0.0| 0.0| 3| 125.9199218
75| 127.15966796875| 10| Monday|
|2022-03-14 00:00:...| 4202.75| 4247.56982421875| 4161.7202148437
5| 4173.10986328125|5574920000| 0.0| 0.0| 3| 29.640136718
75| 85.849609375| 11| Monday|
|2022-03-21 00:00:...| 4462.39990234375| 4481.75| 4424.299804687
5| 4461.18017578125|4869820000| 0.0| 0.0| 3| 1.21972656
25| 57.4501953125| 12| Monday|
|2022-03-28 00:00:...| 4541.08984375| 4575.64990234375| 4517.6899414062
5| 4575.52001953125|4312260000| 0.0| 0.0| 3| -34.430175781
25| 57.9599609375| 13| Monday|
|2022-04-04 00:00:...| 4547.97021484375| 4583.5| 4539.209960937
5| 4582.64013671875|4547350000| 0.0| 0.0| 4| -34.6699218
75| 44.2900390625| 14| Monday|
|2022-04-11 00:00:...| 4462.64013671875| 4464.35009765625| 4408.379882812
5| 4412.52978515625|4266290000| 0.0| 0.0| 4| 50.11035156
25| 55.97021484375| 15| Monday|
|2022-04-18 00:00:...| 4385.6298828125| 4410.31005859375| 4370.299804687
5| 4391.68994140625|3910490000| 0.0| 0.0| 4| -6.060058593
75| 40.01025390625| 16| Monday|
|2022-04-25 00:00:...| 4255.33984375| 4299.02001953125| 4200.8198242187
5| 4296.1201171875|5240040000| 0.0| 0.0| 4| -40.78027343
75| 98.2001953125| 17| Monday|
|2022-05-02 00:00:...| 4130.60986328125| 4169.81005859375|4062.51000976562
5| 4155.3798828125|5163790000| 0.0| 0.0| 5| -24.770019531
25|107.300048828125| 18| Monday|
|2022-05-09 00:00:...| 4081.27001953125| 4081.27001953125| 3975.4799804687
5|3991.239990234375|5954520000| 0.0| 0.0| 5| 90.0300292968
75| 105.7900390625| 19| Monday|
|2022-05-16 00:00:...| 4013.02001953125| 4046.4599609375|3983.98999023437

```



```
5|4008.010009765625|4415030000|0.0|0.0|5|5.0100097656
25|62.469970703125|20|Monday|
|2022-05-23 00:00:...|3919.419921875|3981.8798828125|3909.040039062
5|3973.75|4420030000|0.0|0.0|5|-54.3300781
25|72.83984375|21|Monday|
|2022-06-06 00:00:...|4134.72021484375|4168.77978515625|4109.1801757812
5|4121.43017578125|4332700000|0.0|0.0|6|13.29003906
25|59.599609375|23|Monday|
|2022-06-13 00:00:...|3838.14990234375|3838.14990234375|3734.30004882812
5|3749.6298828125|5636890000|0.0|0.0|6|88.520019531
25|103.849853515625|24|Monday|
|2022-06-27 00:00:...|3920.760009765625|3927.719970703125|3889.65991210937
5|3900.110107421875|4325310000|0.0|0.0|6|20.649902343
75|38.06005859375|26|Monday|
|2022-07-11 00:00:...|3880.93994140625|3880.93994140625|3847.21997070312
5|3854.429931640625|3423480000|0.0|0.0|7|26.5100097656
25|33.719970703125|28|Monday|
|2022-07-18 00:00:...|3883.7900390625|3902.43994140625|3818.629882812
5|3830.85009765625|4046870000|0.0|0.0|7|52.939941406
25|83.81005859375|29|Monday|
|2022-07-25 00:00:...|3965.719970703125|3975.300048828125|3943.459960937
5|3966.840087890625|3568340000|0.0|0.0|7|-1.12011718
75|31.840087890625|30|Monday|
|2022-08-01 00:00:...|4112.3798828125|4144.9501953125|4096.0200195312
5|4118.6298828125|4202810000|0.0|0.0|8|-6.
25|48.93017578125|31|Monday|
|2022-08-08 00:00:...|4155.93017578125|4186.6201171875|4128.9702148437
5|4140.06005859375|4221090000|0.0|0.0|8|15.87011718
75|57.64990234375|32|Monday|
|2022-08-15 00:00:...|4269.3701171875|4301.7900390625|4256.8999023437
5|4297.14013671875|3696830000|0.0|0.0|8|-27.770019531
25|44.89013671875|33|Monday|
|2022-08-22 00:00:...|4195.080078125|4195.080078125|4129.8598632812
5|4137.990234375|3907430000|0.0|0.0|8|57.089843
75|65.22021484375|34|Monday|
|2022-08-29 00:00:...|4034.580078125|4062.989990234375|4017.41992187
5|4030.610107421875|3396510000|0.0|0.0|8|3.9699707031
25|45.570068359375|35|Monday|
|2022-09-12 00:00:...|4083.669921875|4119.27978515625|4083.66992187
5|4110.41015625|3814200000|0.0|0.0|9|-26.7402343
75|35.60986328125|37|Monday|
|2022-09-19 00:00:...|3849.909912109375|3900.449951171875|3838.
5|3899.889892578125|3766850000|0.0|0.0|9|-49.979980468
75|61.949951171875|38|Monday|
|2022-09-26 00:00:...|3682.719970703125|3715.669921875|3644.76000976562
5|3655.0400390625|4886140000|0.0|0.0|9|27.6799316406
25|70.909912109375|39|Monday|
|2022-10-03 00:00:...|3609.780029296875|3698.35009765625|3604.92993164062
5|3678.429931640625|4806680000|0.0|0.0|10|-68.649902343
75|93.420166015625|40|Monday|
|2022-10-10 00:00:...|3647.510009765625|3652.169921875|3588.1000976562
5|3612.389892578125|3834320000|0.0|0.0|10|35.12011718
75|64.06982421875|41|Monday|
|2022-10-17 00:00:...|3638.64990234375|3689.72998046875|3638.6499023437
5|3677.949951171875|4352780000|0.0|0.0|10|-39.3000488281
25|51.080078125|42|Monday|
|2022-10-24 00:00:...|3762.010009765625|3810.739990234375|3741.6499023437
5|3797.340087890625|4747930000|0.0|0.0|10|-35.3300781
25|69.090087890625|43|Monday|
|2022-10-31 00:00:...|3881.85009765625|3893.72998046875|3863.17993164062
5|3871.97998046875|4820620000|0.0|0.0|10|9.87011718
```

```

75| 30.550048828125| 44| Monday|
|2022-11-07 00:00:...| 3780.7099609375|3813.949951171875|3764.69995117187
5|3806.800048828125|4341620000| 0.0| 0.0| 11|-26.0900878906
25| 49.25| 45| Monday|
|2022-11-14 00:00:...|3977.969970703125|4008.969970703125| 3956.3999023437
5| 3957.25|4561930000| 0.0| 0.0| 11| 20.7199707031
25| 52.570068359375| 46| Monday|
|2022-11-21 00:00:...| 3956.22998046875| 3962.0|3933.34008789062
5| 3949.93994140625|3850690000| 0.0| 0.0| 11| 6.29003906
25| 28.659912109375| 47| Monday|
|2022-11-28 00:00:...|4005.360107421875| 4012.27001953125| 3955.7700195312
5| 3963.93994140625|3615430000| 0.0| 0.0| 11| 41.4201660156
25| 56.5| 48| Monday|
|2022-12-05 00:00:...| 4052.02001953125|4052.449951171875|3984.48999023437
5|3998.840087890625|4280820000| 0.0| 0.0| 12| 53.1799316406
25| 67.9599609375| 49| Monday|
|2022-12-12 00:00:...| 3939.2900390625| 3990.7099609375|3935.30004882812
5| 3990.56005859375|3904130000| 0.0| 0.0| 12| -51.270019531
25| 55.409912109375| 50| Monday|
|2022-12-19 00:00:...| 3853.7900390625|3854.860107421875| 3800.040039062
5|3817.659912109375|3969610000| 0.0| 0.0| 12| 36.1301269531
25| 54.820068359375| 51| Monday|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
--+-----+-----+-----+

```

In [106]:

```

from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear,
from pyspark.sql.functions import date_format
from pyspark.sql.functions import col
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select *, date_format(Date, "EEEE") from SPXDaily \
          where ((date_format(Date, "EEEE") in ("Monday", "Friday")) and \
                 year(Date) == 2022)').show(200)

```

```

+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+
|          Date|          Open|          High|
Low|          Close|          Volume|Dividends|Stock Splits|Month|
diff|  high_low_diff|week_of_year|week_day|date_format(CAST(Date AS TI
MESTAMP), EEEE)|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+-----+-----+-----+
+-----+
|2022-01-03 00:00:...| 4778.14013671875| 4796.64013671875| 4758.16992
1875| 4796.56005859375|3831020000| 0.0| 0.0| 1| -18.4
19921875| 38.47021484375| 1| Monday|
Monday|
|2022-01-07 00:00:...| 4697.66015625| 4707.9501953125| 4662.74023
4375| 4677.02978515625|4181510000| 0.0| 0.0| 1| 20.630
37109375| 45.2099609375| 1| Friday|
Friday|

```

In [107]:

```
spark.sql('select month(Date), date_format(Date, "EEEE"), avg(diff) from SPXDaily \
where year(Date) == 2022 \
group by month(Date), date_format(Date, "EEEE") \
order by month(Date), date_format(Date, "EEEE")').show(200)
```

```

+-----+-----+-----+
-----+
|month(CAST(Date AS DATE))|date_format(CAST(Date AS TIMESTAMP), EEEE)|
avg(diff)|
+-----+-----+-----+
-----+
|1|Friday|-
6.6124267578125|
|1|Monday|-4
2.7349853515625|
|1|Thursday|4
7.6402587890625|
|1|Tuesday|
8.090087890625|
|1|Wednesday|
50.85498046875|
|2|Friday|
4.8299560546875|
|2|Monday|4.
349934895833333|
|2|Thursday|1
2.5001220703125|
|2|Tuesday|-
20.574951171875|
|2|Wednesday|
4.2501220703125|
|3|Friday|
3.12744140625|
|3|Monday|
30.58740234375|
|3|Thursday|
-3.45615234375|
|3|Tuesday|
-11.2080078125|
|3|Wednesday|
-25.97392578125|
|4|Friday|
59.050048828125|
|4|Monday|-
7.8499755859375|
|4|Thursday|1
5.3900146484375|
|4|Tuesday|2
9.7073974609375|
|4|Wednesday|
-5.97509765625|
|5|Friday|
-27.392578125|
|5|Monday|
3.9849853515625|
|5|Thursday|
5.5999755859375|
|5|Tuesday|
0.3958984375|
|5|Wednesday|
3.76763916015625|
|6|Friday|
0.9073486328125|
|6|Monday|40.
819986979166664|
|6|Thursday|

```

8.683984375		
	6	Tuesday
1.45001220703125		
	6	Wednesday
6.843994140625		
	7	Friday -
21.353955078125		
	7	Monday 26.
109944661458332		
	7	Thursday -3
9.9849853515625		
	7	Tuesday
-12.35498046875		
	7	Wednesday -3
2.9925537109375		
	8	Friday 2
3.61517333984375		
	8	Monday
8.581982421875		
	8	Thursday
-8.357666015625		
	8	Tuesday
13.673876953125		
	8	Wednesday
-7.72412109375		
	9	Friday
23.073876953125		
	9	Monday -16.
346761067708332		
	9	Thursday
5.12197265625		
	9	Tuesday
46.39501953125		
	9	Wednesday -1
5.33758544921875		
	10	Friday -
3.50750732421875		
	10	Monday -
19.657958984375		
	10	Thursday -1
7.85748291015625		
	10	Tuesday -2
2.71002197265625		
	10	Wednesday -
3.22747802734375		
	11	Friday -
8.6275634765625		
	11	Monday 1
0.58502197265625		
	11	Thursday -36.
806722005208336		
	11	Tuesday
3.553955078125		
	11	Wednesday
4.7439453125		
	12	Friday
-2.668017578125		
	12	Monday 12.
680013020833334		
	12	Thursday
8.902001953125		

	12	Tuesday	2
7.00994873046875			
	12	Wednesday	
6.7425537109375			
+-----+			
-----+			

In [108]:

```
from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear,
from pyspark.sql.functions import date_format
from pyspark.sql.functions import col
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select *, date_format(Date, "EEEE") from SPXDaily \
          where ((date_format(Date, "EEEE") in ("Thursday")) and \
                year(Date) == 2022) and \
                month(Date) == 7').show(200)
```

```

+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+
|          Date|          Open|          High|          Lo
w|          Close|        Volume|Dividends|Stock Splits|Month|          di
ff|  high_low_diff|week_of_year|week_day|date_format(CAST(Date AS TIMESTAM
P), EEEE)|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+
|2022-07-07 00:00:...| 3858.85009765625|  3910.6298828125| 3858.8500976562
5|  3902.6201171875|4057770000|        0.0|        0.0|    7| -43.770019531
25| 51.77978515625|          27|Thursday|
Thursday|
|2022-07-14 00:00:...|3763.989990234375|3796.409912109375| 3721.5600585937
5|  3790.3798828125|4199690000|        0.0|        0.0|    7|-26.3898925781
25|74.849853515625|          28|Thursday|
Thursday|
|2022-07-21 00:00:...|3955.469970703125| 3999.2900390625|3927.63989257812
5|3998.949951171875|4132790000|        0.0|        0.0|    7| -43.479980468
75|71.650146484375|          29|Thursday|
Thursday|
|2022-07-28 00:00:...| 4026.1298828125|4078.949951171875|3992.96997070312
5|4072.429931640625|4413000000|        0.0|        0.0|    7|-46.3000488281
25| 85.97998046875|          30|Thursday|
Thursday|
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
+-----+

```

In [109]:

```
daily_df.createOrReplaceTempView("SPXDaily")
spark.sql('select min(Date), max(Date) from SPXDaily where diff < 20 or diff < -20' ).sh
```

```
+-----+-----+
|          min(Date)|          max(Date)|
+-----+-----+
|2021-01-05 00:00:...|2022-12-30 00:00:...|
+-----+-----+
```

In [110]:

```
import yfinance as yf
import csv

ticker = 'TSLA'
#^GSPC #TSLA
start_dt = '2022-8-15'
end_dt = '2022-8-24'
frequency = '1d'

#get data for TSLA
data = yf.Ticker(ticker)
#ticker_name = data.info['LongName']

#get weekly historical prices for this ticker
df = data.history(interval = frequency, start = start_dt, end = end_dt)
# write data into a csv file
df.to_csv('spx.csv')

# Let Spark know about the header and infer the Schema types!
df = spark.read.csv('spx.csv',inferSchema=True,header=True)
df.head(5)
```

Out[110]:

```
[Row(Date='2022-08-15 00:00:00-04:00', Open=301.78668212890625, High=313.133312988281, Low=301.2300109863281, Close=309.32000732421875, Volume=89359200, Dividends=0.0, Stock Splits=0.0),
 Row(Date='2022-08-16 00:00:00-04:00', Open=311.6666564941406, High=314.6666564941406, Low=302.8833312988281, Close=306.5633239746094, Volume=88136400, Dividends=0.0, Stock Splits=0.0),
 Row(Date='2022-08-17 00:00:00-04:00', Open=303.39666748046875, High=309.65667724609375, Low=300.0333251953125, Close=303.9966735839844, Volume=68766000, Dividends=0.0, Stock Splits=0.0),
 Row(Date='2022-08-18 00:00:00-04:00', Open=306.0, High=306.5, Low=301.85333251953125, Close=302.8699951171875, Volume=47500500, Dividends=0.0, Stock Splits=0.0),
 Row(Date='2022-08-19 00:00:00-04:00', Open=299.0, High=300.3599853515625, Low=292.5, Close=296.6666564941406, Volume=61395300, Dividends=0.0, Stock Splits=0.0)]
```

In [111]:

```
df.head()[4]
```

Out[111]:

309.32000732421875

In [112]:

```
from pyspark.sql.functions import dayofmonth, hour, dayofyear, month, year, weekofyear,
from pyspark.sql.functions import date_format, col
df.select(dayofyear(df['Date']), weekofyear(df['Date']), dayofweek(df['Date']), date_for
```

dayofyear(Date)	weekofyear(Date)	dayofweek(Date)	date_format(Date, EEEE)
227	33	2	Monday
228	33	3	Tuesday
229	33	4	Wednesday
230	33	5	Thursday
231	33	6	Friday
234	34	2	Monday
235	34	3	Tuesday

In [113]:

```
df.filter((dayofmonth(df['Date']) > 20) & (month(df['Date']) > 3)).show()
```

Date	Open	High	Lo
Close	Volume	Dividends	Stock Splits
2022-08-22 00:00:...	291.913330078125	292.3999938964844	286.296661376953
1 289.913330078125	55843200	0.0	0.0
2022-08-23 00:00:...	291.4533386230469	298.82666015625	287.9233398437
5 296.4533386230469	63984900	0.0	0.0

In [114]:

```
#df.filter((df['Volume'].between (40000000, 42780400))).show()
#df.where((df['Volume'].between (40000000, 42780400))).show()
df.filter((df['Volume'] >= 40000000) & (df['Volume'] < 63984900)).show()
```

```
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|          Date|          Open|          High|          Lo
w|          Close|  Volume|Dividends|Stock Splits|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
|2022-08-18 00:00:...|          306.0|          306.5|301.8533325195312
5|302.8699951171875|47500500|          0.0|          0.0|
|2022-08-19 00:00:...|          299.0|300.3599853515625|          292.
5|296.6666564941406|61395300|          0.0|          0.0|
|2022-08-22 00:00:...|291.913330078125|292.3999938964844| 286.296661376953
1| 289.913330078125|55843200|          0.0|          0.0|
+-----+-----+-----+-----+
+-----+-----+-----+-----+
```

In [115]:

```
month_df = df.withColumn('Month', month(df['Date']))
month_df.groupBy('Month').mean()[['avg(Month)', 'avg(Volume)']].orderBy('avg(Month)').sh

# newdf = df.withColumn("Year",year(df['Date']))
# newdf.groupBy("Year").mean()[['avg(Year)', 'avg(Close)']].show()
```

```
+-----+-----+
|avg(Month)|          avg(Volume)|
+-----+-----+
|          8.0|6.785507142857143E7|
+-----+-----+
```

In [116]:

```
month_df = df.withColumn('Month', month(df['Date']))
month_df.groupBy('Month').mean()[['avg(Month)', 'avg(Volume)']].sort('avg(Month)').show(
```

```
+-----+-----+
|avg(Month)|          avg(Volume)|
+-----+-----+
|          8.0|6.785507142857143E7|
+-----+-----+
```

In [117]:

```
month_df = df.withColumn('Month', month(df['Date']))
month_df.groupBy('Month').mean()[['avg(Month)', 'avg(Volume)']].orderBy('avg(Month)').sh
```

+-----+-----+	
avg(Month)	avg(Volume)
+-----+-----+	
	8.0 6.785507142857143E7
+-----+-----+	