Solution Design: Mini Project 4 How Do Markets React to Republicans and Democrats? Version 1.0

Project Instructions:

- 1. Create a csv file with a list of all presidents, their parties from 1920 onwards
- 2. Using Pandas load the .csv file into a Pandas dataframe.
- 3. Download data from an appropriate financial website such as Google Finance, Yahoo Finance, Quandl, CityFALCON, or another similar source.
- 4. Calculate yearly returns for both the downloaded indices from 1920 onwards
- 5. Segregate returns in terms of Presidency i.e. stock market returns during Democratic and Republican years
- 6. Calculate measures of central tendency (mean return, median return, variance of returns) for each of the two groups.
- 7. Represent the findings through suitable comparative graphical studies

Python Libraries Used:

The following libraries are used and can be installed using pip, e.g. "pip install pandas"

pandas 0.22.0 - see https://pandas.pydata.org/pandas-docs/stable/for more information.

Urllib3 1.22 - see https://urllib3.readthedocs.io/en/latest/ for more information.

fix_yahoo_finance - see https://pypi.org/project/fix-yahoofinance/#description

BeautifulSoup from bs4 - see

https://www.crummy.com/software/BeautifulSoup/bs4/doc/

Code Classes and Methods:

I created a Market object(class) that impplements the following methods. The __init__() method is the class constructor where all the necessary variable were declared and intialized either as an empty string or empty list and or empty pandas DataFrame.

1, get_dow_data(self)

This method retrieve the DJIA index from yahoo finance using the fix_yahoo_finance library with yf.pdr_override(). The download index time series is returned as a pandas DataFrame

2, get_snp500_data(self)

This method retrieve the S&P 500 index from yahoo finance using the fix_yahoo_finance library with yf.pdr_override(). The download index time series is returned as a pandas DataFrame

3, calc_daily_returns(self, pandas.DataFrame dataset)

This method reads the Adj Close column of the pandas DataFrame to calculate daily returns over the period of time the downloaded data spread through. The calculated daily return is then returned as a list.

4, get_years(self, pandas.DataFrame dataset)

In other to be able to map returns to presidency the year of the traded was decided and this method reads in the downloaded DataFrame after reseting the index using pandas DataFrame reset_index() method to make the Date column of the DataFrame available to be accessed as a column also assigning a sequential column with a zero base index as the DataFrame frame index. This method also returns a list.

5, do_yearly_returns(self, pandas.DataFrame dataset)

This method receives the pandas DataFrame with the daily returns and year columns now added to it to calculates yearly returns that is returned as a list

6, get_party(self, pandas.DataFrame dataset, string lookup)

The method take as input DataFrame made from csv by reading the presidents csv file with pandas and a string lookup the name of a political party to returns a list of dictionary containing the party name and the year for which the party was in power.

7, split_returns_by_presidency(self, pandas.DataFrame rets, pandas.DataFrame presidency)

The method takes two pandas DataFrame as input, the list of yearly returns loaded into pandas DataFrame and the dictionary of presidency also loaded into pandas DataFrame, the presidency DataFrame is that of a political party at a time and the method returns the dictionary of yearly returns for the presidency of the political party queried for.

Solution Steps

Step 1:

I wrote a utility python script to scrap the list of presidents, their period of presidency and their political party from wikipedia page http://en.wikipedia.org/wiki/List_of_Presidents_of_the_United_States using the following library urllib3, bs4 from where the BeautifulSoup module was imported and used and pandas which enabled me to load the list of dictionary I generated into a pandas DataFrame from where I wrote it into excel using pandas.DataFrame.to_excel() function and converted the excel file into a csv file after removing some data and columns not needed.

Step 2:

I then loaded the csv into pandas DataFrame using the pandas.read csv() method

Step 3:

I then headed on to yahoo finance to download the DOW JONES INDUSTRIAL AVERAGE index and the S&P 500 index using fix_yahoo_finance from within my code specifying startdate as 1920 and end date as today.

Step 4:

Using the **do_yearly_returns()** method I wrote then calculated the yearly returns. Before this I had to calculate daily returns using **calc_daily_returns()** method

Step 5:

I then segregated the yearly returns by presidency of the two main political party under consideration using the **split_returns_by_presidency()** method. I had to prepare the various input that this method will use to accomplish the purpose of the segregation process this is where the **get_party()** and **get_year()** methods comes in handy.

Step 6:

Now I have the datasets I need to calculate the measures of central tendency viz-a-viz mean return, median return, variance of returns for each of the two groups i.e. presidency by political party. I used the built-in method of the pandas DataFrame to find the measures. pandas.DataFrame.mean() from mean, pandad.DataFrame.median() for median and pandas.DataFrame.var() for variance.

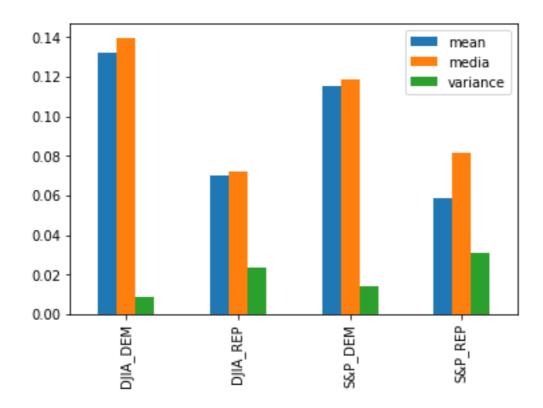
Step 7:

I decided to use barchart for easy comparism of the measures of central tendencies calculated. So I made up a pandas DataFrame from them here is the DataFrame:

SOLUTION DESIGN

```
Variance:
           0.03081363017671629
                        media variance
              mean
          0.132094
DJIA DEM
                     0.139812
                               0.008909
DJIA REP
          0.070095
                     0.071948
                               0.023531
S&P DEM
                     0.118406
          0.115611
                               0.013918
S&P REP
          0.058862
                    0.081627
                               0.030814
```

calling the pandas.DataFrame.plot.bar() function on the DataFrame displayed above gives the bar-chart below:



Using DEM - Democratics and REP - Republicans it could be observed from the bar-chart that yearly returns are higher during Democratics presidency than during Republicans presidency. The variance of the population during Republicans presidency is also higher than that of Democratics showing that during the presidency of Republicans the market is more volatile than that of the Democratics. The same showed pattern is noticed for both indexes under consideration and thus can be generalized to be the entire market reaction for both Republicans and Democratics

Tables:

```
Dow Jones During Democratics
      returns
               year
   0.129935
              1993
   0.026449
              1994
   0.291429
              1995
              1996
   0.226749
   0.222698
             1997
   0.161815
             1998
   0.237701
             1999
   -0.030042
              2000
   0.172250
             2009
   0.102646
              2010
10
   0.067858
              2011
11
   0.062219
              2012
12
   0.216784
             2013
             2014
13
   0.086577
14 -0.011247
              2015
   0.149689
              2016
      0.13209429159974323
Mean:
Median:
        0.13981197679800306
Variance: 0.008908996310214854
```

```
S&P 500 During Democratics
      returns
                year
               1950
    0.215093
    0.140543
               1951
2
    0.113206
               1952
3
4
5
6
7
    0.222499
               1961
   -0.103703
               1962
    0.183266
               1963
    0.117900
               1964
               1965
    0.095296
   -0.130632
               1966
9
    0.186028
               1967
10
    0.081481
               1968
11 -0.113774
               1977
12
    0.031907
               1978
13
    0.115544
               1979
14
    0.263174
               1980
15
    0.072641
               1993
16
   -0.008515
               1994
17
    0.297047
               1995
18
    0.183772
               1996
19
    0.291764
               1997
20
    0.252341
               1998
21
    0.195589
               1999
22 -0.072741
               2000
23
    0.216227
               2009
24
    0.120520
               2010
25
    0.015751
               2011
26
    0.118406
               2012
27
    0.240127
               2013
28
    0.123207
               2014
29
    0.005018
               2015
30
    0.114950
               2016
       0.11561061921147454
Mean:
Median:
          0.11840559700141137
            0.013917932637644894
Variance:
```

```
Jones During Republicans
      returns
               year
    0.183954
              1985
1
    0.221284
              1986
    0.069757
              1987
3
    0.089159
              1988
4
5
6
    0.260001
              1989
   -0.051724
              1990
    0.204495
              1991
7
    0.045126
              1992
8
   -0.038074
              2001
   -0.156254
             2002
    0.207618 2003
10
    0.041069
              2004
11
12
    0.004165
              2005
    0.143652
13
              2006
14
    0.071948
              2007
15 -0.324608
             2008
    0.220037
16
              2017
Mean:
       0.07009452191250096
Median: 0.07194819496171954
Variance: 0.023531227203181248
```

```
500 During Republicans
       returns
                 vear
   -0.062838
                1953
Θ
1
    0.370754
                1954
2
    0.224837
                1955
3
    0.040523
                1956
456
   -0.135963
                1957
    0.318236
                1958
    0.081627
                1959
7
   -0.025130
                1960
8
   -0.116268
                1969
9
    0.002261
                1970
10
    0.118513
                1971
11
    0.152552
                1972
   -0.187099
                1973
12
   -0.330016
13
                1974
    0.261880
14
                1975
15
    0.173549
                1976
   -0.097568
                1981
16
    0.152750
17
                1982
18
    0.184588
                1983
19
    0.027396
                1984
    0.250175
20
                1985
    0.155343
21
                1986
22
    0.057410
                1987
23
    0.095762
                1988
24
    0.258345
                1989
25
   -0.072896
                1990
    0.255268
26
                1991
27
    0.047975
                1992
   -0.088908
28
                2001
   -0.238119
29
                2002
30
    0.215521
                2003
    0.095356
31
                2004
32
    0.042956
                2005
    0.116267
33
                2006
    0.048602
34
                2007
   -0.387021
35
                2008
36
    0.171271
                2017
        0.058861903820286984
Mean:
Median:
          0.0816269083472667
```