Introduction

In this project, you will act as a data visualization developer at Yahoo Finance! You will be helping the "Netflix Stock Profile" team visualize the Netflix stock data. In finance, a *stock profile* is a series of studies, visualizations, and analyses that dive into different aspects a publicly traded company's data.

For the purposes of the project, you will only visualize data for the year of 2017. Specifically, you will be in charge of creating the following visualizations:

- · The distribution of the stock prices for the past year
- · Netflix's earnings and revenue in the last four quarters
- The actual vs. estimated earnings per share for the four quarters in 2017
- A comparison of the Netflix Stock price vs the Dow Jones Industrial Average price in 2017

Note: We are using the Dow Jones Industrial Average to compare the Netflix stock to the larter stock market. Learn more about why the Dow Jones Industrial Average is a general reflection of the larger stock market https://www.investopedia.com/terms/d/djia.asp).

During this project, you will analyze, prepare, and plot data. Your visualizations will help the financial analysts asses the risk of the Netflix stock.

After you complete your visualizations, you'll be creating a presentation to share the images with the rest of the Netflix Stock Profile team. Your slides should include:

- · A title slide
- · A list of your visualizations and your role in their creation for the "Stock Profile" team
- · A visualization of the distribution of the stock prices for Netflix in 2017
- · A visualization and a summary of Netflix stock and revenue for the past four quarters and a summary
- · A visualization and a brief summary of their earned versus actual earnings per share
- A visualization of Netflix stock against the Dow Jones stock (to get a sense of the market) in 2017

Financial Data Source: Yahoo Finance (https://finance.yahoo.com/quote/DATA/)

Step 1

Let's get our notebook ready for visualizing! Import the modules that you'll be using in this project:

- from matplotlib import pyplot as plt
- import pandas as pd
- import seaborn as sns

```
In [1]: from matplotlib import pyplot as plt
import pandas as pd
import seaborn as sns
```

Step 2

Let's load the datasets and inspect them.

Load **NFLX.csv** into a DataFrame called netflix_stocks . Then, quickly inspect the DataFrame using print() .

Hint: Use the pd.read csv() function).

Note: In the Yahoo Data, Adj Close represents the adjusted close price adjusted for both dividends and splits. This means this is the true closing stock price for a given business day.

```
In [2]: netflix_stocks = pd.read_csv('NFLX.csv')
        print(netflix stocks.head())
                Date
                           Open
                                       High
                                                   LOW
                                                             Close
                                                                    Adj Close \
         2017-01-01 124.959999 143.460007 124.309998 140.710007 140.710007
       1 2017-02-01 141.199997 145.949997 139.050003 142.130005 142.130005
       2 2017-03-01 142.839996 148.289993 138.259995 147.809998 147.809998
       3 2017-04-01 146.699997 153.520004 138.660004 152.199997 152.199997
        4 2017-05-01 151.910004 164.750000 151.610001 163.070007 163.070007
             Volume
       0
         181772200
           91432000
       1
       2
          110692700
       3
          149769200
       4 116795800
```

Load **DJI.csv** into a DataFrame called dowjones stocks. Then, quickly inspect the DataFrame using print().

Note: You can learn more about why the Dow Jones Industrial Average is a industry reflection of the larger stock market here (here (<a href="https://www.investopedia.com/terms/d/djia.as

```
In [3]: dowjones_stocks = pd.read_csv('DJI.csv')
        print(dowjones_stocks.head())
                 Date
                               Open
                                             High
                                                            Low
                                                                        Close \
        0 \quad 2017 - 01 - 01 \quad 19872.859375 \quad 20125.580078 \quad 19677.939453 \quad 19864.089844
        1 2017-02-01 19923.810547 20851.330078 19831.089844 20812.240234
        2 2017-03-01 20957.289063 21169.109375 20412.800781 20663.220703
        3 2017-04-01 20665.169922 21070.900391 20379.550781 20940.509766
        4 2017-05-01 20962.730469 21112.320313 20553.449219 21008.650391
              Adj Close
                             Volume
        0 19864.089844 6482450000
        1 20812.240234 6185580000
        2 20663.220703 6941970000
        3
           20940.509766 5392630000
           21008.650391 6613570000
```

Load **NFLX_daily_by_quarter.csv** into a DataFrame called netflix_stocks_quarterly. Then, quickly inspect the DataFrame using print().

```
In [4]: netflix stocks quarterly = pd.read csv('NFLX daily by quarter.csv')
       print(netflix_stocks_quarterly.head())
                Date
                            Open
                                       High
                                                    Low
                                                             Close
                                                                     Adj Close
         2017-01-03 124.959999 128.190002 124.309998 127.489998
                                                                    127.489998
        1 2017-01-04 127.489998 130.169998 126.550003 129.410004
                                                                    129.410004
        2 2017-01-05 129.220001 132.750000 128.899994 131.809998
                                                                    131.809998
       3 2017-01-06 132.080002 133.880005 129.809998 131.070007 131.070007
        4 2017-01-09 131.479996 131.990005 129.889999 130.949997 130.949997
            Volume Quarter
        0
           9437900
                        Q1
       1
           7843600
                        Q1
       2 10185500
                        Q1
       3 10657900
                        Q1
           5766900
                        Q1
```

Let's learn more about our data. The datasets are large and it may be easier to view the entire dataset locally on your computer. Open the CSV files directly from the folder you downloaded for this project.

- . NFLX is the stock ticker symbol for Netflix and ^DJI is the stock ticker symbol for the Dow Jones industrial Average, which is why the CSV files are named accordingly
- In the Yahoo Data, Adj Close is documented as adjusted close price adjusted for both dividends and splits.
- · You can learn more about why the Dow Jones Industrial Average is a industry reflection of the larger stock market here (https://www.investopedia.com/terms/d/djia.asp).

Answer the following questions by inspecting the data in the NFLX.csv,DJI.csv, and NFLX_daily_by_quarter.csv in your computer.

What year is represented in the data? Look out for the latest and earliest date.

```
In [5]: 2017
Out[5]: 2017
```

- · Is the data represented by days, weeks, or months?
- · In which ways are the files different?
- What's different about the columns for netflix_stocks versus netflix_stocks_quarterly?

```
In [6]: "Months"
        "NFLX tracks Netflix \and DJI tracks the Dow Jones \while Quarterly tracks NFLX each day"
        "Quarterly shows what Quarter of trading we are in for each day"
```

Out[6]: 'Quarterly shows what Quarter of trading we are in for each day'

Step 4

Great! Now that we have spent sometime looking at the data, let's look at the column names of the DataFrame netflix stocks using .head().

```
In [7]:
       print(netflix stocks.head())
                Date
                           Open
                                       High
                                                   Low
                                                             Close
                                                                    Adj Close
         2017-01-01 124.959999 143.460007 124.309998 140.710007 140.710007
         2017-02-01 141.199997 145.949997 139.050003 142.130005 142.130005
       1
       2 2017-03-01 142.839996 148.289993 138.259995 147.809998 147.809998
       3
          2017-04-01 146.699997
                                153.520004
                                            138.660004 152.199997
                                                                   152.199997
          2017-05-01 151.910004 164.750000 151.610001 163.070007 163.070007
             Volume
         181772200
       0
       1
          91432000
       2 110692700
       3 149769200
       4 116795800
```

What do you notice? The first two column names are one word each, and the only one that is not is Adj Close!

The term Adj Close is a confusing term if you don't read the Yahoo Documentation. In Yahoo, Adj Close is documented as adjusted close price adjusted for both dividends and splits.

This means this is the column with the true closing price, so these data are very important.

Use Pandas to change the name of of the column to Adj Close to Price so that it is easier to work with the data. Remember to use inplace=True.

Do this for the Dow Jones and Netflix Quarterly pandas dataframes as well. Hint: Use .rename() (https://pandas.pydata.org/pandasdocs/stable/generated/pandas.DataFrame.rename.html)).

```
In [12]: netflix_stocks.rename(columns={'Adj Close' : 'Price'}, inplace=True)
    dowjones_stocks.rename(columns={'Adj Close' : 'Price'}, inplace=True)
    netflix_stocks_quarterly.rename(columns={'Adj Close' : 'Price'}, inplace=True)
```

Run netflix_stocks.head() again to check your column name has changed.

```
In [13]: print(netflix_stocks.head())
                 Date
                             Open
                                        High
                                                    Low
                                                              Close
                                                                         Price
        0 2017-01-01 124.959999
                                  143.460007
                                             124.309998
                                                         140.710007
                                                                     140.710007
        1 2017-02-01 141.199997
                                  145.949997
                                              139.050003
                                                         142.130005
                                                                     142.130005
        2 2017-03-01 142.839996
                                  148.289993 138.259995 147.809998
                                                                     147.809998
        3 2017-04-01 146.699997
                                  153.520004 138.660004 152.199997
                                                                     152.199997
        4 2017-05-01 151.910004 164.750000 151.610001 163.070007 163.070007
              Volume
          181772200
           91432000
        2 110692700
        3 149769200
        4 116795800
```

Call .head() on the DataFrame dowjones_stocks and netflix_stocks_quarterly.

```
In [16]: dowjones_stocks.head()
   netflix_stocks_quarterly.head()
```

Out[16]:

	Date	Open	High	Low	Close	Price	Volume	Quarter
0	2017-01-03	124.959999	128.190002	124.309998	127.489998	127.489998	9437900	Q1
1	2017-01-04	127.489998	130.169998	126.550003	129.410004	129.410004	7843600	Q1
2	2017-01-05	129.220001	132.750000	128.899994	131.809998	131.809998	10185500	Q1
3	2017-01-06	132.080002	133.880005	129.809998	131.070007	131.070007	10657900	Q1
4	2017-01-09	131.479996	131.990005	129.889999	130.949997	130.949997	5766900	Q1

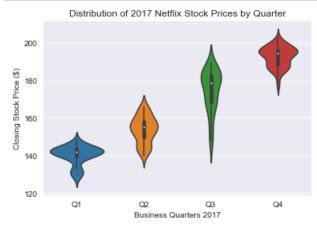
Step 5

In this step, we will be visualizing the Netflix quarterly data!

We want to get an understanding of the distribution of the Netflix quarterly stock prices for 2017. Specifically, we want to see in which quarter stock prices flucutated the most. We can accomplish this using a violin plot with four violins, one for each business quarter!

- 1. Start by creating a variable ax and setting it equal to sns.violinplot(). This will instantiate a figure and give us access to the axes through the variable name ax.
- 2. Use sns.violinplot() and pass in the following arguments:
- \bullet The Quarter column as the \boldsymbol{x} values
- The Price column as your y values
- The netflix_stocks_quarterly dataframe as your data
- 1. Improve the readability of the chart by adding a title of the plot. Add "Distribution of 2017 Netflix Stock Prices by Quarter" by using ax.set_title()
- 2. Change your ylabel to "Closing Stock Price"
- 3. Change your xlabel to "Business Quarters in 2017"
- 4. Be sure to show your plot!

```
In [49]: sns.set_style('darkgrid')
    ax = sns.violinplot()
    sns.violinplot(data=netflix_stocks_quarterly, x='Quarter', y='Price')
    ax.set_title('Distribution of 2017 Netflix Stock Prices by Quarter')
    plt.ylabel('Closing Stock Price ($)')
    plt.xlabel('Business Quarters 2017')
    plt.show()
```



Graph Literacy

- · What are your first impressions looking at the visualized data?
- In what range(s) did most of the prices fall throughout the year?
- What were the highest and lowest prices?

Step 6

Next, we will chart the performance of the earnings per share (EPS) by graphing the estimate Yahoo projected for the Quarter compared to the actual earnings for that quarters. We will accomplish this using a scatter chart.

- 1. Plot the actual EPS by using x_positions and earnings_actual with the plt.scatter() function. Assign red as the color.
- 2. Plot the estimated EPS by using $x_positions$ and $earnings_estimate$ with the plt.scatter() function. Assign blue as the color
- 3. Often, estimates and actual EPS are the same. To account for this, be sure to set your transparency alpha=0.5 to allow for visibility pf overlapping datapoint.
- 4. Add a legend by using plt.legend() and passing in a list with two strings ["Actual", "Estimate"]
- 5. Change the x_ticks label to reflect each quarter by using plt.xticks(x_positions, chart_labels)
- 6. Assing " "Earnings Per Share in Cents" as the title of your plot.

```
In [51]: sns.set_style('darkgrid')
    x_positions = [1, 2, 3, 4]
    chart_labels = ["1Q2017","2Q2017","3Q2017","4Q2017"]
    earnings_actual = [.4, .15,.29,.41]
    earnings_estimate = [.37,.15,.32,.41]

plt.scatter(x_positions, earnings_actual, color='red', alpha=0.5)
    plt.scatter(x_positions, earnings_estimate, color='blue', alpha=0.5)
    plt.legend(["Actual", "Estimate"])
    plt.ylabel("US$")
    plt.xticks(x_positions, chart_labels)
    plt.title("Earnings Per Share in Cents")
    plt.show()
```



Graph Literacy

• What do the purple dots tell us about the actual and estimate earnings per share in this graph? Hint: In color theory red and blue mix to make purple.

Step 7

Next, we will visualize the earnings and revenue reported by Netflix by mapping two bars side-by-side. We have visualized a similar chart in the second Matplotlib lesson <u>Exercise 4 (https://www.codecademy.com/courses/learn-matplotlib/lessons/matplotlib-ii/exercises/side-by-side-bars).</u></u>

As you may recall, plotting side-by-side bars in Matplotlib requires computing the width of each bar before hand. We have pasted the starter code for that exercise below.

- 1. Fill in the $\,n\,,\,\,t\,,\,\,d\,,\,\,w\,$ values for the revenue bars
- 2. Plot the revenue bars by calling plt.bar() with the newly computed x values and the revenue by quarter data
- 3. Fill in the $\,n\,$, $\,t\,$, $\,d\,$, $\,w\,$ values for the earnings bars
- 4. Plot the revenue bars by calling plt.bar() with the newly computed x values and the earnings by quarter data
- 5. Create a legend for your bar chart with the labels provided
- 6. Add a descriptive title for your chart with plt.title()
- 7. Add labels to each quarter by assigning the position of the ticks through the code provided. Hint: plt.xticks(middle_x, quarter_labels)
- 8. Be sure to show your plot!

```
In [52]: sns.set_style('darkgrid')
         # The metrics below are in billions of dollars
         revenue_by_quarter = [2.79, 2.98,3.29,3.7]
         earnings_by_quarter = [.0656,.12959,.18552,.29012]
         quarter_labels = ["2Q2017","3Q2017","4Q2017", "1Q2018"]
         # Revenue
         n = 1 # This is our first dataset (out of 2)
         t = 2 # Number of dataset
         d = 4 # Number of sets of bars
         w = 0.8 \# Width of each bar
         bars1 x = [t*element + w*n for element]
                      in range(d)]
         plt.bar(bars1_x, revenue by quarter)
         # Earnings
         n = 2 # This is our second dataset (out of 2)
         t = 2 # Number of dataset
         d = 4 # Number of sets of bars
         w = 0.8 \# Width of each bar
         bars2_x = [t*element + w*n for element
                      in range(d)]
         plt.bar(bars2_x, earnings_by_quarter)
         middle_x = [ (a + b) / 2.0  for a, b in zip(bars1_x, bars2_x)]
         labels = ["Revenue", "Earnings"]
         plt.legend(labels)
         plt.ylabel('Billions USD')
         plt.title('Netflix Earnings & Revenue by Quarter')
         plt.xticks(middle_x, quarter_labels)
         plt.show()
```



Graph Literacy

What are your first impressions looking at the visualized data?

- · Does Revenue follow a trend?
- · Do Earnings follow a trend?
- · Roughly, what percentage of the revenue constitutes earnings?

```
In [24]: 'Yes, both are increasing exponentially. Earnings is around 4% of Revenue.'
Out[24]: 'Yes, both are increasing exponentially. Earnings is around 4% of Revenue.'
```

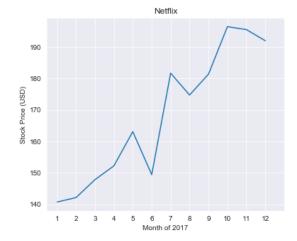
Step 8

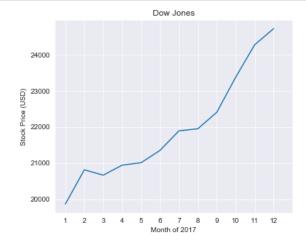
In this last step, we will compare Netflix stock to the Dow Jones Industrial Average in 2017. We will accomplish this by plotting two line charts side by side in one figure.

Since Price which is the most relevant data is in the Y axis, let's map our subplots to align vertically side by side.

- We have set up the code for you on line 1 in the cell below. Complete the figure by passing the following arguments to plt.subplots() for the first plot, and tweaking the third argument for the second plot
 - 1 -- the number of rows for the subplots
 - 2 -- the number of columns for the subplots
 - 1 -- the subplot you are modifying
- Chart the Netflix Stock Prices in the left-hand subplot. Using your data frame, access the Date and Price charts as the x and y axes respectively. Hint: (netflix_stocks['Date'], netflix_stocks['Price'])
- Assign "Netflix" as a title to this subplot. Hint: ax1.set_title()
- For each subplot, set xlabel to "Date" and set ylabel to "Stock Price"
- Chart the Dow Jones Stock Prices in the left-hand subplot. Using your data frame, access the Date and Price charts as the x and y axes respectively. Hint: (dowjones_stocks['Date'], dowjones_stocks['Price'])
- Assign "Dow Jones" as a title to this subplot. Hint: plt.set_title()
- There is some crowding in the Y axis labels, add some space by calling plt.subplots adjust(wspace=.5)
- Be sure to .show() your plots.

```
In [53]: sns.set_style('darkgrid')
         plt.figure(figsize=[15, 5])
         # Left plot Netflix
         ax1 = plt.subplot(1, 2, 1)
         plt.plot(netflix_stocks['Date'], netflix_stocks['Price'])
         ax1.set_title('Netflix')
         ax1.set xlabel("Month of 2017")
         ax1.set_xticks(range(13))
         ax1.set_xticklabels(range(1, 13))
         ax1.set_ylabel("Stock Price (USD)")
         plt.subplots_adjust(wspace=.5)
         plt.subplots_adjust(hspace=10)
         # Right plot Dow Jones
         ax2 = plt.subplot(1, 2, 2)
         plt.plot(dowjones_stocks['Date'], dowjones_stocks['Price'])
         ax2.set_title('Dow Jones')
         ax2.set_xlabel("Month of 2017")
         ax2.set_xticks(range(13))
         ax2.set_xticklabels(range(1, 13))
         ax2.set_ylabel("Stock Price (USD)")
         plt.subplots adjust(wspace=.5)
         plt.subplots adjust(hspace=10)
         plt.show()
```





- How did Netflix perform relative to Dow Jones Industrial Average in 2017?
- · Which was more volatile?
- · How do the prices of the stocks compare?

Step 9

It's time to make your presentation! Save each of your visualizations as a png file with plt.savefig("filename.png").

As you prepare your slides, think about the answers to the graph literacy questions. Embed your observations in the narrative of your slideshow!

Remember that your slideshow must include:

- A title slide
- A list of your visualizations and your role in their creation for the "Stock Profile" team
- A visualization of the distribution of the stock prices for Netflix in 2017
- · A visualization and a summary of Netflix stock and revenue for the past four quarters and a summary
- · A visualization and a brief summary of their earned versus actual earnings per share
- A visualization of Netflix stock against the Dow Jones stock (to get a sense of the market) in 2017

In []:	