

```
In [15]: from matplotlib import pyplot as plt
import pandas as pd
import seaborn as sns
import os
os.chdir('E:/Netflix/Netflix Stocks Capstone')
```

```
In [16]: netflix_stocks=pd.read_csv('NFLX.csv')
print(netflix_stocks)
```

	Date	Open	High	Low	Close	Adj Close	\
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	
5	2017-06-01	163.520004	166.869995	147.300003	149.410004	149.410004	
6	2017-07-01	149.800003	191.500000	144.250000	181.660004	181.660004	
7	2017-08-01	182.490005	184.619995	164.229996	174.710007	174.710007	
8	2017-09-01	175.550003	189.949997	172.440002	181.350006	181.350006	
9	2017-10-01	182.110001	204.380005	176.580002	196.429993	196.429993	
10	2017-11-01	197.240005	202.479996	184.320007	195.509995	195.509995	
11	2017-12-01	186.990005	194.490005	178.380005	191.960007	191.960007	

	Volume
0	181772200
1	91432000
2	110692700
3	149769200
4	116795800
5	135675800
6	185144700
7	136523100
8	111427900
9	208657800
10	161719700
11	115103700

```
In [17]: dowjones_stocks=pd.read_csv('DJI.csv')
print(dowjones_stocks)
```

	Date	Open	High	Low	Close \
0	2017-01-01	19872.859375	20125.580078	19677.939453	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
5	2017-06-01	21030.550781	21535.029297	20994.220703	21349.630859
6	2017-07-01	21392.300781	21929.800781	21279.300781	21891.119141
7	2017-08-01	21961.419922	22179.109375	21600.339844	21948.099609
8	2017-09-01	21981.769531	22419.509766	21709.630859	22405.089844
9	2017-10-01	22423.470703	23485.250000	22416.000000	23377.240234
10	2017-11-01	23442.900391	24327.820313	23242.750000	24272.349609
11	2017-12-01	24305.400391	24876.070313	23921.900391	24719.220703

	Adj Close	Volume
0	19864.089844	6482450000
1	20812.240234	6185580000
2	20663.220703	6941970000
3	20940.509766	5392630000
4	21008.650391	6613570000
5	21349.630859	7214590000
6	21891.119141	5569720000
7	21948.099609	6150060000
8	22405.089844	6342130000
9	23377.240234	7302910000
10	24272.349609	7335640000
11	24719.220703	6589890000

```
In [18]: netflix_stocks_quarterly=pd.read_csv("NFLX_daily_by_quarter.csv")
print(netflix_stocks_quarterly)
```

	Date	Open	High	Low	Close	Adj Close	\
0	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	
..	
246	2017-12-22	188.330002	190.949997	186.800003	189.940002	189.940002	
247	2017-12-26	189.779999	189.940002	186.399994	187.759995	187.759995	
248	2017-12-27	187.800003	188.100006	185.220001	186.240005	186.240005	
249	2017-12-28	187.179993	194.490005	186.850006	192.710007	192.710007	
250	2017-12-29	192.509995	193.949997	191.220001	191.960007	191.960007	

	Volume	Quarter
0	9437900	Q1
1	7843600	Q1
2	10185500	Q1
3	10657900	Q1
4	5766900	Q1
..
246	3878900	Q4
247	3045700	Q4
248	4002100	Q4
249	10107400	Q4
250	5187600	Q4

[251 rows x 8 columns]

```
In [19]: A=max(netflix_stocks.Date)
print(A)
B=max(netflix_stocks_quarterly.Date)
print(B)
C=min(netflix_stocks.Date)
print(C)
```

```
2017-12-01
2017-12-29
2017-01-01
```

The data is represented on monthly basis.\ The netflix and dow jones in based on months whereas netflix quarterly is daily data on quatermonth based.\ netflix_stocks_quarterly contains quarter column

```
In [20]: print(netflix_stocks.head())
```

	Date	Open	High	Low	Close	Adj Close	\
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
0	181772200
1	91432000
2	110692700
3	149769200
4	116795800

```
In [21]: netflix_stocks=netflix_stocks.rename(columns={'Adj Close':'Price'})
dowjones_stocks=dowjones_stocks.rename(columns={'Adj Close':'Price'})
netflix_stocks_quarterly=netflix_stocks_quarterly.rename(columns={'Adj Close':'Price'})
```

```
In [22]: 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
0	181772200
1	91432000
2	110692700
3	149769200
4	116795800

```
In [23]: print(dowjones_stocks.head())
print(netflix_stocks_quarterly.head())
```

	Date	Open	High	Low	Close \
0	2017-01-01	19872.859375	20125.580078	19677.939453	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

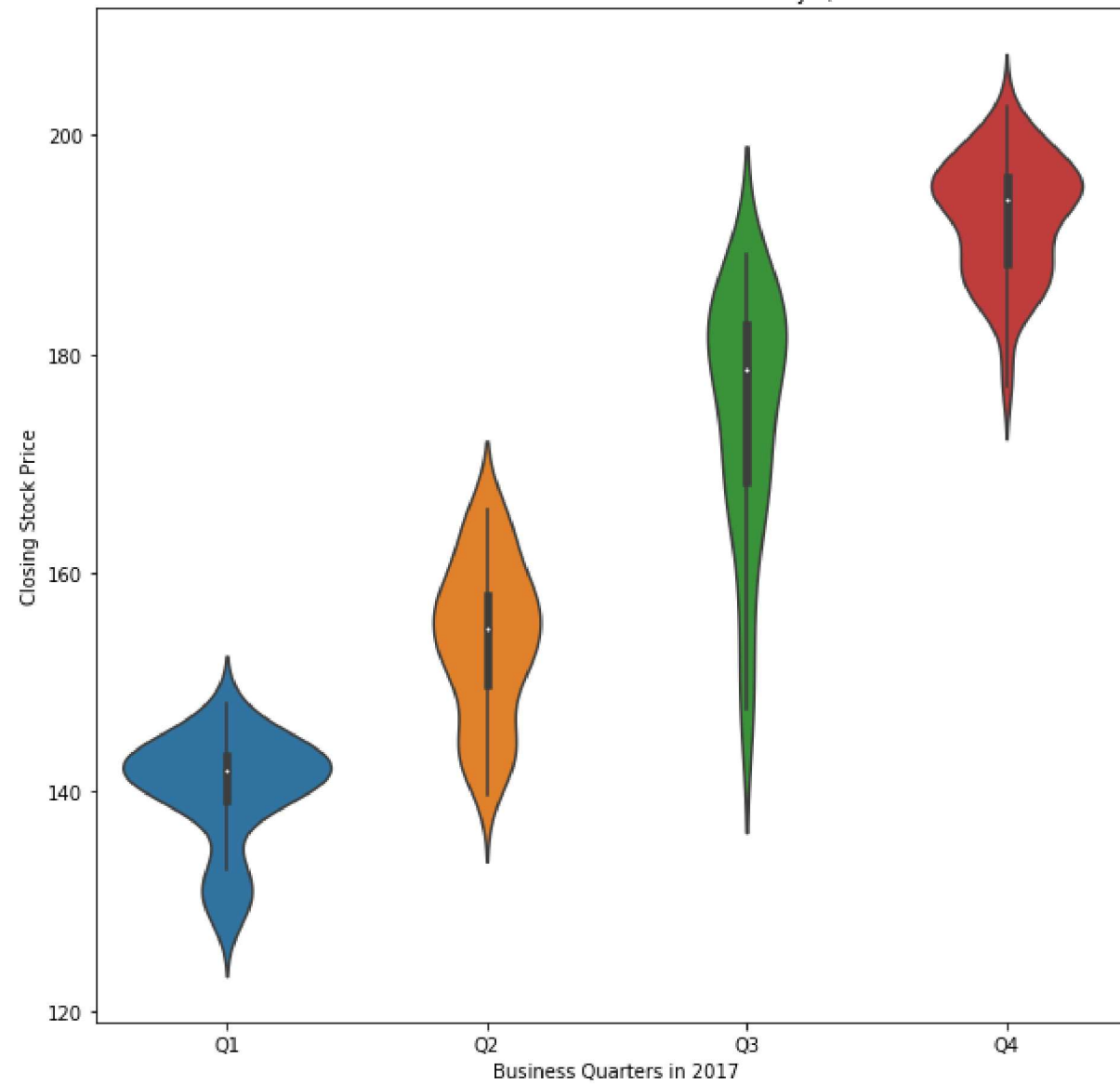
	Price	Volume
0	19864.089844	6482450000
1	20812.240234	6185580000
2	20663.220703	6941970000
3	20940.509766	5392630000
4	21008.650391	6613570000

	Date	Open	High	Low	Close	Price \
0	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
4	5766900	Q1

```
In [24]: plt.figure(figsize = (10,10))
ax=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 in 2017")
plt.show
plt.savefig("Distribution of 2017 Netflix Stock Prices by Quarter.png")
```

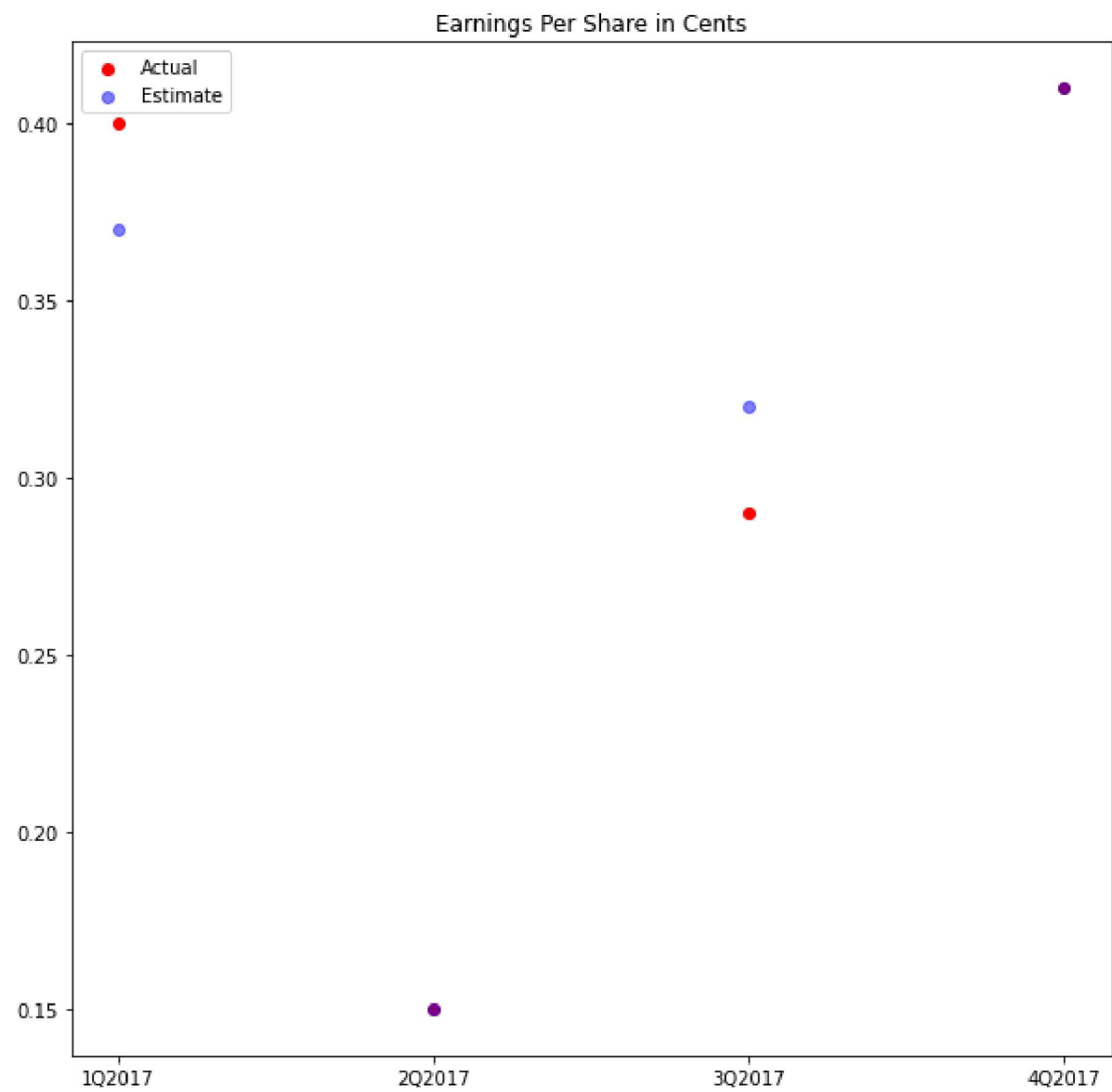
Distribution of 2017 Netflix Stock Prices by Quarter



Graph literacy

Netflix Stock Prices should a bull trend throughout the course of 2017 (Q1 - Q4).\ Prices fell in the 140 - 180 range throughout the year.\ The lowest price was in (Q1) at 130 and highest was in (Q4) at 210.

```
In [29]: 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.figure(figsize = (10,10))
plt.scatter(x_positions,earnings_actual,color="red")
plt.scatter(x_positions,earnings_estimate,color="blue",alpha=0.5)
plt.legend(["Actual", "Estimate"])
plt.xticks(x_positions, chart_labels)
plt.title("Earnings Per Share in Cents")
plt.savefig("Earnings Per Share in Cents.png")
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.

Purple dot shows where the estimated value and actual value overlapped.

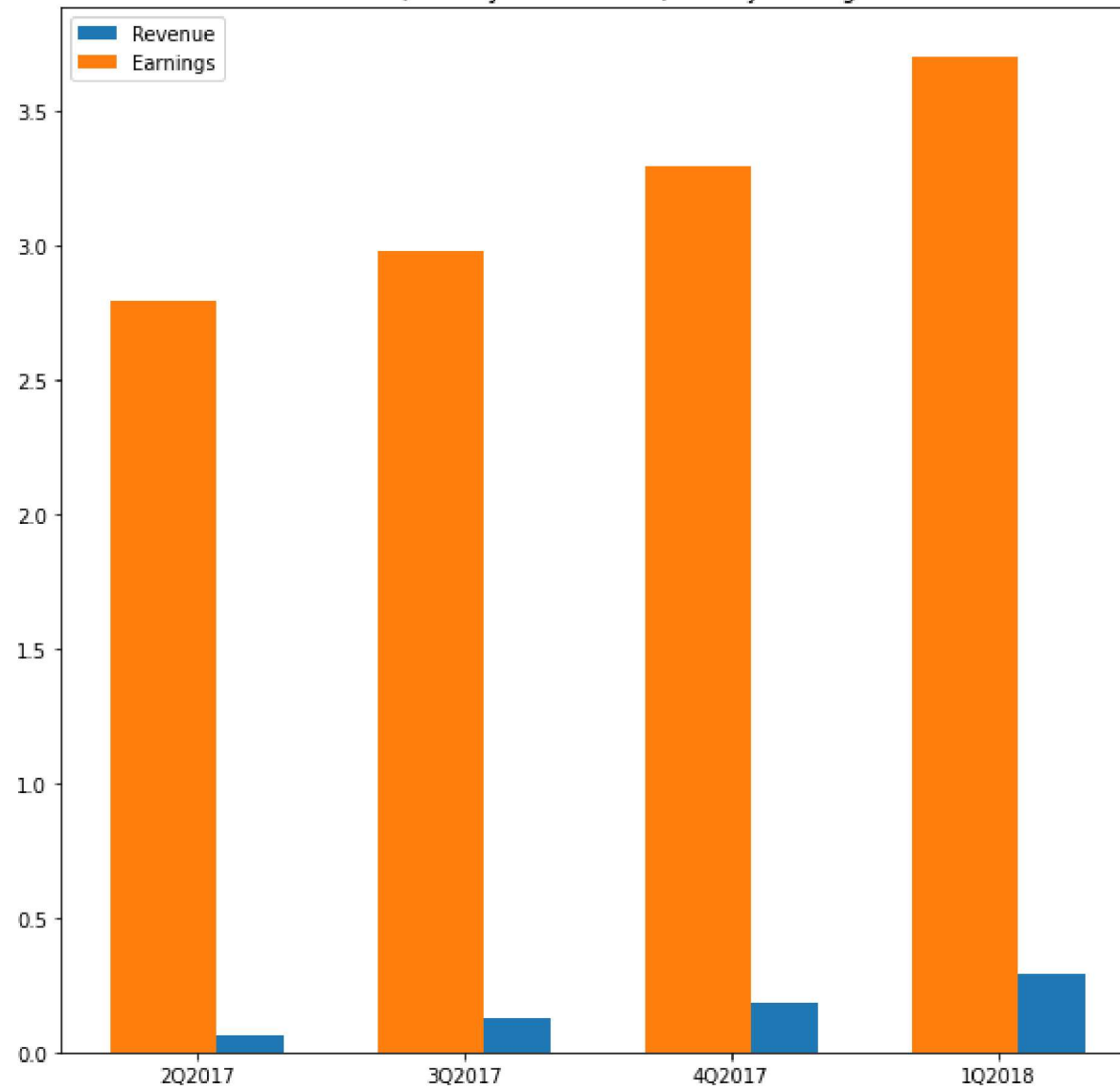
```
In [26]: # The metrics below are in billions of dollars
revenue_by_quarter = [2.79, 2.98, 3.29, 3.7]
earnings_by_quarter = [.0656, .12959, .18552, .29012]
quarter_labels = ["2Q2017", "3Q2017", "4Q2017", "1Q2018"]

# Revenue
n = 1
t = 2
d = 4
w = 0.5
bars1_x = [t*element + w*n for element
            in range(d)]

# Earnings
n = 2
t = 2
d = 4 # Number of sets of bars
w = 0.5 # Width of each bar
bars2_x = [t*element + w*n for element
            in range(d)]

middle_x = [ (a + b) / 2.0 for a, b in zip(bars1_x, bars2_x)]
labels = ["Revenue", "Earnings"]
plt.figure(figsize = (10,10))
plt.bar(bars2_x, earnings_by_quarter)
plt.bar(bars1_x, revenue_by_quarter)
plt.legend(labels)
plt.title('2017 Quarterly Revenue vs Quarterly Earnings')
plt.xticks(middle_x, quarter_labels)
plt.savefig("2017 Quarterly Revenue vs Quarterly Earnings.png")
```

2017 Quarterly Revenue vs Quarterly Earnings



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?

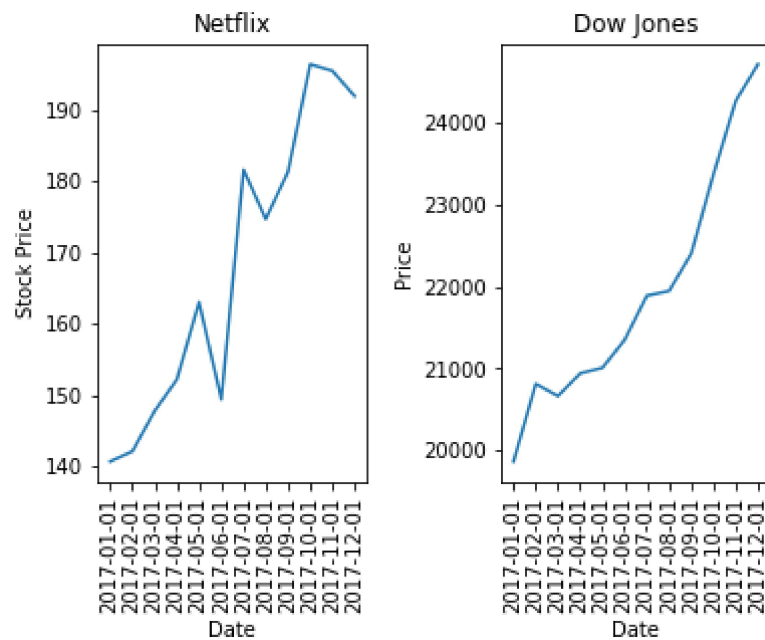
Graph Literacy answers

The revenue follows an upward trend throughout.\ The earnings also tend to increase overtime.\ Roughly 14% of revenue constitutes earnings

```
In [33]: # Left plot Netflix
ax1 = plt.subplot(1,2,1)
plt.plot(netflix_stocks['Date'], netflix_stocks['Price'])
plt.title("Netflix")
ax1.set_xlabel("Date")
ax1.set_ylabel("Stock Price")
plt.subplots_adjust(wspace = 0.5)
plt.xticks(rotation = 'vertical')

# Right plot Dow Jones
ax2 = plt.subplot(1,2,2)
plt.plot(dowjones_stocks['Date'], dowjones_stocks['Price'])
plt.title('Dow Jones')
ax2.set_xlabel('Date')
ax2.set_ylabel('Price')
plt.subplots_adjust(wspace = 0.5)
plt.xticks(rotation = 'vertical')

plt.savefig("Netflix vs Dow Jones stocks.png", bbox_inches='tight')
```



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

Netflix stock performed well relative to Dow Jones industrial average in 2017.\ The Netflix share was more volatile as seen from graph of different peak and valleys.\ The Dow Jones stock price is more.

In []: