

Analysis of Tesla's & NIKE vs. S&P 500

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```
In [103]: #Import required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import pandas_datareader as pdr
from scipy import stats
import seaborn as sns
```

```
In [104]: #Import TSLA from yahoo finance from 2011-2020

tesla= pdr.get_data_yahoo("TSLA",
                           start = "2011-01-01",
                           end = "2020-11-01")

nike= pdr.get_data_yahoo("NKE",
                          start = "2011-01-01",
                          end= "2020-11-01")
```

```
In [105]: #Import TSLA from yahoo finance from 2011-2020
SP500= pdr.get_data_yahoo("SPY",
                           start = "2011-01-01",
                           end= "2020-11-01")
```

```
In [106]: #Making sure that the dataset looks okay
nike.head()
```

Out[106]:

	High	Low	Open	Close	Volume	Adj Close
Date						
2011-01-03	21.645000	21.315001	21.457500	21.522499	8566400.0	16.666460
2011-01-04	21.437500	20.937500	21.400000	20.992500	13797600.0	16.256046
2011-01-05	21.207500	20.877501	20.912500	21.129999	11598800.0	16.362526
2011-01-06	21.125000	20.889999	21.112499	20.940001	8057200.0	16.215391
2011-01-07	20.987499	20.817499	20.924999	20.882500	8174400.0	16.170864

```
In [107]: SP500.head()
```

```
Out[107]:
```

	High	Low	Open	Close	Volume	Adj Close
Date						
2011-01-03	127.599998	125.699997	126.709999	127.050003	138725200.0	104.119293
2011-01-04	127.370003	126.190002	127.330002	126.980003	137409700.0	104.061905
2011-01-05	127.720001	126.459999	126.580002	127.639999	133975300.0	104.602806
2011-01-06	127.830002	127.010002	127.690002	127.389999	122519000.0	104.397934
2011-01-07	127.769997	126.150002	127.559998	127.139999	156034600.0	104.193031

2. Graph the Adjusted Close over the years for Tesla

In [108]: *#Define the x and y labels for the graph*

```
plt.figure(figsize= (10,5))

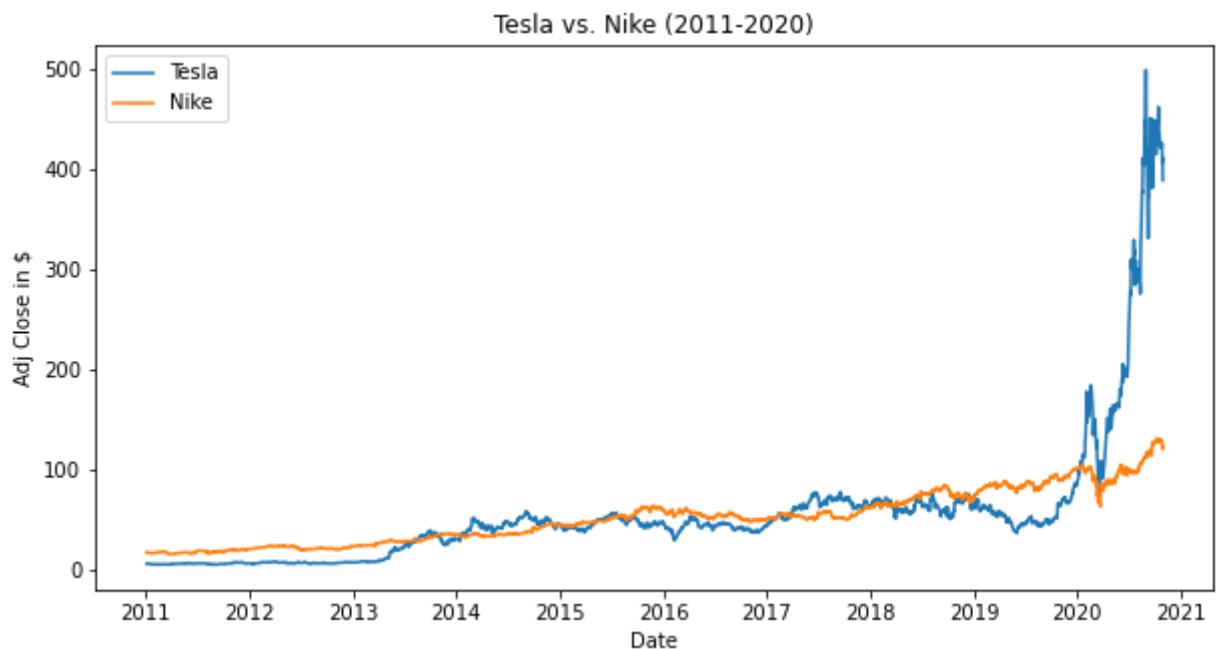
x = tesla.index      #Date
y= tesla['Adj Close'] #Price

x1 = nike.index
y1= nike["Adj Close"]

plt.xlabel("Date")
plt.ylabel("Adj Close in $")

plt.plot(x,y)
plt.plot(x1,y1)
plt.title("Tesla vs. Nike (2011-2020)")
plt.legend(["Tesla", "Nike"])
```

Out[108]: <matplotlib.legend.Legend at 0x7f855a6202b0>



3. Calculate the monthly percentage change

In [109]: *#Calculate the percentage change of the adjusted close, including all rows #but the first one (daily& monthly)*

```
tesla_daily_returns = tesla['Adj Close'].pct_change()[1:]
tesla_monthly_returns= tesla["Adj Close"].resample("M").ffill().pct_change()

nike_daily_returns = nike["Adj Close"].pct_change()[1:]
nike_monthly_returns = nike["Adj Close"].resample("M").ffill().pct_change()
```

In [110]: *#Do the same for SP500*

```
SP500_daily_returns = SP500["Adj Close"].pct_change()[1:]
SP500_monthly_returns= SP500["Adj Close"].resample("M").ffill().pct_change()
```

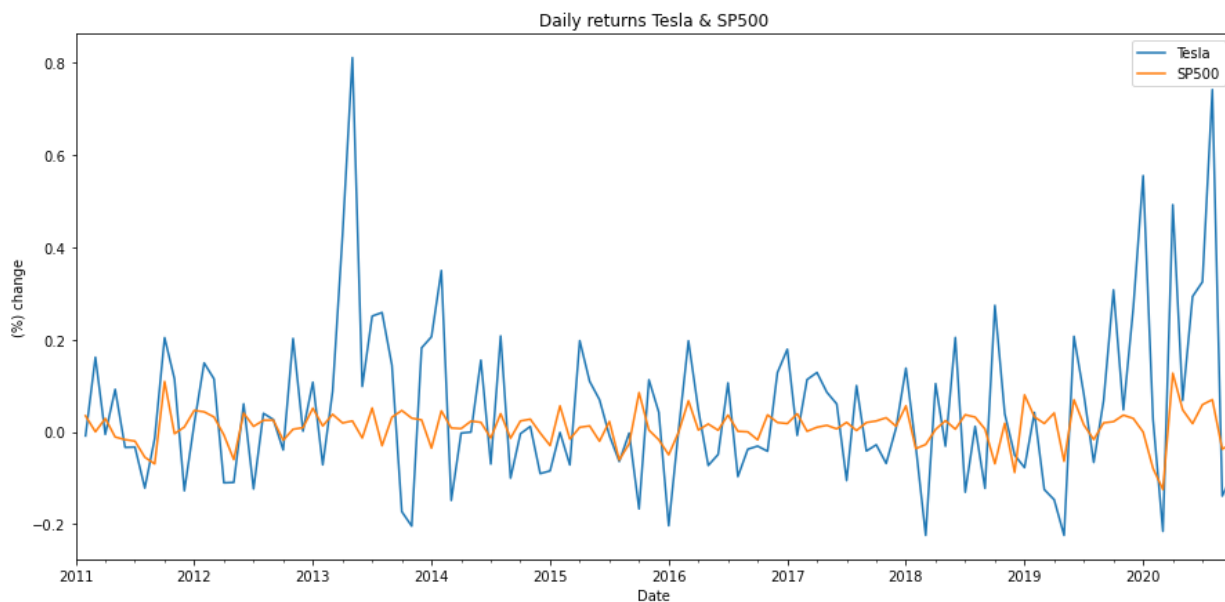
```
In [111]: nike_daily_returns.head()
```

```
Out[111]: Date
2011-01-04    -0.024625
2011-01-05     0.006550
2011-01-06    -0.008992
2011-01-07    -0.002746
2011-01-10     0.007662
Name: Adj Close, dtype: float64
```

4. Plot the monthly returns of Tesla and SP500

```
In [112]: plt.figure(figsize= (15,7))
tesla_monthly_returns.plot()
SP500_monthly_returns.plot()

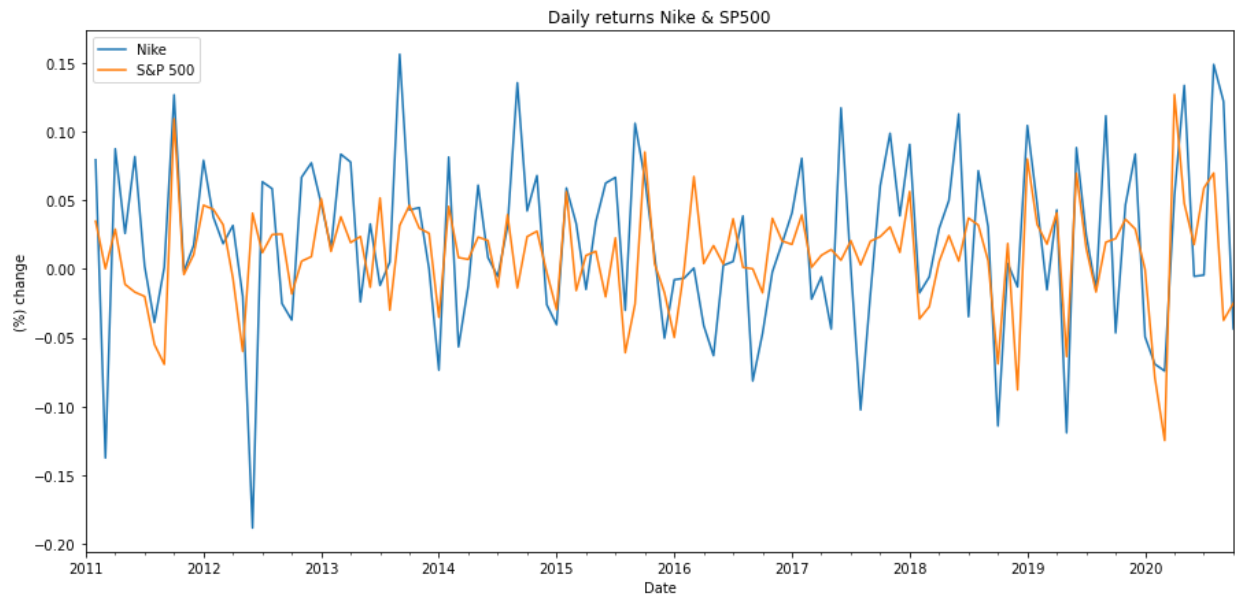
plt.xlabel("Date")
plt.ylabel("(% ) change")
plt.title("Daily returns Tesla & SP500")
plt.legend([ "Tesla", "SP500" ])
plt.show()
```



Based on the graph, Tesla looks like an aggressive stock ($\text{Beta} > 1$) but we still have to calculate it. Nike seems less aggressive and a little more defensive.

```
In [113]: plt.figure(figsize= (15,7))
nike_monthly_returns.plot()
SP500_monthly_returns.plot()

plt.xlabel("Date")
plt.ylabel("(% ) change")
plt.title("Daily returns Nike & SP500")
plt.legend(["Nike", "S&P 500"])
plt.show()
```

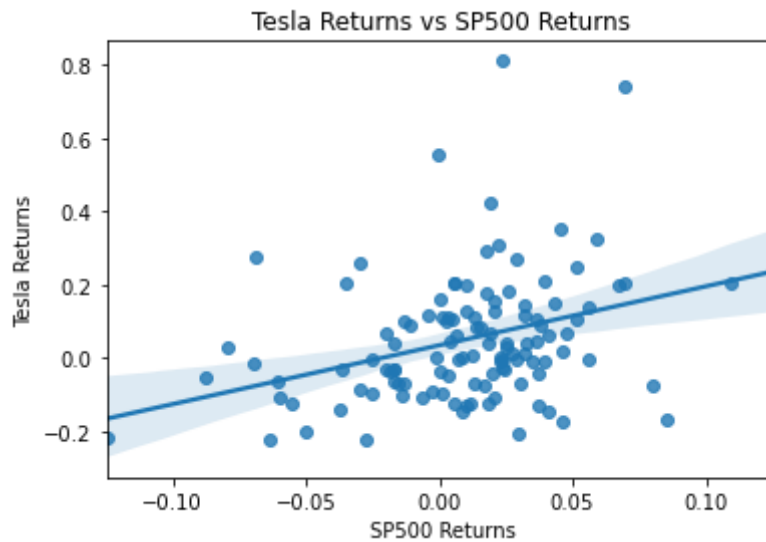


Nike seems quite similar to the market and less aggressive.

5. Graph the monthly returns of Tesla vs. SP500 returns

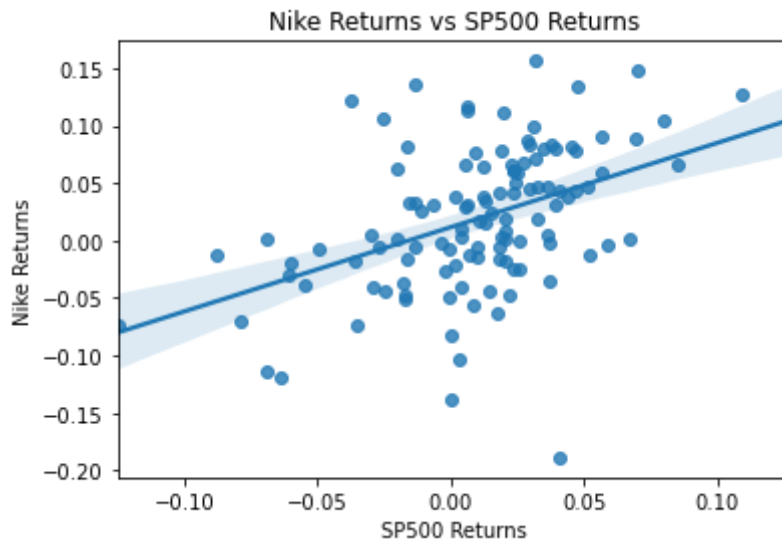
```
In [114]: fig, ax = plt.subplots()

sns.regplot(SP500_monthly_returns.values,
            tesla_monthly_returns.values)
plt.xlabel("SP500 Returns")
plt.ylabel("Tesla Returns")
plt.title("Tesla Returns vs SP500 Returns")
plt.show()
```



Interpretation: This graph illustrates the relationship between the returns of Tesla and the returns of S&P 500. The slope of the line is pretty flat and means that the beta is not that high. The scatter around the line explains the firm's specific risk. Tesla has less firm-specific risk than Nike --> see below!

```
In [115]: sns.regplot(SP500_monthly_returns.values,
nike_monthly_returns.values)
plt.xlabel("SP500 Returns")
plt.ylabel("Nike Returns")
plt.title("Nike Returns vs SP500 Returns")
plt.show()
```



Interpretation: The slope is pretty flat and shows that Nike's beta is low. Nike has a wider scatter showing that the firm had more firm-specific risks.

6. Calculate the Beta and Alpha Value of Tesla against SP500

```
In [116]: (beta, alpha) = stats.linregress(SP500_daily_returns.values,
tesla_daily_returns.values)[0:2]
```

```
In [117]: print ("Tesla is considered an aggressive stock, because its beta \nis", st
"and its alpha is ", alpha)
```

```
Tesla is considered an aggressive stock, because its beta
is 1.3108530183065377 and its alpha is 0.0016543311240569615
```

```
In [118]: (beta1, alpha1) = stats.linregress(SP500_daily_returns.values,  
                                             nike_daily_returns.values)[0:2]
```

```
In [119]: print ("Nike is considered a defensive stock, because its beta \nis", str(b  
              "and its alpha is ", alpha1)
```

Nike is considered a defensive stock, because its beta
is 0.9604703247998598 and its alpha is 0.00043455240556991527

8. Find the Standard deviation on monthly returns

```
In [120]: print("Standard Deviation Tesla:", tesla_monthly_returns.std())  
          print("Standard Deviation Nike: ", nike_monthly_returns.std())
```

Standard Deviation Tesla: 0.17663128492548535
Standard Deviation Nike: 0.062399998307264636

The Standard Deviation shows us that a Tesla's stock is more volatile than Nike.

```
In [ ]:
```