Lets take a raun of our model and see how it performs. We will set a indicator for the daily returns for SPY between 7.2% and 11%, which is rear but does happen in the mist of economic crisis. This is a perfect time to look that the 08 crash as well as the current COViD-19 epidemic

```
In [ ]:
```

Libraries

```
In [2]: import pandas as pd
import numpy as np
from pandas_datareader import data as web
import matplotlib.pyplot as plt
import seaborn as sns
```

import data

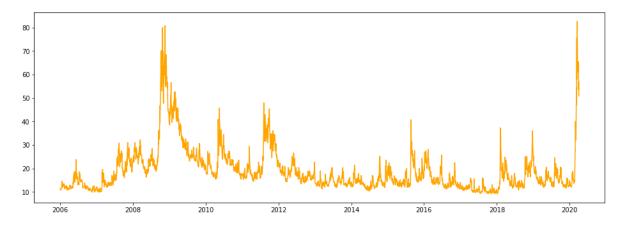
We will be looking at the adjustable close for both the S&P and VIX

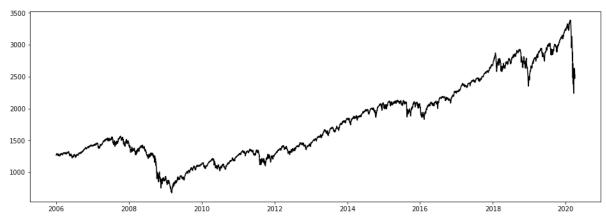
```
In [4]: tickers = ["^VIX", "^GSPC"]
    data = pd.DataFrame()
    for t in tickers:
        data[t] = web.DataReader(t, data_source="yahoo", start = "2006-1-1")
        ["Adj Close"]
```

plotting visual movement for both VIX and SPY .As expected the plots show an inverse correlation

```
In [15]: fig, axes = plt.subplots(2,1,figsize = (16,12))
    axes[0].plot(data["^VIX"], c = "orange")
    axes[1].plot(data["^GSPC"], c = "black")
```

Out[15]: [<matplotlib.lines.Line2D at 0x1286160b8>]





Tracking thre daily returns for each security. We will look at the volitility of the daily returns on a histogram below. what we see is most of the daily returns of SPY between -0.05, and 0.05, where VIX is between -0.2, and .2 with a few outlyers

```
sns.set_style("whitegrid")
In [23]:
            data[["^VIX daily returns","^GSPC daily returns"]].hist(figsize = (16,5)
            ), bins = 70);
                              AGSPC daily returns
                                                                                  ^VIX daily returns
             800
             700
             600
             500
                                                                 300
             400
             300
                                                                 200
             200
                                                                 100
```

The reason for this analysis is the test the theory that as the daily returns of SPY is between 7% and 11% we can see a sharp increase in VIX. We will test this below but first we will creat a hypithetical portfolio containing both SPY and VIX . We will weigh the portfolio 40% SPY and 60% VIX with a position of 150K in the portfolio. Our other assumption is that we are getting in the market on the first date of 1-3-2006

Normal Returns

	^VIX	^GSPC	^VIX daily returns	^GSPC daily returns	^VIX normal returns	^GSPC normal returns
Date						
2006-01- 03	11.14	1268.800049	NaN	NaN	1.000000	1.000000
2006-01- 04	11.37	1273.459961	0.020646	0.003673	1.020646	1.003673
2006-01- 05	11.31	1273.479980	-0.005277	0.000016	1.015260	1.003688
2006-01- 06	11.00	1285.449951	-0.027409	0.009399	0.987433	1.013123
2006-01- 09	11.13	1290.150024	0.011818	0.003656	0.999102	1.016827

Allocations

	^VIX	^GSPC	^VIX daily returns	^GSPC daily returns	^VIX normal returns	^GSPC normal returns	^VIX allocation	^GSPC allocation	
Date									
2006- 01-03	11.14	1268.800049	NaN	NaN	1.000000	1.000000	0.400000	0.600000	•
2006- 01-04	11.37	1273.459961	0.020646	0.003673	1.020646	1.003673	0.408259	0.602204	
2006- 01-05	11.31	1273.479980	-0.005277	0.000016	1.015260	1.003688	0.406104	0.602213	
2006- 01-06	11.00	1285.449951	-0.027409	0.009399	0.987433	1.013123	0.394973	0.607874	
2006- 01-09	11.13	1290.150024	0.011818	0.003656	0.999102	1.016827	0.399641	0.610096	

Position -> 150K

In [33]:

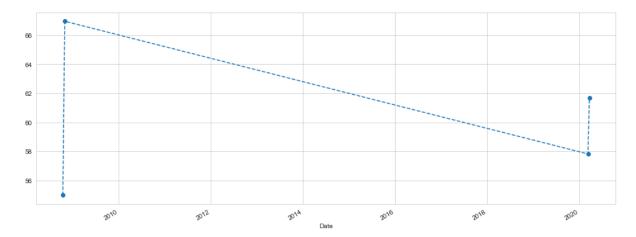
Out[33]:

```
data.head()
                                 ^VIX
                                         ^GSPC
                                                      ^VIX
                                                              ^GSPC
                                                                           ^VIX
                                                                                    ^GSPC
         ^VIX
                    ^GSPC
                                                                                             ^VIX p
                                 daily
                                           daily
                                                    normal
                                                              normal
                                                                      allocation
                                                                                 allocation
                               returns
                                         returns
                                                              returns
                                                   returns
 Date
2006-
              1268.800049
                                 NaN
                                                 1.000000
                                                           1.000000
                                                                       0.400000
                                                                                  0.600000 60000.0
        11.14
                                            NaN
01-03
2006-
        11.37
              1273.459961
                             0.020646
                                       0.003673
                                                  1.020646
                                                            1.003673
                                                                       0.408259
                                                                                  0.602204
                                                                                             61238.
01-04
2006-
        11.31
               1273.479980
                            -0.005277
                                       0.000016
                                                  1.015260
                                                            1.003688
                                                                       0.406104
                                                                                  0.602213 60915.6
01-05
2006-
        11.00
              1285.449951
                            -0.027409
                                       0.009399
                                                  0.987433
                                                            1.013123
                                                                       0.394973
                                                                                   0.607874
                                                                                            59245.9
01-06
2006-
        11.13
             1290.150024
                             0.011818 0.003656
                                                  0.999102 1.016827
                                                                       0.399641
                                                                                  0.610096 59946.
01-09
```

Lets take a raun of our model and see how it performs. We will set a indicator for the daily returns for SPY and 7.2%, which is rear but does happen in the mist of economic crisis. This is a perfect time to look that the 08 crash as well as the curren COViD-19 epidemic

We can see there are 4 dates that this occurred and we just missed 2. in 2020

Out[57]: <matplotlib.axes._subplots.AxesSubplot at 0x133390d30>

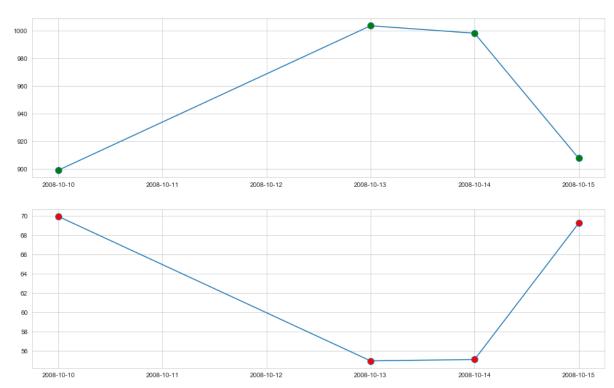


lets take a look at each date where SpY was 7.2 or above. To get a better look at this we will take a few days before and after the set date

2008-10-13

```
In [78]: fig, axes = plt.subplots(2,1, figsize = (16,10))
    axes[0].plot(data.loc["2008-10-10":"2008-10-15"]["^GSPC"], marker = "o",
    mfc = "g", ms = 10)
    axes[1].plot(data.loc["2008-10-10":"2008-10-15"]["^VIX"], marker = "o",
    mfc = "r", ms = 10)
```

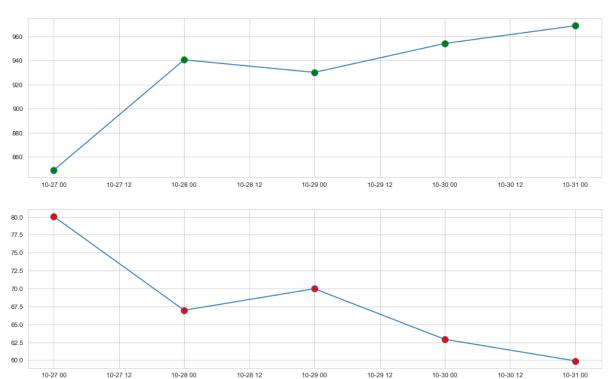
Out[78]: [<matplotlib.lines.Line2D at 0x135f66b70>]



2008-10-28

```
In [83]: fig, axes = plt.subplots(2,1, figsize = (16,10));
    axes[0].plot(data.loc["2008-10-25":"2008-10-31"]["^GSPC"], marker = "o",
    mfc = "g", ms = 10)
    axes[1].plot(data.loc["2008-10-25":"2008-10-31"]["^VIX"], marker = "o",
    mfc = "r", ms = 10)
```

Out[83]: [<matplotlib.lines.Line2D at 0x136762d30>]



2020-03-13

```
In [86]: fig, axes = plt.subplots(2,1, figsize = (16,10));
    axes[0].plot(data.loc["2020-03-10":"2020-03-17"]["^GSPC"], marker = "o",
    mfc = "g", ms = 10)
    axes[1].plot(data.loc["2020-03-10":"2020-03-17"]["^VIX"], marker = "o",
    mfc = "r", ms = 10)
```

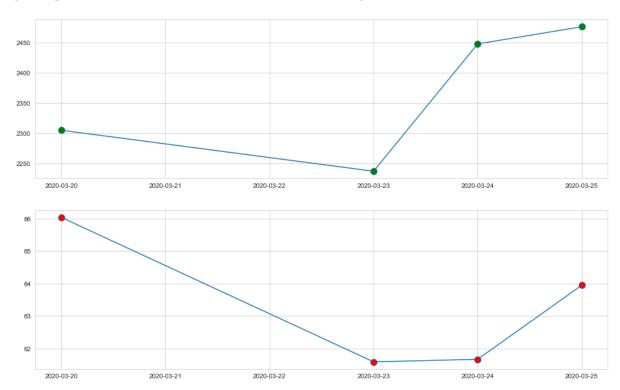
Out[86]: [<matplotlib.lines.Line2D at 0x136da27f0>]



2020-03-24

```
In [89]: fig, axes = plt.subplots(2,1, figsize = (16,10));
    axes[0].plot(data.loc["2020-03-20":"2020-03-25"]["^GSPC"], marker = "o",
    mfc = "g", ms = 10)
    axes[1].plot(data.loc["2020-03-20":"2020-03-25"]["^VIX"], marker = "o",
    mfc = "r", ms = 10)
```

Out[89]: [<matplotlib.lines.Line2D at 0x1375d5d30>]



in summary ad SPY daily returns is in the range of 7.2 and 11 it is almost certain that VIX will have a significant gain. We cna see this clearly in 2020 on the days of 3/13 and 3 /24

In []: