Importing Libraries

```
In [1]:
        import numpy as np
        import pandas as pd
         import pandas datareader.data as web
        import datetime
        import matplotlib.pyplot as plt
        %matplotlib inline
```

Getting Data

```
In [2]:
         start = datetime.date(1982, 4, 20)
         end = datetime.date(2021,12,31)
         data = web.DataReader(name='^GSPC',data source='yahoo',start=start,end=end)
         data.drop(columns=['Volume','Adj Close'], inplace=True)
In [3]:
         #creating raw Excel file with data
         #data.to excel(excel writer='S&P500 Daily.xlsx',sheet name='Data')
In [4]:
         #check for NaN values
         if data.shape == data.dropna().shape:
             print('No empty cells \n')
         else:
             num drop = data.shape[0]-data.dropna().shape[0]
             print('{} rows dropped due to empty cells \n'.format(num drop))
         print('Database data types:\n{}'.format(data.dtypes))
         data.head()
        No empty cells
        Database data types:
        High float64
                float64
        Low
                float64
        Open
        Close float64
        dtype: object
Out[4]:
                       High
                                 Low
                                          Open
                                                    Close
             Date
        1982-04-20 117.139999 114.830002 115.800003 115.440002
        1982-04-21 115.870003 115.300003 115.480003 115.720001
        1982-04-22 117.250000 115.720001 115.720001 117.190002
        1982-04-23 118.639999 117.190002 118.019997 118.639999
```

Returns Calculations

1982-04-26 119.330002 118.250000 118.940002 119.260002

```
In [5]:
        data['Daily Change'] = ((data['Close']-data['Open'])/data['Open'])*100
        data['True Range'] = ((data['High']-data['Low'])/data['Low'])*100
        data.head()
```

Date						
1982-04-20	117.139999	114.830002	115.800003	115.440002	-0.310881	2.011667
1982-04-21	115.870003	115.300003	115.480003	115.720001	0.207826	0.494362
1982-04-22	117.250000	115.720001	115.720001	117.190002	1.270309	1.322156
1982-04-23	118.639999	117.190002	118.019997	118.639999	0.525337	1.237304
1982-04-26	119.330002	118.250000	118.940002	119.260002	0.269043	0.913321

Open

Statistical calculations

High

Low

Out[5]:

Close Daily_Change True_Range

```
Out[6]:
                   Daily_Change True_Range
                        10012.00
                                     10012.00
           count
                            0.04
                                          1.24
           mean
                                          0.96
                            1.07
             std
                                          0.00
             min
                          -20.47
                                          0.67
            25%
                           -0.42
            50%
                            0.06
                                          1.00
            75%
                            0.54
                                          1.50
                           10.79
                                         25.74
            max
```

```
In [7]:
                  calculations1 = {
                           'average returns':np.average(data.Daily Change),
                           'average positive return':data.Daily Change[data.Daily Change > 0].agg(func='mean'),
                           'average negative return':data.Daily Change[data.Daily Change < 0].agg(func='mean'),</pre>
                           'positive change': (data.Daily Change[data.Daily Change > 0].agg(func='count')/len(data
                           'negative change': (data.Daily Change[data.Daily Change < 0].agg(func='count')/len(date
                           'no change':(data.Daily Change[data.Daily Change == 0].agg(func='count')/len(data.Dail
                  calculations2 = {
                           'adjusted poitive return': (calculations1['average positive return']*calculations1['pos
                           'adjusted negative return': (calculations1['average negative return']*calculations1['ne
                           '1 std dev upper':descriptive statistics.Daily Change.loc['mean']+(1*descriptive stati
                           '1 std dev lower':descriptive statistics.Daily Change.loc['mean']-(1*descriptive stati
                           '2 std dev upper':descriptive statistics.Daily Change.loc['mean']+(2*descriptive stati
                                 std dev lower':descriptive statistics.Daily Change.loc['mean']-(2*descriptive statistics.Daily Change.l
                           '3 std dev upper':descriptive statistics.Daily Change.loc['mean']+(3*descriptive stati
                           '3 std dev lower':descriptive statistics.Daily Change.loc['mean']-(3*descriptive stati
                           '1 std dev normal dist':68.20,
                            '2_std_dev_normal_dist':95.40,
                           '3 std dev normal dist':99.80}
                  calculations3 = {
                           '1 std dev actual dist':(len(data[data.Daily Change.ge(calculations2['1 std dev lower
                                            data.Daily Change.le(calculations2['1 std dev upper'])])/descriptive statistic
                            '2 std dev actual dist':(len(data[data.Daily Change.ge(calculations2['2 std dev lower
                                            data.Daily Change.le(calculations2['2 std dev upper'])])/descriptive statistic
                           '3 std dev actual dist': (len(data[data.Daily Change.ge(calculations2['3 std dev lower'
```

data.Daily Change.le(calculations2['3 std dev upper'])])/descriptive statistic

```
# concatinating all calculations into a dataframe

calc_df = pd.concat([
    pd.DataFrame(calculations1,index=[0]).transpose(),
    pd.DataFrame(calculations2,index=[0]).transpose(),
    pd.DataFrame(calculations3,index=[0]).transpose()])

calc_df.rename(columns={0:'Values (in %)'},inplace=True)
calc_df['Values (in %)'] = calc_df['Values (in %)'].round(3)
```

```
In [8]: calc_df
```

	Values (in %)
average_returns	0.038
average_positive_return	0.701
average_negative_return	-0.729
positive_change	53.546
negative_change	46.314
no_change	0.140
adjusted_poitive_return	0.375
adjusted_negative_return	-0.338
1_std_dev_upper	1.110
1_std_dev_lower	-1.030
2_std_dev_upper	2.180
2_std_dev_lower	-2.100
3_std_dev_upper	3.250
3_std_dev_lower	-3.170
1_std_dev_normal_dist	68.200
2_std_dev_normal_dist	95.400
3_std_dev_normal_dist	99.800
1_std_dev_actual_dist	79.405
2_std_dev_actual_dist	95.076
3_std_dev_actual_dist	98.532

Out[8]:

Grouping data into bins and calculating frequencies

```
In [9]:
    data['Daily Change Bins'] = pd.cut(data.Daily_Change, bins=np.linspace(-4,4,33))
    data['True Range Bins'] = pd.cut(data.True_Range, bins=np.linspace(0,4,33))

#daily change frequencies
    freq1 = pd.DataFrame(data['Daily Change Bins'].value_counts().sort_index())
    freq1.reset_index(inplace=True)
    freq1.rename(columns={'index':'Bins_DC','Daily Change Bins':'Frequency_DC'},inplace=True)

freq1_sum = np.sum(freq1.Frequency_DC)
    freq1['Probabilities_DC'] = (freq1.Frequency_DC/freq1_sum)*100
    freq1['Cumulative_DC'] = freq1.Probabilities_DC.cumsum()
```

```
#true range frequencies
freq2 = pd.DataFrame(data['True Range Bins'].value_counts().sort_index())
freq2.reset_index(inplace=True)
freq2.rename(columns={'index':'Bins_TR','True Range Bins':'Frequency_TR'},inplace=True)

freq2_sum = np.sum(freq2.Frequency_TR)
freq2['Probabilities_TR'] = (freq2.Frequency_TR/freq2_sum)*100
freq2['Cumulative_TR'] = freq2.Probabilities_TR.cumsum()

#concatinating 2 frequencies
freq = pd.concat([freq1,freq2],axis=1)
```

In [10]:

freq

Out[10]:	Bins_DC Frequency_DC		Probabilities_DC	Cumulative_DC	Bins_TR	Frequency_TR	Probabilities_TR	Cumulative_T	
	0	(-4.0, -3.75]	11	0.110709	0.110709	(0.0, 0.125]	0	0.000000	0.00000
	1	(-3.75, -3.5]	9	0.090580	0.201288	(0.125, 0.25]	40	0.406339	0.40633
	2	(-3.5, -3.25]	9	0.090580	0.291868	(0.25, 0.375]	347	3.524990	3.93132
	3	(-3.25, -3.0]	23	0.231481	0.523349	(0.375, 0.5]	795	8.075985	12.00731
	4	(-3.0, -2.75]	23	0.231481	0.754831	(0.5, 0.625]	973	9.884193	21.89150
	5	(-2.75, -2.5]	43	0.432770	1.187601	(0.625, 0.75]	1004	10.199106	32.09061
	6	(-2.5, -2.25]	58	0.583736	1.771337	(0.75, 0.875]	974	9.894352	41.98496
	7	(-2.25, -2.0]	74	0.744767	2.516103	(0.875, 1.0]	901	9.152783	51.13774
	8	(-2.0, -1.75]	119	1.197665	3.713768	(1.0, 1.125]	786	7.984559	59.12230
	9	(-1.75, -1.5]	182	1.831723	5.545491	(1.125, 1.25]	665	6.755384	65.87769
	10	(-1.5, -1.25]	213	2.143720	7.689211	(1.25, 1.375]	566	5.749695	71.62738
	11	(-1.25, -1.0]	305	3.069646	10.758857	(1.375, 1.5]	454	4.611946	76.23933
	12	(-1.0, -0.75]	479	4.820853	15.579710	(1.5, 1.625]	387	3.931329	80.17066
	13	(-0.75, -0.5]	672	6.763285	22.342995	(1.625, 1.75]	320	3.250711	83.42137
	14	(-0.5, -0.25]	969	9.752415	32.095411	(1.75, 1.875]	274	2.783421	86.20479
	15	(-0.25, 0.0]	1424	14.331723	46.427134	(1.875, 2.0]	204	2.072328	88.27712
	16	(0.0, 0.25]	1545	15.549517	61.976651	(2.0, 2.125]	206	2.092645	90.36976

	Bins_DC	Frequency_DC	Probabilities_DC	Cumulative_DC	Bins_TR	Frequency_TR	Probabilities_TR	Cumulative_T
17	(0.25, 0.5]	1181	11.886071	73.862721	(2.125, 2.25]	160	1.625356	91.99512
18	(0.5, 0.75]	843	8.484300	82.347021	(2.25, 2.375]	142	1.442503	93.43762
19	(0.75, 1.0]	583	5.867552	88.214573	(2.375, 2.5]	100	1.015847	94.45347
20	(1.0, 1.25]	404	4.066023	92.280596	(2.5, 2.625]	98	0.995530	95.44900
21	(1.25, 1.5]	242	2.435588	94.716184	(2.625, 2.75]	72	0.731410	96.18041
22	(1.5, 1.75]	175	1.761272	96.477456	(2.75, 2.875]	69	0.700935	96.88134
23	(1.75, 2.0]	109	1.097021	97.574477	(2.875, 3.0]	62	0.629825	97.51117
24	(2.0, 2.25]	68	0.684380	98.258857	(3.0, 3.125]	35	0.355547	97.86672
25	(2.25, 2.5]	64	0.644122	98.902979	(3.125, 3.25]	48	0.487607	98.35432
26	(2.5, 2.75]	34	0.342190	99.245169	(3.25, 3.375]	35	0.355547	98.70987
27	(2.75, 3.0]	23	0.231481	99.476651	(3.375, 3.5]	33	0.335230	99.04510
28	(3.0, 3.25]	12	0.120773	99.597424	(3.5, 3.625]	25	0.253962	99.29906
29	(3.25, 3.5]	19	0.191224	99.788647	(3.625, 3.75]	28	0.284437	99.58350
30	(3.5, 3.75]	11	0.110709	99.899356	(3.75, 3.875]	21	0.213328	99.79683
31	(3.75, 4.0]	10	0.100644	100.000000	(3.875, 4.0]	20	0.203169	100.00000

Plotting Returns

```
In [11]:
```

```
#returns plot function
def returns dc():
    fig = plt.figure(figsize=(13,8))
    n, bins, patches = plt.hist(data.Daily_Change,bins=np.linspace(-4,4,33),rwidth=0.5,alk
   patches[16].set fc('red')
   plt.xticks(np.linspace(-4,4,33))
   plt.yticks(range(0,1700,100))
    x = plt.gca().xaxis
    for item in x.get ticklabels():
        item.set rotation(60)
    plt.title('Open to Close Returns',fontdict={'size':20,'weight':'bold','family':'Times
    plt.xlabel('Bins',fontdict={'size':18,'family':'Times New Roman'})
   plt.ylabel('Frequency', fontdict={'size':18,'family':'Times New Roman'})
    plt.grid(alpha=0.3)
```

```
plt.tight_layout()

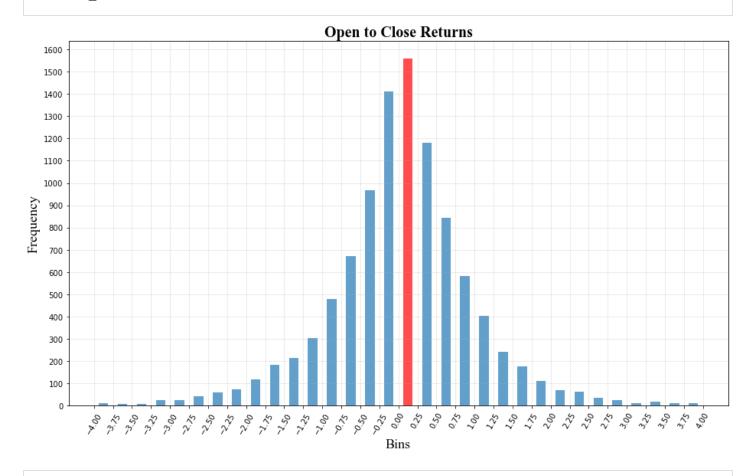
def returns_tr():
    fig = plt.figure(figsize=(13,8))
    n, bins, patches = plt.hist(data.True_Range,bins=np.linspace(0,4,33),rwidth=0.5,alpha=patches[5].set_fc('red')

plt.xticks(np.linspace(0,4,33))
    plt.yticks(range(0,1100,100))
    x = plt.gca().xaxis
    for item in x.get_ticklabels():
        item.set_rotation(60)

plt.title('High to Low Returns',fontdict={'size':20,'weight':'bold','family':'Times Neplt.xlabel('Bins',fontdict={'size':18,'family':'Times New Roman'})
    plt.ylabel('Frequency',fontdict={'size':18,'family':'Times New Roman'})
    plt.grid(alpha=0.3)

plt.tight_layout()
```

In [12]: returns dc()



```
In [13]: returns_tr()
```

