

Project Day -16

[Gold Price Prediction Project Using Machine Learning]

```
In [138]: import pandas as pd
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn import metrics
```

Data Collection and Processing

```
In [139]: data=pd.read_csv('gold_price_data.csv')
data
```

	Date	SPX	GLD	USO	SLV	EUR/USD
0	1/2/2008	1447.160034	84.860001	78.470001	15.1800	1.471692
1	1/2/2008	1447.160034	85.570000	78.370003	15.2850	1.474491
2	1/4/2008	1411.630005	85.129997	77.309998	15.1670	1.475492
3	1/7/2008	1416.180054	84.769997	75.500000	15.0530	1.468299
4	1/8/2008	1390.189941	86.779999	76.059998	15.5900	1.557099
...
2285	5/8/2018	2671.919922	124.589996	14.060000	15.5100	1.186789
2286	5/9/2018	2697.790039	124.330002	14.370000	15.5300	1.184722
2287	5/10/2018	2723.070068	125.180000	14.410000	15.7400	1.191753
2288	5/14/2018	2730.129883	124.489998	14.380000	15.5600	1.193118
2289	5/16/2018	2725.780029	122.543800	14.405800	15.4542	1.182033

2290 rows × 6 columns

```
In [141]: # Top 5 rows of dataset
data.head()
```

	Date	SPX	GLD	USO	SLV	EUR/USD
0	1/2/2008	1447.160034	84.860001	78.470001	15.1800	1.471692
1	1/3/2008	1447.160034	85.570000	78.370003	15.285	1.474491
2	1/4/2008	1411.630005	85.129997	77.309998	15.167	1.475492
3	1/7/2008	1416.180054	84.769997	75.500000	15.053	1.468299
4	1/8/2008	1390.189941	86.779999	76.059998	15.590	1.557099

```
In [142]: # Last 5 rows of dataset
data.tail()
```

	Date	SPX	GLD	USO	SLV	EUR/USD
2285	5/8/2018	2671.919922	124.589996	14.0600	15.5100	1.186789
2286	5/9/2018	2697.790039	124.330002	14.3700	15.5300	1.184722
2287	5/10/2018	2723.070068	125.180000	14.410000	15.7400	1.191753
2288	5/14/2018	2730.129883	124.489998	14.380000	15.5600	1.193118
2289	5/16/2018	2725.780029	122.543800	14.405800	15.4542	1.182033

```
In [143]: # number of rows and column in dataframe
data.shape
```

```
Out[143]: (2290, 6)
```

```
In [144]: # basic information of dataset
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2290 entries, 0 to 2289
Data columns (total 6 columns):
 #   Column      Non-Null Count  Dtype   Format
 --  --          --          --          --
 0   Date        2290 non-null  object  
 1   SPX         2290 non-null  float64 
 2   GLD         2290 non-null  float64 
 3   USO         2290 non-null  float64 
 4   SLV         2290 non-null  float64 
 5   EUR/USD    2290 non-null  float64 
dtypes: float64(5), object(1)
memory usage: 107.5+ KB
```

```
In [145]: # check null values in dataset
data.isnull().sum()
```

```
Out[145]: Date          0
SPX          0
GLD          0
USO          0
SLV          0
EUR/USD     0
dtype: int64
```

```
In [151]: # getting the statistical measurement of the dataset
data.describe()
```

	SPX	GLD	USO	SLV	EUR/USD
count	2290.000000	2290.000000	2290.000000	2290.000000	2290.000000
mean	1654.315776	122.732875	31.842221	20.084997	1.283863
std	519.115140	23.283346	19.523517	7.092566	0.131547
min	676.530029	70.000000	7.960000	8.850000	1.039047
25%	1239.674969	109.725000	14.380000	15.570000	1.171313
50%	1551.434998	120.580002	33.869998	17.268950	1.303297
75%	2073.010070	132.840004	37.827501	22.882500	1.369971
max	2872.870118	184.589996	117.498003	47.259998	1.598798

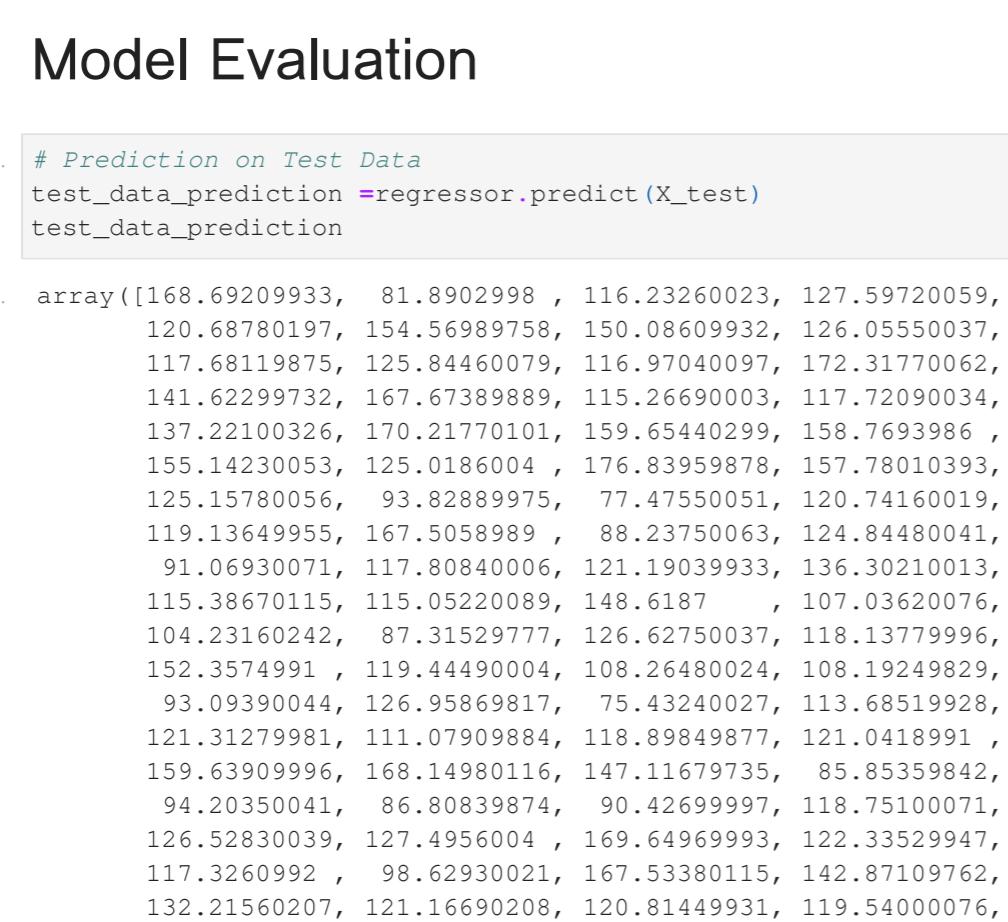
Correlation:

```
In [154]: # Positive Correlation
Correlation = data.select_dtypes(include='number').corr()
Correlation
```

```
Out[154]: SPX      GLD      USO      SLV      EUR/USD
```

```
SPX    1.000000 -0.049345 -0.591573 -0.274055 -0.672017
GLD    0.049345  1.000000 -0.186360  0.866632 -0.024375
USO   -0.591573  1.000000  0.167549  0.167549  0.829149
SLV    0.866632  0.167549  1.000000  0.321631  0.024375
EUR/USD -0.672017  0.024375  0.829149  1.000000  0.705000
```

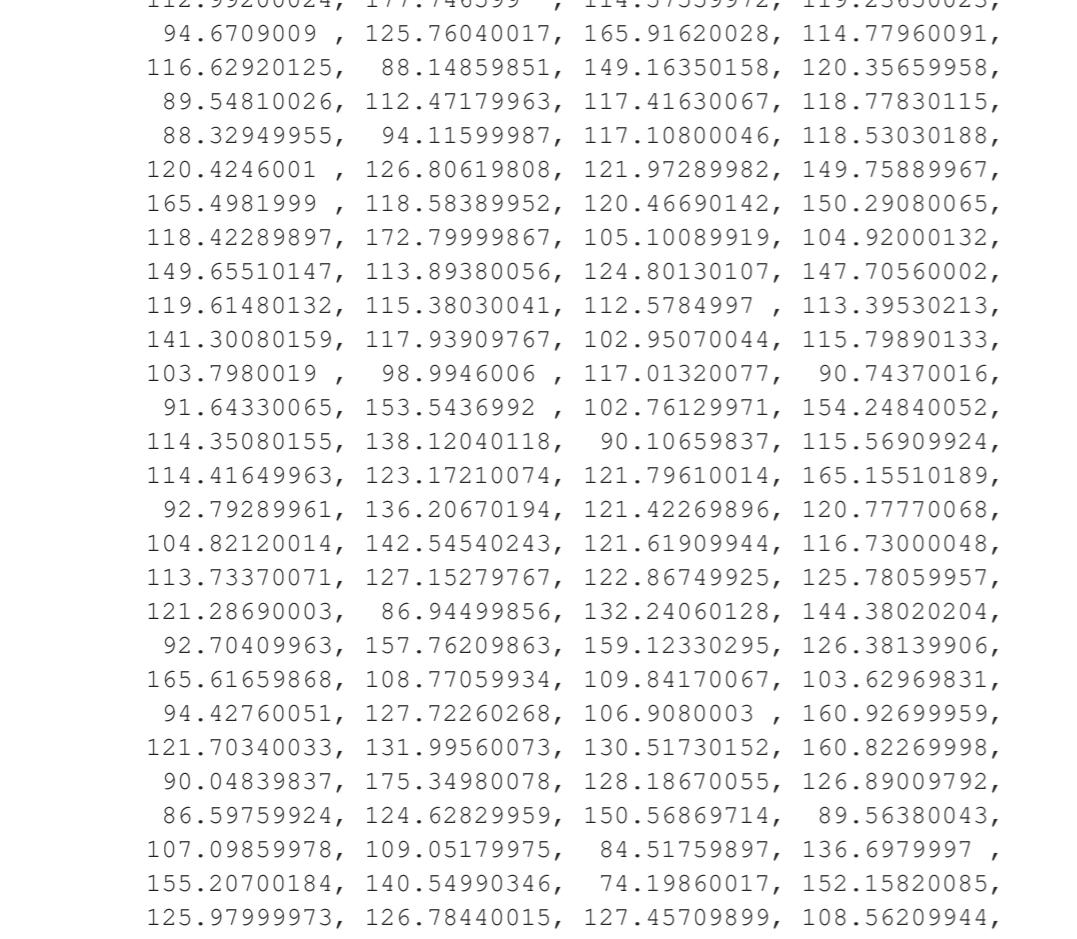
```
In [156]: # constructing a heatmap to understand the correlations
plt.figure(figsize=(8,8))
sns.heatmap(Correlation, annot=True, square=True, fmt=".2f", annot_kws={"size": 8}, cmap="Blues")
plt.show()
```



```
In [158]: # correlation values of GLD
print(Correlation['GLD'])
```

```
SPX    0.049345
GLD    1.000000
USO   -0.186360
SLV    0.866632
EUR/USD -0.024375
Name: GLD, dtype: float64
```

```
In [160]: # checking the distribution of the GLD Price
sns.histplot(data['GLD'], kde=True, color='green')
plt.show()
```



Splitting the Features and Target

```
In [161]: X=data.drop(['Date','GLD'],axis=1)
ydata['GLD']
```

```
Out[161]: X
```

```
SPX      USO      SLV      EUR/USD
0   1447.160034  78.470001  15.1800  1.471692
1   1447.160034  78.370003  15.2850  1.474491
2   1411.630005  77.309998  15.1670  1.475492
3   1416.180054  75.500000  15.0530  1.468299
4   1390.189941  76.059998  15.5900  1.557099
...
2285  2671.919922  14.060000  15.5100  1.186789
2286  2697.790039  14.370000  15.5300  1.184722
2287  2723.070068  14.410000  15.7400  1.191753
2288  2730.129883  14.380000  15.5600  1.193118
2289  2725.780029  14.405800  15.4542  1.182033
2290 rows × 4 columns
```

```
In [167]: Y
```

```
Out[167]: 0   84.860001
1   85.570000
2   85.129997
3   84.769997
4   86.779999
...
2285 124.589996
2286 124.330002
2287 125.180000
2288 124.889998
2289 122.543800
Name: GLD, Length: 2290, dtype: float64
```

Splitting into Training data and Test Data

```
In [170]: X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.2,random_state=2)
```

Model Training : Random Forest Regressor

```
In [173]: regressor = RandomForestRegressor(n_estimators=100)
```

```
In [175]: # Training the model
regressor.fit(X_train,Y_train)
```

```
Out[175]: RandomForestRegressor()
```

Model Evaluation

```
In [177]: # Prediction on Test Data
test_data_prediction = regressor.predict(X_test)
```

```
test_data_prediction
```

```
array([168.66903935,  91.8502989,  116.23260003,  127.59790059,
       120.69780137, 154.55899758, 150.88609922, 126.35530007,
       117.68119875, 125.84460078, 116.39704007, 172.31770006,
       141.62299732, 167.6738988, 115.26690034, 117.72909934,
       137.22100326, 125.0186004, 176.83959875, 157.78010933,
       155.14230052, 125.1800004, 125.64402007, 123.68119895,
       121.65190165, 114.18499814, 121.65190165, 121.65190165,
       119.13649955, 167.5058989, 88.23750063, 124.84480041,
       91.06930071, 117.80840008, 121.19399393, 136.50210013,
       115.38670124, 115.05220089, 148.6187 , 107.03620076,
       104.38160124, 87.3110777, 126.62750057, 118.13770006,
       121.12800059, 125.5710000, 125.5710000, 125.5710000,
       119.79309044, 126.95869815, 125.43240027, 113.48519328,
       131.31729983, 111.10132007, 117.0132007, 117.0132007,
       123.50770027, 110.17439841, 108.26779983, 127.26089909,
       116.13901065, 114.18249982, 121.65190164, 121.65190164,
       115.85699858, 126.82330021, 125.86119894, 127.45939695,
       124.20350041, 86.80400074, 107.41630067, 118.75100015,
       113.50770027, 110.17439841, 108.26779983, 127.26089909,
       116.13901065, 114.18249982, 121.65190164, 121.65190164,
       115.85699858, 126.82330021, 125.86119894, 127.45939695,
       123.51200031, 121.1800004, 121.1800004, 121.1800004,
       121.1800004, 120.1690004, 120.1690004, 120.1690004,
       120.1690004, 118.12800059, 118.12800059, 118.12800059,
       118.12800059, 116.12700059, 116.12700059, 116.12700059,
       116.12700059, 114.12600059, 114.12600059, 114.12600059,
       114.12600059, 112.12500059, 112.12500059, 112.12500059,
       112.12500059, 110.12400059, 110.12400059, 110.12400059,
       110.12400059, 108.12300059, 108.12300059, 108.12300059,
       108.12300059
```

** END - PROJECT **

In []: