1. Impoting Dependencies:

C-Number of row & columns:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn import svm
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestRegressor
from sklearn import metrics
```

2. Data Collcetion and Analysis:

A- Loading dataset:

```
In [ ]: data_gold = pd.read_csv("C:/Machine_learning Python/projets/GoldPrice/gld_price_
```

B- View the data (head)

In []: data_gold.head()

Out[]:		Date	SPX	GLD	USO	SLV	EUR/USD
	0	1/2/2008	1447.160034	84.860001	78.470001	15.180	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

C- View the last 5 row of the data:

In []: data_gold.tail()

0

Out[]:		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	14.4100	15.7400	1.191753
	2288	5/14/2018	2730.129883	124.489998	14.3800	15.5600	1.193118
	2289	5/16/2018	2725.780029	122.543800	14.4058	15.4542	1.182033

D- type of the data (head)

```
type(data_gold)
Out[]: pandas.core.frame.DataFrame
        E-Number of row & columns:
        data_gold.shape
Out[]: (2290, 6)
        F- Information about the data:
In [ ]: data_gold.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 2290 entries, 0 to 2289
       Data columns (total 6 columns):
            Column
                    Non-Null Count Dtype
            ____
                     -----
           Date
                     2290 non-null
                                     object
        0
            SPX
                     2290 non-null float64
                     2290 non-null float64
            GLD
        2
            US0
                     2290 non-null
                                     float64
                     2290 non-null
                                     float64
            SLV
            EUR/USD 2290 non-null
                                     float64
       dtypes: float64(5), object(1)
       memory usage: 107.5+ KB
          2. Statisctical measures:
        A- General Statistic:
        data gold.describe()
Out[ ]:
                      SPX
                                  GLD
                                               USO
                                                            SLV
                                                                   EUR/USD
               2290.000000 2290.000000 2290.000000 2290.000000 2290.000000
         count
               1654.315776
                                                      20.084997
         mean
                             122.732875
                                          31.842221
                                                                    1.283653
                519.111540
                                                       7.092566
                                                                    0.131547
           std
                              23.283346
                                          19.523517
                                                                    1.039047
          min
                676.530029
                              70.000000
                                           7.960000
                                                       8.850000
                             109.725000
          25%
               1239.874969
                                          14.380000
                                                      15.570000
                                                                    1.171313
          50% 1551.434998
                                                                    1.303297
                             120.580002
                                          33.869999
                                                      17.268500
          75% 2073.010070
                             132.840004
                                          37.827501
                                                      22.882500
                                                                    1.369971
```

B- Number of missing value in each column;

184.589996

117.480003

47.259998

1.598798

In []: data_gold.isnull().sum()
#Any missing values

max 2872.870117

```
Out[]: Date 0

SPX 0

GLD 0

USO 0

SLV 0

EUR/USD 0

dtype: int64
```

3. Correlation:

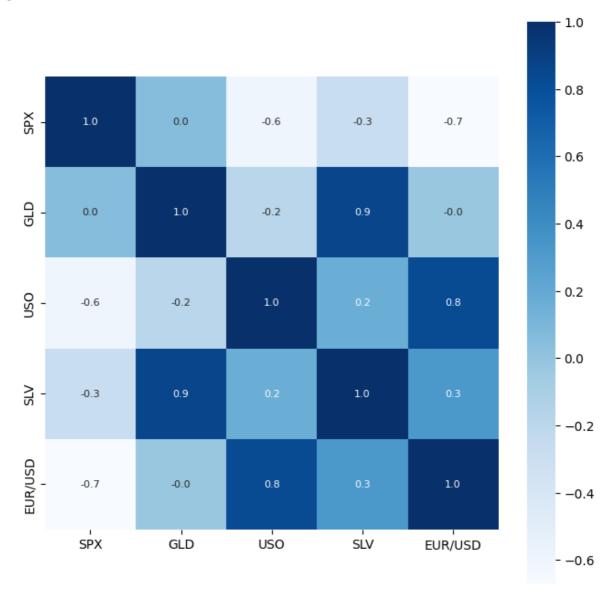
A- Positive correlation:

```
In [ ]: data_goldCorr = data_gold.iloc[:,1:]
    correlation = data_goldCorr.corr()
```

B- Contruction of the heatmap

```
In [ ]: plt.figure(figsize= (8,8) )
    sns.heatmap(correlation, cbar=True, square=True, fmt = ".1f", annot=True, annot_
```





B- Correlation value of GLD:

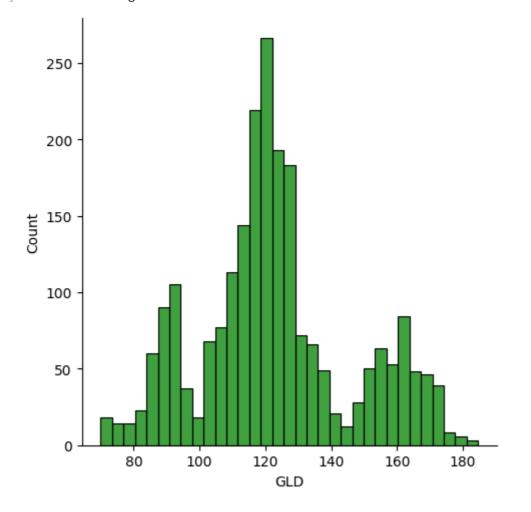
```
In []: print(correlation['GLD'])

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

C- Cheking the distribution of the GOLDQ PRICE:

```
In [ ]: sns.displot(data_gold['GLD'], color='green')
```

Out[]: <seaborn.axisgrid.FacetGrid at 0x2226208cd40>



5. Train test split:

A- Separating a data & label

```
In [ ]: X = data_gold.drop(['Date', 'GLD'], axis=1)
Y = data_gold["GLD"]
print(X)
print(Y)
```

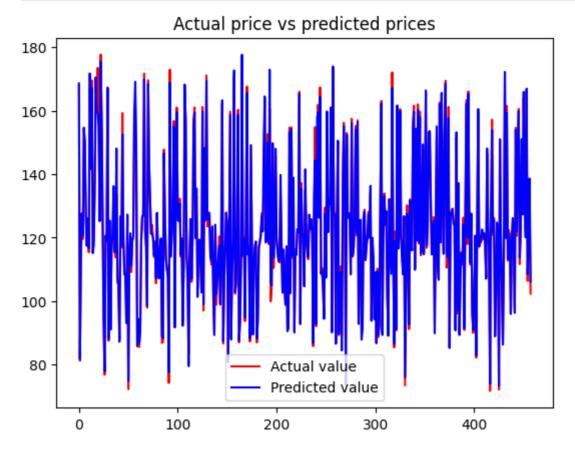
```
SPX
                               US0
                                        SLV
                                             EUR/USD
            1447.160034 78.470001 15.1800 1.471692
       0
       1
            1447.160034 78.370003 15.2850 1.474491
       2
            1411.630005 77.309998 15.1670 1.475492
       3
            1416.180054 75.500000 15.0530 1.468299
            1390.189941 76.059998 15.5900 1.557099
       4
                                        . . .
       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 x 4 columns]
               84.860001
       1
               85.570000
       2
               85.129997
       3
               84.769997
               86.779999
       2285
              124.589996
       2286 124.330002
       2287
              125.180000
       2288
              124.489998
       2289
              122.543800
       Name: GLD, Length: 2290, dtype: float64
        B- Test Split
In [ ]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.2, rando
In [ ]: print(X.shape, X_train.shape, X_test.shape)
       (2290, 4) (1832, 4) (458, 4)
          6. Training the model (Randon Forest Regression):
        A- Loading the model
       model = RandomForestRegressor()
In [ ]:
        B- Training the model:
In [ ]: model.fit(X train, Y train)
Out[]: ▼ RandomForestRegressor
        RandomForestRegressor()
          7. Model Evaluation:
        A- Prediction of test data:
In [ ]: test_train_prediction = model.predict(X_test)
        error_score = metrics.r2_score(Y_test, test_train_prediction)
```

```
print("R squared Error : ", error_score)
```

R squared Error: 0.9894278870106328

B- Compare the actual valuead and Predicted value:

```
In [ ]: Y_test = list(Y_test)
   plt.plot(Y_test, color = 'red', label ='Actual value')
   plt.plot(test_train_prediction, color = 'blue', label ='Predicted value')
   plt.title("Actual price vs predicted prices")
   plt.legend()
   plt.show()
```



8- Example:

```
In []:
    def    predictionF(SPX,USO,SLV,EUR_USD):
        input_data = (SPX,USO,SLV,EUR_USD)
        #Input the data into the numpy array:
        input_dataNumpuy = np.asarray(input_data)
        #Reshape the data:
        input_dataReshaped = input_dataNumpuy.reshape(1,-1)
        #Predict the model:
        prediction = model.predict(input_dataReshaped)
        print("The predcition price of the gold is:", prediction[0])
        predictionF(1400.160034,90.470001,13.18,1.471692)
```

The predcition price of the gold is: 82.44929941000002

c:\Users\HP\AppData\Local\Programs\Python\Python312\Lib\site-packages\sklearn\bas
e.py:465: UserWarning: X does not have valid feature names, but RandomForestRegre
ssor was fitted with feature names
warnings.warn(