```
import pandas as pd
In [1]:
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.model_selection import train_test_split
        from sklearn.ensemble import RandomForestRegressor
        from sklearn import metrics
        df = pd.read_csv("gld_price_data.csv")
In [2]:
In [5]: df
Out[5]:
                   Date
                                SPX
                                           GLD
                                                     USO
                                                              SLV EUR/USD
                                      84.860001 78.470001 15.1800
                1/2/2008 1447.160034
                                                                    1.471692
                1/3/2008 1447.160034
                                      85.570000 78.370003 15.2850
                                                                    1.474491
            2
               1/4/2008 1411.630005
                                      85.129997 77.309998
                                                                    1.475492
                                                          15.1670
               1/7/2008 1416.180054
                                      84.769997 75.500000 15.0530
                                                                    1.468299
            3
                1/8/2008 1390.189941
                                      86.779999 76.059998 15.5900
                                                                    1.557099
               5/8/2018 2671.919922 124.589996 14.060000 15.5100
         2285
                                                                    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
```

localhost:8890/doc/workspaces/auto-j/tree/gold price prediction random forest regression.ipynb?

df.tail()

In [7]:

```
Out[7]:
                   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
 In [9]:
         df.shape
 Out[9]: (2290, 6)
In [11]:
         df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 2290 entries, 0 to 2289
        Data columns (total 6 columns):
            Column
                     Non-Null Count Dtype
                     -----
                     2290 non-null
                                     object
            Date
            SPX
                     2290 non-null float64
         1
                     2290 non-null float64
            GLD
                     2290 non-null float64
         3
            US0
         4
                     2290 non-null float64
            SLV
             EUR/USD 2290 non-null float64
        dtypes: float64(5), object(1)
        memory usage: 107.5+ KB
In [13]: df.isnull().sum()
Out[13]:
         Date
                    0
         SPX
                    0
         GLD
                    0
         USO
                    0
         SLV
         EUR/USD
         dtype: int64
```

Out[15]: **SPX GLD** USO **SLV EUR/USD count** 2290.000000 2290.000000 2290.000000 2290.000000 2290.000000 **mean** 1654.315776 1.283653 31.842221 20.084997 122.732875 519.111540 23.283346 19.523517 7.092566 std 0.131547 676.530029 70.000000 7.960000 8.850000 1.039047 min **25%** 1239.874969 109.725000 14.380000 15.570000 1.171313 **50%** 1551.434998 120.580002 33.869999 17.268500 1.303297 **75%** 2073.010070 132.840004 37.827501 22.882500 1.369971 **max** 2872.870117 1.598798 184.589996 117.480003 47.259998

In [17]: ##co relation postive if one variable increase and the other one also increases otherwise it is negative corelation

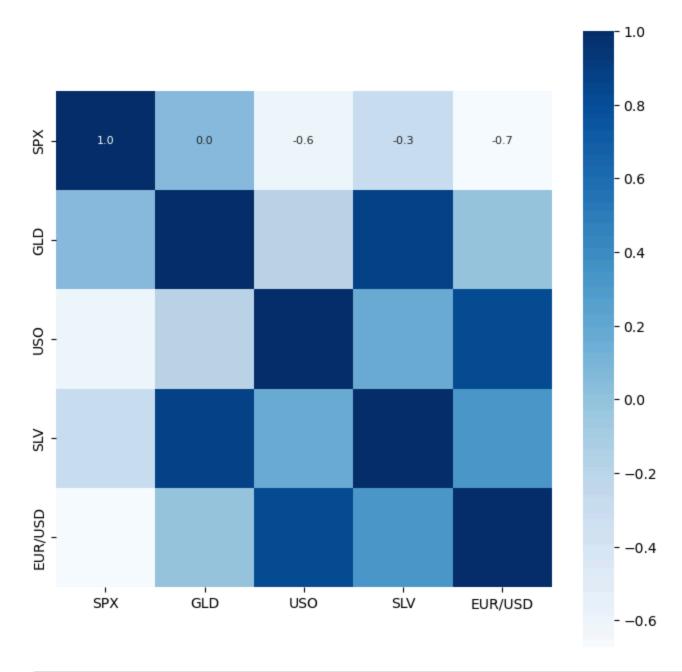
In [32]: z = df.drop(['Date'],axis = 1)

In [34]: **z**

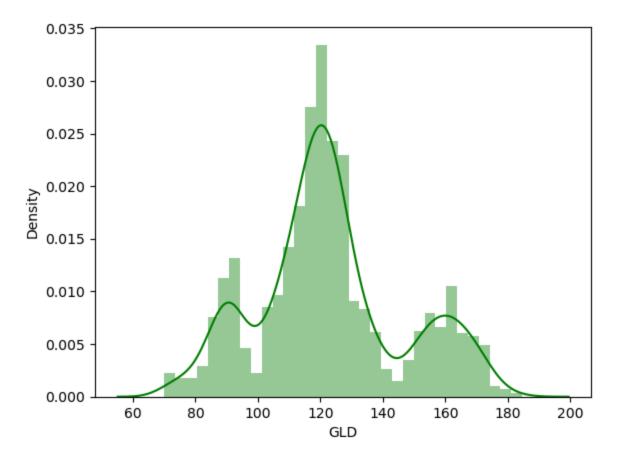
Out[34]:		SPX	GLD	USO	SLV	EUR/USD
	0	1447.160034	84.860001	78.470001	15.1800	1.471692
	1	1447.160034	85.570000	78.370003	15.2850	1.474491
	2	1411.630005	85.129997	77.309998	15.1670	1.475492
	3	1416.180054	84.769997	75.500000	15.0530	1.468299
	4	1390.189941	86.779999	76.059998	15.5900	1.557099
	•••					•••
	2285	2671.919922	124.589996	14.060000	15.5100	1.186789
	2286	2697.790039	124.330002	14.370000	15.5300	1.184722
	2287	2723.070068	125.180000	14.410000	15.7400	1.191753
	2288	2730.129883	124.489998	14.380000	15.5600	1.193118
	2289	2725.780029	122.543800	14.405800	15.4542	1.182033

2290 rows × 5 columns

```
In [44]: correlation = z.corr()
In [60]: ##create a heatmap
    plt.figure(figsize = (8,8))
        sns.heatmap(correlation,cbar = True , square = True ,fmt = '.1f',annot = True,annot_kws={'size':8},cmap ='Blues')
Out[60]: <Axes: >
```



```
Out[62]: SPX
                     0.049345
         GLD
                    1.000000
         US0
                    -0.186360
          SLV
                     0.866632
          EUR/USD
                    -0.024375
         Name: GLD, dtype: float64
         sns.distplot(df['GLD'],color='green')
In [70]:
        C:\Users\parth\AppData\Local\Temp\ipykernel_23648\3736964373.py:1: UserWarning:
        `distplot` is a deprecated function and will be removed in seaborn v0.14.0.
        Please adapt your code to use either `displot` (a figure-level function with
        similar flexibility) or `histplot` (an axes-level function for histograms).
        For a guide to updating your code to use the new functions, please see
        https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
          sns.distplot(df['GLD'],color='green')
        C:\Users\parth\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecate
        d and will be removed in a future version. Convert inf values to NaN before operating instead.
          with pd.option_context('mode.use_inf_as_na', True):
Out[70]: <Axes: xlabel='GLD', ylabel='Density'>
```



In [72]: **z**

Out[72]:		SPX	GLD	USO	SLV	EUR/USD
	0	1447.160034	84.860001	78.470001	15.1800	1.471692
	1	1447.160034	85.570000	78.370003	15.2850	1.474491
	2	1411.630005	85.129997	77.309998	15.1670	1.475492
	3	1416.180054	84.769997	75.500000	15.0530	1.468299
	4	1390.189941	86.779999	76.059998	15.5900	1.557099
	•••					
	2285	2671.919922	124.589996	14.060000	15.5100	1.186789
	2286	2697.790039	124.330002	14.370000	15.5300	1.184722
	2287	2723.070068	125.180000	14.410000	15.7400	1.191753
	2288	2730.129883	124.489998	14.380000	15.5600	1.193118
	2289	2725.780029	122.543800	14.405800	15.4542	1.182033

2290 rows × 5 columns

```
In [74]: y = df['GLD']
In [76]: y
Out[76]: 0
                   84.860001
                   85.570000
          1
          2
                   85.129997
          3
                   84.769997
          4
                   86.779999
                     . . .
          2285
                  124.589996
          2286
                  124.330002
          2287
                  125.180000
                  124.489998
          2288
          2289
                  122.543800
          Name: GLD, Length: 2290, dtype: float64
```

```
In [78]: x = df.drop(['GLD', 'Date'], axis = 1)
In [80]: x
Out[80]:
                      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
         ##splitting into training and test data
In [82]:
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2 , random_state=2)
         a = RandomForestRegressor()
In [84]:
         a.fit(x_train,y_train)
Out[84]:
          ▼ RandomForestRegressor
         RandomForestRegressor()
```

```
In [86]: y_pred_test = a.predict(x_test)
          error_test = metrics.r2_score(y_test,y_pred_test)
In [88]:
In [90]:
          error_test
          0.9892769343291239
Out[90]:
          #visualize
In [100...
          plt.scatter(y_test,y_pred_test,color='green')
          plt.xlabel('Actual Value')
          plt.ylabel('predicted value')
          plt.show()
            180
            160
         predicted value
            140
            120
            100
             80
                                                 120
                         80
                                    100
                                                             140
                                                                         160
                                                                                      180
                                               Actual Value
```

In []: