In [10]:
```
import pandas as pd
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

import warnings
warnings.filterwarnings("ignore")
```

In [2]:
```
df = pd.read_csv('gold_prices_data.csv')
```

In [3]:
```
df.head()
```

Out[3]:

In [4]:
```
df.tail()
```

Out[4]:

						1.182033

In [5]:
```
df.shape
```

Out[5]:
```
(2290, 6)
```

In [6]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2290 entries, 0 to 2289
Data columns (total 6 columns):
 #   Column   Non-Null Count  Dtype  
---  ------   --------------  -----  
 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 [7]:
```
df.isnull().sum()
```

Out[7]:
```
Date       0
SPX        0
GLD        0
USO        0
SLV        0
EUR/USD    0
dtype: int64
```

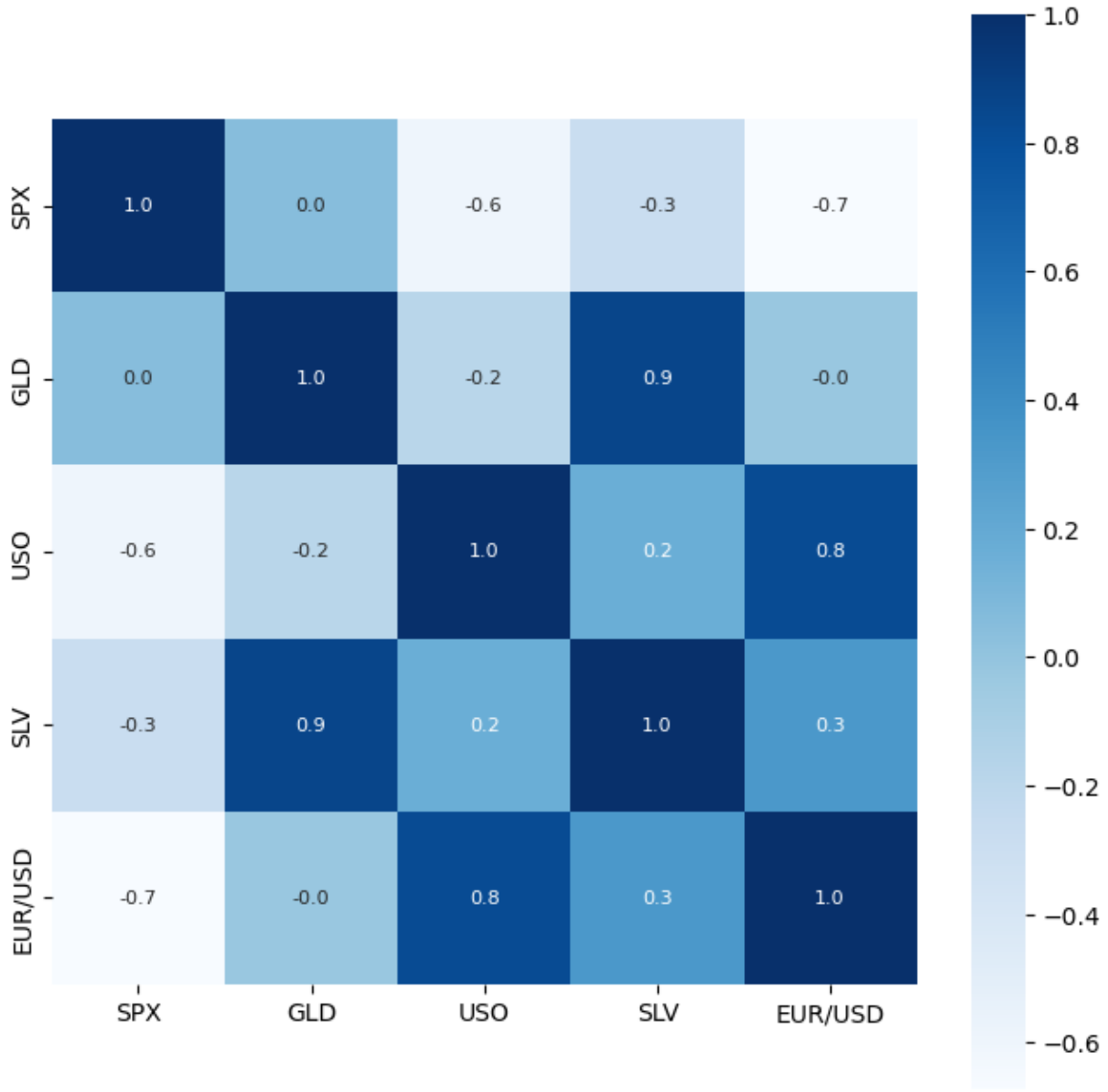
In [8]:
```
df.describe()
```

Out[8]:

Data Analysis

In [11]:
```
# Correlation
correlation = df.corr()
plt.figure(figsize = (8,8))
sns.heatmap(correlation,cbar=True,square=True,fmt='.1f',annot=True, annot_kws={'size'
```

Out[11]:
```
<Axes: >
```

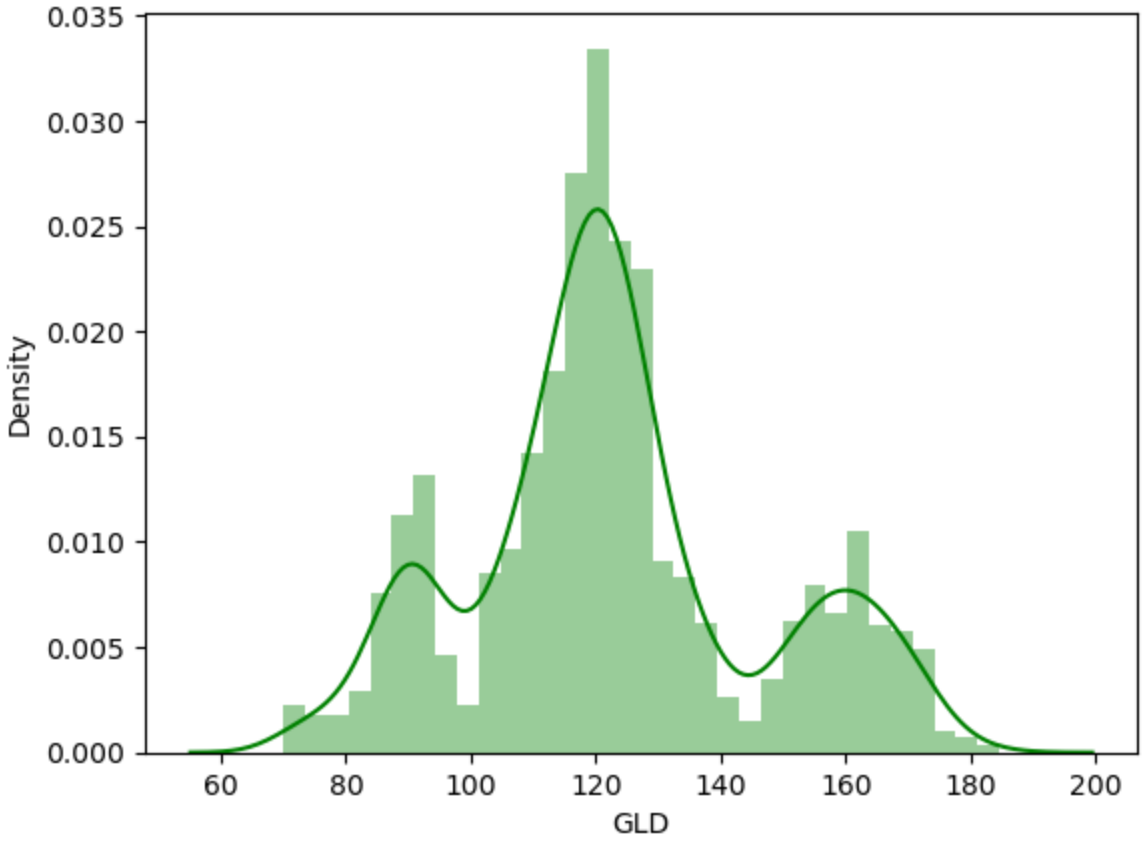


In [12]:
```
# correlation values of Gld
print(correlation['GLD'])
```

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

In [13]:
```
# distribution of the GLD Price
sns.distplot(df['GLD'],color='green')
```

Out[13]:
```
<Axes: xlabel='GLD', ylabel='Density'>
```



Applying Model

In [14]:
```
X = df.drop(['Date','GLD'],axis=1)
y = df['GLD']
print(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 x 4 columns]
```

In [15]:
```
print(y)
```

```
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    124.489998
2289    122.543800
Name: GLD, Length: 2290, dtype: float64
```

In [16]:
```
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=42)
```

In [17]:
```
regressor =RandomForestRegressor(n_estimators=100)
regressor.fit(X_train,y_train)
```

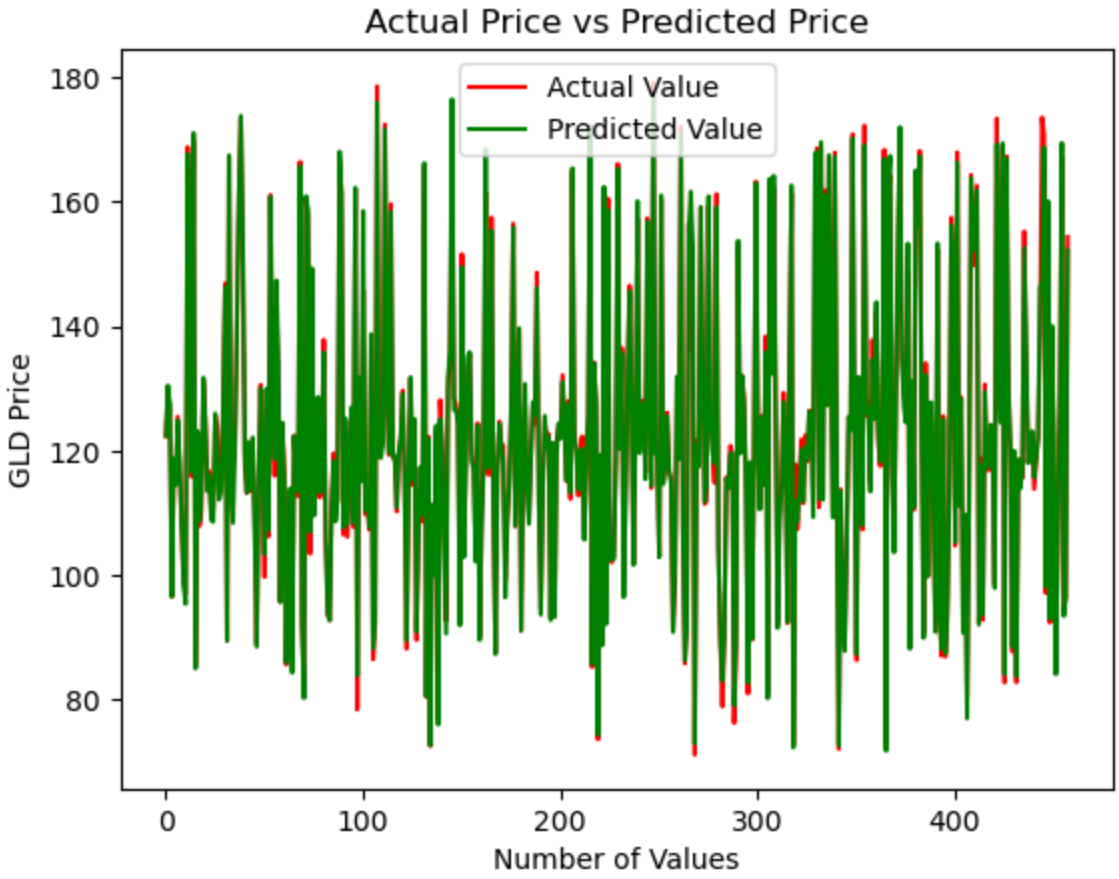
Out[17]:
```
▾ RandomForestRegressor

RandomForestRegressor()
```

In [18]:
```
test_prediction = regressor.predict(X_test)
```

In [19]:
```
error_score = metrics.r2_score(y_test,test_prediction)
print("R squared error : ", error_score)
```

```
R squared error :  0.9886669789501786
```

In [23]:
```
# Comparing Actual and predicted values
y_test = list(y_test)
plt.plot(y_test,color='red',label='Actual Value')
plt.plot(test_prediction, color='green', label='Predicted Value')
plt.title('Actual Price vs Predicted Price')
plt.xlabel('Number of Values')
plt.ylabel('GLD Price')
plt.legend()
plt.show()
```


Actual Price vs Predicted Price
Actual Value
Predicted Value
GLD Price
Number of Values
180
160
140
120
100
80
0
100
200
300
400

In [ ]: