

# Simple Linear Regression Project (Gold Price Prediction)

Of all the precious metals, gold is the most popular as an investment. Investors generally buy gold as a way of diversifying risk, especially through the use of futures contracts and derivatives. The gold market is subject to speculation and volatility as are other markets. Compared to other precious metals used for investment, gold has been the most effective safe haven across a number of countries.

The Dataset contain gold prices (in USD) from 2001 to 2019. Our goal is to predict where the gold prices will be in the coming years

## Import the necessary libraries

```
In [92]: import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns  
import numpy as np
```

## Read 'gold\_price\_usd.csv' & store it in a variable

```
In [93]: df=pd.read_csv('gold_price_usd.csv')  
df
```

Out[93]:

	Date	USD (AM)
0	2001-01-02	272.80
1	2001-01-03	269.00
2	2001-01-04	268.75
3	2001-01-05	268.00
4	2001-01-08	268.60
...	...	...
4713	2019-08-27	1531.85
4714	2019-08-28	1541.75
4715	2019-08-29	1536.65
4716	2019-08-30	1526.55
4717	2019-09-02	1523.35

4718 rows × 2 columns

## View the first 5 rows

```
In [94]: df.head()
```

```
Out[94]:
```

	Date	USD (AM)
0	2001-01-02	272.80
1	2001-01-03	269.00
2	2001-01-04	268.75
3	2001-01-05	268.00
4	2001-01-08	268.60

## Check the information

```
In [95]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4718 entries, 0 to 4717
Data columns (total 2 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Date        4718 non-null    object 
 1   USD (AM)    4718 non-null    float64
dtypes: float64(1), object(1)
memory usage: 73.8+ KB
```

## Find the columns

```
In [96]: df.columns
```

```
Out[96]: Index(['Date', 'USD (AM)'], dtype='object')
```

## Rename USD (AM) to Price

```
In [97]: df.rename(columns={'USD (AM)':'price'}, inplace='true')
df
```

Out[97]:

	Date	price
0	2001-01-02	272.80
1	2001-01-03	269.00
2	2001-01-04	268.75
3	2001-01-05	268.00
4	2001-01-08	268.60
...	...	...
4713	2019-08-27	1531.85
4714	2019-08-28	1541.75
4715	2019-08-29	1536.65
4716	2019-08-30	1526.55
4717	2019-09-02	1523.35

4718 rows × 2 columns

### Check if there are any missing values in the dataset

In [98]: `df.isna().sum()`

Out[98]:

### Gather the basic statistical information about the dataset

In [99]: `df.describe()`

Out[99]:

	price
count	4718.000000
mean	959.990812
std	449.456217
min	256.700000
25%	449.112500
50%	1113.125000
75%	1293.750000
max	1896.500000

### Convert Date column from object to datetime format

In [100...]: `df["year"] = pd.DatetimeIndex(df["Date"]).year  
df["month"] = pd.DatetimeIndex(df["Date"]).month`

df

Out[100]:

	Date	price	year	month
0	2001-01-02	272.80	2001	1
1	2001-01-03	269.00	2001	1
2	2001-01-04	268.75	2001	1
3	2001-01-05	268.00	2001	1
4	2001-01-08	268.60	2001	1
...	...	...	...	...
4713	2019-08-27	1531.85	2019	8
4714	2019-08-28	1541.75	2019	8
4715	2019-08-29	1536.65	2019	8
4716	2019-08-30	1526.55	2019	8
4717	2019-09-02	1523.35	2019	9

4718 rows × 4 columns

In [101...]

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4718 entries, 0 to 4717
Data columns (total 4 columns):
 #   Column   Non-Null Count  Dtype  
---  --  
 0   Date      4718 non-null   object 
 1   price     4718 non-null   float64
 2   year      4718 non-null   int64  
 3   month     4718 non-null   int64  
dtypes: float64(1), int64(2), object(1)
memory usage: 147.6+ KB
```

## Create a new column with Year

In [102...]

```
df["year"] = pd.DatetimeIndex(df["Date"]).year
df
```

Out[102]:

	Date	price	year	month
0	2001-01-02	272.80	2001	1
1	2001-01-03	269.00	2001	1
2	2001-01-04	268.75	2001	1
3	2001-01-05	268.00	2001	1
4	2001-01-08	268.60	2001	1
...	...	...	...	...
4713	2019-08-27	1531.85	2019	8
4714	2019-08-28	1541.75	2019	8
4715	2019-08-29	1536.65	2019	8
4716	2019-08-30	1526.55	2019	8
4717	2019-09-02	1523.35	2019	9

4718 rows × 4 columns

## Create a new column with Months

In [103...]

```
df["month"] = pd.DatetimeIndex(df["Date"]).month
```

In [104...]

```
df
```

Out[104]:

	Date	price	year	month
0	2001-01-02	272.80	2001	1
1	2001-01-03	269.00	2001	1
2	2001-01-04	268.75	2001	1
3	2001-01-05	268.00	2001	1
4	2001-01-08	268.60	2001	1
...	...	...	...	...
4713	2019-08-27	1531.85	2019	8
4714	2019-08-28	1541.75	2019	8
4715	2019-08-29	1536.65	2019	8
4716	2019-08-30	1526.55	2019	8
4717	2019-09-02	1523.35	2019	9

4718 rows × 4 columns

## See all the years and Months in our dataset

In [105...]

```
#Years  
df[["year"]]
```

Out[105]:

	year
0	2001
1	2001
2	2001
3	2001
4	2001
...	...
4713	2019
4714	2019
4715	2019
4716	2019
4717	2019

4718 rows × 1 columns

In [106...]

```
#Months  
df[["month"]]
```

Out[106]:

	month
0	1
1	1
2	1
3	1
4	1
...	...
4713	8
4714	8
4715	8
4716	8
4717	9

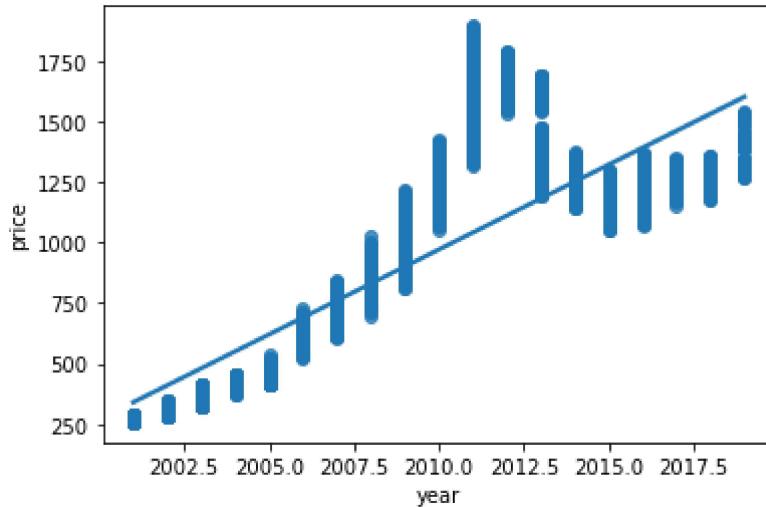
4718 rows × 1 columns

## Visualization

## Create a regression plot with x-axis as years and y-axis as Price

```
In [107... import matplotlib.pyplot as plt
import seaborn as sns
sns.regplot(x=df['year'],y=df['price'])

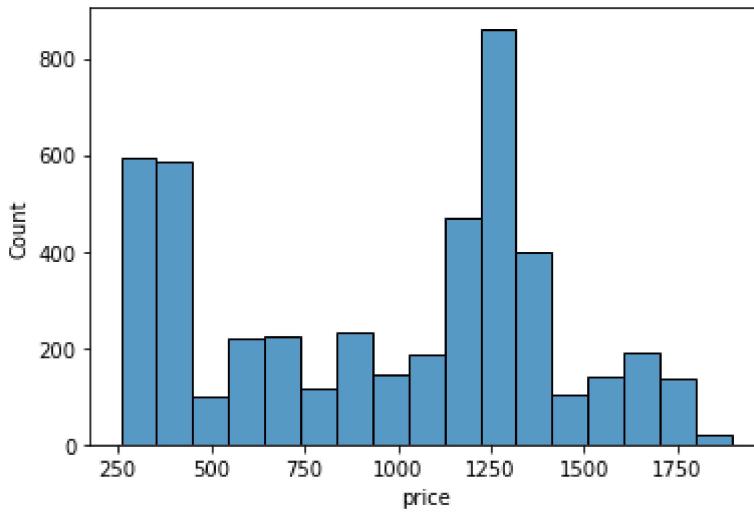
Out[107]: <AxesSubplot:xlabel='year', ylabel='price'>
```



## Plot a histplot to find the variation in price

```
In [108... sns.histplot(x=df['price'])

Out[108]: <AxesSubplot:xlabel='price', ylabel='Count'>
```



## Assign year and price in x and y variables

```
In [109... x = df[['year']]
y = df['price']
```

## Split the data into training and testing set

We will train our model on the training set and then use the test set to evaluate the model

```
In [110]: # import train_test_split  
from sklearn.model_selection import train_test_split  
  
In [111]: x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.3, random_state=42)
```

## Train Data

```
In [112]: # import LinearRegression from sklearn  
from sklearn.linear_model import LinearRegression
```

## Create Linear Regression Model

```
In [113]: model=LinearRegression()
```

## Train the model

```
In [114]: model.fit(x_train,y_train)  
Out[114]: LinearRegression()
```

## Check the score of our model

```
In [115]: model.score(x_train,y_train)  
Out[115]: 0.7048691960223057
```

## Check the coefficient and Intercept

```
In [116]: #print the intercept  
print(model.intercept_)  
-140074.32374779382  
  
In [117]: #print the coefficient  
print(model.coef_)  
[70.17366927]
```

## Make Prediction with Test data

```
In [118]: # Also store the predicted values in a variable  
y_pred=model.predict(x_test)  
len(y_pred)  
  
Out[118]: 1416
```

```
In [119...]: c=[i for i in range(1,1417)]  
c
```

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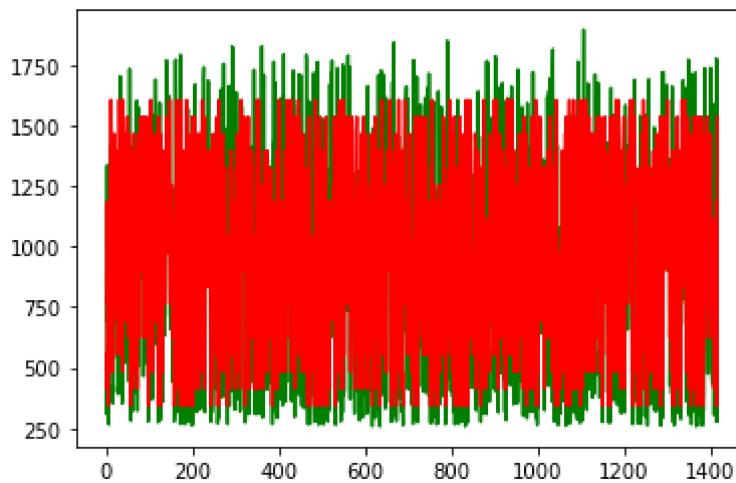
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990,  
991,  
992,  
993,  
994,  
995,  
996,  
997,  
998,  
999,  
1000,  
...]
```

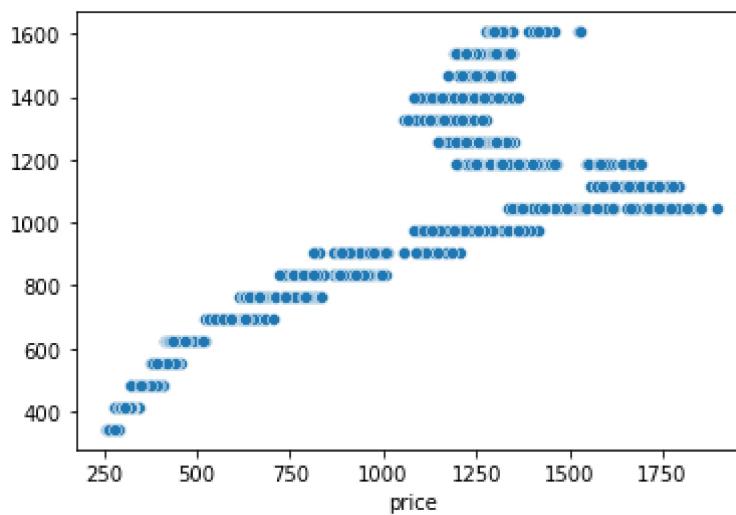
```
In [120]: plt.plot(c,y_test,color='green')  
plt.plot(c,y_pred,color='red')
```

```
Out[120]: [<matplotlib.lines.Line2D at 0x10b204f0670>]
```



In [121]: `sns.scatterplot(x=y_test,y=y_pred)`

Out[121]:



**Create a new dataframe with actual and predicted values with year(X\_test) as index**

In [ ]:

**Check the mean absolute error, mean square error**

In [122]: `from sklearn.metrics import mean_absolute_error, mean_squared_error`

In [123]: `# Mean absolute error  
mean_absolute_error(y_test,y_pred)`

Out[123]: 186.2427389387351

In [124]: `# Mean squared error  
mean_squared_error(y_test,y_pred)`

Out[124]: 58032.97376893088

# predict the prices for following years

.2025,2026,2027,2028,2030

```
In [125]:  
data=[2025,2026,2027,2028,2030]  
df1=pd.DataFrame(data,columns=['year'])  
df1
```

Out[125]:

	year
0	2025
1	2026
2	2027
3	2028
4	2030

# Great Job!

In [ ]:

In [ ]: