

Simple Linear Regression Model (CalBank Case Study)

August 31, 2020

1 Question: How Liquid is the CalBank shares on Ghana Stock Exchange ?

2 Data Source : Ghana Stock Exchange - CALBank (2007 - 2019)

3 True Model: Closing Stock Price and Opening Stock Price

3.1 Suppose that Closing Stock Price is determined by the Model

4 Closing Stock Price = $B_0 + B_1$ Opening Stock Price + U

4.1 y = Closing Stock price and X = Opening Stock Price

4.1.1 y is also known as Dependent variable, Explained variable, Response variable, Predicted variable, Regressand and Target

4.2 X is also known as independent variable, Explanatory variable, Control variable, Predictor variable Regressor and Features

4.3 U is also known as error term or disturbance term: Representing factors other than X that affect y

4.4 We are interested in the effect of Opening Stock Price on Closing Stock Price holding other factors fixed or constant

4.5 This effect is given by B_1

4.6 The error term U contain factors such as Dividend yield, Interest rate, Firm growth, CEO performance and so on

5 Gather Data:

5.1 Cleaning the Data

5.2 Data on Features is Opening Stock Price in GHS

5.3 Data on Targets is Closing Stock Price in GHS

6 Exploring and Visualization:

```
[28]: import pandas as pd
```

```
[29]: data = pd.read_csv("CalBank.csv")
```

```
[30]: print(data)
```

	Date	Share_Code	Year_High_GHS	Year_Low_GHS	\
0	25/06/2007	CAL	0.2700	0.20	
1	26/06/2007	CAL	0.2702	0.20	
2	27/06/2007	CAL	0.2705	0.20	
3	28/06/2007	CAL	0.2715	0.20	

4	29/06/2007	CAL	0.2720	0.20
...
3078	23/12/2019	CAL	1.0800	0.64
3079	24/12/2019	CAL	1.0800	0.64
3080	27/12/2019	CAL	1.0800	0.64
3081	30/12/2019	CAL	1.0800	0.64
3082	31/12/2019	CAL	1.0800	0.64

	Previous_Closing_Price_VWAP_GHS	Opening_Price_GHS	Closing_Price_GHS	\
0	0.2646	0.2646	0.2700	
1	0.2700	0.2700	0.2702	
2	0.2702	0.2702	0.2705	
3	0.2705	0.2705	0.2715	
4	0.2715	0.2715	0.2720	
...	
3078	0.8100	0.8100	0.8100	
3079	0.8100	0.8100	0.8100	
3080	0.8100	0.8100	0.8200	
3081	0.8200	0.8200	0.8200	
3082	0.8200	0.8200	0.8900	

	Price_Change_GHS	Closing_Bid_Price_GHS	Closing_Offer_Price_GHS	\
0	0.0054	0.2700	0.00	
1	0.0002	0.2700	0.00	
2	0.0003	0.2702	0.00	
3	0.0010	0.2715	0.00	
4	0.0005	0.2715	0.00	
...	
3078	0.0000	0.8100	NaN	
3079	0.0000	0.8100	NaN	
3080	0.0100	0.8100	0.82	
3081	0.0000	0.8200	NaN	
3082	0.0700	NaN	0.82	

	Total_Shares_Traded	Total_Value_Traded	Last_Transaction_Price_GHS
0	2142400	NaN	0
1	6500	NaN	0
2	6000	NaN	0
3	741200	NaN	0
4	3000	NaN	0
...
3078	0	0	0.82
3079	1,000	810	0.82
3080	50,000	41,000.00	0.82
3081	0	0	0.82
3082	216,200	193,159.00	0.82

[3083 rows x 13 columns]

```
[31]: data.describe()
```

```
[31]:
```

	Year_High_GHS	Year_Low_GHS	Previous_Closing_Price_VWAP_GHS	\
count	3083.000000	3083.000000	3083.000000	
mean	0.810767	0.555953	0.689800	
std	0.401668	0.310753	0.348338	
min	0.200000	0.170000	0.170000	
25%	0.390000	0.260000	0.310000	
50%	0.800000	0.442000	0.760000	
75%	1.060000	0.850000	0.980000	
max	1.970000	1.400000	1.970000	

	Opening_Price_GHS	Closing_Price_GHS	Price_Change_GHS	\
count	3083.000000	3083.000000	3083.000000	
mean	0.689631	0.690109	0.001272	
std	0.348286	0.348606	0.029425	
min	0.170000	0.170000	-0.200000	
25%	0.310000	0.310000	0.000000	
50%	0.760000	0.760000	0.000000	
75%	0.980000	0.980000	0.000000	
max	1.970000	1.970000	0.690000	

	Closing_Bid_Price_GHS	Closing_Offer_Price_GHS
count	2389.000000	2571.000000
mean	0.626448	1.596433
std	0.385603	38.729524
min	0.000000	0.000000
25%	0.273500	0.290000
50%	0.750000	0.740000
75%	0.950000	0.980000
max	3.290000	1900.000000

```
[32]: type(data)
```

```
[32]: pandas.core.frame.DataFrame
```

```
[33]: stockPrice = data[["Opening_Price_GHS", "Closing_Price_GHS"]]
```

```
[34]: type(stockPrice)
```

```
[34]: pandas.core.frame.DataFrame
```

```
[35]: print(stockPrice)
```

	Opening_Price_GHS	Closing_Price_GHS
0	0.2646	0.2700
1	0.2700	0.2702

2	0.2702	0.2705
3	0.2705	0.2715
4	0.2715	0.2720
...
3078	0.8100	0.8100
3079	0.8100	0.8100
3080	0.8100	0.8200
3081	0.8200	0.8200
3082	0.8200	0.8900

[3083 rows x 2 columns]

```
[36]: stockPrice.describe()
```

```
[36]:
```

	Opening_Price_GHS	Closing_Price_GHS
count	3083.000000	3083.000000
mean	0.689631	0.690109
std	0.348286	0.348606
min	0.170000	0.170000
25%	0.310000	0.310000
50%	0.760000	0.760000
75%	0.980000	0.980000
max	1.970000	1.970000

```
[37]: #y = data[["Closing_Price_GHS"]]
      #X = data[["Opening_Price_GHS"]]
```

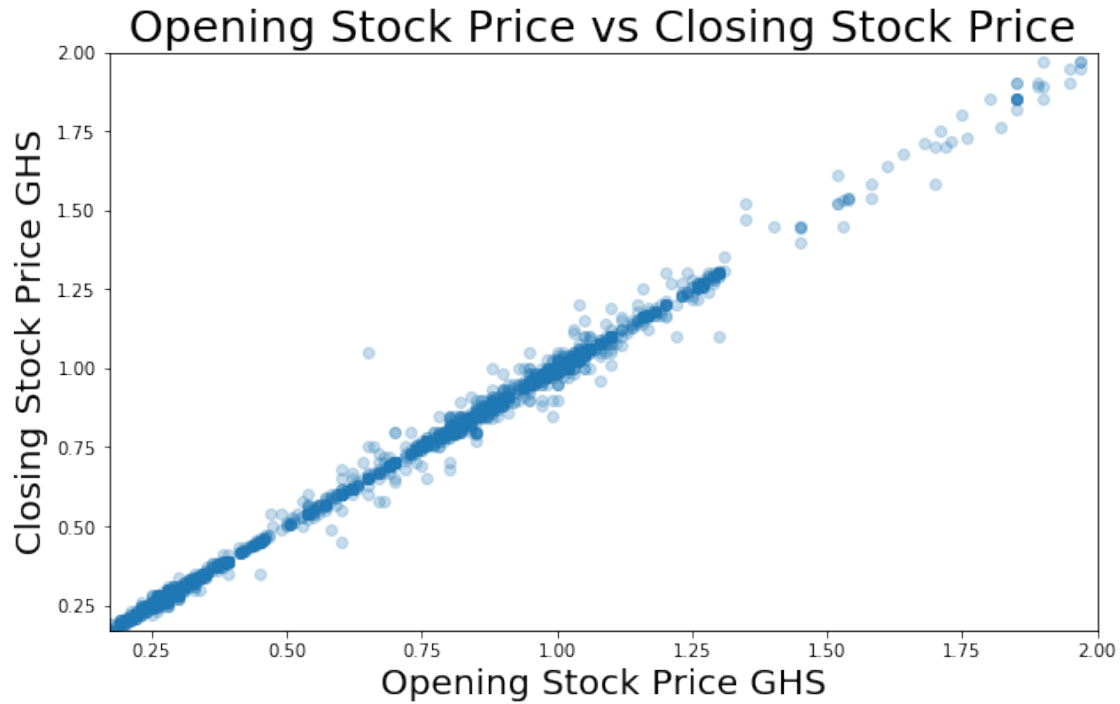
```
[38]: y = stockPrice[["Closing_Price_GHS"]]
      X = stockPrice[["Opening_Price_GHS"]]
```

```
[43]: import matplotlib.pyplot as plt
```

6.1 Scatter Plot

```
[66]: %matplotlib inline
      plt.figure(figsize=(10,6))
      plt.scatter(X,y, alpha=0.25)
      plt.title("Opening Stock Price vs Closing Stock Price", fontsize=25)
      plt.xlabel("Opening Stock Price GHS" , fontsize=20)
      plt.ylabel("Closing Stock Price GHS" , fontsize=20)
      plt.ylim(0.17,2)
      plt.xlim(0.17,2)
      plt.show
```

```
[66]: <function matplotlib.pyplot.show(*args, **kw)>
```



7 Hypothesize or Fitted Model:

7.1 we can draw whole lot of line through the data to estimate the parameter B_0 and B_1 But our job is to find the best possible line that minimize this residual and the method we use is OLS - Ordinary Least Square

8 Run and Evaluate Rgression

```
[67]: from sklearn.linear_model import LinearRegression
```

```
[68]: reg = LinearRegression()
```

```
[69]: reg.fit(X,y)
```

```
[69]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)
```

8.1 slope coefficient: B_1

```
[71]: reg.coef_
```

```
[71]: array([[0.99950251]])
```

8.2 Intercept: B_0

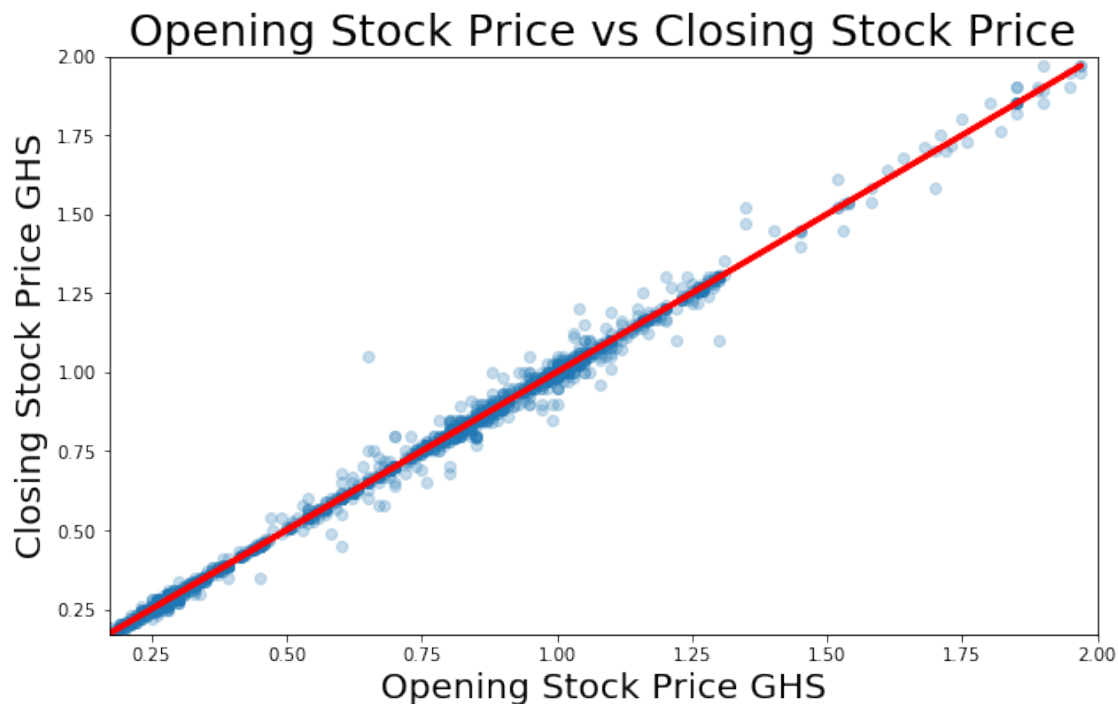
```
[72]: reg.intercept_
```

```
[72]: array([0.00082142])
```

9 Plotting

```
[74]: %matplotlib inline
plt.figure(figsize=(10,6))
plt.scatter(X,y, alpha=0.25)
plt.title("Opening Stock Price vs Closing Stock Price", fontsize=25)
plt.xlabel("Opening Stock Price GHS", fontsize=20)
plt.ylabel("Closing Stock Price GHS", fontsize=20)
plt.ylim(0.17,2)
plt.xlim(0.17,2)
plt.plot(X,reg.predict(X), color="red",linewidth=3)
plt.show
```

```
[74]: <function matplotlib.pyplot.show(*args, **kw)>
```



10 Estimated or Fitted Model

11 $y = 0.00 + 0.99X$

11.1 where y = Closing Stock Price , X = Opening Stock Price , $B_0 = 0.00$ and $B_1 = 0.99$

11.2 The intercept B_0 literally means an investor with zero investment in CalBank shares or equities will earn zero returns

11.3 The slope B_1 implies that an investor with 1GHS investment on CalBank shares will loss money after going short. Thus wont make profit on his investment

12 Goodness of fit of the model: R^2

```
[75]: reg.score(X,y)
```

```
[75]: 0.9971702603458394
```

12.1 $R^2 = 0.997$

12.2 The Opening Stock Price explains over 99% of the variation in the Closing Stock Price of CalBank

13 Conclusion

13.1 In conclusion liquidity is not permanant, CalBank shares is not actively traded on the Ghana Stock Exchange(GSE) and as a result investor would hold their stocks beacuse they not willing to traded at fall price. my advice to investor is to invest in money market intrument for short term investment

14 Prepared by PAUL OFFEI

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[ ]:
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