

Stock Price Prediction Using Machine Learning Algorithm.



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About Me

- I'm a data scientist with core interest in data analytics technology, Machine Learning and Artificial intelligence.
- I love building tech communities. I'm currently the Co-organizer & team lead for Pydata Lagos.
- I love Fried plantain - Dodo



What is a stock?

Stocks also referred to as shares or equity represents a fractional ownership in a company.

What is the stock market?

The stock market refers to public markets that exist for issuing, buying and selling stocks that trade on a stock exchange or over-the-counter.

Simply put, the stock market is a place where investors can sell or buy ownership of companies in the form of stocks or shares.

What's the fuss about the Stock Market?



Global Market Capitalization = \$80 Trillion.

Heading steadily towards \$100 Trillion

Nigeria Stock Exchange (NSE)



The Market Capitalization of NSE as at January 2018 stood at N14.18 Trillion.

Why Stock Price Prediction?



First, let me tell you a story...

Me after Investing in Stock



Grab some
popcorn
&
Watch my
money grow
like grass.

Me after seeing my investment yield in 3 months



Me seeing my
investment yield
after 1 year.



Here is what happened:

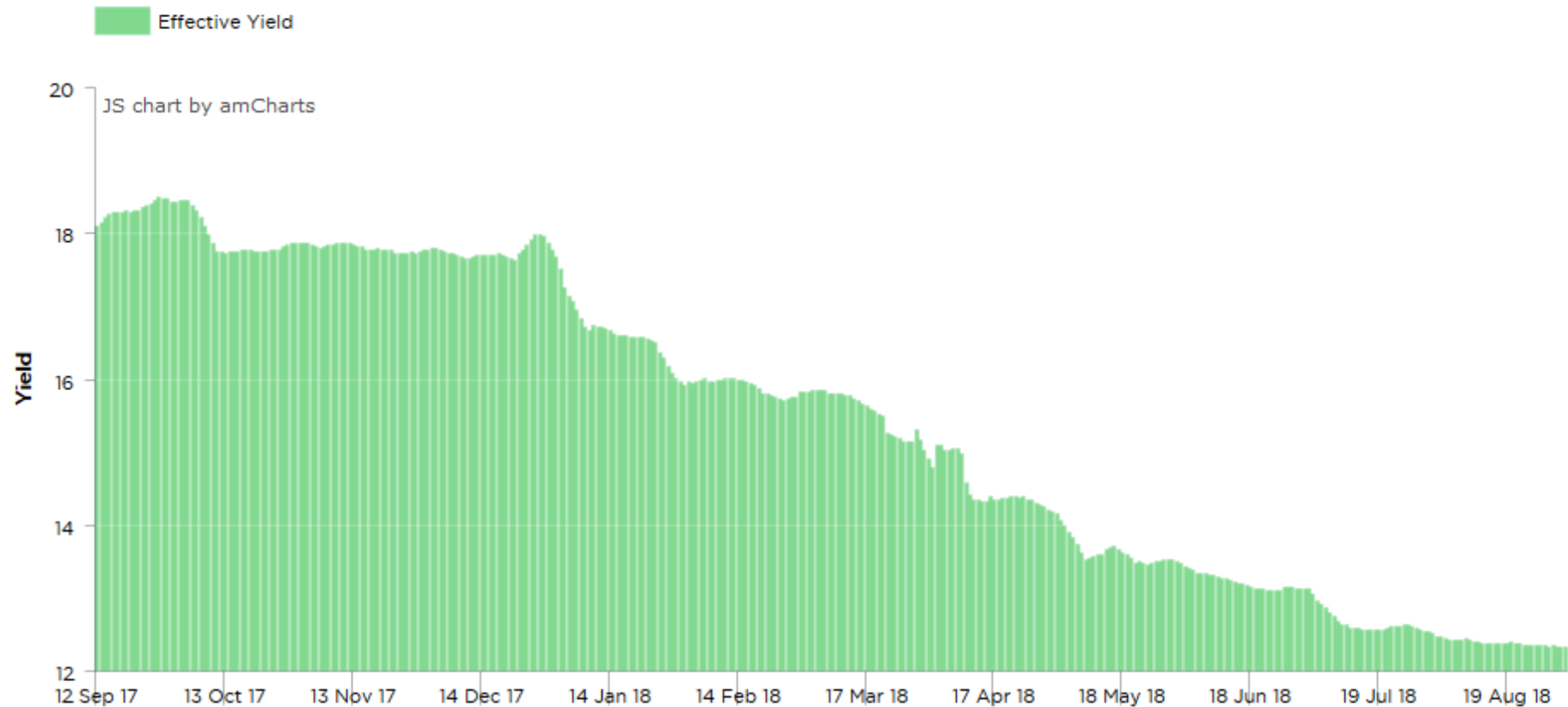
FUND PERFORMANCE

Money Market Fund

Discovery Fund

Aggressive Growth Fund

Ethical Fund



Think About this:



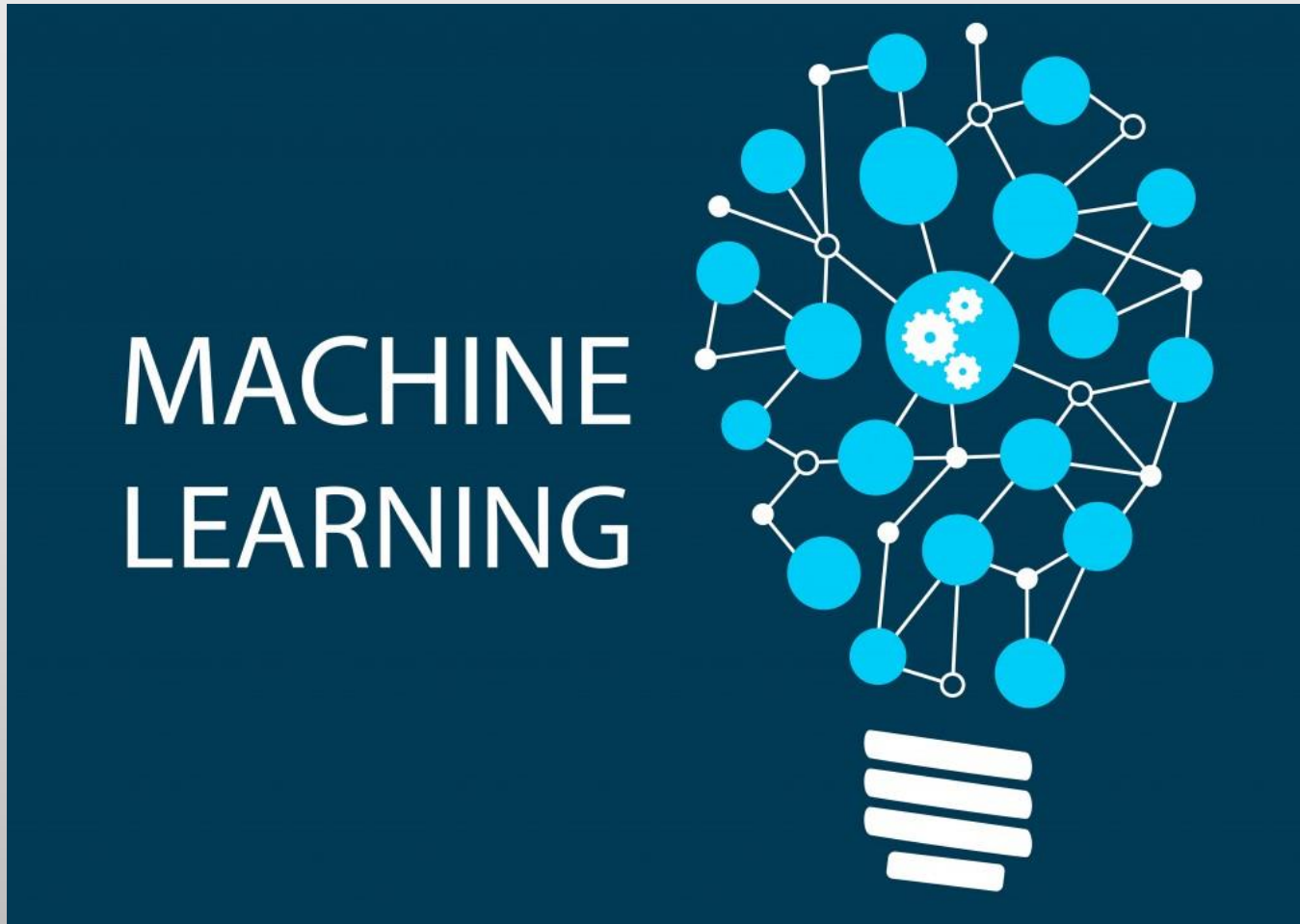
Imagine I understand all the
underlying factors that
influence the prices of stocks
so I can purchase stocks that
will yield better returns.

Imagine I can build an automated trading system that can understand these factors, accurately predict future prices of stocks and tell me which stock is safe/profitable to buy.

Your guess is as good
as mine:



Machine Learning to the rescue!



“Machine Learning at its most basic is the practice of using algorithms to parse data, learn from it, and then make a determination or prediction about something in the world.”

- Nvidia

Let's dive into implementing a
Machine Learning Algorithm
to predict stock prices...

The Goal:

Our goal is to build a Machine Learning Model to predict stock prices.

Outcome:

The outcome of our work will be predicted prices & a measure of the percentage of accuracy of our model in predicting the prices.



Importing Libraries

```
In [136]: 1 import pandas as pd
          2 import requests
          3 import numpy as np
          4 from bs4 import BeautifulSoup
```

Create an empty dataframe to store our data

```
In [137]: 1 data = pd.DataFrame()
          2 pd.options.mode.chained_assignment = None #Prevent error Warning of a copy in memory
```



Build a simple Web Scraper to import Stock Market data from Ngtradeonline.com

In [138]:

```
1 count = 0
2 while count < 20 :
3     res = requests.get("http://www.ngtradeonline.com/Home/PriceHistory?page="+str(count)+"&s
4     soup = BeautifulSoup(res.content, 'lxml') #Use BeautifulSoup to prettify our data
5     table = soup.find_all('table')[0] #Find all the html tag named 'table'
6     html_data = pd.read_html(str(table))[0] #Read the content of found html tables
7     data = data.append(html_data) # Append to the empty Dataframe above
8     print("page "+str(count)+" Done.") #Keep track of each web page read
9     count = count + 1
10 data.to_csv('GTB_data.csv') # Export the data to a CSV file
```

page 0 Done.

page 1 Done.

page 2 Done.

page 3 Done.

page 4 Done.

page 5 Done.

page 6 Done.

page 7 Done.

page 8 Done.

Let's look at the Data

In [172]:

```
1 data.head(20)
```

Out[172]:

	Symbol	Low	Open	Price	Volume	High	Change	Date
0	GUARANTY	31.95	34.50	32.50	0	33.35	-2.00	9/11/2018
1	GUARANTY	34.50	35.00	34.50	0	35.00	-0.50	9/10/2018
2	GUARANTY	34.40	34.95	35.00	0	35.30	0.05	9/7/2018
3	GUARANTY	34.95	36.65	34.95	0	36.50	-1.70	9/6/2018
4	GUARANTY	36.65	37.05	36.65	0	36.85	-0.40	9/5/2018
5	GUARANTY	36.50	36.50	37.05	0	37.45	0.55	9/4/2018
6	GUARANTY	36.05	36.00	36.50	0	36.75	0.50	9/3/2018
7	GUARANTY	36.00	38.00	36.00	0	37.85	-2.00	8/31/2018
8	GUARANTY	37.90	39.05	38.00	0	38.00	-1.05	8/30/2018
9	GUARANTY	38.65	39.00	39.05	0	39.05	0.05	8/29/2018
0	GUARANTY	31.95	34.50	32.50	0	33.35	-2.00	9/11/2018
1	GUARANTY	34.50	35.00	34.50	0	35.00	-0.50	9/10/2018
2	GUARANTY	34.40	34.95	35.00	0	35.30	0.05	9/7/2018
3	GUARANTY	34.95	36.65	34.95	0	36.50	-1.70	9/6/2018
4	GUARANTY	36.65	37.05	36.65	0	36.85	-0.40	9/5/2018
5	GUARANTY	36.50	36.50	37.05	0	37.45	0.55	9/4/2018
6	GUARANTY	36.05	36.00	36.50	0	36.75	0.50	9/3/2018
7	GUARANTY	36.00	38.00	36.00	0	37.85	-2.00	8/31/2018
8	GUARANTY	37.90	39.05	38.00	0	38.00	-1.05	8/30/2018
9	GUARANTY	38.65	39.00	39.05	0	39.05	0.05	8/29/2018

Data transformation

```
In [139]: 1 #Create a column to hold the percentage spread of the closing price
          2 data['HL_PCT'] = (data['High']- data['Low'])/data['Price'] *100.0
          3
          4 #Create a column to hold the Percentage of Price Change
          5 data['PCT_change'] = (data['Price'] - data['Open']) / data['Open'] * 100.0
```

Generate a new set of data from the transformation above

```
In [140]: 1 new_data = data[['Price', 'HL_PCT', 'PCT_change', 'Volume']]
```

```
In [174]: 1 new_data.head(10)
```

Out[174]:

	Price	HL_PCT	PCT_change	Volume	label
0	32.50	4.307692	-5.797101	0	35.00
1	34.50	1.449275	-1.428571	0	34.95
2	35.00	2.571429	0.143062	0	36.65
3	34.95	4.434907	-4.638472	0	37.05
4	36.65	0.545703	-1.079622	0	36.50
5	37.05	2.564103	1.506849	0	36.00
6	36.50	1.917808	1.388889	0	38.00
7	36.00	5.138889	-5.263158	0	39.05
8	38.00	0.263158	-2.688860	0	32.50
9	39.05	1.024328	0.128205	0	34.50

Import the Machine Learning Algorithms from Scikit Learn library

```
In [142]: 1 from sklearn import preprocessing, cross_validation, svm
          2 from sklearn.linear_model import LinearRegression
          3 import math
```

Feature Engineering - Forecasting the stock prices

```
In [144]: 1 forecast_value = int(math.ceil(0.01 * len(new_data))) # Deduce the number of times to forecast
          2 # Add a new column that a shifting of the price values based on the size of data
          3 new_data['label'] = new_data['Price'].shift(-forecast_value)
```

```
In [145]: 1 new_data.head()
```

Out[145]:

	Price	HL_PCT	PCT_change	Volume	label
0	32.50	4.307692	-5.797101	0	35.00
1	34.50	1.449275	-1.428571	0	34.95
2	35.00	2.571429	0.143062	0	36.65
3	34.95	4.434907	-4.638472	0	37.05
4	36.65	0.545703	-1.079622	0	36.50

```
In [146]: 1 new_data.dropna(inplace= True)
```

Creating the features and the label for training the model

```
In [147]: 1 x = np.array(new_data.drop(['Price'], axis = 1))  
          2 y = np.array(new_data['Price'])
```

Pre-processing: Scaling the data to normalize it's values

```
In [148]: 1 #Pre-processing  
          2 x = preprocessing.scale(x)
```

```
In [149]: 1 # Splitting the data into training and test set  
          2 x_train, x_test, y_train, y_test = cross_validation.train_test_split(x, y, test_size=0.2)
```

Building the Machine Learning Model

Using Linear Regression Machine Learning Algorithm

```
In [150]: 1 linear_model = LinearRegression()  
          2 linear_model.fit(x_train, y_train)
```

```
Out[150]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=1, normalize=False)
```

Evaluating the Accuracy of the model

```
In [151]: 1 accuracy = linear_model.score(x_test, y_test)  
          2 accuracy = accuracy * 100  
          3 print('Percentage Accuracy = ', np.round(accuracy), '%')
```

```
Percentage Accuracy = 94.0 %
```

Using Support Vector Regression (SVR) Machine Learning Algorithm

```
In [189]: 1 clf = svm.SVR() #Creating an instance of SVR model
          2 clf.fit(x_train, y_train) # Fitting the model on the data
```

```
Out[189]: SVR(C=1.0, cache_size=200, coef0=0.0, degree=3, epsilon=0.1, gamma='auto',
            kernel='rbf', max_iter=-1, shrinking=True, tol=0.001, verbose=False)
```

Evaluating the Accuracy of the model

```
In [190]: 1 accuracy = clf.score(x_test, y_test)
          2 accuracy = accuracy * 100
          3 print('Percentage Accuracy = ', np.round(accuracy), '%')
```

```
Percentage Accuracy = 88.0 %
```

Using multiple kernels for the SVR

In [192]:

```
1 for k in ['linear', 'poly', 'rbf', 'sigmoid']:
2     clf = svm.SVR(kernel=k)
3     clf.fit(x_train, y_train)
4     confidence = clf.score(x_test, y_test)
5     print('Percentage Accuracy = ', k, np.round(accuracy), '%')
6
```

```
Percentage Accuracy = linear 88.0 %
Percentage Accuracy = poly 88.0 %
Percentage Accuracy = rbf 88.0 %
Percentage Accuracy = sigmoid 88.0 %
```


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


General ▾

 ▾ % , ←.0 .00
.00 →.0

Number □

Conditional Formatting Format as Table Cell Styles

Three icons for grid manipulation: 'Insert' shows a grid with an arrow pointing left to insert a column; 'Delete' shows a grid with a red 'X' over a column; 'Format' shows a grid with a blue cell and a double-headed arrow indicating formatting. Below the icons are the labels 'Insert', 'Delete', and 'Format', each with a small downward arrow. The word 'Cells' is centered below these labels.

Σ AutoSum ▾ A Z  
 Fill ▾ Sort & Find &
 Clear ▾ Filter ▾ Select ▾
 Editing

[illegible]

Conclusion

This model can be deployed as a web service on an automated trading platform.

Link to Datasets & Code on GitHub:

[https://github.com/KelvinOyanna/Pyc
on-Nigeria-2018-Resource.git](https://github.com/KelvinOyanna/Pyc
on-Nigeria-2018-Resource.git)

Got Questions?



Thank you!