

A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light green. They are positioned diagonally, with the blue one partially covering the green one.

# Predicting the Stocks

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# Topics

- The Problem
- The Goal
- Our Approach
- Conclusion





# The Problem

- People lose lots of money in stocks
- One of the most difficult jobs is to analyze and predict the trend of a stock
- Market is extremely volatile, nearly everything online can affect its market price
- An algorithm can help people make passive income



# The Goal

To program a software capable enough to take existing data from the market and utilize it to train itself and eventually predict future trends for any given stock



## Our Approach

We'll be utilizing Long Short-Term Memory method to train the data and accurately predict the future trends

LSTM is a a deep learning artificial recurrent neural network (RNN) architecture.

Unlike traditional feed-forward neural networks, LSTM has feedback connections. It can handle single data points (such as pictures) as well as full data sequences (such as speech or video).

# Prepare the data for analysis:

Download using yahoo finance:

```
[3] !pip install yfinance
```

Raw Data:

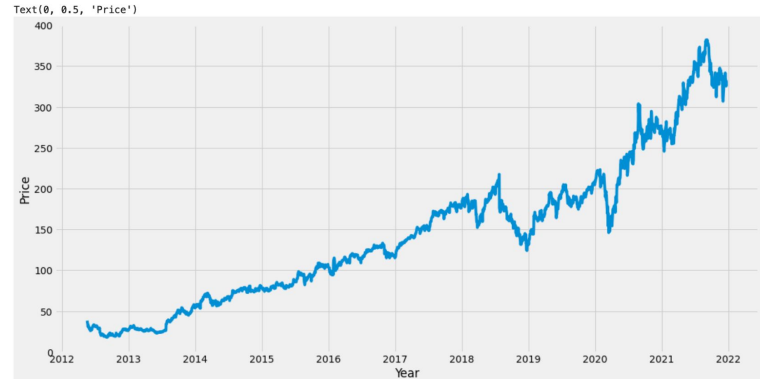
DataFrame with shape (2415, 7)

```
hist = fb.history(period="max", auto_adjust=True)
#print histogram
hist
```

Date	Open	High	Low	Close	Volume	Dividends	Stock Splits
2012-05-18	42.049999	45.000000	38.000000	38.230000	573576400	0	0
2012-05-21	36.529999	36.660000	33.000000	34.029999	168192700	0	0
2012-05-22	32.610001	33.590000	30.940001	31.000000	101786600	0	0
2012-05-23	31.370001	32.500000	31.360001	32.000000	73600000	0	0
2012-05-24	32.950001	33.209999	31.770000	33.029999	50237200	0	0
...	...	...	...	...	...	...	...
2021-12-15	332.489990	342.109985	323.980011	341.660004	24681300	0	0
2021-12-16	338.980011	344.459991	333.739990	334.899994	22635000	0	0
2021-12-17	332.799988	337.109985	330.750000	333.790009	39711700	0	0
2021-12-20	329.779999	329.899994	322.529999	325.450012	17868700	0	0
2021-12-21	326.410004	333.135803	323.750000	332.640015	7842779	0	0

2415 rows x 7 columns

```
[33] import matplotlib.pyplot as plt
plt.figure(figsize=(16,8))
year = hist.index
output_var = hist['Close'].values
output_var=np.array(output_var)
plt.plot(year, output_var)
plt.xlabel("Year")
plt.ylabel("Price")
```

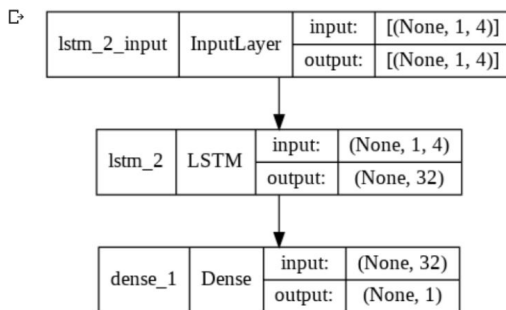


## Split the data into training and test:

```
[10] #Creating training and test data using TimeSeriesSplit
from sklearn.model_selection import TimeSeriesSplit
timesplit= TimeSeriesSplit(n_splits=10)
for train_index, test_index in timesplit.split(feature_transform):
    X_train, X_test = feature_transform[:len(train_index)], feature_transform[len(train_index): (len(train_index)+len(test_index))]
    y_train, y_test = output_var[:len(train_index)].ravel(), output_var[len(train_index): (len(train_index)+len(test_index))].ravel()
```

## Building first LSTM model with one layer:

```
lstm = Sequential()
lstm.add(LSTM(32, input_shape=(1, trainX.shape[1]), activation='relu', return_sequences=False))
lstm.add(Dense(1))
lstm.compile(loss='mean_squared_error', optimizer='adam', metrics=['accuracy'])
plot_model(lstm, show_shapes=True, show_layer_names=True)
```



## Training the data:

```
[18] #Model Training
history=lstm.fit(X_train, y_train, epochs=100, batch_size=8, verbose=1, shuffle=False)
```

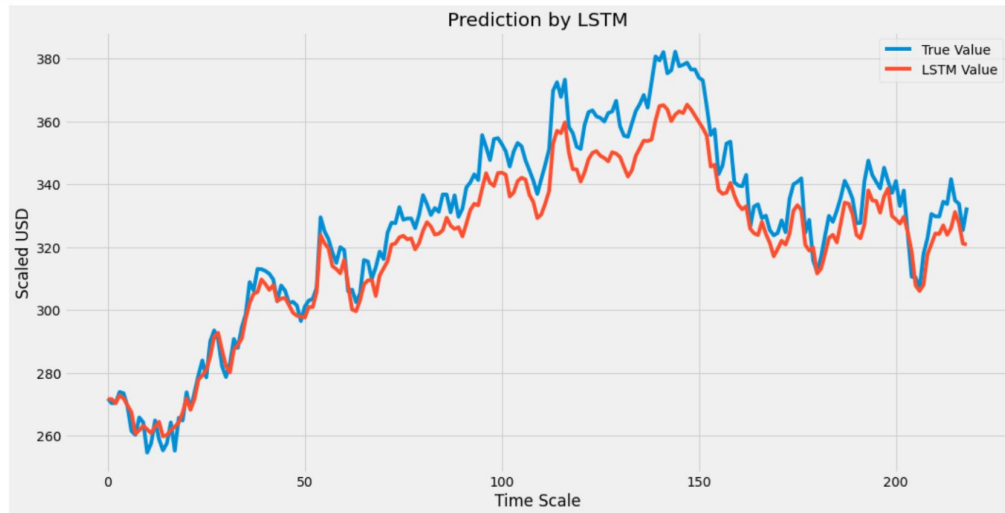
Loss

Epoch 1/100

275/275 [=====] - 2s 2ms/step - loss: 19715.5996 -

Epoch 100/100

275/275 [=====] - 1s 2ms/step - loss: 3.2842

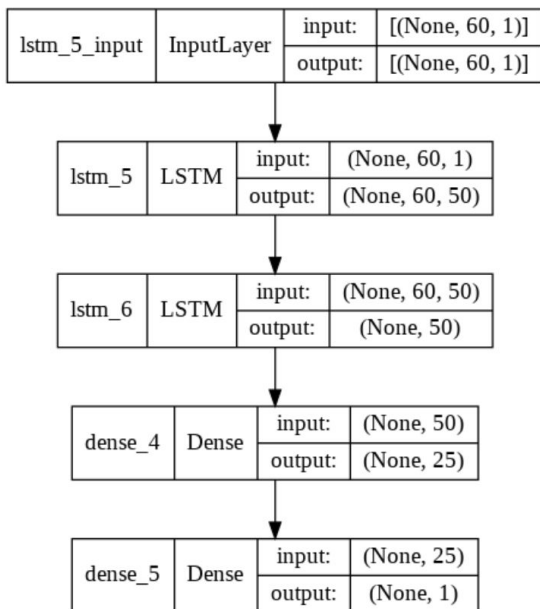


```
[21] print(np.mean(y_test==y_pred))
```



## Building second LSTM model with multiple layers:

```
[30] model = Sequential()  
model.add(LSTM(50, return_sequences=True, input_shape=(x_train.shape[1],1) ))  
model.add(LSTM(50, return_sequences=False))  
model.add(Dense(25))  
model.add(Dense(1))  
plot_model(model, show_shapes=True, show_layer_names=True)
```



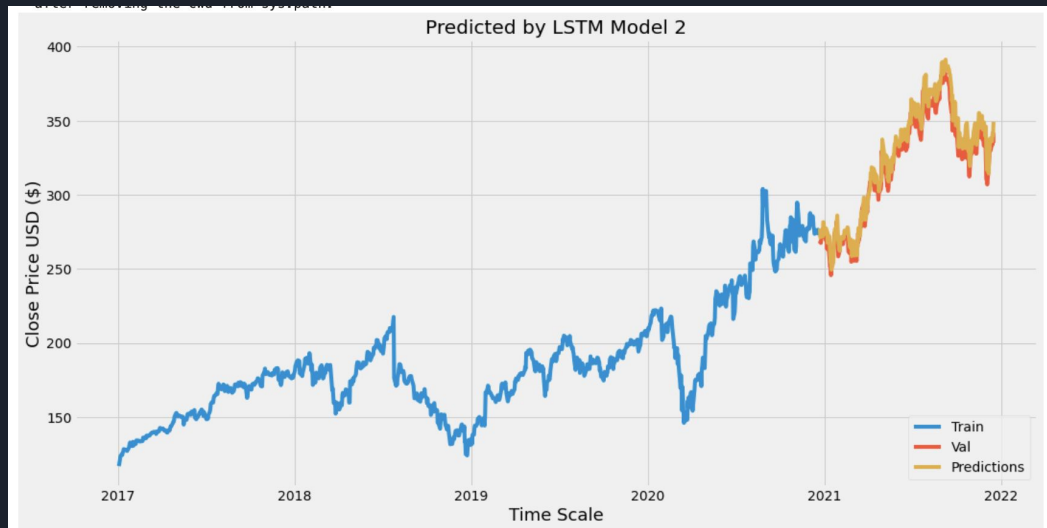
## Training the data for second Model:

```
[31] #compile model
      model.compile(optimizer = 'adam', loss= 'mean_squared_error')
```

Smaller Loss

```
#model fit
model.fit(x_train, y_train, batch_size = 1, epochs =1)
```

```
940/940 [=====] - 26s 28ms/step - loss: 4.8372e-04
<keras.callbacks.History at 0x7f108e50ddd0>
```



```
[ ] valid
```

Close Predictions		
Date		
2020-12-18	276.399994	278.946350
2020-12-21	272.790009	278.604950
2020-12-22	267.089996	278.031799
2020-12-23	268.109985	276.757935
2020-12-24	267.399994	275.450134
...	...	...
2021-12-08	330.559998	321.547028
2021-12-09	329.820007	322.355347
2021-12-10	329.750000	323.856262
2021-12-13	334.489990	325.555634
2021-12-14	333.739990	327.765259

249 rows x 2 columns



# Conclusion

- Over 50% of citizens in the US aren't invested into the stock market
- All billionaires and other financially successful people are
- We want to change that = encourage others to invest in the stock market
- Easy-to-use software that can also help teach you about the market while making you passive income



## References

1. <https://www.analyticsvidhya.com/blog/2021/10/machine-learning-for-stock-market-prediction-with-step-by-step-implementation/>
2. <https://youtu.be/QIUxPv5PJOY>
3. [https://en.wikipedia.org/wiki/Long\\_short-term\\_memory](https://en.wikipedia.org/wiki/Long_short-term_memory)
4. <https://journalofbigdata.springeropen.com/articles/10.1186/s40537-020-00333-6>