

```
dataset_train = pd.read_csv("Google_Stock_Price_Train.csv")
dataset_train.head()
```

	Date	Open	High	Low	Close	Volume
0	1/3/2012	325.25	332.83	324.97	663.59	7,380,500
1	1/4/2012	331.27	333.87	329.08	666.45	5,749,400
2	1/5/2012	329.83	330.75	326.89	657.21	6,590,300
3	1/6/2012	328.34	328.77	323.68	648.24	5,405,900
4	1/9/2012	322.04	322.29	309.46	620.76	11,688,800

# Understanding Long Short Term Memory Network

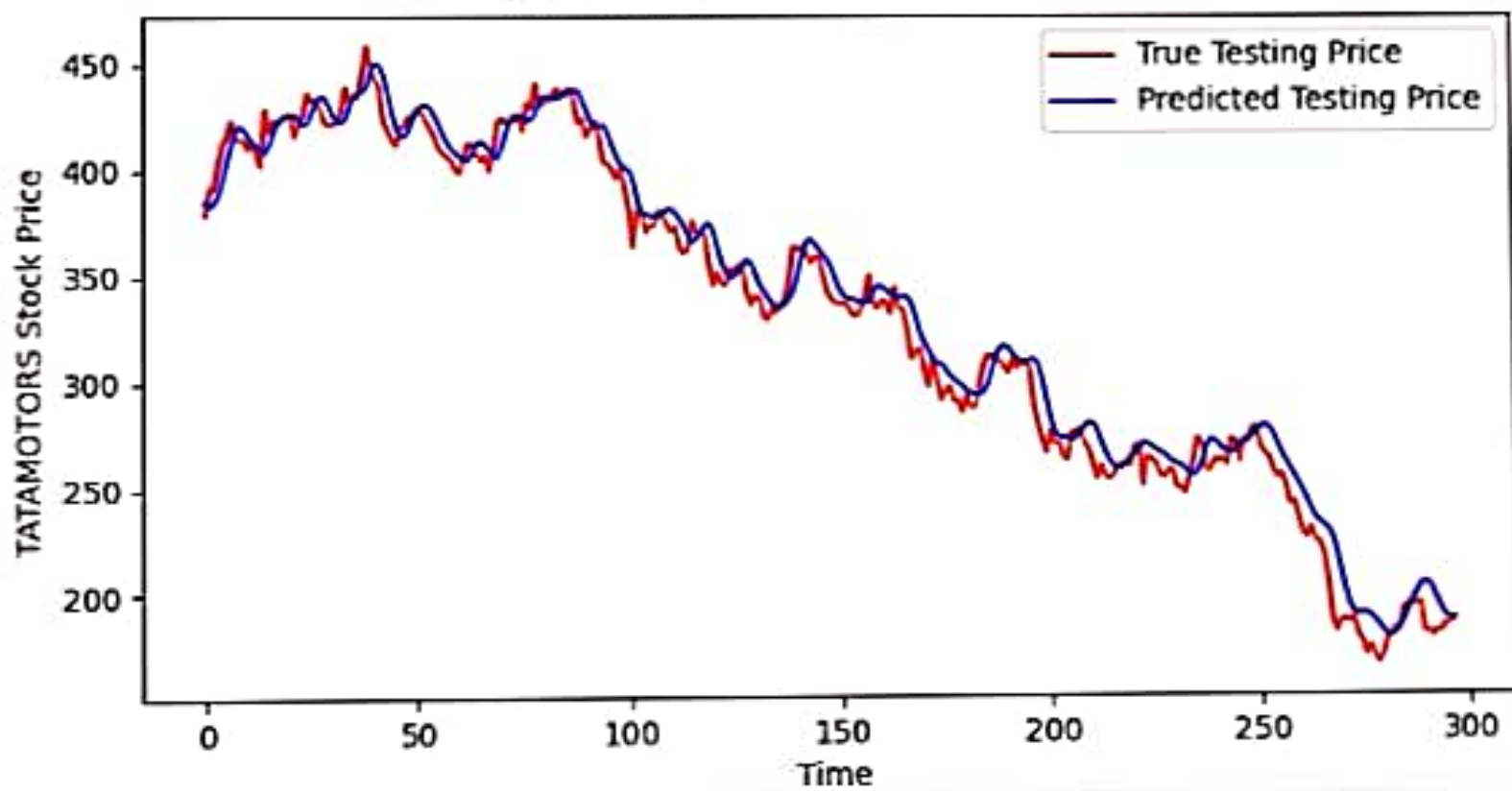
Here, you will use a Long Short Term Memory Network (LSTM) for building your model to predict the stock prices of Google.

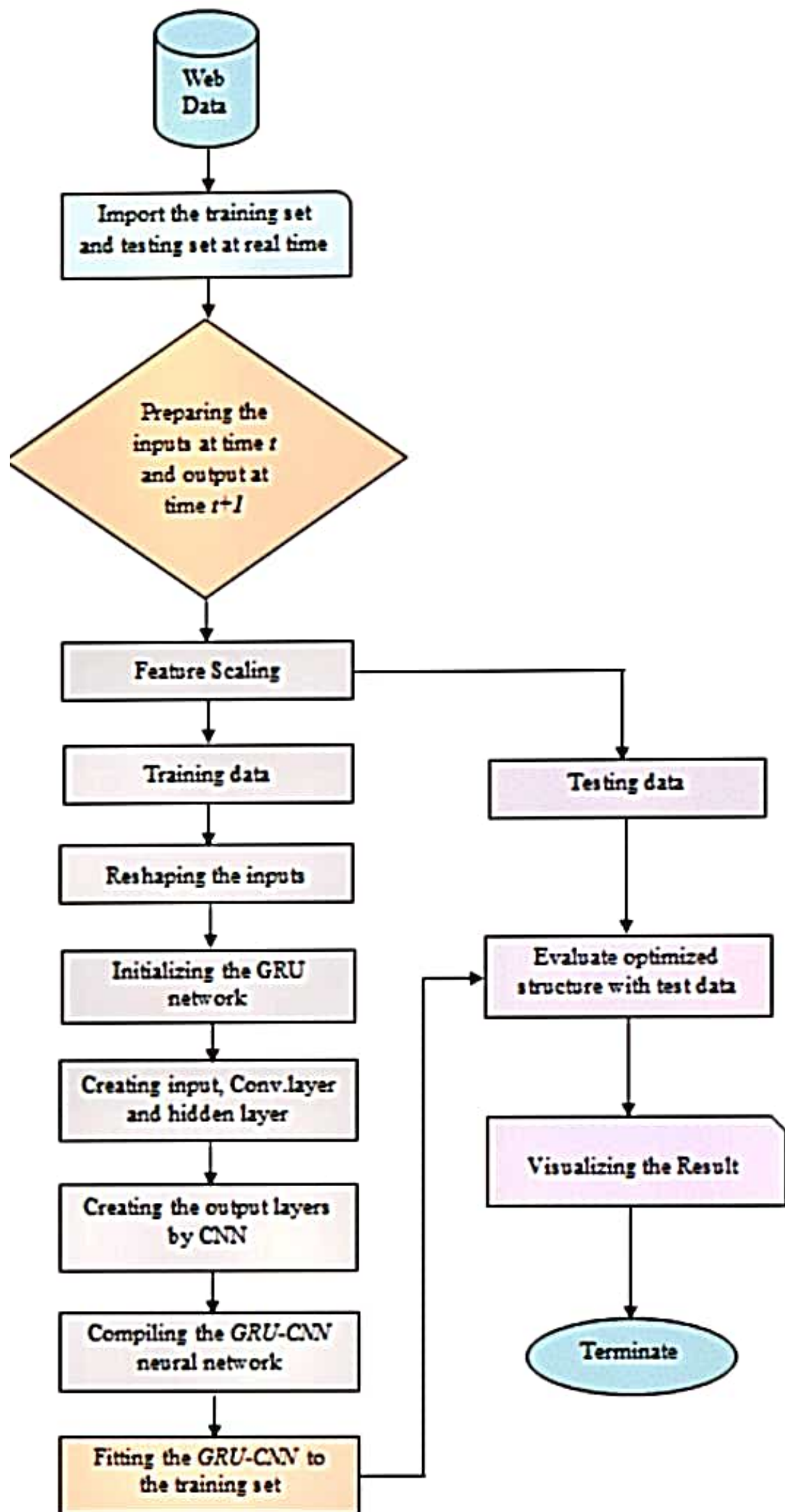
LTSMs are a type of [Recurrent Neural Network](#) for learning long-term dependencies. It is commonly used for processing and predicting time-series data.

# Importance of Stock Market

- Stock markets help companies to raise capital.
- It helps generate personal wealth.
- Stock markets serve as an indicator of the state of the economy.
- It is a widely used source for people to invest money in companies with high growth potential.

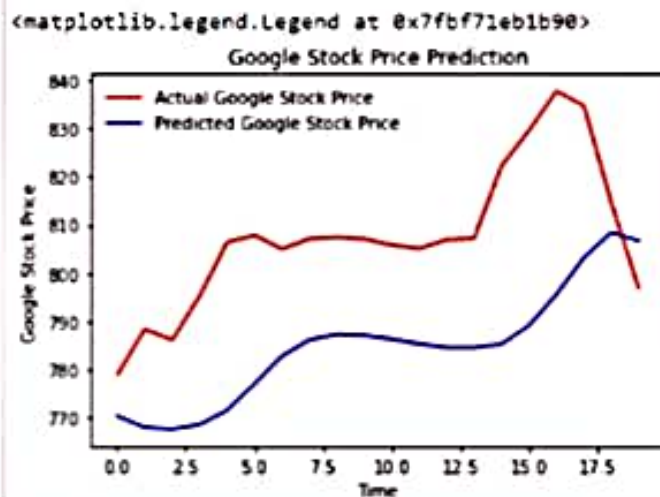
TATAMOTORS Stock Price Prediction





## 12. Plotting the Actual and Predicted Prices for Google Stocks.

```
plt.plot(actual_stock_price, color = 'red', label = 'Actual Google Stock Price')  
plt.plot(predicted_stock_price, color = 'blue', label = 'Predicted Google Stock Price')  
plt.title('Google Stock Price Prediction')  
plt.xlabel('Time')  
plt.ylabel('Google Stock Price')  
plt.legend()
```



As you can see above, the model can predict the trend of the actual stock prices very closely. The accuracy of the model can be enhanced by training with more data and increasing the LSTM layers.





## 7. Building the Model by Importing the Crucial Libraries and Adding Different Layers to LSTM.

```
from keras.models import Sequential
from keras.layers import LSTM
from keras.layers import Dense
from keras.layers import Dropout
```

```
regressor = Sequential()

regressor.add(LSTM(units = 50, return_sequences= True, input_shape = (X_train.shape[1], 1)))
regressor.add(Dropout(0.2))

regressor.add(LSTM(units = 50, return_sequences= True))
regressor.add(Dropout(0.2))

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regressor.add(Dropout(0.2))

regressor.add(LSTM(units = 50))
regressor.add(Dropout(0.2))

regressor.add(Dense(units=1))
```

### 3. Use the Open Stock Price Column to Train Your Model.

```
training_set = dataset_train.iloc[:,1:2].values

print(training_set)
print(training_set.shape)

[[325.25]
 [331.27]
 [329.83]
 ...
 [793.7 ]
 [783.33]
 [782.75]]
(1258, 1)
```

### 4. Normalizing the Dataset.

```
from sklearn.preprocessing import MinMaxScaler

scaler = MinMaxScaler(feature_range = (0,1))
scaled_training_set = scaler.fit_transform(training_set)

scaled_training_set

array([[0.08581368],
       [0.09701243],
       [0.09433366],
```





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