

# Predicting Stock Prices

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# 01

## The Idea

Why this project excited us!



# PROBLEM

The stock market is exciting and a great way to invest and make money. But how to choose which stocks to invest in, without any financial knowledge?

# SOLUTION

As data Scientists, we can leverage available data to create a prediction model that helps us understand the performance of stocks in the market and then invest money accordingly



# 02

## The Data

What data did we use?

# The Data

- We extracted the data from **<https://finance.yahoo.com/>** which hosts the stock prices till date for the companies listed on the stock market.
- We used the **pandas\_datareader** to get data from Yahoo Finance.
- We took Google stock data from 2010- 2022(April).
- The API returns the Closing price, Start price and the volume of stocks, traded for each day.

# 03

## The Modelling

How did Deep Learning Help?



# How does Deep Learning Help

- For the modelling of this data, we used the LSTM model.
- LSTM layers have an inherent property to control the inflow and outflow of only the relevant information from the past combined with the present information.
- This is vital in the context of stock price prediction as they largely depend on the historical data.



# How does Deep Learning Help

```
# Defining the model
# This model uses LSTM layers

class LSTM(nn.Module):
    def __init__(self, input_dim, hidden_dim, num_layers, output_dim):
        super(LSTM, self).__init__()
        self.hidden_dim = hidden_dim
        self.num_layers = num_layers
        self.lstm = nn.LSTM(input_dim, hidden_dim,
                             num_layers, batch_first=True)
        self.fc = nn.Linear(hidden_dim, output_dim)

    def forward(self, x):
        h0 = torch.zeros(self.num_layers, x.size(
            0), self.hidden_dim).requires_grad_()
        c0 = torch.zeros(self.num_layers, x.size(
            0), self.hidden_dim).requires_grad_()
        out, (hn, cn) = self.lstm(x, (h0.detach(), c0.detach()))
        out = self.fc(out[:, -1, :])
        return out
```

# 04

## The Process

What went well and what  
issues did we face?



# What went well

- The process of downloading the data was very smooth and fast.
- The training was also very fast on the CPU.

# What issues did we have

- A lot of variation and fluctuation in data causes poor predictions in long term
- Predictions flat-line because of the same
- The need to train the model frequently to incorporate the historical effects with the new data

# 05

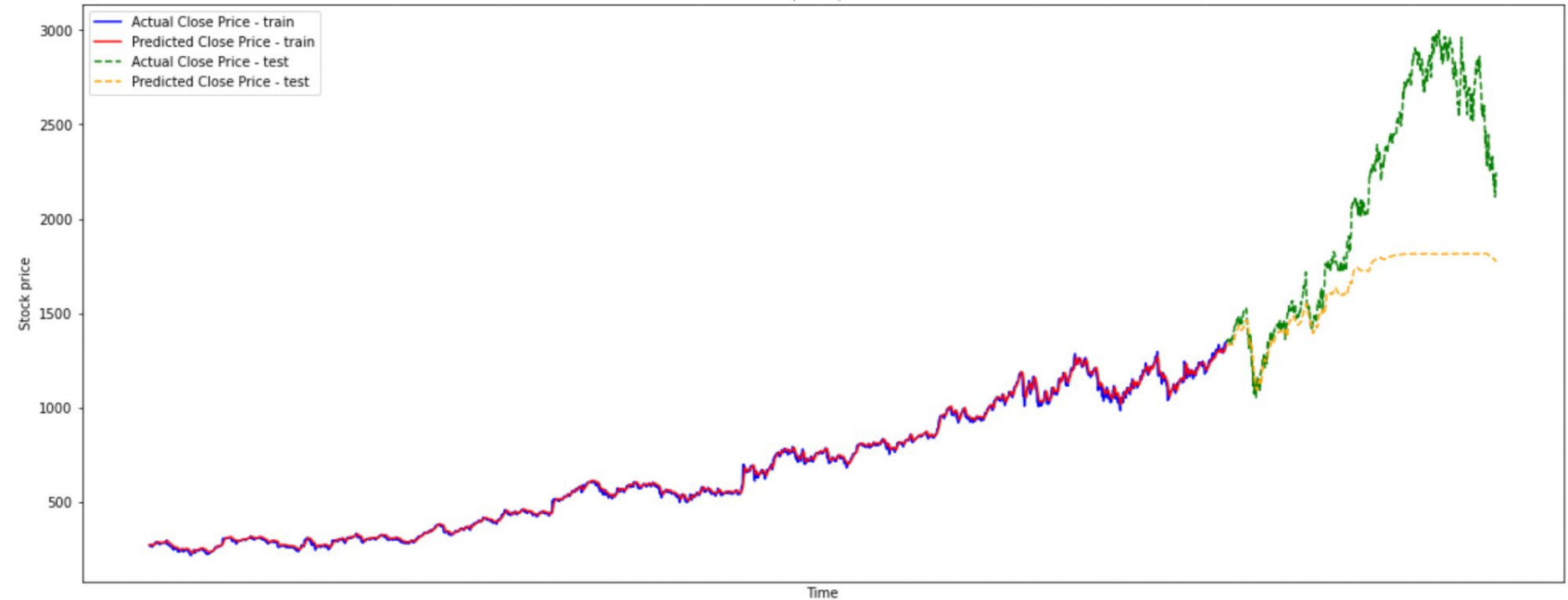
## The Conclusion

How is the final result?



# Results and Conclusion

GOOGL price prediction



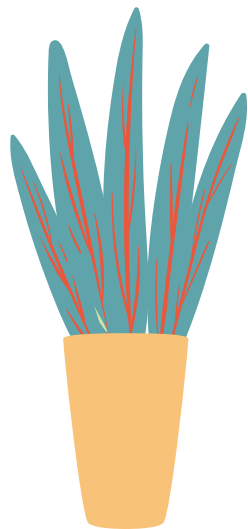
# Results and Conclusion

- Number of Epochs run : 800
- Training time : 300s
- Mean Squared Error (train) : 0.000139
- Mean Squared Error (test) : 0.185

# Conclusion and Future Goals

- The predictions are not much accurate for the future data
- A more complex model might be helpful that can capture the trend and fluctuations
- A good approach might be to consider the extrinsic factors of the market as well.





**Thank You**