

Jesse Mailhot

Capstone Project 2

Comparing LSTM & Prophet on
Predicting Stock Price Movement

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Introduction

- ❖ Predicting stock price movement
- ❖ LSTM
- ❖ Facebook Prophet

Data Set

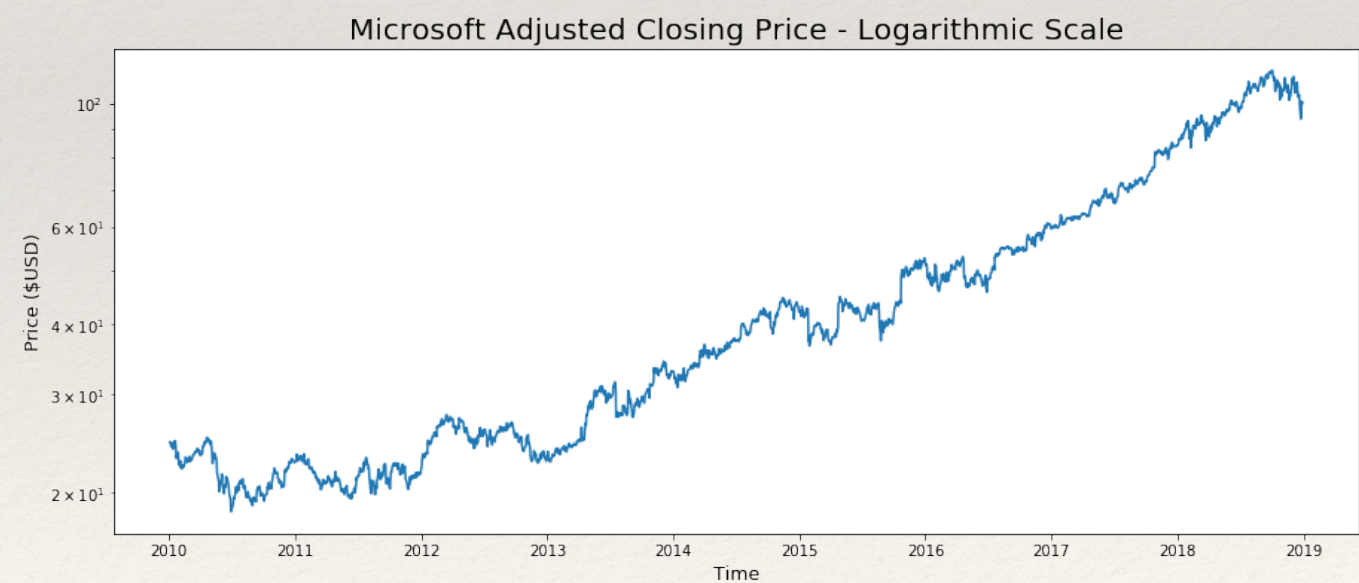
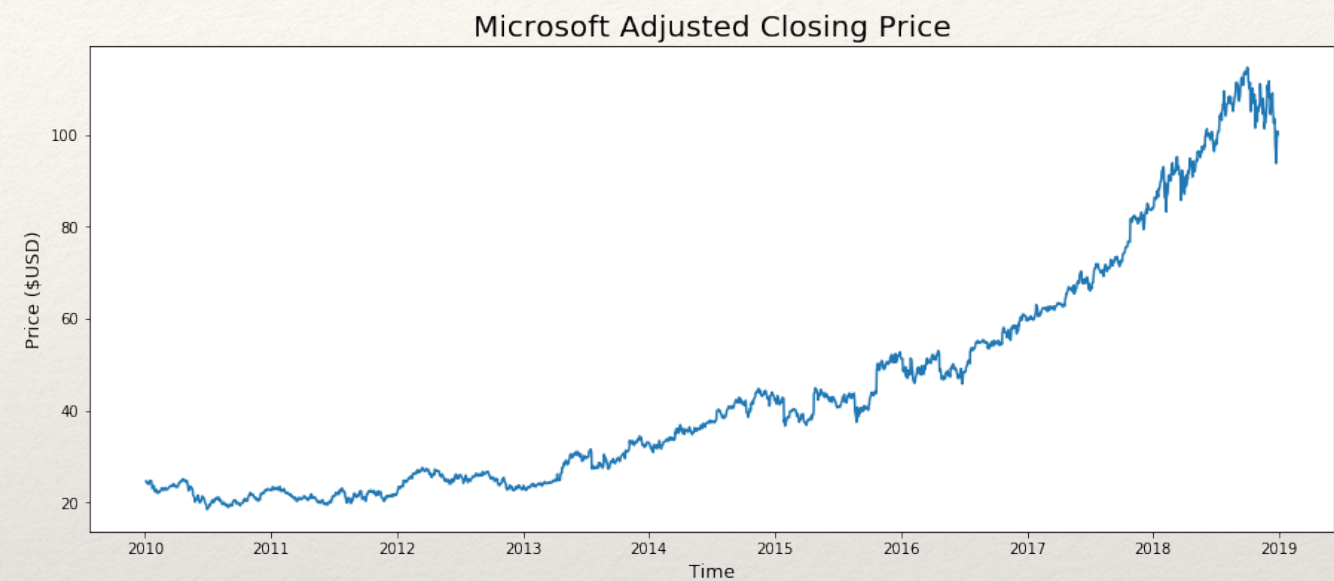
- ❖ Alphavantage API
- ❖ Microsoft Stock
 - ❖ 1 / 2010 - 12 / 2018

Data Wrangling

- ❖ Custom function to pull data from API
 - ❖ Symbol, start date, end date
 - ❖ Returns pandas data frame
- ❖ Save data frame as CSV file for further analysis

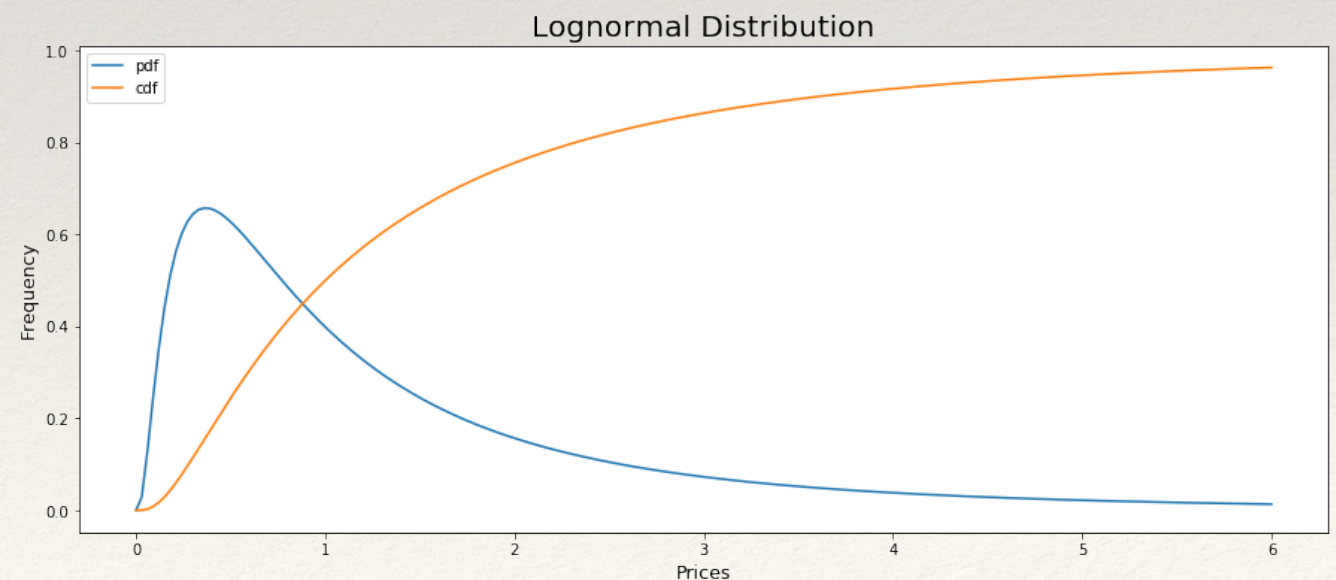
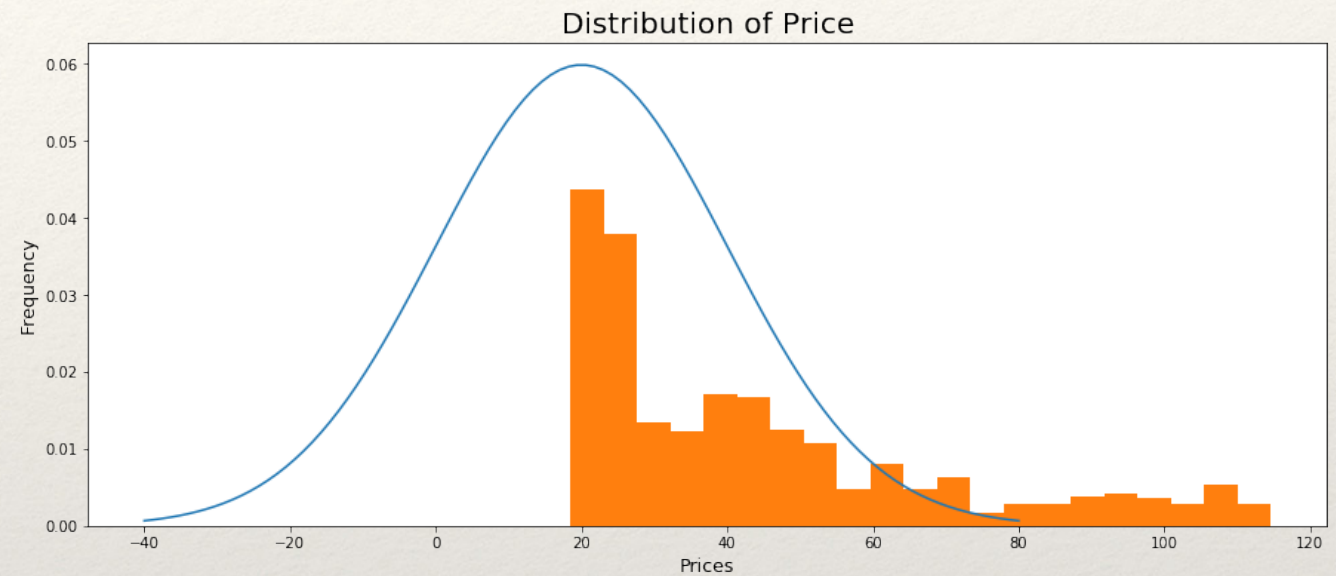
Data Analysis

- ❖ Bias associated with price
- ❖ Logarithmic scale
- ❖ Long term trend



Data Analysis

- ❖ Lognormal returns
- ❖ Lognormal distribution
- ❖ Instantaneous rate of change of return



Preprocessing

LSTM

- ❖ Supervised learning
 - ❖ ± 60 days
- ❖ Train test split
 - ❖ Avoid times series bias
- ❖ LSTM format
 - ❖ data, steps, features

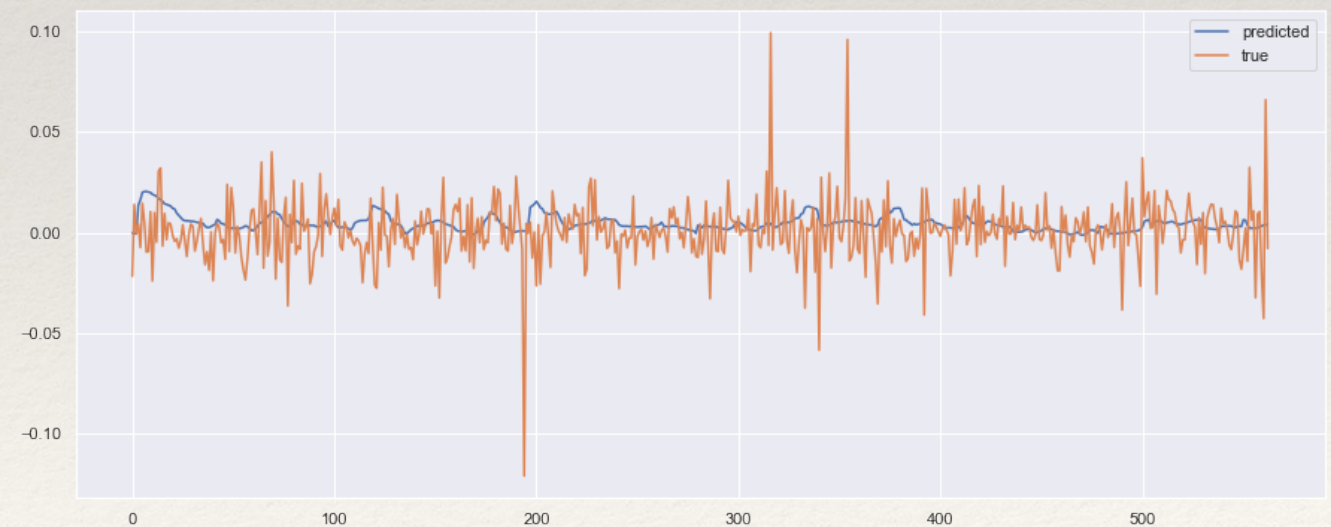
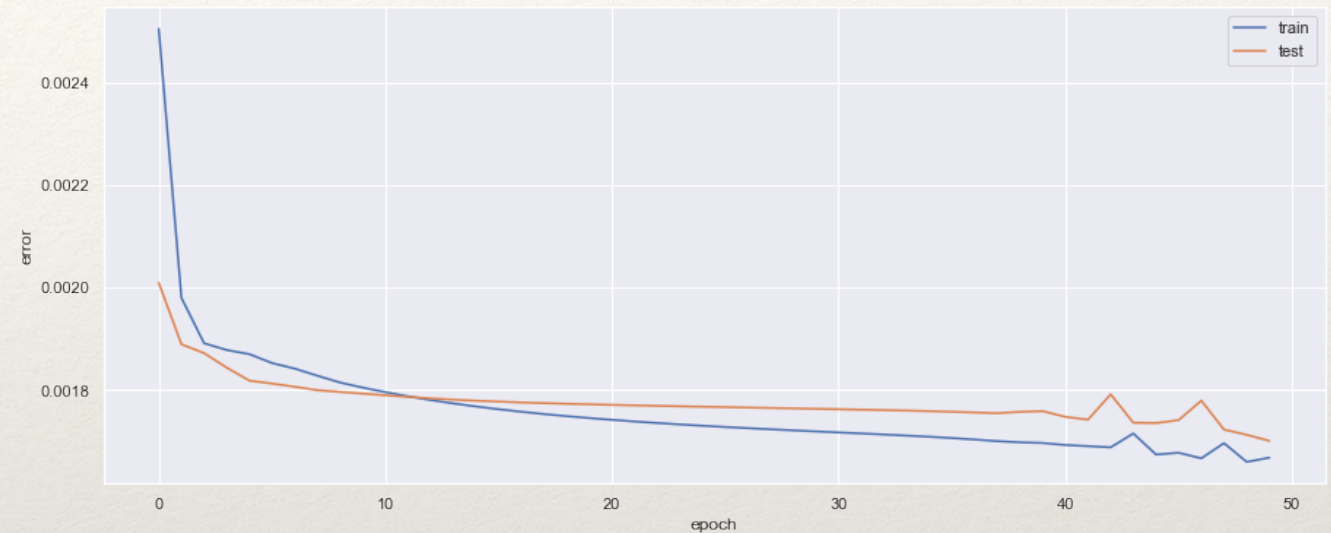
Prophet

- ❖ 'ds' & 'y'

Model Analysis

LSTM

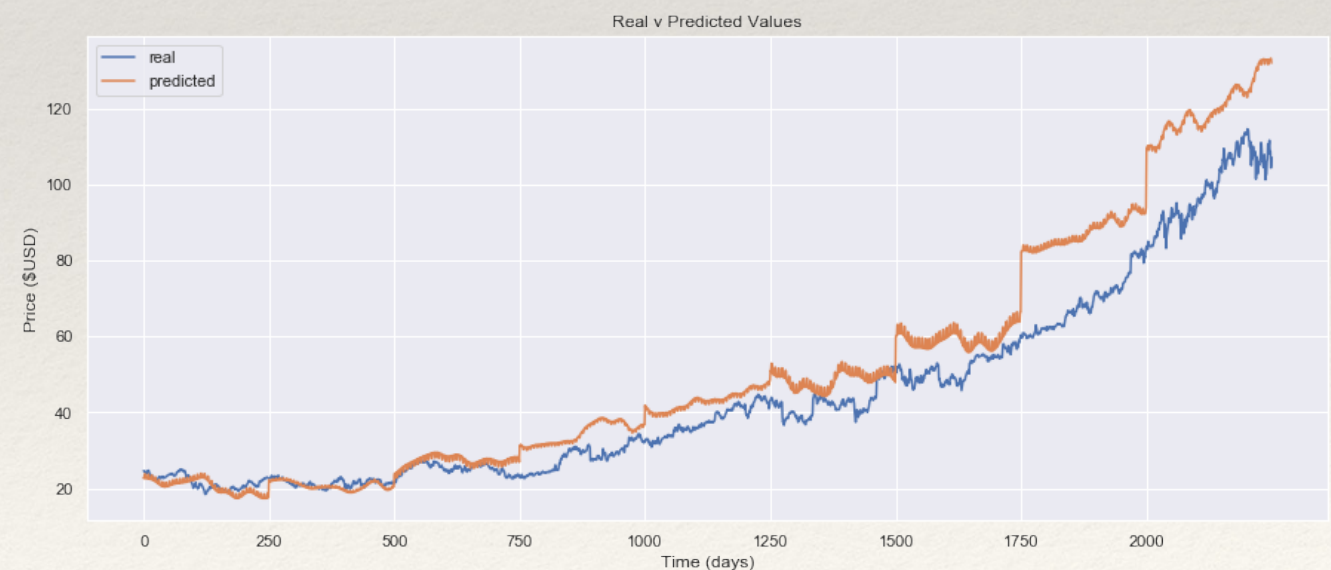
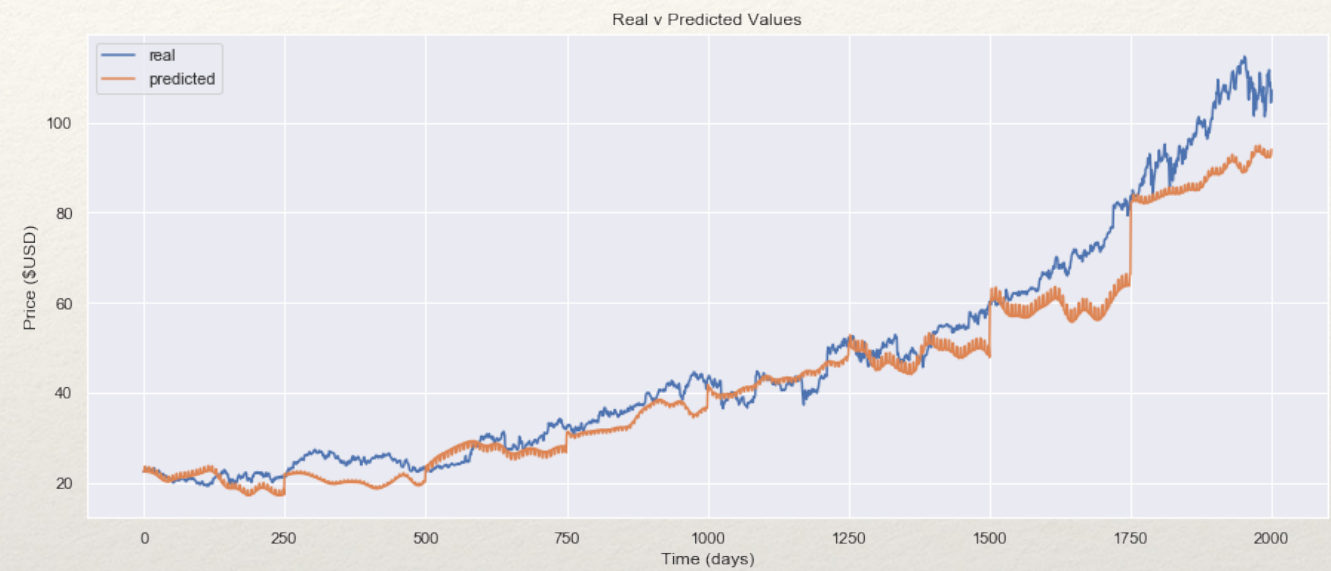
- ❖ 3 layers
- ❖ 50 nodes
- ❖ Loss: MSE
- ❖ Optimizer: ADAM
- ❖ Training / testing loss
- ❖ Daily return predictions
- ❖ Direction vs. Value



Model Analysis

Prophet

- ❖ Yearly seasonality
- ❖ For loop vs manual



Model Comparison

LSTM

- ❖ Short vs long term predictions
 - ❖ 53% daily return direction
 - ❖ 74% quarterly return direction
- ❖ Averse to time series bias

Prophet

- ❖ 100% correct predictions
 - ❖ Small sample size
 - ❖ Stock market assumptions

Next Steps

- ❖ Other stock prices
- ❖ Shorter predictions with prophet
 - ❖ Takes a long time to train and predict
 - ❖ May not be able to properly predict using a loop
- ❖ Other deep learning models
 - ❖ CNN, GRU