

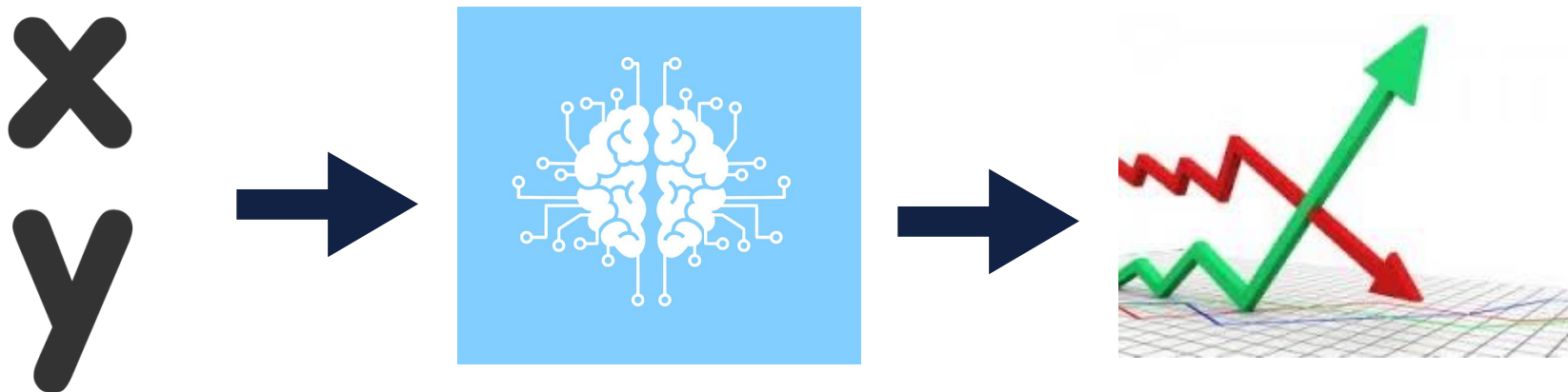
Predicting Closing Stock Exchange Prices on the NYSE & NASDAQ

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Big Question

Based on a given set of variables, can machine learning predict whether a stock exchange index will close higher or lower than previous close?



Applicability

Personal Finance



charles
SCHWAB

SoFi 

Mobile Investing

Robinhood 



TD
Ameritrade

Project Background

Dataset

We utilized two datasets from Kaggle (please see link below)

<https://www.kaggle.com/mattiuzc/stock-exchange-data>

Columns

The two CSVs were cross referenced by the Index to allow for more detail.

- **indexProcessed.csv** - Stock Exchange Index, Date, Open, High, Low, Close, Adj Close, Volume
- **indexInfo.csv** - Region, Exchange, Index, Currency

Technology

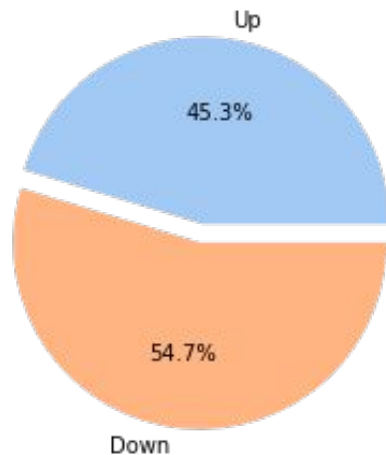
We used 3 main technologies for different aspects of the project:

- **SQL - Postgres** - database
- **Python** - data preparation and machine learning
- **Tableau** - data visualization

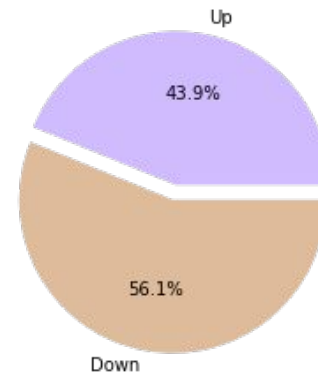
Exploratory Analysis

These charts show the overall distribution of Up Vs. Down since 1966

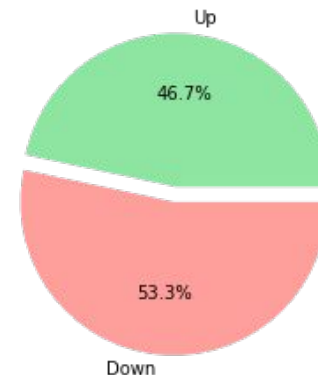
NYSE and NASDAQ - Up/Down from 1966-2021



NASDAQ - Up/Down from 1966-2021

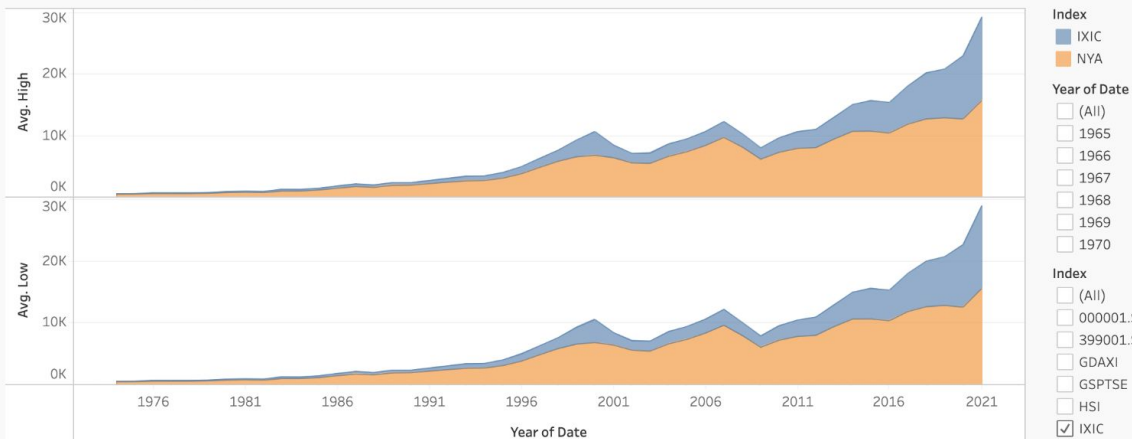


NYSE - Up/Down from 1966-2021



EXPLORATORY ANALYSIS

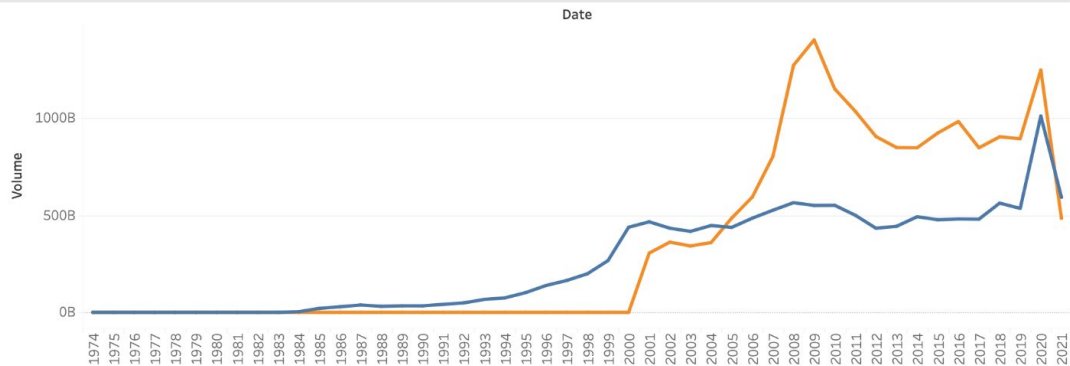
NYA V NASDAQ High Low Chart



New York Stock Exchange and NASDAQ Over Time

This dashboard shows the importance of our exploration into the NYSE and NASDAQ data. We can see that over time people are buying indices more and the average high is increasing.

NYA and NASDAQ Volume



ER Diagram

We started with 2 tables from Kaggle:

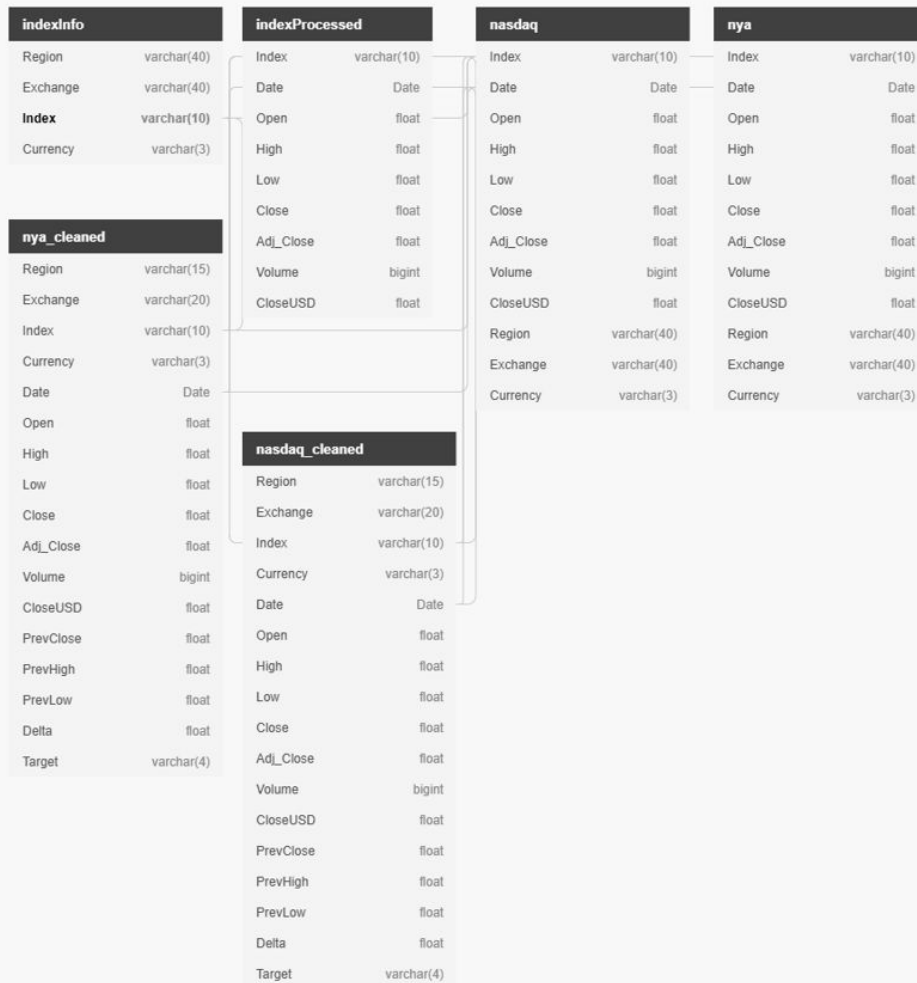
- indexInfo
- indexProcessed

We created 2 tables:

- Nasdaq & NYA - This was done by using a left Join on the “indexProcessed” and “indexInfo” table with a filter by index

We exported 2 tables back to SQL:

- NASDAQ_cleaned & NYA_cleaned - These tables are copied of the Nasdaq and NYA tables with new column names for data cleansing purposes



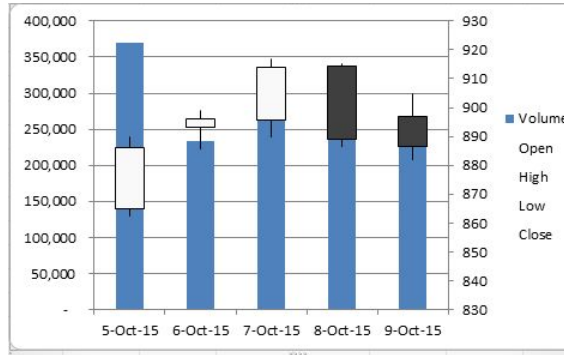
VARIABLES

Name	Type	Explanation	NYA	NASDAQ
Index	Object	Index name (Eg: NYA, IXIC)	✓	✓
Date	Datetime	Date in question	✓	✓
Open	Float64	Opening value of the Index on the date	✓	✓
High	Float64	High value of the Index on the date	✓	✓
Low	Float64	Low value of the Index on the date	✓	✓
Close	Float64	Closing value of the Index on the date	✓	✓
Adj Close	Float64	Adjusted Closing value of the Index on the date	✓	✓
Volume	Int64	Volume traded of the Index on the date	✓	✓
CloseUSD	Float64	Closing value of the Index on the date in USD	✓	✓
Region	Object	Region where Index is located	✓	✓
Exchange	Object	Full name (Eg: New York Stock Exchange)	✓	✓
Currency	Object	Currency index is traded in	✓	✓

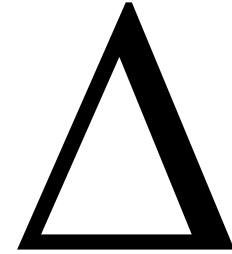
Data Cleansing & Wrangling



Dropped NaN/ Nulls



Added previous day's High, Low, Close to today's row



Added Delta and Target Columns

- Delta: % gain or loss based on Close price
- Target: Determine Up/Down Trend based on Delta

Machine Learning Models

Supervised Machine Learning Models Used:

- Logistic Regression
- Balanced Random Forest Classifier
- SMOTE oversampling
- Undersampling
- Decision Tree Model

Machine Learning Preparation:

We decided on the following **Features** for the model:

- Open
- PrevHigh
- PrevLow
- PrevClose

We tried to predict the **Target**: whether or not the index closed higher or lower than the previous day.

Logistic Regression

NYSE

Accuracy: 0.85

NASDAQ

Accuracy: 0.62

Drawback: Assumes linearity between dependent and independent variables

Success: Solid baseline!

Decision Tree Model

NYSE

Accuracy: 0.76
Precision: 0.76
F1: 0.76

NASDAQ

Accuracy: 0.63
Precision: 0.63
F1: 0.63

Drawback: Sensitive to small changes in data

Success: Visually intuitive and efficient!

Balanced Random Forest Classifier

NYSE

Accuracy: 0.77
Precision: 0.77
F1: 0.77

NASDAQ

Accuracy: 0.64
Precision: 0.64
F1: 0.64

Drawback: Value gained from additional samples drops off over time

Success: Handles linear and non-linear relationships well!

SMOTE Oversampling

NYSE

Accuracy: 0.87
Precision: 0.88
F1: 0.87

NASDAQ

Accuracy: 0.78
Precision: 0.78
F1: 0.78

Drawback: Overfitting more likely

Success: Doesn't lose any information!

Undersampling

NYSE

Accuracy: 0.87
Precision: 0.87
F1: 0.87

NASDAQ

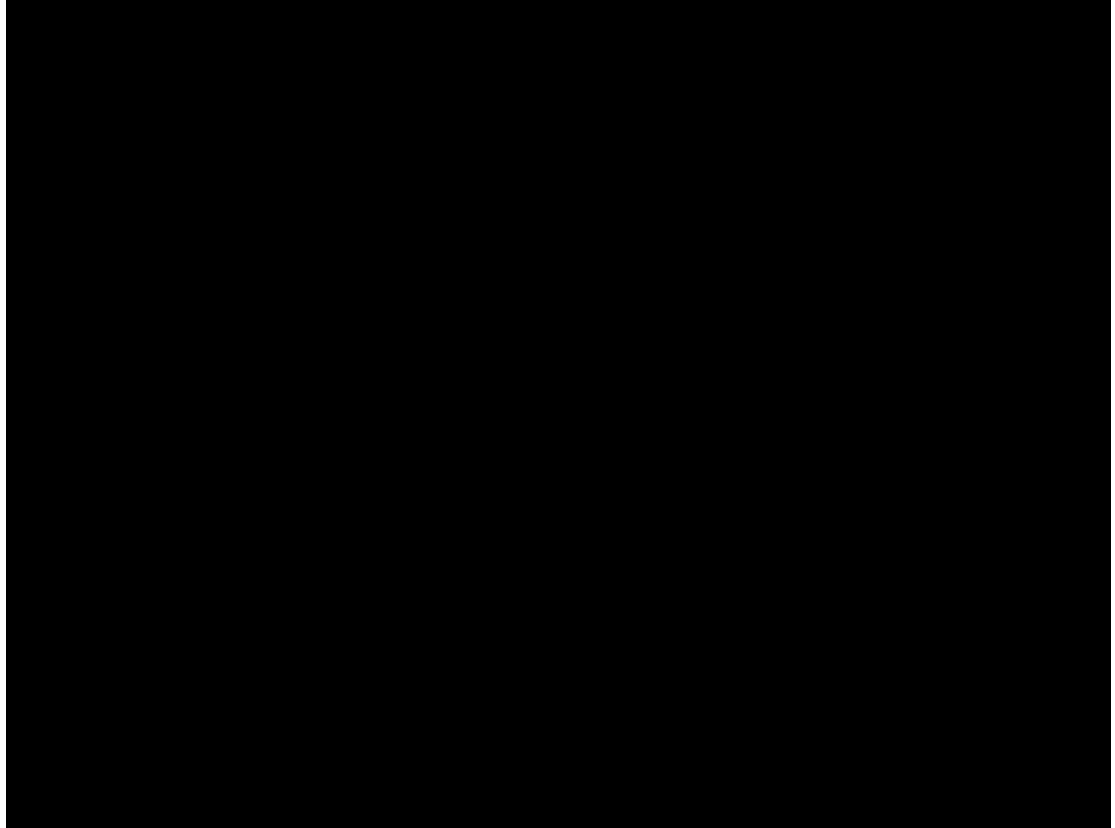
Accuracy: 0.77
Precision: 0.78
F1: 0.78

Drawback: Potentially discarding useful data

Success: Helps against skewing towards the majority class!

Machine Learning Model Results and Score Trends

[Tableau](#)



Results



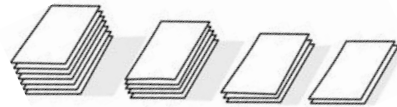
It was better at predicting for the NYSE than NASDAQ



The model was more accurate at predicting **higher** closings than it was at predicting **lower** closings



SMOTE Oversampling and Undersampling had the highest accuracy, precision, and F1 scores



Accuracy was better when resampled

What We Would Do Differently

1

Zoom in on a particular stock or industry to see if there will be an accuracy improvement

2

Look up existing machine learning progress/accuracy with stock predictions

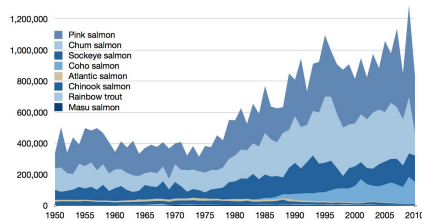
3

Use a different model such as one that predicts continuous variables

Recommendations for Future Analysis



Predict closing *price*



Account for time series



Run model with different
stock exchanges



Predict closing direction for
individual stocks

