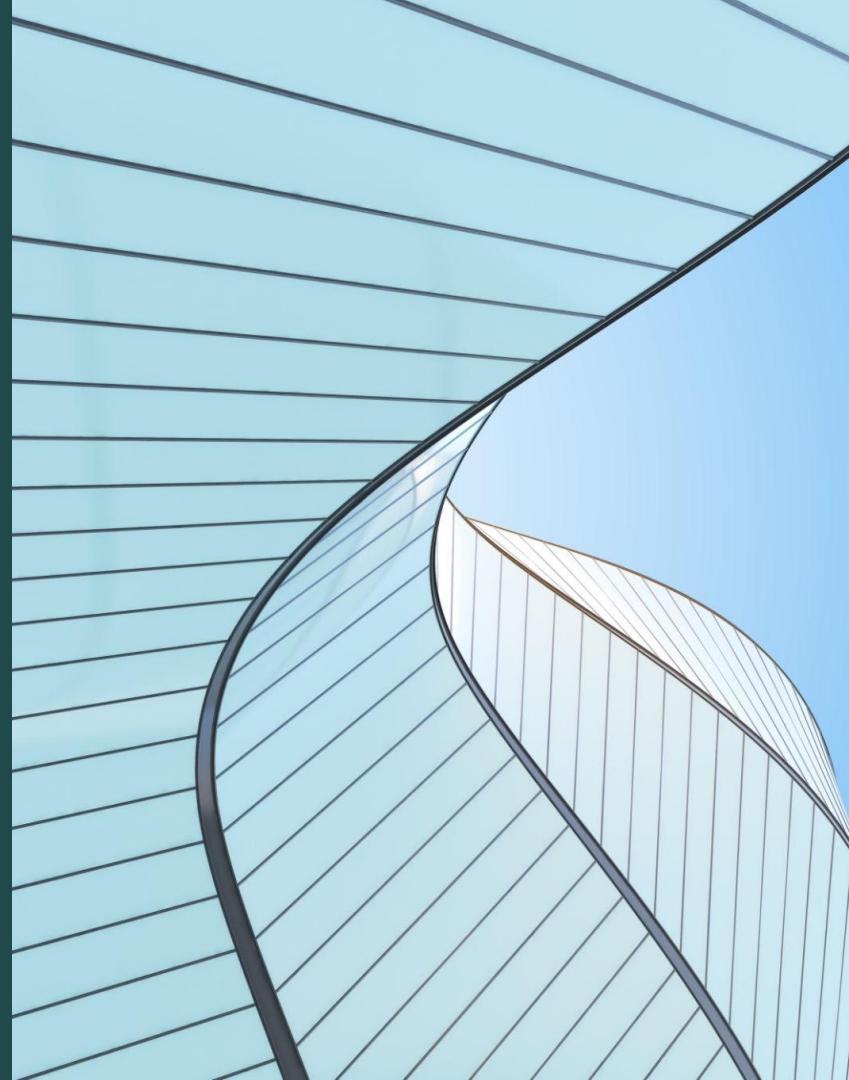


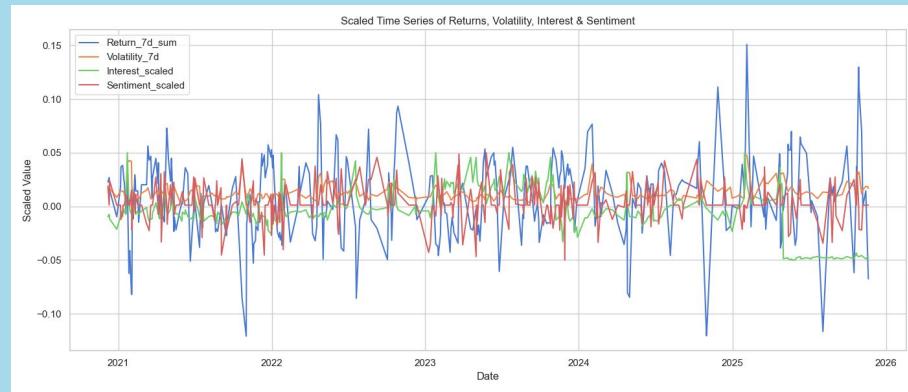
11/25/2025

Modeling IBM Stock Movements with Public Interest and Sentiments

Principles of Programming: Final Project Presentation
DSCI 510 – Fall 2025 – Dr. Alexey Tregubov
University of Southern California
Madeleine Willson



Project Overview



Goal: Predict daily stock returns

Motivation: Understand Market Behavior

Scope: Why IBM? Why 5 years?

Modeling: Binary Classification

Proxy for Interest: Search Volume

Proxy for Sentiment: Article Headlines

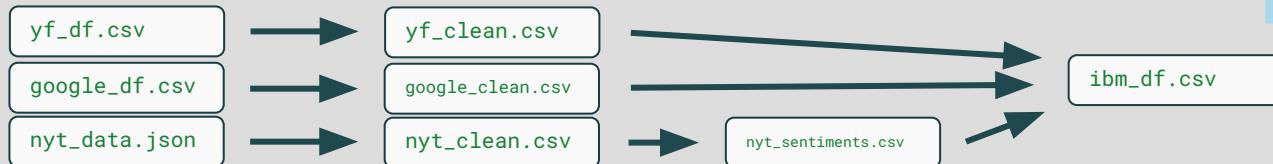
Data Sources



	DESCRIPTION	KEY FIELDS	RAW DATA SIZE	ACCESS METHOD
Yahoo Finance	Historical IBM Stock Prices	<ul style="list-style-type: none"> • Open • Close • Volume • Return 	~1250 rows	<ul style="list-style-type: none"> • yfinance library • Gives DataFrame
Google Trends	Search interest for "IBM" keyword	<ul style="list-style-type: none"> • Date • Interest 	~1825 rows	<ul style="list-style-type: none"> • pytrends library • Gives DataFrame
NYT Articles	Headlines mentioning "IBM" keyword	<ul style="list-style-type: none"> • Date • Abstract • Headline • URL 	~600 articles	<ul style="list-style-type: none"> • NYT Article Search API • Gives JSON

Summary of Results

Data Cleaning & Integration



- Aligned on a common date range, cleaned missing values
- Created lagged & rolling features for returns, volatility, interest, sentiment
- Sentiment scored using VADER on NYT headlines
- Ended with 14 features for 359 days

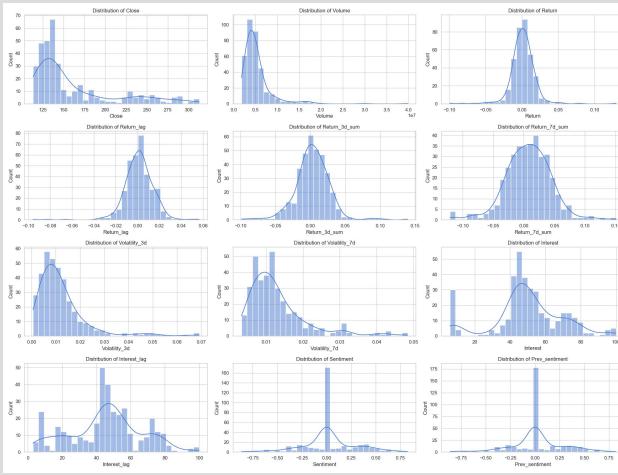
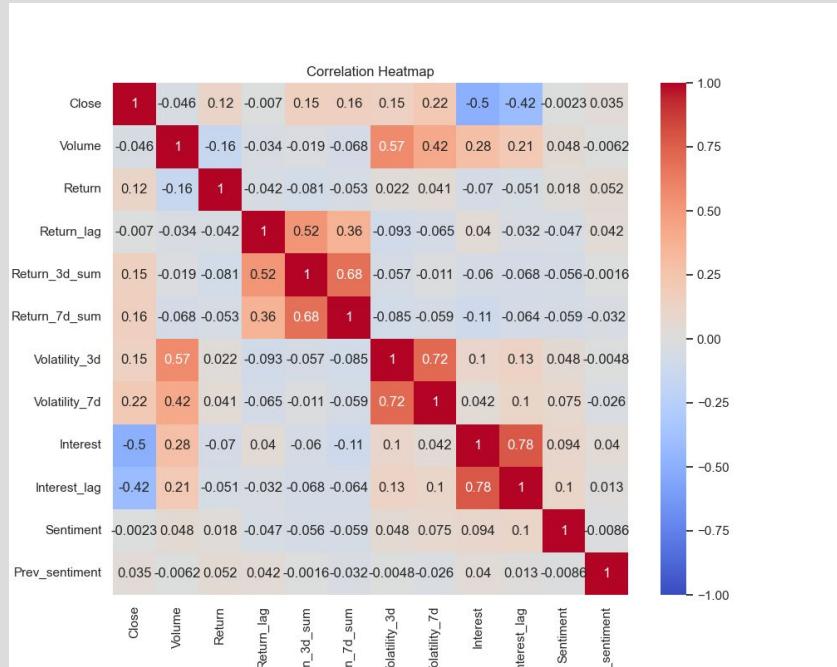
```
1   Date,Close,Volume,Return,Return_lag,Return_Su_Sun  
2   2020-11-23,114.8087921142578,5910318,,,,,,Monday  
3   2020-11-24,118.94837188720705,8109115,0.036056295
```

```
Date,Interest,Interest_lag  
2020-11-23,44,  
2020-11-24,51,44.0  
2020-11-25,42,51.0
```

```
1   Date,Headline  
2   2025-11-04,IBM to Cut Thousands of Workers Amid A.  
3   2025-04-28,IBM Plans to Invest $150 Billion Domest
```

Summary of Results

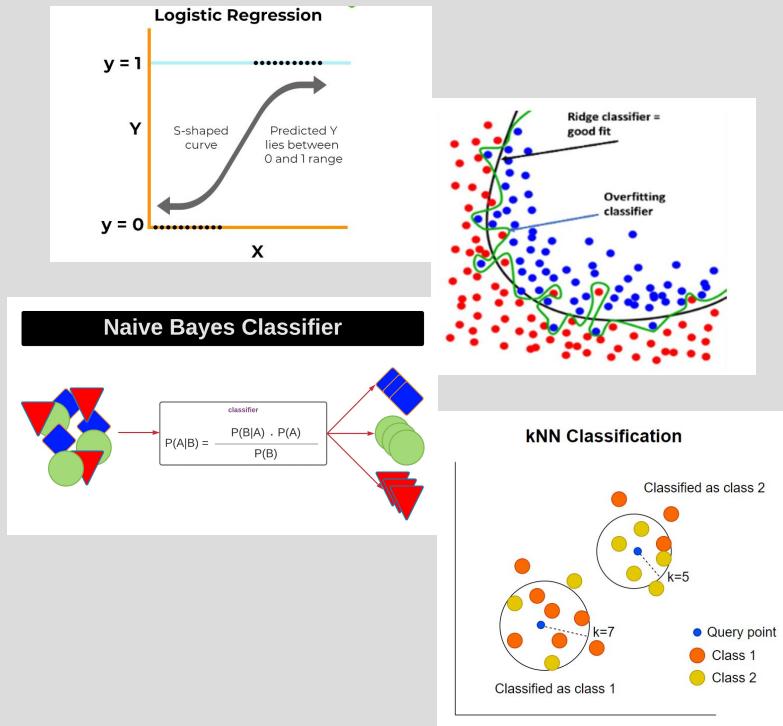
Exploratory Data Analysis



- Mean return (~0.001) positive
- Mean sentiment (~0.03) positive
- Mean return (~50.1) positive
- No missing values
- Distribution and correlation heatmaps to explore relationships among variables
- Day of the week & time series trends explored

Summary of Results

Modeling Approach

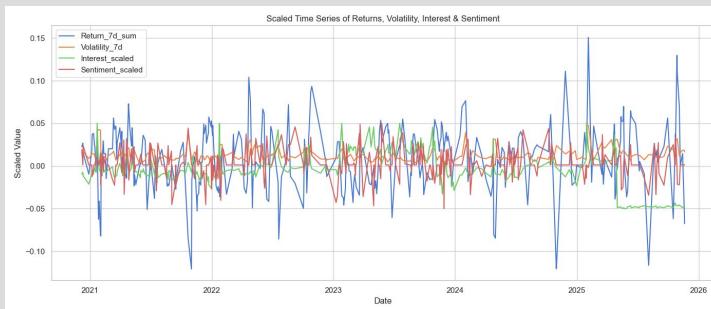


- ◆ Binary classification
80% train, 20% test
- ◆ Class Imbalance
Handled via weighted training & extra .5%
- ◆ Feature Selection
Lagged return, rolling returns(3&7), rolling volatility (3&7), lagged interest, previous sentiment
- ◆ Models Tested
Logistic Regression, Ridge Classifier, Naive Bayes, KNN

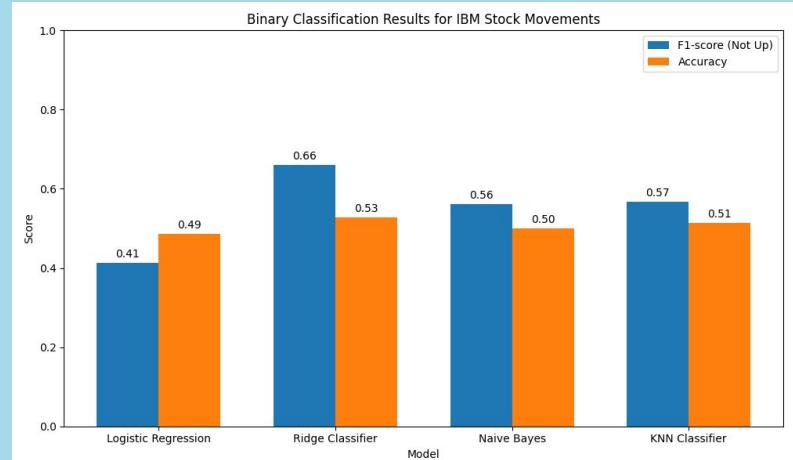
Summary of Results

Model Performance

- Evaluated with f1 score (minority class) & accuracy
- Logistic Regression and Ridge performed best
- Reasonable predictive power despite very noisy data



Logistic Regression:					Naive Bayes:				
f1-score (Stock not up): 0.4127					f1-score (Stock not up): 0.5610				
Preformance Metrics:					Preformance Metrics:				
	precision	recall	f1-score	support		precision	recall	f1-score	support
0	0.4815	0.3611	0.4127	36	0	0.5000	0.6389	0.5610	36
1	0.4889	0.6111	0.5432	36	1	0.5000	0.3611	0.4194	36
accuracy			0.4861	72	accuracy			0.5000	72
macro avg	0.4852	0.4861	0.4780	72	macro avg	0.5000	0.5000	0.4902	72
weighted avg	0.4852	0.4861	0.4780	72	weighted avg	0.5000	0.5000	0.4902	72



Ridge Classifier:					KNN Classifier:				
f1-score (Stock not up): 0.6600					f1-score (Stock not up): 0.5679				
Preformance Metrics:					Preformance Metrics:				
	precision	recall	f1-score	support		precision	recall	f1-score	support
0	0.5156	0.9167	0.6600	36	0	0.5111	0.6389	0.5679	36
1	0.6250	0.1389	0.2273	36	1	0.5185	0.3889	0.4444	36
accuracy			0.5278	72	accuracy			0.5139	72
macro avg	0.5703	0.5278	0.4436	72	macro avg	0.5148	0.5139	0.5062	72
weighted avg	0.5703	0.5278	0.4436	72	weighted avg	0.5148	0.5139	0.5062	72

Challenges

First time API user, and overly optimistic about my proxies.

01

API limits & date
window restrictions
complicated data
collection

02

Aligning datasets on
datetime variables
took a lot of time to
resolve reactively

03

Handling noisy,
small datasets to
prevent overfitting

04

Experimenting with
feature selection
and model tuning to
improve
performance

Thank you

Questions?