#PredictingTheDow

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Agenda

- Theories
- Problem
- Exploration
- Experimenting
- Refining
- Feature Selection/Importance
- Backtesting
- Conclusions and Future Work

Github Repo

https://github.com/ddm7018/-PredictingTheDow

Supporting Theory

It has long been theorized that the news affects the stock market.

September 11th - when the markets opened on 9/17, NYSE went down 680 points (7.1%) Lehman Brothers collapse led Dow closing 4.4% or 504 point down

Quants are already doing this - Hathaway effect

Correlating Financial Time Series with Micro-Blogging Activity

Can we predict the stock market movement from the news?

Efficient Market (Opposing) Theory

it is impossible to "beat the **market**" because stock **market efficiency** causes existing share prices to always incorporate and reflect all relevant information

Problem

Dataset: https://www.kaggle.com/aaron7sun/stocknews

Using /r/worldnews (Reddit) to predict whether the stock market(measured by Dow Jones Index) will go up or down

We are given Top 25 News Items of the day, along with

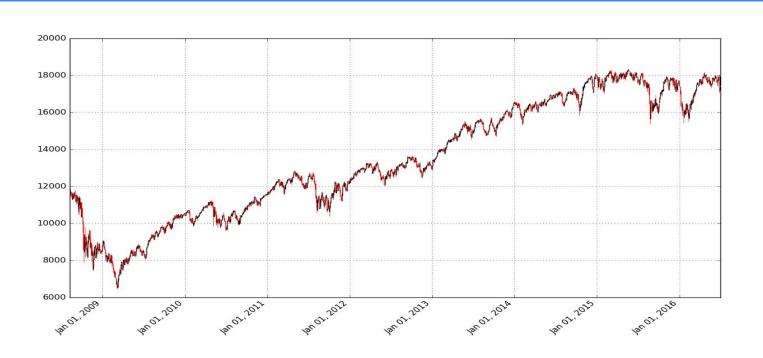
'Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close'

Predictor: 'Label' (1 if Open-Close >0 else 0)





Exploration Time...(candlestick issues)



Word Clouds





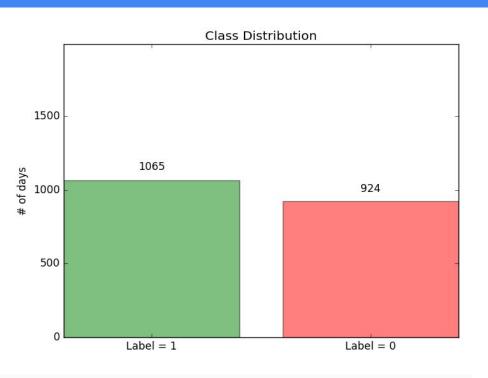
Exploration Time...(cont..)

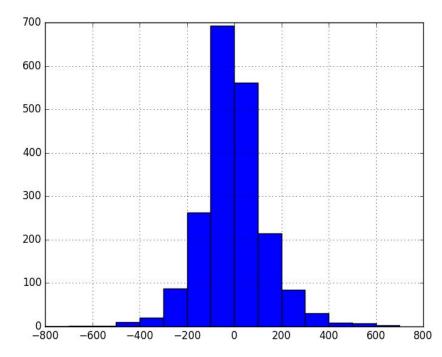
06/08/2008 -> 07/01/2016

Stock Market went up 53% during that time (the goal to beat)

Can we beat it?!

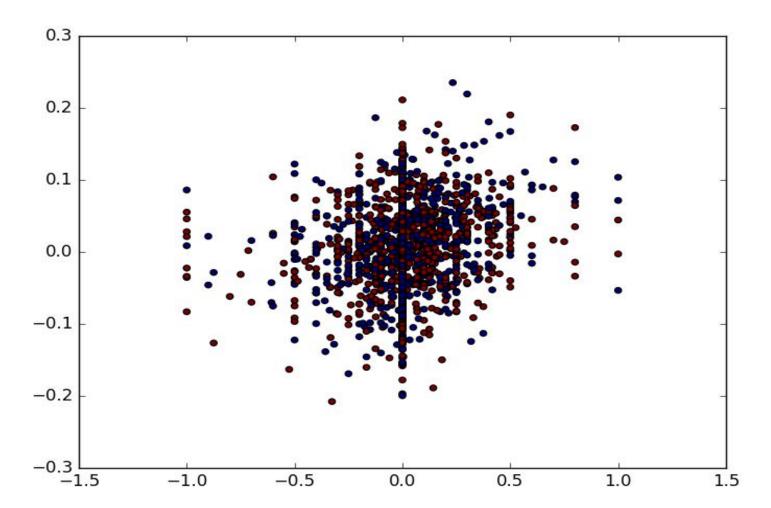
Distribution

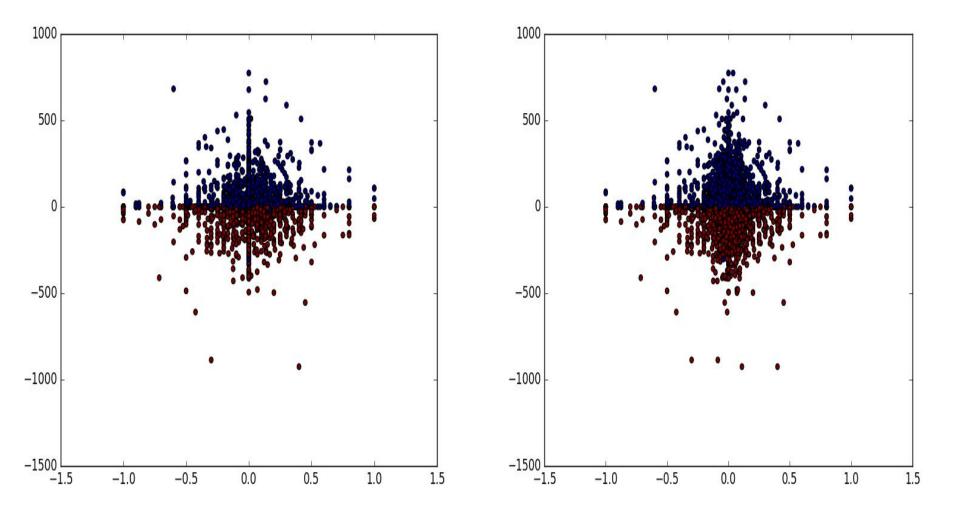


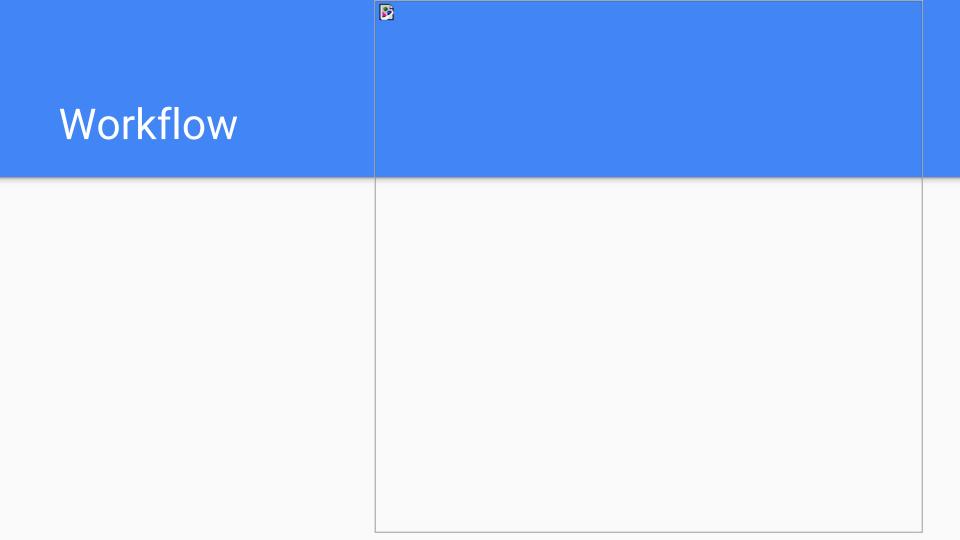


Distribution (cont.) - 100 Label









Classifiers

- KNN
- AdaBoost
- DecisionTree
- RandomForest
- LogisticRegression
- SVC
- ExtraTrees
- BernoulliNB

Vectors

- Count Vector
- Count Vector of n = 2
- Tdidf Vectorizer
- Tdidf Vectorizer n = 2

New Method - Ada Boosting

- short for "Adaptive Boosting", is a machine learning meta-algorithm formulated by Yoav Freund and Robert Schapire who won the Gödel Prize in 2003
- popular boosting technique which helps you combine multiple "weak classifiers" into a single "strong classifier". A weak classifier is simply a classifier that performs poorly, but performs better than random guessing

AdaBoosting in Scikit-learn

Other New Terms Used

- ExtraTreesClassifer Extremely Randomized Trees
- Truncated SVD Singular Value Decomposition

We ran into problems...

In our first or second checkpoint, we reported accuracies of 80% and better but..we split our training and testing data incorrectly..and some of our training data found its way into our test data as well.

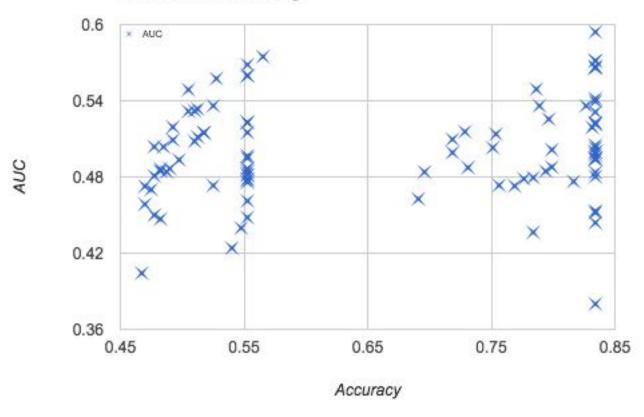
Calculated AUC wrong!

The data was quite messy and Count Vectors and Td-idfVectors required stopword removal and stemming

95% of words found when the stock market went down were also found in news articles when the stock market went up

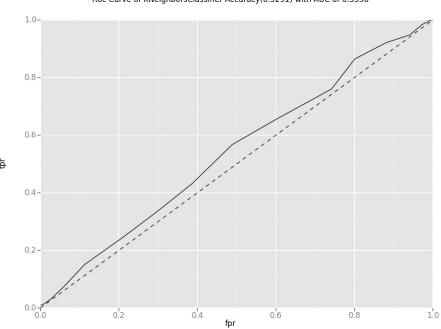
Minor foreign event did not seem to have any effect of the stock market

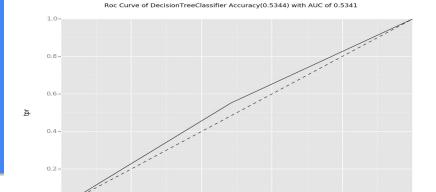
AUC vs. Accuracy



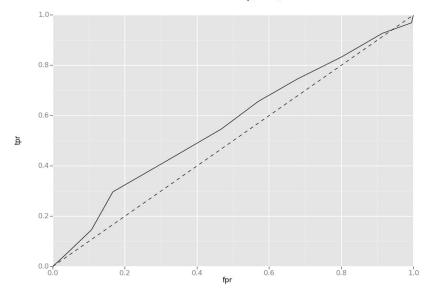
Our Best AUC Curves...







Roc Curve of AdaBoostClassifier Accuracy(0.5397) with AUC of 0.5632



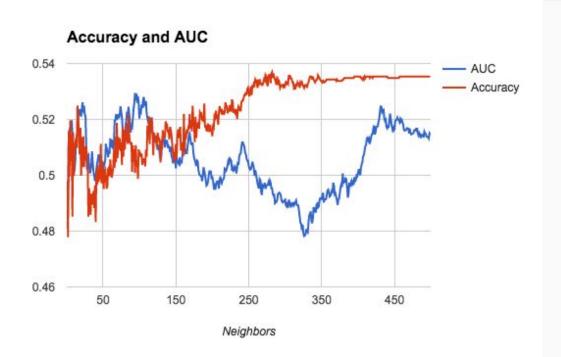
Cross Validation and Model Refining

Cross Validating brought most algorithms to have an accuracy of 53%

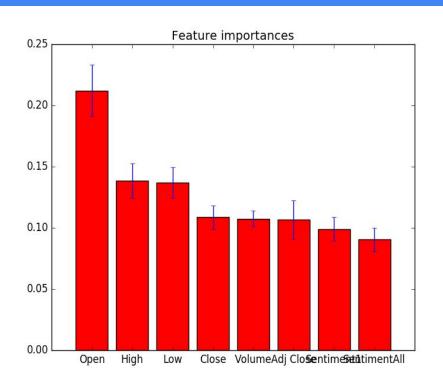
We concentrated on refining KNN. When we optimizing AUC, accuracy was neglected and vice-versa. We were not able to solidly get AUC above >.5 and accuracy above 53%

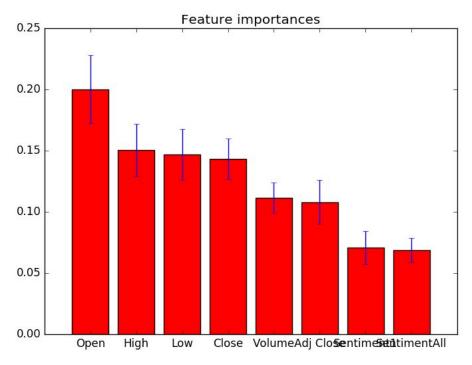
AUC < .6 is quite poor!

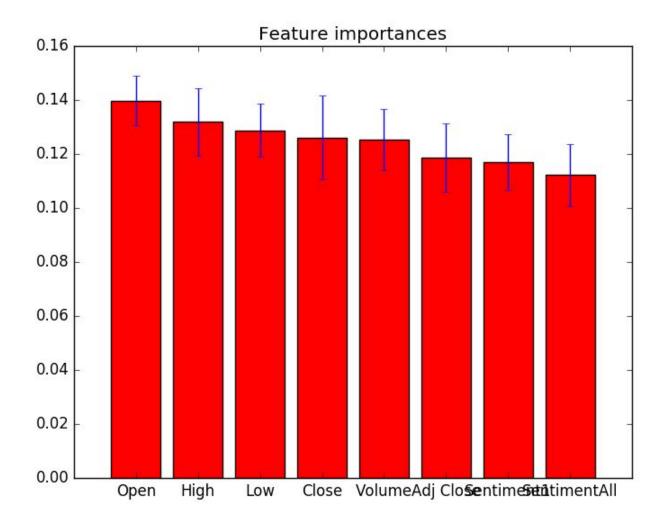
Refining KNN (k-fold k = 5)

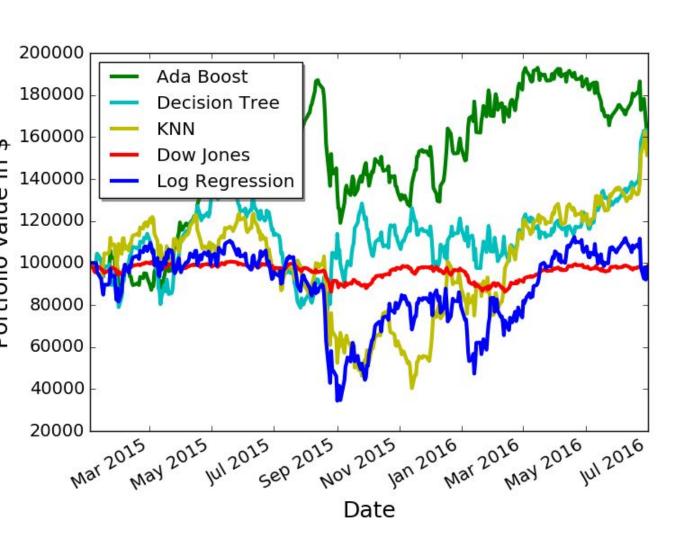


Feature Importance - ExtraTreesClassifier









AdaBoost - 62%

DT - 57.4%

KNN - 51.8%

LR -1.8 %

Future Work

Try different text data or more specific data (research with surprisingly good accuracy in this domain) Correlating Financial Time Series with Micro-Blogging Activity

More, different vectorization

Incorporate numeric data with text data

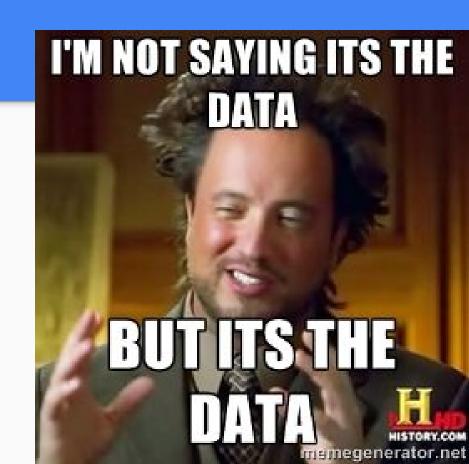
More backtesting (maybe with Tomorrow LAbel and 100 Label)

Stanford Named Entity Recognizer

YOU GOT 80% RECOGNITION RATE



WHY NOT 100%



Conclusions

Given the text data that we have, we can't accurately predict whether the stock market will go up or down

Dataset was created for Deep Learning course, maybe a deep learning approach would be beneficial

There's a lot a foreign news that doesn't impact a the stock market. What would happen if we tried a different text source?

What if we analyzed Apple Reddit News vs Apple Stock?