# #PredictingTheDow

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## 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

Can we predict the stock market movement from the news?

## **Efficient Market 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

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

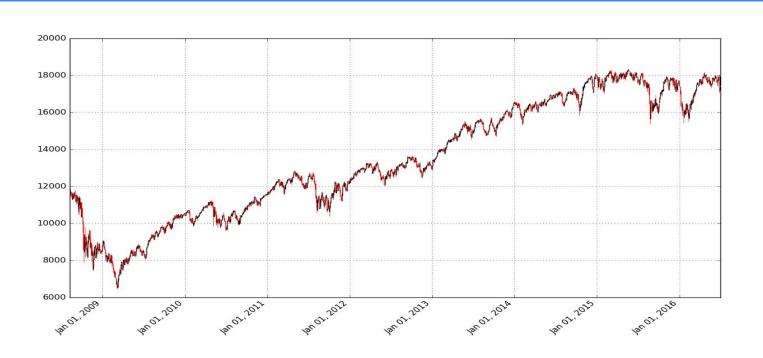
'Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close'

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





# Exploration Time...(candlestick issues)



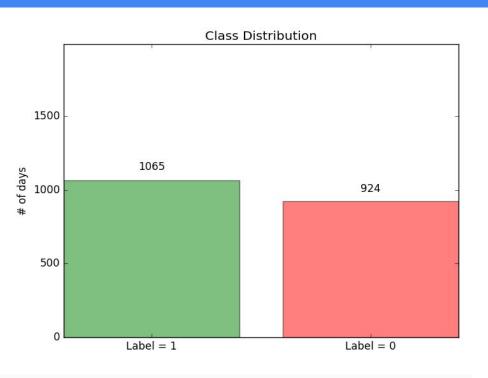
#### Exploration Time...(cont..)

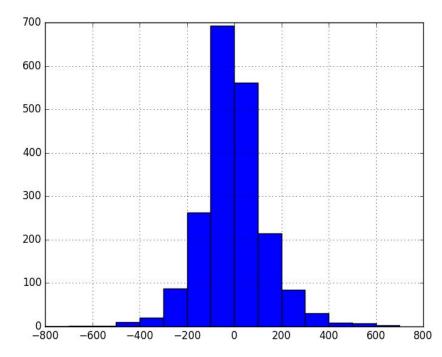
06/08/2008 -> 07/01/2016

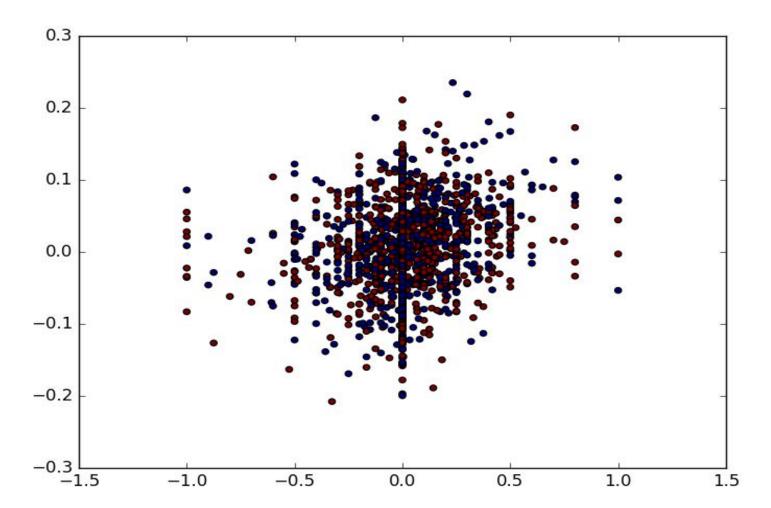
Stock Market went up 53% during that time (the goal to beat)

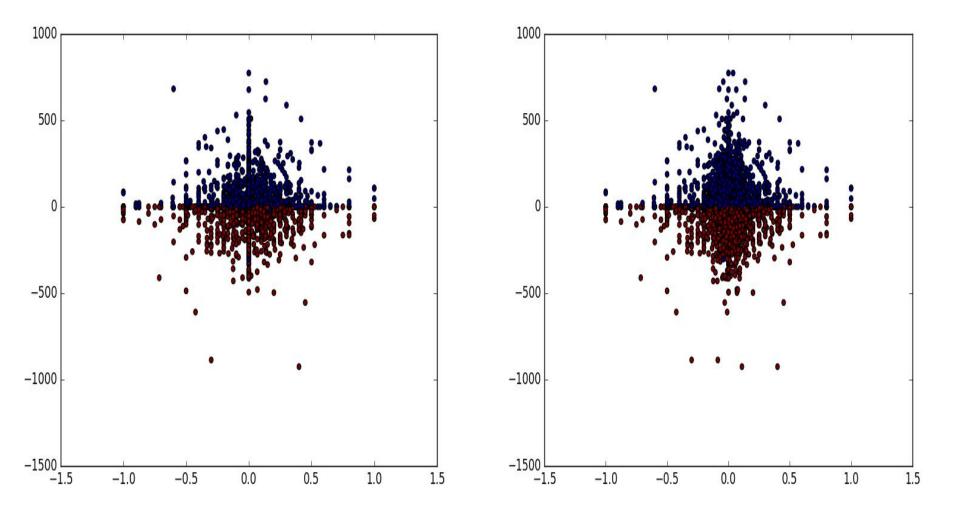
Can we beat it?!

#### Distribution









#### **Word Clouds**





#### We ran into problems...

In our first or second checkpoint, we reported accuracies of 80% and better

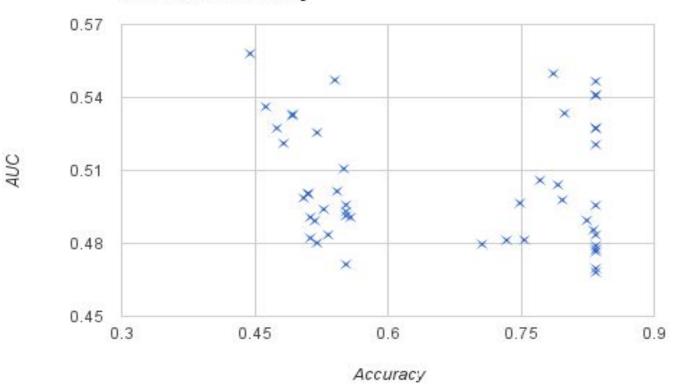
But..we split our training and testing data incorrectly..and some of our training data found its way into our test data as well.

The data was quite messy and CountVectors and Td-idfVectors required stemword removal and stemming

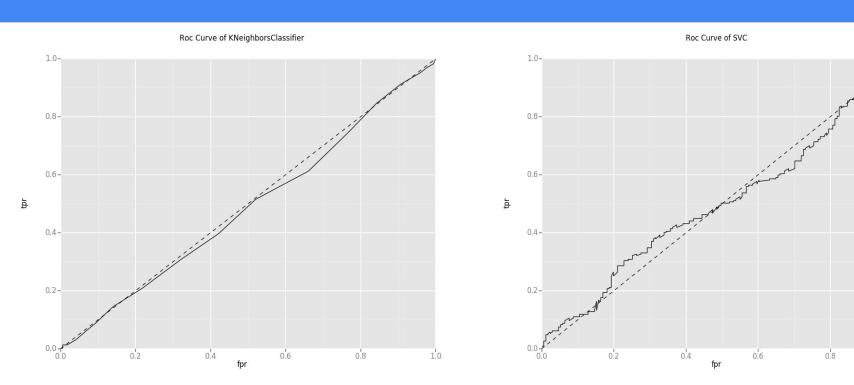
95% of words found when the stock market went down were also found in news articles when the stock market went up

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

#### **AUC vs Accuracy**



#### Our Best AUC Curves...



#### Cross Validation and Model Refining

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

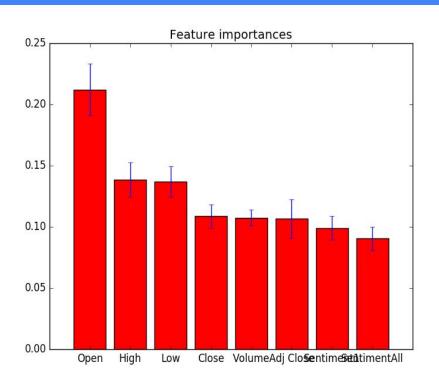
We concentrated on refining KNN. When we optimizing AUC, accuracy was neglected and vice-versa. We were not able to soldily get AUC above >.5 and accuracy above 53%

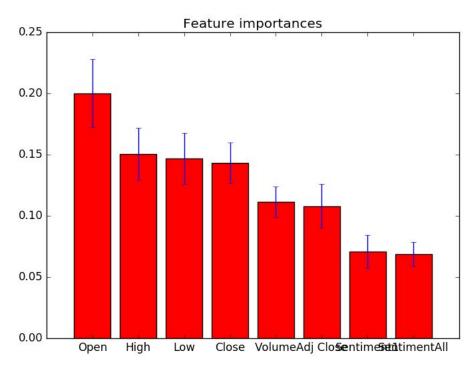
AUC < .6 is quite poor!

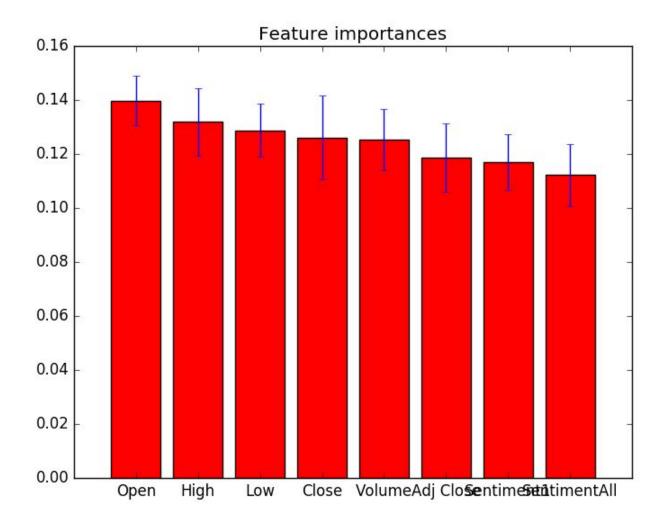
# Refining KNN

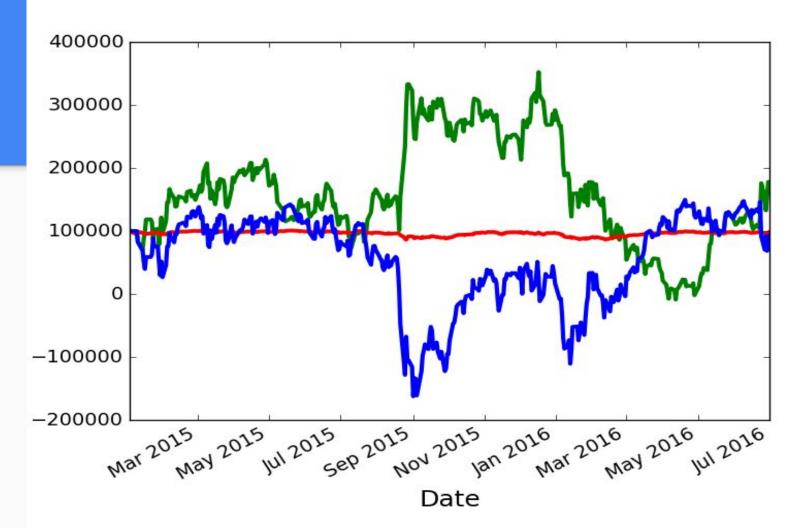
```
[4]+ Stopped
                            python -i refineKNN.py
Daniels-MacBook-Air:Project danielmooney$ python -i refineKNN.py
KNN using CountVector with neigh 238 had the best AUC of 0.527938221256an accuracy of 0.499953608247
[5]+ Stopped
                            python -i refineKNN.py
Daniels-MacBook-Air:Project danielmooney$ python -i refineKNN.py
KNN using Tdidf with neigh 4 had the best AUC of 0.517641765621an accuracy of 0.473829896907
>>>
[9]+ Stopped
                             python -i refineKNN.py
Daniels-MacBook-Air:Project danielmooney$ python -i refineKNN.py
KNN using CountVector with neigh 299 had the best AUC of 0.516656727079an accuracy of 0.532170103093
[10]+ Stopped
                             python -i refineKNN.py
Daniels-MacBook-Air:Project danielmooney$ python -i refineKNN.py
KNN using TfidfVectorizer with neigh 280 had the best AUC of 0.528955796047an accuracy of 0.525855670103
```

#### Feature Selection









#### **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

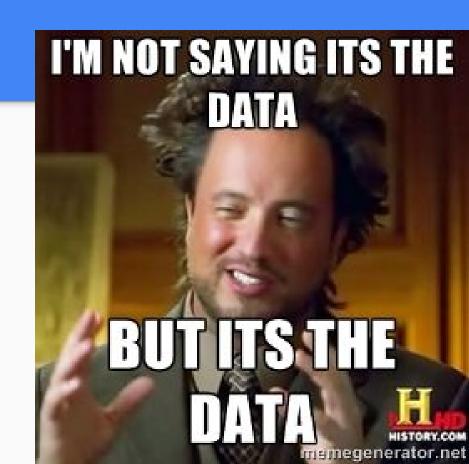
More backtesting

**Regression Analysis** 

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?