# Predicting the impact of President Trump's tweets on US stock market

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# Agenda

- 1. Problem Statement
- 2. Data collection & Preprocessing
- 3. Topic Modelling
- 4. Sentiment Analysis
- 5. Modelling & Result
- 6. Backtesting trading strategy
- 7. Conclusion

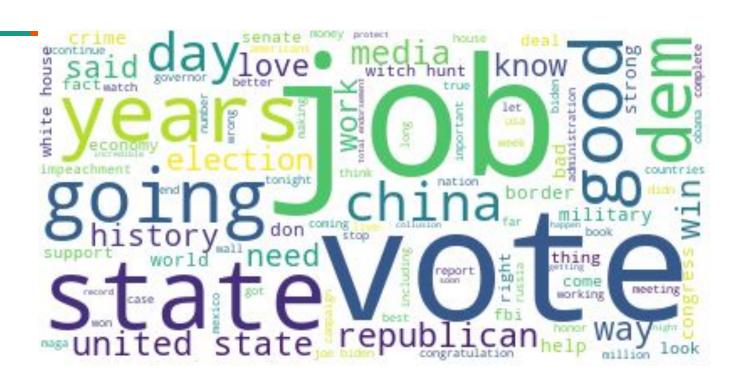
#### **Problem Statement**

- Stock market is always unpredictable but full of opportunities
- President Trump has 87.2 million followers on Twitter
- Binary classification model that predict President Trump's trade related tweets on it's impact (positive/ negative) on the S&P 500 index
- Evaluation metrics: Accuracy
- Using these predictions to create a profitable trading strategy



### **Data Collection & Preprocessing**

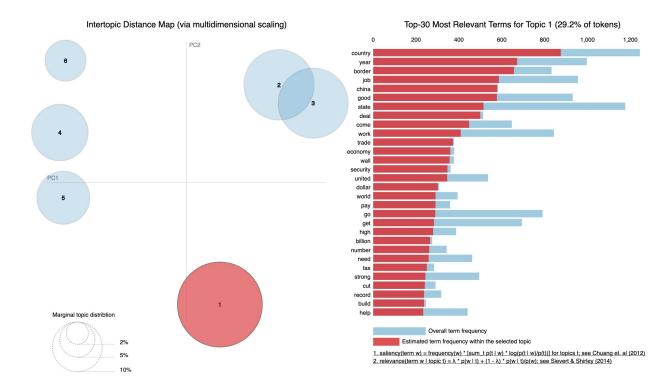
- 23,357 tweets were collected from 20 Jan 2017 till 08 Oct 2020
- Emojis, url links and non english characters were removed
- Retweets were dropped
- Tweets that have less than 3 word count were treated as noise
- Default stopwords added (thank, will, great, just, etc..)
- S&P 500 financial data collected for same period
- Tweets were labeled on the performance of the index



### **Topic Modelling**

- A statistical modelling for discovering abstract topics
- Latent Dirichlet Allocation (LDA) assumes each topic is a mixture over an underlying set of words, and each document is a mixture of over a set of topic probabilities
- 3 key parameters:
  - N\_components Number of topics
  - Alpha higher alpha documents are assumed to be made up of more topics and result in more specific topic distribution per document
  - Beta with high beta, topics are assumed to made of up most of the words and result in a more specific word distribution per topic
- Topic that capture words relevant to trade and economy

N\_components= 6
alpha=0.2
beta=0.8





Topic\_1: 0.817 Topic\_2: 0.015 Topic\_3: 0.015 Topic\_4: 0.122 Topic\_5: 0.015 Topic\_6: 0.015

Dominant Topic: Topic\_1 (Trade/Economy)

#### **Sentiment Analysis**

- Identify and extract opinions with a given text
- VADER (Valence Aware Dictionary and Sentiment Reasoner)
- Returns a positive, neutral, negative, and compound sentiment score
- Compound score range from -1 (extremely negative) to +1 (extremely positive)
- Tweets with substantial positive and negative sentiment were isolated

Positive Sentiment: Compound Score > 0.7

Negative Sentiment: Compound Score < -0.7



Dominant Topic: Trade/Economy

Neg: 0.053 Neu: 0.639 Pos: 0.308 **Compound: 0.8481** 

# Modelling

| 1. | Logistic | Regression |
|----|----------|------------|
|    | 0        |            |

#### 2. MultinomialNB

3. Random Forest Classifier

4. XGBoost Classifier

5. SVC

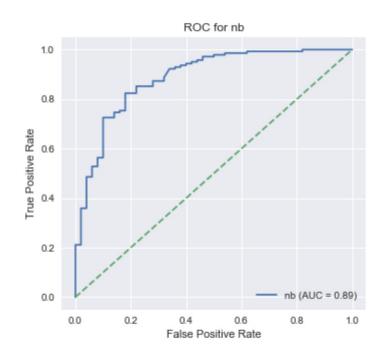
Train Test 0.983 0.829

0.925 0.833

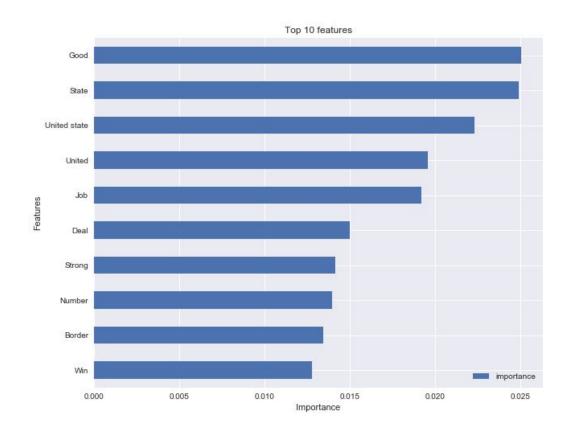
0.956 0.843

0.976 0.849

1.000 0.864



# Top word features



# **Backtesting Trading Strategy**

- Focusing only on buy signals (where model predicted 1, ignore if 0)
- Buy at tomorrow's open and sell at tomorrow's closing price
- Benchmark to compare is Buy and Hold strategy

72%

VS

52%



#### Conclusion

- We are all going to be rich??
- Trending bull market will generate more buy signals
- Trump tweets coincide with FED economic data release
- Backtest is a test on the past
- Gather other data such as real time news articles, other high ranking politicians tweets/press release
- Real time market data
- Trump tweets does have certain relationship with market returns
- Further studies on NLP and market returns

#### THE END



If Biden somehow manages to win this election, you won't see me again, I'll simply leave the country!



Bi den