

# **Predicting Stock Return with Text Data**

#### **RECENT NEWS**

Recent breaking news all over the world

#### Africa

 'This is huge': Locust swarms in Africa are worst in decades





#### **Australia**

 Australia fires: The fires are still burning. And they'll be burning for months to come





#### China

 The coronavirus from China is new, and that makes everything dicier





US

 Kobe Bryant, daughter killed in copter crash, 7 others dead

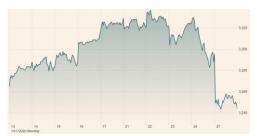




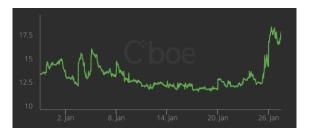
Review of the current market

#### Financial Market

• S&P 500 Index (01/13/2020—01/27/2020):



• VIX Index (01/02/2020 – 01/27/2020):





#### **Theories**

- Keynes (1936): Animal Spirits
- Eugene Fama (1970): Efficient Market Hypothesis
- Shiller (2002): Feedback Loop & Attention Cascade

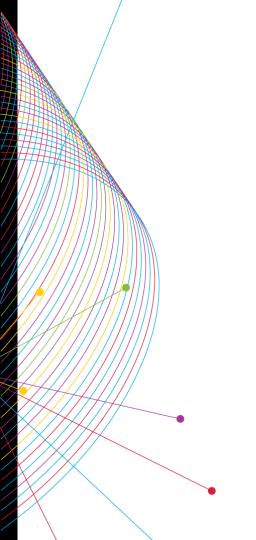
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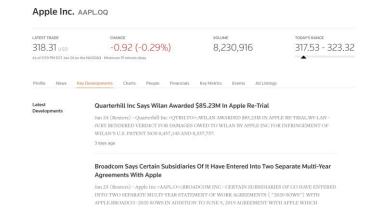
# PART 1. DATA

#### **DATA MINING**

Scratching all the news of the stocks in NYSE, NASDAQ and ASE

## Tickers

- Download all the tickers from NYSE, NASDAQ, and ASE;
- Search for each ticker's website on Thomson Reuters;





- Use Python Selenium and Scrapy to scratch 200,000+ news articles;
- Utilize NLTK and Beautifulsoup package in Python to conduct text preprocessing;







#### **TEXT PRE-PROCESSING**

Using NLTK package to clean the news we got by the web crawler



The raw news (example):

German biotech group Morphosys seeks European partner for debut drugNovartis said on Thursday it will pay 95 million euros (\$110.49 million) to Galapagos and MorphoSys to license a prospective medicine targeting the skin condition atopic dermatitis, as the Swiss drugmaker strengthens its foothold in immunology.

Converting text to lowercase:

german biotech group morphosys seeks european partner for debut drugnovartis said on thursday it will pay 95 million euros (\$110.49 million) to galapagos and morphosys to license a prospective medicine targeting the skin condition atopic dermatitis, as the swiss drugmaker strengthens its foothold in immunology.

Regular expression:

german biotech group morphosys seeks european partner for debut drugnovartis said on thursday it will pay million euros million to galapagos and morphosys to license a prospective medicine targeting the skin condition atopic dermatitis as the swiss drugmaker strengthens its foothold in immunology

Tokenizing:

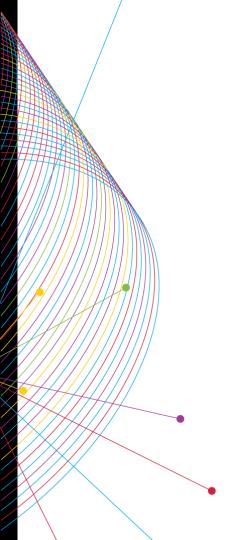
['german', 'biotech', 'group', 'morphosys', 'seeks', 'europea n', 'partner', 'for', 'debut', 'drugnovartis', 'said', 'on', 't hursday', 'it', 'will', 'pay', 'million', 'euros', 'million', 'to', 'galapagos', 'and', 'morphosys', 'to', 'license', 'a', 'p rospective', 'medicine', 'targeting', 'the', 'skin', 'conditio n', 'atopic', 'dermatitis', 'as', 'the', 'swiss', 'drugmaker', 'strengthens', 'its', 'foothold', 'in', 'immunology']

Deleting Stop words:

['german', 'biotech', 'group', 'morphosys', 'seeks', 'europea n', 'partner', 'debut', 'drugnovartis', 'said', 'thursday', 'pa y', 'million', 'euros', 'million', 'galapagos', 'morphosys', 'l icense', 'prospective', 'medicine', 'targeting', 'skin', 'condi tion', 'atopic', 'dermatitis', 'swiss', 'drugmaker', 'strengthe ns', 'foothold', 'immunology']

• Stemming & Lemmatizing:

['german', 'biotech', 'group', 'morphosi', 'seek', 'european', 'partner', 'debut', 'drugnovarti', 'said', 'thursday', 'pay', 'million', 'euro', 'million', 'galapago', 'morphosi', 'licens', 'prospect', 'medicin', 'target', 'skin', 'condit', 'atop', 'der mat', 'swiss', 'drugmak', 'strengthen', 'foothold', 'immunolo g']



# **PART 2. METHODOLOGY**

#### **MODEL SETUP**

Create the word matrix with the sentiment-charged words for the different news

#### Word Matrix

- Record the word (or phrase) counts of the certain article in a vector;
- Combine those vectors to form the word matrix.

$$D = \begin{bmatrix} d_1 & \dots & d_n \end{bmatrix}$$

$$\begin{bmatrix} d_{1,1} & \cdots & d_{1,m} \\ \vdots & \ddots & \vdots \\ d_{n,1} & \cdots & d_{n,m} \end{bmatrix}$$

#### Sentiment-charged words

 Calculate the frequency with which word j co-occurs with a positive return;

$$f_j = \frac{\text{\# articles including word } j \text{ AND having } \operatorname{sgn}(y) = 1}{\text{\# articles including word } j}$$

 Set the threshold values to filter the sentiment-charged words;

$$\widehat{S} = \{j : f_j \ge 1/2 + \alpha_+, \text{ or } f_j \le 1/2 - \alpha_-\} \cap \{j : k_j \ge \kappa\}.$$



Based on the Multinomial Distribution, estimate the crucial estimators, O+ and O-

#### Preparation

 Distribution assumption: sentimentcharged word counts are generated by a mixture multinomial:

$$d_{i,[S]} \sim \text{Multinomial}\left(s_i, \ p_i O_+ + (1 - p_i) O_-\right)$$

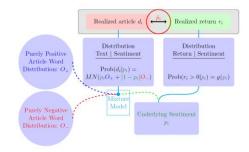
 P value setup: for each news, we assign a sentiment value p based on the rank of the relevant stock's return:

$$\widehat{p}_i = \frac{\text{rank of } y_i \text{ in } \{y_l\}_{l=1}^n}{n}$$



#### **Distribution Estimation**

 By setting up the sentiment values for each news, to estimate the estimators, O+ and O- (two topic vectors):



F: a vector of frequency

$$F = \frac{1}{2}(O_+ + O_-)$$

T: a vector of tone

$$T = \frac{1}{2}(O_+ - O_-)$$



With the result of O+ and O-, add a penalty term to optimize the calculation process



Calculation process:

$$\mathbb{E}\widetilde{d}_{i,[S]} = \mathbb{E}\frac{d_{i,[S]}}{s_i} = p_i O_+ + (1 - p_i) O_-$$

$$\mathbb{E}\widetilde{D}' = OW$$
, where  $W = \begin{bmatrix} p_1 & \cdots & p_n \\ 1 - p_1 & \cdots & 1 - p_n \end{bmatrix}$ , and  $\widetilde{D} = [\widetilde{d}_1, \widetilde{d}_2, \dots, \widetilde{d}_n]'$ .

$$\widetilde{O} = \widehat{D}\widehat{W}'(\widehat{W}\widehat{W}')^{-1}$$

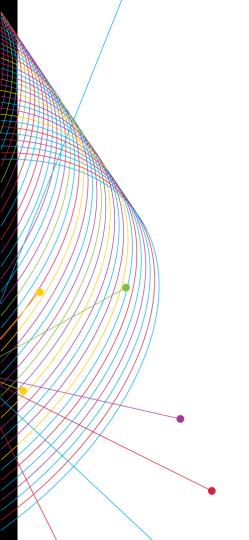
• Example:

	0+	О-
repurchase	0.030759	0.009670
fell	0.020964	0.039305



 Combining the penalty term, the estimated value, p, could be shrunk to 0.5, which avoid the extreme result:

$$\lambda \log(p_i(1-p_i))$$

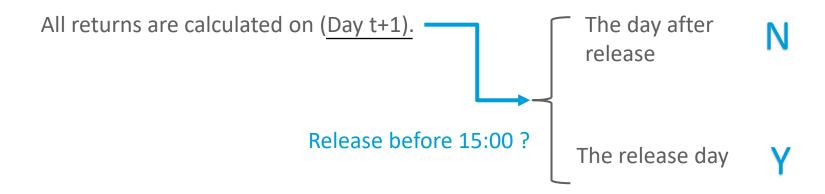


# **PART 3. EMPIRICAL ANALYSIS**

#### MATCH TEXT DATA WITH RETURN

Get the excessive return of day t+1

Excessive Return = True Return — S&P 500 return



#### **NEWS DATA**

#### Example of news data collected from Thomson Reuters

	ticker	head	description	txt	datadate	ret
0	XXII	BRIEF-22nd Century ?CEO Henry Sicignano To Dis	* 22ND CENTURY - CO'S CEO HENRY SICIGNANO III	March 19 (Reuters) - 22Nd Century Group Inc: *	20180319	0.014204
1	XXII	BRIEF-22ND CENTURY GROUP TO DISCONTINUE U.S. S	* 22ND CENTURY GROUP INC - WILL DISCONTINUE U	Nov 22 (Reuters) - 22nd Century Group Inc: * 2	20171122	0.017992
2	XXII	BRIEF-22nd century receives guidance from FDA	* 22Nd century receives guidance from FDA on p	June 22 (Reuters) - 22nd Century Group Inc : *	20170622	0.022555
3	XXII	BRIEF-22nd Century entered into warrant exerci	* 22nd Century - entered into warrant exercise	June 19 (Reuters) - 22nd Century Group Inc: *	20170619	-0.024741
4	XXII	BRIEF-22nd Century, Dent Neurosciences Researc	* 22nd century and Dent Neurosciences Research	June 8 (Reuters) - 22nd Century Group Inc: * 2	20170608	-0.000267



Threshold: times = 400 posi\_proportion = 0.54 nega\_proportion = 0.46











Construct sentiment word vectors for each news document

#### News 1:

"Apple just releases a milestone product that beats all the competencies and reduce this year's loss."

#### News 2:

"Microsoft has to face a drop due to their loss on new products this year. Stock price may drop by 5%"

	Milestone	Beat	Loss	Drop
News 1	1	1	1	0
News 2	0	0	1	2

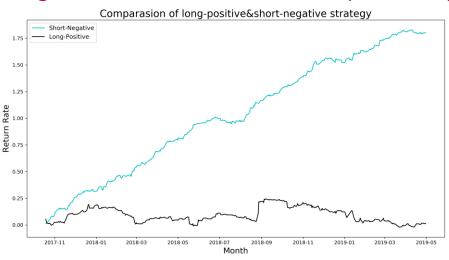


#### PREDICTION AND RESULTS

Investor are more sensitive about negative news.

#### Larger Effect of the Negative News

- Short-Negative: Short all the related stocks with the predicted negative scores
- Long-Positive: Long all the related stocks with the predicted positive scores

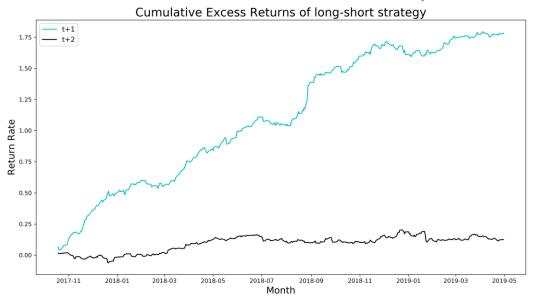


#### PREDICTION AND RESULTS

Based on the predicted absorption speed of the news, market is fairly efficient.

## $\mathcal{N}$ The timeliness of news

- **T+1**: calculated with **t+1** returns of the related stocks (Correlation: **0.043**)
- **T+2**: calculated with **t+2** returns of the related stocks (Correlation: **0.002**)





- Figure out a better method to decide the threshold for word subsets selection.
- Limitations of bag-of-words: Meaning (Context, semantic, tone)
- Potential application in the other areas

# Thank you!

