# **Sentiment Analysis**

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### **Sentiment Analysis Overview**

#### **Definition of Sentiment Analysis**

Sentiment Analysis is the use of Natural Language Processing techniques to systematically identify and study useful information such as emotion and preferences.

#### **Importance of Sentiment Analysis**







Centralized criteria



Provide timely analysis

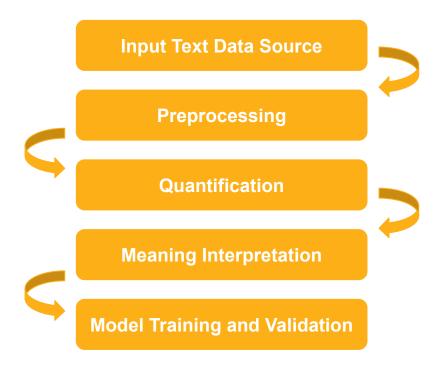
Build a pre-trained model that can be applied to a wide variety of textual datasets

- Predicting voter preferences on candidates based on political surveys
- Product improvement using public opinion

**Motivation** 

### **Natural Language Processing**

Steps and Processes





### **Introduction and Summary of Methodologies**

Three Methodologies

#### **TextBlob**

- Python library
- Based on Naive Bayes
- Determines the polarity of phrases ranging from negative to positive

### Neural Network

- Based on a paper done by Liu
  Bing and Mingqin Hu
- Identified the sentiment of opinion sentences
- Used WordNet: a lexical database of semantic relations between words for over 200 languages

#### TF-IDF

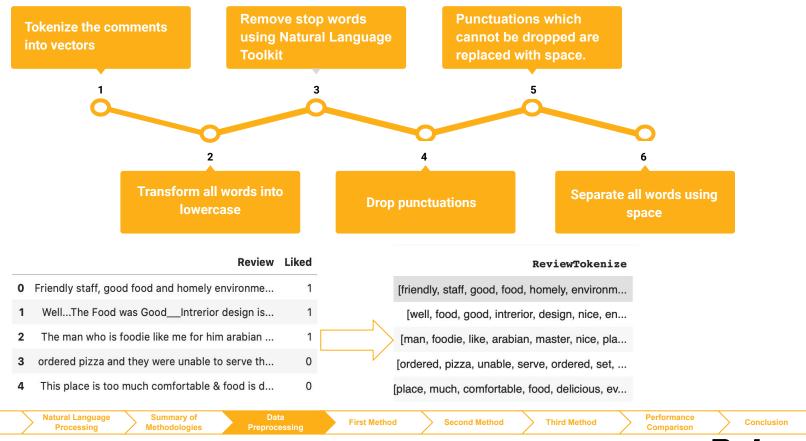
- Converting text data into vector
- Counts the frequency of words to determine sentiment of word (frequency within and across document)
- Two types of algorithm: Naive
  Bayes and SVM



### **Data Preprocessing**

Sentiment

**Analysis** 





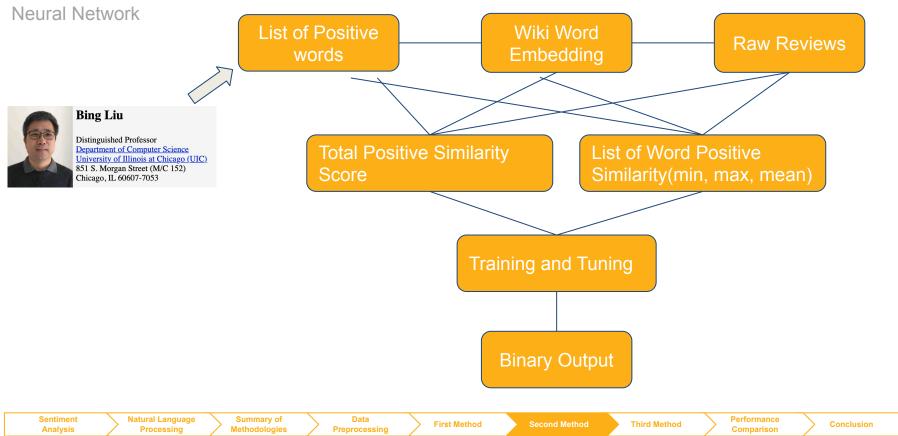
TextBlob

**TextBlob** uses a Movies Reviews dataset in which reviews have already been labelled as positive or negative. The data is trained on a **Naive Bayes Classifier**.

Passed the tokens to a **TextBlob sentiment classifier** which classifies the reviews as positive, negative or neutral by assigning it a polarity between -1.0 to 1.0. If the polarity > 0, the review is classified as positive, and if the polarity < 0, the review is classified as negative.

Data Type	Accuracy
Validation Data	66.75%
Test (New Data)	68.82%







#### **Neural Network**

Data Type	Accuracy
Validation Data	47%
Test (New Data)	54%

#### **Reasons for Failure**

- Too much preprocessing, **lost meaning** of the original text
- The research is based on social media data from 2004, which is too old



TF-IDF with Naive Bayes and Support Vector Machine

Term Frequency - Inverse Document Frequency (TF-IDF) Term Frequency looks how often each word appears in the document Inverse Document Frequency is the log to the all document divided by the document words appears

Since TF-IDF is a vectorization method, we need to use classification model to categorize the vectorized data points.

#### **Naive Bayes**

Туре	Accuracy Score	
Validation data	81.42	
New data	75.4	

#### **Support Vector Machine**

Kernel	SVM Accuracy Score (Validation)	SVM Accuracy Score (New Data)	
Linear	81.75	72.06	
Polynomial	79.44	64.50	
rbf	82.27	67.20	
sigmoid	79.98	69.30	

Performance

### **Performance Comparison**

Accuracy

Data Type	Accuracy Scores			
	TextBlob (Naive Bayes)	TF-IDF (Naive Bayes)	TF-IDF (SVM)	Neural Network
Validation Data	66.75%	81.42%	82.27%	54.24%
Test (New) Data	68.82%	73.40%	72.06%	51.20%



### Conclusion

The sentiment analysis is one of the most commonly used NLP to help us understand public opinions towards certain matter. Based on the experiments, there is no "the best" algorithm for text mining.

#### Recommendation

According to the performance results, we recommend TF-IDF with both Naive Bayes and SVM

### **Next Step**

- Cross-validation to split dataset
- Customized kernel for SVM
- Train the model based on more datasets across all industries
- Neural Network may perform better if resources are allowed





## Thank you!

**Any Questions?** 

