# Capstone 2 - Final Report

Conducting NLP on Amazon Store Review Data

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### Project Aim and Background

- Goal was to develop Machine learning models using Natural Language Processing (NLP)
- Used Amazon store review data to predict review sentiment
  - Dataset from "Electronics"



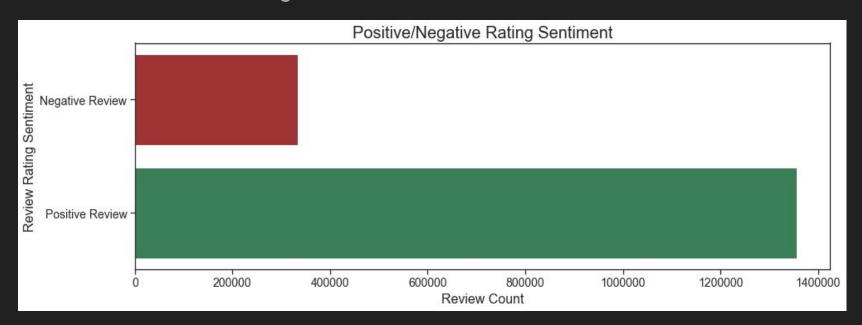
### **Initial Findings**

 The first step I took in examining the data was to look at a simple count plot of reviews broken out by their respective 5-star rating.



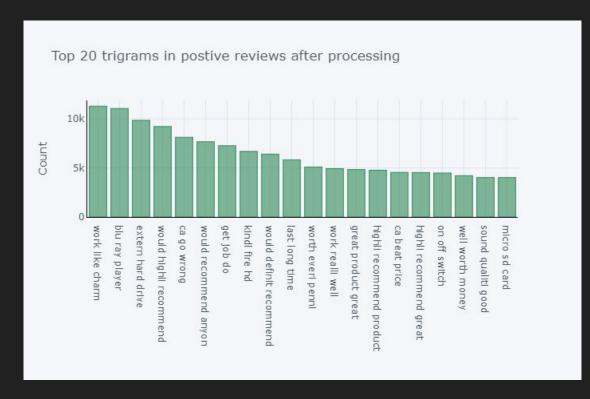
#### **Review Sentiment**

 I created a new binary target feature, which represents 1,2 and 3 star reviews as 0 or "Negative" and 4 and 5 star reviews as 1 or "Positive".

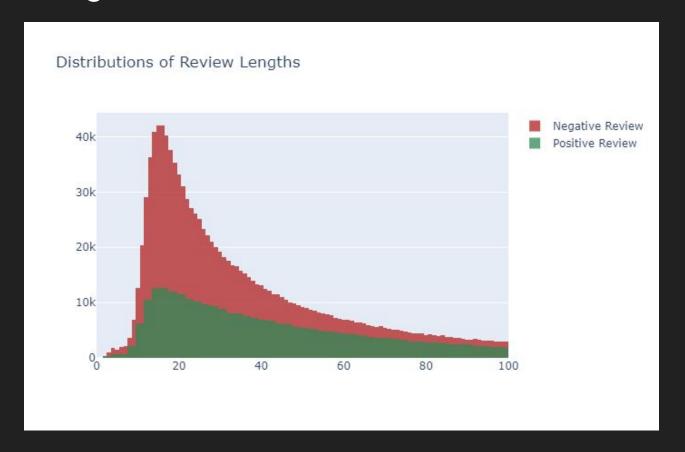


## Text Feature Engineering

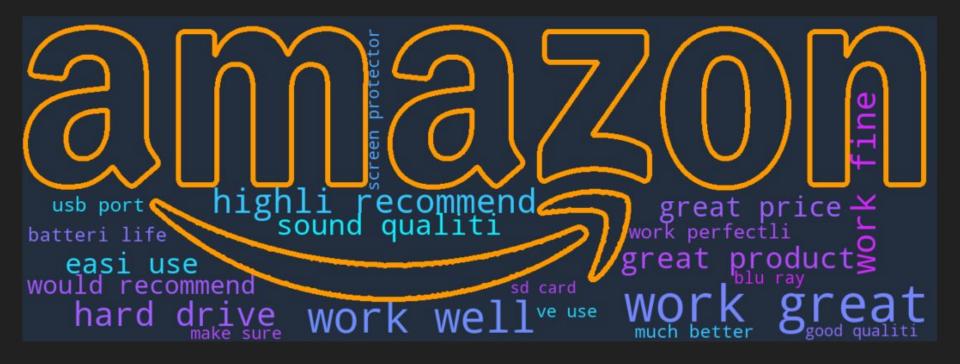
- 1. Remove Stop Words
- 2. Tokenization
  - a. CountVectorizer()
- 3. Lemmatization
- 4. Stemming
- 5. N-Gram Features



# Review Length



### **Word Cloud**



Based on to 20 Bigrams

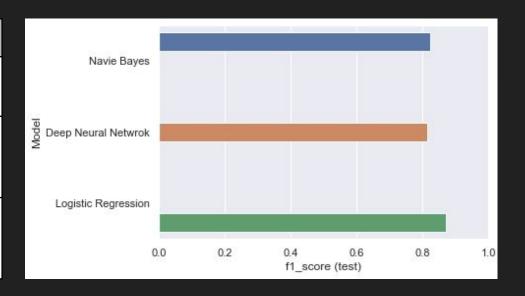
# Model Preprocessing

- 1. Bootstrapping Samples
- 2. Shuffling Dataset
- 3. Text Feature Vectorization
- 4. Split Training and Test Set



# Machine Learning

Model	F1 - Score	Wall Time
Naive Bayes	0.8225	1min 26s
Deep Neural Network	0.8134	~ 24 hrs (for full data set)
Logistic Regression	0.8721	14 min 46s



### Conclusion

- Based on the results, it seems like Logistic Regression may be the best classifier for this particular type of review data.
- Not only did it have a much better F1-Score than both the Naive Bayes model and DNN, it took a reasonable time to fit the model.
- However, the results for the DNN are based on only a 3% sample of the entire dataset.

## Considerations for Future Development

- Moving forward I would like to be able to train the DNN on all the review data, but since the dataset is so large I will likely need a GPU to accelerate the training process. It's quite possible that the DNN may perform better after training on all of the data.
- Additionally, I would like to develop models utilizing the numerical features within this dataset.