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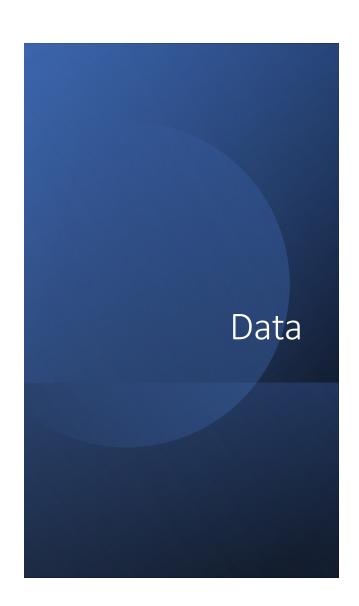


Background

 Customers buy apple products everyday. They also hop on twitter to voice their opinions about products everyday on Twittter. Using Natural Language Processing (NLP), we can determine what are the most common words being sent and with what tone. And Identify the best model for future predictions.



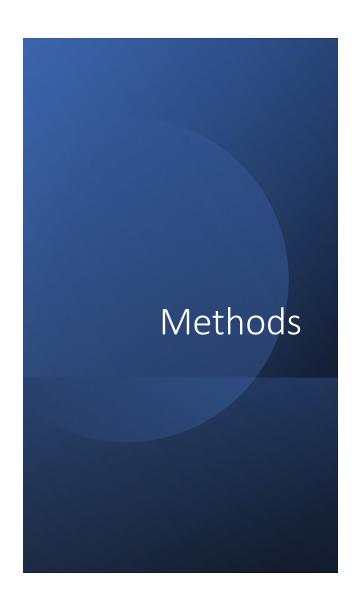
- What positive things are customers saying about Apple Products or Apple Company?
- What negative things are customers saying about Apple Products or Apple Company?



- Using the tweet product dataset. Dataset has over 9000 tweets, and 3 columns.
- Contains data including
 - Tweets
 - Focus of the tweet
 - Sentiment



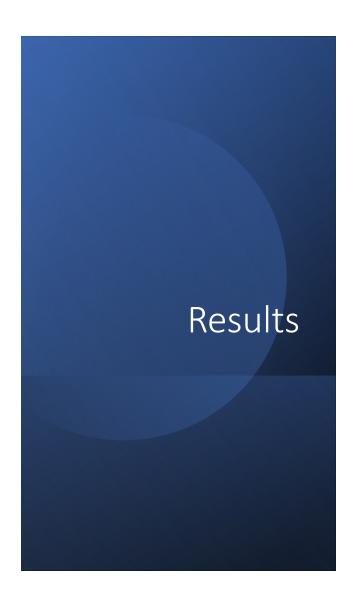
- Removal of special characters (@,#, etc.)
- Applying NLP Techniques such as Stemming, Lemmatization, and Tokenization.
- Removal of small words (less than 3 characters in length)
- Removing stop words



Build machine learning models with Word Vectorizers to predict future sentiments (using CountVectorizer and Term Frequency - Inverse Document Frequency (TFIDF)

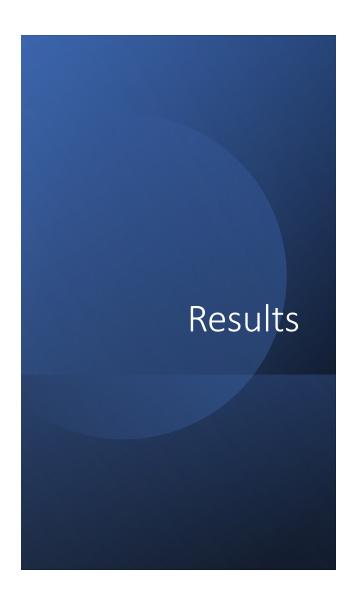
Developed several models to evaluate performance

- Precision correct predictions vs total predictions
- Recall Correct predictions vs actual positive predictions
- F1 score harmonic mean of Precision and Recall Accuracy

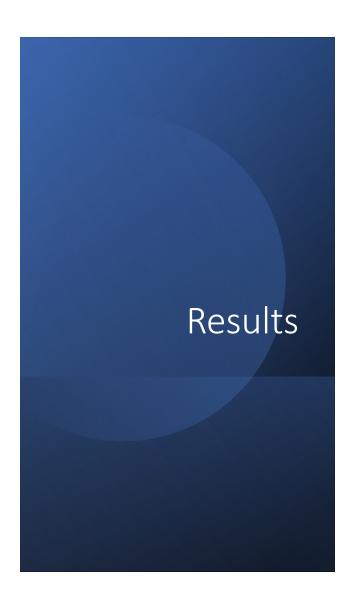


WordCloud for positive Emotion Sentiment

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without read open population application a
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WordCloud for Negative Emotion Sentiment



WordCloud for all Apple Sentiments

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app market expert size and a state of the size of the
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Summary of Findings / Conclusion

- Most positive words were associated with events.
- Most negative words associated with the product or negative words (such as battery)
- The TFIDF Vectorizer is the best for predicting future sentiments.
 - TFIDF Vectorizer outperformed Precision, Recall, F1 score against CountVectorizer

