

# Analyzing Consumer Sentiment using Twitter

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# Overview

This notebook aims to analyze Twitter data for sentiment and build a model that can best predict Twitter sentiment.

- Run Natural Language Processing to discover details about twitter user sentiment
- Build machine learning models to predict future sentiment

# Outline

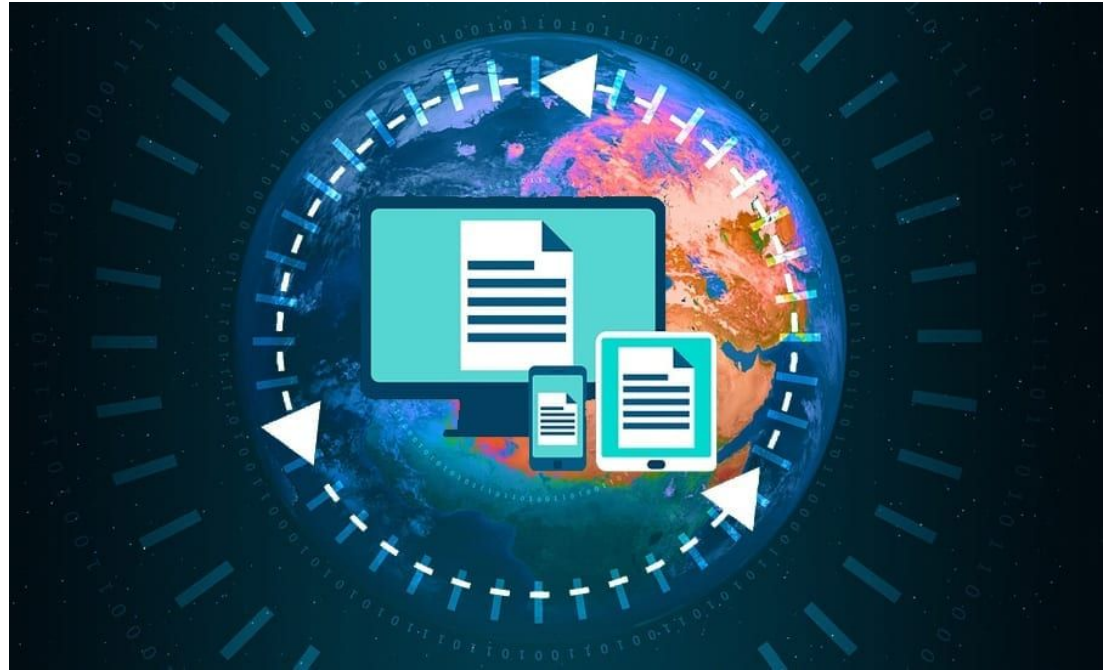
- Business Problem
- Data
- Methods Used
- NLP Analysis
- Conclusions
- Recommendations
- Next Steps

# Business Problem

Tech news company Engadget need to analyze trends to provide appropriate coverage on their content platform.

Mining sentiment on Twitter can help entities quickly gauge user opinion.

Furthermore, having a deep understanding of general consensus will help the recommendation process.



# Data

Comparing consumer  
sentiment between Apple  
Google products

9093 samples in total

3291 usable samples



# Methods Used

## Phase 1

- Exploratory Data Analysis
- Natural Language Processing
- Word Clouds

## Phase 2

- Class Imbalances
- Scikit Learn
- Keras

# NLP Analysis - Apple

# Apple Words



## Apple Words without Products



# NLP Analysis - Apple

## Positive Word Cloud



## Negative Word Cloud





# NLP Analysis - Google

# Google Words



## Google Words without Products



# NLP Analysis - Google

Positive Word Cloud



Negative Word Cloud



# Conclusions

# NLP Findings

## Apple

- We see much excitement for a popup store & its location downtown, which sells the latest products.
- Unfortunately, we can see that some apple users are experiencing issues with their battery life.
- Some users also do not approve of Apple's new product design. Also, it seems some people are calling Apple fascist.

# NLP Findings

## Google

- There is much buzz for Google Circle (a social network) in this dataset and Google Maps.
- We also see much mention of Marissa Mayer, a prolific businesswoman.
- Some users are experiencing interface issues in the same breath.

# NLP Findings

## **What can we gain from this information?**

- We can gain that though there was a bunch of positivity in our dataset of tweets.
- We can perceive a level of mixed reception since the negatives show discontent with the launch of products, new features & new designs of various products.

# Conclusions

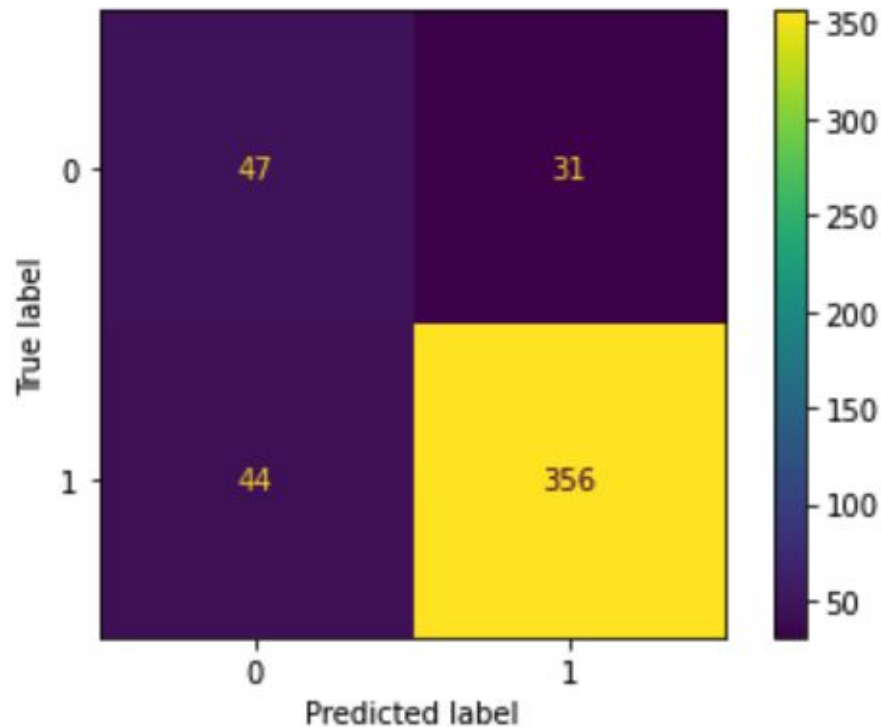
## **The best model is a Keras Model with Parameter Tuning.**

- The confusion matrix reports a highest correct classification of negative sentiment.
- The Scikit-Learn models all performed well on the accuracy, but most of their accuracy scores don't perform better than either Keras model.
  - The Linear SVC model has 86.82% accuracy, beating the tuned Keras model.
- The Logistic Regression model is the second-best performing model for predicting negative sentiment, with 44 correct predictions.

# Output

Interpretation of churn with a Tuned SVM:

- **47 True Negatives:** Tweets with negative sentiment
- **31 False Positives:** Predicted tweets with positive sentiment, but they're negative
- **44 False Negatives:** Predicted tweets with negative sentiment, but they're positive
- **356 True Positives:** Tweets with positive sentiment





# Recommendations

## Usage for our model:

- Use this model to predict Twitter sentiment for future events, or even with a live API
- Use this model to predict if negative tweets are about Apple or Google
  - You can do this by splitting the original dataframe only to include negative tweets and have the products be the target column
- Build a web app using techniques in this notebook to predict twitter sentiment

# Recommendations

- To further increase model performance, we could implement a grid search on all models to discover if higher model performance is probable.
- Utilize a Live API to collect tweets and store in a database for future activities

## Next Steps

- Engadget can publish various news articles targeted to their audience about the different findings this notebook uncovers
  - Apple articles targeted at Apple enthusiasts
  - Google articles targeted at Google enthusiasts
  - Negative or positive press about this particular SXSW as a whole
- Run similar analysis on future events or tech releases. Thus providing timely updates to enthusiasts & avid readers.

Questions?