



# Airline Sentiment Analysis

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# Problem

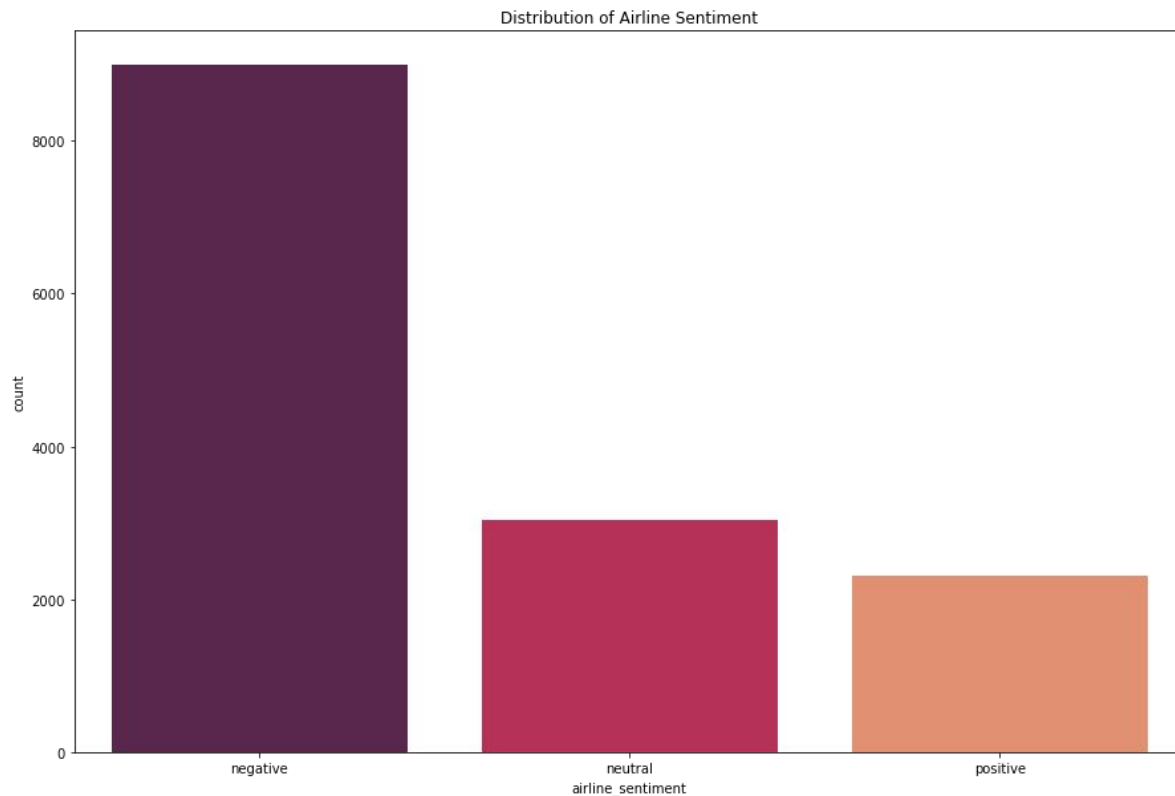
- ❖ As businesses change, so do their customers sentiments.
- ❖ Companies need to track their customer sentiment.
- ❖ Create a sentiment analysis for airline companies.
- ❖ Identify business solutions for airline companies

# Data Overview and Cleaning

- ❖ The dataset is from kaggle.
- ❖ The data was scraped from Twitter between Feb 23, 2015 and Feb 22, 2015 and classified as negative, positive, or neutral.
- ❖ Mainly focusing on the following columns: “airline”, “airline\_sentiment”, “negativereason”, “text”

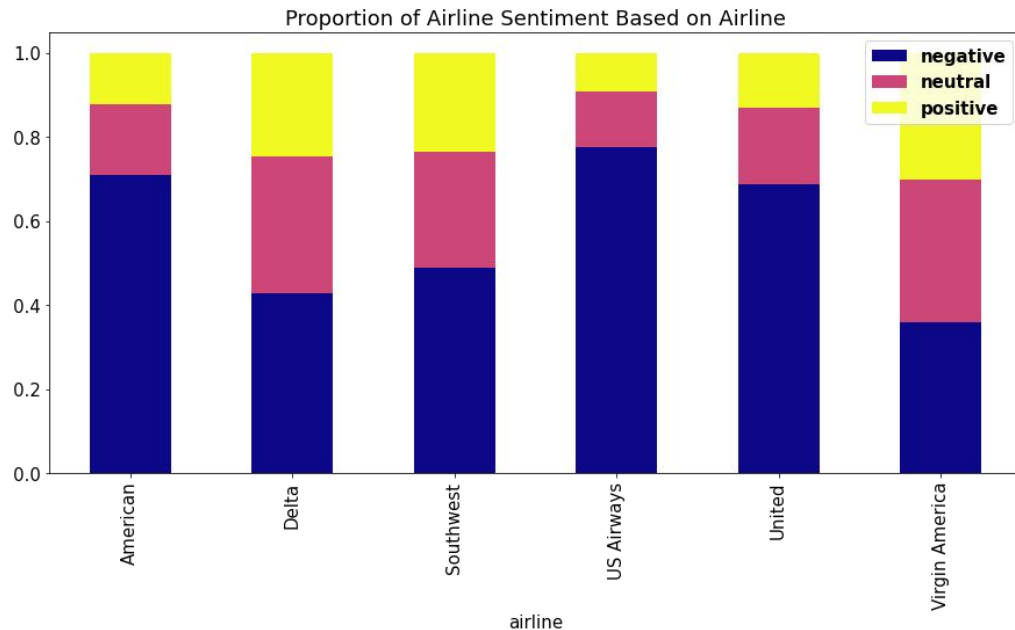
# Analysis: Distribution of Airline Sentiment

- ❖ Most of the sentiments are negative.
- ❖ This is an imbalanced dataset.



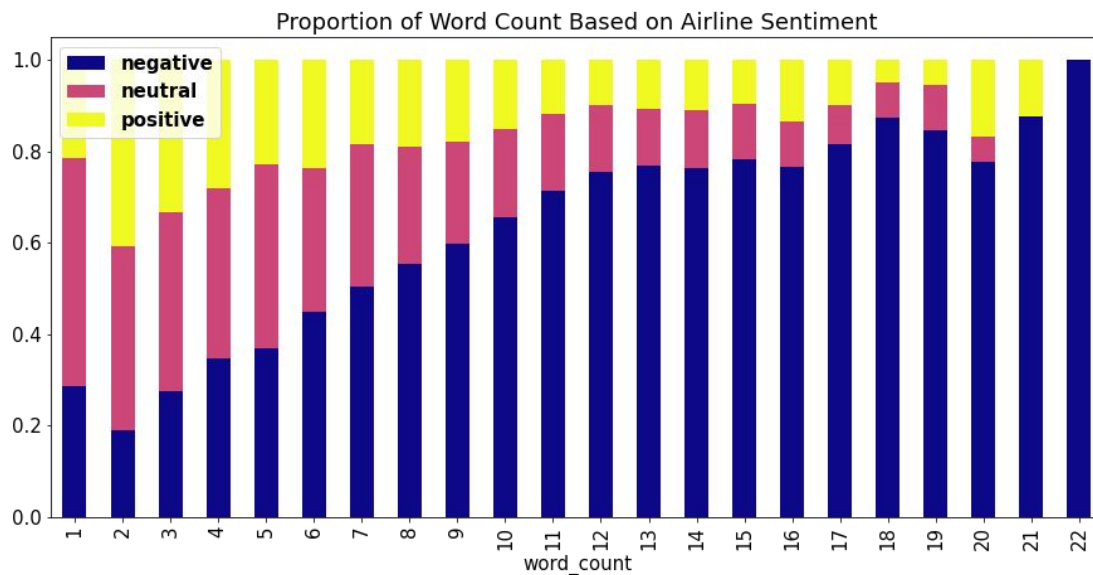
# Analysis: Proportion of Airline Sentiment

- ❖ US Airways has the most proportion of negative sentiment
- ❖ Virgin America has the least amount of positive sentiment, but it only accounts for 3.4% of the data



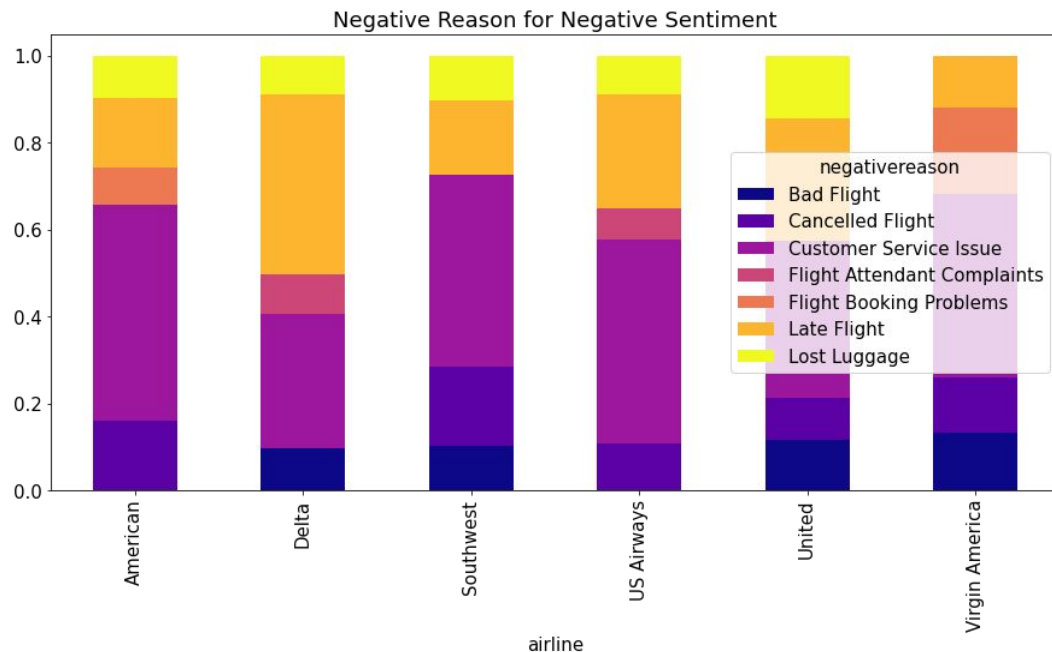
# Proportion of Word Count

- ❖ The negative sentiment tweets have more words in their text.
- ❖ The tweets with a positive or neutral sentiment have the least amount of words per tweet.



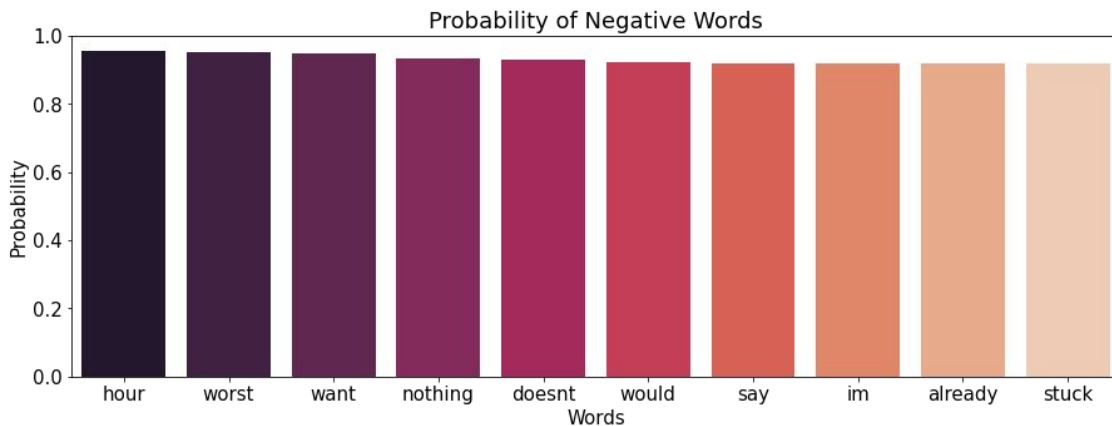
# Negative Reason for Negative Sentiment

- ❖ This is top 5 reasons for negative sentiment
- ❖ Customer service issues and late flights are the top issues across airlines.



# Predictive Words: United

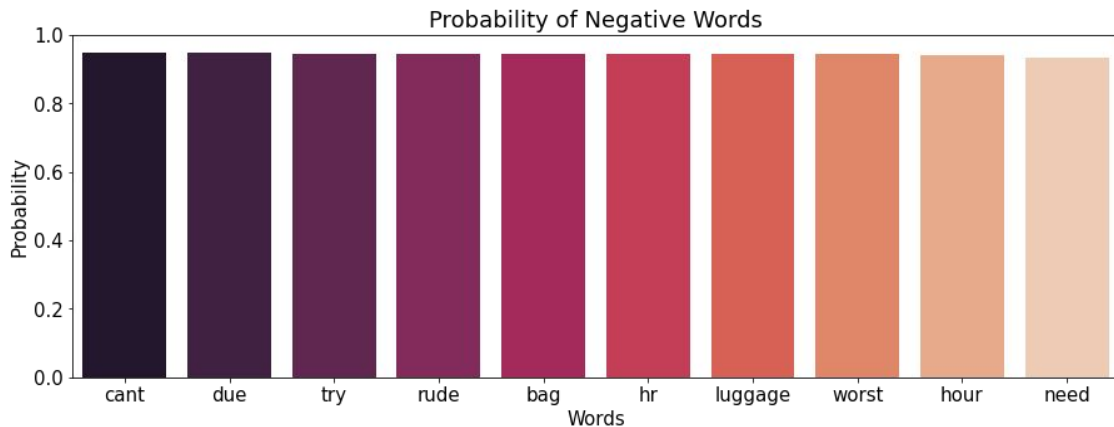
- ❖ Late and delayed flights
- ❖ Stuck for several hours
- ❖ Couldn't reach customer service
- ❖ Lost luggage and had a hard time recovering them
- ❖ “@united I sure did. I had to drive a total of 3 hours to get my own bag. I'd like to explain that debacle but no one wants to talk to me.”





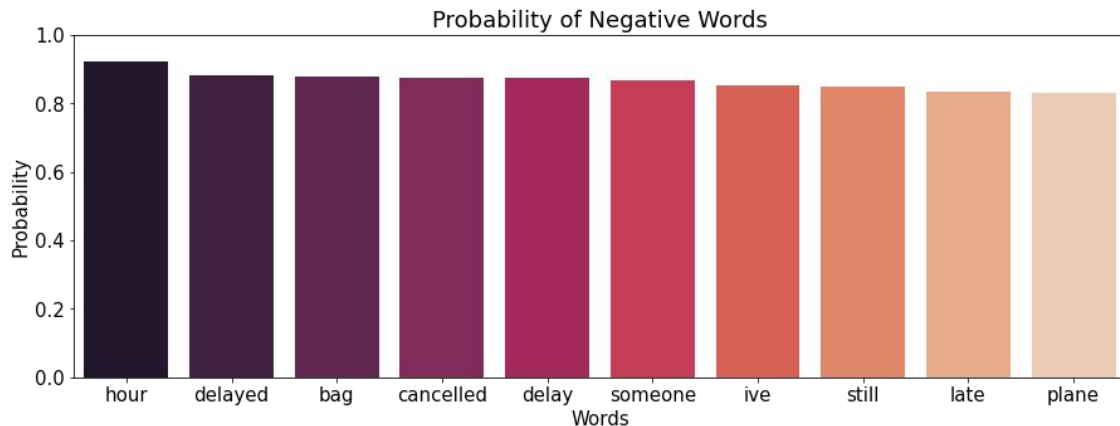
# Predictive Words: American

- ❖ Late flights & customer service issues
- ❖ Booking and re-scheduling issues
- ❖ Rude flight attendants, and lost luggage
- ❖ “@AmericanAir im tryin to book a flight but cant get ahold of anyone!”
- ❖ “@AmericanAir Right. But more than two hours Late Flight, and it seems due to poor communication, which sounded like it was annoying on-plane staff”
- ❖ “@AmericanAir extremely upset that your baggage handlers decide to go in my luggage and take my belongings”



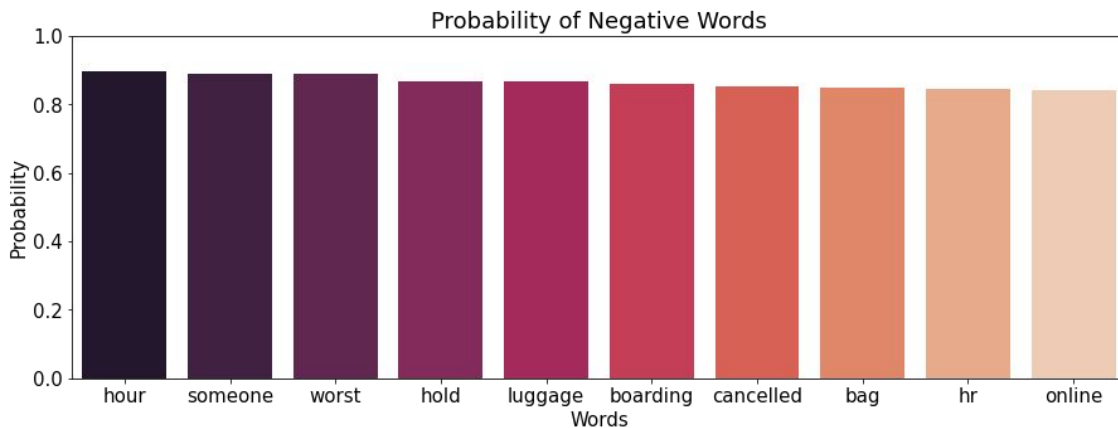
# Predictive Words: Delta

- ❖ Delayed and canceled flights and customer service issues
- ❖ Trouble contacting customer service and had to call several times
- ❖ “@JetBlue I had to call back five times to get someone on the phone who knew what they were doing. By that time my getaway went up by \$200.”
- ❖ “@JetBlue what is the deal with flt 460 today? Departure keeps changing. When is it going why is it so Late Flight?”
- ❖ “@JetBlue great job getting flight 28 in 10 minutes early. Too bad we're at 50 minutes and counting waiting for our bags.”



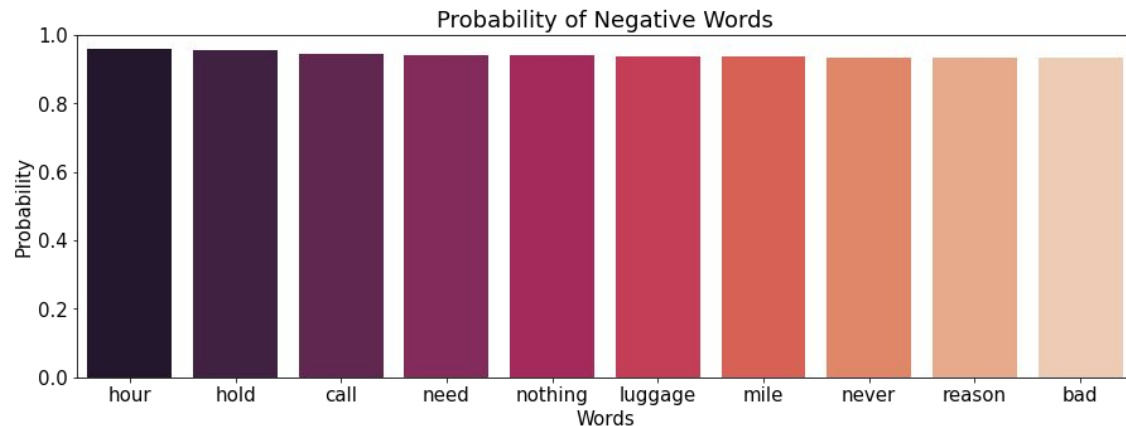
# Predictive Words: Southwest

- ❖ Customer service issues
- ❖ If customers have to change their reservations, get boarding passes, etc, they have to call customer service instead of doing it online
- ❖ Customers are also experiencing delayed and late flights and lost and damaged luggage
- ❖ “@SouthwestAir Why can we no longer change trips with a companion online? Been doing it for years, now get message can't be done online?”
- ❖ “@SouthwestAir can you have someone call me back? I have been on hold two times today for over 20 min and still haven't gotten through”



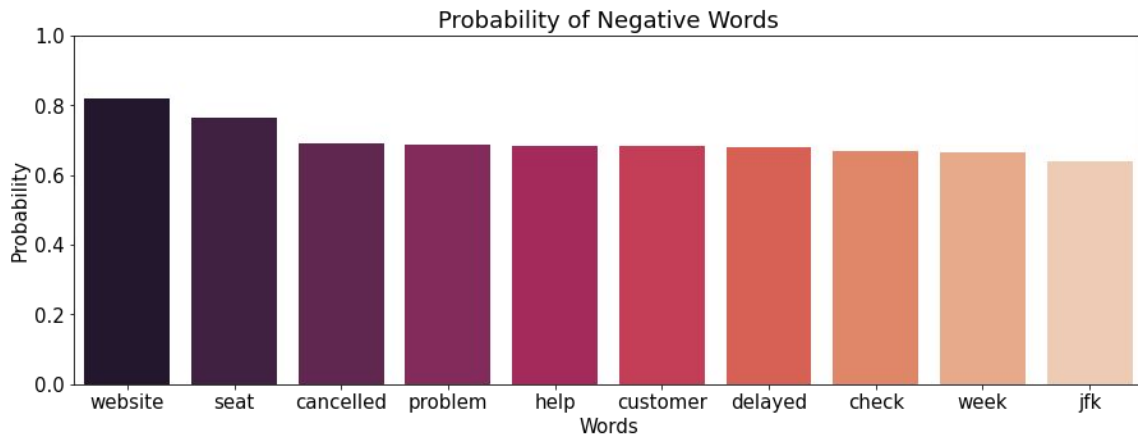
# Predictive Words: US Airways

- ❖ Customer service issues
- ❖ Not being able to connect with customer service and being on hold
- ❖ Being hung up on
- ❖ also experiencing lost luggage
- ❖ “@US Airways but I've been trying to call them since yesterday and I keep getting hung up on? Can you get me through to them??”
- ❖ “@US Airways I've been on hold to change a date on a ticket for over 3 hours. Can someone please assist me? Unacceptable.”
- ❖ “@US Airways @Beamske how about a real live person talk to the person whose luggage was lost for 4 days and vacation wrecked . @yorkshire2002”



# Predictive Words: Virgin America

- ❖ Flight booking problems and cancelled flights
- ❖ Customers weren't able to reschedule a flight online.
- ❖ Website was down
- ❖ With customers' flights being canceled, it is hard for them to reschedule their flight
- ❖ “@VirginAmerica Is it me, or is your website down? BTW, your new website isn't a great user experience. Time for another redesign.”
- ❖ “@VirginAmerica How do I reschedule my Cancelled Flightled flights online? The change button is greyed out!”



# Modeling: Data Text Preprocessing

- ❖ The following needs to be done to preprocess text.
- ❖ Make text lowercase and remove punctuations and special characters.
- ❖ Lemmatize text – returns a root word
- ❖ Remove stopwords (“this”, “is”, etc.)
- ❖ Tokenize text – puts words in a list

# Modeling: Data Vectorizing

- ❖ Vectorizing text to raw data
- ❖ Tested CountVectorizer and TfidfVectorizer
- ❖ Based on ROC-AUC Score TfidfVectorizer performed the best.

| Vectorizer      | ROC-AUC Score | Accuracy |
|-----------------|---------------|----------|
| CountVectorizer | 0.9577        | 92.18%   |
| TfidfVectorizer | 0.9583        | 90.94%   |

# Model Selection: Selecting Model

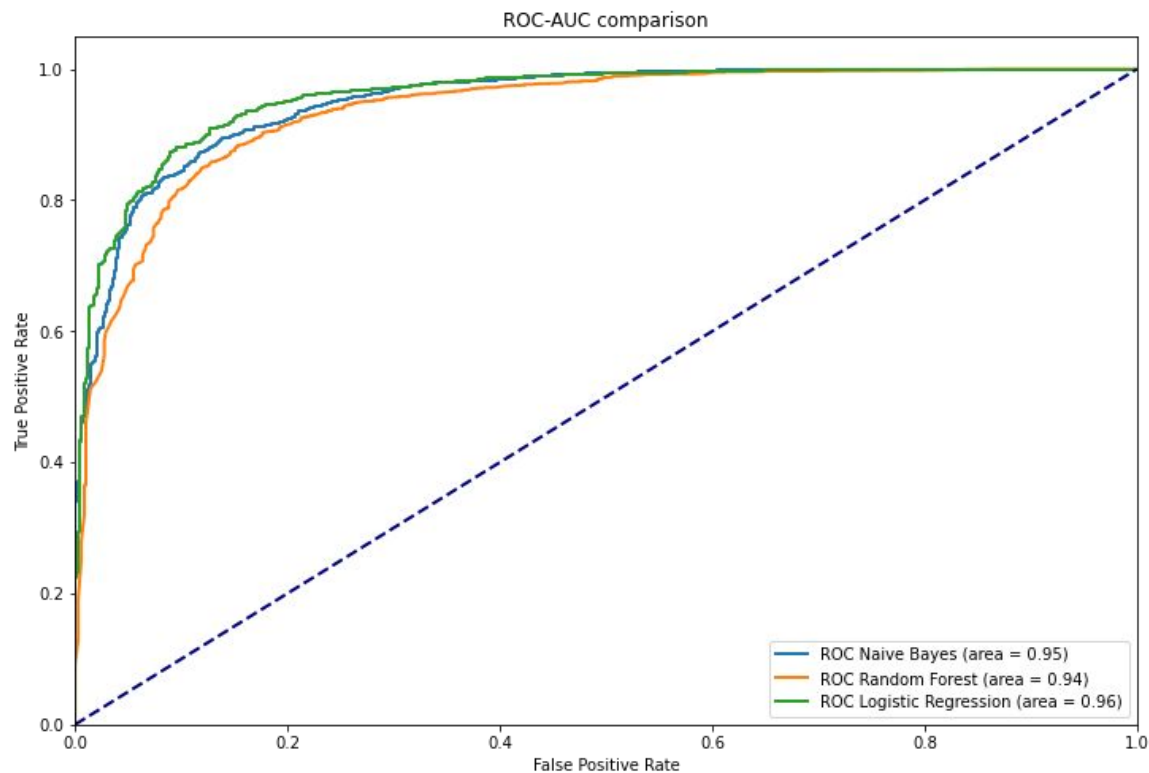
| Model                 | Best Parameters   | ROC-AUC Score | Accuracy |
|-----------------------|---|---------------|----------|
| Naive Bayes           | n/a   | 0.9512        | 89.17%   |
| Random Forest         | <b>max_depth:</b> None,<br><b>max_features:</b> auto,<br><b>n_estimators:</b> 400 | 0.9373        | 89.91%   |
| Logistic Regression 1 | <b>C:</b> 1,<br><b>max_iter:</b> 1000   | 0.9583        | 90.94%   |



# Model Selection: Selecting Model Cont.

| Model   | Best Parameters         | ROC-AUC Score | Accuracy |
|---|-------------------------|---------------|----------|
| Logistic Regression 2 -<br>CountVect with n-grams | C: 1,<br>max_iter: 1000 | 0.9570        | 91.44%   |
| Logistic Regression 3 -<br>Tf-idf with n-grams    | C: 1,<br>max_iter: 1000 | 0.9517        | 87.72%   |

# ROC-AUC Comparisons



# Best Threshold

- ❖ Change the threshold for classification problems to improve performance.
- ❖ Threshold will change based on specific business needs

| Threshold     | Recall | Precision | Accuracy | F1-Score | Ba  |
|---------------|--------|-----------|----------|----------|-----|
| Default (0.5) | 0.99   | 0.91      | 0.91     | 0.95     | 0.8 |

# Business Case 1: Track Sentiment Analysis

- ❖ Track overall customer sentiment on Twitter
- ❖ Managers can make assessments for improvements.
- ❖ Focus on balanced accuracy which is good for imbalance dataset and when we care about the positive and negative classes.
- ❖ Having a threshold of 0.70 allows the model to have a better balanced accuracy.

| Threshold         | Default (0.50) | Optimal Balanced Accuracy (0.70) |
|-------------------|----------------|----------------------------------|
| Recall            | 0.99           | 0.95                             |
| Precision         | 0.91           | 0.95                             |
| Accuracy          | 0.91           | 0.92                             |
| F1-Score          | 0.95           | 0.95                             |
| Balanced Accuracy | 0.80           | 0.88                             |

## Business Case 2: Customer Service

- ❖ Use model to alert bot to alert customers with appropriate message based on sentiment.
- ❖ A false negative would predict a tweet to be a positive sentiment tweet, but it was actually a negative sentiment tweet.
- ❖ A false negative will result a customer not receiving the proper customer service help.

| Threshold         | Default<br>(0.50) | Optimal<br>F1-Measure(0.664) |
|-------------------|-------------------|------------------------------|
| Recall            | 0.99              | 0.96                         |
| Precision         | 0.91              | 0.95                         |
| Accuracy          | 0.91              | 0.92                         |
| F1-Score          | 0.95              | 0.95                         |
| Balanced Accuracy | 0.80              | 0.87                         |

## Business Case 2: Customer Service Cont.

- ❖ A false positive would predict a negative sentiment tweet, but it was actually a positive sentiment tweet.
- ❖ A false positive will result in a bot responding to a customer that had a positive experience and offer a discount or refund
- ❖ We could use F1-Score since it uses precision and recall.
- ❖ The metric might change based on a cost analysis of the false positives and compare to the false negatives.

| Threshold                | Default (0.50) | Optimal<br>F1-Measure(0.664) |
|--------------------------|----------------|------------------------------|
| <b>Recall</b>            | 0.99           | 0.96                         |
| <b>Precision</b>         | 0.91           | 0.95                         |
| <b>Accuracy</b>          | 0.91           | 0.92                         |
| <b>F1-Score</b>          | 0.95           | 0.95                         |
| <b>Balanced Accuracy</b> | 0.80           | 0.87                         |

# Conclusion

- ❖ Tracking customer sentiment can help businesses improve products and services.
- ❖ This model can be used for airlines based on specific business needs.
- ❖ The next steps would include looking deeper at the causes of misclassification to improve the model because some tweets can have sarcasm.
- ❖ I would also gather tweets from a longer timeframe because all of the tweets collected are from Feb 16, 2015 and Feb 24, 2015.

**End**

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