

Topic Modeling Using Natural Language Processing

Johnny McGregor
General Assembly

Topic Modeling Using Natural Language Processing

Johnny McGregor
General Assembly

Objective

- Build a model that can help a science website identify whether posts to their website is relevant.
- Collect posts from the science and comedy subreddits.
- Use Natural Language Processing to make the text compatible with the model.
 - Represent the words as numbers
- Evaluate how accurate the model is at distinguishing between the two topics.

Data

- Collected using the Reddit API
 - Collected as a JSON file (Java Script Object Notation)
 - Using Python we can convert into a Pandas Dataframe
- 2868 total posts
 - 1344 science
 - 1524 comedy

Natural Language Processing

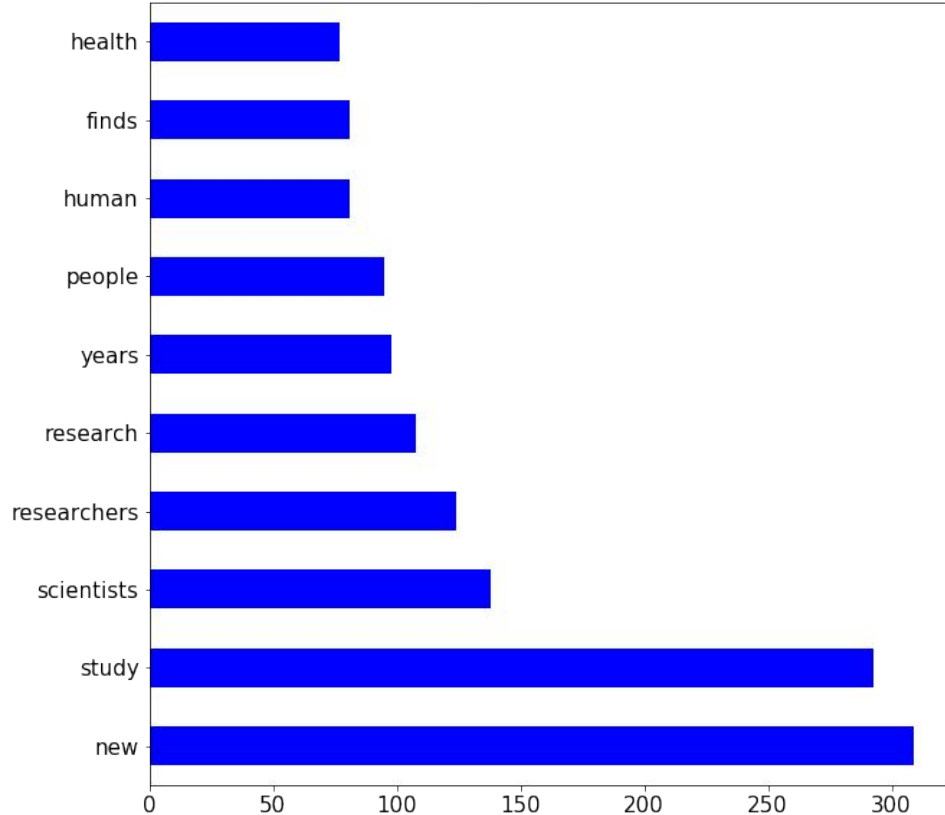
- “Hello, how are you?”
 - hello
 - how
 - are
 - you
- Filter out everything except the letters
 - Comma, questions mark, and quotes are removed above.
 - Each word can become a feature in the model

Natural Language Processing

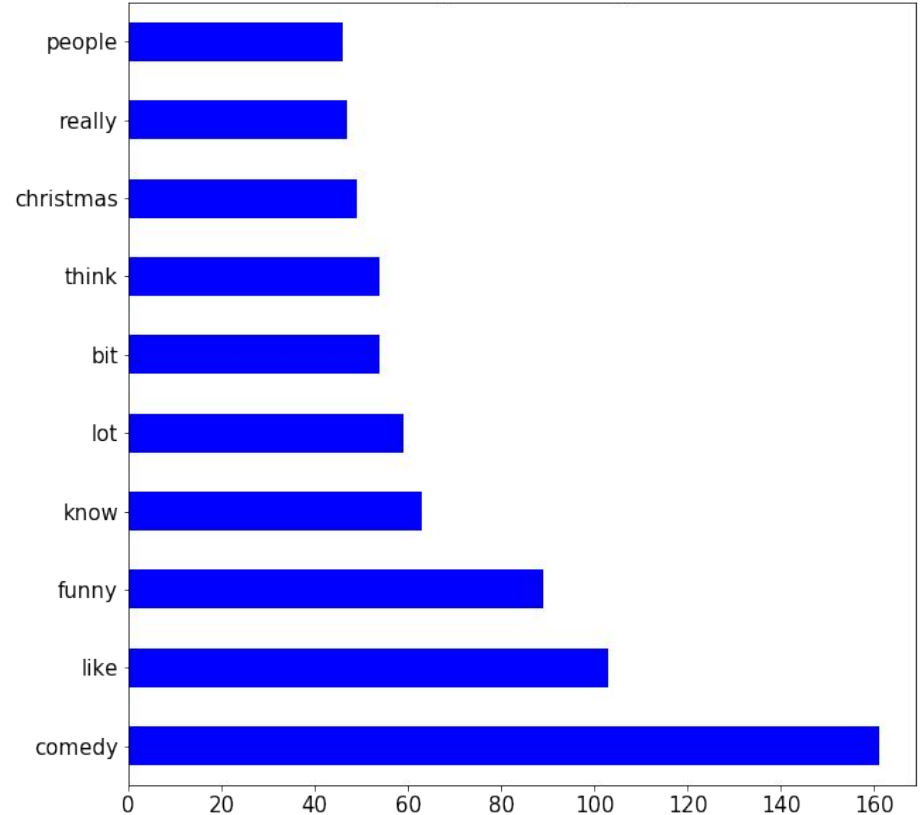
- Word Count
 - Tracks the amount of times a word occurs.
 - Ignores grammar, structure, order
- Frequency
 - $\text{Number of times a word is in a document (post) / Total words in document}$
 - $\text{Number of total documents (2868 in this case) / Number of documents that contain the word}$
- Often a good idea to discard “stop words”
 - Due to high frequency they are usually not meaningful in distinguishing between categories
 - I, by, me, my, of, we you
- Feature Engineering
 - Shortening words to their roots
 - Combining words into pairs/trios

Most Common Words

Most Frequent Science Terms



Most Frequent Comedy Terms



Modeling

- Split up our data
 - 2151 posts used for training data set
 - 717 posts used for test data set
- Fit Training data to the model
 - Set certain parameters like maximum features, maximum document frequency
- Predict whether each post in the test data came from science or comedy subreddit

How Do We Measure Success?

- Of the 717 posts in the test data set how many were correctly identified?
 - Accuracy
- Of the posts that were about science how many were correctly identified?
 - True Positive Rate
- Of the posts that weren't about science how many were correctly identified?
 - True Negative Rate
- Of all science predictions how many were correct?
 - Precision

How Do We Measure Success?

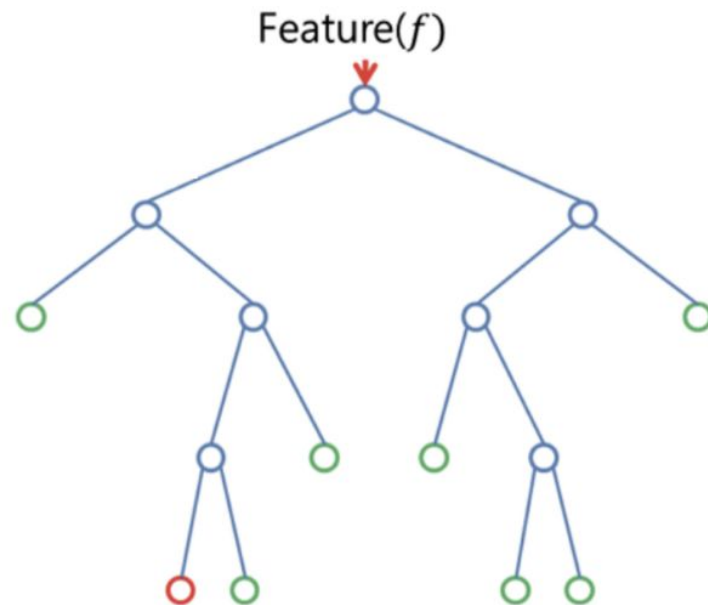
- Accuracy
- Low False Positives Rate ($1 - \text{True Positive Rate}$)
- Baseline Score
 - 1524 comedy posts/ 1344 science posts
 - If we predicted every post to be comedy (or not science) we would be right 53.1% of the time

Logistic Regression

- Makes predictions based on the probability of a post being in either science or not science. 1 or 0.
- Can quantify the likelihood of a 1 or a zero based on each feature. The word study in a post makes it x times more likely to be a science post.
 - Three and half times more likely to be classified science If “study” is in the post
 - Probability .78

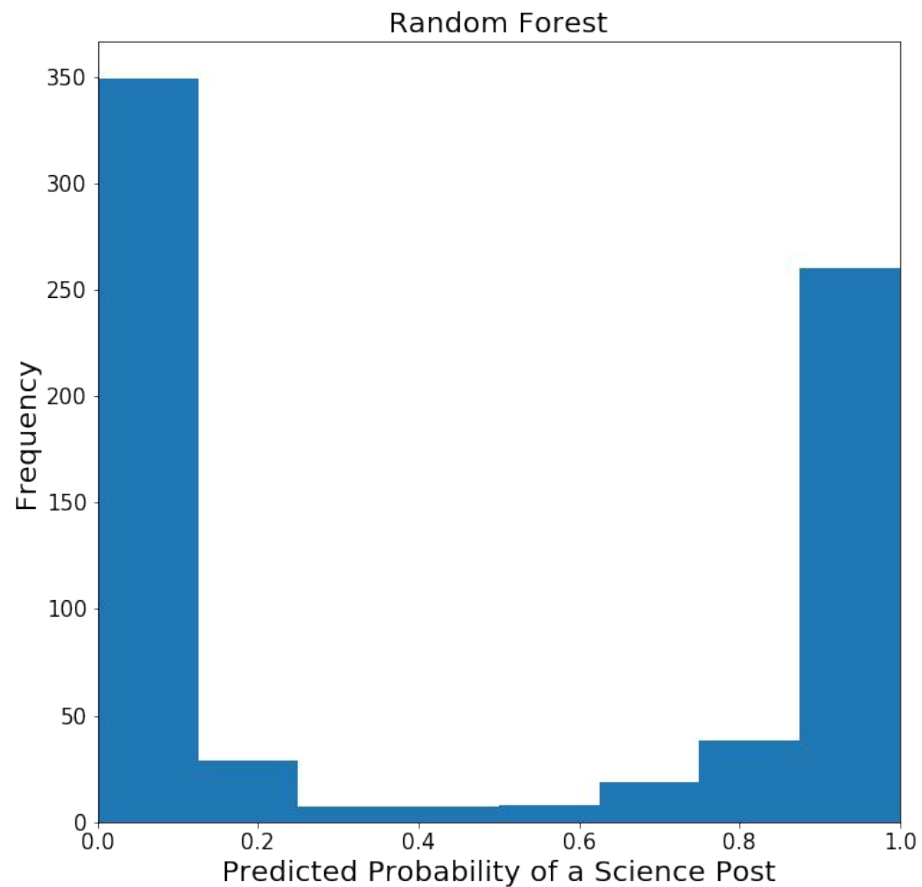
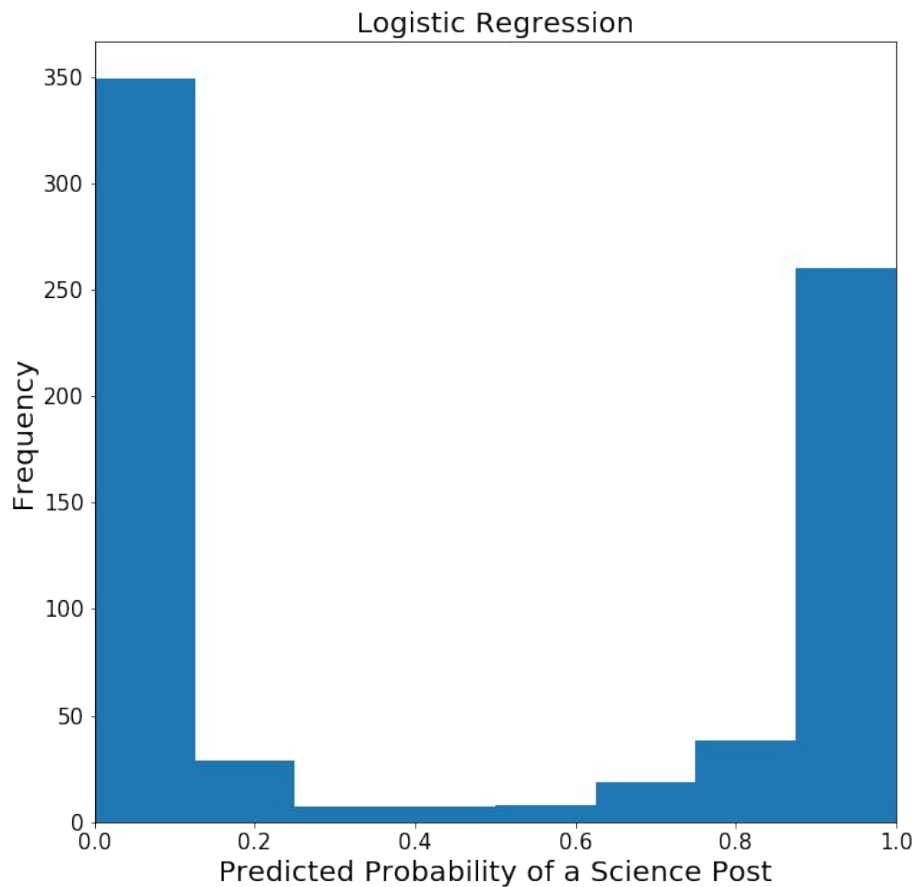
Random Forest

- Splits the data up based on random subset of features
- Can set the depth of the “trees”, maximum amount of features,
- The goal is to separate the two classes as much as possible and this is done many times over with a many Decision Tree Classifier



Decision Tree

Model Comparison



Model Comparison (717 Possible Outcomes)

Logistic Regression:

Accuracy - 97.6

True Negatives - 378 (Rate = 99.2)

True Positives - 322 (Rate = 95.8)

False Negatives - 14

False Positives - 3

Precision = 99.0

Random Forest:

Accuracy - 96.9

True Negatives - 378 (Rate = 99.2)

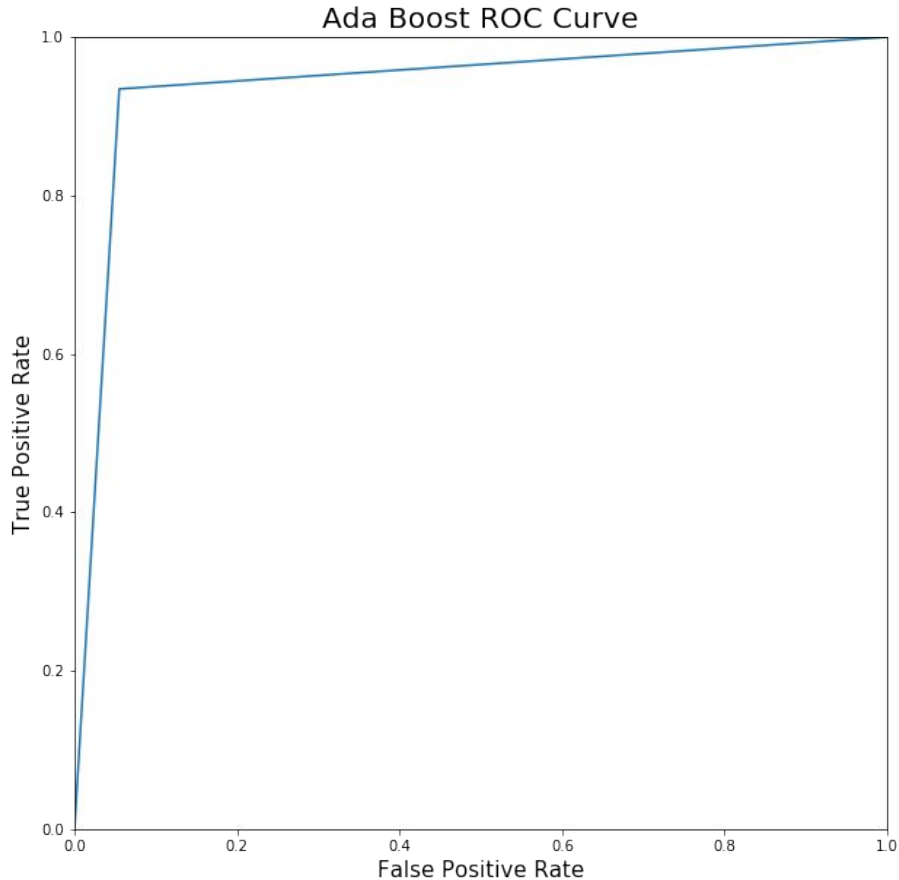
True Positives - 317 (Rate = 94.3)

False Negatives - 19

False Positives - 3

Precision = 99.0

Ada Boost ROC Curve



True Negatives: 363

False Positives: 18

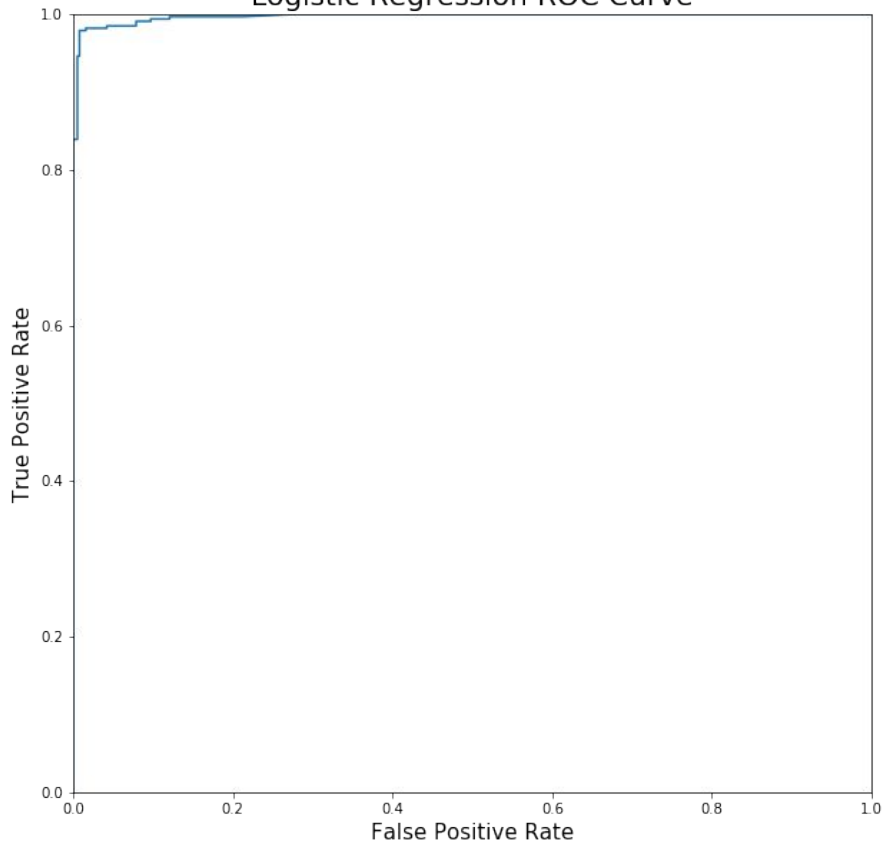
False Negatives: 22

True Positives: 314

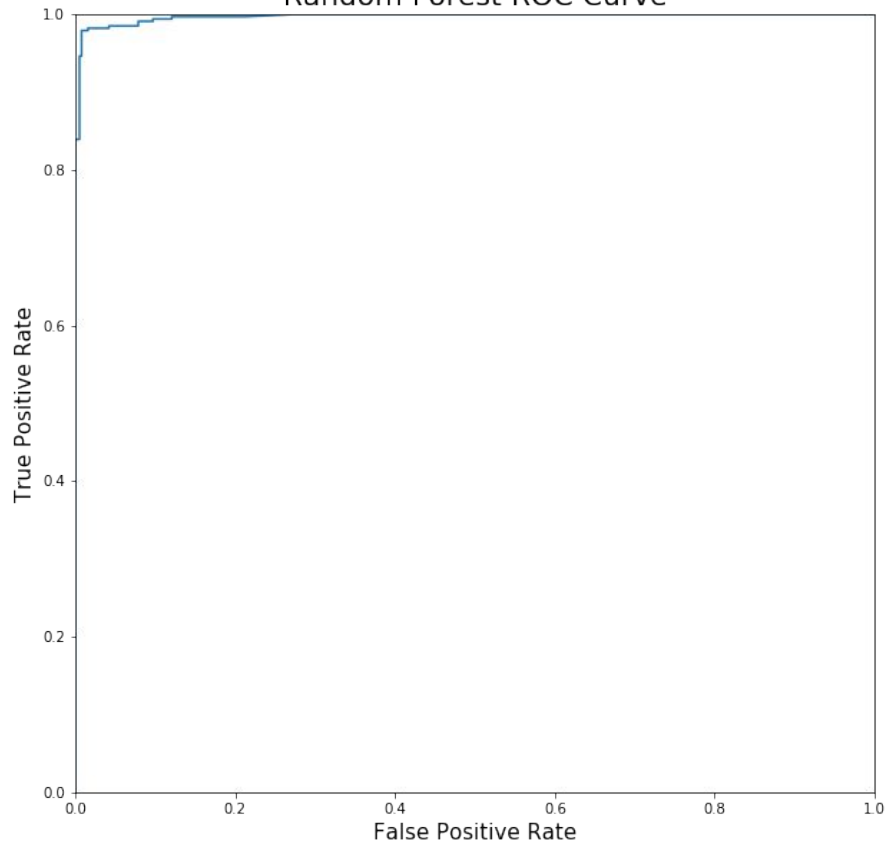
Adjust Probability threshold lower to
lessen false positives, adjust higher for
false negatives

ROC Curves

Logistic Regression ROC Curve



Random Forest ROC Curve



Misclassifications

	true_science_post	predict_proba	predictions	post
1338	1	0.197532	0	repeatedly watching video touching filthy bedpan reduced people ocd symptoms
534	1	0.408786	0	freeze dried polio vaccine require refrigeration offered full protection polio virus tested mice
760	1	0.137316	0	dietary versatility early pleistocene hominins
1291	1	0.387267	0	university leeds news young star caught forming like planet
1143	1	0.408786	0	freeze dried polio vaccine require refrigeration offered full protection polio virus tested mice
1289	1	0.419122	0	pathogenic copy number variants affect gene expression contribute genomic burden cerebral palsy
1333	1	0.474157	0	tourists may making antarctica penguins sick
151	1	0.137316	0	dietary versatility early pleistocene hominins
2073	0	0.562806	1	cannot unsee snake space space odyssey
1311	1	0.297073	0	modern humans round heads
1281	1	0.110851	0	genetically modified pigs protected classical swine fever virus
1290	1	0.479111	0	active cognitive lifestyle potential neuroprotective factor huntington disease
1329	1	0.078276	0	painless adhesives
1274	1	0.448082	0	human sex reversal caused duplication deletion core enhancers upstream sox
1264	1	0.123501	0	chromatin loop extrusion chromatin unknotting

Recommendations

- If you want to eliminate posts incorrectly being classified as science, increase the threshold higher than .5
- Random Forest over Logistic Regression
- Look at some common words from misclassified posts to better train the model, and possibly flag inappropriate posts

Objective

- Build a model that can help a science website identify whether posts to their website is relevant.
- Collect posts from the science and comedy subreddits.
- Use Natural Language Processing to make the text compatible with the model.
 - Represent the words as numbers
- Evaluate how accurate the model is at distinguishing between the two topics.

Data

- Collected using the Reddit API
 - Collected as a JSON file (Java Script Object Notation)
 - Using Python we can convert into a Pandas Dataframe
- 2868 total posts
 - 1344 science
 - 1524 comedy

Natural Language Processing

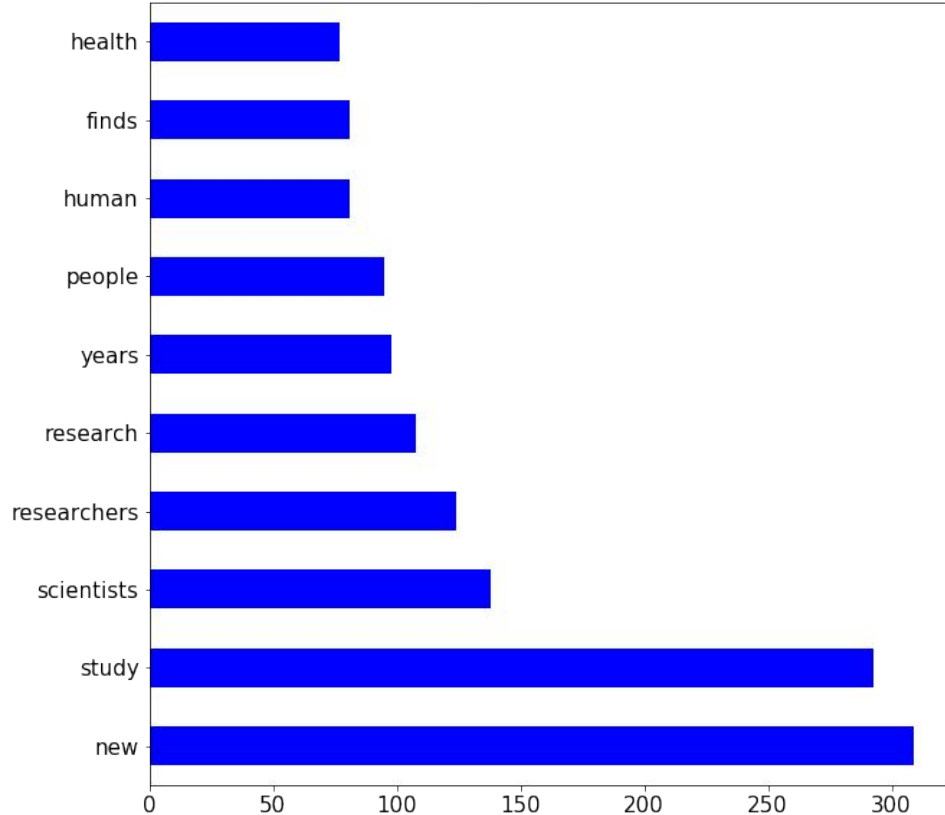
- “Hello, how are you?”
 - hello
 - how
 - are
 - you
- Filter out everything except the letters
 - Comma, questions mark, and quotes are removed above.
 - Each word can become a feature in the model

Natural Language Processing

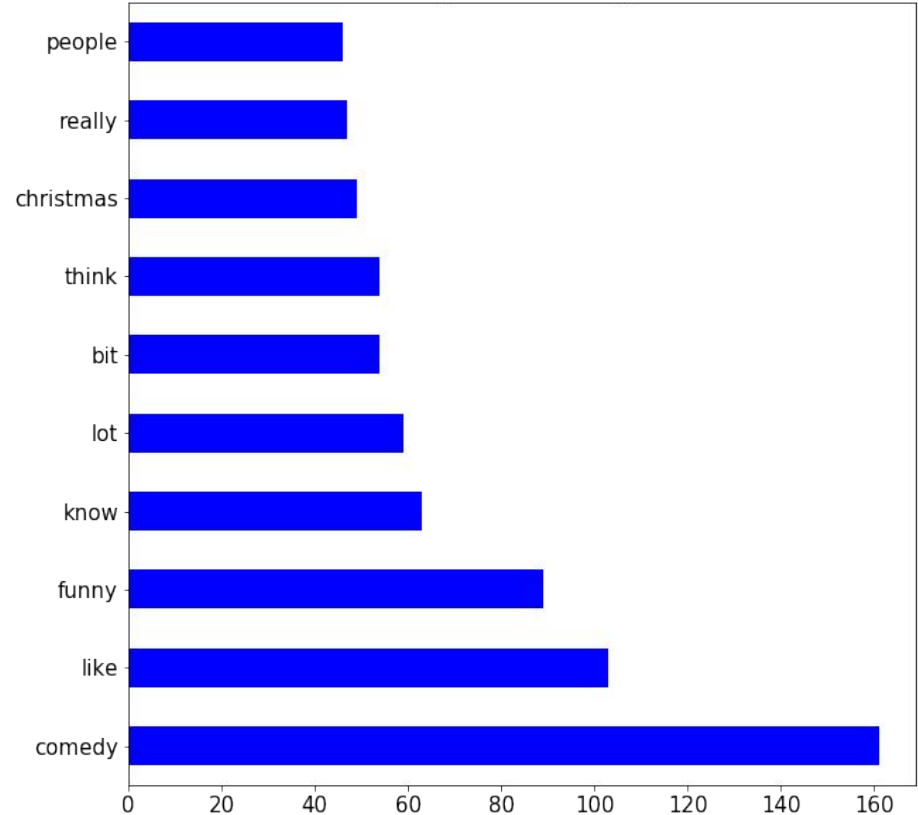
- Word Count
 - Tracks the amount of times a word occurs.
 - Ignores grammar, structure, order
- Frequency
 - $\text{Number of times a word is in a document (post)} / \text{Total words in document}$
 - $\text{Number of total documents (2868 in this case)} / \text{Number of documents that contain the word}$
- Often a good idea to discard “stop words”
 - Due to high frequency they are usually not meaningful in distinguishing between categories
 - I, by, me, my, of, we you
- Feature Engineering
 - Shortening words to their roots
 - Combining words into pairs/trios

Most Common Words

Most Frequent Science Terms



Most Frequent Comedy Terms



Modeling

- Split up our data
 - 2151 posts used for training data set
 - 717 posts used for test data set
- Fit Training data to the model
 - Set certain parameters like maximum features, maximum document frequency
- Predict whether each post in the test data came from science or comedy subreddit

How Do We Measure Success?

- Of the 717 posts in the test data set how many were correctly identified?
 - Accuracy
- Of the posts that were about science how many were correctly identified?
 - True Positive Rate
- Of the posts that weren't about science how many were correctly identified?
 - True Negative Rate
- Of all science predictions how many were correct?
 - Precision

How Do We Measure Success?

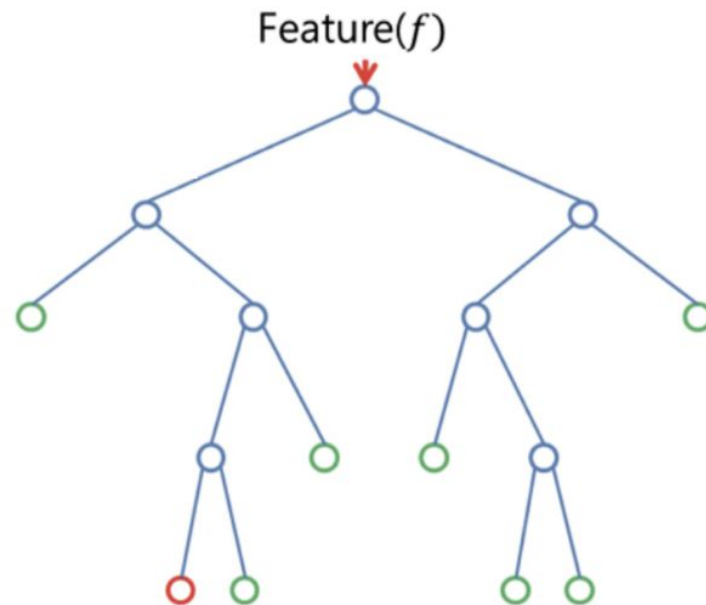
- Accuracy
- Low False Positives Rate ($1 - \text{True Positive Rate}$)
- Baseline Score
 - 1524 comedy posts/ 1344 science posts
 - If we predicted every post to be comedy (or not science) we would be right 53.1% of the time

Logistic Regression

- Makes predictions based on the probability of a post being in either science or not science. 1 or 0.
- Can quantify the likelihood of a 1 or a zero based on each feature. The word study in a post makes it x times more likely to be a science post.
 - Three and half times more likely to be classified science If “study” is in the post
 - Probability .78

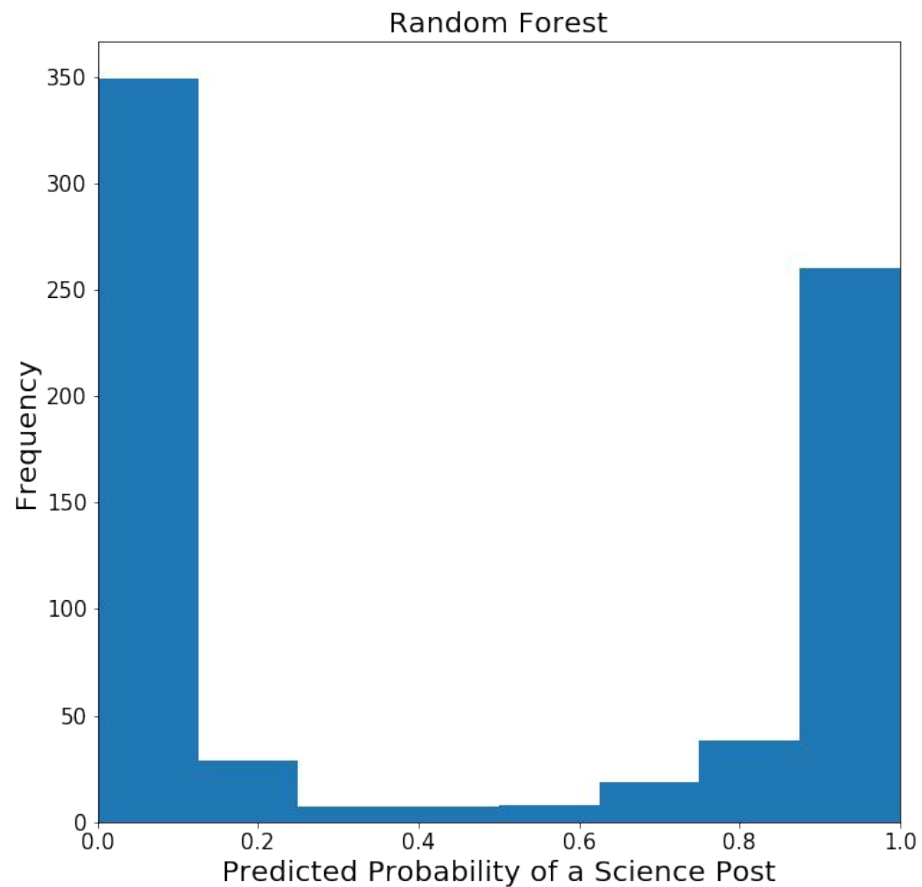
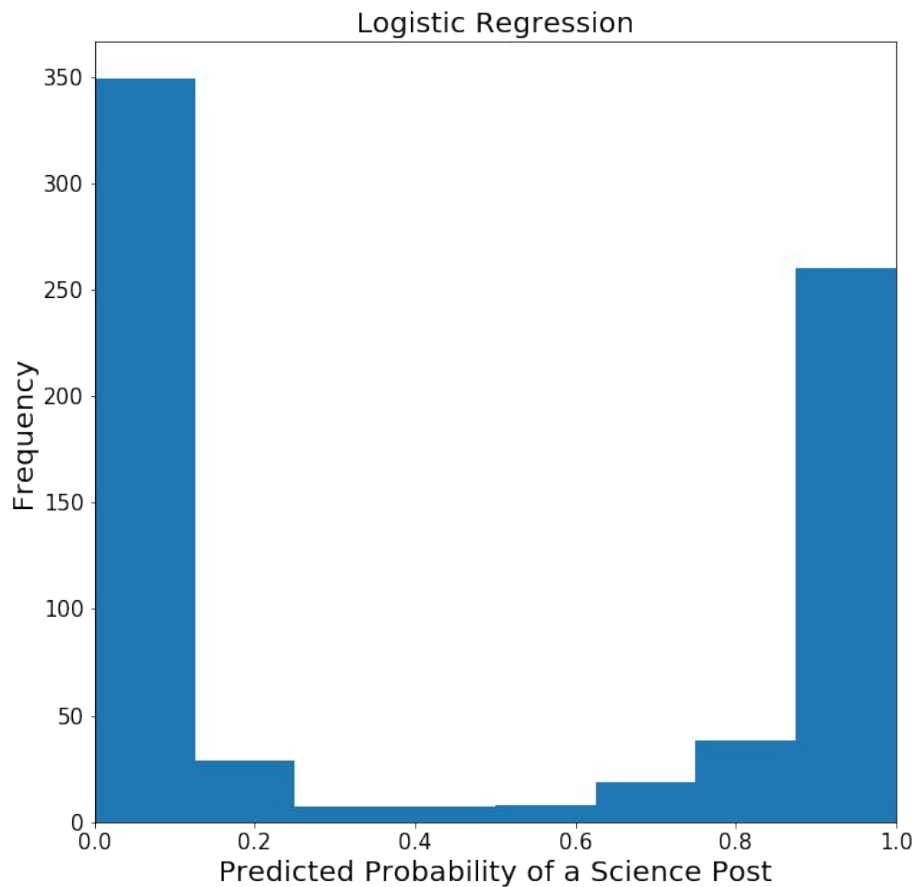
Random Forest

- Splits the data up based on random subset of features
- Can set the depth of the “trees”, maximum amount of features,
- The goal is to separate the two classes as much as possible and this is done many times over with a many Decision Tree Classifier



Decision Tree

Model Comparison



Model Comparison (717 Possible Outcomes)

Logistic Regression:

Accuracy - 97.6

True Negatives - 378 (Rate = 99.2)

True Positives - 322 (Rate = 95.8)

False Negatives - 14

False Positives - 3

Precision = 99.0

Random Forest:

Accuracy - 96.9

True Negatives - 378 (Rate = 99.2)

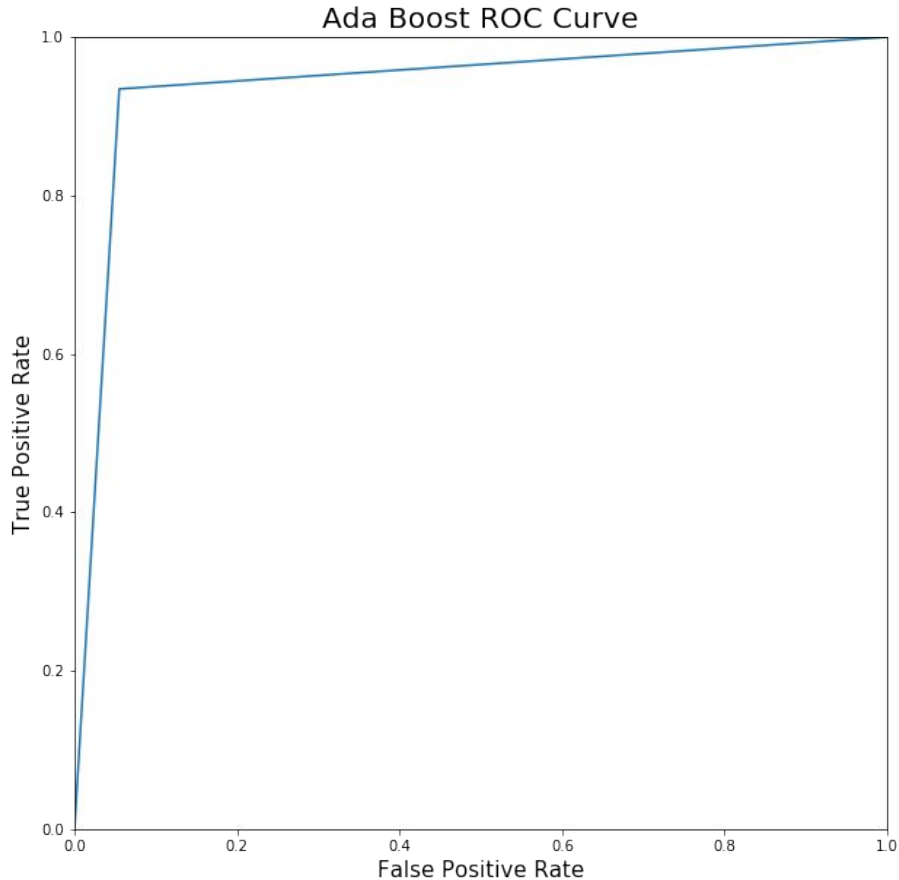
True Positives - 317 (Rate = 94.3)

False Negatives - 19

False Positives - 3

Precision = 99.0

Ada Boost ROC Curve



True Negatives: 363

False Positives: 18

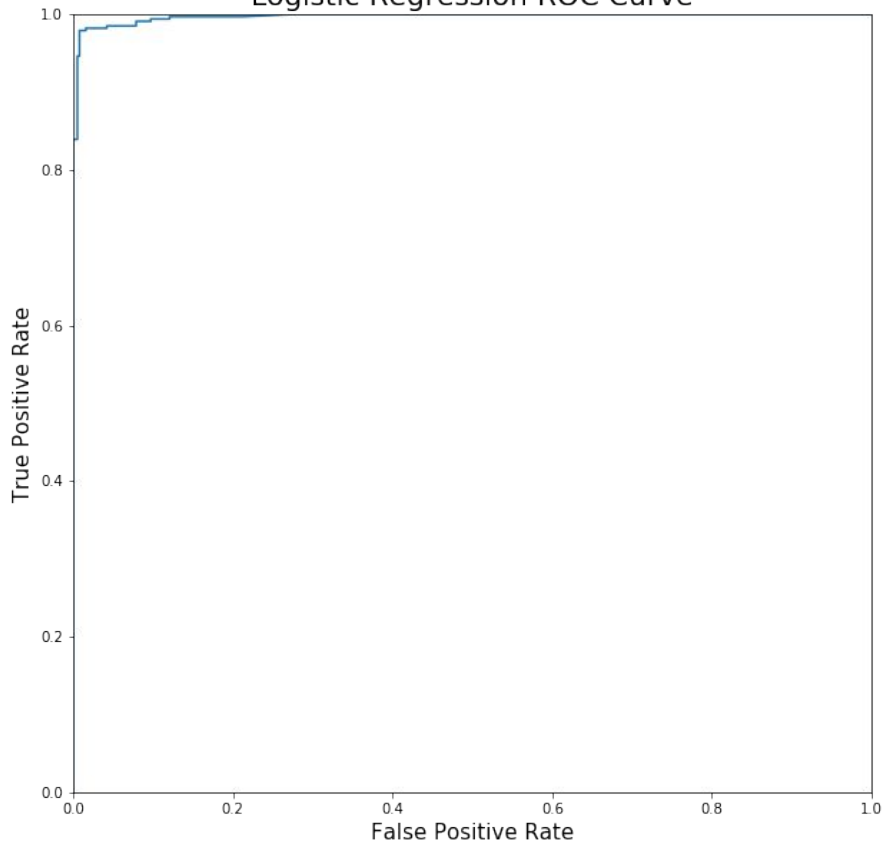
False Negatives: 22

True Positives: 314

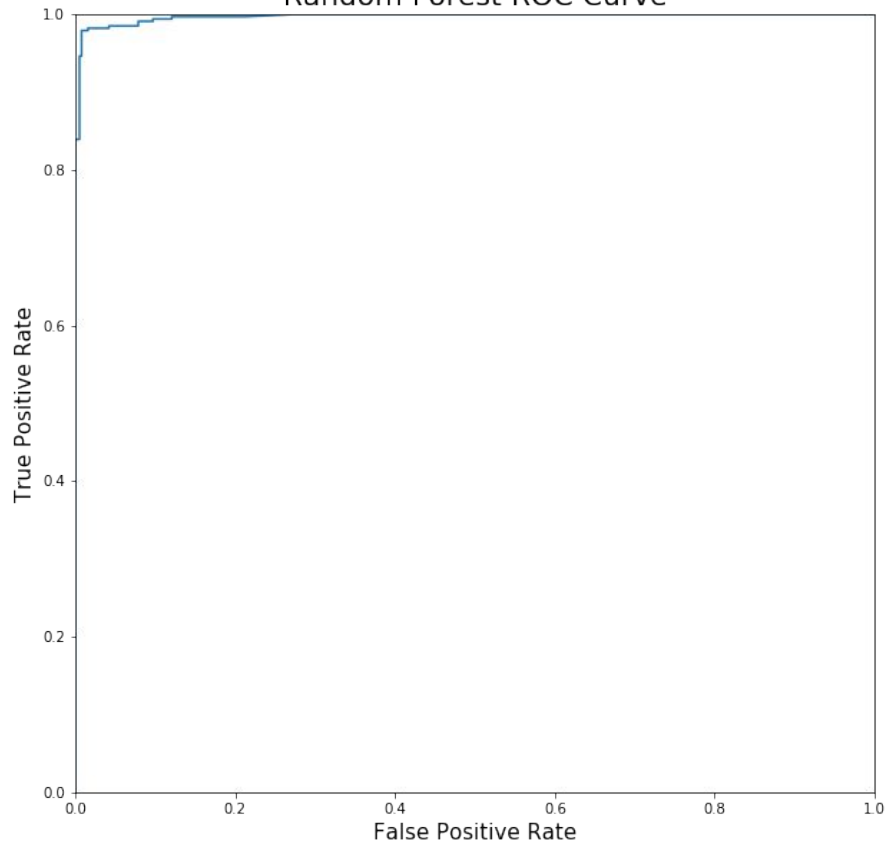
Adjust Probability threshold lower to
lessen false positives, adjust higher for
false negatives

ROC Curves

Logistic Regression ROC Curve



Random Forest ROC Curve



Misclassifications

	true_science_post	predict_proba	predictions	post
1338	1	0.197532	0	repeatedly watching video touching filthy bedpan reduced people ocd symptoms
534	1	0.408786	0	freeze dried polio vaccine require refrigeration offered full protection polio virus tested mice
760	1	0.137316	0	dietary versatility early pleistocene hominins
1291	1	0.387267	0	university leeds news young star caught forming like planet
1143	1	0.408786	0	freeze dried polio vaccine require refrigeration offered full protection polio virus tested mice
1289	1	0.419122	0	pathogenic copy number variants affect gene expression contribute genomic burden cerebral palsy
1333	1	0.474157	0	tourists may making antarctica penguins sick
151	1	0.137316	0	dietary versatility early pleistocene hominins
2073	0	0.562806	1	cannot unsee snake space space odyssey
1311	1	0.297073	0	modern humans round heads
1281	1	0.110851	0	genetically modified pigs protected classical swine fever virus
1290	1	0.479111	0	active cognitive lifestyle potential neuroprotective factor huntington disease
1329	1	0.078276	0	painless adhesives
1274	1	0.448082	0	human sex reversal caused duplication deletion core enhancers upstream sox
1264	1	0.123501	0	chromatin loop extrusion chromatin unknotting

Recommendations

- If you want to eliminate posts incorrectly being classified as science, increase the threshold higher than .5
- Random Forest over Logistic Regression
- Look at some common words from misclassified posts to better train the model, and possibly flag inappropriate posts