



Capstone: What makes a Yelp review useful?

Problem Statement

Yelp is a platform that allows users to make reservations, leave reviews, and find businesses. In addition, users can mark which reviews they find useful. Reviews can be both very helpful and very harmful for businesses. Being able to determine what makes a useful review can help businesses create better, more targeted listings.

What is a “Yelp”?

- Find and connect with businesses
 - Appointments
 - Waitlists
 - Reservations
- Business info
 - Hours of operation
 - Location
- Reviews, photos
 - Funny/Cool/Useful



The Data!

- Available from <https://www.yelp.com/dataset>
- 6,990,280 reviews
- 150,346 businesses
- 200,100 pictures



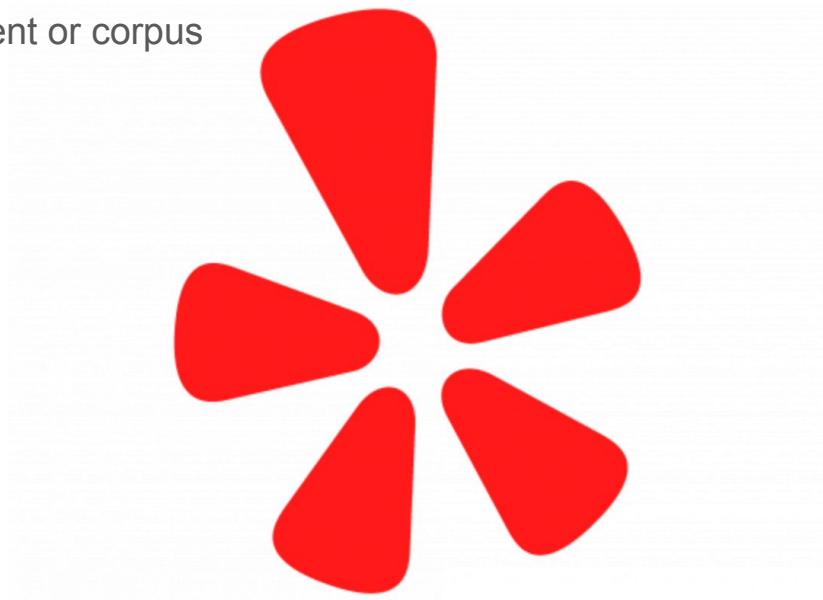
Data Preparation

- Data was randomly sampled to approx 20,000 records
- No imputation was necessary
- New lines and punctuation were cleaned up
- Outliers removed
 - Useful column had extreme outliers



Vectorization

- TF-IDF
 - Term frequency-inverse document frequency
 - A measure of how relevant a word is in a document or corpus
 - Max depth: 200



LassoCV

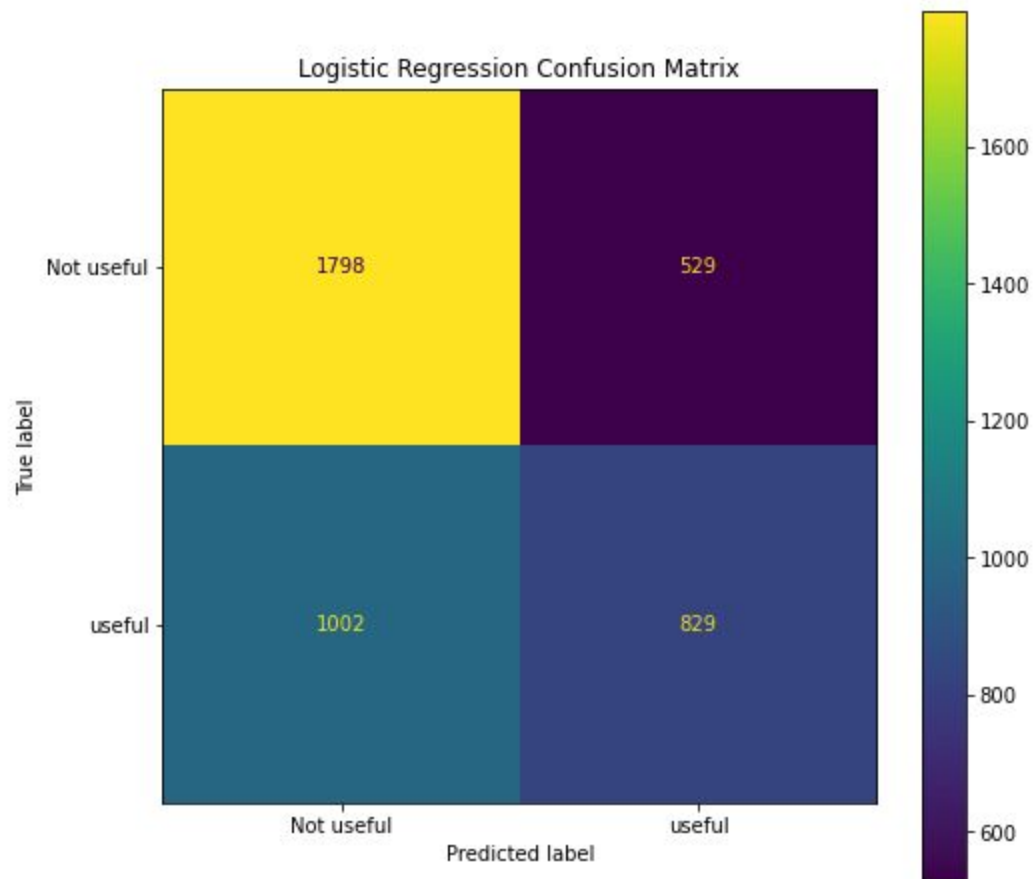
- Reduce columns / multicollinearity



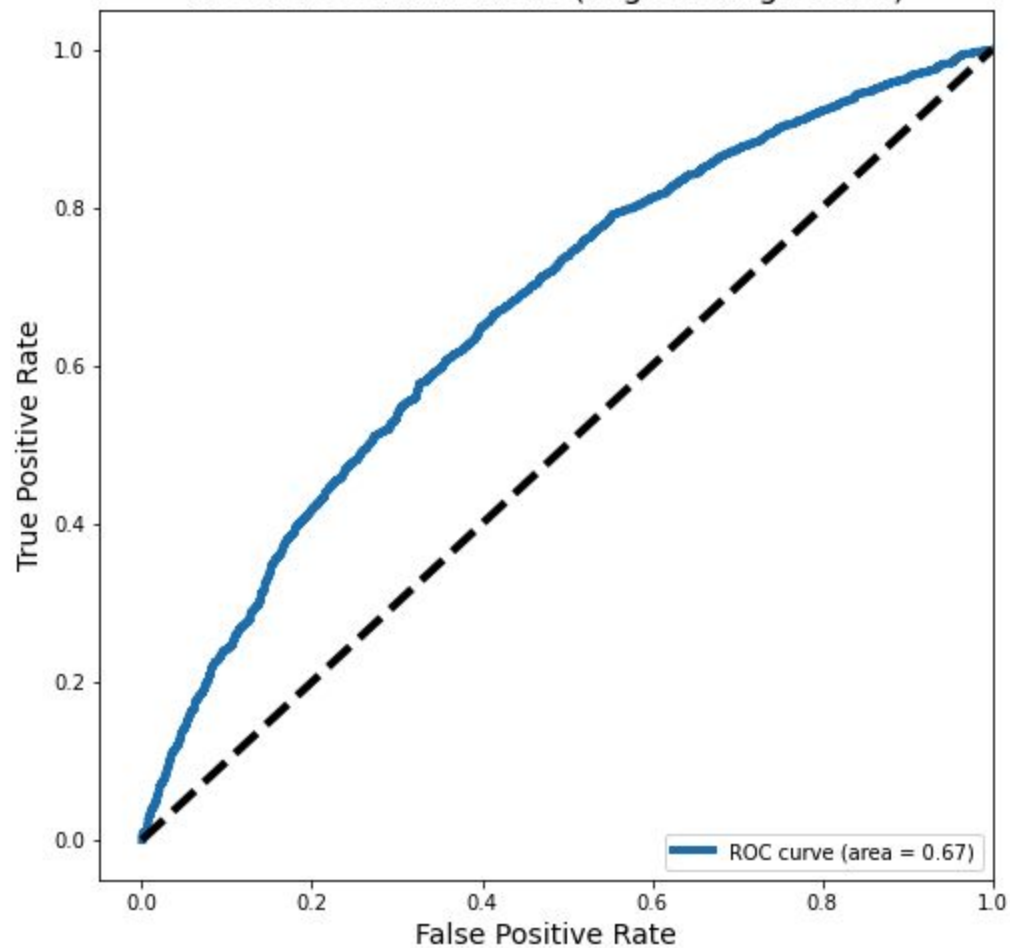
Logistic Regression Model

- Parameters
 - $C=100$
 - Solver: liblinear
 - Penalty: l2
- Metrics
 - 0.62 ± 0.036
 - Precision: 0.61
 - Recall: 0.45
 - F1-score: 0.52





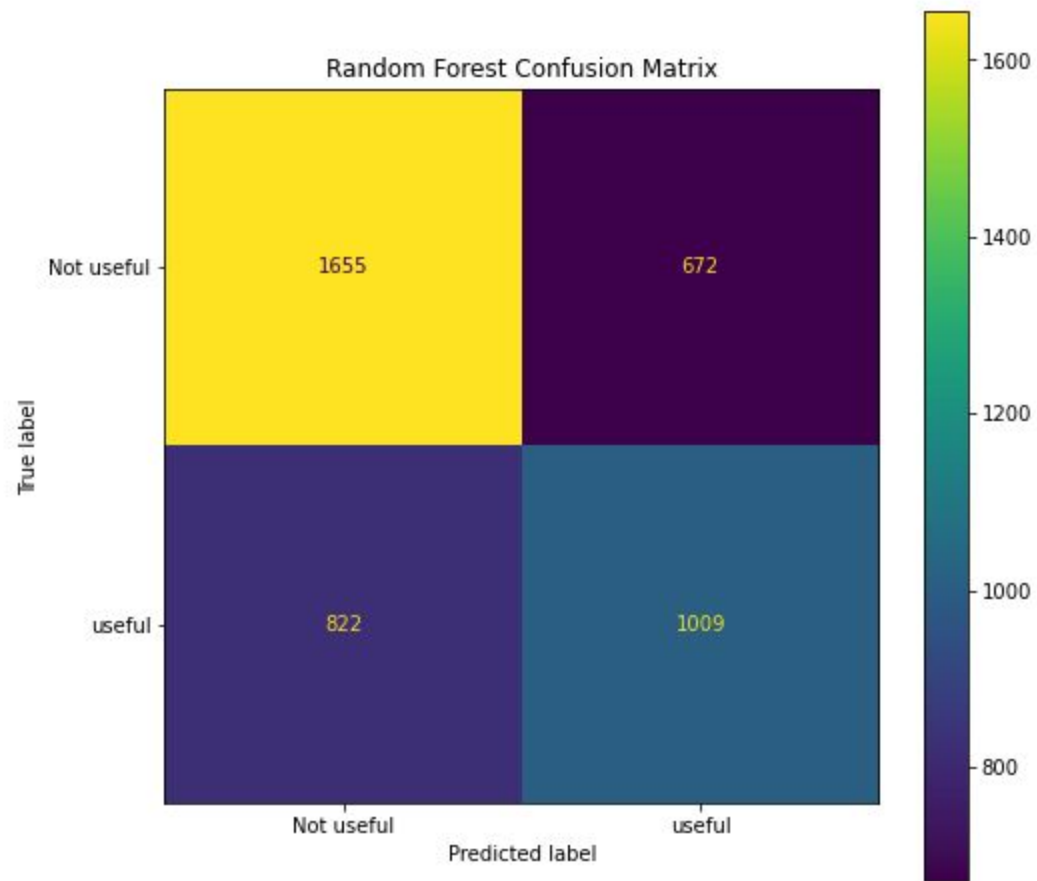
ROC for Useful Reviews (Logistic Regression)



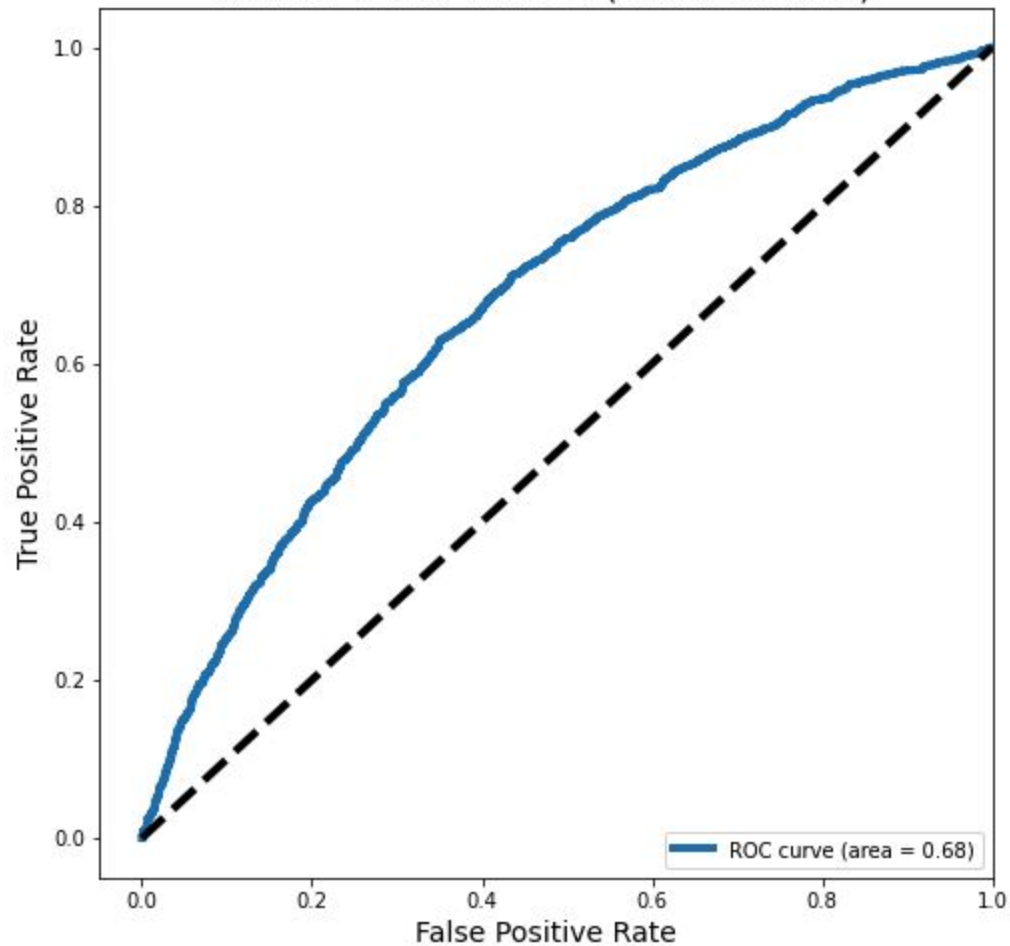
Random Forest

- Parameters
 - N_estimators: 200
 - Min_samples_leaf: 10
 - Warm start: true
- Metrics
 - 0.64 +/- 0.038
 - Precision: 0.60
 - Recall: 0.55
 - F1-score: 0.57





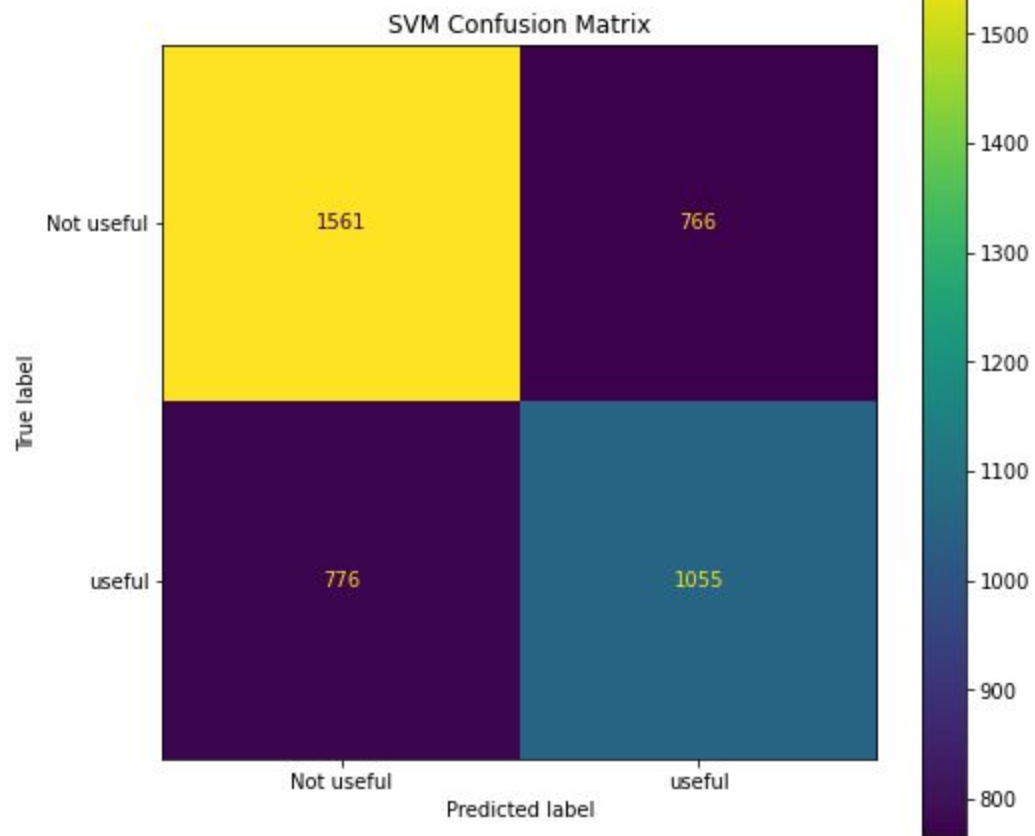
ROC for Useful Reviews (Random Forest)



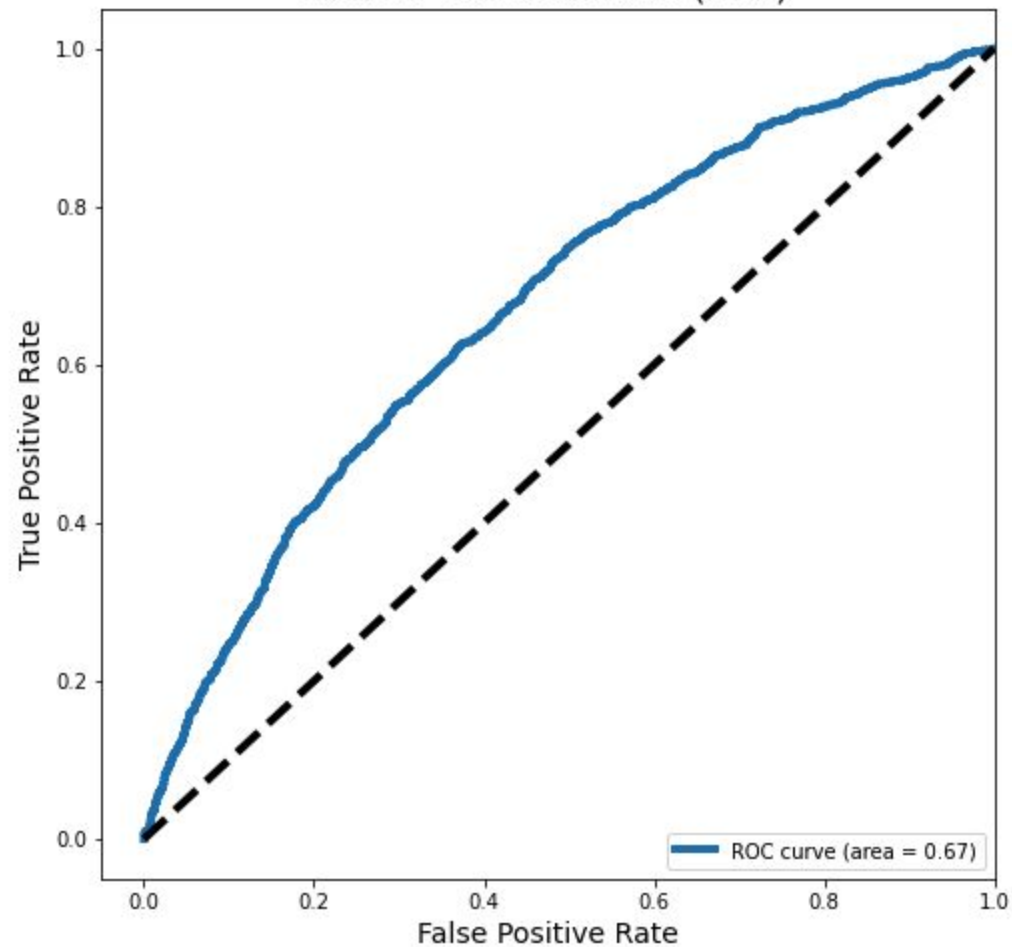
SVM

- Parameters
 - $C=0.0001$
 - Max_iter: 900
 - Class_weight: balanced
- Metrics
 - 0.62 +/- 0.043
 - Precision: 0.58
 - Recall: 0.58
 - F1-score: 0.58





ROC for Useful Reviews (SVM)



Neural Network

- Parameters
 - 1 input, 3 hidden, 1 output layer
 - Kernel regularization = l2
 - Activation function = ReLU for non-output layers
 - Activation function = sigmoid for output layer
- Metrics
 - 0.63



Themes

- Location!

- Shop
- Place
- Store

- Praise

- Great
- Good
- Like

- Temporal

- New
- Time



Best Model?

- All models beat the baseline (0.56)
- SVM?
 - Best true positive predictions
 - Lowest false negative predictions
 - Highest recall score
 - Highest f1-score
 - BUT score is less than Random Forest and NN
 - AND highest false positive rate



Conclusion

- Somewhat limited in how accurate we can get the models
 - Accuracy around 62-64%
- All models performed comparably
 - No clear winners based on multiple metrics
- Themes
- Future analysis might want to pull in business data
 - Location
 - Type

