Topic Model Classification via Yelp Reviews

Can Topic Model distributions predict review sentiment on unseen text?



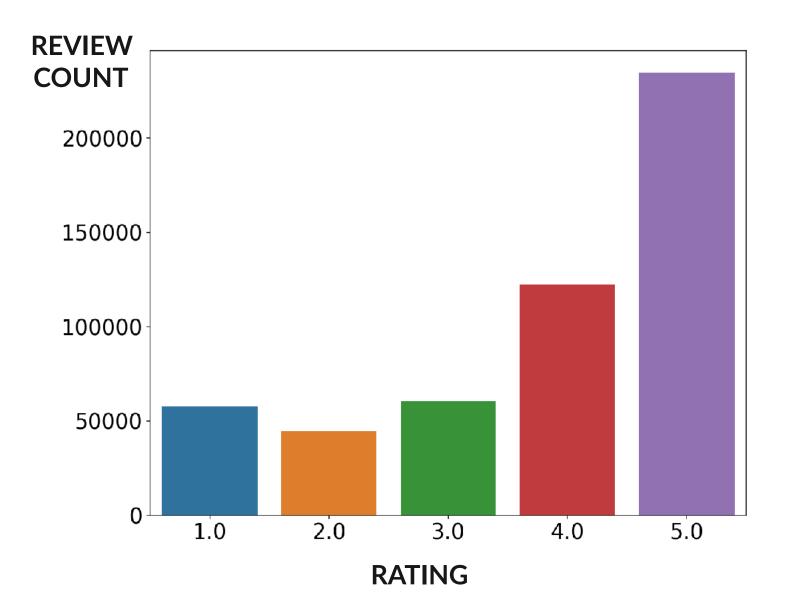
Use Cleaned & Bigram-ed Review Text Feature Vector is a Probability Distribution

Predict Review Sentiment

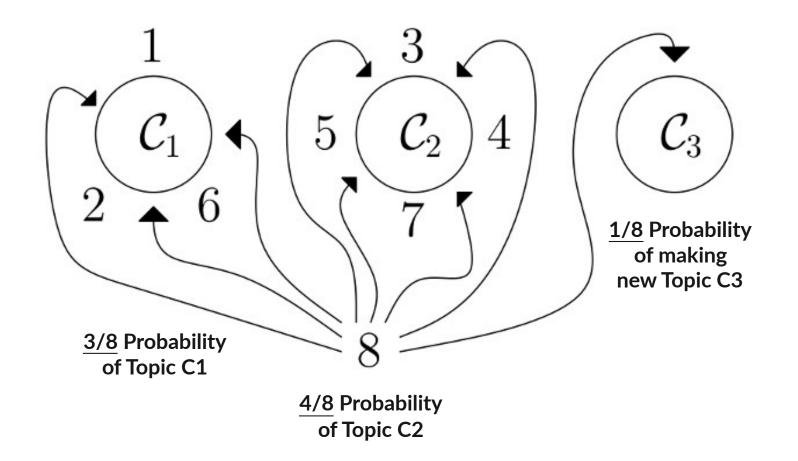


2016 Review Ratings

- Heavily tilted toward 4s and 5s.
- Will correct for this imbalance when training classifiers.



How to choose the Number of Topics?



- Hierarchical Dirichlet Process does <u>not</u> assume fixed number of topics.
- Above is rough explanation of doc-topic grouping.

Source(s): http://gerin.perso.math.cnrs.fr,

http://blog.echen.me

FIT HIERARCHICAL DIRICHLET PROCESS

Extract Topic Distributions with LDA





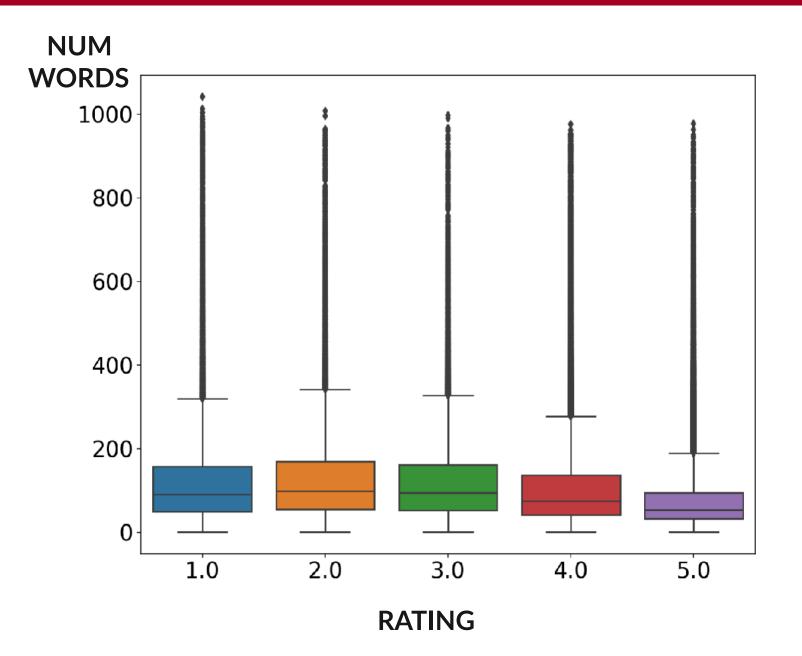


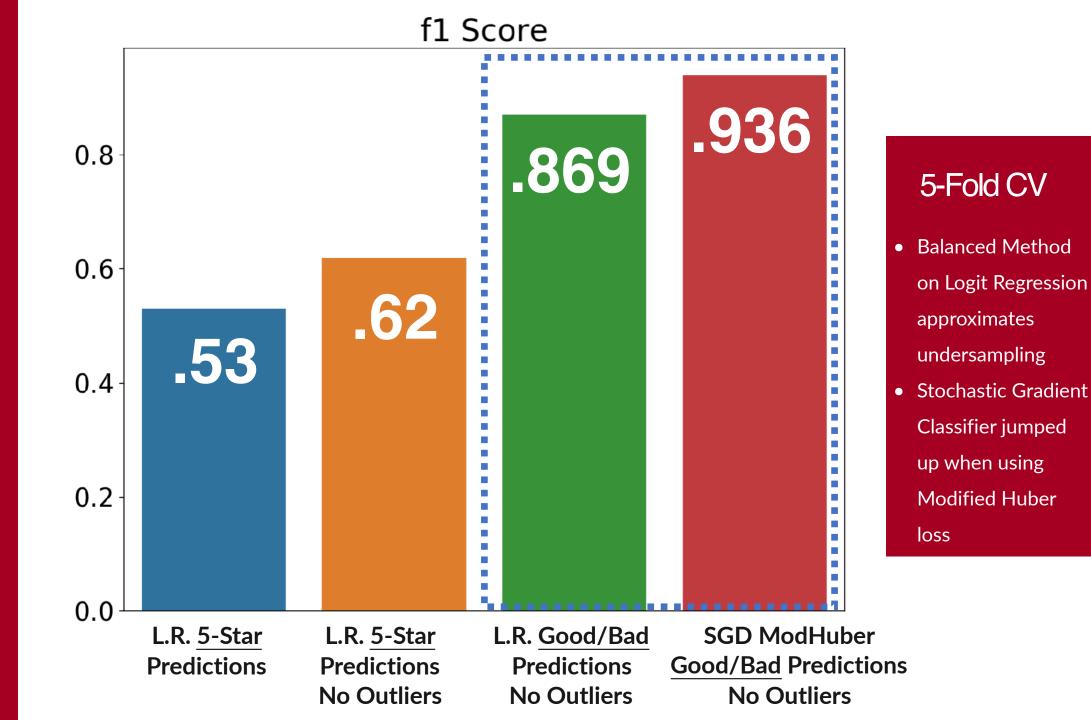
Train LDA Model on Year 1 Reviews w/ 20 Topics (implied by HDP) Apply Year 1 Reviews to trained LDA Model and get Features
(Probability Distribution over TOPICS for each Review)

Add some hand-engineered features and pass to supervised classifier CV-Loop

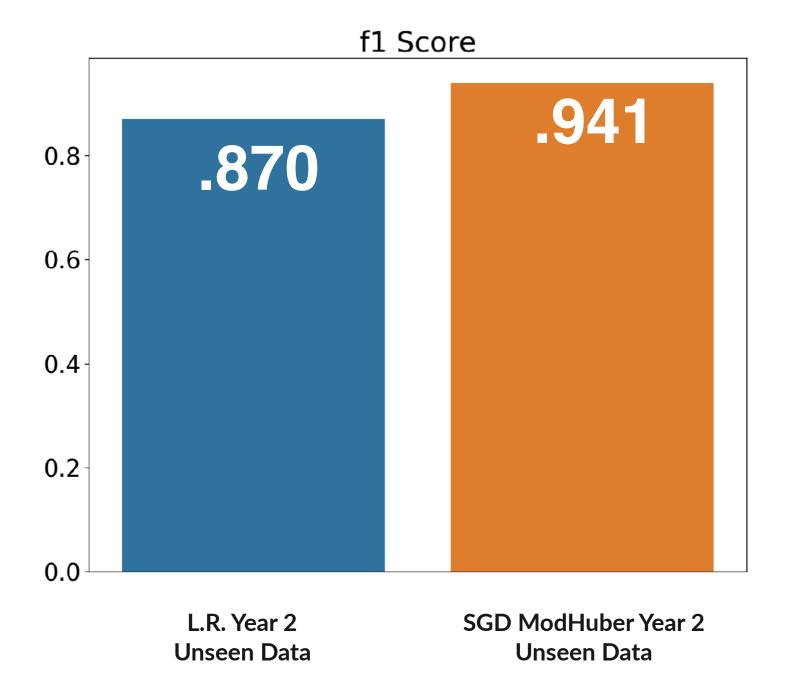
2016 Review Word Count

- Large outlier range.
- Removing these outliers helped immensely with classifier training.





UNSEEN DATA TEST RESULTS



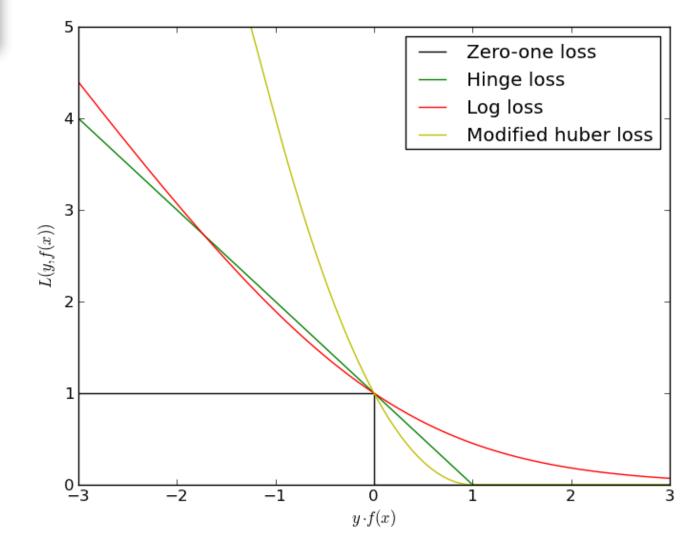
Hypothesis Testing

- Ran a McNemar
 Test (Chi-Squared
 for models) to see if
 scores are truly
 different
- Near 0 p-value on chi-square indicates models are truly different and SGD Mod. Huber wins.

What is Modified Huber Loss Exactly?

- Note Modified Huber punishes you more on outliers
- Suggests to me that that with one-byone gradient descent, it learns faster to avoid misclassifications

Source: http://ogrisel.github.io/scikit-learn.org/sklearn-tutorial/auto_examples/linear_model/
plot_sgd_loss_functions.html



Future Work







Very, very slow LDA training.
Would like to explore with
trigrams and different
lemmatization options.

More by-hand feature engineering on things like if restaurants offer takeout or not.

Extend to predict Neutral reviews as well, and also find high-dim visual projections to 2-D



Thank You!