

AMEE AMIN | GENERAL ASSEMBLY DATA SCIENCE

PREDICTING HELPFULNESS OF AMAZON CUSTOMER REVIEWS

BACKGROUND

- ▶ Amazon displays top reviews to help customers with their buying decision process. Top reviews are from Top reviewers.
- ▶ A reviewer's rank is determined by the overall helpfulness of all their reviews, factoring in the number of reviews they have written.
- ▶ Review helpfulness plays an important part in determining rank. Writing thousands of reviews that customers don't find helpful won't move a reviewer up in the standings.

EXAMPLE PRODUCT – BUBBA’S PALEO GRANOLA

Top Customer Reviews

★★★★★ **Vanilla and Uber Chocolate. The taste is fresh and really good**

By [L.A.G.](#) on July 19, 2016

Flavor Name: Vanilla | Size: 1 | **Verified Purchase**

Tried cinn-ful Apple, Vanilla and Uber Chocolate. The taste is fresh and really good. I have to be careful not to eat the whole bag! Loved all 3 favors, cannot pick a favorite. Great for a snack or when craving something sweet. Did not expect to love the taste. Just reordered 6 bags of each. Happy that it is non gmo. Wish it was organic.

[Comment](#) | 3 people found this helpful. Was this review helpful to you? [Report abuse](#)

★★★★★ **especially something that was a little bit sweet. This is so delicious**

By [SLD500](#) on April 5, 2016

Flavor Name: Uber Chocolate | Size: 1 | **Verified Purchase**

Arrived way ahead of schedule. I am new to the Paleo lifestyle and I missed the snacks...especially something that was a little bit sweet. This is so delicious!!! No guilt...just pure good food. I LOVE it!!! Thank you Bubba's Fine Foods!!!

[Comment](#) | 3 people found this helpful. Was this review helpful to you? [Report abuse](#)

PROBLEM

- ▶ When people don't vote, how should Amazon determine Helpfulness?
- ▶ New products are viewed less by customers, lower chance of review votes
- ▶ Problem: Build a model that predicts the helpfulness of a review

DATA

- ▶ Dataset consists of 568,454 fine food reviews from Amazon users left up to October 2012.
- ▶ Features include product id, user ratings, helpfulness score, timestamp, review summary, and review text.

	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summary	Text
Id									
1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	1	5	1303862400	Good Quality Dog Food	I have bought several of the Vitality canned d...

HYPOTHESIS

- ▶ Reviews that mention whether the customer would buy the product again & score the product highly, will be more helpful to other customers.

METHOD

- ▶ Make Helpfulnesses a binary classification problem
- ▶ “Bag of words” model. Tf-idf vectorizer -> Logistic regression.
- ▶ K-means clustering. Select Top 10 words -> Logistic Regression

CLEANING THE DATA

- ▶ Punctuation
- ▶ Capitalization
- ▶ HTML tags

BINARIZE HELPFULNESS

- ▶ If $(\# \text{ of helpful votes}) / (\# \text{ of total votes}) > 0.5$
 - ▶ Review is helpful
- ▶ If $(\# \text{ of helpful votes}) / (\# \text{ of total votes}) \leq 0.5$
 - ▶ Review is not helpful

EXPLORATORY ANALYSIS

- ▶ For reviews with > 10 votes
- ▶ If $(\# \text{ of helpful votes}) / (\# \text{ of total votes}) > 0.5$
 - ▶ Review is helpful
- ▶ If $(\# \text{ of helpful votes}) / (\# \text{ of total votes}) \leq 0.5$
 - ▶ Review is not helpful

EXPLORATORY ANALYSIS

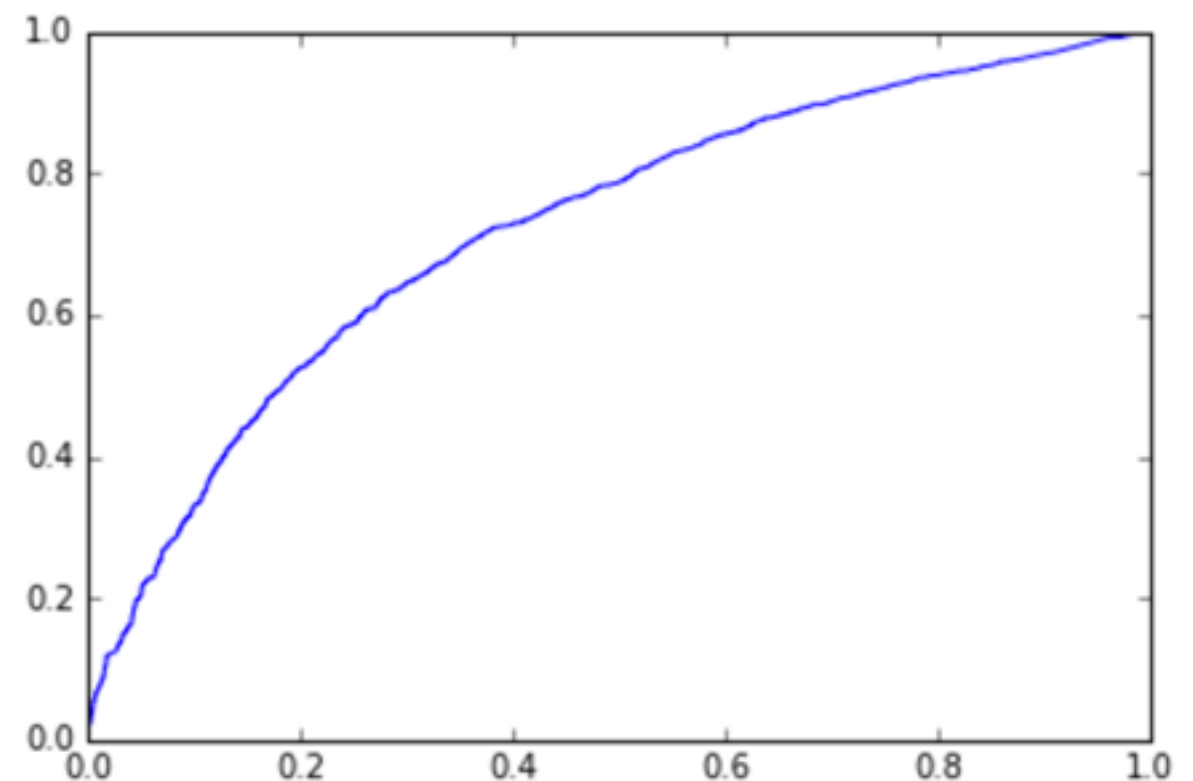
- ▶ Many more helpful reviews than not helpful, will bias the model
- ▶ Use stratified K-fold to compensate

	HelpfulnessNumerator	HelpfulnessDenominator	Score	Text
Helpful				
0	4100	4100	4100	4100
1	17363	17363	17363	17363

EXPLORATORY ANALYSIS

- ▶ Bag of words model: good AUC value, but difficult to interpret meaning from best coefficients

```
(u'eat', 1.1988757821801048),  
(u'love', 1.327079240641496),  
(u'used', 1.4182934021649025),  
(u'time', 1.4314137846470354),  
(u'amazon', 1.4488825074938934),  
(u'try', 1.7436129020024635),  
(u'make', 1.8089264785709085),  
(u'good', 1.9259597318553792),  
(u'think', 2.2960891065828037),  
(u'people', 2.4230429756217542)]
```



ROC / AUC SCORE: 0.72

K-MEANS CLUSTERING

- ▶ Hypothesis: There's a natural clustering to review vocabulary. I can use the most descriptive clusters to simplify the model.

```
model = KMeans(n_clusters=4, init='k-means++', max_iter=100, n_init=1, random_state=5)
```

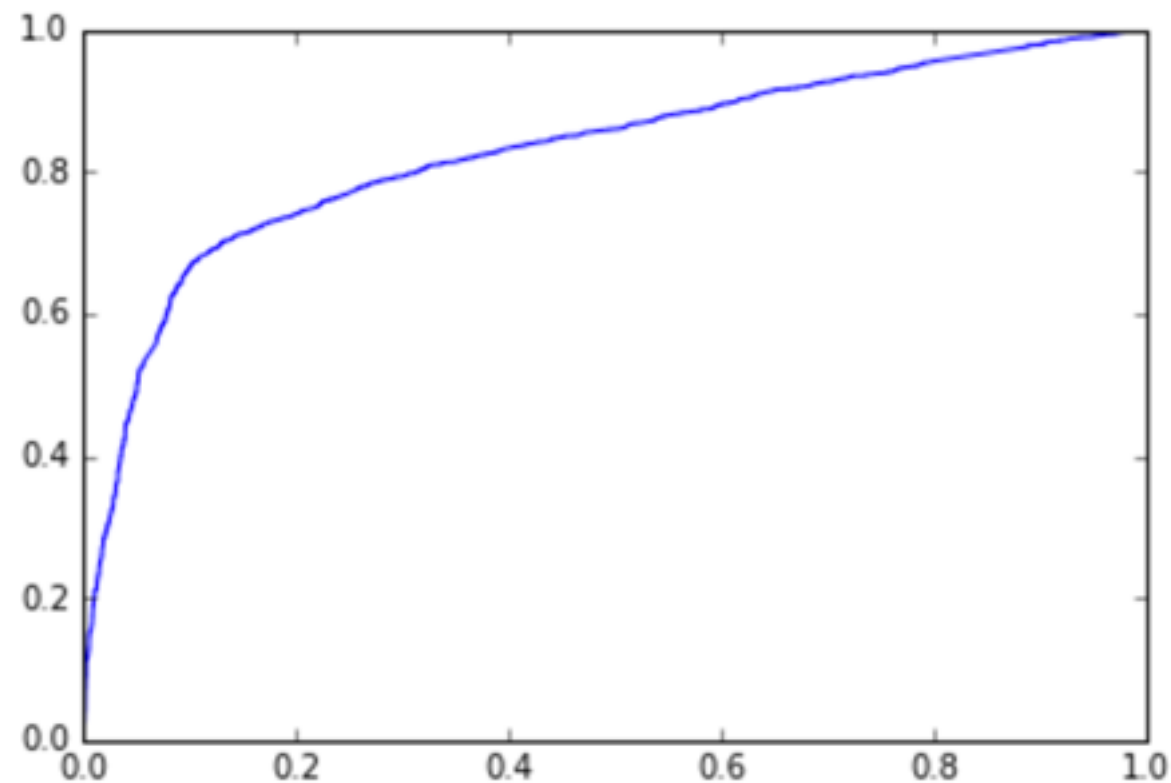
```
vectorizer = TfidfVectorizer(min_df = 0.05, max_df=0.95,  
                             ngram_range=(1, 2),  
                             stop_words='english')
```

- ▶ Select Top 10 words per cluster

LOGISTIC REGRESSION

- Features: Top cluster words + Product score

```
gs = grid_search.GridSearchCV(  
    estimator=LogisticRegression(),  
    param_grid={'C': [10**i for i in range(-5, 5)], 'class_weight': [None, 'balanced']},  
    cv=cross_validation.StratifiedKFold(dfl.Helpful, n_folds=10),  
    scoring='roc_auc'  
)
```



ROC / AUC SCORE: 0.82

(Bag of words ROC / AUC: 0.72)

LOGISTIC REGRESSION

- ▶ Best coefficients. Can interpret better than bag of words.

```
(u'price', 1.2031181875427359),  
(u'great', 1.3016982468873803),  
(u'food', 1.4150144194030678),  
(u'flavor', 1.5746769214376584),  
(u'cup', 1.9186985579466715),  
(u'br br', 2.1964875164571946)]
```

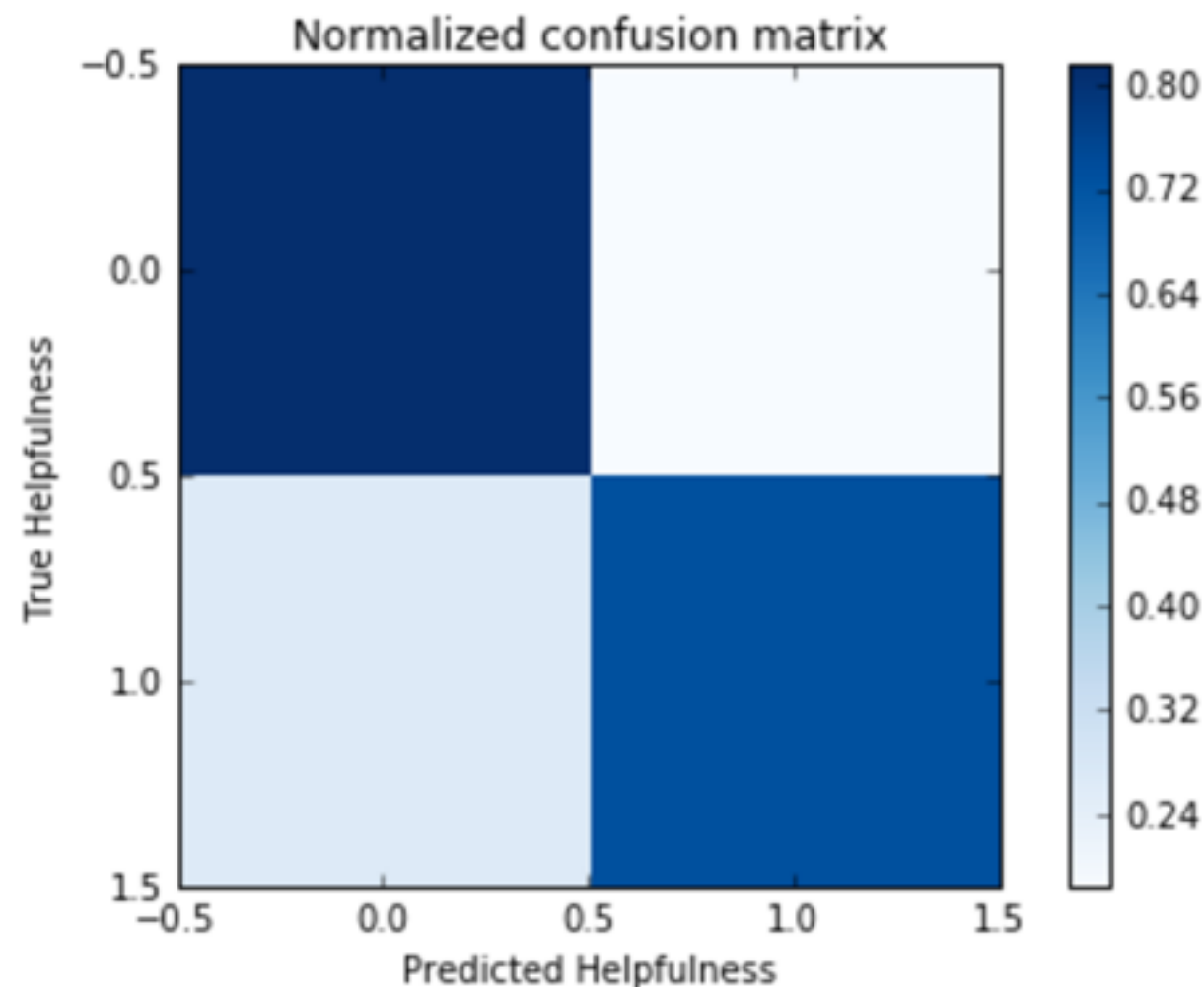
- ▶ My html parser isn't working :-/

```
(u'drink', -1.3446412638531735),
```

LOGISTIC REGRESSION

- ▶ Confusion Matrix
- ▶ Helpful reviews are being predicted as unhelpful.

Optimized against
false positives



IMPLICATIONS

- ▶ Offer customers a guide or tips when writing reviews
 - ▶ “Describe the flavor”
 - ▶ “What’s the value for the price you paid?”
- ▶ Need to understand the features, customer behavior, and domain better. More “unhelpful” reviews.
- ▶ Alternative methods: LDA for top “topical” words, Random Forest

QUESTIONS TO EXPLORE

- ▶ Is there natural clustering? Examine silhouette coefficient
- ▶ Binarize score - are reviews with score of 1 or 5 more helpful?
- ▶ Do customers tend to vote on reviews more when reviews are helpful?
- ▶ What should be defined as “unhelpful”?

ACKNOWLEDGEMENTS

- ▶ Thanks Sri!
- ▶ Thanks Dan!
- ▶ Thanks Kaggle!
- ▶ Thanks Amazon!