

Introduction

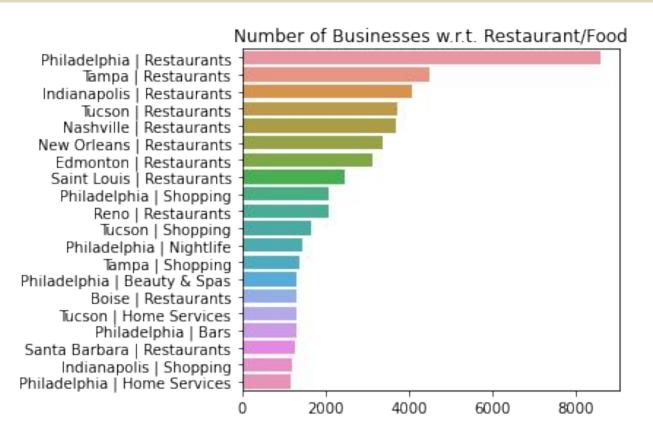
Model:

- a) Define potential customers of a restaurant as people who do not rate that restaurant now.
- b) Generate personalized wordclouds for restaurant owners in Philadelphia by predicting potential customers' star rating to the restaurants and analyzing their previous reviews.

Process:

- a) Narrow down our research to restaurants in Philadelphia in EDA & Preprocessing part.
- b) Use Matrix Factorization model and Neural Network model to build customers' star ratings prediction system.
- c) Extract previous reviews of potential customers to the similar restaurants to generate wordclouds for each restaurant.
- d) Build up Shiny App to display the personalized wordclouds.

EDA & Preprocessing

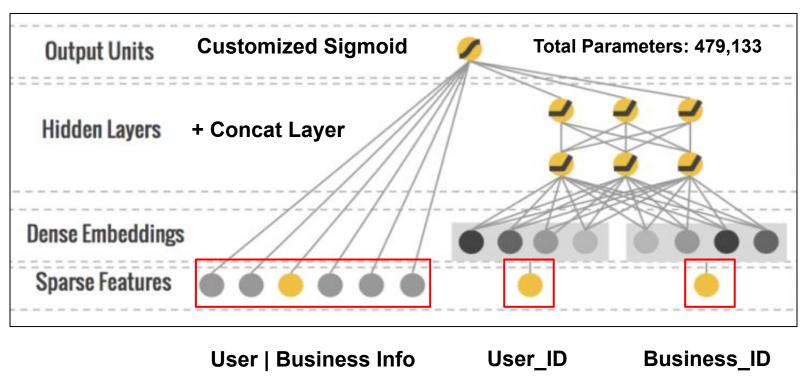


- Philadelphia has the largest number of restaurant and food entities, which will lead to a more robust model.
- Users and restaurants has more than 20 reviews are chosen to ensure an accurate model.

Recommendation System (Star Rating Prediction)

- Neural Network Architecture
- Regression Problem (1 to 5 stars)

Google's Wide & Deep Learning Architecture (Heng-Tze Cheng, 2016)



Recommendation System (Star Rating Prediction)

Train/Val/Test Set Splitting

Test: Most Recent Review for each user

Val: Second Recent

Train: Remaining

Keras functional API

Evaluation Metrics - RMSE
 Regression Problem
 Small Difference Preferred More
 RMSE 1.125

Neural Net vs Matrix Factorization
 RMSE 1.125 vs 1.134
 Neural Net model Selected

B3 **B**4 B2 **B**5 U1 ? ? U2 5 3 U3 ? 3 ? 2 1 U4 ? 5

How Do We Find Similar Restaurants?

Potential positive reviewer's negative review history

1

?

U5

?

5

Word Embedding

- 1. Select 'categories' Attribute
 - comma separated attribute
 - 150,346 entities
- 2. Build 'word2vec' model with 'gensim' library
 - category embedding vectors
- 3. Extract top 5 categories for each category
 - cosine similarity score

Category	Similarity		
Brazilian	1.000000		
Argentine	0.662890		
Peruvian	0.532551		
Basque	0.531147		
Spanish	0.501794		

4. Analyze Purple cells' Textual Reviews!

Similar to B1

	B1	B2	ВЗ	B4	B5
		H		E	
U1	3	2	3	2	3
U2	5	2	5	1	3
U3	5	3	1	5	2
U4	1	4	2	5	2
U5	5	1	1	5	3

TOP 5 Most Similar To 'Brazilian'

Potential positive reviewer's **negative** review history

4. Analyze Purple cells' Textual Reviews!

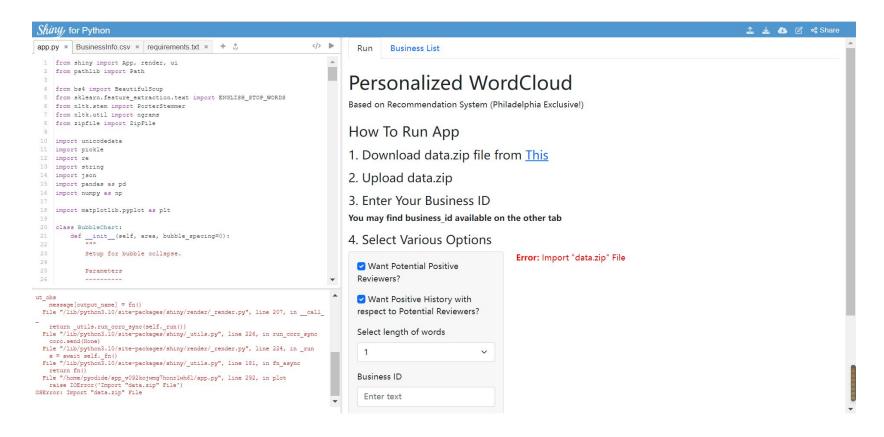
- Remove accents, white spaces, stopwords
- Build N-Gram Model
- etc.

Similar to B1

	B1	B2	В3	В4	B5
U1	3	2	3	2	3
U2	5	2	5	1	3
U3	5	3	1	5	2
U4	1	4	2	5	2
U5	5	1	1	5	3

Potential positive reviewer's negative review history

Here is the entrance interface to our shiny app:

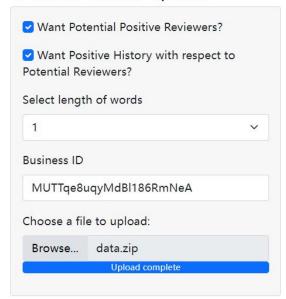


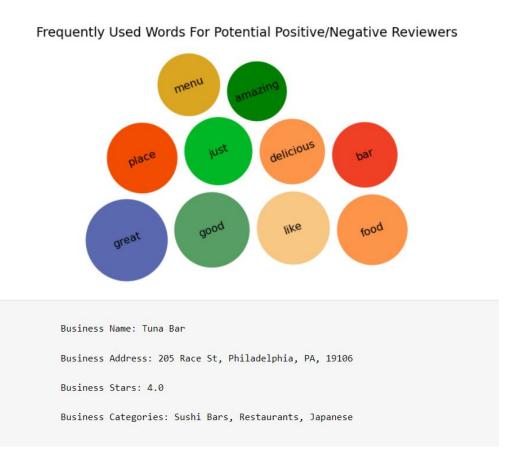
It needs some time to initialize and upload the full data set.

Then, by entering a business ID from the business list, we can get the word cloud.

You may find business id available on the other tab

4. Select Various Options

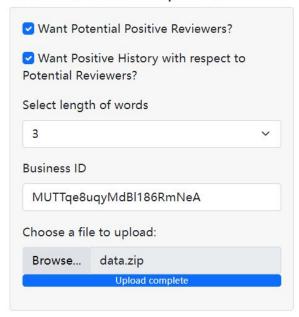


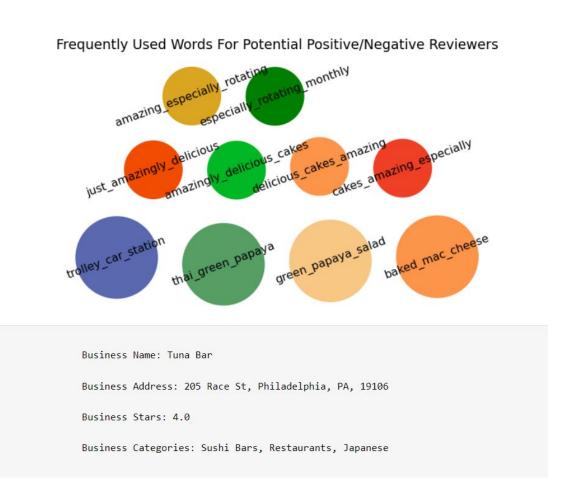


Here is a example with one word length

You may find business id available on the other tab

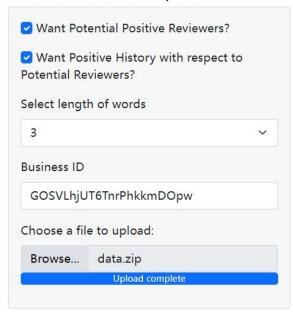
4. Select Various Options



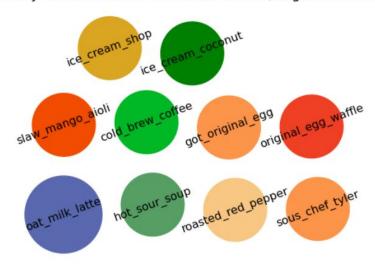


You may find business id available on the other tab

4. Select Various Options



Frequently Used Words For Potential Positive/Negative Reviewers



Business Name: Vineyards Cafe

Business Address: 847 N 16th St, Philadelphia, PA, 19130

Business Stars: 4.5

Business Categories: Coffee & Tea, Cafes, Food, Creperies, Internet Cafes, Restaurants

Conclusion

- Final model: Neural Network model
- Recommendation system: predicted unrated star reviews
- Word embedding technique: similar business categories
- RMSE: 1.125
- NLP: Generating the wordcloud
- Shiny visualization: Wordcloud for every business owner
- Next step: business with few reviews? hyperparameter tuning with Keras-Tuner?

Thank You