

Personalized Yelp Recommendations

GA Tech CSE6424 Data Visualization - Team Project

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About - Me

► Education

- Biochemical Engineering (2015)
- Masters in Analytics (2020)



► Work Experience



Virology Research (1.5 years)

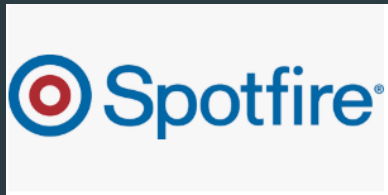
Process Scientist I (2.75 years)

Process Engineer II (1.0 years)

Consultant (2 mo)

About - Software/Hardware & Programming Experience

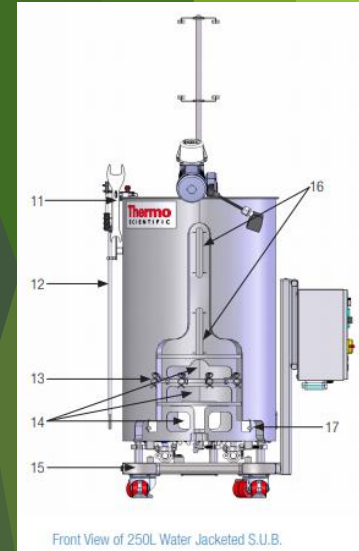
► Software



► Programming



- Equipment (5-layer flasks, Ambr15, Waves, Stainless Steel reactors, disposable reactors)



Problem Statement

- ▶ Yelp is a great resource for finding places to eat!
- ▶ However, unless you're an avid Yelp user -- Yelp cannot recommend a personalized search.
 - ▶ Multiple written multiple reviews,
 - ▶ Multiple business visits/ratings
- ▶ Heck, Yelp's Recommendations go on for pages and pages like google searches!
 - ▶ This isn't personalization more than just recommendations!

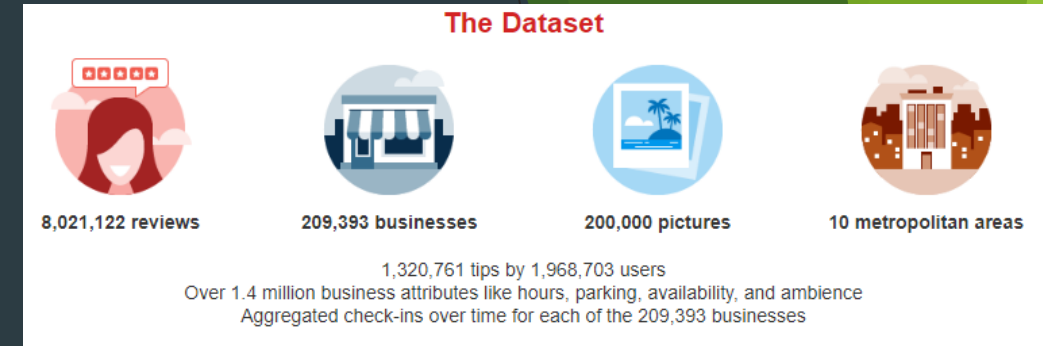
Project Goal

- ▶ Create an app that can allow users to input their food preferences & output a personalized restaurant search
- ▶ Additionally, provide metrics & limit the results as to allow users to make informed decisions quicker and with confidence!

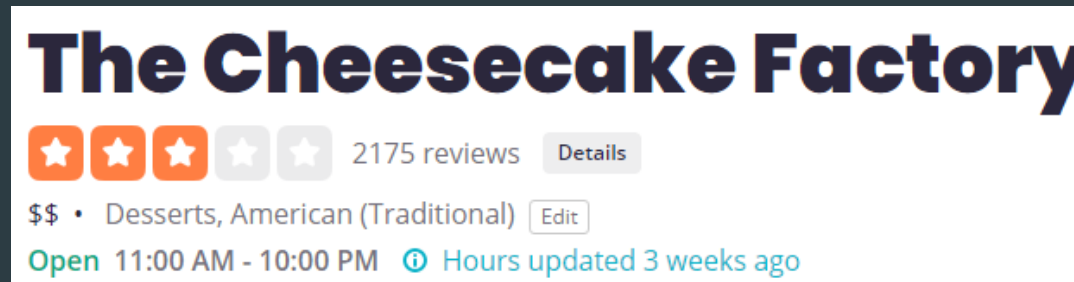
Recommendation System Details

- ▶ Package: LightFM
- ▶ LightFM maps user/item features and performs dimensionality reduction to find latent factors (see Matrix Factorization) that exist in your model
- ▶ The model of choice here was a user-based collaborative filtering model
 - ▶ A method that exploits user's features (cuisine preference) to create a model that can be used to predict for a **new user (you!)** based on their own preferences

What exactly are the inputs? How was the model trained?



- ▶ The Yelp Challenge Dataset contains lots of information
- ▶ Specifically, we used what the business tagged itself as
- ▶ Ex: The Cheesecake Factory lists under {Desserts,American (Traditional)}



- ▶ These business tags are used as inputs and the LightFM model is trained on what users that exist in the system that have reviewed +3 locations.
- ▶ Recommended restaurants are populated based on highest similarity score

Example of how User-based Collaborative
Filtering works (Next slide)

Model Input & Output Overview

User_profile = ["Japanese","Thai","Burgers","Korean","Soups"]

Feature Vector for this user:

"Japanese"	"Burgers"	"American (New)"	"Wigs"	"Thai"	"Hair Salon"
1	1	0	0	1	0

Sparse Matrix{
Dim: 1xN
Values: {("Japanese",1)
...
("Soups",1)
}
N = # unique features

All_items = {0,1,2,...,M}
M = # businesses

Model.predict(users_id=0,

items_id=

All_items = {0,1,2,...,M}
M = # businesses

user_features=



)

Sort

Top X recs

Putting it all together!

- 1) Preferences are fed into the model in the 1st drop-down menu
- 2) Recommendations are automatically populated from the LightFM model
- 3) Graphics are also automatically updated once businesses are populated

Yelp Recommender - Team99(198)



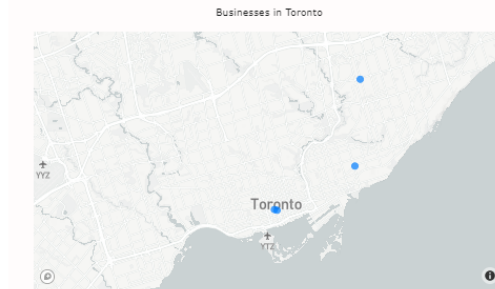
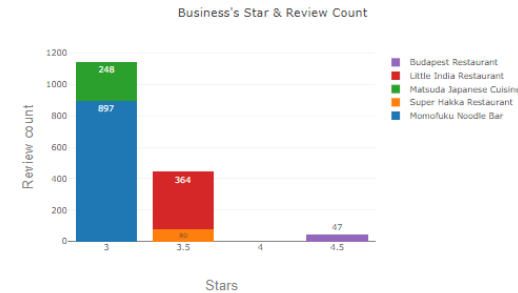
1) Select your Preferences -- As many as you like! Preferably > 5

Burgers French Active Life Hawaiian Vegetarian

2) Businesses will be populated based on your preference in ranked order (left to right) -- Feel free to add more/change the selection(s)

Momofuku Noodle Bar Super Hakka Restaurant Matsuda Japanese Cuisine Little India Restaurant Budapest Restaurant

Note: Graphics may take ~15s to load & will automatically update with any changes to above selectors.



Momofuku Noodle Bar

Most Positive Attributes

- pork
- chicken
- service

Least Positive Attributes

- bill
- disappointment
- problem

