

2018 Data Science Workshop

FoodFlix

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FoodFlix

A Food Recommender Engine (Netflix for Food)

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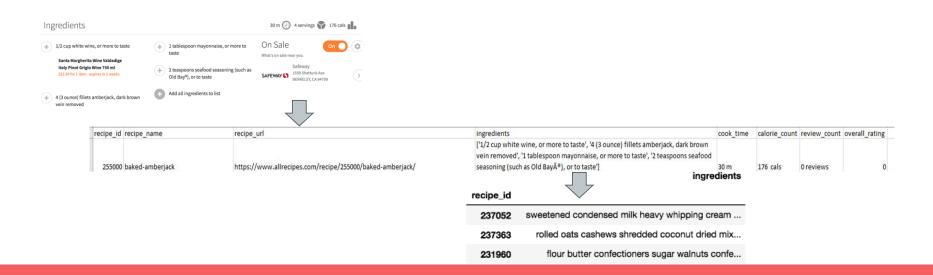
Introduction

- Objective
 - Build a food recommendation system from the ground up
- Why FoodFlix?
 - Make it easy for users to find healthy diet choices
 - Build weekly schedules or shopping lists

User's height & weight Restrictions (eg. nut) User preferences (eg. pumpkin, milk) Feedback on recommended recipes (i.e., like or dislike)

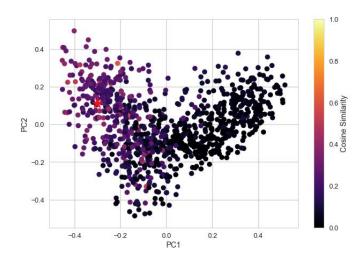
1. Obtaining data

- 18K recipes scraped from <u>allrecipes.com</u> using Beautiful Soup
 - Recipe with ingredients , Cooking time, Number of reviews, Rating, Nutrition information. Calories
- Data Cleaning
 - Removing Stop words, non-alphabetical character, recipes without ingredients, etc.



2. Feature engineering

- **1. TF-IDF**: term frequency inverse document frequency
- Word2vec: average of ingredients' word vector
- 3. Doc2vec: train recipe vectors using their ingredients



2D Principal component weights for TF-IDF vectors

3. Model Development

- 1. Find recipes with preferred ingredients using word search
 - a. Pumpkin \rightarrow use recipes that contain pumpkin in ingredients
- 2. Find recipes with calories within range set by BMI input
 - a. Height/Weight → BMI → Recommended calories range per meal → Filter recipes
- 3. Remove recipes with restricted ingredients
 - a. Nut \rightarrow Filter out recipes that contain nuts
- **4. Find recipes similar to the 'liked' ones**, using cosine similarity of TF-IDF, or Word2Vec, or Doc2Vec
 - a. Among the recipes that are filtered from (step 1, 2, 3), recommend more recipes that reflect user's personal preference shown by 'likes' and 'dislikes'

4. User interface development





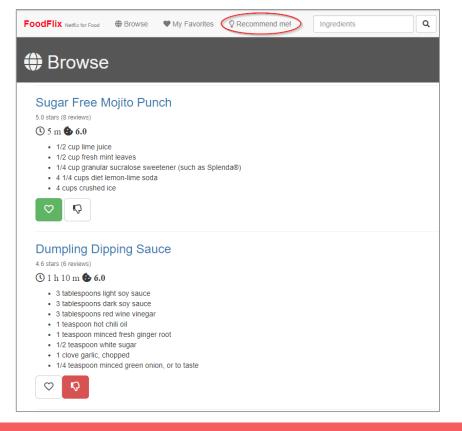


- Easy app/web development
- Individual sessions
- User Profile
- Compute BMI, Calories per day
- Model Training & Prediction
- Database to store recipe data
- Hold user account information



FoodFlix

https://foodflix-api-heroku.herokuapp.com/



- Login on website / validation
- Recipes 'Liked' or 'Disliked'
- Invoke Recommender Engine
- Predict based on:
 - Likes / dislikes
 - Calories per day
- Provide top 3 recommendations

FoodFlix LIVE DEMO



5. Conclusion

- Built a food recommendation system from data scraping to user-interface
- Evaluation results:
 - Blind test comparing FoodFlix results to randomly recommended recipes shows that it performs better than random guesses
 - Subjectively rate recommendations 0 5 based on how relevant they are

Mean recommendation rating (out of 5)

TF-IDF /	W2V /	D2V /
Random	Random	Random
3.7 / 1.53	3.7 / 1.4	3.5 / 0.4

6. Potential improvements

- Perform more thorough and quantitative evaluation (A/B testing if possible)
- Ensemble of TF-IDF, W2V, and D2V may result in better recommendation
- Emphasis on healthy recommendations (eg. nutrition information)
- Classify recipes to use in filters and recommendations
- Recommend based on similar profiles (collaborative filtering)

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Any Questions?



Graduate Data Science Organization

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