LA Restaurant Health Inspection

Recommendation System

Team HAL9000

CONTENT

Overview

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Models

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Demo Future Works

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Project Objective and Definition

Pain Points

- Public health issue has attracted more and more attention.
 Few smart city projects studied on restaurant inspection.
- For the citizens, health inspection information can be hard to find on platforms.
- For the governments, the restaurant inspection is a source of cost and efficient resource allocation is needed.

Our Project

- Predict the restaurant health violation risk
- Segment the restaurants
- Offer personalized restaurant recommendations

OVERVIEW Simplified Process Map **EDA** COUNTY OF LOS ANGELES OPEN DATA (Understand the datasets) **Feature Engineering** (Deal with null values, outliers, yelp Streamlit etc. Select/Extract/Synthesize useful features for the model. e.g. Transform the open time Selenium 9K+ restaurants Streamlit per day on Yelp to open hours (Imitate a human 13k+ inspection (For website) records per week) clicking on a web page) Customers EDA & Feature Record Problem looking for Integration & Definition healthy Engineering Linkage Data Collection Modeling restaurants Visualization ML/DS Skills Government Domain Knowledge Restaurants' **Prediction NLP** owners (Random Forest, Boosting, Neural (Text preprocessing Networks, etc.) RLTK and cleaning) Segmentation (KMeans, LDA Topic Modeling) **RLTK** Recommendation (Blocking, (Content-based) **Entity Linking**) **Optimization & Evaluation**



Prediction

Goal: Classify the restaurants into 3 health risk levels (low/medium/high risk)

Features:

Baseline.

Final Model

- Categorical: Location, Price, Size, Photos(bool), Take_out(bool), Q&A(bool)
- Numerical: Open_duration, employees, Review_counts, Review_sentiment_polarity

Machine Learning Models & Evaluation Metrics:

Model Accuracy Roc auc Logistic Regression 0.835 0.58 SVM 0.839 0.63 Random Forest 0.842 0.67 XGBoost 0.827 0.71 **Neural Networks** 0.51 0.857

Segmentation

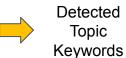
Pipeline: PCA - KMeans - t-SNE & Topic Modeling

	Model	Objective	Evaluation		
X	PCA	Dimension Reduction	Explained variance on the first 3 PCs: 0.5		
	KMeans	Learn the representation of the data	Scree Plot: K=5		
× ×	t-SNE	2D visualization of the clusters based on PCA	KL divergence = 0.47		
\	Topic Modeling	Extract the keywords from comments	We sampled 10 restaurants and check if the keywords are really included in the descriptions of the restaurants. Included Rate = 70%		

Number of Correct Topic/ Detected Topics

Randomly Selected Restaurants



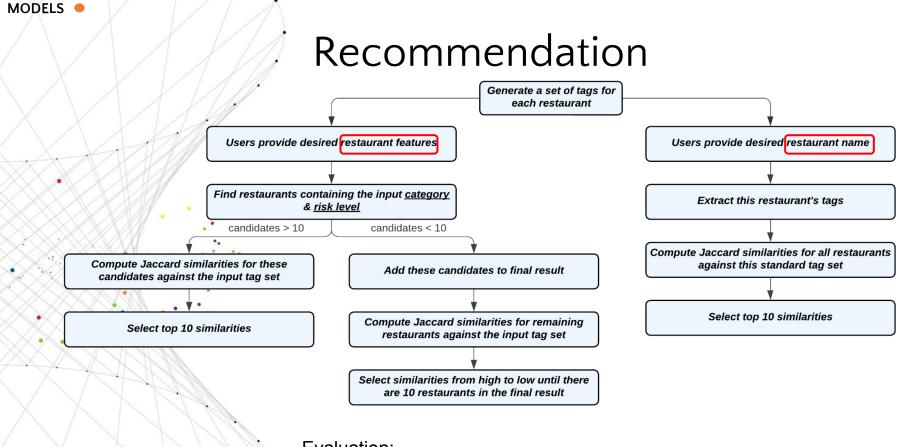


e.g. 10 Restaurants

Topic Model Really in the Yelp **Description?**



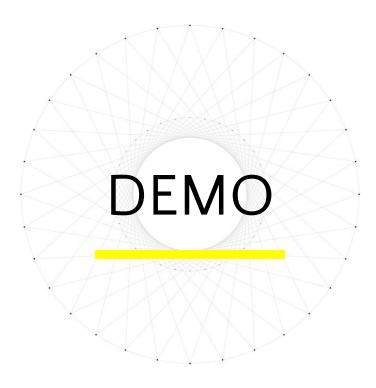
		restaurant	cluster	is_included
	0	Pharaoh Karaoke Lounge	0	False
	1	Sanamluang Cafe Hollywood	0	True
Results	2	Eduardo's Border Grill	1	True
	3	Mom's Donut and Chinese Food to Go	1	False
	4	Ministry Of Coffee	1	True
	5	Papa Johns Pizza	2	True
	6	Moishe's Restaurant	3	True
	7	Hot Thai Restaurant	3	True
	8	Kai Ramen West Hollywood	3	True
	9	Ginger's Divine Ice Creams	3	False



Evaluation:

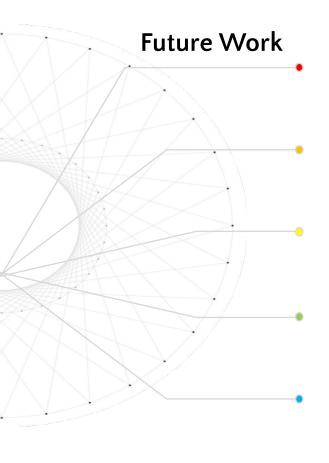
- Among all 10 recommended results, how many are actually relevant?
- Ran some tests average **74%** relevancy.

Backup Demo



Web application

File Use Case



- 1. Optimize UI adding images, organize layouts, etc.
- 2. Add an option for users to provide feedbacks on the models

- 3. Add a rating system for users to provide feedbacks on the recommendation systems
- Crawl more features to feed into the model to improve the performance
- 5. Integrate ChatGPT for explaining the final results

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