Rest o Recommend

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Motivation:

Many dream of owning a restaurant. However, restaurant business is hard. According to CNBC, 60% of restaurants fail in first year and 80% of restaurants fail before their fifth year. Main reasons of failure are: poor location choice, unable to get actionable insights from customer feedback and lack of competitive intelligence. We want to help restaurant owners use data to make decisions, thereby increasing their chance of success.

Approach:

Currently, restaurant owners rely on experience, intuition or approach consulting firms such as restaurantowner.com to run their business. Our approach, when compared to others, is as follows:



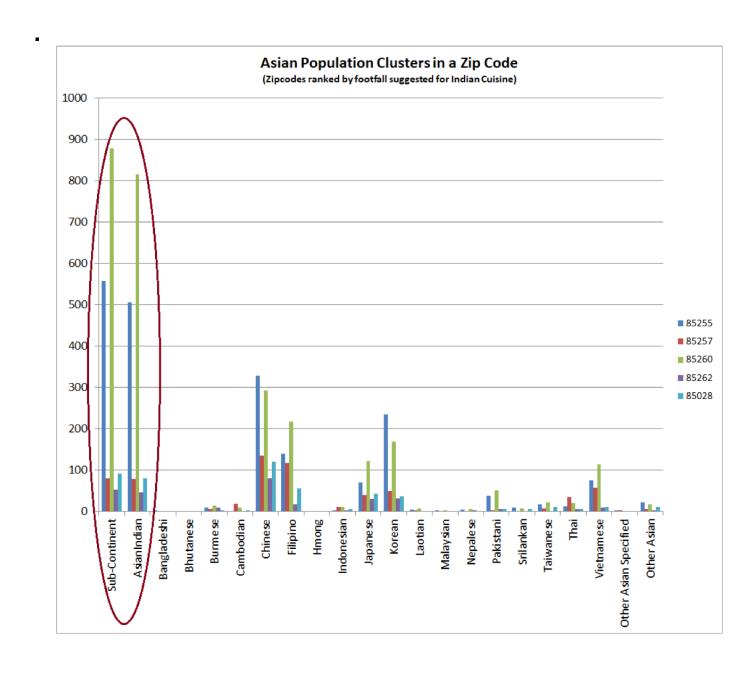
Data:

| | Number of restaurants | 11072 |
|---------------------------|--|--|
| Yelp Academic Dataset: | Average Number of reviews per restaurant | 489 |
| | Number of Cuisines accounted for | 30 |
| | Number of restaurants attributes analyzed | 36 |
| | Source | Yelp.com |
| | Method obtained | Direct download and Fusion API |
| | size on disk | 1.3 GB |
| | Attributes used | Restaurant facilities, cuisines, customer reviews, footfall information. |
| Walkscore: | Attributes used | Walk Score, Zip code |
| | Source | www.walkscore.com |
| | Method | Scraped using the API. |
| Census Data: | Attributes used | ethnicity distribution, demographic details, median income, # of establishments |
| | Source | Census 2010, ACS 2016 (www.census.gov) |
| | Method obtained | Scraped using API |

Experiments and Results:

Location Prediction:

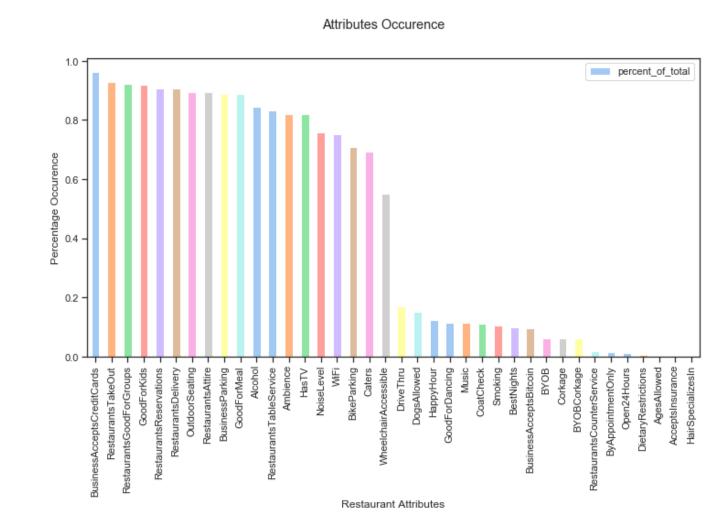
We obtained an R-Square of 42.47%, RMSE of 2.67 for footfall using XGBoost Regressor. We observed that the zip codes predicted for higher footfall have a good representation of the ethnic groups that are associated with the selected cuisine

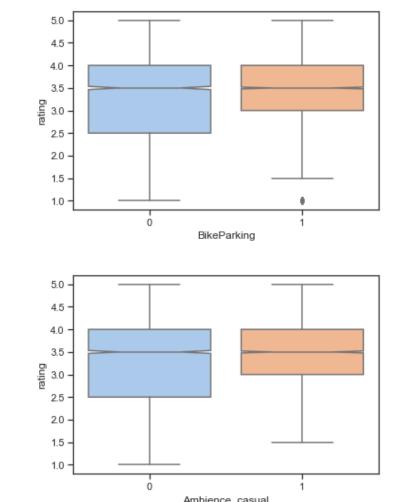


| Model | RMSE | R-Square |
|-------------------|------|----------|
| Linear regression | 1.7 | 24.19% |
| Random Forrest | 2.72 | 38.20% |
| XGBoost | 2.67 | 42.47% |

Attributes Evaluation:

We evaluated whether customers care about the attributes predicted by our model. We mined through reviews for 25 restaurants to evaluate the topics suggested by our model and observed the following prediction accuracy:



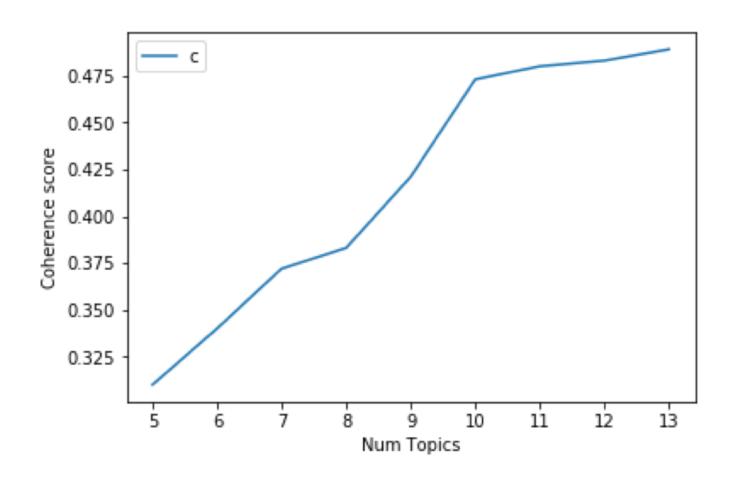


| Percent of Cases | Result |
|------------------|---------------------------------------|
| 56% of cases | 3 or more suggestions were relevant. |
| 12% of cases | 2 out of 5 suggestions were relevant |
| 8% of cases | 1 out of 5 suggestions were relevant |
| 24% of cases | None of the suggestions were relevant |

| Attribute | Sample review excerpt | |
|------------|--|--|
| 'Ambience' | This place feels a bit out-of-the way, with a small interior and not a whole lot to the ambiance. The sushi, however, is superb. | |

Text Analysis:

To determine the best number of topics to be retrieved from the reviews we computed topic coherence metric with different number of topics and obtained the optimal topic coherence as 10.



Summary Review: All and all the place is a great burger at a decent price. I would certainly recommend.



Conclusion:

The closest work related to our project is done by Lu et.al. But the work mainly concentrated on predicting whether an existing restaurant would be successful or not based on the yelp reviews. Our work is novel as we tried to suggest restaurant location where it would earn more revenues based on footfall metric. We later used LDA for topic modeling and TextRank algorithm for automatic summarization. Our model is unique as it combines various models and gives a one-stop solution for various restaurant setup and improvement requirements.