





Explore gym business--- Give suggestions to gym owners to improve their business

- 1. Do explanatory data analysis
- 2. Build models to predict stars from reviews
- Extract features from data and do association rules mining





Data Cleaning - preparation



- Select the 'Gym' data from data set: Review_train: 35033 rows 26.1MB
- Split the attributes of business data and add it to the data as columns, such as 'WheelchairAccessible';
- Keep the attributes having missing values less than 35%: 'BusinessParking', 'ByAppointmentOnly', 'GoodForKids';

Data Cleaning - from review to words

Worst Gyms I met. And I would not come again.

Change n't into not

['Worst', 'Gyms', 'I', 'met', 'And', 'I', 'would', 'not', 'come', 'again']

Remove punctuation

['worst', 'gyms', 'i', 'met', 'and', 'i', 'would', 'not', 'come', 'again']

Convert to lower case

['worst', 'gym', 'meet', 'would', 'not', 'come']

Lemmatization

Worst Gyms I met. And I wouldn't come again.











Split into words

['Worst', 'Gyms', 'I', 'met', '.', 'And', 'I', 'would', 'not', 'come', 'again', '.']

Remove non-alphabetic

['Worst', 'Gyms', 'I', 'met', 'And'. 'I'. 'would'. 'not'. 'come'. 'again']

Remove stop words

(except for not no nor)

['worst', 'gyms', 'met', 'would', 'not', 'come'l

Analysis of reviews wrt words



- Word cloud for overall reviews
- ★ Top 5 frequent words in reviews of different stars



- ★ Top 3 keywords in all review
- ★ Top 3 keywords for reviews of different stars



Use word2vec in python to create feature vectors for all words

Analysis of reviews wrt words





freque ncy

stars

| | not | not | not | not | not |
|---|------------|--------|-----------|---------|-------|
| | no | like | like | like | great |
| • | membership | no | room | great | love |
| | month | people | machine | room | like |
| | tell | room | equipment | machine | work |
| | | | | | |

* ** ** *** ***



Analysis of reviews wrt words

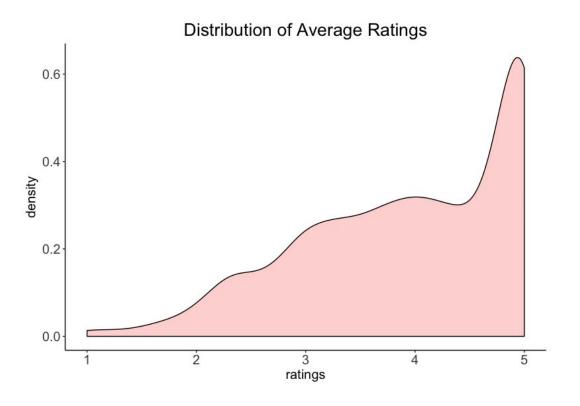


| TFIDF | wigwam | funk | normal | valued | love | great |
|-------|-----------|----------|--------|----------|------------|------------|
| | disgusted | parking | wifi | helpful | convenient | hospitable |
| | invested | terribly | fix | flawless | overall | affordable |
| stars | * | ** | *** | **** | **** | all |



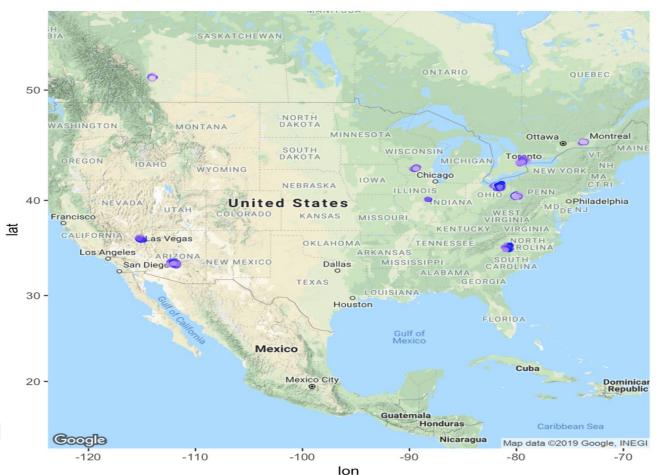
Analysis of stars - Distribution of Average Ratings

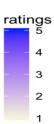




- Calculate average ratings for each business
- Most businesses in gym field have pretty high ratings

Distribution of gym businesses



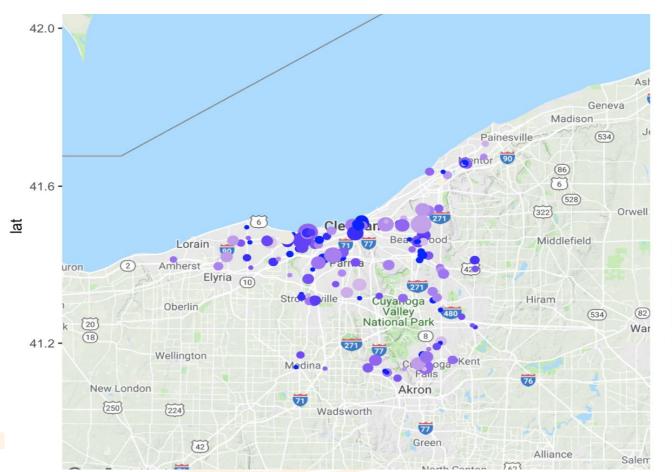




Distribution of gym businesses in Cleveland

num

ratings

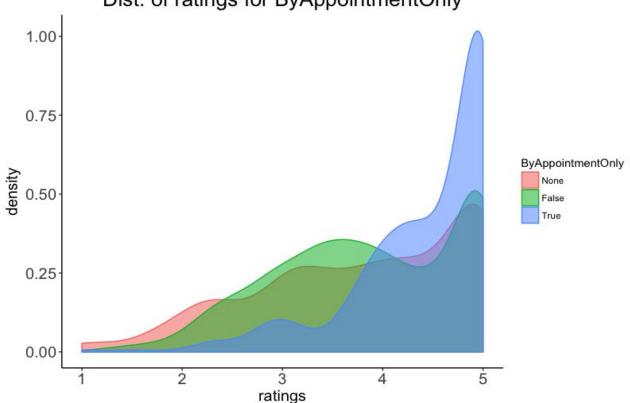




Analysis of stars - Ratings w.r.t. attributes



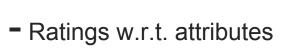




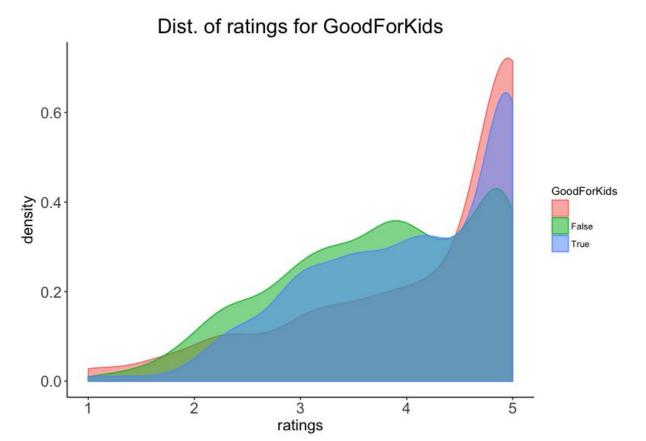


Those require appointment only tends to have higher ratings

Analysis of stars - Ratings w.r.t. attributes







No significant differences about this attribute

Analysis of stars - Ratings w.r.t. attributes





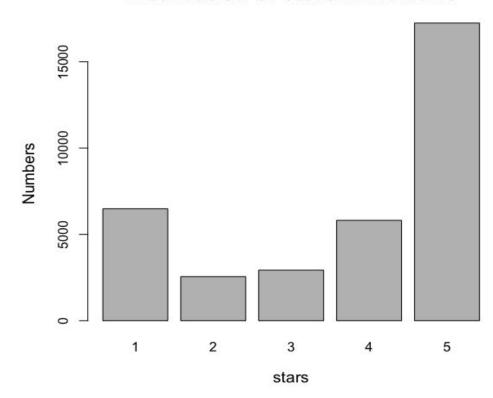
Those have lot parking and street parking tend to have higher ratings



Analysis of stars - Ratings w.r.t. reviews



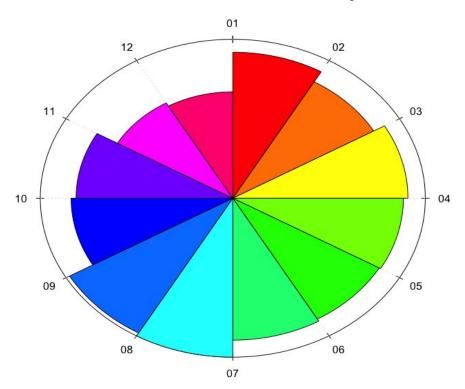
Distribution of Stars in Reviews



Analysis of stars - Reviews distribution by month



Number of reviews for each month of the year

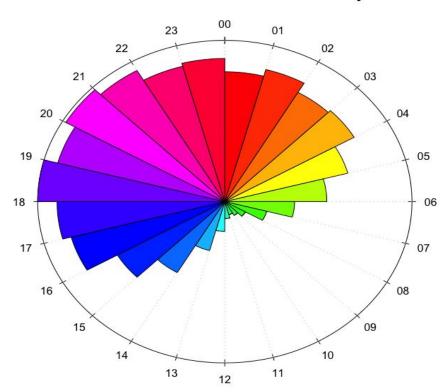


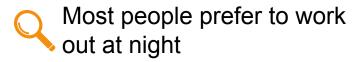
- People tend to start working out in January
- People tend to stop working out in September
- The number of reviews reaches to its maximum in July

Analysis of stars - Reviews distribution by day



Number of reviews for each hour of the day







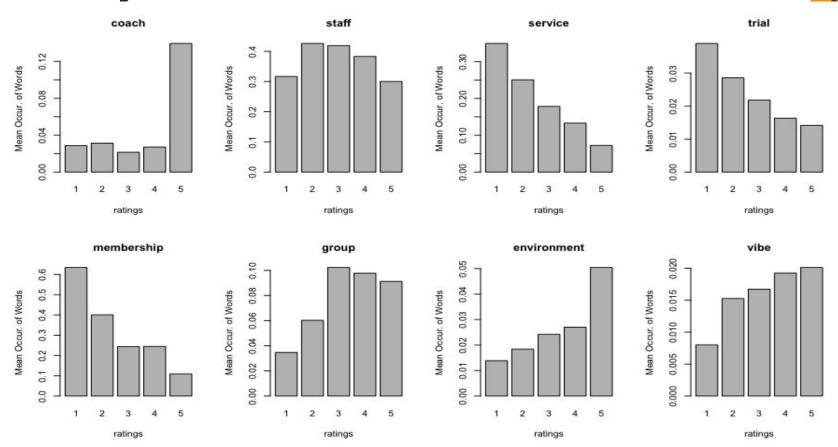
Analysis of stars - Mean occurrence of words





- Find some common nouns in gym from background knowledge.
 (coach, environment, membership etc.)
- Calculate mean occurrence of each word among ratings.
- Describe potential relationship.

Analysis of stars - Mean occurrence of words



Future Plan



Feature based summary

- Find features from data
- Identify sentences discussing these features
- Identify positive or negative sentence



Prediction Models

- Words embedding and frequency
- Logistic Model
- Neural Network



