# Data Mining Capstone Task 1

September 14, 2015

#### 1 Abstract

# 2 Implementations

In this task, I used *python* toolkits to perform the analysis. In particular, I used packages such as *gensim*, *sklearn* for the topic extraction.

For data visualisation, I used D3 to do all the drawings.

## 2.1 Topic mining of all restaurant reviews (Task 1.1)

To understand what people are talking about among the review data, I used topic model LDA to extract 10 topics from all the reviews that are for restaurants.

I applied *TfidfVectorizer* to vectorize the raw review data, after filtering all reviews and only collect those for restaurants. The TF transformation was made to be linear, and I enabled IDF reweighting. I specified the n-gram range to be 1 to 2, i.e. terms with 1 or two words were collected.

By using D3, I applied  $word\ cloud$  to visualise the data. Essentially, each cloud represents a topic. The font size shows the significance of the corresponding term in this topic. Ten colours are used to distinguish the topics.

Figure 1 shows the result.

Here are several observations:

1. Cuisines / foods are very popular topics – topics 4, 5, 7, 9 more or less emphesize a particular cuisine, e.g. Mexican tacos, pizza, Japanese sushi, and Thai / Chinese food, while topics 0, 6, 8 emphesize certain foods or drinkings.

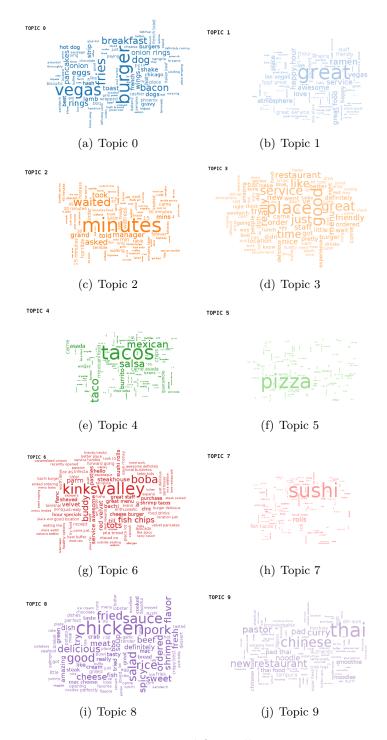


Figure 1: Ten topics mined from all restaurant reviews

- 2. General comments (usually good) to the restaurants topics 1 and 3 clearly show a good impression, while topic 2 possibly suggests an inferior impression in terms of waiting (and then asking for managers). In particular, topic 2 also suggests that *time* is a very important topic, or factor, when customers are reviewing a restaurant.
- 3. Locations are also mentioned, especially "Las Vagas Valley", spreaded amongst topics 0, 1, 6.

### 2.2 Topic mining of positive and negative reviews (Task 1.2)

In this task, I managed to explore the topic distribution for subsets of all the reviews. In particular, this section introduces the observations made upon subsets of negative reviews (reviews with star number  $\leq 2$ ) and positive reviews (reviews with star number  $\geq 4$ ).

The topic model used is still LDA, with identical configurations as in task 1.1.

Figures 2 and 3 show the result.

From these results, it is shown that:

- 1. Both positive and negative reviews still talk much about food or cuisines themselves frequently. For instance, in negative review topics, tacos, crab legs, sushi, carne asada, pizza, hot dogs are the top topics. While for positive reviews, still pizza, sushi, Indian food, chicken, breakfast, oysters are mentioned.
- 2. Compared with all reviews, these subsets now shows something different in different subsets.
  - In the positive review topics, there is no negative impression when looking at the top words in each word cloud of each topic. And even for the positive topics, the phrasing are quite general ("great place", "amazing", "definitely coming") and not touching specific factors that influence customers' rating.
  - However in the negative review topics, many specific terms indicate what the customers are looking for "portion size", "short ribs", "limited menu", "service", "time", i.e. amount of food, menues, service, and time. At the same time, general comments (e.g. "just okay") can still be seen.

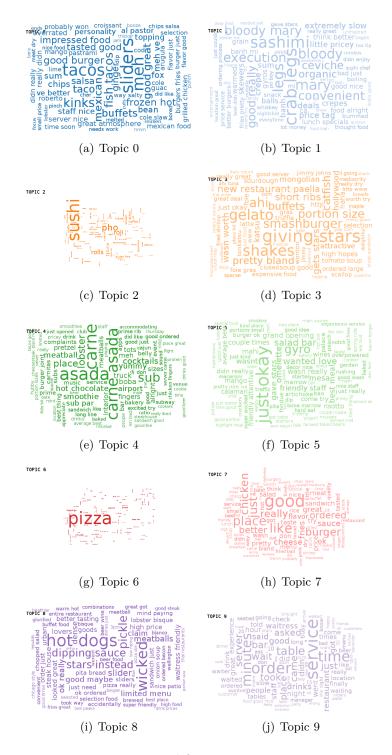


Figure 2: Ten topics mined from negative restaurant reviews

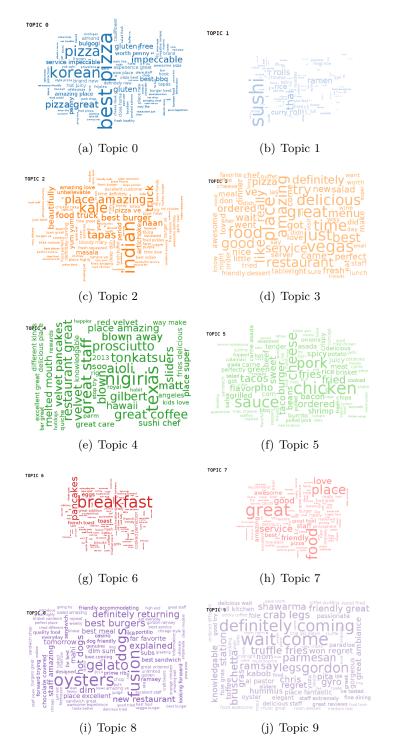


Figure 3: Ten topics mined from positive restaurant reviews  $\overset{\cdot}{5}$ 

#### 2.3 Topic mining of reviews from different cuisines

Since reviews are made to different restaurants, and different restaurants do have different categories, or cuisines. Intuitively, one could imagine the topics mentioned in the reviews could vary.

Here I did some topic extraction upon reviews from different categories, and hereby present the comparison between category "Automotive", "Tours", and "Seafood Market".

Figure 4 shows a collection of all three categories, 2 topics sampled for each.

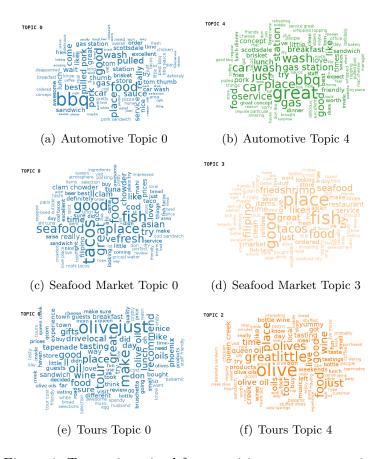


Figure 4: Ten topics mined from positive restaurant reviews