Fitness Studio Preference Analysis

Kushal Shah ks5574@nyu.edu

Rupesh Bagwe rpb337@nyu.edu

Vinayak Natarajan vn686@nyu.edu

Abstract:—Fitness Studios play an important role in the lifestyle of every individual. These Fitness Studios can be Gyms, Yoga centres, Martial Arts centres, Dance centres, clubs for swimming and so on. It is observed that the choice of the people for choosing the fitness studios for their respective workouts are generally influenced by the quality of some specific fitness studio in their vicinity. So proper business knowledge is required to implement any specific decision regarding marketing or even setting up a new business. This Analysis work is aimed at securing investment capital from Business developers and planning the marketing strategies accordingly.

1. Introduction

The Fitness Studios in any locality can sometimes play a crucial role in the fitness choices of the people living in that locality. As people prefer to workout on a daily basis to maintain a healthy lifestyle, the Fitness Studios need to be close to their homes and if not people prefer some different kind of workout mode rather than travelling to some distant fitness studio. The main theme of this study is to analyse the preferences of the people on what kind of Fitness Studio do they use and thus secure private business setups in that locality based on the insights received from this analysis. This can also help other local businesses to give targeted advertisements and change their business plans accordingly in that area. So a detailed analysis on people's choices for Fitness Studios and the kind of ratings received by the existing Fitness Studios is the primary goal of this study.

2. Problem Statement & Formulation

2.1 Problem Statement

The formal problem statement being tackled is, Leveraging Fitness Studio Preferences patterns of people to attract private investments in a specific locality and for existing businesses to adopt newer plans based on the insights of this analysis.

2.2 Formulation

This process can be performed by aid of external data or corporate investors who can choose to "Promote Certain Regions". An example if a locality in New York has great reviews for a gym and many people are using that gym then secondary businesses like gym supplies can be promoted in this locality. And if this region has bad reviews for gym, then newer business opportunities for setting up gym can be promoted in this region.

3. Datasets and Tools

3.1 Datasets

We are primarily looking at the Yelp data but have possible plans for other datasets. Following is a complete list:-

- Yelp business data
- Yelp review data
- Yelp User data

3.2 Tools

Major tools that have been used are:

• PySpark

- SparkML
- ElasticSearch
- Kibana

4. Architecture & Proposed Approach

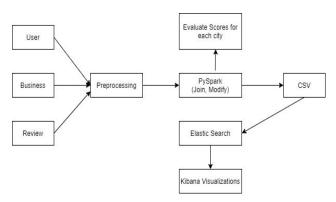


Figure 1: Architecture

This work leverages Pyspark and SparkML to guide the data flow to calculate scores for each Fitness Studio in each locality. ElasticSearch and Kibana are used for visualization of the Analysis. Figure 1 shows the layout of the architecture being developed.

4.1 Data Cleaning

The data we get is a dump from Yelp Business data. Figure 2 shows the raw data. We have around 10GB of data which is unordered, un-grouped, and full of issues. Figure 3 shows the data post cleaning. The Figure 4 shows the exact flow of the score generation model in our project. For this microanalysis we performed the following operations:

- Broke the data apart into separate lines
- Joined the datasets. (a)Yelp Business Dataset and (b)Yelp Review dataset
- Filtered based on the Fitness Studio types
- Filtered according to the positive and negative reviews.
- Used RegexTokenizer, StopWordRemover and Word count functions to find the occurences of Fitness Studio names.
- Took top 1000 words and found the points/score for each word.
- For required city and fitness studio type, these steps are repeated to find score of that studio.



Figure 2: Raw Data

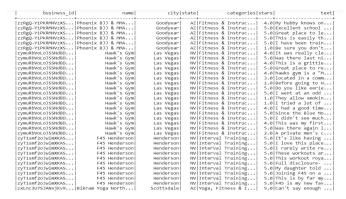


Figure 3: Filtered Data

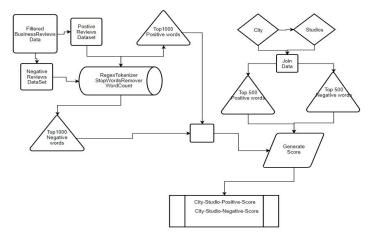


Figure 4: Studio Preference Microanalysis

4.2 Producer - Consumer Model

The Producer Consumer model relies on a Producer, a Consumer and Multiple topics. Currently there is a single producer and a single consumer. This might scale because the producer for the model is the data that is city wide. Later on the producer model can be expanded by having precise data from each zip code. The Consumer can then perform the data cleaning and manipulation part and then the data can be passed to the analysis engine.

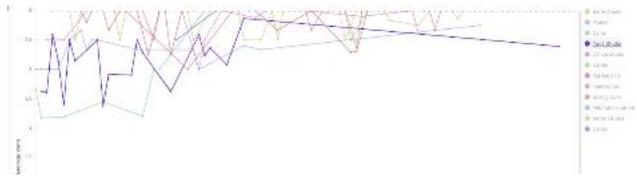


Figure 5: Line chart for Average Stars received by each Fitness Studio type

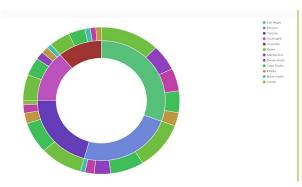


Figure 6: Donut Chart for Fitness Studio types per city

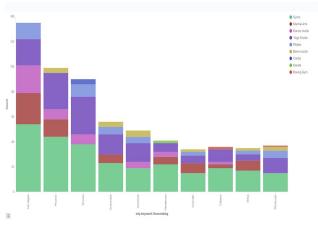


Figure 7: Stacked bar chart for Fitness Studio types per city

4.3 Analysis

ElasticSearch is used to make the dataset ready for visualization using Kibana. The Steps used in ElasticSearch are:

- •Created es mapping to create a document for every row
- •Inserted each document into elasticsearch
- •Created an es index for data aggregation and visualization

The visualizations created using Kibana helps us understand the trends of preferences for Fitness Studios. Figure 7 is a stacked bar chart and each bar represents a city. Each city again shows the distribution of various Fitness Studios in that city. Figure 6 is a donut chart where inner donut

represents cities and outer donut represents Fitness Studio types. Figure 5 in a multiple line chart. Here y-axis represents the Average score for each of the Fitness Studio and x-axis represents time, The line chart shows the trend of changing average scores over time for each Fitness Studio type.

5. Achieved Goals

As a result of the Microanalysis (figure 4) done, we have the keywords that are derived from positive reviews of the customers and keywords that are derived from negative reviews of the customer.

cspwordpoints	csptotalocc	word
0.024875159679479734	1071	gym
0.017466031819765415	752	great
0.012170479619091859	524	like
0.012147253512948554	523	place
0.012054349088375334	519	class
0.011961444663802114	515	fitness
0.011822088026942283	509	get
0.011752409708512369	506	classes
0.010567878295203809	455	time
0.009940773429334571	428	one
0.009894321217047962	426	workout
0.009824642898618046	423	really
0.009731738474044826		love
0.009104633608175589	392	staff
0.008059458831726861	347	always
0.00801300661944025	345	work
0.007478806178144234	322	people
0.007246545116711183	312	friendly
0.007200092904424573		good
0.007153640692137	308	also
0.006828475206131692	294	best
0.006294274764835676	271	clean
0.00627104865869237		fun
0.00627104865869237		feel
0.005852978748112879	252	training
0.005783300429682964		nice
0.005713622111253048		well
0.005713622111253048	246	first

Figure 8: Positive Keywords

The above table shows the positive keywords table and similarly there is a table with negative keywords. This gives us the sense of the number of occurrences of these words in the positive reviews and the last column is the score of that word with respect to the total number of occurrences of that word in the entire text.

```
for items in finalscores:
    print(items)

('Phoenix', 'Gym', 'pos', 2.621279999999983)
('Phoenix', 'Gym', 'neg', 2.837080000000007)
('Phoenix', 'Yoga', 'pos', 2.8239299999999)
('Phoenix', 'Yoga', 'neg', 2.81131999999999)
('Tempe', 'Yoga', 'neg', 2.71541)
('Tempe', 'Yoga', 'neg', 2.84229999999996)
('Tempe', 'Gym', 'pos', 2.6683800000000004)
('Tempe', 'Gym', 'neg', 2.764079999999999)
('Tempe', 'Dance Studios', 'pos', 2.28832)
('Tempe', 'Dance Studios', 'pos', 2.28832)
('Pittsburgh', 'Gym', 'pos', 2.6739699999999)
('Pittsburgh', 'Gym', 'pos', 2.759480000000001)
('Pittsburgh', 'Yoga', 'pos', 2.76892)
('Pittsburgh', 'Yoga', 'neg', 2.771429)
```

Figure 9: Final scores

The second part of microanalysis is majorly for the generation of score pertaining to a specific fitness studio in a specific city. So figure 9 shows us the positive and negative scores of fitness studios in some cities. Looking at the first 2 results we see that yoga has a more positive score than gym in Phoenix which implies that people in Phoenix prefer Yoga centres over gyms.



Figure 10: Negative Keywords

Figure 10 shows the Negative Keywords which are visualized in this format to make it easy for the user to understand.

Thus using the outcomes shown in figure 8, 9 and 10, we come to know the keywords present in positive/negative reviews and thus existing fitness studio managements can work on these factors to improve the quality of their service.

The score based on city and fitness studio makes it very easy for the entrepreneurs to study the trend in some specific city before and analyse the business requirements.

6. Conclusion

A model for finding the scores for Fitness Studios in different cities by analysing the user reviews of that specific fitness studio in that city. Our main goal was to provide the entrepreneurs with an analysis study of the preferences of people when it comes to choosing fitness studios and that we have achieved using Kibana visualizations. The extraction of positive and negative keywords from the user reviews is also done which would help the franchises to improve the service quality and design marketing plans.

7. Future Scope

Scalability needs to be achieved in the production environment and the plans for its implementation is:

- For each city filter business by category Active Life/Fitness Instruction(Upto 200k business, 20M+ reviews).
- Create indexes on elasticsearch based on metric to be evaluated
- Display both macro and micro level metric on Kibana to derive insight.
- We have used only Unigram but the feature will be more robust and useful after we use bigram and trigrams