### Big Data Analysis of Local Business and Reviews

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#### **ABSTRACT**

In this paper, we have analyzed the local business data and reviews to get insights on the popularity of a business and factors responsible for it. We have also analyzed the sentiments of the customer reviews for better understanding of business popularity. The total size of the dataset is 180 MB and we have adopted cloud computing service like Big Data Hadoop using HiveQL and Pig for query. Our preferred choice of framework for this project was Big Data Hadoop primarily because it is an open source software and its scalability and flexibility best suited our requirements.

This project has helped us in understanding various aspects of the Local Businesses; factors driving their popularity, customer review patterns and regions that favor certain businesses the most, are a few of the many aspects to name. Analyzing the customer sentiments based on their reviews has helped us in realizing the importance of customer satisfaction, also it is possible to derive action plans to improve business performance to keep up with the competition.

### **CCS Concepts**

### **Keywords**

Local Business; Data Analysis; Hadoop; Big Data; Reviews; Visualization; IBM Bluemix; Azure; Hive.

#### 1. INTRODUCTION

Yelp and Google Local engages a large percentage of customers for reviews about local business. It can be great source of information for business owners to improve their services and users to choose best services available. More than 3 Exabyte of data is created every day on the internet, making it a challenge to store this huge amount of data and retire useful information from it.

Our objective is to provide insightful analytics on existing business registered on these platforms, help future business owners to make important decisions regarding services or business expansion and find business trends. We have analyzed a data set from Yelp and Google Local that spans a variety of businesses such as restaurants, shopping, nightlife, medical, education, entertainment, common services, etc. in various cities of the world, using Big Data, a non-expensive framework that can store a huge variety of data set and process it in parallel [1][2]. We have worked on Hadoop ecosystem to perform the Big data analytics and have built the project on IBMs Biginsight as well as Azure HDinsight.

Biginsight is an open source service provided by IBM Open Platform, and includes all the Hadoop ecosystem components integrated with Apache Spark for a good processing power. We also built the project on Microsoft's ready to roll-out service called "HDInsight" using Hortonworks Hadoop on the Azure platform. The Hive ODBC driver from Microsoft and Hortonworks provides more options to Excel and PowerPivot users to query data within Hadoop running the cloud computing service.

The technical specifications for the platforms used for the project are as follows:

Azure Spec	ifications	IBM Bluemix Specifications			
Worker Node	4	Management	vCPU	12	
Header Node	4	Nodes: 1	RAM	48 GB	
Number of Cores	24		vCPU	4	
RAM	14 GB	Data Nodes: 1	RAM	24 GB	
Disk Size	200 GB		Data Disk	1 TB SATA	
Operation System	LINUX	CPU Speed		2.4 GHz Intel Xeon ES-2673	

**Table 1: Technical Specifications** 

HiveQL and Pig are the querying tools built on top of Hadoop that are used to query data within HDFS. They inherit all of Hadoop's fault tolerance features and are scalable for Big Data. The Hive language resembles SQL, which makes it useful for creating reports by Data Analysts whereas Pig is a Procedural Data Flow language used for programming by researchers and programmers.

Visualizations for this project are generated in Tableau and Excel power view that create multi-faceted views of data and help communicate complex ideas simply.

### 2. RELATED WORK

Published work related to local business analysis is focused on specific categories of business in general. Some of them have performed a detail analysis of business of a location, city or country unlike this paper which studies countries like US and UK. For instance, Gwo-Hshiung Tzeng [3] concentrates on the criteria for a good restaurant location in Taipei. Whereas Tsung-Yu Chou [4] evaluates the importance of infrastructure cost and environmental factors responsible for setting up a hotel business. Predicting Usefulness of Yelp Reviews by Xinyue et al. uses MATLAB to perform language processing techniques.

Our work has a more holistic and complete approach to the data analysis taking in account every aspect of local business. The goal of this paper is to provide all the possible insights from the data set, like defining a method that analyzes the factors of success of all kinds of businesses, instead of focusing on a single category of business, finding business trends, customer response, popularity of every business etc. Furthermore, using Hadoop Big Data and Cloud Computing services to store and process massive business data set.

### 3. DATA SPECIFICATIONS

The Local Business dataset collected from Yelp and Google APIs dated between 2005 and 2016, is rich in information about the local businesses in various cities in 14 different states in U.S and 4 other cities that include: U.K: Edinburgh, Germany: Karlsruhe, Canada: Montreal and Waterloo. The yelp data set is of size 90MB in CSV file format with 334,335 rows and 108 columns. The Google Local reviews dataset is of size 85MB in JASON file format with 117,486 and 10 columns. To be able to use this data in Hadoop, the JSON file was converted to a CSV file using SerDe (Serializer/Deserializer) in Hive. Alternatively, JSON to CSV file format conversion can also be done in PIG using the built-in function JSONLoader().

The business data contained a lot of junk data which made it difficult to import into Hive tables for analysis. To clean the data, we removed the duplicate rows, eliminated complete NULL valued columns. We also had to format the date column to yyyymm-dd time columns to timestamps.

### 4. ANALYSIS OF DATASET

The clean data was imported into HDFS, exploratory data analysis was performed on the dataset using LOAD and DUMP in Pig to gather some initial facts about our data. For example, the local business data included 2,903,217 entries having attributes such as business id, address, name, longitude, latitude, etc. The reviews data included 1,174,850 hive entries having attributes that included business id, user id, review id, stars, review date, and the user comments. Separate tables were created for business and reviews in HDFS using Hive QL. Reviews table creation is as follows:

"CREATE EXTERNAL TABLE IF NOT EXISTS reviews(funny int, useful int, cool int, userid string, reviewid string, stars int, reviewday date, type string, businessid string, comment string)ROW FORMAT DELIMITED FIELDS TERMINATED BY ',' STORED AS TEXTFILE LOCATION '/user/rsingh26/LocalBussinessReviews/' TBLPROPERTIES ('skip.header.line.count'='1');"

A master table was created by joining the business and reviews table on common business id in each table. There are 452 categories of business, we have grouped them broadly into 6 categories listed below. For each category, separate tables were created to further analyze the business grouped in them: - Education, Entertainment, Food, Medical, Services and Shopping.

#### 4.1 INFORMATION EXTRACTION

The data has been analyzed in the form of answers to the following questions related to business data.

4.1.1 What is the review count for each category of business to understand which business category gets most number of reviews by the customers?

The Hive QL to get the sum of the count for each category is as follows and its visualization is in Figure 1.

select sum(review\_count) from Education; select sum(review\_count) from Entertainment; select sum(review\_count) from Food; select sum(review\_count) from Medical; select sum(review\_count) from Services; select sum(review count) from Shopping;

#### **Review Count based on Category**

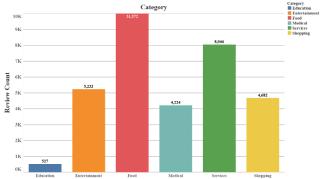


Figure 1. Review count category-wise

**Inference**: Users tend to review the food businesses more compared to the other categories of business like education, medical, shopping and services. Although categories like Education and Medical are equally vital categories as Food for day to day life. This insight can be used to drive medical and educational organizations to encourage customers to rate the services in-turn having an impact on business and service quality.

### 4.1.2 Which sub categories are most popular in different states in US?

The business data was collected from 14 states in US as shown in Figure 2. and Figure 3. It has been visualized using Excel powerview.

State	Category			
Alabama	Food			
Arizona	Home Services			
Arkansas	Home Services			
California	Entertainment			
Illinois	Food			
Minnesota	Local Services			
Nevada	Food			
New Mexico	Steak Houses			
North Carolina	Active Life			
Oregon	Hotels and Travel			
Pennsylvania	Food			
South Carolina	Food			
Texas	Banks and Credit Union			
Wisconsin	Food			

Figure 2. Business Categories and States

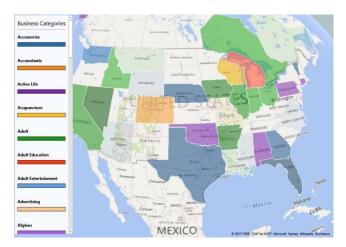


Figure 3. Sub categories grouped by state

### 4.1.3 Which is the most popular city for each business category?

The Tree map representation in Tableau is used to visualize the business categories grouped by city. The size of the rectangle represents the review count for a city and every business category is represented with a different color, refer Figure 4.

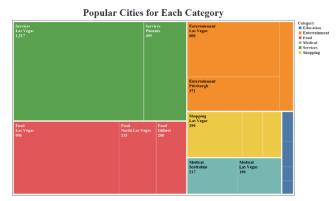


Figure 4. Business categories grouped by city

**Inference**: In every category of business maximum number of review count comes from Las Vegas. The possible reason for this could be the large number of tourists that visit this city every year, use the services and leave feedbacks. As another possibility, the business in this city encourage customers to review their business to improve the customer satisfaction and increase popularity.

## 4.1.4 What is the count of reviews for the subcategories of Shopping?

Bubble chart in Tableau is used to visualize the shopping subcategories grouped by city in Figure 5. The distinct color of the bubble represents different sub-category and size of the bubble represents the review count. Bubbles with city names are the most prominent sub category of Shopping in the respective city and the number below the city name is the review count.



Figure 5. Review count for the sub categories of Shopping

### 4.1.5 What is the sentiment analysis the customer reviews for all the Local Businesses in a city?

Customer's sentiment is analyzed by the sum of the positive, negative and neutrals words, used in the review, in accordance to the English dictionary. Cumulative sentiment is calculated for every review and summed all the business in a city and represented in pie chart for each city. The size of the pie graph is the number of people who voted the review useful.

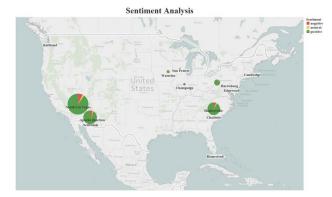


Figure 6. Sentiment Analysis based on customer reviews

**Inference:** More than 60% people in a city write positive reviews for the business and services they use. Las Vegas has the highest number of useful review votes.

# 4.1.6 What is the percentage of positive, neutral and negative reviews by the customers for the sub category of Local Businesses?

Sentiment analysis for the sub categories of Services is done in Figure 7. Similar analysis can be done for Education, Entertainment, Food, Medical and Shopping as well.

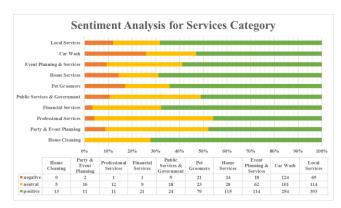


Figure 7. Percentage of positive, neutral and negative customer reviews for the Services category

**Inference**: Services businesses mostly received positive reviews from the customers and the Home cleaning services did not receive any negative feedback.

### 4.1.7 What is the maximum number of business reviewed by individual users?

The following Hive QL query was used to find the maximum number of business reviewed by a single reviewer and Figure. 8 visualizes it in Excel charts.

select userid, count(businessid) from review\_analysis group by userid limit 10;

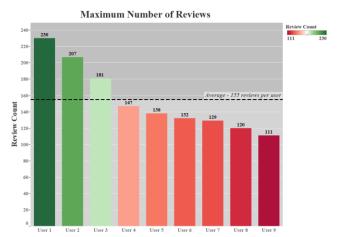


Figure 8. Maximum count of reviews made by individual users

**Inference**: One of the user has written reviews for almost 250 different business over a span of 10 years. Further text analysis of the reviews could help us in understanding the authenticity of these reviews.

### 4.1.8 What are the factors responsible for the popularity of the Local Businesses?

The top and bottom rated Food business are listed below in Table. 1 and Table.2 respectively. Various attributes like happy hour, reservation, parking, vegan etc. were compared and analyzed. It led to the conclusion that the attributes listed in Table. 3 are hugely responsible for the good reviews and ratings of a food business. Similar analysis can be done with other categories of business as well.

Top Rated Food Businesses					
City	Name	Reviews	Stars		
Las Vegas	Art of Flavours	359	5		
Las Vegas	Poke Express	315	5		
Las Vegas	Brew Tea Bar	306	5		
Gilbert	Frost Gelato	260	5		
Las Vegas	Dutch Bros. Coffee	241	5		
Phoenix	Handcrafted American	232	5		
Las Vegas	Karaoke Bar	192	5		
Montreal	Kem CoBa	156	5		
Mesa	Gelato Dolce Vita	149	5		
Las Vegas	Tast Crepes	148	5		

Table. 2 Top rated food businesses

Bottom Rated Food Businesses					
City	Name	Reviews Star			
Las Vegas	KFC	19	1		
Las Vegas	MCDonald's	17	1		
Charlotte	Pizza Hut	15	1		
Chandler	Dairy Queen	14	1		
Maricopa	Dairy Queen	14	1		
Scottsdale	Food Truck Festival	12	1		
Pittsburgh	Pizza Hut	11	1		
Carnegie	Walmart	10	1		
Queens	Burger King	9	1		
Surprise	Church's Chicken	9	1		

Table. 3 Bottom rated food businesses

Reser	vation	Ambi	ence	Wheelchair T		V WiFi		iFi	
Тор	Low	Тор	Low	Тор	Low	Тор	Low	Тор	Low
No	No	Yes	No	Yes	No	Yes	No	Free	No
Yes	No	No	Yes	Yes	No	No	No	Free	No
Yes	No	Yes	No	No	No	No	No	Free	No
No	No	No	No	Yes	No	No	No	No	No
Yes	No	Yes	No	No	No	Yes	Yes	Free	No
Yes	No	No	Yes	Yes	No	Yes	No	Free	No
Yes	Yes	Yes	Yes	Yes	No	Yes	No	Free	Free
Yes	No	Yes	No	Yes	Yes	Yes	Yes	No	Free
Yes	Yes	Yes	Yes	No	No	Yes	Yes	Free	No
Yes	No	Yes	No	Yes	No	No	No	Free	Free
80%	20%	70%	40%	70%	10%	60%	30%	80%	30%

Table. 4 Factors influencing the popularity of the food businesses

**Inference**: More than 70% of the top 10 food business have reservation, ambience, wheelchair facility, TV and free Wi-Fi whereas less than 30% of bottom 10 food business have these facilities.

#### 5. CONCLUSION

Using Hadoop, Hive, Pig, and Tableau enhanced the exploratory possibilities and analytical capability, to store and process Big Data in parallel computing. The findings from the research can be concluded as: food is the most popular category of Local Business based on the review count. Las Vegas is the most popular city for local business in every category of business. Reservation, ambience, Wi-Fi are the main factors responsible for the popularity of food businesses. More than 60% of people in a city write positive reviews for local business. And on record the maximum number of business reviewed by a single reviewer is 250.

### 6. ACKNOWLEDGMENTS

We are grateful to Mahsa Tayer Farahani and Hemamalini Madhanguru for their help and support in this project.

### 7. REFERENCES

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