Exploratory Data Analysis and Data Mining on Yelp Restaurant Review Using Ada Boosting and MLP Techniques

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Abstract:- Exploratory data analysis (EDA), which provides both descriptive and inferential analysis, plays a crucial role in comprehending the significance of the data's hidden information. The text corpus's subjects are identified using the data mining method. The datasets from Yelp, which contain information about businesses. users, ratings, and signups, have been analyzed in this study. In addition to timing of check-ins at company sites, our study also looks at firm performance, regional distribution, reviewer ratings, and other factors. We discovered that Yelp check-ins, tips, and elite users had all declined over time. Additionally, our analysis showed that Canadians have more reliable star ratings and sentiment ratings than Americans. To improve on this effort, we suggest a new project that comprises gathering a dataset, cleaning the data by removing null values, applying a machine learning algorithm with Ada Boosting, and forecasting the accuracy score with MLP. The proposed technique for EDA and data mining on Yelp restaurant reviews has various potential flaws. Because the information was selected depending on the needs of the research, it may not be representative of all restaurants on Yelp. This might lead to skewed findings. Pre-processing processes such as data cleaning and sampling may remove vital information or inject noise into the dataset. The model's performance and generalizability may not be adequately assessed using hold-out and cross-validation procedures.

Keywords:- Exploratory Data Analysis (EDA), Descriptive Analysis, Inferential Analysis, Data Mining, Yelp, Datasets, User Information, Ratings, Performance, Regional Distribution, Star Ratings, Sentiment Ratings, Machine Learning Algorithm, Ada Boosting, MLP, Accuracy Score, Data Cleaning.

I. INTRODUCTION

The Internet is a large and incredibly astounding reservoir of information, there is no question about it. Due to the growth of websites, the expansion of electronic commerce (e-commerce), and the fact that many companies allowed customers to rate their items, the Internet has developed into a valuable resource for consumer reviews of a variety of goods and services.

Reviews are statements made by customers about products, services, brands, or enterprises on social networks, instant messaging, blogs, microblogs, websites, or other online communities. "Peer-shared product reviews on companies' or third parties' websites" are the terms used to describe reviews. E-commerce websites, such as Amazon, and ranking product websites, such as Yelp, provide customers with a 5-point scale on which they can rate products or the quality of services. where 5 is the highest possible score and I is the lowest. Customers can rate products or the quality of services using a 5-point scale supplied by ranking products on websites like Yelp and e-Commerce websites like Amazon, where 5 is the highest possible score and I is the lowest.

Reviews and rating systems have developed into a significant resource that prospective or new consumers rely on and use to inform significant decisions in a variety of areas of their lives, from what they invest in to what they eat to where they get treatment. The fact that business owners rely on customer reviews as a source of information for making decisions about their operations highlights the requirement for more examination in the space of electronic surveys. Tracking reviews online assists service businesses in improving their goods and services by recognizing client needs and highlighting areas of dissatisfaction. Restaurant operators can better determine what customers want by studying feedback that has been shared on electronic platforms.

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A 1-star improvement in Cry's positioning likewise makes a 5 9% lift in eatery space deals. Therefore, it is important for those working in the restaurant industry to understand what works for their customers.

It takes a lot of time and effort to use and comprehend the massive amount of evaluations. To simplify, summarize, and comprehend data, however, exploratory analysis and data mining approaches are vital. The information is delivered on time with the least amount of work and the maximum profit. The main goal of this study is to shed light on how to make the most of consumer information and experiences shared about restaurants through internet review sites. The purpose of the project is to create a new dataset with relevant qualities using exploratory information examination and information mining methods on a Howl eatery survey dataset, preprocess the data, extract features, and test the models using Ada Boosting and MLP approaches. Data gathering, preprocessing, feature extraction, and model evaluation utilising hold-out and cross-validation approaches are the specific procedures involved. The end goal is to create a classification model that is highly accurate and predictively relevant.

II. RELATED WORK

Roger D. Peng's exploratory data analysis This book provides a thorough analysis of EDA as of 2012[1]. Written in Python, Utilizing Textual Analysis: Enabling Language-Aware Data Products by Benjamin Bengfort 2017[2]: This study utilizes Chapters 3 and 6 on text clustering and preprocessing, respectively. Think Stats by Allen B. Downey: Exploratory Data Analysis This book, published in 2014 [3], covers the complete data analytics process, including data collection and statistical result generation. Good, I. J. Exploratory Data Analysis: A Philosophical Approach, 1983[4]:-paper makes an attempt to understand ED philosophically. Modeling a topic: - Topic modeling offers a method for studying unlabeled text, The authors of a 2015 [5] paper titled "A Survey of Topic Modeling in Text Mining" describe the various topic modeling methodologies and how they are applied.

Text Similarity Computing Based on Word Co-occurrence and the LDA Topic Model Minglai Shao and Liangxi Qin (2014) [6] developed a text similarity computation method based on word occurrences and hidden themes models in this study. Idle Dirichlet Allotment and the Regular Number of Subjects: A few Perceptions 201081papershowshttps://www.yelp.com/dataset/challenge[7].Bar plots are a significant point in the Four Examinations on the Impression of Bar Graphs - Scene Exploration 2014 [8] article. Data analytics experience in EDA and testing: concepts, expectations, and difficulties 2016 [9] Review of machine learning and data mining techniques for electrical design automation.

Data mining techniques and machine learning in electrical design automation and test are reviewed in the [10] study. The 2014 [11] article on the use of exploratory information examination in evaluating exhibits how EDA is utilized in reviewing. Using word clouds as a basis for text

analytics, Word Cloud Explorer was created in 2014. [12] discusses word clouds and their word cloud explorer tool for text visualization. Visualizing words in clouds across several text documents 2015. This study [13] discusses word cloud analysis of numerous texts. David M.P. Ennock, Steve Lawrence, and Kushal Dave, 2003 [14] has created a methodology for automatically differentiating between good and bad reviews, using SVM with -grammes and metrics (precision and recall) to gauge performance. (Lee, Srivakumar, and Bo Pang 2002) [15], categorized by general emotion rather than by topic.

III. DATA COLLECTION AND DESCRIPTION

A. Data Background

Yelp.com is regarded as a comprehensive review site. A multinational company, Yelp is headquartered in San Francisco, California. The firm operates the Yelp smart phone app and website, which collects reviews of nearby businesses from the general public. Howl was established in 2004 and extended all through Europe and Asia somewhere in the range of 2009 and 2012. In 2019, Yelp saw a monthly average of 61.8 million unique desktop visits and 76.7 million unique website users [7].Yelp stated that it has 192 million reviews as of June 30, 2019 [8]. The website has sections for particular types of companies, including cafes, hospitals, hotels, spas, and schools. It uses a one to five star rating system to allow users to publish text reviews and submit reviews on products or services from companies.

B. Data Collection

The dataset may be accessible through the Howl Dataset Challenge, which is accessible on the Cry site, as well as on the Kaggle website. Only two of the five CS files from the Yelp dataset—yelp_business and yelp review—have been used because they are appropriate for this study.

C. Data Description

The business dataset contains 174,567 entries over 13 descriptive variables and several company types. The review dataset contains information about users' commercial experiences. There are 5,261,668 documents in the review dataset, along with nine descriptive characteristics.

IV. IMPLEMENTATION

A. Collection of Data

Obtain the Restaurant review dataset for Canada and the US from a reliable source. Filter the dataset to ensure that the data only includes reviews with star ratings and sentiment ratings. Create a new dataset with attributes relevant to the analysis, such as the restaurant name, location, star rating, sentiment rating, and country.

B. Pre-Processing the Data

Clean the data by removing any duplicates, missing values, or irrelevant data. Convert the text data to numerical data by using techniques such as bag-of-words or word embeddings. Part the dataset into preparing and testing sets, with a proportion of 80:20.

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C. Extraction of Features

Extract relevant features from the pre-processed data using techniques such as Principal Component Analysis (PCA) or t-Distributed Stochastic Neighbor Embedding (t-SNE). Train the Ada Boosting and MLP classifiers on the training dataset. Use the trained classifiers to predict the star rating and sentiment rating of the testing dataset. Generate comparison charts to visualize the performance of the classifiers.

D. Evaluating the Model

Assess the presentation of the classifiers utilizing measurements like exactness, accuracy, review, and F1 score. Use wait or cross-approval procedures to guarantee that the classifiers are not overfitted. Compare the performance of the Ada Boosting and MLP classifiers to determine which one provides more accurate forecasts. Generate a graph representation of the categorized data to visualize the consistency of star ratings and sentiment ratings between Canada and the US.

Overall, this process involves collecting, implementation pre-processing, feature extraction, and model evaluation steps to predict the consistency of star ratings and sentiment ratings between Canada and the US. The Ada Boosting and MLP classifiers are used to achieve more accurate forecasts, and comparison charts are generated to visualize the results.

> Correlation Matrix:

The correlation matrix of the mean values of the 'cool', 'useful', and 'funny' columns in a pandas Data Frame named 'df' grouped by the 'stars' column. The resulting correlation matrix shows the correlation coefficients between the 'cool', 'useful', and 'funny' columns. The connection coefficient is a worth between - 1 and 1 that actions the direct connection between two factors.

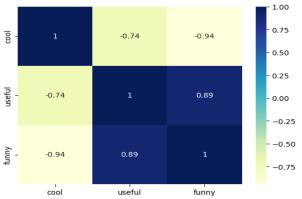


Fig 1:CORRELATION MATRIX.

➤ Viualization of the Reviews:

The number of reviews for each star rating (1-5) in a bar chart. It uses the seaborn library to set the color palette for the bars and matplotlib to plot the bar chart. The pd.Series() method creates a pandas series object from the "stars" column of the dataframe and then the value_counts() method counts the number of occurrences of each star rating. The resulting frequency counts are plotted as a bar chart using the

plot() method. The figsize parameter is used to set the size of the figure, and the rot parameter sets the rotation angle of the x-axis labels. Finally, the xlabel() and ylabel() methods set the x and y-axis labels, respectively. Where x-axis tells about Stars and y-axis tells about Frequency

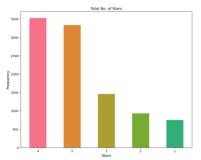


Fig 2.1:THE NUMBER OF REVIEWS OF EACH STAR 1-

We will create a new column that combines the stars 1-3 as negative and 4-5 as positive, 0 if the star rating is 3 or less, and 1 if the star rating is more than 3.

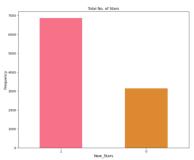


Fig 2.2:BAR GRAPH WITH POSITIVE NEGITVE COLUMNS

Word Cloud Generartion For Positive And Negitive Reviews:

A "word cloud" depicts the frequency of words visually. "Depending on how frequently it appears in the text being analyzed, the phrase appears larger in the picture created. Word clouds are becoming more popular as a quick method for determining written content's main idea. They have been utilized, for example, to picture the substance of political discourses in governmental issues, business, and training. Word clouds were used to look at the content of Board committee papers in the Health Board that I support to see if the organization's most important operations get enough attention..



Fig 3.1: Word Cloud for negative and neutral reviews (stars 1-3)



Fig 3.1: Word Cloud for positive reviews (stars 4-5)

There after, by utilizing a logistic regression model to classify text. The text is first cleaned up by getting rid of non-alphabetic letters, changing it to lowercase, breaking it up into words, and getting rid of stop words. The cleaned text is then divided into training and test sets and converted into a numerical vector representation using the TF-IDF Vectorizer class. Using a variety of assessment measures, the logistic regression model is tested on the test set after being trained on the training set. Using the pickle library, the trained model is saved to disc and then loaded once more to verify that it functions as intended.

	precision	recall	f1-score	support
9	0.81	0.58	0.68	1028
1	0.83	0.94	0.88	2272
accuracy			0.83	3300
macro avg	0.82	0.76	0.78	3300
weighted avg	0.82	0.83	0.82	3308

Fig 4:Test Accuracy Score

V. CONCLUSION

This study applied EDA and data mining techniques to Yelp restaurant review data to analyze company performance, geographic distribution, reviewer ratings, and timing of check-ins. The proposed approach involved data collection, pre-processing, feature extraction, and model evaluation using machine learning algorithms. The results showed a decrease in Yelp reviews, tips, elite users, and check-ins over time, as well as differences in star and sentiment ratings between Canadians and Americans. The accuracy of the model was predicted using Ada Boosting and MLP algorithms. Overall, this study provides valuable insights into the trends and patterns of Yelp restaurant review data and demonstrates the potential of data mining techniques in analyzing large datasets.

FUTURE WORK

Future work for EDA and data mining on Yelp restaurant reviews dataset can be achieved through the following steps:

 Sentiment Analysis: Implement sentiment analysis to determine the overall positive or negative tone of the reviews. Machine learning algorithms such as logistic regression or Naive Bayes can be used. The results can be used to identify the strengths and weaknesses of the restaurants.

- Topic Modeling: Use topic modeling to identify the key topics that people discuss in their reviews. Techniques such as Latent Dirichlet Allocation (LDA) or Nonnegative Matrix Factorization (NMF) can be used. This can help restaurant owners to understand what customers are saying about their food, service, and other aspects of their business.
- User Profiling: Profile users who leave positive or negative reviews to identify their characteristics. AI calculations, for example, choice trees or arbitrary timberlands can be utilized. The results can be used to tailor the marketing strategies of the restaurant to different user groups.
- Predictive Modeling: Utilize prescient demonstrating to foresee the rating of an eatery in light of different factors like area, food, and cost range. AI calculations, for example, choice trees, irregular backwoods, or brain organizations can be utilized. The outcomes can be utilized to distinguish the variables that are most significant in deciding the rating of a café.
- Time-Series Analysis: Conduct time-series analysis to identify the trends in the reviews over time. Techniques such as moving averages or exponential smoothing can be used. The results can be used to identify the changes in the preferences of customers over time.
- Text Summarization: Summarize the reviews into short paragraphs that capture the key points. Techniques such as text clustering or text summarization algorithms can be used.
- Interactive Visualization: Use interactive visualization to create dashboards that allow restaurant owners to explore the data in an interactive way. Tools such as Tableau or PowerBI can be used

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