Swiggy Data Analysis

Low Level Design (LLD)

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Abstract

The online food ordering market includes foods prepared by restaurants, prepared by independent people, and groceries being ordered online and then picked up or delivered. The first online food ordering service, World Wide Waiter (now known as Waiter.com), was founded in 1995. Online food ordering is the process of ordering food from a website or other application. The product can be either ready-to-eat food or food that has not been specially prepared for direction consumption.

In the world of rising new technology and innovation, Food industry is advancing with the role of Data Science and Analytics. Data analysis can help them to understand their business in a quiet different manner and helps to improve the quality of the service by identifying the weak areas of the business. This study demonstrates the how different analysis help to make better business decisions and help analyze customer trends and satisfaction, which can lead to new and better products and services. Different analysis performed such as Extract, Transformed Load(ETL) Analysis and Descriptive Analysis on variety of use cases to get the key insights from this data based on which business decisions will be taken.

Introduction

1.1 What is Low-Level design document?

The goal of the LDD or Low-level design document (LLD) is to give the internal logic design of the actual program code for the House Price Prediction dashboard. LDD describes the class diagrams with the methods and relations between classes and programs specs. It describes the modules so that the programmer can directly code the program from the document.

1.2 Scope

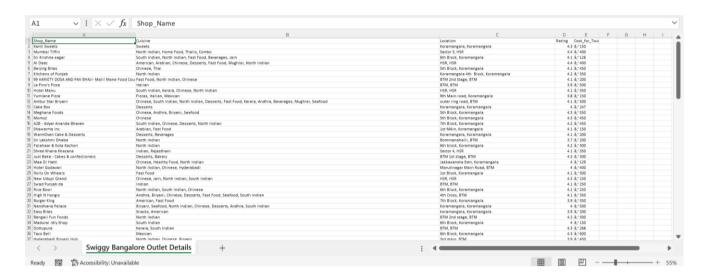
Low-level design (LLD) is a component-level design process that follows a step-by-step refinement process. The process can be used for designing data structures, required software architecture, source code and ultimately, performance algorithms. Overall, the data organization may be defined during requirement analysis and then refined during data design work.

1.3 Constraints

The analysis must be user friendly, code must be neat & clean, ETL must be automated as much as possible because it will save huge amount of time. Moreover, users should not be required to have any of the coding knowledge as the insights they are looking for are mentioned in-detail with respective visuals.

2 Technical Specifications

2.1 Swiggy Bangalore Outlet Dataset



2.1.1 Swiggy Bangalore Outlet Dataset Overview –

The Listings dataset consists of a table with 118 records and 5 features. Features are distributed as 2 Continuous features and 3 Categorical features. There are a total 0% of records having Missing Values. In short, there are no Missing Values present in the dataset.

```
Basic Information:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 118 entries, 0 to 117
Data columns (total 5 columns):
   Column
                  Non-Null Count Dtype
0 Shop_Name 118 non-null object
1 Cuisine 118 non-null object
 1 Cuisine
     Location 118 non-null object Rating 118 non-null object
   Location
 2
    Cost_for_Two 118 non-null
                                     object
dtypes: object(5)
memory usage: 4.7+ KB
None
```

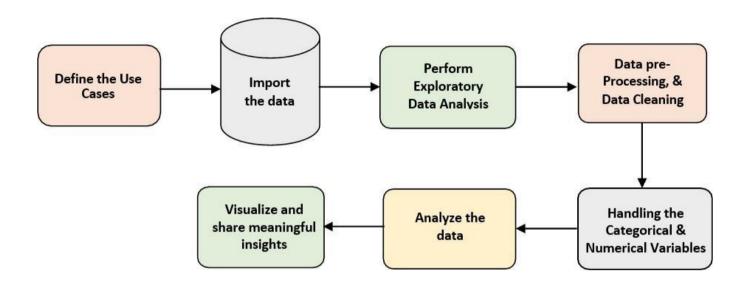
2.1.2 Overview of Variables:

Number of unique values for each attribute:

shop_name	114
cuisine	79
location	64
rating	12
cost_for_two(₹)	30
avg_cost_per_person	30
cost_category	2
area	4

dtype: int64

3 Architecture



3.1 Architecture Description -

3.1.1 Data Description -

As we have seen earlier, in our Swiggy dataset, we have around 118 records with 5 different features. Features are distributed as 2 Continuous features and 3 Categorical features. These datasets are given in the form of Comma Separated Value (.csv) format.

3.1.2 Define the Use Cases -

At this stage, based on the given dataset and business problems we have defined the several Use Cases to perform the analysis on and this will definitely help out get the key insights from this data based on which business decisions will be taken. Furthermore, It helps in not only understanding the meaningful relationships between attributes but it also allows us to do our own research and come-up with our findings.

3.1.3 Import the Dataset -

As we have received the dataset in the form of Comma Separated Value (.csv) format, therefore we can import the same using Pandas read_csv() function.

Reading Data In [2]: df = pd.read_csv("Swiggy Bangalore Outlet Details.csv") Location Rating Cost_for_Two Shop Name 0 Kanti Sweets Koramangala, Koramangala 4.3 ₹ 150 Mumbai Tiffin North Indian, Home Food, Thalis, Combo Sector 5, HSR ₹ 400 2 Sri Krishna saqar South Indian, North Indian, Fast Food, Beverag... 41 ₹ 126 6th Block, Koramangala HSR. HSR 4.4 Al Daaz American, Arabian, Chinese, Desserts, Fast Foo... ₹ 400 4.1 Beijing Bites Chinese Thai 5th Block, Koramangala ₹ 450 Pan-Asian, Chinese, Asian 113 Wok Paper Scissors JNC Road, Koramangala 3.9 ₹ 219 114 Savoury Restaurant Arabian, Middle Eastern, North Indian, Grill... Madiwala, BTM 4.1 ₹ 600 115 Royal Treat North Indian, Chinese, Seafood, Biryani 5th block Koramangala, Koramangala 4.2 ₹ 193 116 Thali 99 North Indian ₹ 200 Koramangala, Koramangala 117 Mani's Dum Biryani Andhra, Biryani 1st Block, Koramangala 4.2 ₹ 400

3.1.4 Exploratory Data Analysis (EDA) –

- "Exploratory Data Analysis" (EDA) is a "Data Exploration" step in the Data Analysis Process, where a number of techniques are used to better understand the dataset being used.
- Understanding the Dataset can refer to a number of things including but not limited to...
 - Extracting Important "Variables".
 - Identifying "Outliers", "Missing Values", or "Human Error".
 - Understanding the Relationships between variables.
 - Ultimately, maximizing our insights of a dataset and minimizing potential "Error" that may occur later in the process.

In other words, it will gives you a better Understanding of the "Variables" and the "Relationships" between them.

3.1.5 Data Pre-processing, Data Cleaning & Imputation (Handling the Categorical & Numerical Variables) –

Data pre-processing is a process of preparing the raw data and making it suitable for our analysis purpose, where we have to do lot of Data Cleaning, handle the missing values by using appropriate imputation techniques and based on that variable nature i.e. either of Categorical & Numerical variable. Here, in this project, we have done the substitution/imputation of missing values using either mean, median or mode according to the nature of those variables. Moreover, we also removed the columns which are does not participate in our analysis.

3.1.6 Analyse the Data –

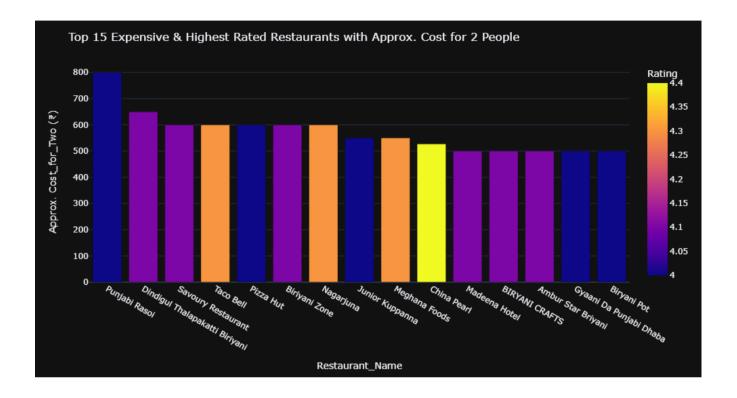
Once the pre-processing is done, we are good to go with our actual analysis where we write lines of codes and logics to prepare our data as per the defined use cases.

3.1.7 Visualize & Share Meaningful Insights –

Finally, it's time to turn our data into some sort of visual representation. In short, Data visualization is the process of translating large data sets and metrics into charts, graphs and other visuals such as Bar Plot, Pie Chart, Heat map, Box Plot, Scatter Plot, and many more. The resulting visual

representation of data makes it easier to identify and share insights about the information represented in the data.

Here is the beautiful glimpse of one of our visuals are -



All those different analysis help out to make better business decisions and help analyse customer trends and satisfaction, which can lead to new and better products and services.

4.Technology Stack

