



Data Collection and Preprocessing Phase

Date	15 July 2024
Team ID	740148
Project Title	Number Oracle:Big Mart Sales Predictive Analysis
Maximum Marks	6 Marks

Data Exploration and Preprocessing Template

Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

Section	Description
Data Overview	Basic statistics, dimensions, and structure of the data.
Univariate Analysis	Exploration of individual variables (mean, median, mode, etc.).
Bivariate Analysis	Relationships between two variables (correlation, scatter plots).
Multivariate Analysis	Patterns and relationships involving multiple variables.
Outliers and Anomalies	Identification and treatment of outliers.
Data Preprocessin g Code Screenshots	

```
# Fill NaN values
                    train df['Item Weight'].fillna(train df['Item Weight'].mean(), inplace=True)
                     test_df['Item_Weight'].fillna(test_df['Item_Weight'].mean(), inplace=True)
Handling
Missing Data
                    train df['Outlet Size'].fillna(train df['Outlet Size'].mode()[0], inplace=True)
                     test_df['Outlet_Size'].fillna(test_df['Outlet_Size'].mode()[0], inplace=True)
                     # Handle outliers using TOD
                     # Handle outliers using IQR
                     def handle_outliers(df, column):
                        Q1 = df[column].quantile(0.25)
                         Q3 = df[column].quantile(0.75)
                        IQR = Q3 - Q1
Data
                        lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR
Transformatio
                        df[column] = np.where(df[column] > upper_bound, upper_bound,
                                             np.where(df[column] < lower_bound, lower_bound, df[column]))</pre>
                     numerical_columns = ['Item_Weight', 'Item_Visibility', 'Item_MRP', 'Item_Outlet_sales']
                     for column in numerical columns:
                        handle_outliers(train_df, column)
if column != 'Item Outlet Sales':
                            handle_outliers(test_df, column)
                    # Encode categorical features
                    le = LabelEncoder()
                    cat col = ['Item Fat Content', 'Item Type', 'Outlet Identifier', 'Outlet Size', |Outlet Location Type', 'Outlet Type']
Feature
Engineering
                    for col in cat col:
                       train df[col] = le.fit transform(train df[col])
                       test df[col] = le.transform(test df[col])
                    # Train-test split
Save
                    X = train_df.drop(columns=['Item_Outlet_Sales'])
Processed
                    v = train df['Item Outlet Sales']
Data
                    X train, X val, y train, y val = train test split(X, y, test size=0.3, random state=42)
Loading Data
                       # Load datasets
                       train df = pd.read csv('/content/train.csv')
                       test df = pd.read csv('/content/test.csv')
                       sample submission df = pd.read csv('/content/sample submission.csv')
```