Lead Scoring Case Study

Summary Report

Steps of Analysis and Approach

Importing Libraries and Checking the Data Missing Value Treatment Outlier Treatment EDA

> Univariate Analysis Bivariate Analysis

Data Preparation

Train Test Split

Feature Scaling

Feature Selection Using RFE & Further by Manual Inspection

Finding Optimal Cutoff Point Using ROC

Metrics beyond accuracy & Plotting the ROC Curve

Precision and Recall

Making predictions on the test set

Importing Libraries and Checking the Data - This involves importing the necessary libraries and understanding the data, its dimensions, and the data types of the columns. Identifying the target variable and the variables out of which feature selection will be done at a later stage. Duplicates were also checked.

Missing Value Treatment - Missing value treatment was done in a number of ways. The missing values in the case of different columns were replaced by either mean or median or mode depending on the type of data in the column, i.e., numerical or categorical, and keeping in mind their business significance.

Outlier Treatment - After looking at the percentile values in the columns outliers were removed from the Total Visits and Page Views Per Visit columns. Boxplots were used to confirm the data after cleaning.

EDA

Univariate Analysis - Data were analyzed after grouping according to the target variable. The relation of some of the variables with the Converted data was seen using countplots and catplots.

Bivariate Analysis - A pairplot was generated to understand the correlation between the numerical variables. Columns with high correlation can be dropped as they will not influence the target variable independently.

Data Preparation - Based on the EDA analysis it is seen that many columns are not adding any information to the model, hence we drop them before further analysis. After dropping the unnecessary columns for categorical variables with multiple levels, dummy features using one- hot encoding was done.

Train Test Split and Feature Scaling -

Data is divided into Train and Test. Features that require scaling are done so using the Standard Scaler.

Feature Selection Using RFE & Further by Manual Inspection -

Feature Selection is done using the RFE method and the number of variables is reduced to 18. Then manually the model summary is analyzed and variables having a p-value of more than 0.05, are dropped one by one. After dropping each variable the model summary is again analyzed.

Finding Optimal Cutoff Point Using ROC - The confusion matrix is created and accuracy, sensitivity, and specificity for various probability cutoffs are calculated. The value of 0.4 is selected as a cutoff.

Metrics beyond accuracy & Plotting the ROC Curve - The fpr, tpr, and threshold values are calculated. Area under the ROC curve is calculated

Precision and Recall - Using sklearn utilities the precision and recall score were calculated. A precision and recall tradeoff is done by plotting both of them and finding out where they intersect.

Making predictions on the test set - Using the cut-off of 0.4 the predictions are made. Accuracy, Sensitivity and Specificity values of test set are matched with trained set. Overall performance is evaluated