

Lead Scoring Case Study Summary

X Education gets a lot of leads, its lead conversion rate is very poor at around 30%. The company requires us to build a model wherein we need to assign a lead score to each of the leads such that the customers with a higher lead score have higher conversion chance. CEO's target for lead conversion rate is around 80%.

1. Reading & Understanding the Data

First, we read the Dataset and Inspect the Data.

2. Data Cleaning

- a) First step to clean the dataset we dropped the columns having NULL values greater than 35%.
- b) Next, we removed the imbalanced and redundant variables. This step also included imputing the missing values as and where required. The outliers were identified and removed.
- c) Then we chose to drop the variables having unique values.

3. Data Preparation

- a) Changed the binary variables into '0' and '1'.
- b) We created dummy variables for the categorical variables.
- c) Then, Removed all the repeated and redundant variables.

4. Train Test Split

The next step was to divide the data set into train and test sections with a proportion of 70:30.

5. Feature Scaling

- a) We used the Min Max Scaling to scale the original numerical variables.
- b) The, we plot the a heatmap to check the correlations among the variables.
- c) Dropped the highly correlated dummy variables.

6. Model Building

- a) Using the Recursive Feature Elimination, we went ahead and selected the 15 top important features.
- b) Then, we recursively tried looking at the P-values to select the most significant values that should be present and dropped the insignificant values.
- c) Then, we arrived at the 11 most significant variables.
- d) Then we tried looking at the VIF values to select the most significant values that should be present and dropped the insignificant values.
- e) Finally, we arrived at the 10 most significant variables.
- f) For our final model we checked the optimal probability cut off by finding points and checking the accuracy, sensitivity, and specificity.
- g) We then plot the ROC curve for the features and the curve came out be pretty decent with an area coverage of 84% which further solidified the of the model.
- h) Then we implemented the learnings to the test model and calculated the conversion probability based on the Sensitivity and Specificity metrics and found out the accuracy value to be 77.36%; Sensitivity= 76.97%; Specificity= 77.73%.

7. Model Evaluation

Evaluation involved creating a confusion matrix and selecting a cutoff point of 0.39 based on accuracy, sensitivity, and specificity plots, yielding metrics around 80%. Sensitivity-specificity view was chosen over precision-recall for the final predictions to align with the CEO's goal.

8. Conclusion

- a) While we have checked both Sensitivity-Specificity, we have considered the optimal cut off based on Sensitivity and Specificity for calculating the final prediction.
- b) Accuracy, Sensitivity and Specificity values of test set are around 77.36%, 76.97% and 77.73% which are approximately closer to the respective values calculated using trained set.
- c) Hence overall this model seems to be good.
- d) Features which contribute more towards the probability of a lead getting converted are:
 - i. Lead Origin_Lead Add Form
 - ii. What is your current occupation_Working Professional
 - iii. Total Time Spent on Website