## Lead Scoring Logistic Regression

Team:-

Elumalai C Hariharan J

#### **Problem Statement**

- An education company named X Education sells online courses to industry professionals. On any given day, many professionals who are interested in the courses land on their website and browse for courses.
- X Education wants to find out the leads that are most likely to convert into paying customers. The company requires to build a model and to assign a lead score to each of the leads such that the customers with higher lead score have a higher conversion chance and the customers with lower lead score have a lower conversion chance. The CEO, in particular, has given a ballpark of the target lead conversion rate to be around 80%.

#### Steps involved in the Analysis

- 1) Reading and Understanding the dataset.
- 2) EDA Building

Removing Missing/Null values

**Outlier Treatment** 

Rescaling

Converting some binary variables

3) Data Preparation

create dummy features

Perform train-test split

Perform Scaling

- 4) Model Building
- 5) Model Evaluation

**ROC Curve** 

Optimal Cut off Point

Predictions on the test set

6)Conclusion

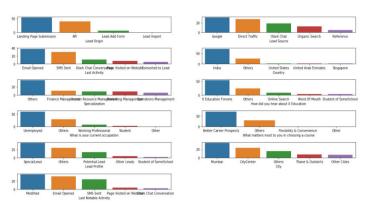
## Reading and Understanding the dataset

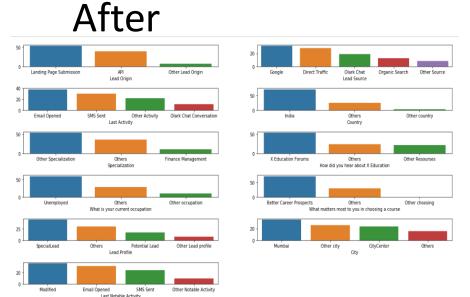
- Importing the data into pandas data frame.
- Inspecting the data frame using inspections methods like Info, shape, describe etc.
- Handling Select Level in Categorical Columns

#### **EDA Building**

- There was no duplicate records in the given data set.
- Unique Categories in the Categorical Columns

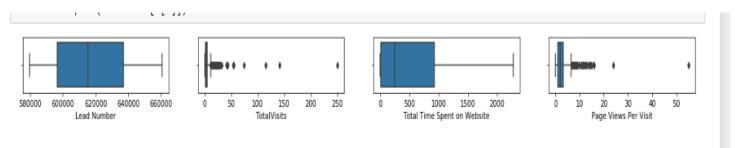
#### Before



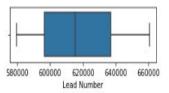


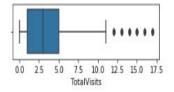
### **EDA Building**

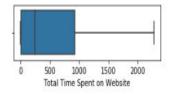
- Outliers
  - Box plot is used to find out the outliers of the variables

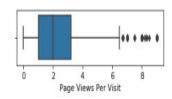


Capping method is used to fix the outliers.







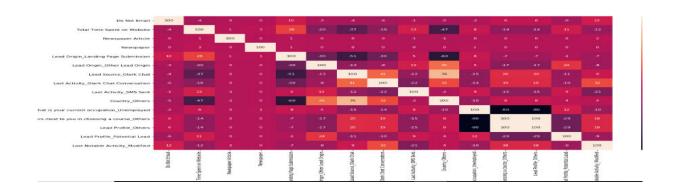


#### **Data Preparation**

- For all categorical variables with multiple levels, created the dummy features (one-hot encoded)using the pandas get dummies functions.
- Data is split into train and test set where 70 percentage of data in train set and 30 percentage in test data
- Using Standard Scalar the variables are scaled.

#### **Model Building**

- Using Stats model build the first model and analyzed the coefficient of each variable.
- Logistic Regression model is used to find out the top 15 variables to build model.
- Heat Map is build to show the correlation of the variables.



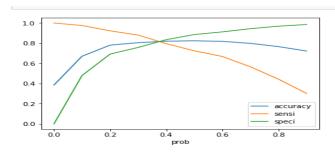
#### **Model Building**

- Using stat model the model is build for the top 15 variables.
- Confusion matrix is build for the model and analyzed the Accuracy, Sensitivity, Precision, Recall.
- VIF score is calculated for the variables and eliminated the variables whose value is more than 5.

- ROC Curve.
  - It shows the trade off between sensitivity and specificity (any increase in sensitivity will be accompanied by a decrease in specificity).
  - The closer the curve follows the left-hand border and then the top border of the ROC space, the more accurate the test.
  - The closer the curve comes to the 45-degree diagonal of the ROC space, the less accurate the test.

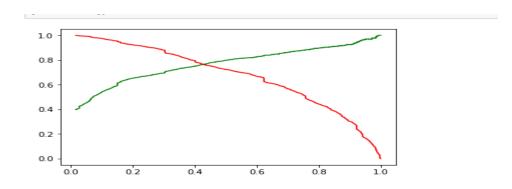
True Positive Rate

- Optimal Cut off Point:-
  - For the different probability from 0.0 to 0.9 calculated the Accuracy, Sensitivity, Precision, Recall.
  - Line chart is plotted for Accuracy, Sensitivity,
     Precision, Recall and found the actual cut of point.



From the curve above, 0.4 is the optimum point to take it as a cut off probability.

- Precision and Recall curve
  - Precision and Recall curve is calculated for thresholds using the precision\_recall\_curve() function that takes the true output values and the probabilities for the positive class as output and returns the precision, recall and threshold values



- Predictions on the test set
  - Using the train model the values are predicated for the test data set.
  - Accuracy, Sensitivity, Precision, Recall is calculated for the test modal.
  - Summary:-

# Summary of the train modal Evaluation Result Accuracy :-82 % Sensitivity :- 79% Specificity :-83 % Precision :-75 % Recall :-79 %

#### Summary of the Test modal

#### **Evaluation Result**

Accuracy :-81 %

Sensitivity :- 78 %

Specificity :-83 %

Precision :-72 %

Recall :-78 %

#### Conclusion

- From the train and test model evaluation we can infer that all Accuracy, Sensitivity & specificity are all close to each other-which determines the ROC cut off 0.4 is good.
- And also we can see all the variables P values ate <0.5 which are more significant.</li>
- And also VIF values of all the variables are < 5, which is also more significant.
- \_\_ve coefficient values variables has less percentage of chance of probability of lead getting converted.
- +ve coefficient values variables has high percentage of chance of probability of lead getting converted.
- High Chance of getting lead to converted variables are listed below with their co-efficient value.
- Lead Origin Other Lead Origin: Coefficient values = 3.1409 times of Lead origin probability of lead getting converted.
- Newspaper Article: Coefficient values = 1.7885 times of Newspaper Article probability of lead getting converted.
- Lead Profile\_Potential Lead: Coefficient values = 1.6258 times of Lead origin probability of lead getting converted.
- Overall, the Company should concentrate on the above variables to increase the number of leads percentage.

**End of the Analysis-Thanks.**