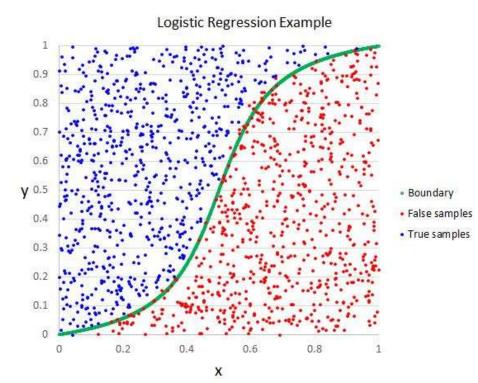
Logistic Regression – Lead Scoring Assignment

SNEHAL VIRWADEKAR (DSC 43)

DEEPTHY T. BABU

Assignment- Technical &

Business Analysis



I. BUSINESS PERSPECTIVE

Problem Statement

- ▶ X Education gets leads from various sources like Referrals , interested professional who visited the website etc.
- ▶ Although X Education gets a lot of leads, its lead conversion rate is very poor.
- To make this process more efficient, the company wishes to identify the most potential leads, also known as 'Hot Leads'.
- ▶ We need to build a model wherein we need 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.

Data

► Leads.csv

- Features 37
- Data Points: 9240
- Target Column: 'Converted'
 - 5679 Label 0
 - 3561 Label 1
- 30 Features Categorical
- 7 Features Numerical

Strategy

- Missing Data Handling
 - Dropping Features with >45% Missing Data
 - GridSearchcv to find best Imputation strategy
- Outliers Trimming using IQR: Total Visits, Pageviews Per Visits
- ► EDA —using Pairplot, Boxplot
- ▶ Removing Constant and Quasi Constant Features : eg Magazine, Receive More updates about Course...etc
- ▶ Dealing with High Cardinal Features by Clubbing Rare Categories : eg Prospect ID, Leadnumber...etc
- ► Feature Scaling using Standard Scaler
- RFE For Feature Elimination.
- ► VIF Removing Multicolinearity
- ► Trained model with reduced dataset
- ▶ Model Evaluation using F1 Score, Accuracy, Recall and Precision

Evaluation Metrices

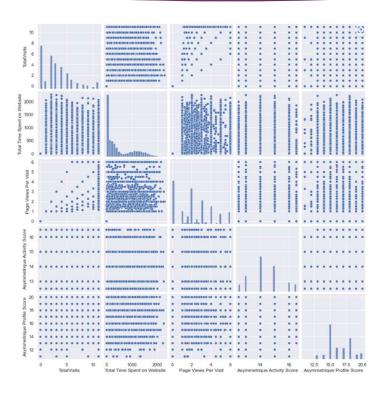
- ► Accuracy ~ 89.90%
- ► Precision ~ 89.55%
- ► Recall ~ 83%

Visualizations :

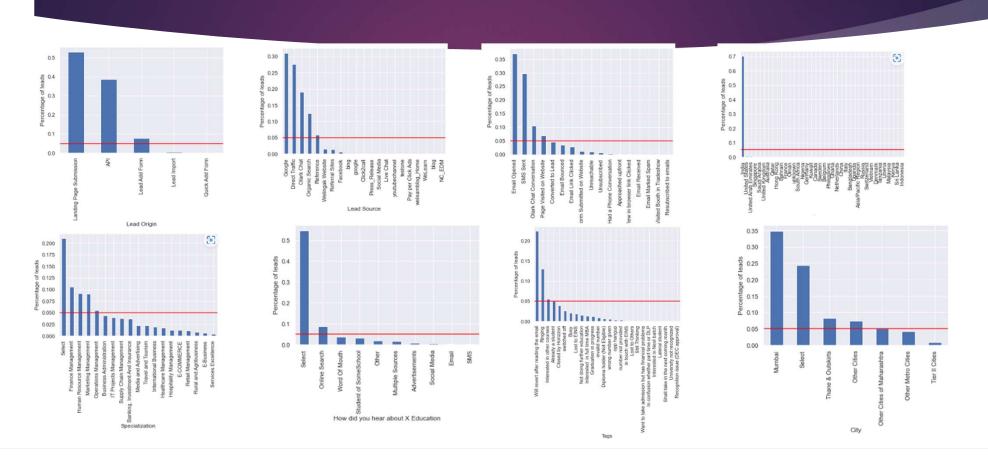
Outlier Analysis:



Pairplot:

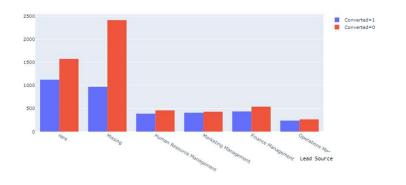


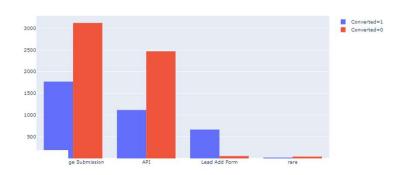
Features with Rare Categories:

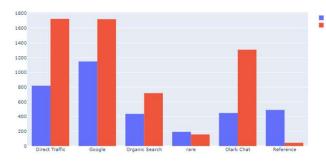


EDA: Categorical vs Target Variable (Converted)

Specialization Lead Origin



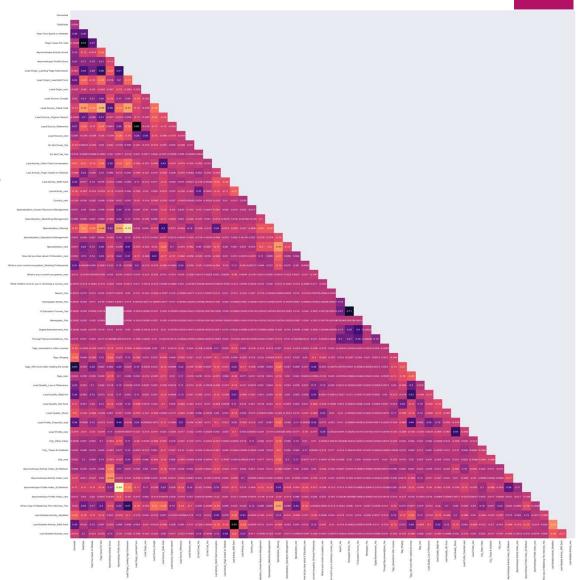




Observations From above plots

- > Higher no of customers converted who were unemployed and Working
- Less customers converted when Last notable activity is Email opened and modified
- Less customers converted when Asymmetrique Profile Index = Medium
- ➤ Highest People Converted when Tags = Will Revert After Reading
- ➤ Highest People Converted when Lead Quality = Mightbe
- ➤ More People Converted when Lead Profile = Potential Lead
- ➤ More number Converted were from MUMbai Clty
- ➤ More number Converted were from GOOGLE and Direct Traffic





Final Dataset after Preprocessing:

➤ Total Columns: 46

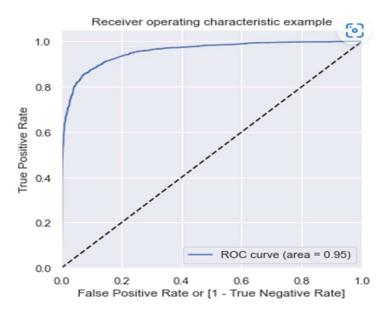
Total Data Points (Rows): 9240

MODEL BUILDING

Model Building Strategy

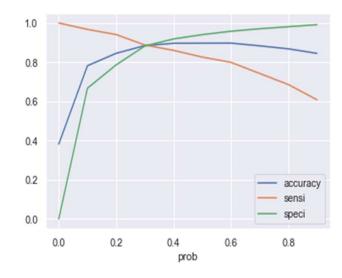
- ▶ Splitting into Train Test Split 70:30 ratio of split
- ▶ RFE For Feature Selection: 15 variables as output
- ▶ Building model by selecting variable with p value < 0.05
- ► Applying Vif to remove multicolinearity
- Prediction on Test set :
- Evaluating using multiple Metrics
- Final Accuracy: 89.90%
- ► Recall: 83%
- ▶ Precision 89%

ROC Curve:



Accuracy Sensitivity Specificity Curve

- ► Finding Optimal Cutoff point
- ► The sensitivity and specificity of a quantitative test are dependent on the cut-off value above or below which the test is positive.
- In general, the higher the sensitivity, the lower the specificity, and vice versa.



Important Features

- ► If Tags_Will revert after reading the email
- Lead Origin is from category Lead Add Form
- ► Tags is Ringing
- ► Total Time Spent on Website is high
- ► Last Notable Activity is SMS Sent
- ► Lead Quality is not Worst
- ► Tags are Interested in other courses
- ▶ Lead Source is Olark Chat
- ▶ Lead Profile is Potential Lead
- ► Last Activity is Olark Chat Conversation
- Page Views Per Visit is more
- Do Not Email is Yes
- ▶ TotalVisits is more
- ► Lead Source is through Reference
- Asymmetrique Activity Indexis from rare

If X Education
Focuses on above
Features and Targets
Customers based on
these Behavior the
Lead Conversion
rate will increase
and There will be
many Hot Leads



Thank you