

LEAD SCORE CASE STUDY

Case study Partners:

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Problem statement:

X Education sells online courses to industry professionals.

- ✓ X Education gets a lot of leads, its lead conversion rate is very poor. For example, if, say, they acquire 100 leads in a day, only about 30 of them are converted.
- ✓ To make this process more efficient, the company wishes to identify the most potential leads, also known as 'Hot Leads'.
- ✓ If they successfully identify this set of leads, the lead conversion rate should go up as the sales team will now be focusing more on communicating with the potential leads rather than making calls to everyone.

Business Objective:

- ✓ X education wants to know most promising leads.
- ✓ For that they want to build a Model which identifies the hot leads.
- ✓ Deployment of the model for the future use.

Steps:

➤ **Data cleaning and data manipulation:**

1. Check and handle duplicate data.
2. Check and handle NA values and missing values.
3. Drop columns, if it contains large amount of missing values and not useful for the analysis.
4. Imputation of the values, if necessary.
5. Check and handle outliers in data.

➤ **EDA:**

1. Univariate data analysis: value count, distribution of variable etc.
2. Bivariate data analysis: correlation coefficients and pattern between the variables etc.

➤ Feature Scaling & Dummy Variables and encoding of the data.

➤ Classification technique: logistic regression used for the model making and prediction.

➤ Validation of the model.

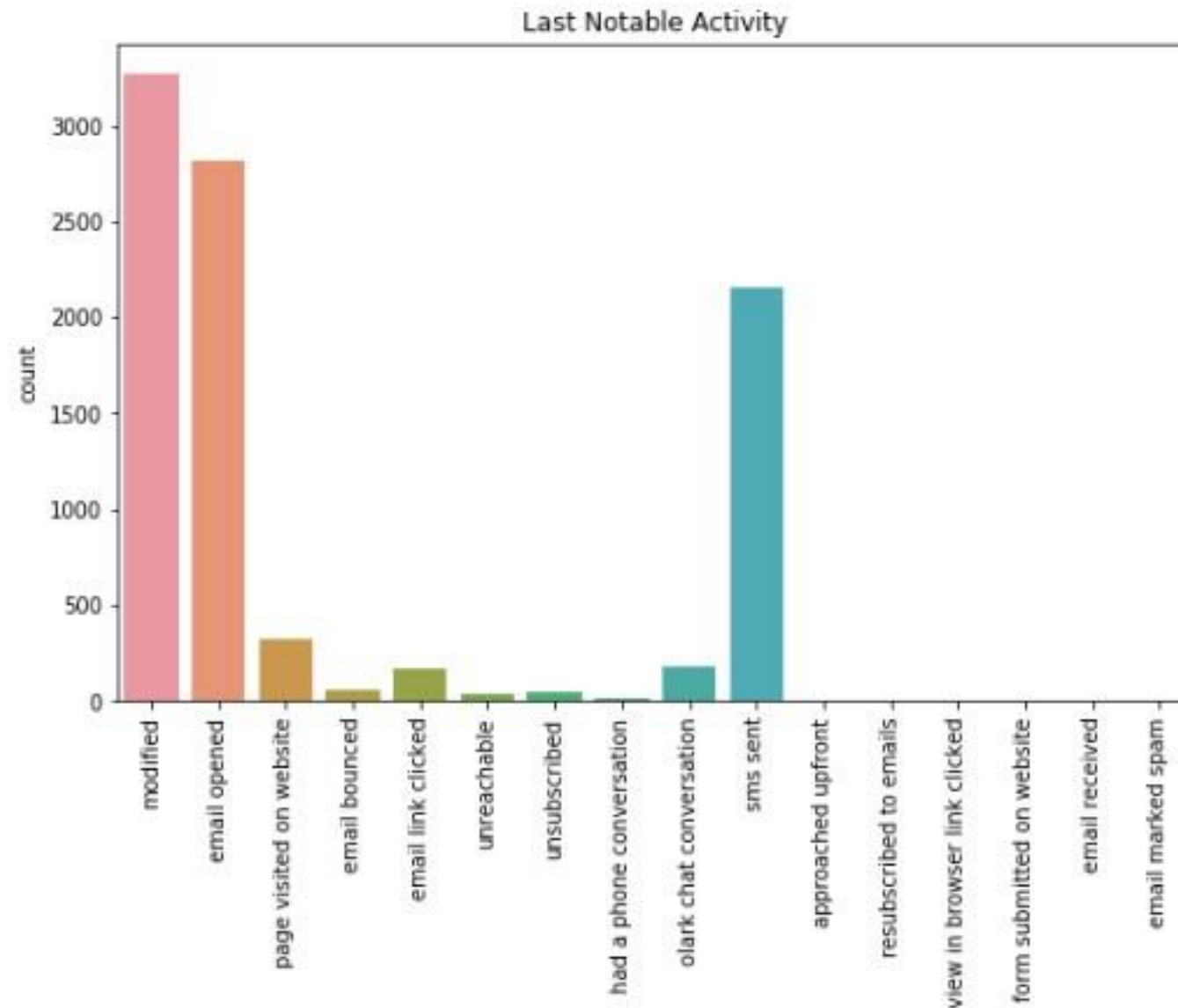
➤ Model presentation.

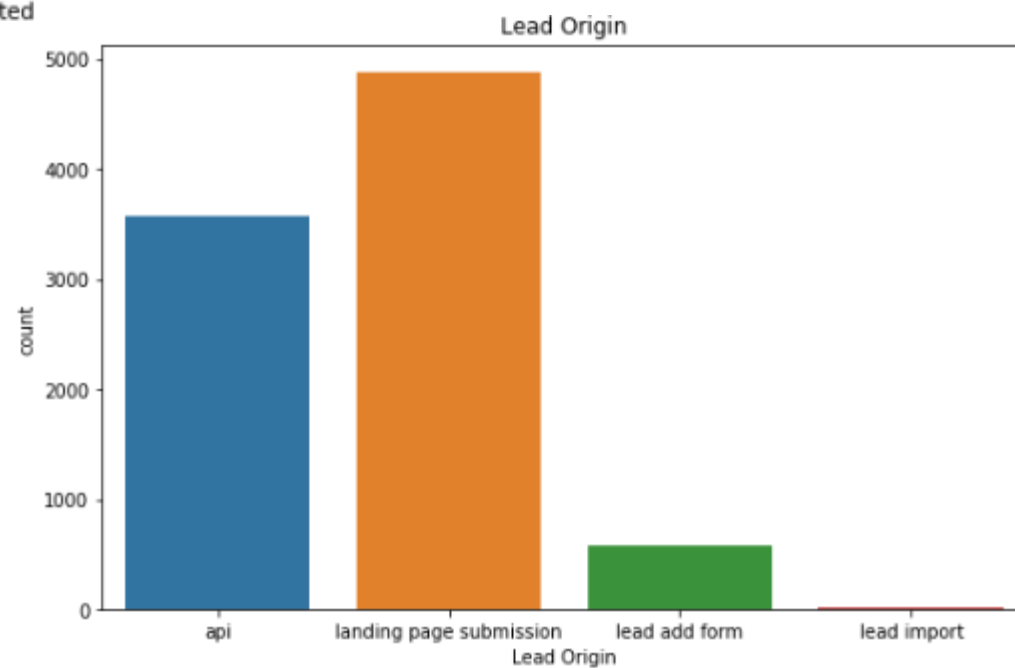
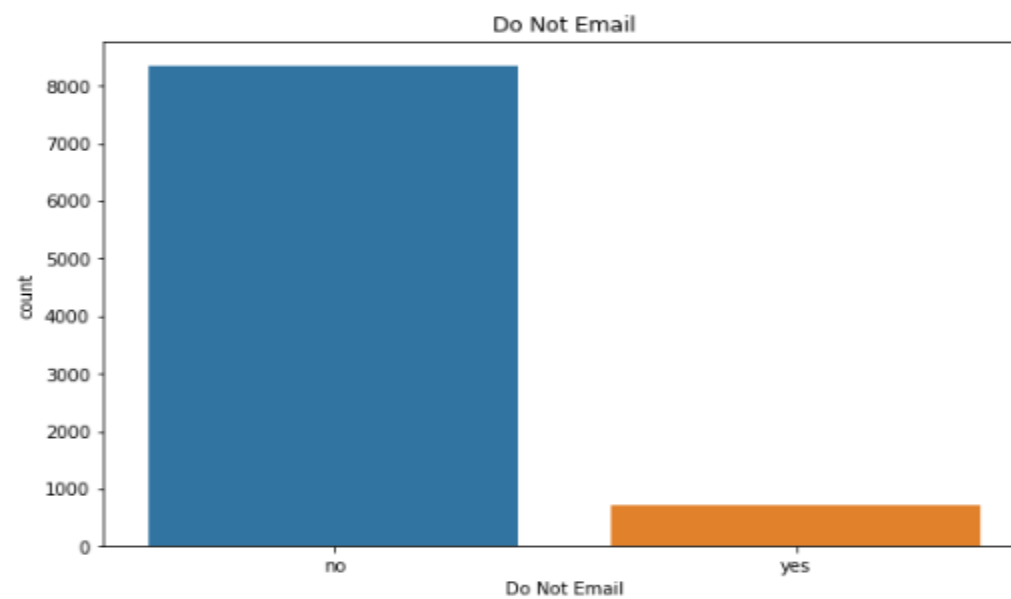
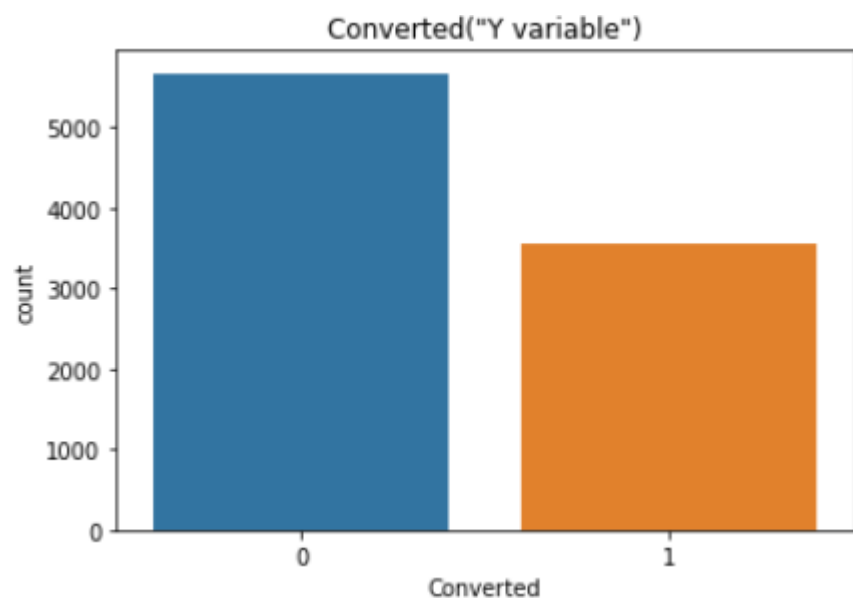
➤ Conclusions and recommendations.

Data Handling:

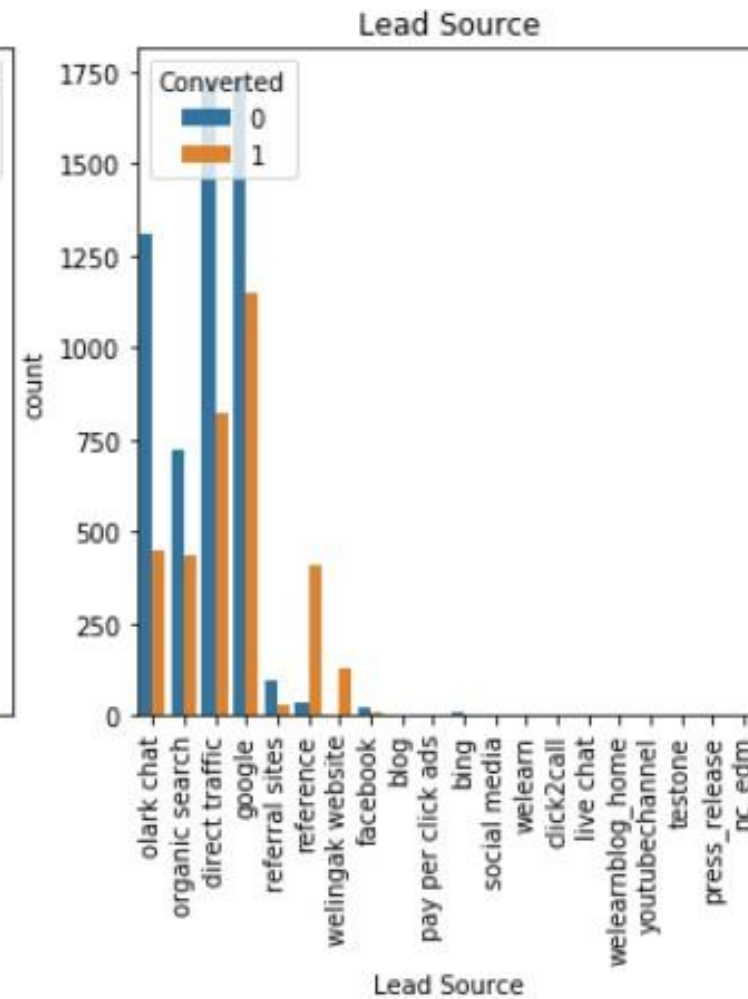
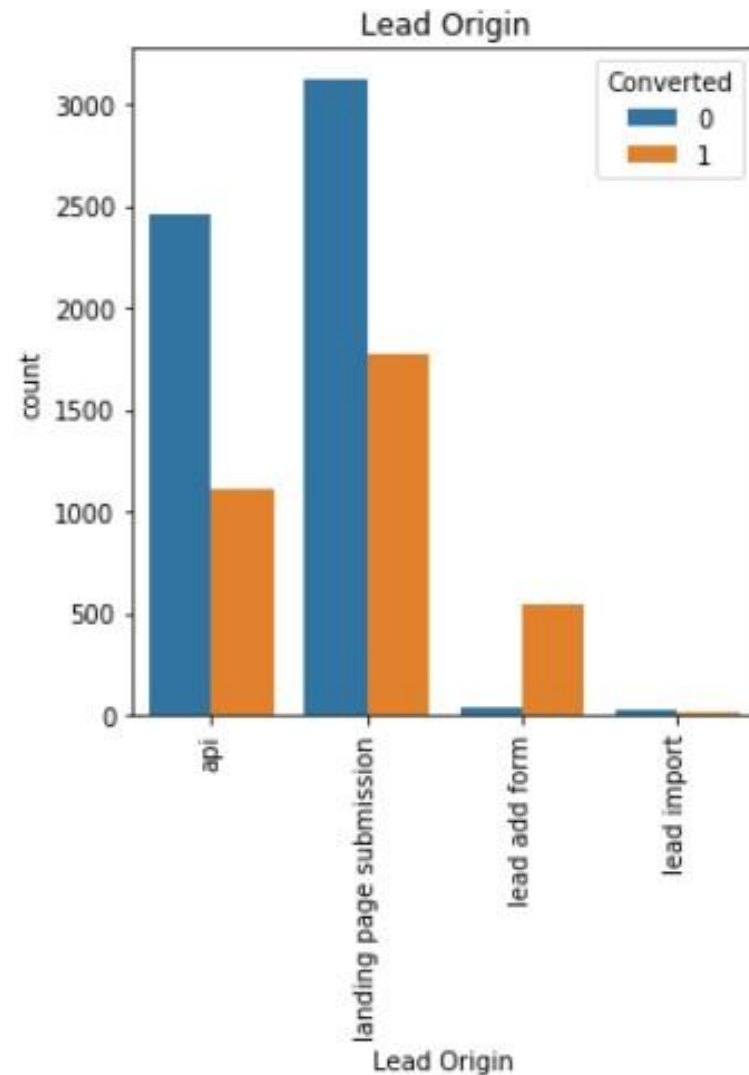
- Total Number of Rows =37, Total Number of Columns =9240.
- Single value features like “Magazine”, “Receive More Updates About Our Courses”, “Update me on Supply”
- Chain Content”, “Get updates on DM Content”, “I agree to pay the amount through cheque” etc. have been dropped.
- Removing the “Prospect ID” and “Lead Number” which is not necessary for the analysis.
- After checking for the value counts for some of the object type variables, we find some of the features which has no enough variance, which we have dropped, the features are: “Do Not Call”, “What matters most to you in choosing course”, “Search”, “Newspaper Article”, “X Education Forums”, “Newspaper”, “Digital Advertisement” etc.
- Dropping the columns having more than 35% as missing value such as ‘How did you hear about X Education’ and ‘Lead Profile’.

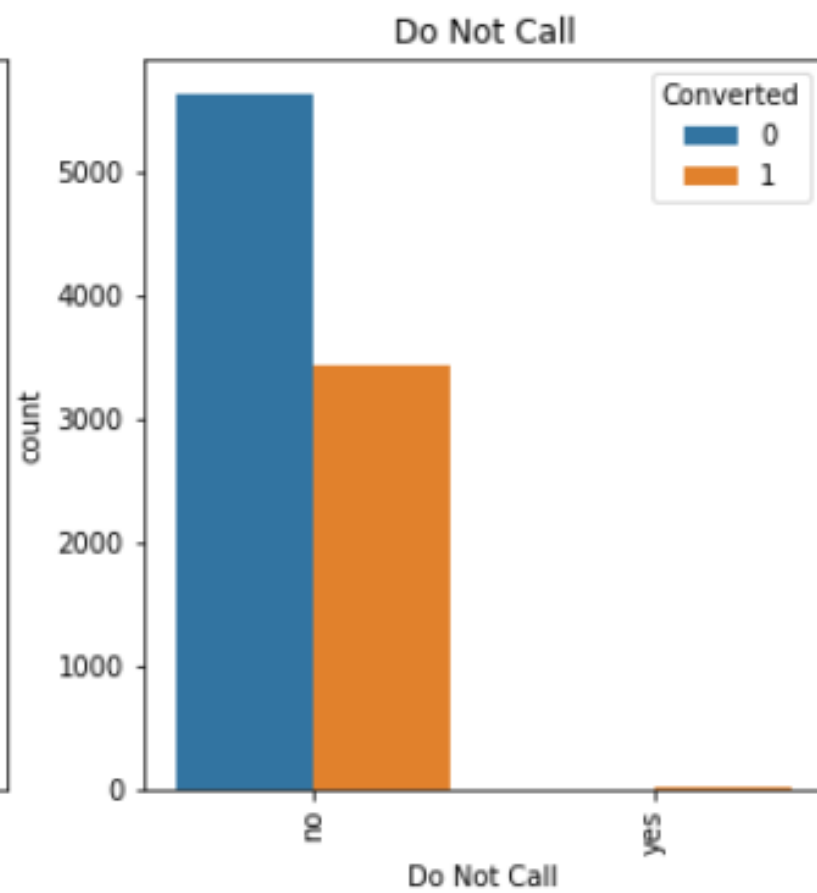
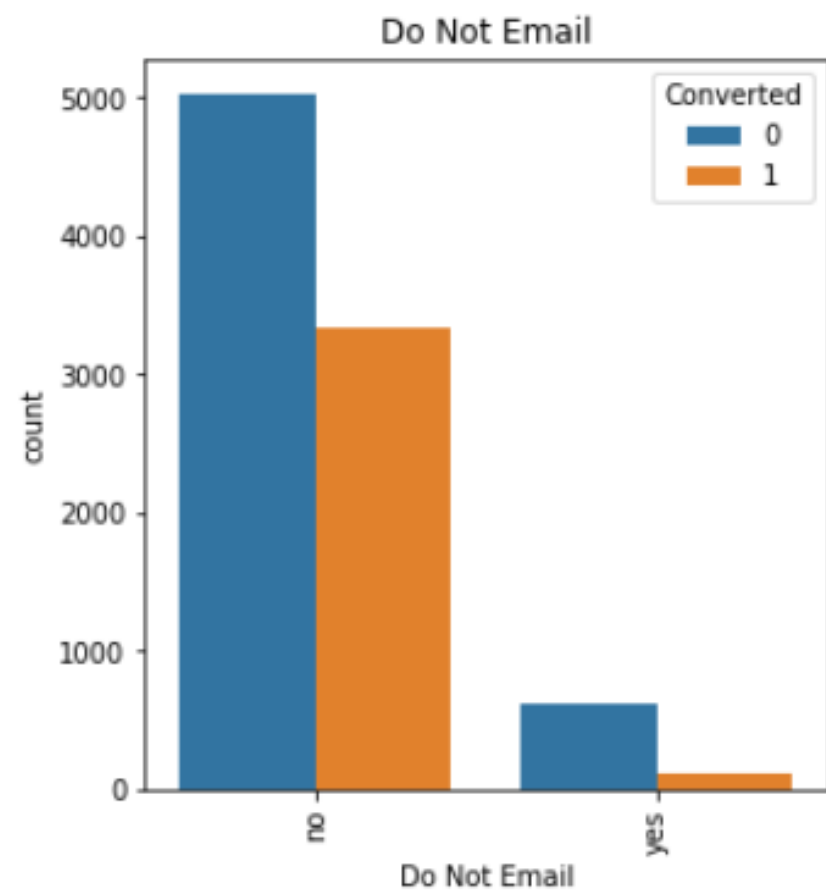
EDA : (visualization Graphs below)

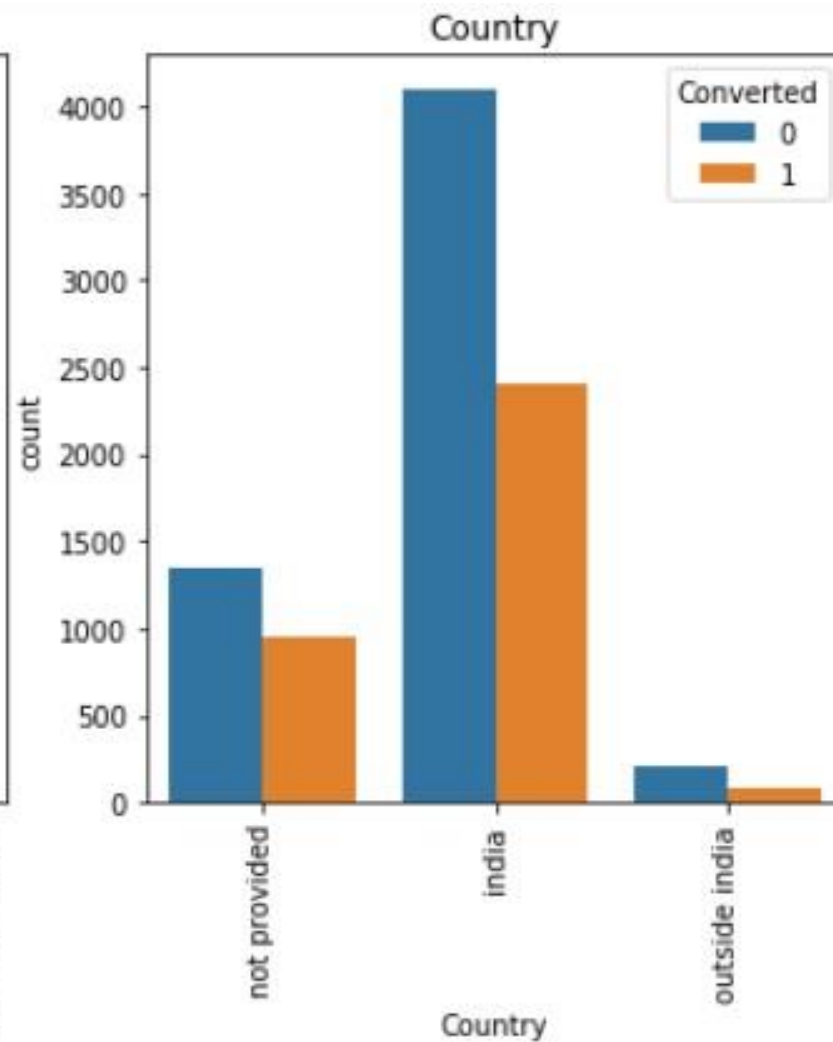
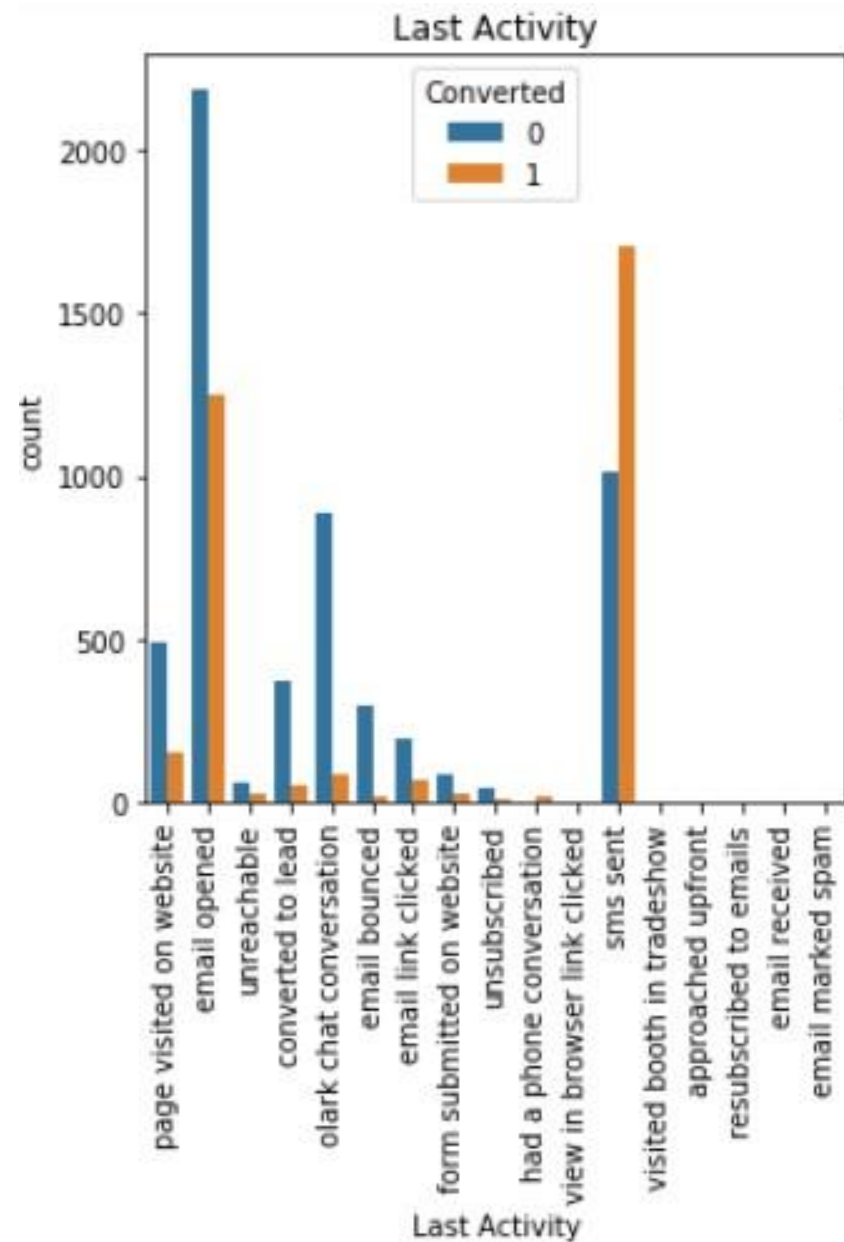




Categorical Variable Relation:







Data Normalization & Standardization :

- Numerical Variables are Normalized (Min-max scaler used)
- Dummy Variables are created for object type variables
- Total Rows for Analysis: 8792
- Total Columns for Analysis: 43

Model Building

- Splitting the Data into Training and Testing Sets
- The first basic step for regression is performing a train-test split, we have chosen 70:30 ratio.
- Use RFE for Feature Selection
- Running RFE with 15 variables as output
- Building Model by removing the variable whose p- value is greater than 0.05 and vif value is greater than 5
- Predictions on test data set
- Overall accuracy 81%

Model Evaluation:

- A confusion matrix was made.
- Later on the optimum cut off value (using ROC curve) was used to find the
 - accuracy
 - sensitivity and
 - specificity which came to be around 80% each

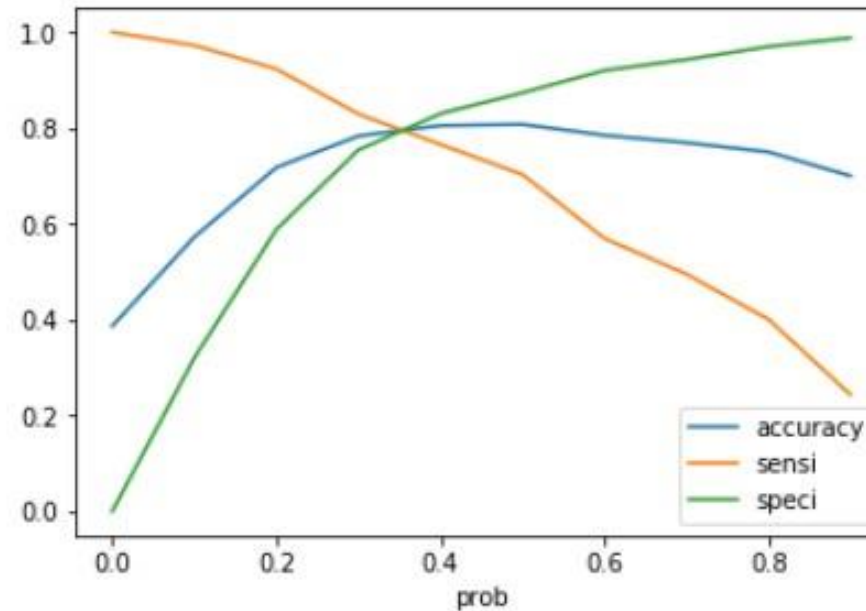
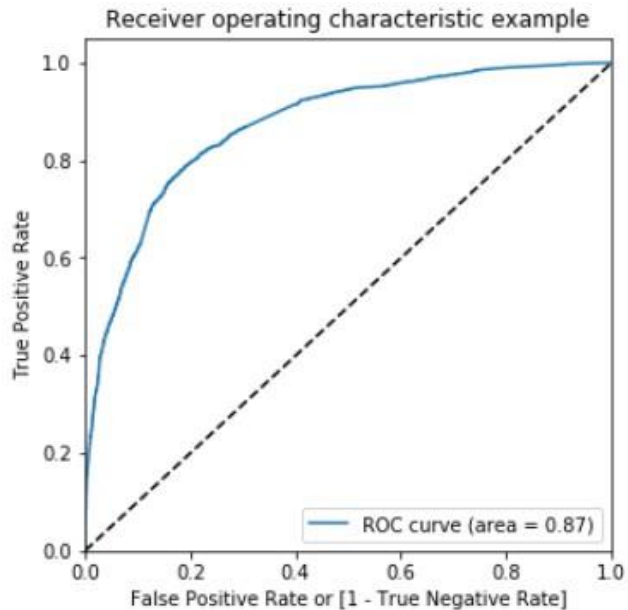
Prediction:

- Prediction was done on the test data frame and with an optimum cut off as 0.35 with accuracy, sensitivity and specificity of 80%.

Precision – Recall:

- This method was also used to recheck and a cut off of 0.41 was found with Precision around 73% and recall around 75% on the test data frame

ROC Curve



Finding Optimal Cut off Point

- Optimal cut off probability is that
- probability where we get balanced sensitivity and specificity.
- From the second graph it is visible that the optimal cut off is at 0.35.

Conclusion

It was found that the variables that mattered the most in the potential buyers are
(In descending order)

- The total time spend on the Website.
- Total number of visits
- When the lead source was:
 - Google
 - Direct traffic
 - Organic search
 - Welingak website
- When the last activity was:
 - SMS
 - Olark chat conversation
- When the lead origin is Lead add format
- When their current occupation is as a working professional.

Thus “X Education company” can flourish as they have a very high chance to get almost all the potential buyers to change their mind and buy their courses.