

Lead Scoring Case Study

Master of Data Science

- upGrad

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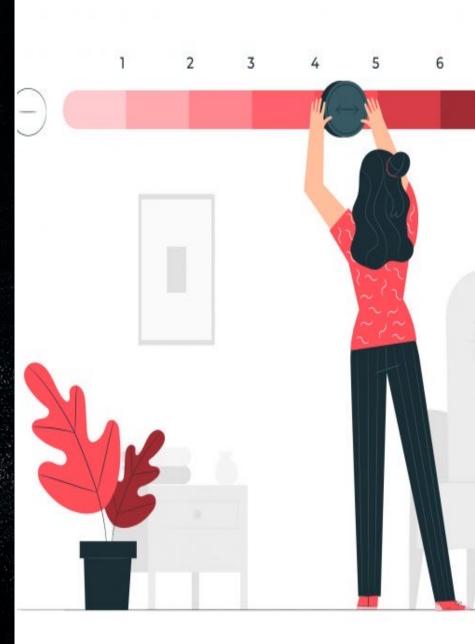
Course 2: Machine Learning - I -

Lead Scoring case Study

February 27th, 2023

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I. Introduction

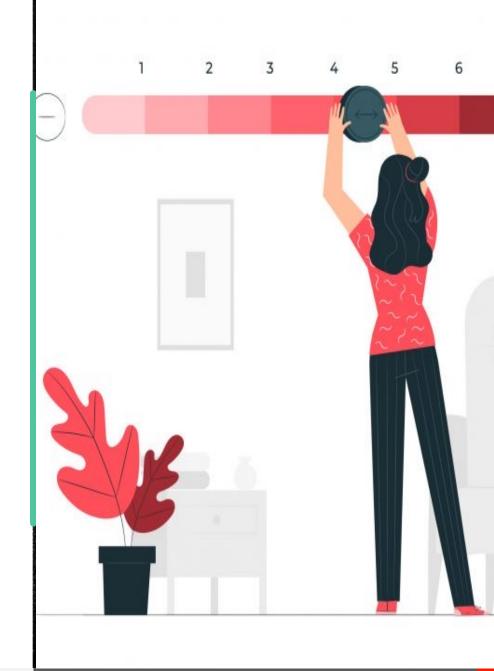
- X Education sells online courses to industry professionals on several websites and search engines.
- The typical lead conversion rate at X education is around 30%.
- The company wants to identify the most potential leads (Hot Leads) and improve lead conversion rate by building a model assigning a lead score to each of the leads.
- The CEO has given a ballpark of the target lead conversion rate around 80%.

• Target variable:

- 1 lead was converted
- 0 lead was not converted

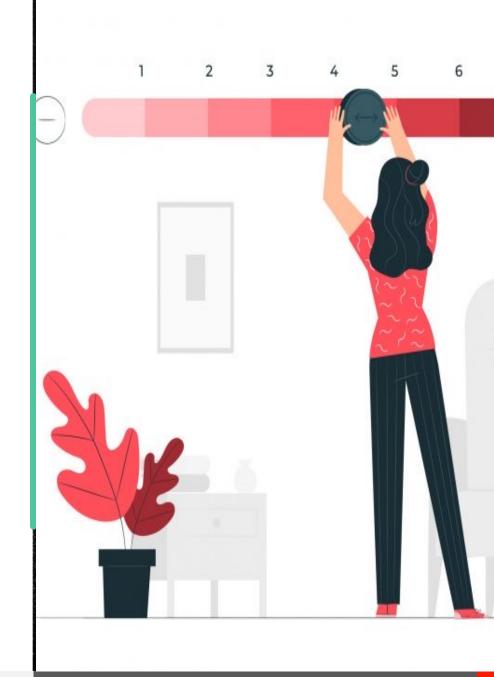
Goals:

- Build a logistic regression model to assign a lead score between 0 and 100 to each of the leads
- Suggest solutions to related problems to cope with changes in the future.



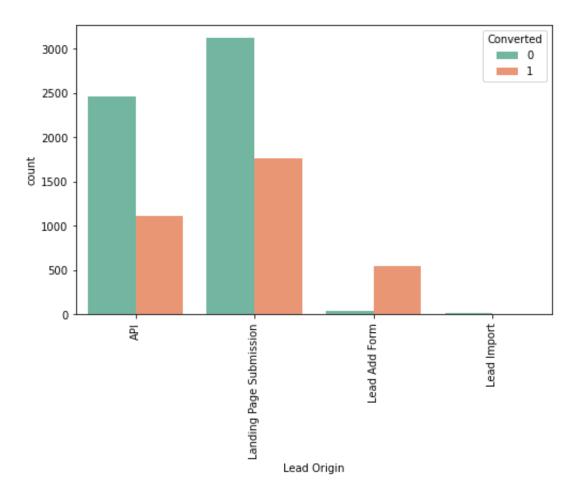
II. Data Cleaning

- Since there are many 'Select' values in our dataset as some customers did not select any option from the list, those are as good as NULL. Hence, we will convert them to null values.
- We will drop the columns with missing values > 40%.
- Columns with moderate missing values:
- ✓ Impute missing values with mode
- ✓ Create 'Others' category for insignificant values
- ✓ Should values fall into only one unique value, we drop other columns.
- ⇒ At this stage, we have retained 98% of the rows after cleaning the data.
- → Null percentages < 1%



3.1. Univariate Analysis and Bivariate Analysis

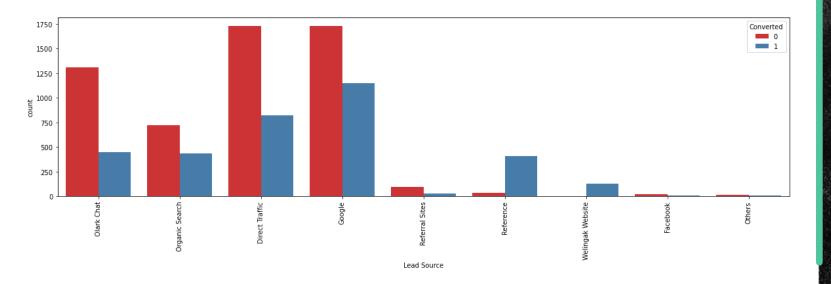
Lead Origin



- API and Landing Page Submission have around 30-35% conversion rate while the number of lead originated from them are considerable.
- Lead Add Form has > 90% conversion rate but the number of lead are not significant.
- Lead Import are also not noticeable.
- => To improve the lead conversion rate, X
 Education should focus on improving lead
 conversion of API and Landing Page Submission
 origin and generate more leads from Lead Add
 Form.

3.1. Univariate Analysis and Bivariate Analysis

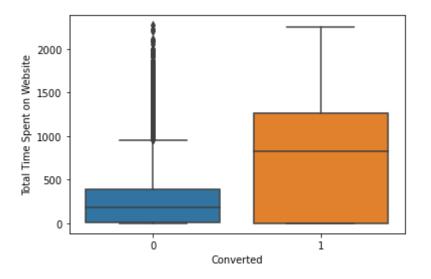
Lead Source



- Google and Direct traffic generates most of leads.
- Conversion Rate of Reference leads and Welingak Website is high.
- => To improve lead conversion rate, X Education should focus on improving lead conversion of Olark Chat, Organic Search, Direct Traffic, and Google leads and generate more leads from Reference and Welingak Website.

3.1. Univariate Analysis and Bivariate Analysis

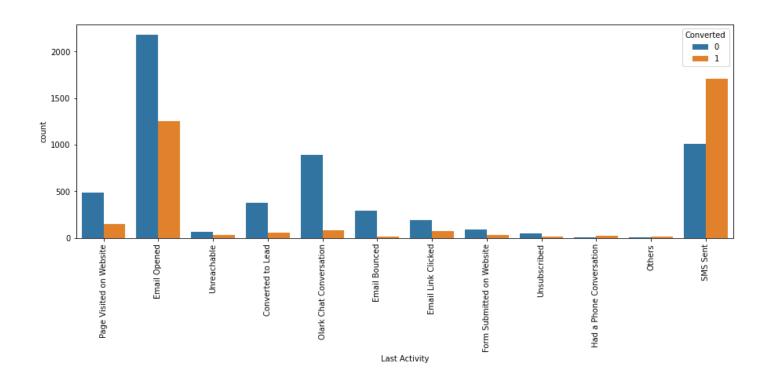
Total Time Spent on Website



- Leads spending more time on the website are more likely to be converted.
- => Website should be made more attractive and engaging to make leads spend more time on the website.

3.1. Univariate Analysis and Bivariate Analysis

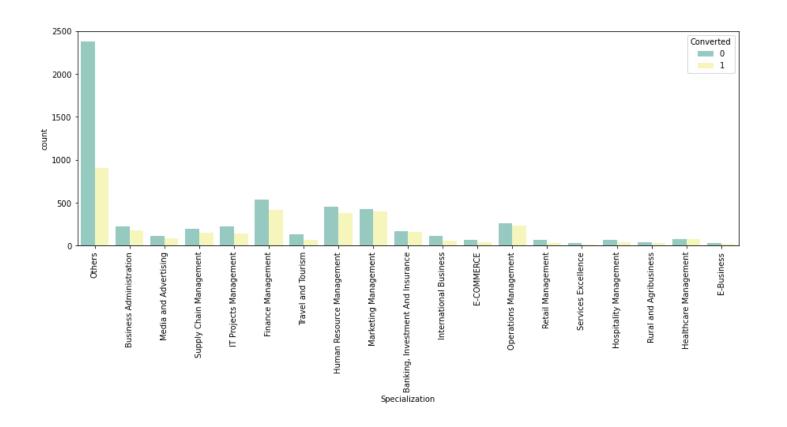
Last Activity



- Most of the leads' last activity is opening their emails.
- Nearly 60% of leads with last activity as SMS
 Sent is converted.

3.1. Univariate Analysis and Bivariate Analysis

Specialization

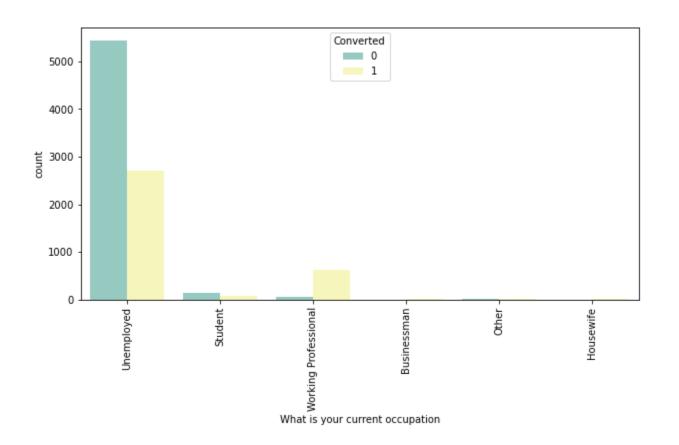


Findings:

 X Education should pay more attention to Specializations with high conversion.
 However, there is no significant difference between these parameters.

3.1. Univariate Analysis and Bivariate Analysis

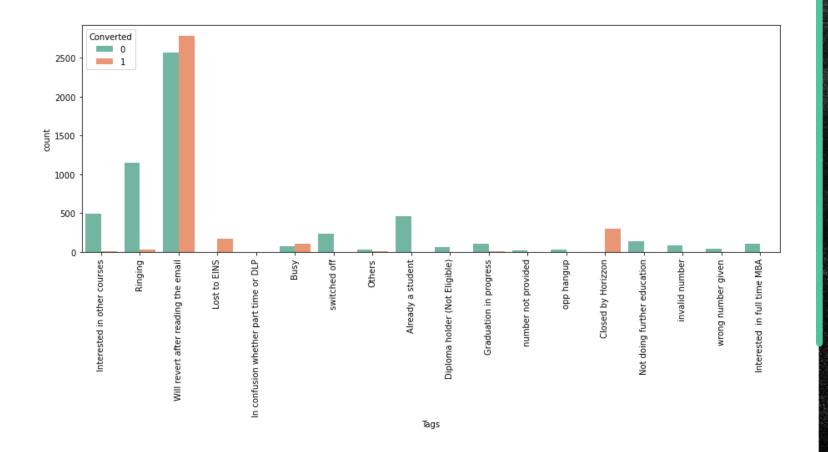
What is your current occupation



- Working Professionals are more likely to be converted comparing to other categories.
- Unemployed leads dominant the dataset; however, it has around solely 30-35% conversion rate.

3.1. Univariate Analysis and Bivariate Analysis

Tags

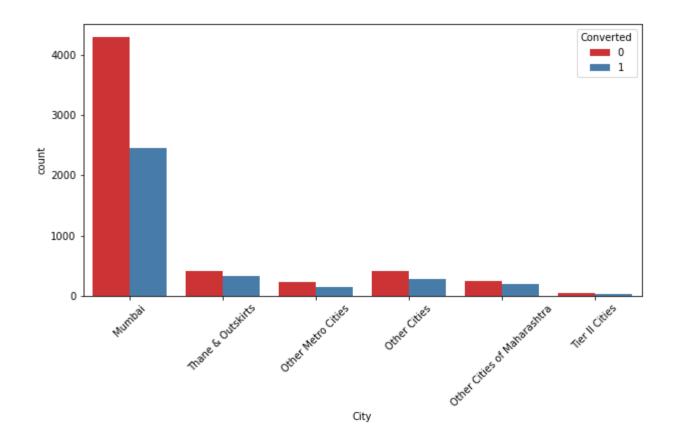


Findings:

 This is not available for model building as they are notes by sales team for tracking.
 Thus, we will drop this column before building the model.

3.1. Univariate Analysis and Bivariate Analysis

City

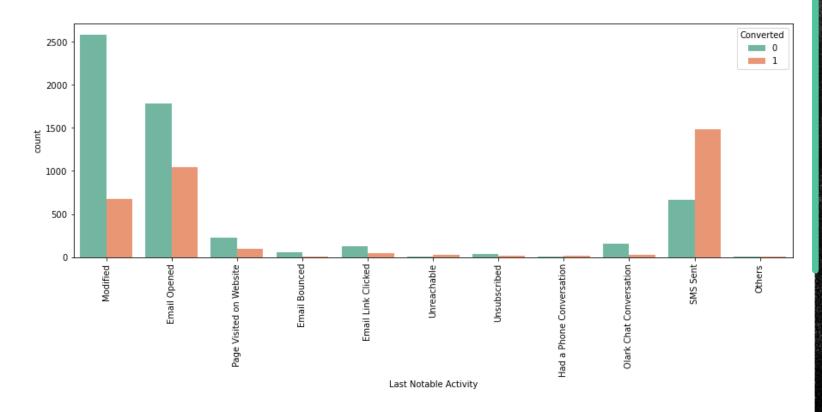


Findings:

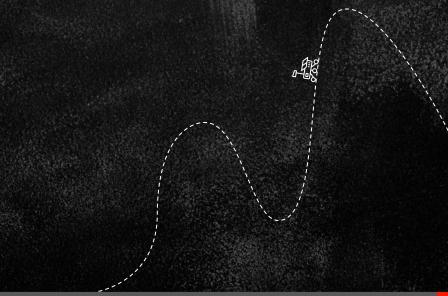
• Most leads are from Mumbai with nearly 40% conversion rate.

3.1. Univariate Analysis and Bivariate Analysis

Last Notable Activity



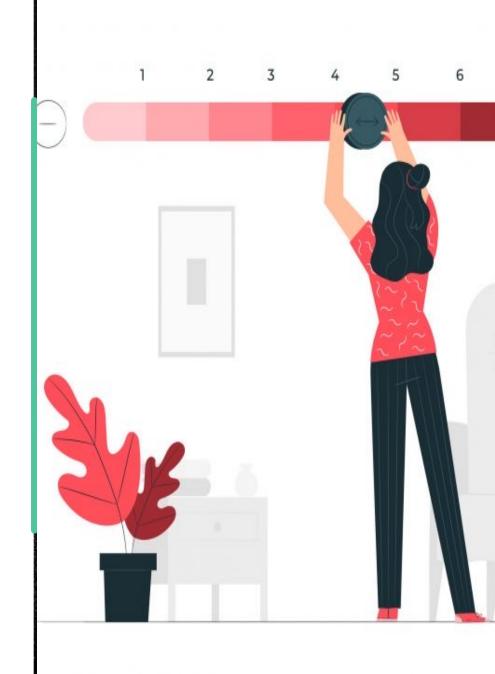
- Most leads came from Modified or Opened Email as their last notable activities. Leads which received SMS has high conversion rate at about 65%.
- From the above-mentioned Univariate
 Analysis and Bivariate Analysis, we notice
 that some columns do not have statistic
 values. Therefore, we can drop them for our
 best model in later analysis.



3.2. Data Preparation

In this session, we will:

- Converting binary variables: 'Do Not Email', 'Do Not Call'
- Handling Dummy variables: 'Lead Origin', 'Lead Source', 'Last Activity',
 'Specialization', 'What is your current occupation', 'City', 'Last Notable Activity'
- Splitting the data into train and test set
- Rescaling the Features: Normalisation



IV. Building Model

4.1. SciKit Learn LogisticRegression

We will use the SciKit Learn LogisticRegression function for its compatibility with RFE (Recursive Feature Elimination - a utility from sklearn). After using statsmodel and Checking VIF, we found our best model at 7th attempt as below:

	Features	VIF
10	Last Notable Activity_Modified	1.64
2	Lead Origin_Landing Page Submission	1.63
3	Lead Source_Olark Chat	1.60
7	Last Activity_Olark Chat Conversation	1.55
8	Last Activity_SMS Sent	1.47
1	Total Time Spent on Website	1.29
4	Lead Source_Reference	1.23
9	What is your current occupation_Working Profes	1.18
0	Do Not Email	1.13
5	Lead Source_Welingak Website	1.05
6	Last Activity_Had a Phone Conversation	1.00

Findings:

The VIFs and p-values are both within acceptable limits. Hence, we proceed to generate predictions using this model.

- For each term, the p-value tests the null hypothesis that the coefficient is less than 0.05. A low p-value (< 0.05) suggests that the null hypothesis may be rejected.
- Generally, if a VIF is more than 10, we have strong multicollinearity. In our scenario, VIFs < 5 indicate that we are in good shape and may continue with our regression.
- Model 7 is our final model. We have 11 variables in our final model.

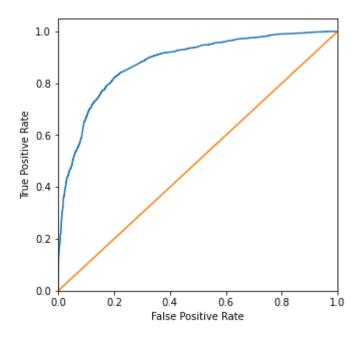
V. Making Predictions

Our specificity was good (\sim 89%) but sensitivity was only well-under 70%. This is due to our initial chosen cut-off point of 0.5.

We use ROC curve to optimize sensitivity as:

- It shows the tradeoff between sensitivity and specificity
- It depicts how accurate the test is

ROC Curve

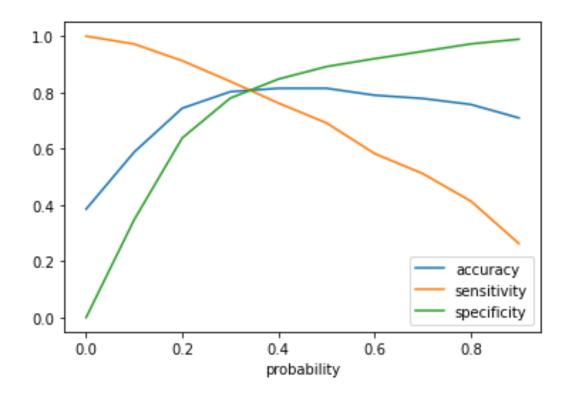


Findings:

• As our ROC curve is close to the left-hand and top border, our model is good, and the test is accurate.

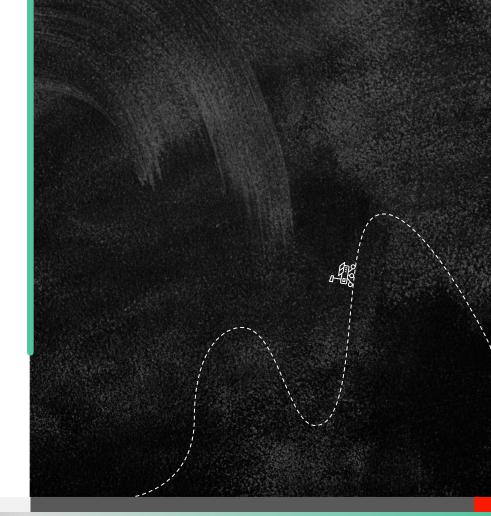
V. Making Predictions

Let's find the optimal cut-off probability where there is balanced sensitivity and specificity.



Findings:

• The chart suggests that the optimal cut-off probability should be about 0.35



VI. Model Evaluation

Train Data:

• Accuracy: 80.9 %

• Sensitivity: 80.4 %

• Specificity: 81.2 %

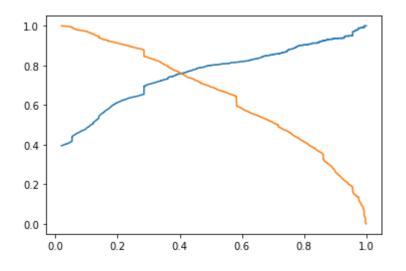
Test Data:

• Accuracy: 80.3 %

• Sensitivity: 79.0 %

• Specificity: 81.3 %

Precision and recall tradeoff



Findings:

• This is a good model which can predict the Conversion Rate well. The CEO has now strong confidence in making strategic decision as the model has given a ballpark of the target lead conversion rate to be around 80%.

VII. Conclusion

• There are 364 hot leads which should be contacted as they are more likely to be converted with the Score >=85

Recommendations:

X Education should target those groups of customers who have high chance to be converted:

- Calls leads coming from the lead sources 'Welingak Websites', 'Reference', and 'Olark Chat'
- Calls leads whose last activity was 'Phone Conversation' and 'SMS Sent'
- Calls leads who are the 'Working Professionals'
- Calls leads who spent more time on the websites ('Total Time Spent on Website')

