

Logistic Regression Lead Score

Updated: 13 Mar 2024



Step 1: Importing and Data general overview

About the dataset

- The data set contains 9240 rows and 37 columns.
- Prospect ID: A unique ID with which the customer is identified.
- Converted: The target variable. Indicates whether a lead has been successfully converted or not.
- 35 independence variables
- Data types: float64(4), int64(3), object(30)

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9240 entries, 0 to 9239
Data columns (total 37 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   Prospect ID                             9240 non-null   object
1   Lead Number                             9240 non-null   int64
2   Lead Origin                             9240 non-null   object
3   Lead Source                             9204 non-null   object
4   Do Not Email                            9240 non-null   object
5   Do Not Call                             9240 non-null   object
6   Converted                               9240 non-null   int64
7   TotalVisits                             9103 non-null   float64
8   Total Time Spent on Website              9240 non-null   int64
9   Page Views Per Visit                    9103 non-null   float64
10  Last Activity                           9137 non-null   object
11  Country                                 6779 non-null   object
12  Specialization                           7802 non-null   object
13  How did you hear about X Education       7033 non-null   object
14  What is your current occupation          6550 non-null   object
15  What matters most to you in choosing a course  6531 non-null   object
16  Search                                  9240 non-null   object
17  Magazine                                9240 non-null   object
18  Newspaper Article                       9240 non-null   object
19  X Education Forums                      9240 non-null   object
...
35  A free copy of Mastering The Interview    9240 non-null   object
36  Last Notable Activity                    9240 non-null   object
dtypes: float64(4), int64(3), object(30)
```

Step 2: EDA and Cleansing Data - Converted variable analysis



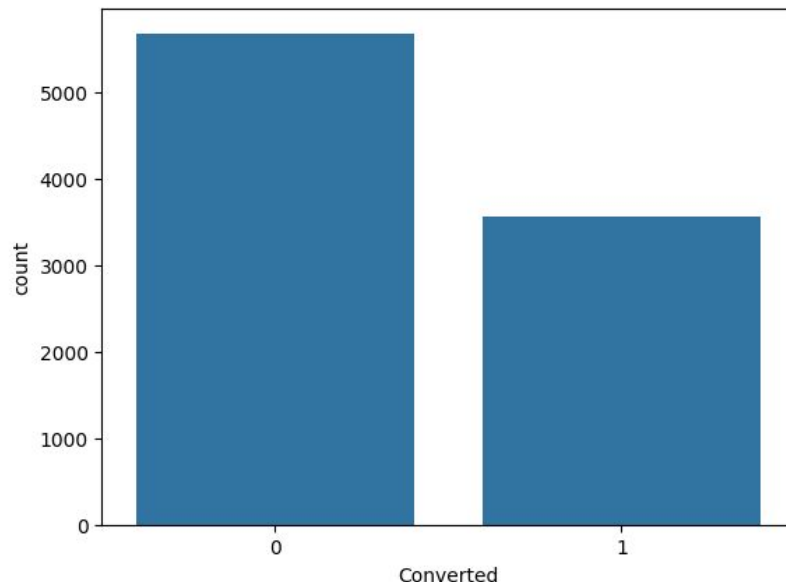
The target variable in your dataset, which represents whether a lead has been successfully converted or not with 0 for "not converted" and 1 for "converted successfully," would be considered a categorical variable.

HOW TO DO

- 1/ Check unique value of target column
- 2/ Change Converted into category
- 3/ Show the distribution of unique value in Converted by countplot
- 4/ Count numbers of each unique values for the target column

RESULT:

- Target var: should be category.
- There are 5679 customers with converted unsuccessfully with value 0 in target column and 3561 are converted successfully with value 1.
- % customer can be converted successfully in data frame: 38.5% in this dataset





Step 2: EDA and Cleansing Data - Missing value treatment

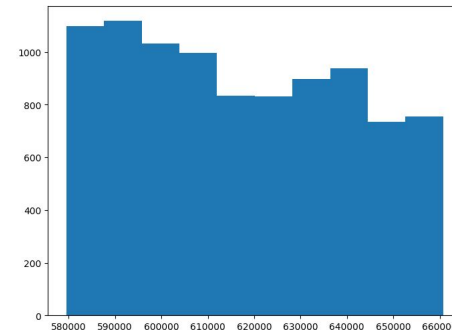
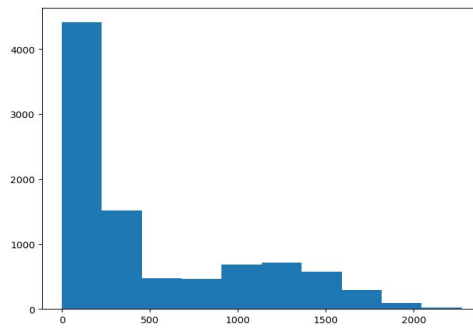
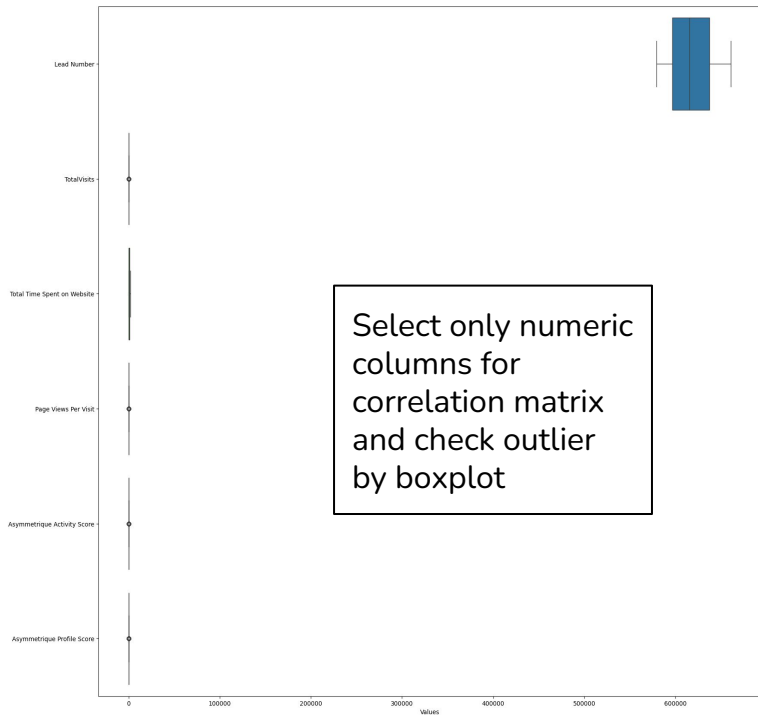
HOW TO DO

- 1/ Finding missing data and check % missing value in each column.
- 2/ Drop cols having missing values.
- 3/ Replace missing values with the mean value.
- 4/ Check unique value of some cols before making decision of treating missing values.
- 5/ Replace missing values in specified columns with "Other" and "Select".
- 6/ Check % missing value in each column.

RESULT: No missing value in data lead

```
Prospect ID      0.00
Lead Number      0.00
Lead Origin      0.00
Lead Source      0.39
Do Not Email     0.00
Do Not Call      0.00
Converted         0.00
TotalVisits      0.00
Total Time Spent on Website 0.00
Page Views Per Visit 0.00
Last Activity     1.11
Country          0.00
Specialization   0.00
How did you hear about X Education 0.00
What is your current occupation 0.00
What matters most to you in choosing a course 0.00
Search           0.00
Magazine         0.00
Newspaper Article 0.00
X Education Forums 0.00
Newspaper        0.00
Digital Advertisement 0.00
Through Recommendations 0.00
Receive More Updates About Our Courses 0.00
Tags             0.00
...
Asymmetrique Profile Score 0.00
I agree to pay the amount through cheque 0.00
A free copy of Mastering The Interview 0.00
Last Notable Activity 0.00
dtype: float64
```

Step 2: EDA and Cleansing Data - Check outliers



Hist for some num col having outliers

Step 3: Data preparation

HOW TO DO

- 1/ Transform binary variables by one hot encoding(Yes/No >> 0/1)
- 2/ Transform remaining category variables into numeric by dummies

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9240 entries, 0 to 9239
Data columns (total 56 columns):
 #   Column                                                                 Non-Null Count  Dtype
---  -
 0   Tags_Interested in other courses                                     9240 non-null  int64
 1   Tags_Other                                                           9240 non-null  int64
 2   Tags_Ringing                                                         9240 non-null  int64
 3   Tags_Will revert after reading the email                           9240 non-null  int64
 4   Last Activity_Email Opened                                          9240 non-null  int64
 5   Last Activity_Olark Chat Conversation                              9240 non-null  int64
 6   Last Activity_Other                                                  9240 non-null  int64
 7   Last Activity_Page Visited on Website                              9240 non-null  int64
 8   Last Activity_SMS Sent                                              9240 non-null  int64
 9   Lead Source_Google                                                  9240 non-null  int64
10   Lead Source_Olark Chat                                              9240 non-null  int64
11   Lead Source_Organic Search                                          9240 non-null  int64
12   Lead Source_Other                                                   9240 non-null  int64
13   Lead Source_Reference                                                9240 non-null  int64
14   Lead Origin_Landing Page Submission                                9240 non-null  int64
15   Lead Origin_Lead Add Form                                           9240 non-null  int64
16   Lead Origin_Lead Import                                             9240 non-null  int64
17   Lead Origin_Quick Add Form                                          9240 non-null  int64
18   City_Other                                                           9240 non-null  int64
19   City_Other Cities                                                   9240 non-null  int64
...
54  Asymmetrique Activity Score                                         9240 non-null  float64
55  Asymmetrique Profile Score                                          9240 non-null  float64
dtypes: float64(4), int64(52)
```



Step 4 - 5: Split into train - test set and Scale Data

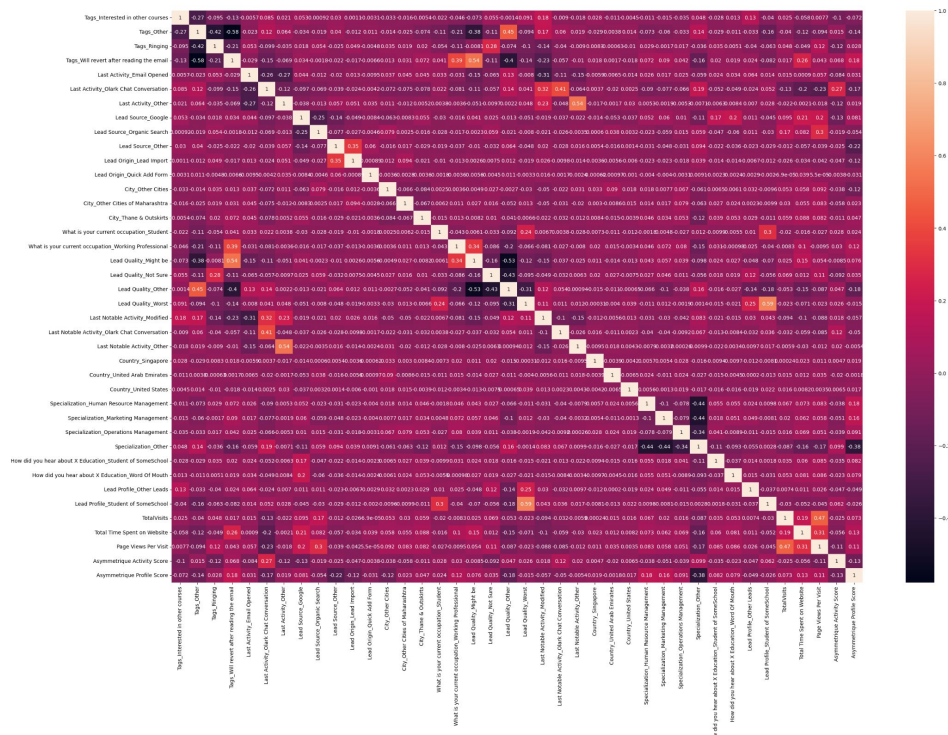
Split into train and test set

- 1/ Create subset y
- 2/ Split subsets X and y into train sets and test sets
- 3/ Check dimension of subsets

Scale Data

For continuous variables, we will use scaling method of Z-score Normalization (Standardization) for scaling these cols

- 1/ Find the variables having high correlation with one another $\geq 70\%$
- 2/ Remove high correlation attributes





Step 7: Model Building

HOW TO DO

- 1/ Train Model on X_train
- 2/ First model
- 3/ Using REF to choose the best attributes

Generalized Linear Model Regression Results

Dep. Variable:	Converted	No. Observations:	6468
Model:	GLM	Df Residuals:	6427
Model Family:	Binomial	Df Model:	40
Link Function:	Logit	Scale:	1.0000
Method:	IRLS	Log-Likelihood:	-1756.2
Date:	Wed, 13 Mar 2024	Deviance:	3512.3
Time:	03:36:30	Pearson chi2:	8.91e+03
No. Iterations:	19	Pseudo R-squ. (CS):	0.5445
Covariance Type:	nonrobust		



Step 8: Data Evaluation

Metrics

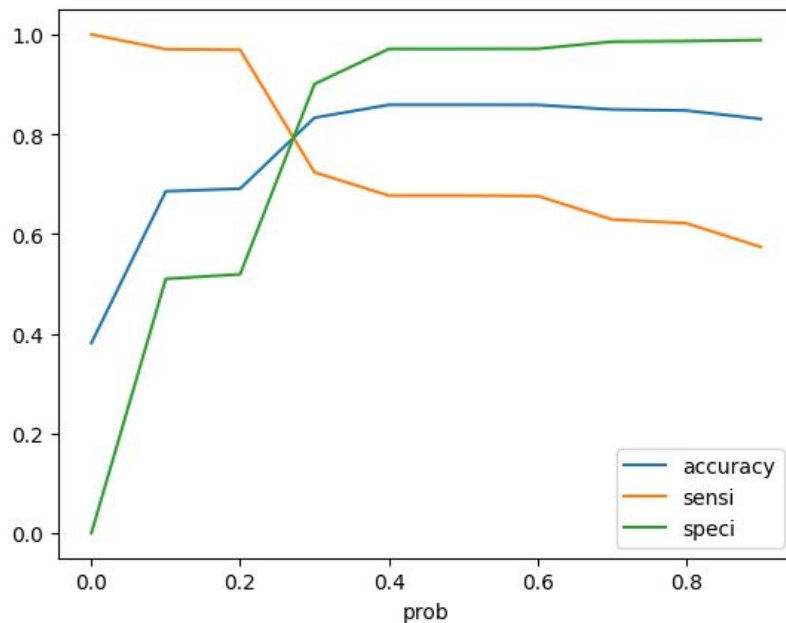
1/ Accuracy reach 85% on train set
2/ Specificity is 87% that mean most of positive samples are predicted correctly



Cut-off

Determine the threshold that give the best model base on ROC Curve

Step 8: Data Evaluation



Cut-off

As shown in the illustration, the cut-off is approximate 0.3



Step 8: Data Evaluation

Test Set with Cut-off 0.3

1/ Accuracy reach 84% on test set
2/ Specificity is 90% that mean most of positive samples are predicted correctly

```
metrics.accuracy_score(y_train_pred_final.Converted, y_train_pred_final.final_predicted)
```

```
0.8330241187384044
```

```
# specificity  
TN / float(TN+FP)
```

```
0.9005497251374313
```

Business Case Study

A user, becoming a lead when they have accessed to X Education's system: website, ads, surveys,...

From these customers' behaviors, the list of leads is built and the staffs will contact to them and persuade they become the real customer. This process could take much effort than it's necessary.



Business Case Study

Base on existing data, we try to build a machine learning model to handle this. The accuracy of model is about 85%, an considerate number to believe in practice. In fact, in some case, accuracy is not the most important metric we need to count. Specificity and Sensitivity are also crucial in special situation. The result of model return a probability for each input sample. Our mission is to choose an appropriate threshold for each condition.

