

LEAD SCORE CASE STUDY

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INTRODUCTION:

Professionals in the business can purchase online courses from X Education, an education company. The business advertises its classes on a number of websites and search engines, including Google.

After visiting the website, users can peruse the available courses, complete the course registration form, or watch some videos. These folks are categorized as leads when they complete a form with their phone number or email address. Additionally, the business receives leads from previous recommendations. After obtaining these leads, sales team members begin calling, emailing, and so on. At X education, the lead conversion rate is typically 30%.

Problem Statement:

A company called X-Education provides professional online education courses and internet marketing through adverts. The company uses a variety of ways to obtain information, and it calls leads who inquire about a particular degree of schooling. Lead conversion is usually 30% of specific education. The company also uses specific criteria to identify Hot Leads. The ratio of leads converted to enrolments is lower. company provided Aim for 80% of the total enrolled.

BUSINESS GOAL:

"Hot Leads" are leads that have the highest potential and are sought after by the company. The business requires a model in which each lead is given a score, with the goal being to increase the conversion chance of a client with a higher lead score and decrease it for a customer with a lower lead score. Specifically, the CEO provided an approximate figure of 80% for the lead conversion rate.

METHODOLOGY

DATA PREPARATION

- Read data from source
- Convert data into clean format suitable for analysis
- Outlier treatment
- Exploratory data analysis

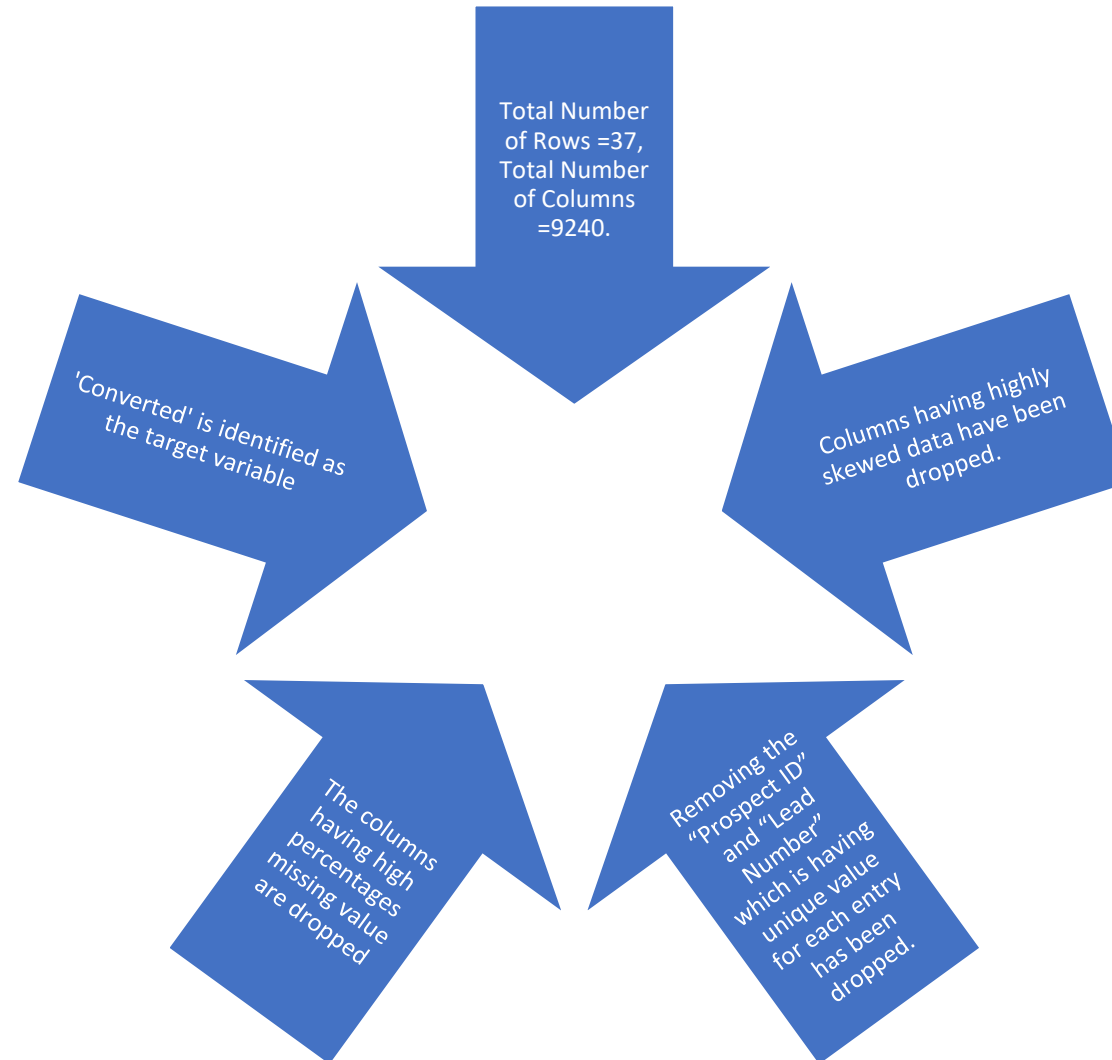
MODEL BUILDING

- Feature selection using RFE, VIF and p-value
- Determine optimal model using Logistic Regression
- Calculate various evaluation metrics

MODEL EVALUATION

- Determine Lead score and check if target final prediction is greater than 80% conversion rate
- Evaluate final prediction on test set

DATA MANIPULATION



EXPLORATORY DATA ANALYSIS – Unique Values

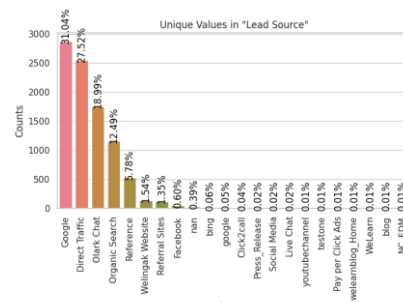


Figure 1

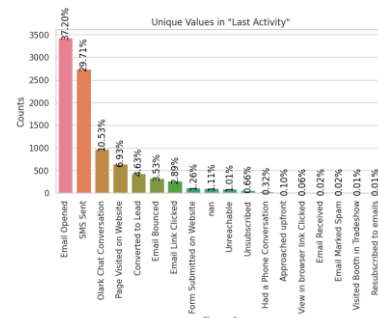
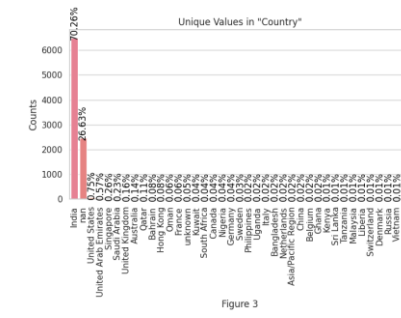
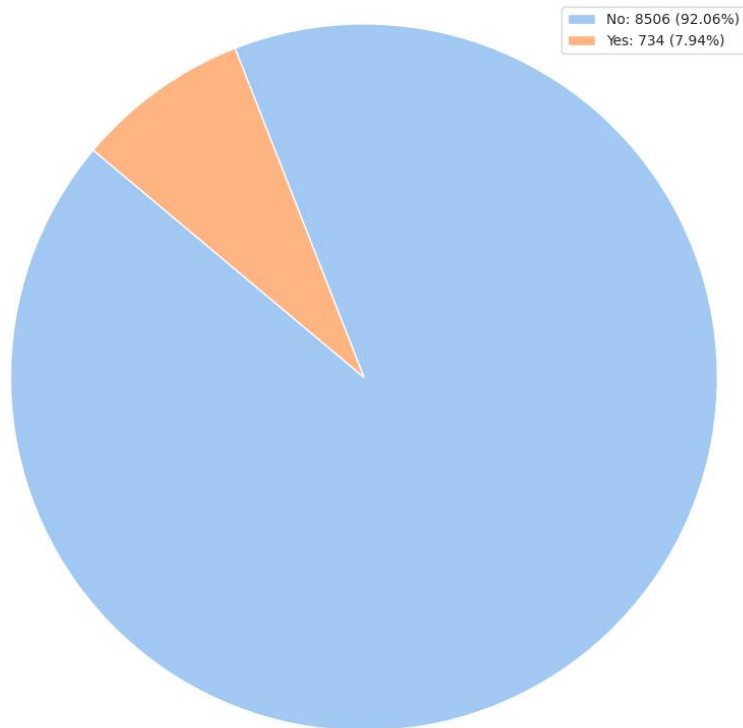


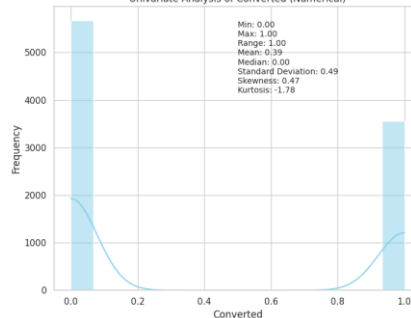
Figure 2



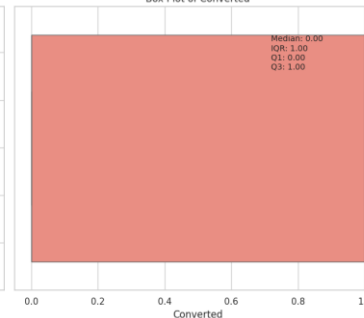
Univariate Analysis of Do Not Email (Categorical)



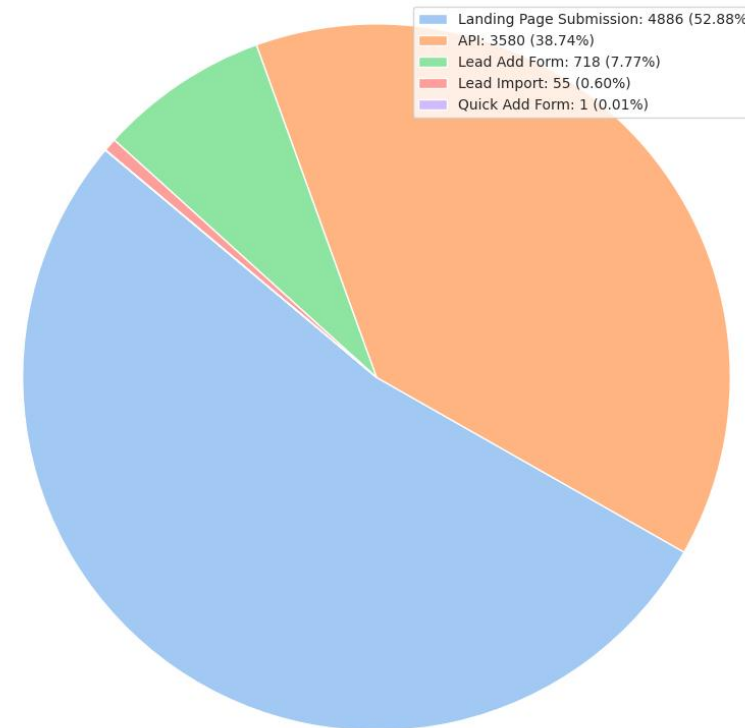
Univariate Analysis of Converted (Numerical)



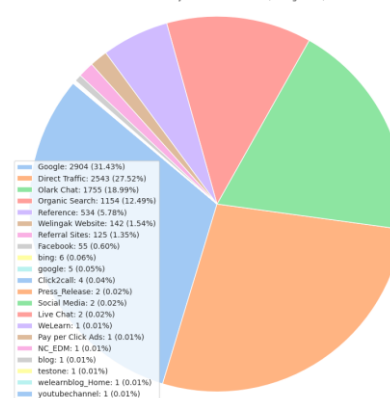
Box Plot of Converted



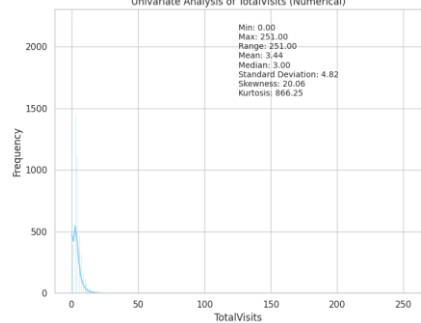
Univariate Analysis of Lead Origin (Categorical)



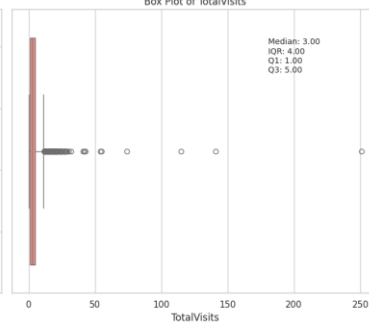
Univariate Analysis of Lead Source (Categorical)



Univariate Analysis of TotalVisits (Numerical)

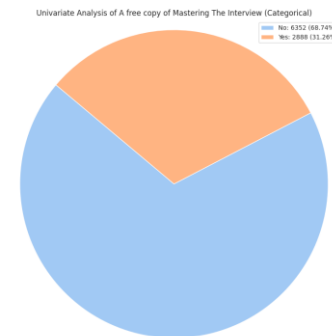
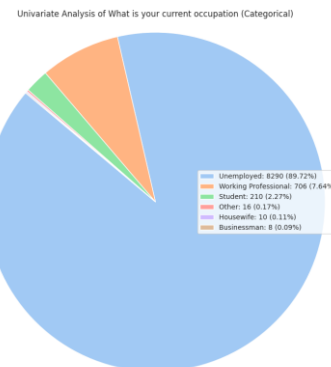
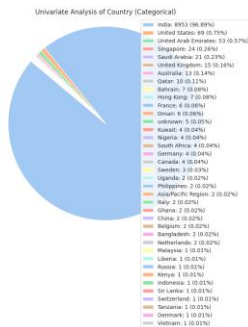
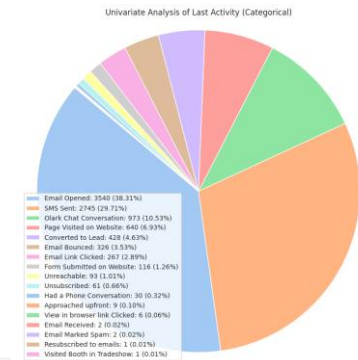
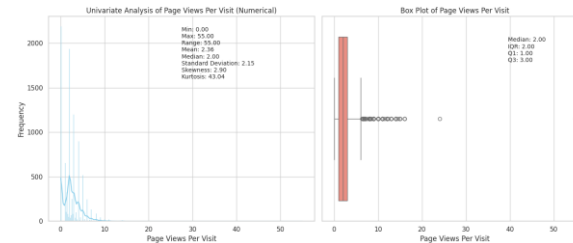
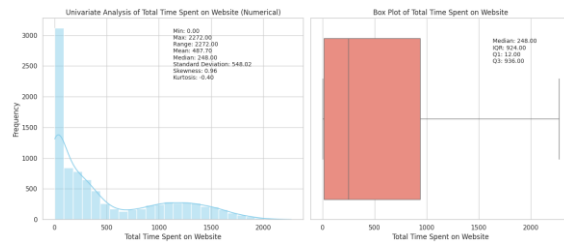


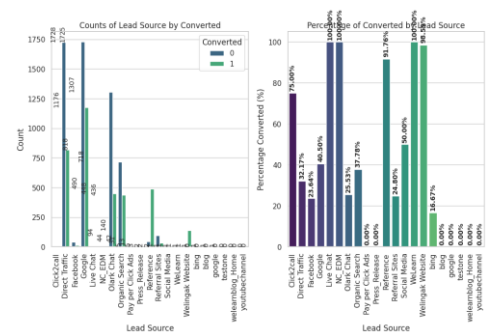
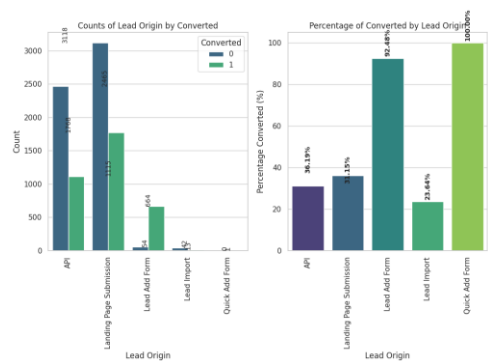
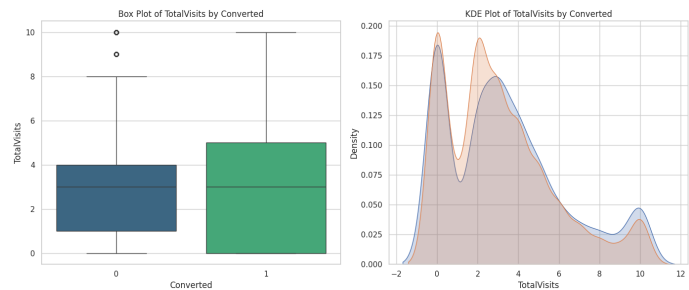
Box Plot of TotalVisits



EXPLORATORY DATA ANALYSIS - Univariate

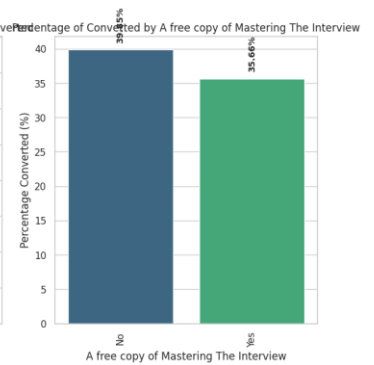
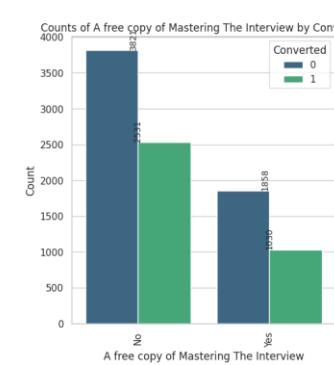
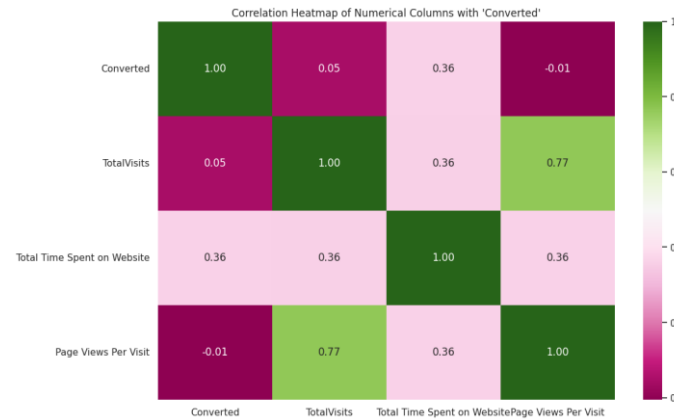
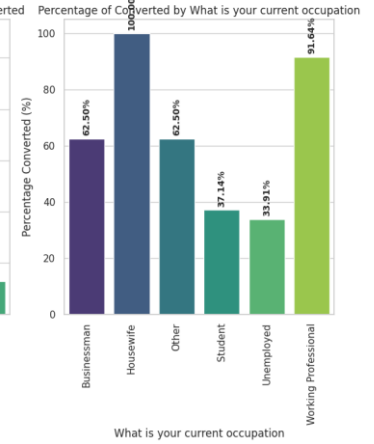
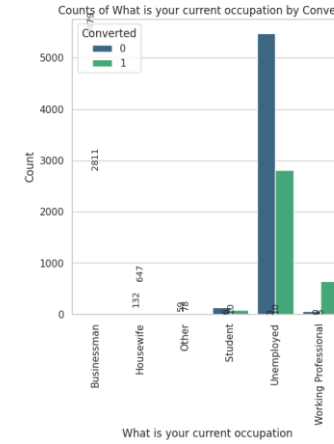
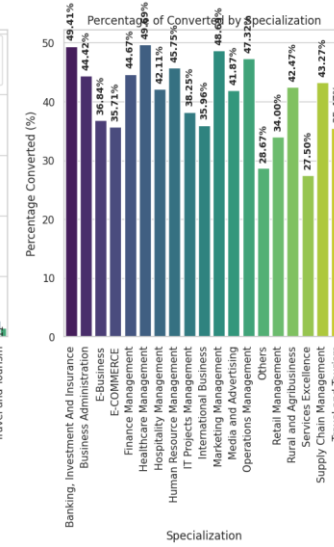
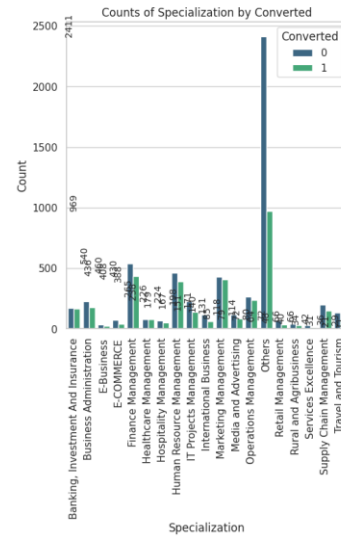
EXPLORATORY DATA ANALYSIS - Univariate



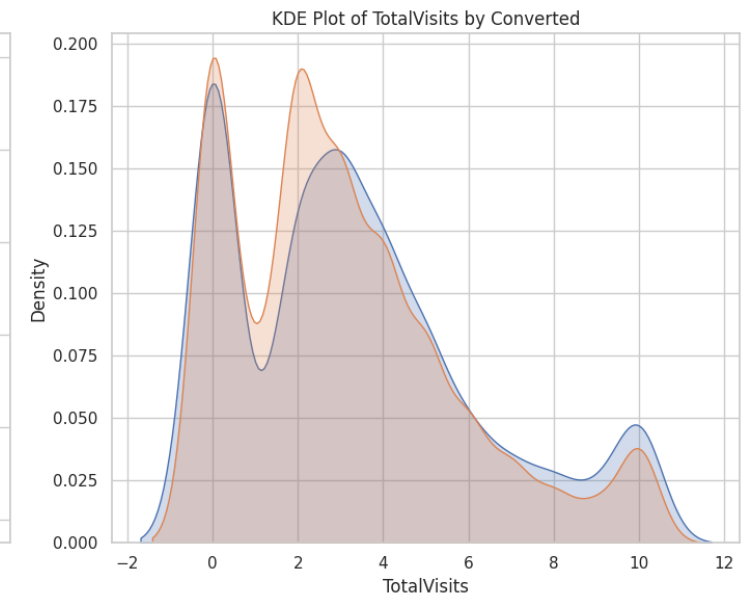
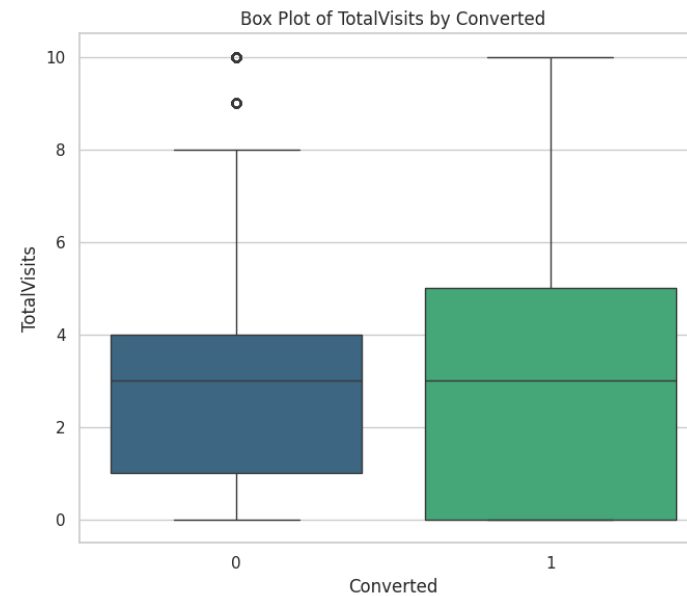
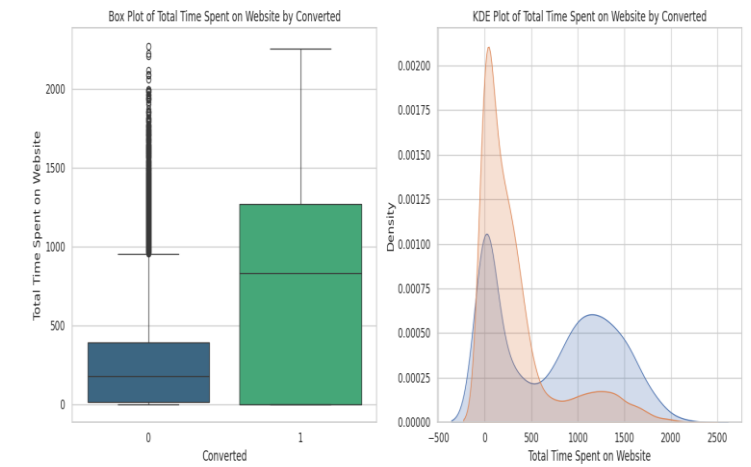
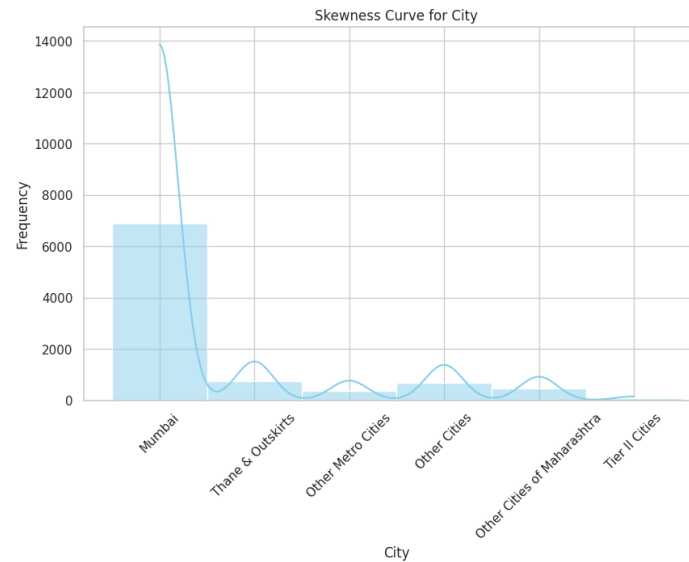


EXPLORATORY DATA ANALYSIS - Bivariate

EXPLORATORY DATA ANALYSIS - Bivariate



EXPLORATORY DATA ANALYSIS - Bivariate



MODEL BUILDING

- Split Data into train and test data with ratio 75:25
- Using RFE to choose variables to be accepted or rejected
- Build the right model by removing variables whose p-value > 0.05 and VIF > 5

Generalized Linear Model Regression Results

Dep. Variable:	Converted	No. Observations:	6930
Model:	GLM	Df Residuals:	6915
Model Family:	Binomial	Df Model:	14
Link Function:	Logit	Scale:	1.0000
Method:	IRLS	Log-Likelihood:	-2977.8
Date:	Mon, 20 May 2024	Deviance:	5955.7
Time:	17:26:59	Pearson chi2:	7.94e+03
No. Iterations:	7	Pseudo R-squ. (CS):	0.3761
Covariance Type:	nonrobust		

	coef	std err	z	P> z	[0.025	0.975]
const	-0.2601	0.085	-3.048	0.002	-0.427	-0.093
Lead Origin_Lead Import	-1.3717	0.492	-2.786	0.005	-2.337	-0.407
Lead Source_Direct Traffic	-1.3074	0.109	-12.004	0.000	-1.521	-1.094
Lead Source_Google	-0.9214	0.103	-8.965	0.000	-1.123	-0.720
Lead Source_Organic Search	-1.0832	0.126	-8.569	0.000	-1.331	-0.835
Lead Source_Reference	2.5451	0.201	12.641	0.000	2.150	2.940
Lead Source_Referral Sites	-1.1901	0.302	-3.938	0.000	-1.783	-0.598
Lead Source_Welingak Website	4.4690	0.725	6.164	0.000	3.048	5.890
Last Activity_Converted to Lead	-1.2082	0.205	-5.906	0.000	-1.609	-0.807
Last Activity_Olark Chat Conversation	-1.3996	0.156	-8.988	0.000	-1.705	-1.094
Last Activity_SMS Sent	1.2572	0.070	17.887	0.000	1.119	1.395
Last Activity_Unsubscribed	1.0743	0.445	2.413	0.016	0.202	1.947
What is your current occupation_Working Professional	2.8511	0.180	15.839	0.000	2.498	3.204
Do Not Email	-1.4495	0.162	-8.975	0.000	-1.766	-1.133
Total Time Spent on Website	1.0593	0.037	28.591	0.000	0.987	1.132

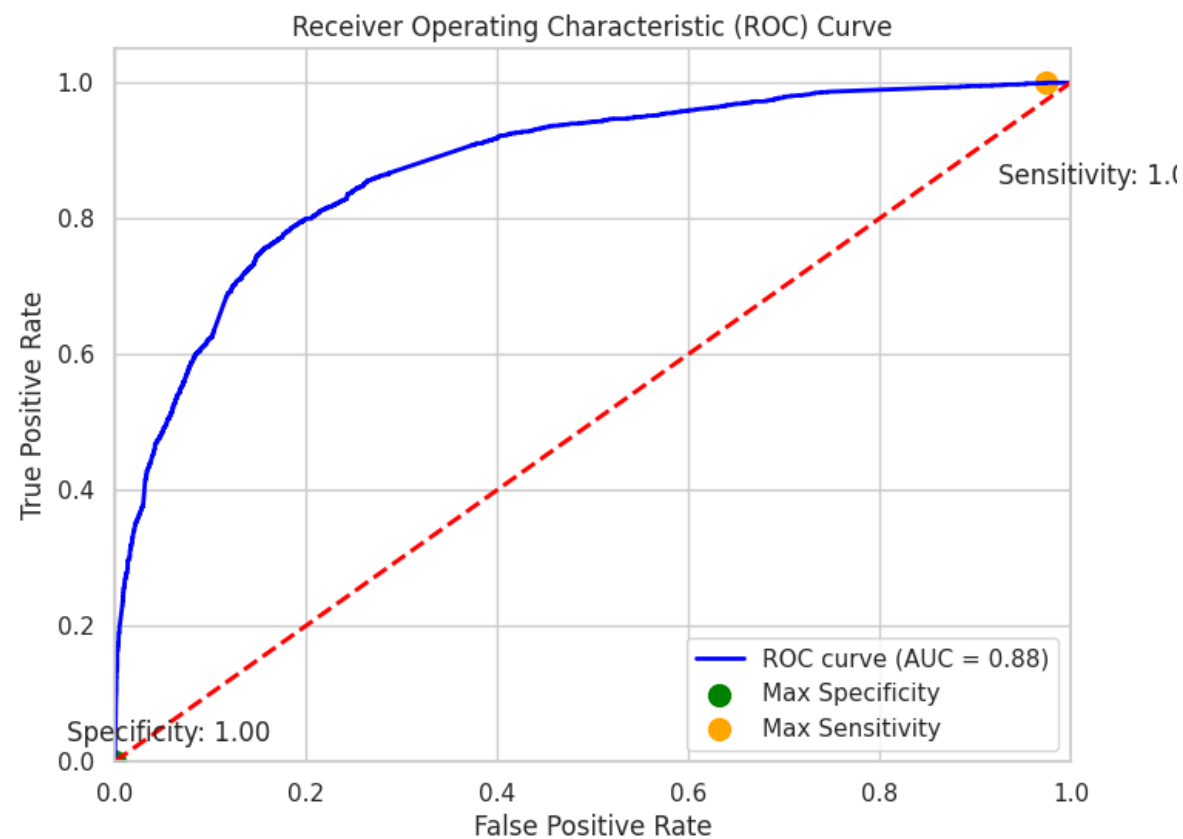
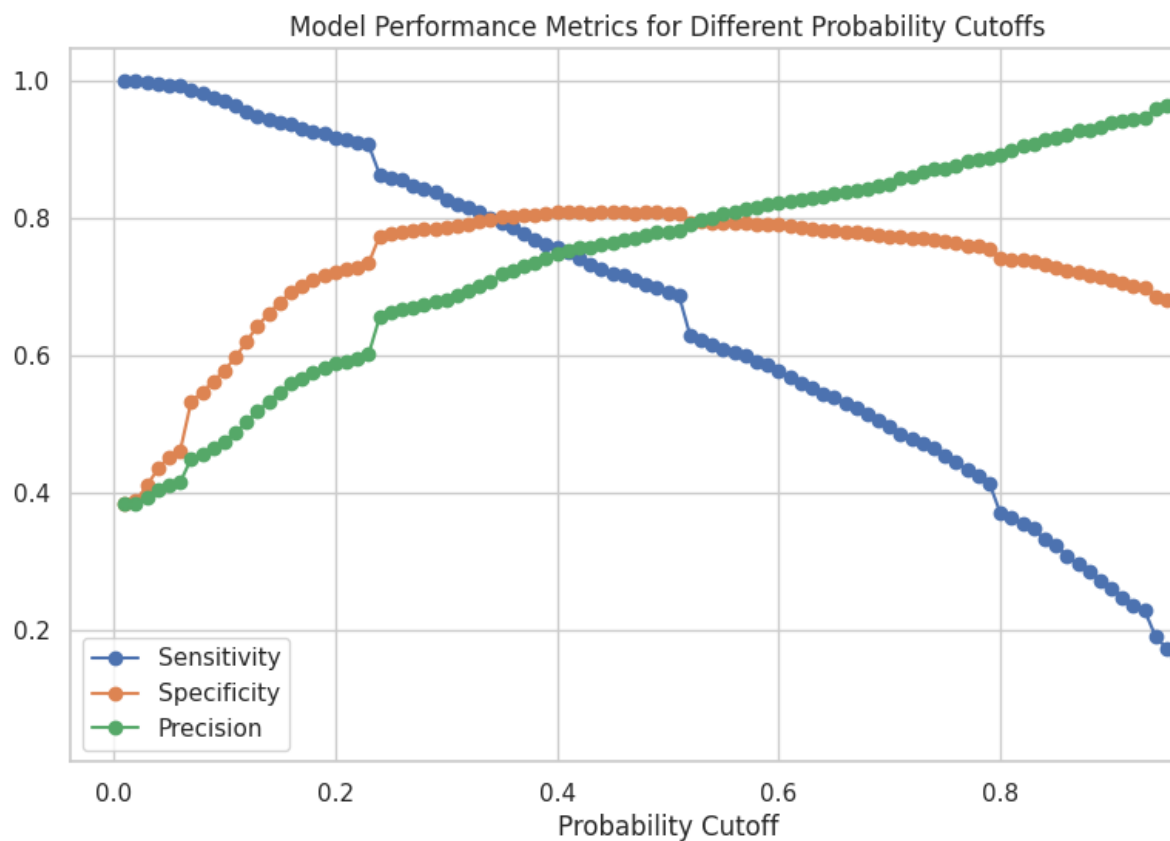
Top 5 Features with highest p-value:

Last Activity_Unsubscribed	1.582255e-02
Lead Origin_Lead Import	5.336486e-03
Lead Source_Referral Sites	8.225501e-05
Last Activity_Converted to Lead	3.495785e-09
Lead Source_Welingak Website	7.098194e-10

dtype: float64

OPTIMAL CUT OFF PROBABILITY

From graph, we are taking Overall Optimal Cutoff: 0.35



FINAL TRAINED MODEL

Evaluation metrics

Metric	Value
True Positive (TP)	714
True Negative (TN)	1144
False Positive (FP)	260
False Negative (FN)	192
Accuracy	0.804329
Sensitivity (Recall)	0.788079
Specificity	0.814815
Precision	0.73306

Main Features that determine the 'Conversion' are:

- Lead Source_Welingak Website
- What is your current occupation_Working Professional
- Lead Source_Reference

Feature Importance

Feature Importance	
const	-0.260093
Lead Origin_Lead Import	-1.371679
Lead Source_Direct Traffic	-1.307420
Lead Source_Google	-0.921376
Lead Source_Organic Search	-1.083223
Lead Source_Reference	2.545115
Lead Source_Referral Sites	-1.190135
Lead Source_Welingak Website	4.468967
Last Activity_Converted to Lead	-1.208157
Last Activity_Olark Chat Conversation	-1.399644
Last Activity_SMS Sent	1.257157
Last Activity_Unsubscribed	1.074337
What is your current occupation_Working Professional	2.851091
Do Not Email	-1.449473
Total Time Spent on Website	1.059300

CONCLUSION



The logistic regression model is used to predict the probability of conversion of a customer.



While we have calculated both sensitivity-specificity as well as Precision-Recall metrics, we have considered optimal cut off on the basis of sensitivity-specificity for final prediction



Lead Score calculated shows the conversion rate of final predicted model is around 80.4329 % in test data as compared to 80.1876% in train data, as both scores are similar, proving a good model



In Business terms, this model has capability to adjust with the company's requirements in coming future