REPORT ON LOGISTIC REGRESSION ASSIGNMENT – LEAD SCORING(Dated 16th Oct 2024)

Presented by

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Batch ID 5702

Business Objective: X Education has appointed you to help them select the most promising leads, i.e. the leads that are most likely to convert into paying customers. The company requires you to build a model wherein you need to assign a lead score to each of the leads such that the customers with a higher lead score have a higher conversion chance and the customers with a lower lead score have a lower conversion chance. The CEO, in particular, has given a ballpark of the target lead conversion rate to be around 80%.

Data Sources

- Leads. csv': contains all the information of the client on leads from the past
- Leads Data Dictionary.xlsx' is data dictionary which describes the meaning of the variables.

Business Problem Solving Methodology

<u>Problem solving approach methodology:</u> Data Science methodology, Machine Learning: Logistic Regression as this is a classification problem, Model evaluation, Insights & recommendations based on feature importance.

Steps followed are as under.

- 1. Importing the available data set
- 2. Data Understanding
- 3./ Data Pre-processing
- A. EDA
- Data Preparation
- 6. Machine Learning/Model building
- 7. Model evaluation
- 8. Final model
- 9. Feature importance
- 10. Recommendations

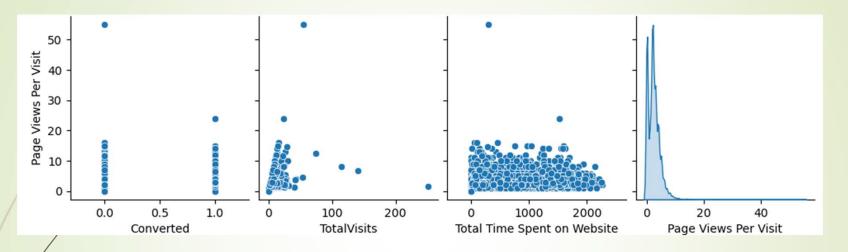
Data Understanding & Pre-processing

This dataset has 9,240 records & 37 features

There are 30 categorical features & 7 numerical features

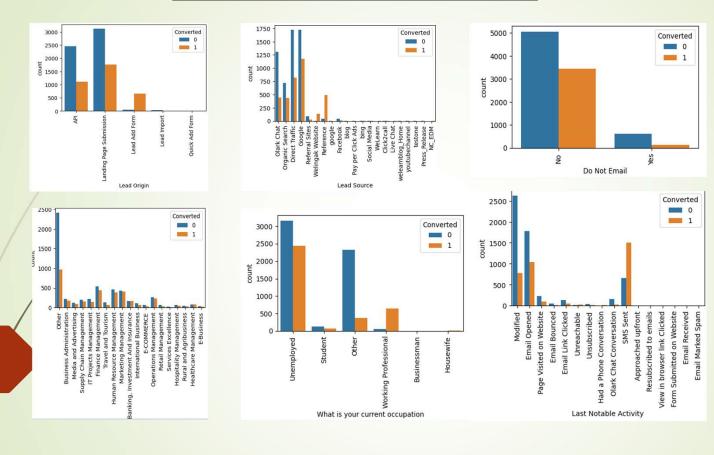
- Checked & removed duplicate records, but in this case there was no duplicate records found.
- There were few features which had 'select' as value which was made 'null' as applicable in this case.
- ➤ Null values handling: 7 features had > 40% null values which we dropped.10 more features had null values < 40% which we dropped few & imputed median & mode values for rest of them. We ended up having 25 features at this stage.
- Checked for class imbalance: Dropped 13 features which had highly imbalanced data.

EDA

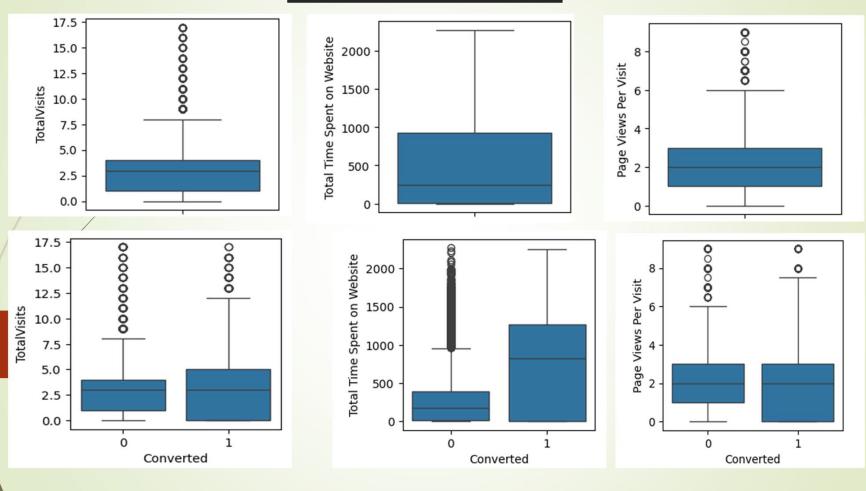


Features 'TotalVisits', 'Total Time Spent on Website', 'Page Views Per Visit' have no linear correlation with target feature 'Converted'

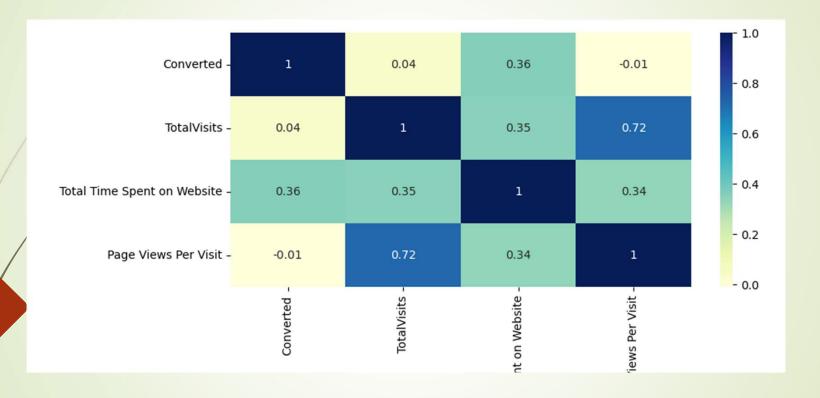
EDA – COUNTPLOTS



EDA - BOXPLOTS



EDA - HEATMAP



'Page Views Per Visit' & 'Total Visits' have a high correlation of 0.72

EDA – Insights

- Lead Origins like 'API' & 'Landing Page Submission' have significant impact on lead conversions.
- Lead Sources like 'Google', 'Direct Traffic', 'Reference', 'Olark Chat' & 'Organic Search' have significant impact on lead conversions.
- Those who have opted for email have significant impact on lead conversions.
- ➤ If the Last Activity is either 'SMS sent' or 'Email opened', then they have significant impact on lead conversions.
- Those who have mentioned specialization as 'Others' have significant impact on lead conversions.
- > Those who are 'Unemployed' have significant impact on lead conversions.
- Those who have not opted for 'A free copy of Mastering The Interview' have significant impact on lead conversions
- Those who have visited the website atleast 5 times have significant impact on lead conversion rate.
- Those who have spent total time on website of around 1200 units have significant impact on lead conversion rate.
- Those who have viewed Average number of pages on the website of atleast 3 have significant impact on lead conversion rate

Outliers & its handling

Total Visits' & 'Page Views Per Visit' have outliers on the upper range & required Outlier Treatment. From describe function previously, it was observed that 75th percentile value of 'Total Visits' is 5 & for 'Page Views Per Visit' is 3. Hence capped maximum values to 99th percentile.

Data Preparation

- Converted 2 features binary values (Yes/No) to 1/0
- Converted 5 features 'Lead Origin', 'Lead Source', 'Last Activity', 'Specialization' & 'What is your current occupation' into numerical features using dummies
- We had totally all 11 as numerical features at this stage which was good enough for model building.
- Feature Scaling: on 3 features using Standard Scaler from sklearn
- We had almost 38% lead converted rate.
- 'Do Not Email' & 'Last Activity_email bounced' have high correlation of 0.63.
- Lead Origin_Lead Add Form' and 'Lead Source_Reference' having higher correlation of 0.85.
- 'Lead Source_Facebook' and 'Lead Origin_Lead Import' having higher correlation of 0.98
- 'TotalVisits' and 'Page Views Per Visit' having correlation of 0.72

Machine Learning/Model Building

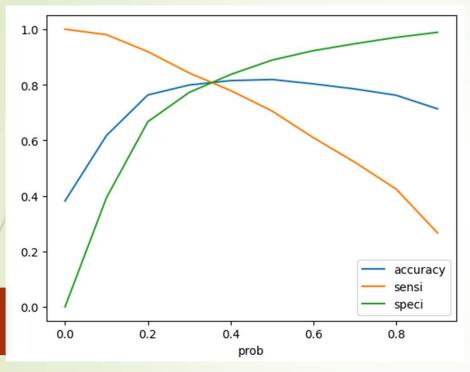
- Used Logistic Regression from sklearn
- Feature selection using RFE(selected 15 important features)
- Used statsmodels to add constant & fit the model
- Used summary()/mainly p-values & VIF scores to tune the model
- Final model(Third model) had 13 features, all significant p-values & all VIF scores below 5.
- ➤ Using final model, predicted train dataset values.
- Found out Optimal cut-off point using accuracy, sensitivity & specificity scores as 0.35
- Tuned predicted values based on 0.35 threshold & added lead scores based on the same.
- The cut-off point of Precision Recall curve was around 0.45
- Plotted ROC curve & had AUC score of 0.89

Machine Learning/Model building(Contd..)

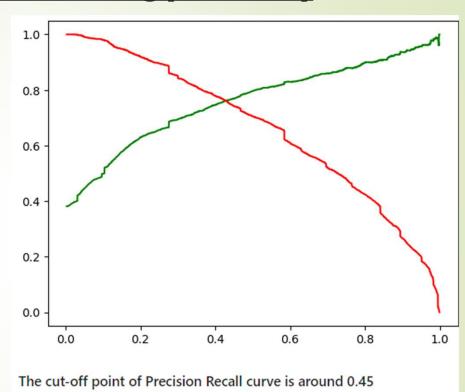
		Gene	ralized Linear Mode	l Regression Results				
Dep. Variable: Model: Model Family: Link Function: Method: Date:			Converted		6363			
			GLM	Df Residuals:		6349		
			Binomial	Df Model:		13		
			Logit	Scale:	1	1.0000		
			22911					
			IRLS Log-Likelihood:					
			Wed, 16 Oct 2024	5	129.0			
	T	ime:	23:32:36	Pearson chi2:	6.79	e+03		
No. Iterations:			7	Pseudo R-squ. (CS):	0.	.4076		
Cova	riance 1	уре:	nonrobust					
/				Feat	ures	VIF		
	10			Specialization_C	ther	2.27		
	4			Lead Source_Olark	Chat	2.06		
	2	Lead Origin_Landing Page Submission						
	11	What is your current occupation_Other						
	8			Last Activity_SMS	Sent	1.54		
	3		Lea	ad Origin_Lead Add I	orm	1.52		
	7		Last Activity	_Olark Chat Converse	ation	1.49		
	5		Lead	Source_Welingak We	bsite	1.32		
	1		Total	al Time Spent on We	bsite	1.25		
	0			Do Not E	mail	1.21		
	12 V	vhat is	s your current occ	upation_Working Pro	fes	1.20		
	9		L	ast Activity_Unsubscr	ibed	1.10		
	6		Last A	ctivity_Converted to	Lead	1.09		

	coef	std err	z	P> z	[0.025	0.975]
const	-0.2785	0.127	-2.195	0.028	-0.527	-0.030
Do Not Email	-1.6051	0.179	-8.962	0.000	-1.956	-1.254
Total Time Spent on Website	1.0675	0.040	26.445	0.000	0.988	1.147
Lead Origin_Landing Page Submission	-0.8760	0.130	-6.753	0.000	-1.130	-0.622
Lead Origin_Lead Add Form	2.8909	0.211	13.726	0.000	2.478	3.304
Lead Source_Olark Chat	1.0952	0.125	8.787	0.000	0.851	1.339
Lead Source_Welingak Website	3.1667	1.029	3.078	0.002	1.150	5.183
Last Activity_Converted to Lead	-1.1395	0.210	-5.422	0.000	-1.551	-0.728
Last Activity_Olark Chat Conversation	-1.2742	0.167	-7.635	0.000	-1.601	-0.947
Last Activity_SMS Sent	1.3083	0.076	17.197	0.000	1.159	1.457
Last Activity_Unsubscribed	1.3908	0.452	3.080	0.002	0.506	2.276
Specialization_Other	-0.8366	0.124	-6.748	0.000	-1.080	-0.594
What is your current occupation_Other	-1.1914	0.089	-13.423	0.000	-1.365	-1.017
hat is your current occupation_Working Professional	2.3946	0.189	12.642	0.000	2.023	2.766

Machine Learning/Model building(Contd..)



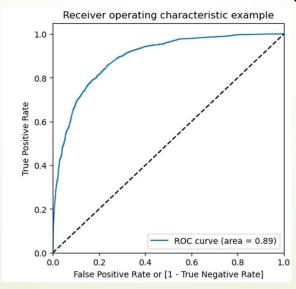
Optimal cut-off point at 0.35



Machine Learning/Model building(Contd..)

	Converted	Converted_Prob	Prospect ID	0.0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	8.0	0.9	final_predicted	Lead_Score
0	0	0.059293	302	1	0	0	0	0	0	0	0	0	0	0	6
1	0	0.022673	6087	1	0	0	0	0	0	0	0	0	0	0	2
2	0	0.241976	1033	1	1	1	0	0	0	0	0	0	0	0	24
3	0	0.153273	7656	1	1	0	0	0	0	0	0	0	0	0	15
4	1	0.752513	3241	1	1	1	1	1	1	1	1	0	0	1	75

Assigning lead score to the leads based on optimal cut off of 0.35



Model Evaluation & Metrics

- Model was evaluated using Precision Recall cut-off of 0.45 which reduced both sensitivity & Recall scores. This was not ideal for the nature of business problem had on hand & hence continued based on 0.35 threshold only.
- Made predictions on test set

Train data evaluation metrics

Accuracy: 81%
Sensitivity: 81%
Specificity: 80%
Precision: 72%

Recall: 81%

Test data evaluation metrics

Accuracy: 80%
Sensitivity: 81%
Specificity: 80%
Precision: 72%

Recall: 81%

Final Model

-0.2785 - 1.6051xDo Not Email + 1.0675xTotal Time Spent on Website - 0.8760xLead Origin_Landing Page Submission + 2.8909xLead Origin_Lead Add Form + 1.0952xLead Source_Olark Chat + 3.1667xLead Source_Welingak Website - 1.1395xLast Activity_Converted to Lead - 1.2742xLast Activity_Olark Chat Conversation + 1.3083xLast Activity_SMS Sent + 1.3908xLast Activity_Unsubscribed - 0.8366xSpecialization_Other - 1.1914xWhat is your current occupation_Other + 2.3946xWhat is your current occupation_Working Professional

Feature Importance

Features importance with respective percentages are as under

	Lead Source_Welingak Website	100.000000
	Lead Origin_Lead Add Form	91.291311
	What is your current occupation_Working Professiona	al 75.617920
	Last Activity_Unsubscribed	43.917779
	Last Activity_SMS Sent	41.314677
	Lead Source_Olark Chat	34.584874
	Total Time Spent on Website	33.709718
/	Specialization_Other	-26.417332
	Lead Origin_Landing Page Submission	-27.661534
	Last Activity_Converted to Lead	-35.984065
	What is your current occupation_Other	-37.624009
	Last Activity_Olark Chat Conversation	-40.236284
	Do Not Email	-50.687170

Recommendations

Top 3 features contributing to lead conversions are

- 1. Lead Source_Welingak Website
- 2. Lead Origin_Lead Add Form
- 3. What is your current occupation_Working Professional

Top 3 features to be focussed more for having significant leads conversions are

- 1. Do Not Email
- 2. Last Activity_Olark Chat Conversation
- 3. What is your current occupation_Other

Recommendations(Contd..)

- ➤ Lead Origins like 'API' & 'Landing Page Submission' have significant impact on lead conversions.
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Recommendations(Contd..)

Few areas they can look into to create more potential leads are as under

- a. E-mail reminder campaigns
- b. Improvise Marketing strategies around preferred lead sources like 'Google', 'Direct Traffic', 'Reference', 'Olark Chat' & 'Organic Search'.
- c. Referral programs with eye-catching incentives.
- d. Target customers whose occupation mentioned as 'others' (to understand more about them & filter out who among these might be really interested in persuing the learning/course)

