# LEAD SCORING ASSIGNMENT

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#### PROBLEM STATEMENT

We are given with three data sets-

- 1. 'Leads.csv' contains all the information of the leads from the past with around 9000 datasets.
- 2. 'Leads Data Dictionary.xlsx' is data dictionary which describes the meaning of the variables.

Primary goal is to help X Education select most promising leads (Hot Leads), i.e. the leads that are most likely to convert into paying customers.

#### **APPROACH**

Reading and understanding the Data

• Importing and understanding the past Data provided by the Company

**Data Cleaning** 

- Missing value imputation
- Removing duplicate data and other redundancies

Exploratory Data Analysis

- Univariate Analysis
- Bivariate Analysis

Data Preparation

- Outlier treatment
- Dropping unnecessary columns
- Dummy variable creation and feature standardization

# Model Building

- Feature selection using RFE
- Manual feature elimination based on p-value and VIFs

# Model Evaluation

- Evaluating Model based on various evaluation metrics
- Finding the optimal probability thresold

# Assigning Lead Scores

- Finalizing the first model
- Using predicted probabilities to calculate Lead Scores
- Lead score=Probability\*100

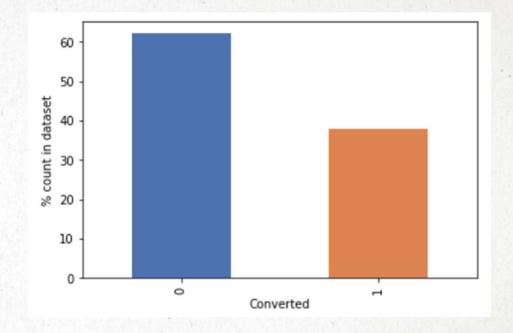
# **EXPLORATORY DATA ANALYSIS**

- For best feature selection
- To check the distribution of the data

## **UNIVARIATE ANALYSIS**

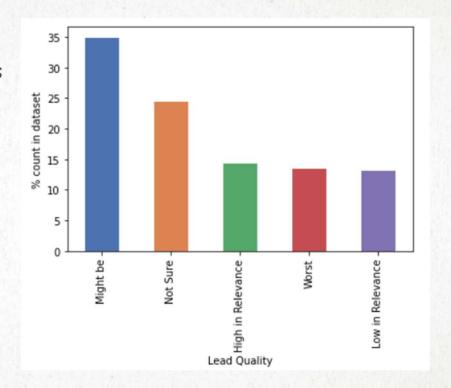
#### **DISTRIBUTION OF LEADS**

- Approximately 38% of the leads are converted.
- Approximately 62% of the leads are not converted.

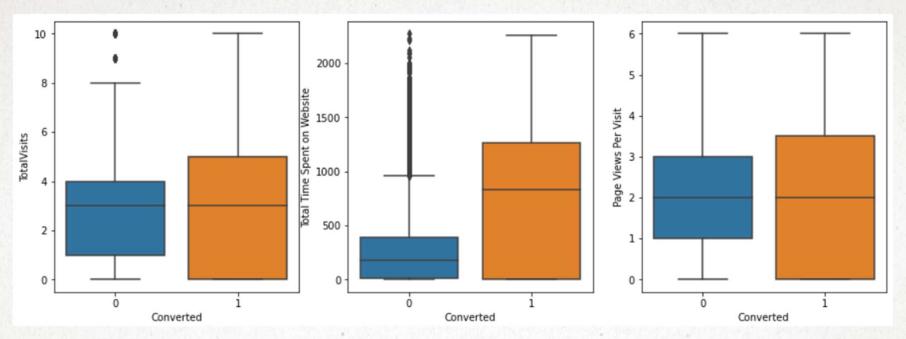


## **DISTRIBUTION OF LEAD QUALITY**

- Approximately 35% of the Leads might be the potential buyers
- Approximately 25% of the leads are not sure followed by High Relevance.

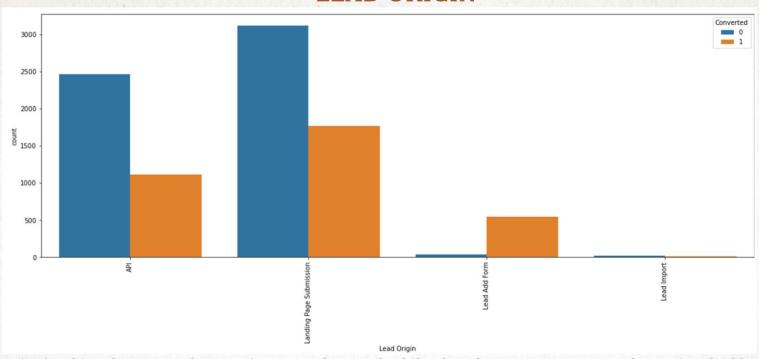


#### **NUMERICAL VARIABLES**



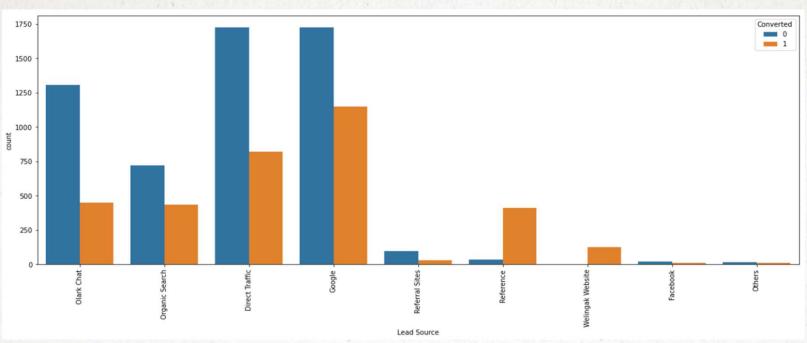
• People spending more time on website are more likely to get converted.

#### **LEAD ORIGIN**



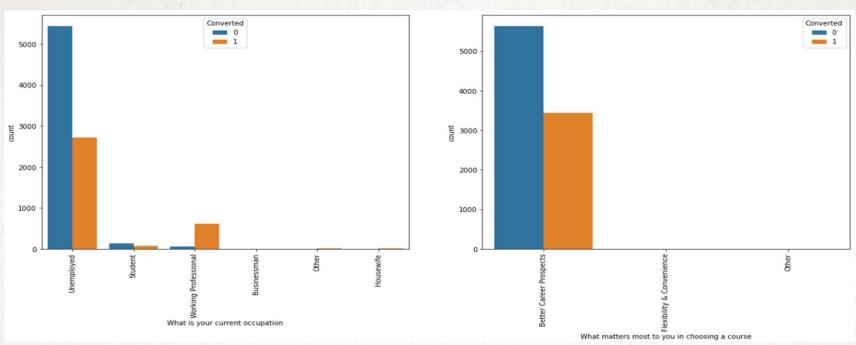
- 'API' and 'Landing Page Submission' generate the most leads but have less conversion rates, whereas 'Lead Add Form' generates less leads but conversion rate is great.
- Try to increase conversion rate for 'API' and 'Landing Page Submission', and increase leads generation using 'Lead Add Form'.

#### **LEAD SOURCE**



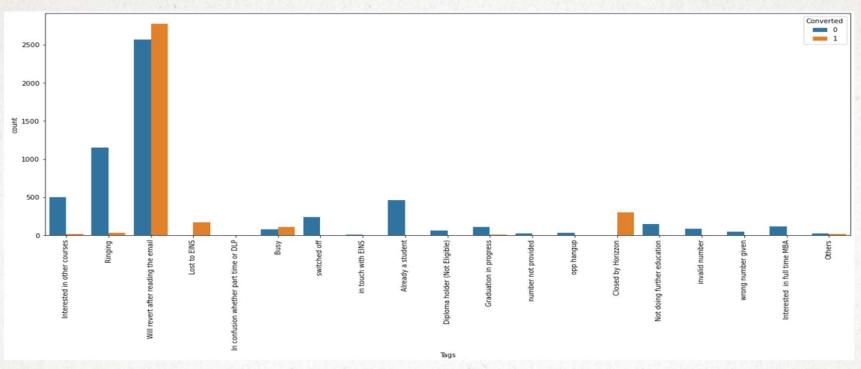
- Very high conversion rates for lead sources 'Reference' and 'Welingak Website'.
- Most leads are generated through 'Direct Traffic' and 'Google'...

## **CURRENT OCCUPATION**



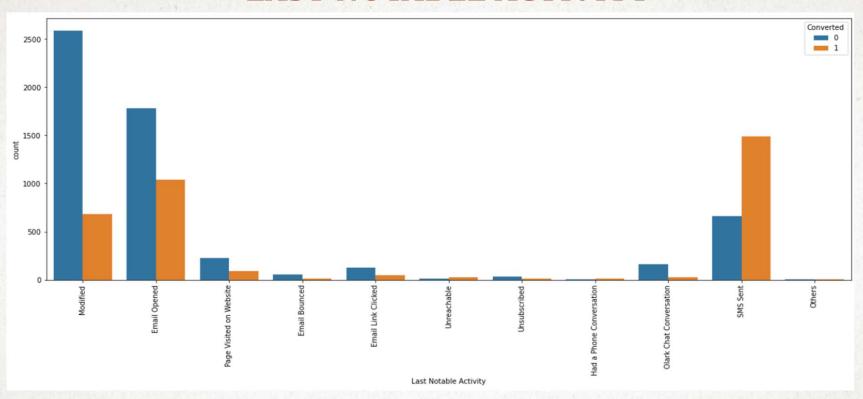
Working Professionals are most likely to get converted.

## **TAGS**



 High conversion rates for tags 'Will revert after reading the email', 'Closed by Horizon', 'Lost to EINS', and 'Busy'.

### LAST NOTABLE ACTIVITY



• Highest conversion rate is for the last notable activity 'SMS Sent'.

# **MODEL EVALUATION**

### Final Model Summary with all p values less than 5%

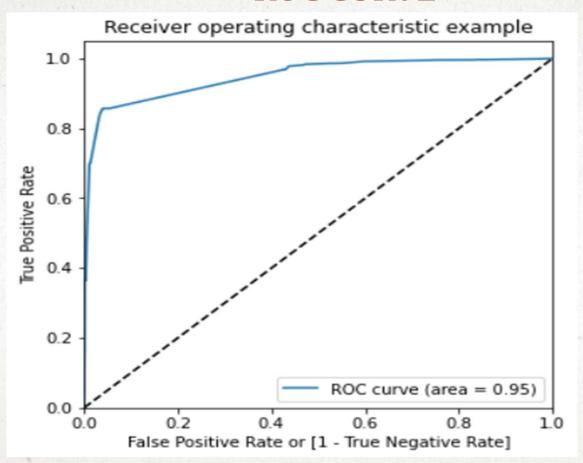
Generalized Linear Model Regression Results								
Dep. Variable:	Converted	No. Observati	ons:	63	== 351			
Model:	GLM	Df Residuals:		63	38			
Model Family:	Binomial	Df Model:			12			
Link Function:	Logit	Scale:		1.00	000			
Method:	IRLS	Log-Likelihood:		-1601.0				
Date:	Tue, 15 Nov 2022	Deviance:		3202	2.0			
Time:	05:09:00	Pearson chi2:		3.48e+	-04			
No. Iterations:	8	Pseudo R-squ.	(CS):	0.56	35			
Covariance Type:	nonrobust							
===========		coef	std err	Z	P> z	[0.025	0.975]	
const		-1.9192	0.211	-9.080	0.000	-2.333	-1.505	
Do Not Email		-1.2835	0.212	-6.062	0.000	-1.698	-0.868	
Lead Origin_Lead Add Form		1.2035	0.368	3.267	0.001	0.482	1.925	
Lead Source_Welingak Website		3.2825	0.820	4.002	0.000	1.675	4.890	
Tags_Busy		3.8043	0.330	11.525	0.000	3.157	4.451	
Tags_Closed by Horizzon		7.9789	0.762	10.467	0.000	6.485	9.473	
Tags_Lost to EINS		9.1948	0.753	12.209	0.000	7.719	10.671	
Tags_Ringing		-1.8121	0.336	-5.401	0.000	-2.470	-1.154	
Tags_Will revert after reading the email		3.9906	0.228	17.508	0.000	3.544	4.437	
Tags_switched off		-2.4456	0.586	-4.171	0.000	-3.595	-1.297	
Lead Quality_Not Sure		-3.5218	0.126	-28.036	0.000	-3.768	-3.276	
Lead Quality_Worst		-3.9106	0.856	-4.567	0.000	-5.589	-2.232	
Last Notable Activity_SMS Sent		2.7395	0.120	22.907	0.000	2.505	2.974	

#### **HEATMAP**

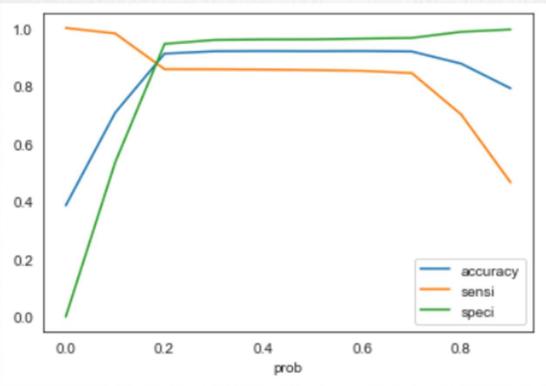


• Correlations between features in the final model are negligible.

## **ROC CURVE**

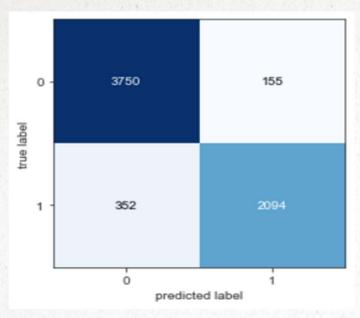


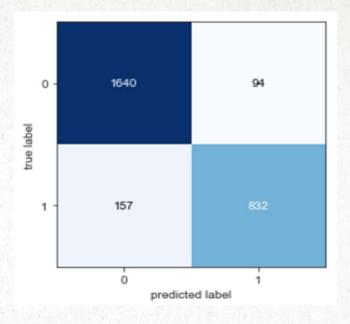
### **OPTIMAL THRESOLD**



 Graph showing changes in Sensitivity, Specificity and Accuracy with changes in the probability. Threshold values Optimal cutoff = 0.20

## **CONFUSION MATRIX**





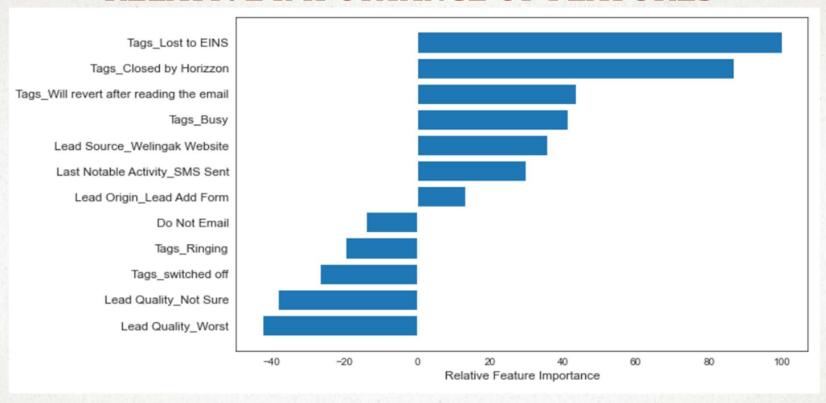
For Train Data

For Test Data

# **FINAL RESULTS**

Data	Train Set	Test Set
Accuracy	0.91	0.907
Sensitivity	0.86	0.84
Specificity	0.94	0.94
False Positive Rate	0.05	0.054
Positive Predictive Value	0.90	0.898
Negative Predictive Value	0.91	0.91
AUC	0.95	0.938

#### RELATIVE IMPORTANCE OF FEATURES



# INFERENCES

#### **IMPORTANT FEATURES**

- Three variables which contribute most towards the probability of a lead conversion in decreasing order of impact are:
  - 1. Tags\_Lost to EINS
  - 2. Tags\_Closed by Horizzon
  - 3. Tags\_Will revert after reading the email
- These are dummy features created from the categorical variable Tags.
- All three contribute positively towards the probability of a lead conversion.
- These results indicate that the company should focus more on the leads with these three tags.

**Situation 1:** Company has interns for 2 months. They wish to make lead conversion more aggressive. They want almost all of the potential leads to be converted and hence, want to make phone calls to as much of such people as possible.

#### Solution:

- > Sensitivity = True Positives / (True Positives + False Negatives)
- ➤ Sensitivity can be defined as the number of actual conversions predicted correctly out of total number of actual conversions. As we saw earlier, sensitivity decreases as the threshold increases.
- ➤ High sensitivity implies that our model will correctly predict almost all leads who are likely to convert. At the same time, it may overestimate and misclassify some of the non-conversions as conversions.
- As the company has extra man-power for two months and wants to make the lead conversion more aggressive, it is a good strategy to go for high sensitivity. To achieve high sensitivity, we need to choose a low threshold value.

**Situation 2:** At times, the company reaches its target for a quarter before the deadline. It wants the sales team to focus on some new work. So during this time, the company's aim is to not make phone calls unless it's extremely necessary. **Solution:** 

- > Specificity = True Negatives/ (True Negatives + False Positives) > Specificity can be defined as the number of actual non-conversions predicted correctly out of total number of actual non-conversions. It increases as the threshold increases.
- ➤ High specificity implies that our model will correctly predict almost all leads who are not likely to convert. At the same time, it may misclassify some of the conversions as non-conversions.
- ➤ As the company has already reached its target for a quarter and doesn't want to make unnecessary phone calls, it is a good strategy to go for high specificity.
- ➤ It will ensure that the phone calls are only made to customers who have a very high probability of conversion. To achieve high specificity, we need to choose a high threshold value.

# RECOMMENDATIONS

- By referring to the data visualizations, focus on
  - Increasing the conversion rates for the categories generating more leads and
  - Generating more leads for categories having high conversion rates.
- Pay attention to the relative importance of the features in the model and their positive or negative impact on the probability of conversion.
- Based on varying business needs, modify the probability threshold value for identifying potential leads.

