## X Education - Lead Scoring Case Study

#### **Group Members:**

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## Background

X Education Company

- X Education , An education company named sells online courses to industry professionals
- Many interested professionals land on their website
- The company markets its courses on several websites like Google. Once these people land on the website, they might browse the courses or fill up a form for the course or watch some videos

## Background

X Education Company

- When these people fill up a form providing their email address or phone number, they are classified to be a lead
- Once these leads are acquired, employees from the sales team start making calls, writing emails, etc. Through this process, some of the leads get converted while most do not

## Problem Statement

X Education Company's

Problem

- X Education gets a lot of leads but its lead conversion rate is very poor
- To make this process more efficient, the company wishes to identify the most potential leads, also known as 'Hot Leads'
- If they successfully identify this set of leads, the lead conversion rate should go up as the sales team will now be focusing more on communicating with the potential leads rather than making calls to everyone

## Problem Statement

X Education Company's Problem

- We will help them to select the most promising leads, i.e. the leads that are most likely to convert into paying customers.
- We are required to build a model wherein we need to assign a lead score to each of the leads such that the customers with higher lead score have a higher conversion chance
- The CEO, in particular, has given a ballpark of the target lead conversion rate to be 80%.

#### Lead - Conversion Process

Lead to Conversion process

Lead Generation:

1. Ads on websites like Google

2. Referrals

Visit to X
Education
website by these
potential
customers
(professionals)

Visitors either provide Email id & Contact Details
Or
View videos etc

Tele calling and Emailing activity to all the leads

~30% leads get converted

Proposed Solution:
A model to filter leads
so that leads to
conversion ratio is
80%+

### Proposed Solution

Selection of Hot Leads

Communicating with Hot Leads

Conversion of Hot Leads

#### **Leads Clustering**

We cluster the leads into certain categories based on their tendency or probability to convert, thus, getting a smaller section of hot leads to focus more on.

#### **Focus Communication**

Since we would have a smaller set of leads to have communication with, we might make more impact with effective communication.

#### Increase conversion

Since we focussed on hot leads, which were more probable to convert, we would have a better conversion rate, and hence we can achieve the 80% target.

### Solution

Selection of Hot Leads

For our Problem Solution, the crucial part is to accurately identify hot leads.

The more accurate we obtain the hot lead, the more chance we get of higher conversion ratio.

Since we have a target of 80% conversion rate, we would want to obtain a high accuracy in obtaining hot leads.

## Implementation

Loading &
Observing the past
data provided by
the Company

Univariate, Bivariate, and Heatmap for numerical and categorical columns

Performing prerequisites for RFE and Logistic Regression

Data
Sathering
Data
Performing
EDA
Preparation
Model
Building

Duplicate removal, null value treatment, unnecessary column elimination, etc.

Outlier Treatment, Feature-Standardization Selection of top 25 features using RFE

Reduction of columns and Model re-building

Verifying our Final Model Accuracy etc. with model built with PCA

Feature Model Model
Selection Building Improvement

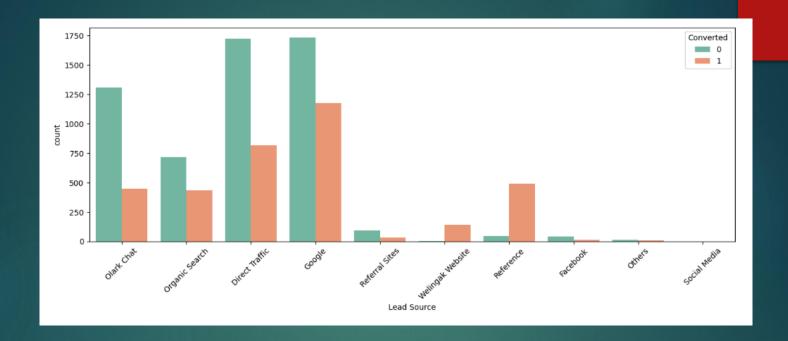
Final Model

Verifying with PCA

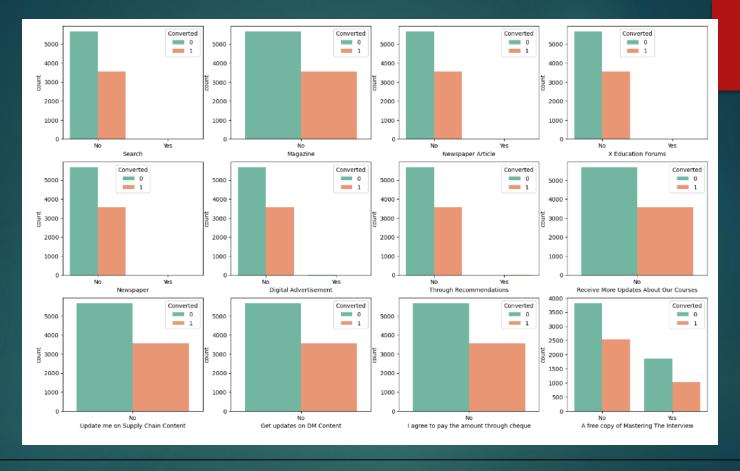
Model building using RFE for selected columns

Final Model Analysis and performance on Test Data

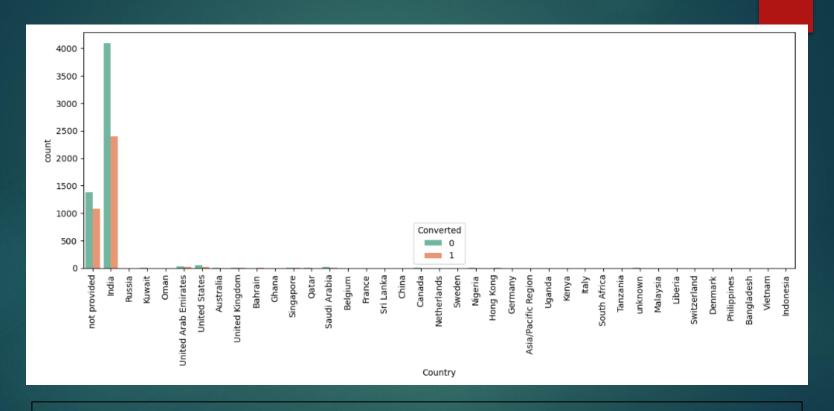
## Plots (Visualization)



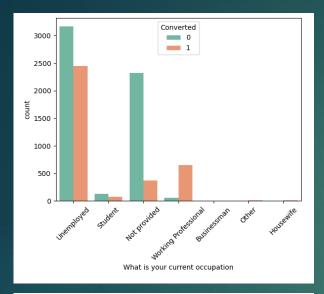
EDA plots depicting variation in categorical column for those who Converted and those who didn't.

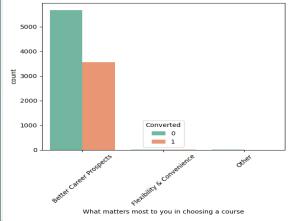


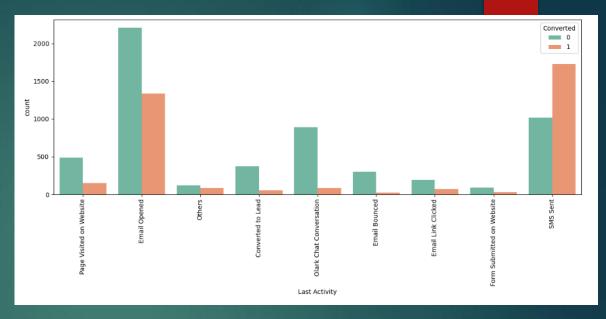
EDA plots depicting variation in categorical column for those who Converted and those who didn't.



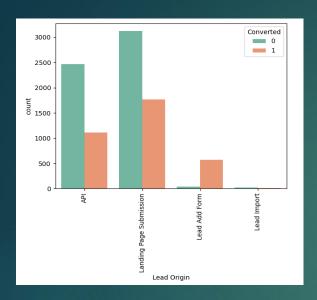
EDA plots depicting variation in categorical column for those who Converted and those who didn't.

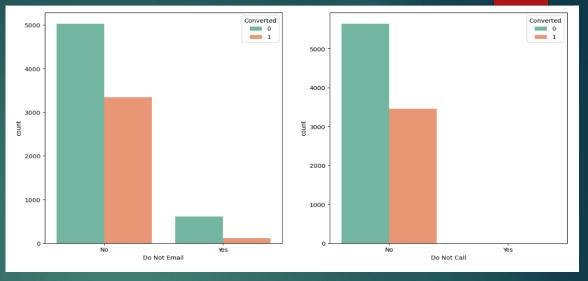


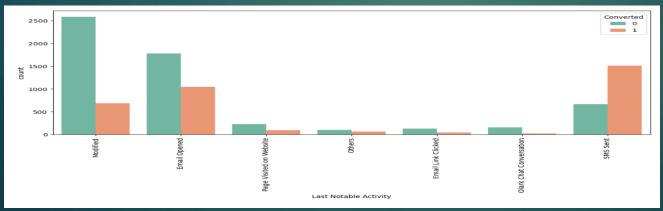




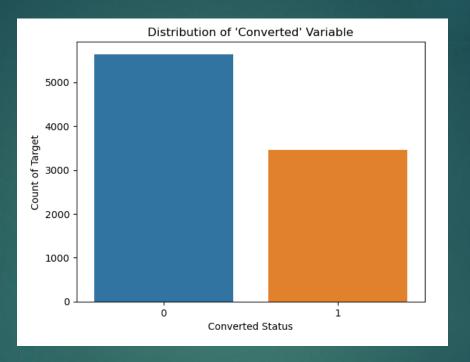
EDA plots depicting variation in categorical column for those who Converted and those who didn't.



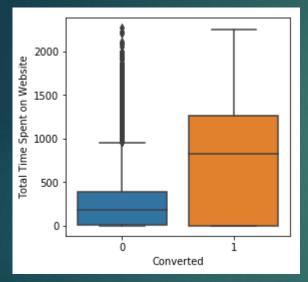


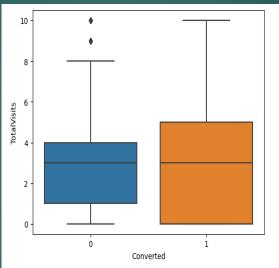


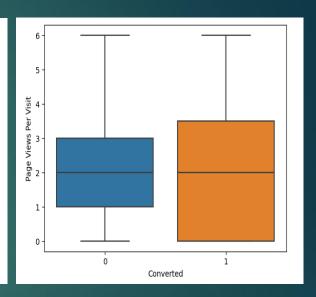
variation in categorical column for those who Converted and those who didn't.



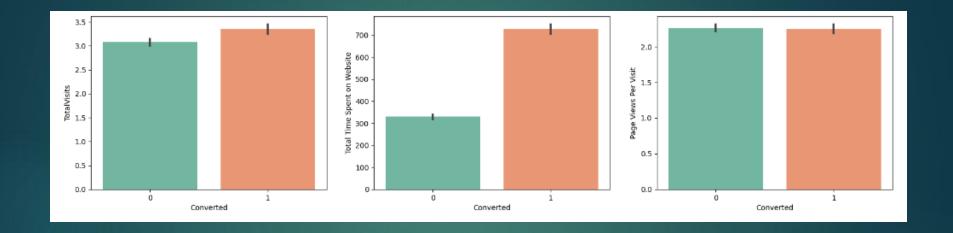
EDA plots depicting variation in numerical columns for those who Converted and those who didn't.



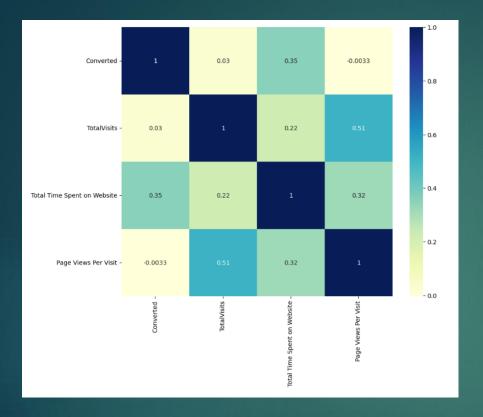




EDA plots depicting variation in numerical columns for those who Converted and those who didn't.



EDA plots depicting variation in numerical columns for those who Converted and those who didn't.

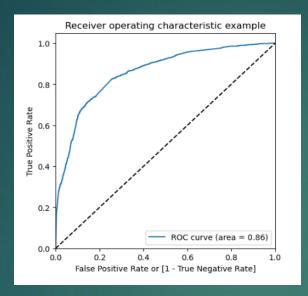


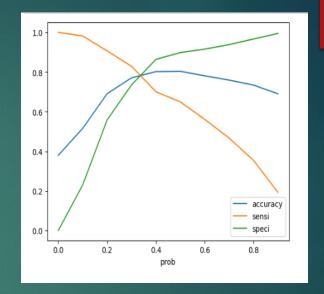
EDA plots depicting correlation (Heat Map) of all selected numerical columns.



EDA plots depicting correlation (Heat Map) of all selected columns (numerical columns and dummy columns).

--0.2





Linear Regression Final Model Parameters
Area under ROC = 0.86
Optimum point for cut-off = 0.30

## Inference / Conclusion

## Model Analysis

Performance of our Final Model

Overall accuracy on Test set: 77.52 %

Sensitivity of our logistic regression model: 83.01%

Specificity of our logistic regression model: 74.13%

# Inferences from Model

Business Insights Derived from our Model

Top 3 variables in model, that contribute towards lead conversion are:

- Total Time Spent on Website
- Last Notable Activity\_SMS Sent
- Total Visits

# Inferences from Model

Business Insights Derived from our Model

Top 3 variables in my model, that should be focused are:

- Last Activity\_SMS Sent (positively impacting)
- Last Activity\_Olark Chat Conversation (negatively impacting)
- Lead Source\_Olark Chat (negatively impacting)

## Conclusion (LR Model)

- OUR LOGISTIC REGRESSION MODEL IS DECENT AND ACCURATE ENOUGH, WHEN
   COMPARED TO THE MODEL DERIVED USING PCA, WITH 77% ACCURACY ON TEST SET, 83
   % SENSITIVITY AND 74 % SPECIFICITY.
- WE CAN VARY THESE PARAMETERS BY VARYING THE CUT-OFF VALUE AND THUS PREDICT
  HOT LEADS BASED ON SCENARIOS LIKE AVAILABILITY OF EXTRA RESOURCES AND VICEVERSA.
- LEAD SCORE CALCULATED IN THE TRAINED SET OF DATA SHOWS THE CONVERSION RATE
   ON THE FINAL PREDICTED MODEL IS AROUND 80%