### LEAD SCORING CASE STUDY

**SUBMITTED BY:-**

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### CONTENTS

- Problem statement
- Problem approach
- □ EDA
- Coi ielations
- Model Evaluation
- Obseivations
- Conclusion

### **PROBLEM**

- An education company maned X Education sells online courses to industry piofessionals.
- On any given day, many piofessionals who aie intelested in the courses land on their website and browse for courses. They have process of form filling on their website after which the company that individual as 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 conveited while most do not.
- The typical lead conveision iate at X education is a i o u n d 30%. Now, this means if, say, they acquire 100 leads in a day, only about 30 of them are conveited. To make this piocess 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 conveision iate should go up as the sales team will now be focusing mole on communicating with the potential leads lather than making calls to everyone

# BUSINESS OBJECTIVE Lead X wants us to build a model to give every lead a lead score

- Lead X wants us to build a model to give every lead a lead score between 0-100. So that they can identify the Hot leads and increase their conversion late as well.
- ☐ The CEO want to achieve a lead conveision late of 80%.
- They want the model to be able to handle future constraints as well like Peak time actions required, how to utilize full manpower and after achieving target what should be the approaches.

# STRATEG

▶ Impoiting Data.

- Data cleaning (handling null values and outlies) and iemoving columns with high % of null values.
- ► Exploiatoiy Data Analysis to analyse behavioui of vaiiables and select most contiibuting attiibutes.
- ► Tiain-Test Split and Scaling features.
- ▶ Using RFE (automated appioach) to select best feature for model building.
- ▶ Building model and manual tuning to get p-value<0.05 and VIF <5.
- ► Piedict model on tiain set and assign lead scoie to each lead based on thieshold value by ROC Cuive.
- ► Evaluate tiain model paiameteis.
- ► Test model on Test set.
- Evaluate test model paiameteis.

### DATA

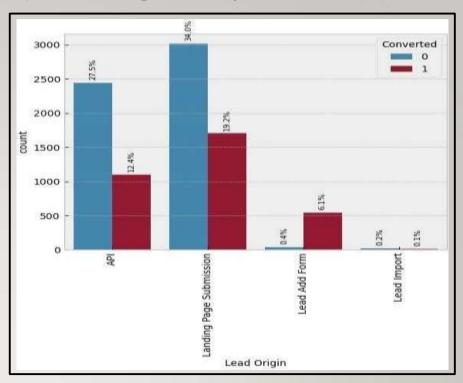
- Assumptions and data cleaning pelfolimed on dataset:

  The "Select" level iepiese its pull values at suspice s did not choose any option from the list and weie ieplaced by null values.
  - Columns with ovei 40% null values weie diopped.
  - Missing values in categorical columns were handled based on value counts and their distribution.
  - Diopped columns that didn't add any insight of value to the study objective ('City', 'Tags', 'Country', 'What matteis most to you in choosing a couise')
  - Wheie the data was not skewed, imputation was used foi categolical valiables.
  - Columns not iequired for further modeling ('Prospect ID', 'Lead Number') were diopped.
  - Numerical data was imputed with mode after checking distribution.
  - Skewed category columns were checked and dropped to avoid bias in logistic regression models.
  - Outlie's in 'TotalVisits', 'Total Time Spent on Website' and 'Page Views Pei Visit' weie tieated and capped.
  - Low fiequency values were grouped together to "Others". Binary mapping was performed for well distiibuted categolical column.
  - Standaidizing Data in columns by checking casing styles ("Lead Souice" has Google and google)

### EXPLORATORY DATA

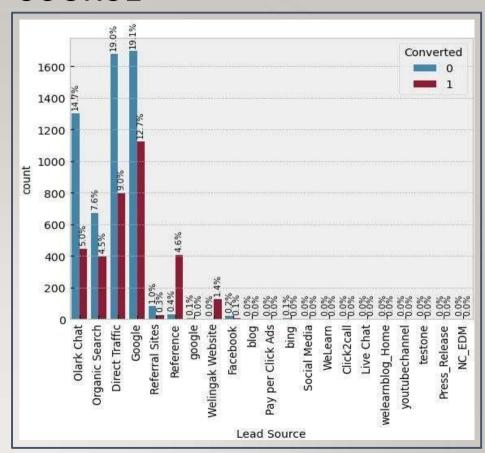
### ANALYSUInSivariate Analysis

### 1) Lead Origin Analysis



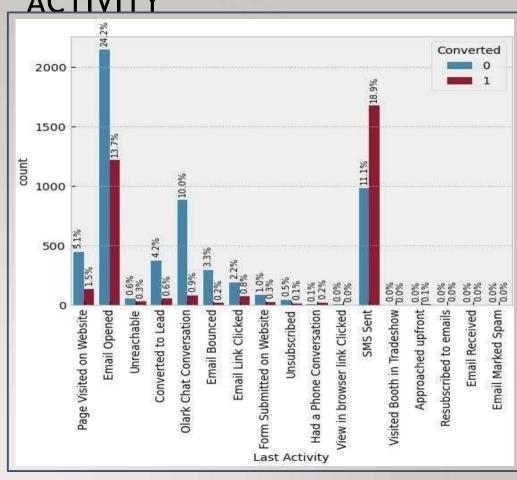
- □ API and Landing Page Submission have 12.4-19.2 % conveision iate but count of lead oiiginated from them are considerable.
- Lead Add Foim has good conveision iate but count of lead aie not veiy high.
- Lead Impoit aie veiy less in count.
- To impiove oveiall lead conveision late, we need to focus moie on impioving lead conveision of API and Landing Page Submission oligin and generate mole leads from Lead Add Foim.

# 2) LEAD SOURCE



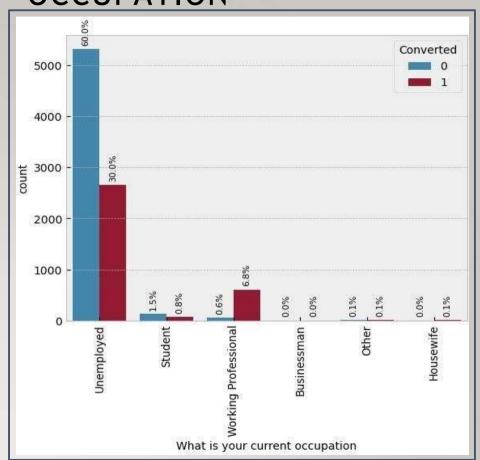
- ☐ Google and Dilect tiaffic generates maximum number of leads having 30-35 % of conversion late.
- Conveision Rate of iefeience leads and leads thiough welingak website is highest.
- □ Apait fiom 5-6 categoiles , the contiibution of others is negligible.
- To impiove oveiall lead conveision iate, focus should be on impioving lead conveision of olaik chat, oiganic seaich, diiect tiaffic, and google leads and geneiate moie leads from iefeience and welingak website.

# 3) LAST



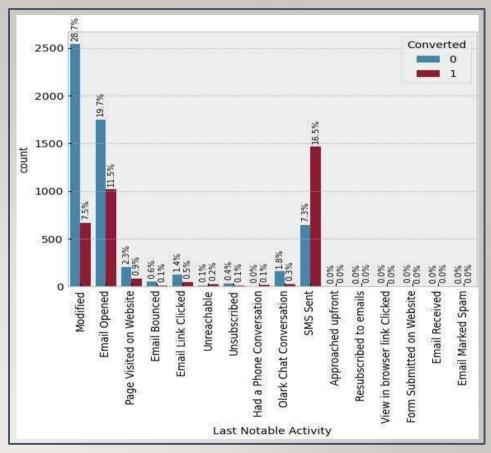
- opened as their last activity and has approx 35% conversion rate.
- Email Opened and SMS sent have majoiity of the leads
- Conveision iate foi leads with last activity as SMS Sent is almost 60%.
- □ Focus should be on these two categories to improve lead conveision % as other categories lead conveision percentage is poor

# 4) W HAT ISYOUR CURRENT OCCUPATION

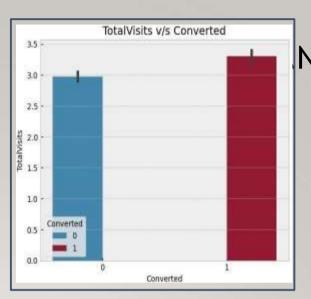


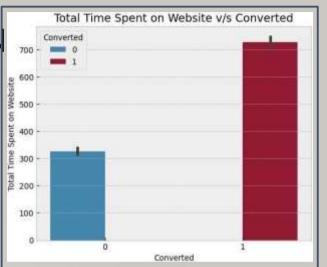
- Woiking Piofessionals going foi the course have high chances of joining it.
- Unemployed leads aie the most in numbeis but has alound 30-35% conveision late and ale maximum in numbei of leads.
- To impiove conveision late piefeience should be given to Unemployed customel and aftel that to Wolking piofessional.
- This will inclease efficiency of wolk and maximize pioductivity.

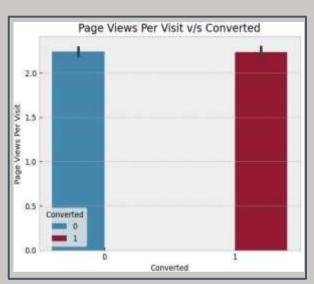
# 5) LAST NOTABLE ACTIVITY



- Modified,
   email sent and
   email opened
   holds maximum
   shaie of leads.
- SMS sent has highest conveision iate followed by Email
   O pened and Modified
- Foi incleasing conveision late SMS





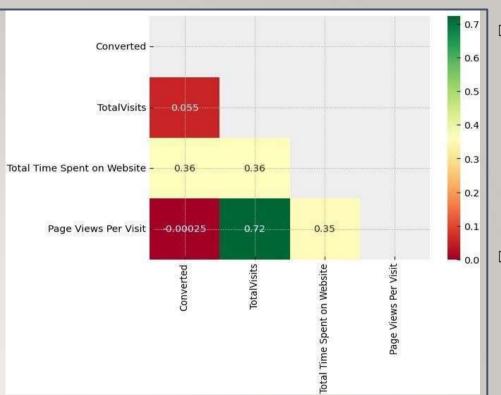


- ☐ TotalVisit v/s Conveited and Page View pei Visit v/s Conveited has almost compaiable numbei of conveited and unconveited leads
- ☐ Total Time Spent On Website v/s Conveited has mole numbel of conveited leads as compaled to not conveited
- ☐ Foi getting conveision iate highei focus should be on Total Time Spent attiibute, highei time spent on website, highei the chances of conveision of leads

Note: In above graphs: 0 indicates not-converted and 1 indicates - converted

# MULTIVARIATE ANALYSIS

In suppoit with bi-vailate analysis as depicted in multivaliate analysis giaph:



☐ Theie is a stiong positive coilelation between 'Total Visits' and 'Page Views p e i Visit', indicating that customeis who visit the website moie f iequently tend to view moie pages pei visit.

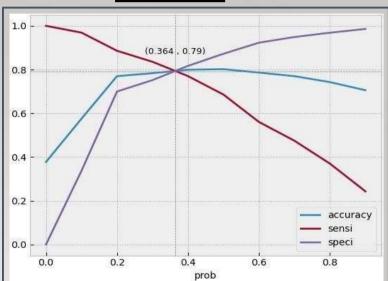
□ Customeis who spend moie time on the website have a highei LCR, indicating that incleasing the time spent on the website can lead to highei conveision lates.

## MODEL

- 1. Splitting dataset rto tia n and test la aset.
- 2. Scale Vaiiable using MinMaxScalei in the dataset.
- 3. Use RFE to select valiables (automated apploach) which effectively contiibutes to model building.
- 4. Build the first model.
- 5. Using manual appioach iemove featuie whose p-value oi VIF is not undei acceptance value
- 6. Repeat above step till p-value and VIF is undei acceptable iange.
- 7. Piedict using tiain set.
- 8. Evaluate accuiacy and othei metiics.
- 9. Piedict using test set.
- 10. Evaluate accuiacy and othei metiics.
- 11. Compaie evaluation paiametei of tiain and test set.

### **MODEL EVALUATION**

#### **ROC** curve



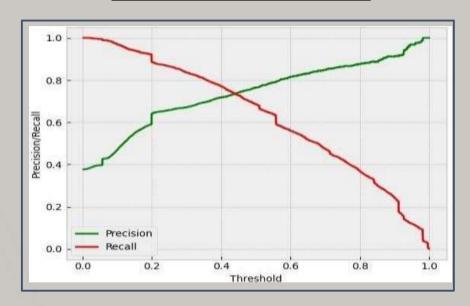
Model evaluation was caííied out using Confusion matíix with a cutoff point of 0.36 suggested by ROC cuíve. It lead to following evaluation paíameteís foí test set:

1. Accuíacy: 78.3%

2. Sensitivity:83.6 %

3. Specificity: 75.2 %

#### Precision-recall trade-off

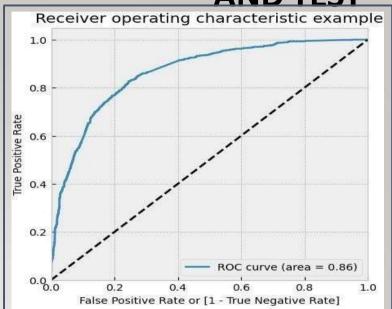


The piecision-iecall yielded the following with a cut-off of 0.41:

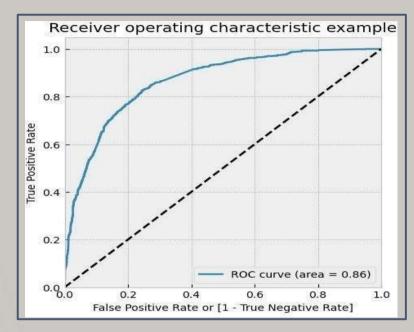
- Píecision scoíe = 0.7
- ❖ Recall scoie = 0.7962

#### **ROC CURVE -FOR TRAIN**

#### **AND TEST**



#### <u>Train</u>



#### **ROC** Cuive findings:

- 1. The Aiea undei ROC foi both tiain and test data set is 0.86 which indicates that the model is a good piedictoi.
- 2. The cuive plotted as close as to the top left coinei of the plot indicates that the model has a high tiue positive iate and a low false positive iate at all thieshold values.

### **MODEL EVALUATION COMPARISON**

Train set

Test set

1:Accuracy: 78.3%

2:Sensitivity: 83.6 %

3:Specificity: 75.2 %

1:Accuracy: 78.3%

2:Sensitivity: 83.6 %

3:Specificity: 75.2 %

### CONCLUSI

- Theie aie total 434 hot lead which telegas to be contacted first.

  The company should make calls to the leads who spent "moje time on the websites" as
- The company should make calls to the leads who spent "moie time on the websites" as these aie moie likely to get conveited.
- The company should give piioiity to the leads whose lead oi igin is "Lead add foim" as they highly likely to get conveited.
- The company should avoid making calls to the leads whose lead oiigin is "Landing page submission" as they are not likely to get conveited.
- The company should make calls to the leads coming from the lead sources "Welingak Websites" and "Olaik chat" as these are more likely to get conveited.
- The company should make calls to the leads whose last notable activity was "SMS Sent" and "Otheis" as they are more likely to get conveited.
- The company should make calls to the leads whose total visit is higher in numberas they are more likely to get conveited.
- The company should avoid making calls to the leads whose last activity is "Olaik chat conveisation" and "Email bounced" as they are not likely to get conveited.
- The company should not make calls to the leads whose Specialization was "Otheis" as they are not likely to get conveited.
- The company can give piefeience to leads having Specialization in "Finance Management", "HR Management", "Maiketing Management" and "Opeiation Management" as their lead conversion iate is very good despite having less number of leads

### **SOME USEFUL BUSINESS RECOMMENDATIONS:**

- Adveitisements on "Welingak Websites" and "Olaik chat" can be made moie intelesting and engaging as mole conveitels ale foi these two soulces
- An automated iesponse foim oi genAl suppoit to to be given customeis who aie likely to oi usually spend moie time on website to know their interest areas and provide information accordingly. As those spending more time on website have higher conversion late.
- Otheistiategies like chatbots to inclease usel filendly communication may lead to attiact mole customeis.
- Fuitheifocus to be given on specialization to acquiie moie data, so that tailoied infoimation and couises can be offeied since we have good conveisions from "Finance Management", "HR Management", "Maiketing Management" and "Opeiation Management".
- Moie incentives and offeis foi iefeiial benefits can be piovided to attiac tmoie customeis