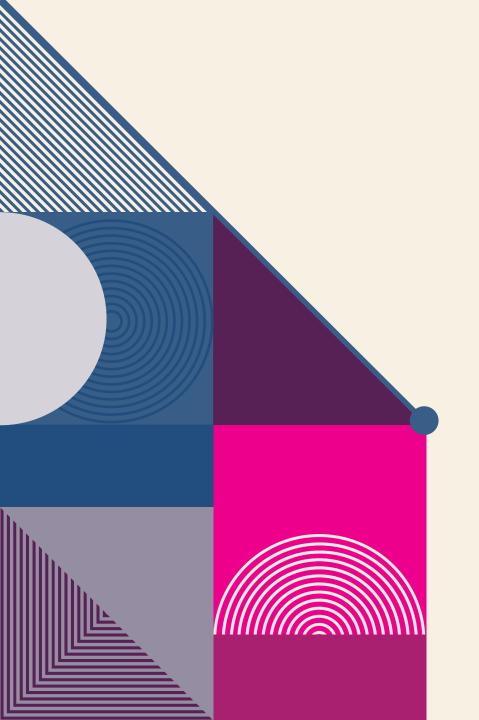
# LEAD SCORING CASE STUDY

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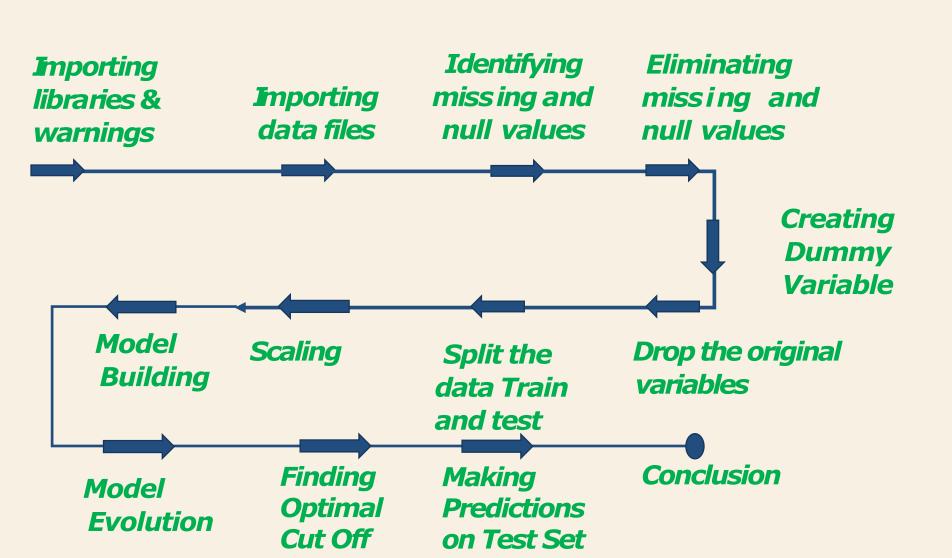
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# PROBLEM STATEMENT

- To help X education to select the most promising leads known as 'hot leads' who are most likely to convert into paid customers.
- Build a logistic regression model to assign a lead score between 0 and 100 to each of the leads where the
  leads with higher lead score have a higher conversion chance and the leads with lower lead score have a
  lower conversion chance.
- Identify the driver variables and understand their significance which are strong indicators of lead conversion.
   Identify the outliers, if any, in the dataset and justify the same.
- Consider both technical and business aspects while building the model.
- Summarize the conversion predictions by using evaluation metrics like accuracy, sensitivity, specificity and precision.

### **WORK FLOW**



### **IMPORTING LIBRARIES AND WARNINGS**

Importing libraries
Imported pandas, numpy, matplotlib & seaborn for data loading & visualization
Imported sklearn for model building and statsmodels for model evolution.

Importing Warnings
Highlights warnings
however the program
runs.

### READING DATASET

Datafile is extracted from the given dataset. namely 'Leads.csv'

Highlighted datafile description, shape etc., in the notebook for elaborated experience in reading the data.

### DATA CLEANING AND PREPARATION

#### Leads.csv:

- Following columns contain more than 3000 null values initially, hence dropped those columns:
  - > Tags
  - Lead Quality
  - > Asymmetrique Activity Index
  - ➤ Asymmetrique Profile Index
  - > Asymmetrique Activity Score
  - > Asymmetrique Profile Score

### **DUMMY VARIABLE CREATION**

- Check the columns which are of type 'object'
- Create dummy variables using the 'get\_dummies' command for following columns 'Lead Origin', 'Lead Source', 'Do Not Email', 'Last Activity', 'What is your current occupation', 'A free copy of Mastering The Interview', 'Last Notable Activity'
- Add the results to the master dataframe
- Creating dummy variable separately for the variable 'Specialization'
- Drop the variables for which the dummy variables have been created

# **TEST-TRAIN SPLIT & SCALING**

- Put all the feature variables in X
- Put the target variable in y
- Split the dataset into 70% train and 30% test
- Scale the three numeric features i.e. 'TotalVisits', 'Page Views Per Visit', 'Total Time Spent on Website' present in the dataset

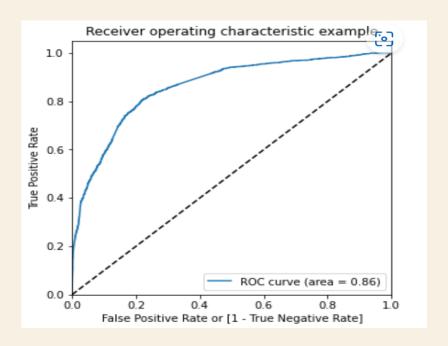
# **MODEL BUILDING**

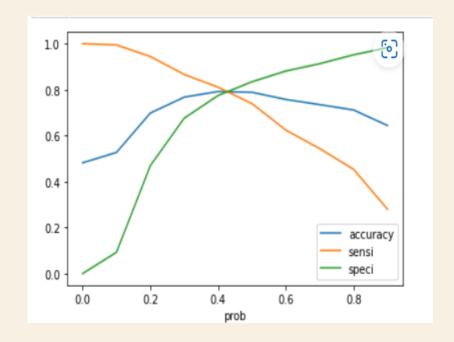
- Import RFE and select 15 variables
- Put all the columns selected by RFE in the variable 'col'
- Fit a logistic Regression model on X\_train after adding a constant and output the summary
- Make a VIF dataframe for all the variables present
- VIFs seem to be in a decent range except for 'Lead Origin\_Lead Add Form', 'Lead 'Source\_Reference' and 'Lead Source\_Welingak Website'. Let's first drop the variable 'Lead Source\_Reference' since it has a high p-value as well as a high VIF.
- Refit the model with the new set of features
- Building Model by removing the variable whose p-value is greater than 0.05 and vif value is greater than 5%
- Predictions on test data set
- Overall accuracy 79%

	Features	VIF
2	Lead Origin_Lead Add Form	84.19
4	Lead Source_Reference	65.18
5	Lead Source_Welingak Website	20.03
11	What is your current occupation_Unemployed	3.65
7	Last Activity_Had a Phone Conversation	2.44
13	Last Notable Activity_Had a Phone Conversation	2.43
1	Total Time Spent on Website	2.38
0	TotalVisits	1.62
8	Last Activity_SMS Sent	1.59
12	What is your current occupation_Working Profes	1.56
3	Lead Source_Olark Chat	1.44
6	Do Not Email_Yes	1.09
10	What is your current occupation_Student	1.09
9	What is your current occupation_Housewife	1.01
14	Last Notable Activity_Unreachable	1.01

20XX Pitch deck title

# **ROC CURVE**





- > Finding Optimal Cut off Point
- Optimal cut off probability is that probability where we get balanced sensitivity and specificity.
- > From the second graph it is visible that the optimal cut off is at 0.41.

### CONCLUSION

It was found that the variables that mattered the most in the potential buyers are (In descending order):

- > The total time spend on the Website.
- > Total number of visits.
- When the lead source was:
  - i. Google
  - ii. Direct traffic
  - iii. Organic search
  - iv. Welingak website
- ➤ When the last activity was:
  - i. SMS
  - ii. Olark chat conversation
- When the lead origin is Lead add format.
- When their current occupation is as a working professional. Keeping these in mind the X Education can flourish as they have a very high chance to get almost all the potential buyers to change their mind and buy their courses.

