# Lead Scoring Case Study

ML Assignment

## High level Approach

# Data Cleansing & Analysis

- Inspecting and cleaning data set
- identify and handle null values
- Univariate, bivariate analysis
- Correlation analysis
- Multivariate analysis

#### **Model Building**

- Transformation- Dummy variable creation
- Train-test split
- Build logistic regression model
- Use RFE to reduce the features

#### **Model Evaluation**

- Accuracy, Precision metrics
- Optimum cutoff point
- RIC curve

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## Fixing Null values

- Identify all columns that have more than 45% null values
- Drop the columns from the dataset
- Remove unwanted columns from the dataset
- Impute the null values wherever possible

- Following columns are removed as don't add to the modeling - 'Prospect ID', 'Lead Number'
- 'Select' is converted to Null for all columns
- Country and City columns are removed as they show high class imbalance
- What matters most to you in choosing a course is dropped as it has too many null values

## Following columns are dropped as they are skewed or have high null values

'Tags',

'Do Not Call',

'Search',

'Magazine',

'Newspaper Article',

'X Education Forums',

'Newspaper',

'Digital Advertisement',

'Through Recommendations',

'Receive More Updates About Our Courses',

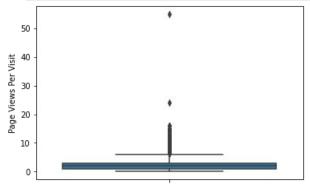
'Update me on Supply Chain Content',

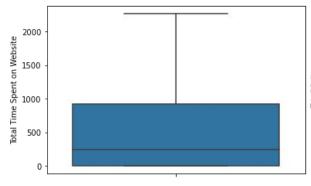
'Get updates on DM Content',

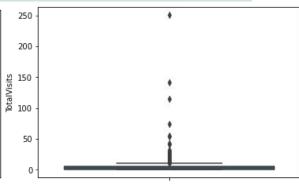
'I agree to pay the amount through cheque'

#### Data corrections

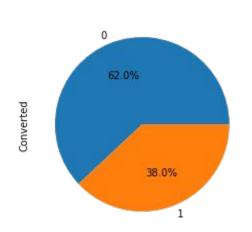
- Dropped the rows with null Total visits
- Replaced Null values and combining with low frequency values of Last Activity
- Wrong values of Lead Source are corrected
- No outliers seen with the variables 'Page View Per Visit', 'Total Time Spent on website', and 'Total Visits'

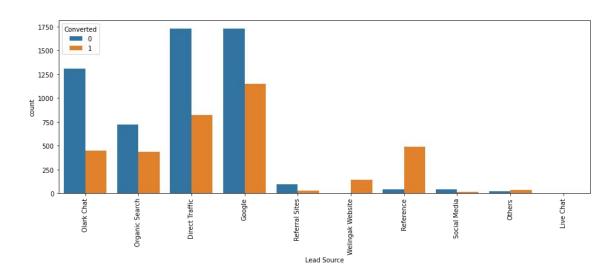






## Data analysis





38% of the clients got converted, which is a good data set to train the model without major class imbalance

Google and Direct Traffic are the highest source of leads

Total columns at this stage : 12

## Data Preparation and fitting the model

- Converted binary variables to dummy variables
- Dataset is split to Train and Test in the ratio of 70:30
- Total number of columns at this stage is 67
- Scaled the numerical variables using MinMax scaler
- Created the logistic regression model using statsmodels

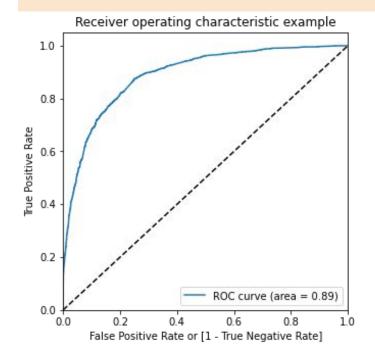
## Model Building and RFE

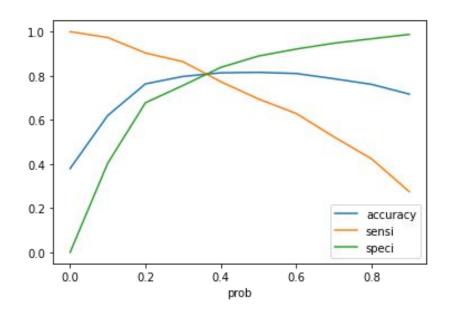
- RFE used to find the most significant 15 features
- Rebuilt the model based these features
- What is your current occupation\_Housewife variable is dropped as it was found to have high p-value
- Rebuild the model after this
- All p-values of the variables are in the reasonable range (below 0.05%)
- VIF values are less than 5 for all variables
- This can be treated as the final model with 14 variables

### **Model Evaluation**

- Initial Confusion matrix calculated for the trained values with 50% as cut off
  - Accuracy: 81%, Sensitivity: 69%
- Area under the ROC curve : 89%

- New cut off arrived based on accuracy plot :
  38% Accuracy : 81%, Sensitivity : 78%
- Lead score is calculated with 38% conversion probability





### Conclusion

- Customers who make more visits and spend more time spent on the website have higher chances of conversion
- Customers whose leads came via Olark chat and Welingak website have high conversion probability
- Working Professionals likely to have higher conversion chances
- People who opted NOT to be emailed have lower chances of conversion
- Last Activity of SMS or others have higher chances of conversion