

# **Summary**

## **Problem statement**

An education company named X Education sells online courses to industry professionals. The company wishes to find the "Hot leads".

X Education focuses on to select the most promising leads, the leads that are most likely to convert into paying customers. The company requires you to build a model wherein you need to assign a lead score to each of the leads such that the customers with a higher lead score have a higher conversion chance and the customers with a lower lead score have a lower conversion chance.

The CEO in particular has given a ballpark of the target lead conversion rate to be around 80%.

## **Solution Summary**

### **Understanding the Data**

Reading and understanding the data. And analyzing their data description and information.

### **Data cleaning**

- Columns with greater than 35% of null values has been dropped, which includes Imputing the missing values as and where required.
- The outliers were identified and removed.
- Numerical categorical data were imputed with mode and columns with only one unique response from customer were dropped.

### **Data Analysis**

- Data imbalance checked- only 38.5% leads converted.
- Performed univariate and bivariate analysis for categorical and numerical variables. Provided the valuable insight on effect on target variable.
- Time spend on website shows positive impact on lead conversion.

## **Data Preparation**

- Created dummy features (one-hot encoded) for categorical variables.
- Splitting Train & Test Sets.
- Feature Scaling using Standardization.
- Dropped few columns, they were highly correlated with each other.

## **Model Building**

- Used RFE to reduce variables from 48 to 15. This will make data frame more manageable.
- Manual Feature Reduction process was used to build models by dropping variables with p – value.
- Total 3 models were built before reaching final Model 4 which was stable with (p-values < 0.05). No sign of multicollinearity with  $VIF < 5$ .
- Logm4 was selected as final model with 12 variables; we used it for making prediction on train and test set.

## **Model Evaluation**

- Confusion matrix was made and cut off point of 0.345 was selected based on accuracy, sensitivity and specificity plot.
- This cut off gave accuracy, specificity and precision all around 80%. Whereas precision recall view gave less performance metrics around 75%.
- As to solve business problem CEO asked to boost conversion rate to 80%, but metrics dropped when we took precision-recall view.
- So, we will choose sensitivity-specificity view for our optimal cut-off for final predictions.
- Lead score was assigned to train data using 0.345 as cut off.

## **Making predictions on test data**

- Making Predictions on Test: Scaling and predicting using final model.
- Evaluation metrics for train & test are very close to around 80%.
- Lead score was assigned.
- Top 3 features are:
  - Lead Source\_Welingak Website
  - Lead Source\_Reference
  - Current\_occupation\_Working Professional