

# LEAD SCORING CASE STUDY

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## **Introduction:**

- X Education has appointed you to help them select the most promising leads, i.e. 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 higher lead score have a higher conversion chance and the customers with 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%.

## **Problem Statement:**

- Build a logistic regression model to assign a lead score between 0 and 100 to each of the leads which can be used by the company to target potential leads. A higher score would mean that the lead is hot, i.e. is most likely to convert whereas a lower score would mean that the lead is cold and will mostly not get converted

## **Data Provided :**

- Two Spread sheets containing all the information of the customer when they inquire or visited website. The other spread sheet gives information about what variable have gathered

# SOLUTION APPROACH

## Approach for case study:

- Load/ Read all data provided and identify the data structure, information about data like data types, missing values.
- Data Cleaning Techniques :
  - ❖ Investigate the missing values and impute that with suitable method viz, mean, mode, median etc
  - ❖ Check outliers in data and take suitable action
  - ❖ Check data imbalance and treat that
- Data analysis
  - Cleaned data to investigate the relationship of variables with that of TARGET variable
  - Get the train and test set by splitting the cleaned data set
  - Use Recursive Feature Elimination (RFE) methodology from sklearn to get variable which are most effective
  - Get the matrices (TP,FP, TN and FN ) and then calculate the necessary numbers such as sensitivity, specificity, precision and recall at optimum probability value. This Optimum number can be fetch by building RoC curve
  - The above process need to done for both train and test data set
  - Finally Calculate the lead score at optimum probability value (lead score = predicted probability x100)

# RESULTS

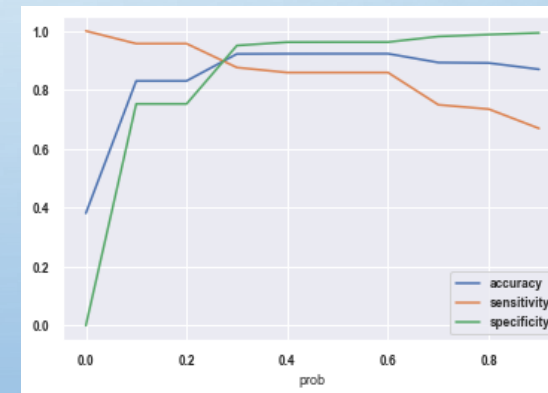
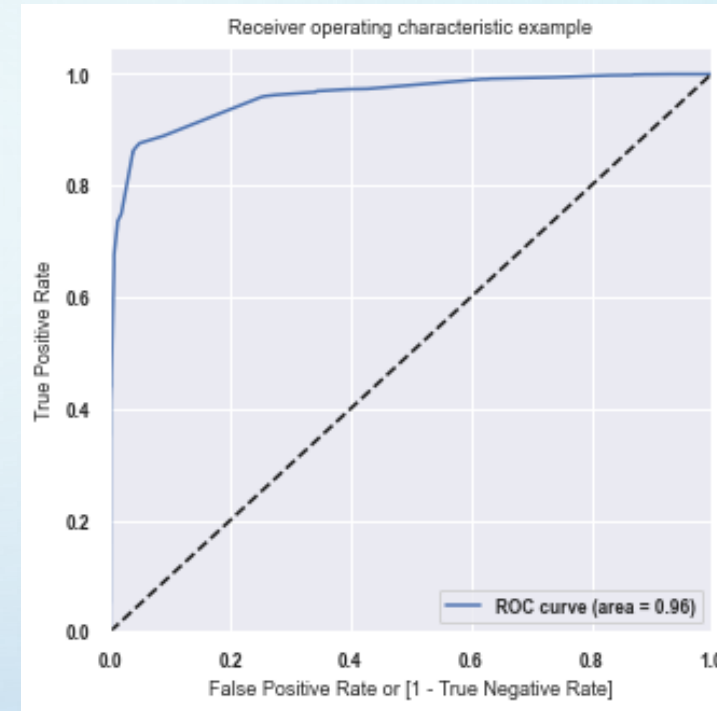
## 1. MODEL 2 SHOWS THE OPTIMUM VIF FOR DIFFERENT VARIABLE AND THAT IS CHOSEN AS GOOD MODEL

## 2. ROC CURVE

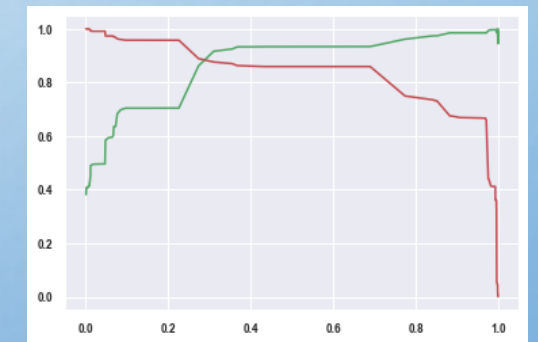
- The area under the curve is ~96% which is good model for prediction.
- Now based on this get the insight for specific parameters viz sensitivity, specificity, precision and recall

## 3. OPTIMUM CURVE

- The accuracy, sensitivity and specificity lines are intersecting at ~0.25 probability. So, we will proceed with this value.



Accuracy, sensitivity and specificity



Precision and recall



# RESULTS

Sr no	Accuracy (%)	Sensitivity(%)	Specificity (%)	Precision (%)	Recall (%)
Train data	92	86	96	93	86
Test Data	91	90	91	87	90

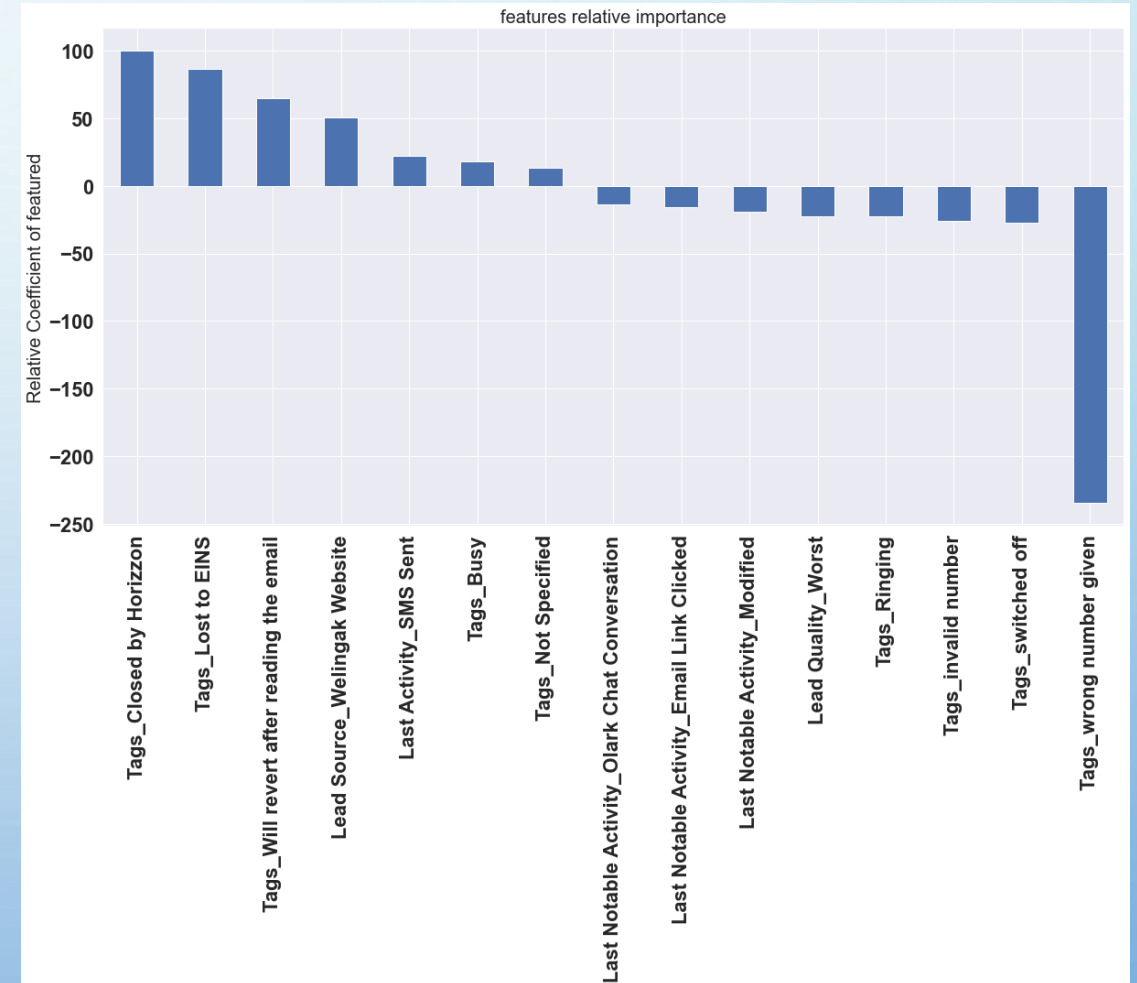
# RESULTS : MAIN FEATURES

## Top 5 variables which can fetch the lead:

- Tags\_Closed by Horizzon
- Tags\_Lost to EINS
- Tag\_We will revert after reading the email
- Lead Source\_Welingak Website

## 5 variables which need more attention for converting to lead:

- Tags\_wrong number given
- Tag\_switched off
- Tag\_invalid number
- Tag ringing
- Lead Quality worse



# CONCLUSION

- Some top leads for conversion need follow up
- Some features which relatively low importance need more focused from business
- Some of leads which talks about do not email, MIGHT BE or worse can be given low priority

THANK YOU