## **Lead Scoring Case Study Summary**

### 1. Problem Statement:

X Education, an online course provider for industry professionals, seeks assistance in identifying the most promising leads likely to convert into paying customers. The objective is to develop a lead scoring model that assigns lead scores, ensuring higher conversion chances for leads with higher scores and lower chances for those with lower scores. The CEO has set a target lead conversion rate of approximately 80%.

## 2. Solution Summary:

### **Step1: Reading and Understanding Data:**

The dataset was reviewed and inspected to understand its structure and content.

### Step2: Data Cleaning:

- a. Identified and dropped variables with unique values.
- b. Replaced 'Select' values with null values in columns where leads did not choose any option.
- c. Removed columns with more than 70% NA values and addressed imbalanced and redundant variables.
- d. Imputed missing values, handled outliers, and standardized labels for consistency. For instance, standardized city names and imputed missing values based on the predominant location (Mumbai in this case).

### **Step3: Data Transformation:**

Converted binary variables to '0' and '1' and transformed the target variable to indicate lead conversion status (1 for converted, 0 for not converted).

### **Step4: Dummy Variables Creation:**

Created dummy variables for categorical variables and eliminated repeated or redundant variables.

### **Step5: Test Train Split:**

Partitioned the dataset into training and testing sets with a 70-30 split.

## **Step6: Feature Rescaling:**

- a. Applied Min-Max Scaling to normalize numerical variables.
- b. Examined variable correlations via heatmap and removed highly correlated dummy variables.

#### **Step7: Model Building:**

- a. Employed Recursive Feature Elimination to select the 15 most important features.
- b. Utilized statistical analysis, including P-values, to identify significant variables and exclude insignificant ones.
- c. Arrived at 13 significant variables with acceptable VIFs.
- d. Determined optimal probability cut-off for the final model by assessing accuracy, sensitivity, and specificity.

- e. Evaluated model performance using ROC curve analysis, achieving an area under the curve of 89%.
- f. Ensured 80% prediction accuracy based on the converted column.
- g. Assessed precision, recall, accuracy, sensitivity, and specificity on the train set.
- h. Determined a cut-off value of approximately 0.3 based on Precision and Recall trade-off.
- i. Implemented learnings on the test model and calculated conversion probability with Sensitivity and Specificity metrics, achieving 89.98% accuracy, 81.79% sensitivity, and 94.95% specificity.

# **Step 8: Conclusion:**

- The lead scoring model predicted a conversion rate of 89.37% on the test set, aligning with the CEO's target rate of around 80%.
- The model's high sensitivity facilitates the identification of promising leads.
- Key features contributing to conversion probability include Lead Origin (Lead Add Form), Current Occupation (Working Professional), and Total Time Spent on Website.