

Machine Learning model to predict if a client will subscribe to the product, given his/her demographic and marketing campaign related information.

1. Initial Findings about Data:

1. There are we have **45211** observations of **17** variables in original dataset (**7**-Numerical Variables and **10**-Categorical Variables).
2. No explicit missing values but there are many '**unknowns**' values for some Categorical Variables that will be treated as missing values.
3. From the distribution of Target variable: "**is_success**" it is found that data is **imbalanced** because there is approx. 88% is 'no' and 12% is 'yes'.

2. Exploratory Data Analysis

A. For Numerical Variables

1. Analysis of each Numerical variable by plotting **Boxplot** with respect to target variable.
2. Some Independent numerical variable ('**balance**', '**duration**', '**campaign**', '**pdays**', '**previous**') contains many outliers.
3. I choose a **range** based on **Maximum** and **Minimum** value for each Numerical variable by observing Boxplot of corresponding variable. Any value out of this range will be treated as Outlier and same will be imputed by **Mean** of corresponding variable.

B. For Categorical Variables

1. Analysis of each Categorical variable by plotting **Crosstab** with respect to target variable.
2. If any Categorical variable has **more than 50%** 'unknown' values ('**poutcome**') or seems **highly unbalanced** ('**default**') or seems having **negligible impact on target** variable ('**contact**'), we can drop that variable from dataset.
3. Variables having **less than 50%** 'unknown' values are imputed by **Mode** of respective variable.

3. Feature Engineering

1. Created new **dummy variables** to convert Categorical into Numerical.
2. Total variables after creating dummies becomes 39.

4. Feature Selection

1. Feature selection by **Principal Component Analysis**. I have selected first **32 components** out of 39.

5. Model Training

1. Implement **Logistic Regression**, **Linear Discriminant Analysis**, **K-Nearest Neighbor**, **Decision Tree**, **Naive Bayes**, **Support Vector Machine** along with **Cross Validation**.

6. Model Selection

1. "Support Vector Machine" has highest Accuracy (**89.35%**) but it is taking **more time** compare to other algorithms.
2. "Logistic Regression" also has nearly same accuracy (**89.17%**) and it is very **faster** than SVM.
3. So I have considered Logistic Regression as Best model for prediction.

7. Prediction

1. Prediction on Validation Dataset by Logistic Regression with following result:
 - ✓ Accuracy – 0.88
 - ✓ F1-score – 0.87