

# Deliverable Week 9

**Group Name:** Destined Data Team

**Specialization:** Data Science

## **Team members:**

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## **Problem description**

ABC bank wants to sell its term deposit product to customers and before launching the product they want to develop a model which will help them in understanding whether a particular customer will buy their product or not (based on customer's past interaction with bank and other Financial Institutions).

## Data Cleaning and Transformation

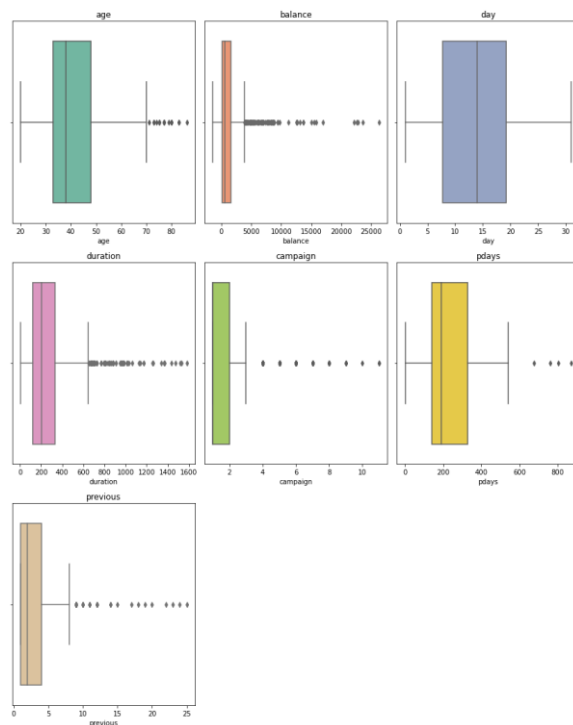
The dataset was checked for missing values, duplicates, outliers, skewness

No missing values or duplicates were found. Summary statistics such as the mean, standard deviation, distribution, and kurtosis, skewness were checked.

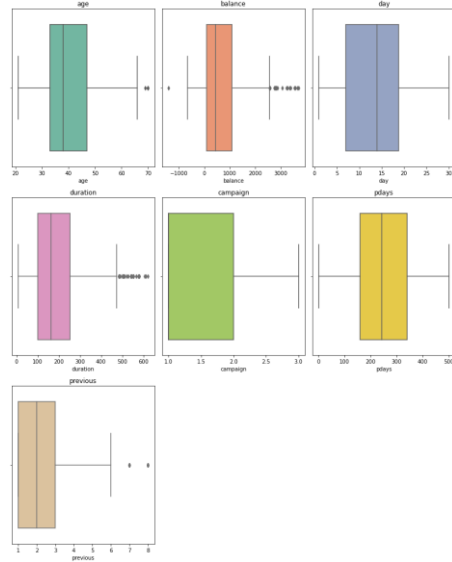
**Github repo link:** <https://github.com/PraneethaRajupalepu/Bank-DataScience-Project>

## Results and approaches

- EDA was performed for continuous and categorical variables
- Suggested methods for dealing with outliers
  - Method 1: Flooring and capping using Interquantile Range (IQR)  
Outliers are removed by dropping any values that are below 25<sup>th</sup> percentile and above 75<sup>th</sup> percentile.



**Figure 1 - Before Transformation**



**Figure 2 - After Transformation**

After transformation of the data using the interquartile range, there is less outliers.

- **Method 2: Binning of outliers**

Continuous variables were binned so that outlier values are converted to count values.

- We decided to keep outliers. A different method of dealing with outliers without removing them is to use the Minkowski error which reduces the impact that outliers will have on the model.
- **Weight of Evidence (WOE) and Information Value (IV)**  
We identified variables that had significant predictive power for modeling using WOE and IV.