# **Credit Card Fraud Detection - Explanation in Points**

# 1. Importing Libraries

• Various libraries are imported to perform data manipulation, visualization, preprocessing, model training, and evaluation.

### Key Libraries Used:

- o pandas, numpy: For data handling and numerical computations.
- o matplotlib.pyplot, seaborn: For visualization.
- sklearn.model\_selection: For splitting data and hyperparameter tuning.
- o sklearn.preprocessing: For scaling numerical features.
- sklearn.ensemble, sklearn.linear\_model: Machine learning models (Random Forest, Logistic Regression).
- o sklearn.metrics: For performance evaluation.
- o imblearn.over\_sampling: To handle class imbalance using SMOTE.

## 2. Loading the Dataset

- The dataset is loaded using pd.read\_csv().
- Basic dataset information is displayed:
  - shape of the dataset.
  - o First 5 rows using data.head().
  - General dataset info (data.info()).
  - Class distribution (data['Class'].value\_counts()) to check fraud vs. non-fraud transactions.

### 3. Data Preprocessing

### Checking for Missing Values:

o data.isnull().sum() is used to detect missing values in the dataset.

#### Feature Scaling:

o The Amount feature is normalized using StandardScaler to standardize its values.

## Dropping Unnecessary Features:

o The Time feature is removed as it is not relevant for fraud detection.

## • Separating Features & Target Variable:

- X contains all features except the target column (Class).
- o y contains the target labels (Class), which indicate fraud (1) or non-fraud (0).

# 4. Handling Class Imbalance

- The dataset is highly imbalanced (fewer fraud cases than non-fraud).
- Steps Taken to Address Imbalance:
  - Handling Missing Values:
    - SimpleImputer(strategy='mean') replaces NaN values with the mean.
  - Applying SMOTE (Synthetic Minority Over-sampling Technique):
    - Generates synthetic samples for the minority class (fraud cases) to balance the dataset.

# 5. Splitting the Dataset

- The resampled data (X\_resampled, y\_resampled) is split into training and testing sets.
- Train-Test Split:
  - o train\_test\_split(X\_resampled, y\_resampled, test\_size=0.3, random\_state=42) is used to:
    - Allocate 70% of data for training.
    - Allocate 30% for testing.

## Conclusion

- This project aims to detect fraudulent transactions in credit card data.
- It involves data preprocessing, handling imbalance using SMOTE, and splitting data for training/testing.
- Further steps would involve training models (e.g., Logistic Regression, Random Forest) and evaluating their performance.