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### **CREDIT CARD FRAUD DETECTION**

It is important that credit card companies are able to recognize fraudulent credit card transactions so that customers are not charged for items that they did not purchase. Credit card frauds are easy and friendly targets. E-commerce and many other online sites have increased the online payment modes, increasing the risk for online frauds. Increase in fraud rates, researchers started using different machine learning methods to detect and analyse frauds in online transactions.

There will be few challenges which would face while predicting the fraud as enormous data is processed every day and the model built must be fast enough to respond to the scam in time. Imbalanced data that is most of the transactions (99.8%) are not fraudulent which makes it hard for detecting the fraudulent ones. Data availability is also a challenge as the data is mostly private and misclassified data can be another major issue, as not every fraudulent transaction is caught and reported.

The dataset contains transactions made by credit cards in September 2013 by European cardholders. This dataset presents transactions that occurred in two days, where we have 492 frauds out of 284,807 transactions. The dataset is highly unbalanced, the positive class (frauds) account for 0.172% of all transactions. It contains only numerical input variables which are the result of a PCA transformation. Unfortunately, due to confidentiality issues, we cannot provide the original features and more background information about the data. Features V1, V2, ... V28 are the principal components obtained with PCA, the only features which have not been transformed with PCA are 'Time' and 'Amount'. Feature 'Time' contains the seconds elapsed between each transaction and the first transaction in the dataset. The feature 'Amount' is the transaction Amount, this feature can be used for example-dependant cost-sensitive learning. Feature 'Class' is the response variable and it takes value 1 in case of fraud and 0 otherwise.

The source data has been attached with it. These datasets are taken from kaggle.

Attributes from the data sets are as follows:

trans\_date\_trans\_time, cc\_num, merchant, category, amt, first, last, gender, street, city, state, zip, lat, long, city\_pop, job, dob, trans\_num, unix\_time, merch\_lat, merch\_long, is\_fraud.