

FRAUD DETECTION

STUDI INDEPENDEN MBKM 2022 FINAL PROJECT

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FRAUD DETECTION



Total Rows

640K++

Total Columns

29

Explanation

The Fraud Detection dataset consist of more than 640k credit account transaction records. Every record includes some detailed information, such as the user info, the transaction date, even the merchant info. It's also already comes with the fraud label to differentiate a fraud transaction with a non-fraud transaction.

Output Target

Find the variable that has the most influence to the fraud label (target variable) is the purpose of this project.

PROJECT BREAKDOWN









EXPLORATORY
DATA
ANALYSIS

DATA
PREPROCESSIN
G

DATA MODELLING

CONCLUSION





EXPLORATORY DATA ANALYSIS

MISSING VALUES

Six columns were removed because all the records are Null. For the other columns that contains Null, it's been calculated that the missing values was below 1% of total records so the null was replaced with the mode of each column.

OUTLIERS

Outliers was seen in numerical records from the transaction such as for creditLimit, availableMoney, transactionAmount, and currentBalance. Since the purpose of this project is to detect the fraud transaction, it's better to keep the outliers since it could be one of the crucial indication of fraud.

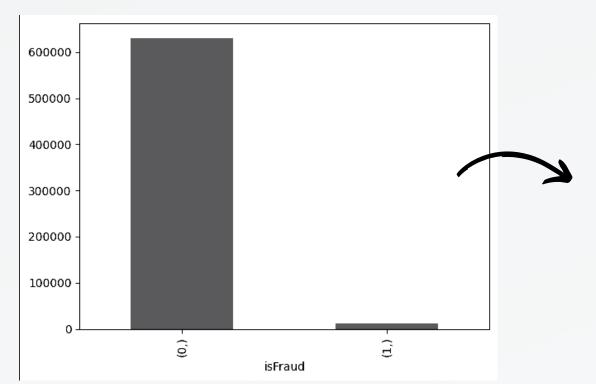


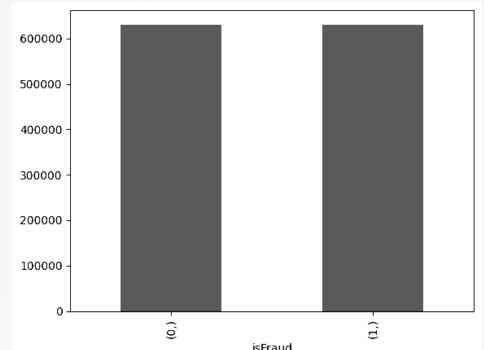


DATA PREPROCESSING

ENCODING

New column ['CVVNotMatch'] created to identify the transaction where the card CVV and the entered CVV was not match. In addition, columns with boolean datatypes has been changed to object datatypes so it's easier for data modelling.





BALANCING DATASET

The target variable, isFraud, has two values: 0 (not a fraud) and 1 (fraud], the number of fraud transaction is very small compared to the non-fraudulent ones. Using SMOTE, the imbalanced fixed in hopes the algorithm would predict better with balanced records between the fraud and non-fraud transaction.



DATA MODELLING



Decision Tree capture non-linear relationships and interactions between variables without any prior assumptions about the data structure.

DECISION TREE



Random Forest combine multiple decision trees, often leading to improve accuracy. This method reduce the risk of overfitting to the training data.

RANDOM FOREST



Logistic Regression model coefficients can be directly interpreted to understand the influence of each predictor variable on the outcome.

LOGISTIC REGRESSION





DECISION TREE

Accuracy	Specificity
93.77%	0.923



RANDOM FOREST

Accuracy	Specificity
96.09%	0.947



LOGISTIC REGRESSION

Accuracy	Specificity
69.52%	0.696

MODEL EVALUATION

Why Accuracy and Specificity?

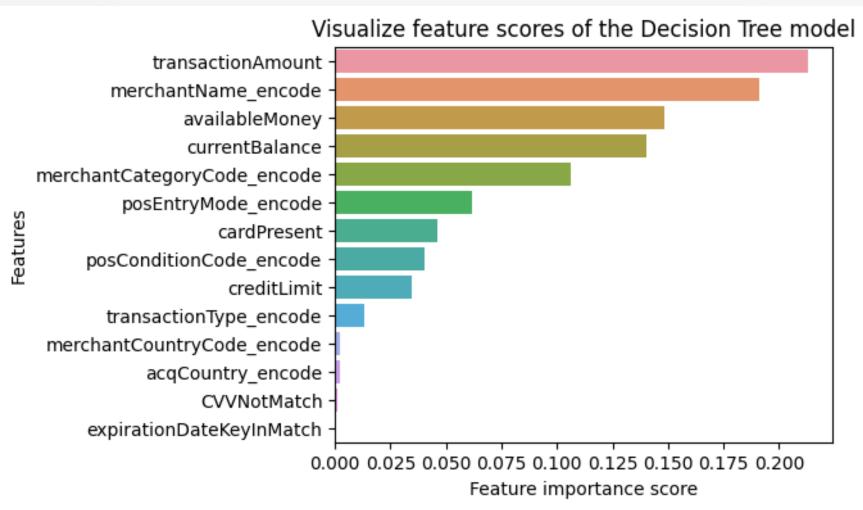
Specificity is important for imbalanced dataset since it's focused on the accuracy of negative values. In this project, specificity shows us how accurate it is to predict the fraud transaction one.

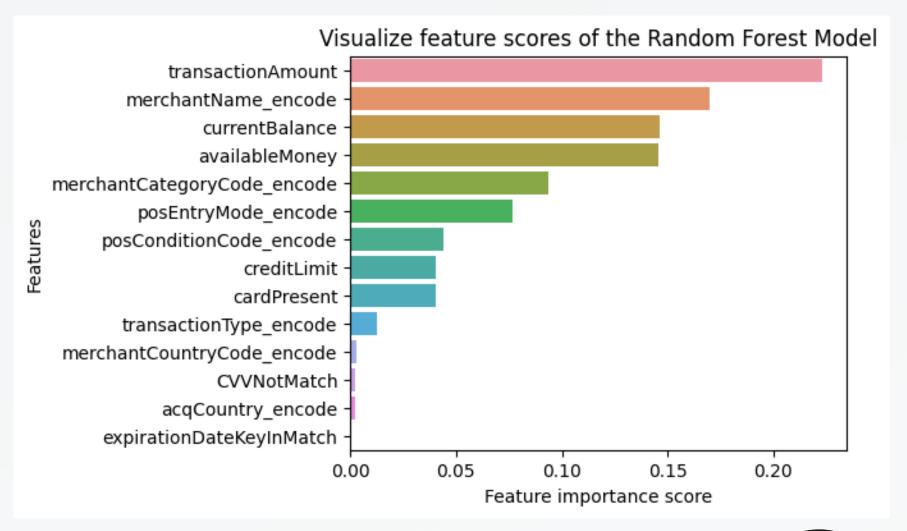
In other side, accuracy score shows us how accurate it is in predicting the fraud label for all of the transaction.

FEATURE IMPORTANCE SCORE



DECISION TREE













CONCLUSION VARIABLE THAT HAS THE MOST INFLUENCE TO THE TARGET VARIABLE

Transaction Amount

The result from decision tree and random forest modelling pointed out that Transaction Amount has the most influence for the target variable, isFraud. This means that transaction amount is a strong predictor of fraud and plays a crucial role in fraud detection models. It could also mean that higher transaction amounts might be more likely to be fraudulent.

THANK YOU!

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