** A**

**Assessment Report**

on

**“Predicting Credit Card Fraud”**

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**BACHELOR OF TECHNOLOGY**

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in

**Introduction To AI**

By

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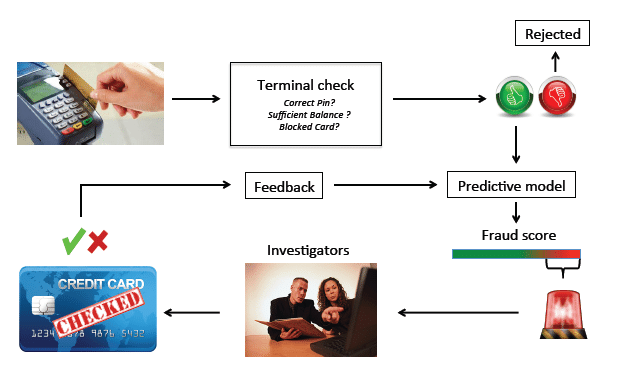
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**Introduction**

Detecting credit card fraud is a critical challenge in the financial sector, where billions are lost every year due to unauthorized transactions. The goal of this project is to build a machine learning classification model that accurately identifies fraudulent transactions by analysing patterns in user behaviour, transaction amount, location, and device usage. Since fraudulent activity is rare and often disguised among normal transactions, this problem involves handling highly imbalanced data and learning subtle anomalies. An effective solution helps banks and payment systems prevent financial loss, improve trust, and enhance security for customers by flagging suspicious transactions in real time.

**Methodology**

To detect fraudulent credit card transactions, we followed a structured machine learning pipeline:

1. **Data Collection & Upload:**  
   The dataset containing transaction records was uploaded and loaded into Google Colab.
2. **Data Preprocessing:**
   * Handled missing values by dropping incomplete rows.
   * Encoded categorical features using Label Encoding to convert them into numeric format.
   * Normalized or scaled features (if needed) to maintain consistency.
3. **Feature & Target Selection:**  
   Separated independent features (e.g., amount, location, device, behavior) and the target variable (Fraud or Not Fraud).
4. **Data Splitting:**  
   Split the dataset into training and testing sets using an 80-20 ratio, preserving class distribution using stratification.
5. **Model Building:**  
   Trained a **Random Forest Classifier**, a robust ensemble algorithm known for handling imbalanced datasets and high-dimensional features.
6. **Model Evaluation:**  
   Evaluated the model using **accuracy**, **precision**, **recall**, and a **confusion matrix heatmap** to visualize correct and incorrect classifications.
7. **Visualization:**  
   Visual tools such as confusion matrix plots were used to better understand model performance and detect any misclassification trends.

**Code Used**

from google.colab import files

uploaded = files.upload()

import pandas as pd

df = pd.read\_csv('creditcard.csv')  # If the file name was different, update it!

df.head()

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.preprocessing import StandardScaler

from sklearn.model\_selection import train\_test\_split

from sklearn.ensemble import RandomForestClassifier

from sklearn.metrics import classification\_report, confusion\_matrix

from imblearn.over\_sampling import SMOTE

# Load dataset

# Feature Scaling for 'Amount' and 'Time'

scaler = StandardScaler()

df['scaled\_amount'] = scaler.fit\_transform(df[['Amount']])

df['scaled\_time'] = scaler.fit\_transform(df[['Time']])

df.drop(['Amount', 'Time'], axis=1, inplace=True)

# Rearranging columns

scaled\_amount = df['scaled\_amount']

scaled\_time = df['scaled\_time']

df.drop(['scaled\_amount', 'scaled\_time'], axis=1, inplace=True)

df.insert(0, 'scaled\_amount', scaled\_amount)

df.insert(1, 'scaled\_time', scaled\_time)

# Splitting features and target

X = df.drop('Class', axis=1)

y = df['Class']

# SMOTE for balancing the classes

sm = SMOTE(random\_state=42)

X\_res, y\_res = sm.fit\_resample(X, y)

# Train-Test Split

X\_train, X\_test, y\_train, y\_test = train\_test\_split(X\_res, y\_res, test\_size=0.3, random\_state=42)

# Model Training

model = RandomForestClassifier(n\_estimators=100, random\_state=42)

model.fit(X\_train, y\_train)

# Prediction and Evaluation

y\_pred = model.predict(X\_test)

print("Classification Report:\n")

print(classification\_report(y\_test, y\_pred))

# Confusion Matrix

plt.figure(figsize=(6,4))

sns.heatmap(confusion\_matrix(y\_test, y\_pred), annot=True, fmt='d', cmap='Blues')

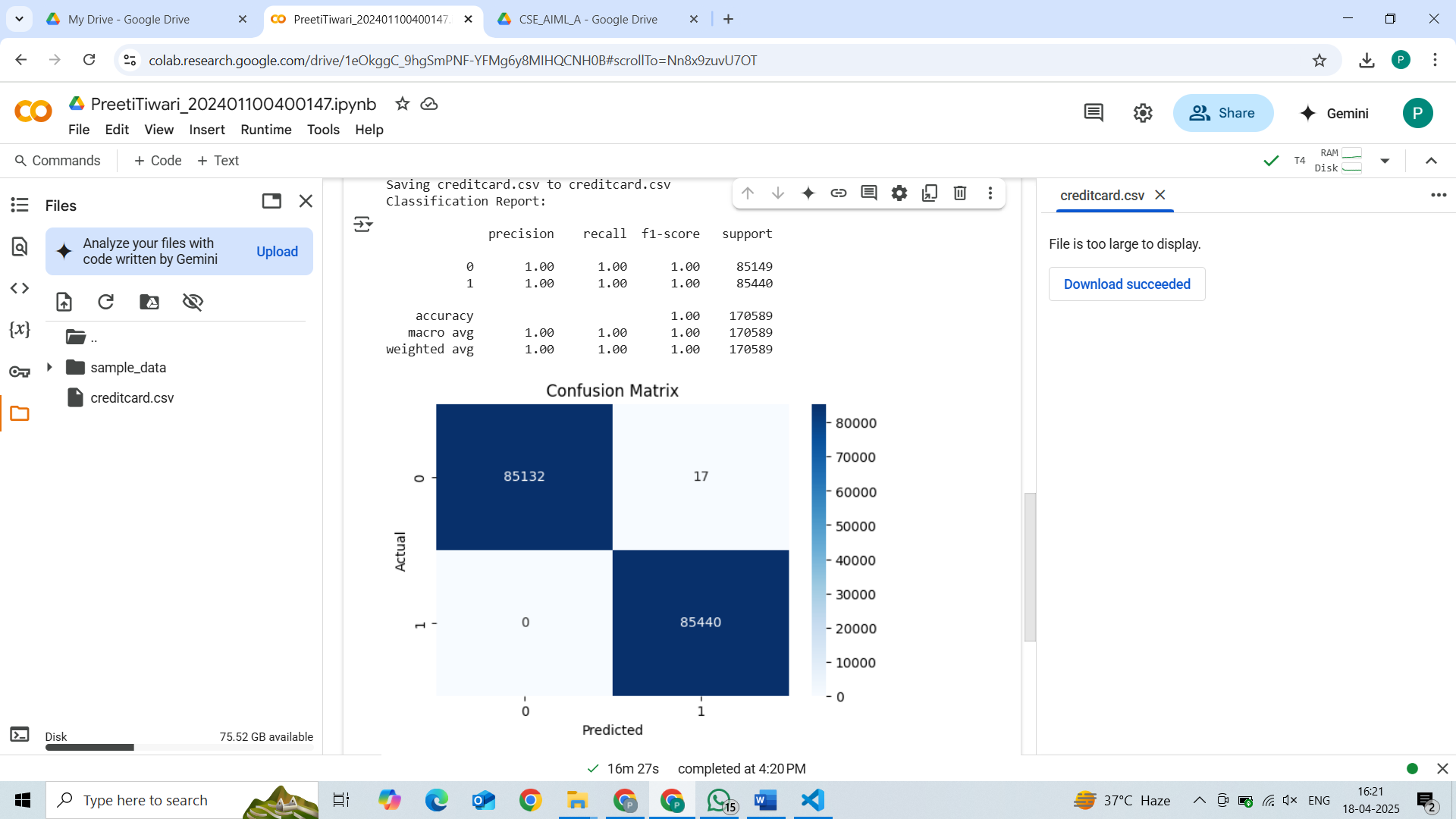
plt.title("Confusion Matrix")

plt.xlabel("Predicted")

plt.ylabel("Actual")

plt.show()

**Output**

 **Refrences**

* Used ChatGPT for basic Understanding.
* Dataset of credit card prediction taken from Kaggle.
* Used image from google images.
* Used google colaboratory for running the program.