INTERNSHIP PROJECT REPORT

CREDIT CARD FRAUD DETECTION USING ML

Name of Project: Credit Card Fraud Detection using ML

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Branch: CSE – DATA SCIENCE

Project Submission Date: 11/09/2023

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- A project to develop a machine learning model to detect fraudulent credit card transactions.
- There are a number of machine learning techniques that can be used to detect credit card fraud. Some of the most common techniques include:
- Logistic regression
- Support vector machines
- Decision trees
- Random forests
- Neural networks

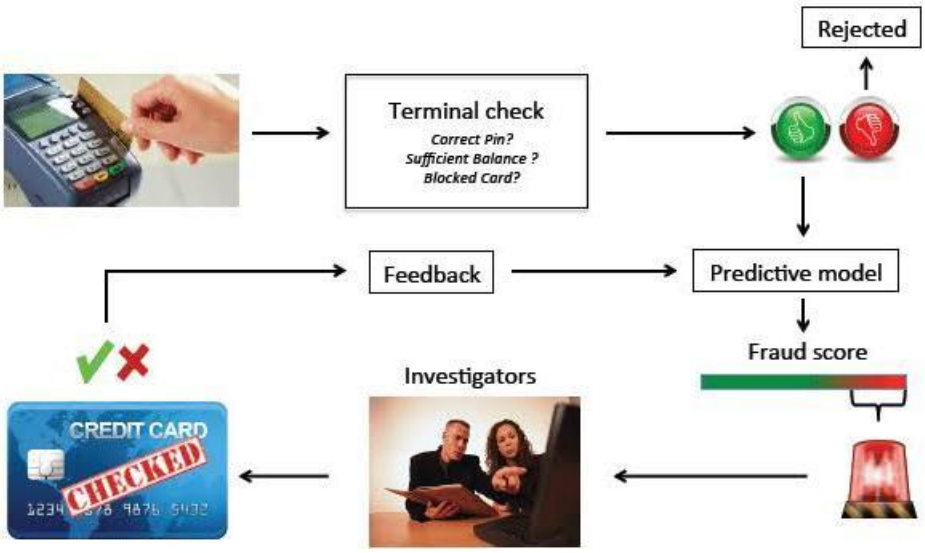


Figure: Fraud Detection Process

List of Software:

- Python Based Software
- Example : - Anaconda , Jupyter Notebook , Google Colab etc.

List of Hardware:

- OS – Windows 7, 8 , 10 and 11 (32 and 64 bit)
- RAM – 4GB

Basically, there are five steps in Credit Card Fraud Detection process
Step 1: Dataset (Credit Card Data)
Step 2: Data Pre-Processing
Step 3: Data Analysis
Step 4: Train Test split
Step 5: Logistics Regression Model
Step 6: Evaluation

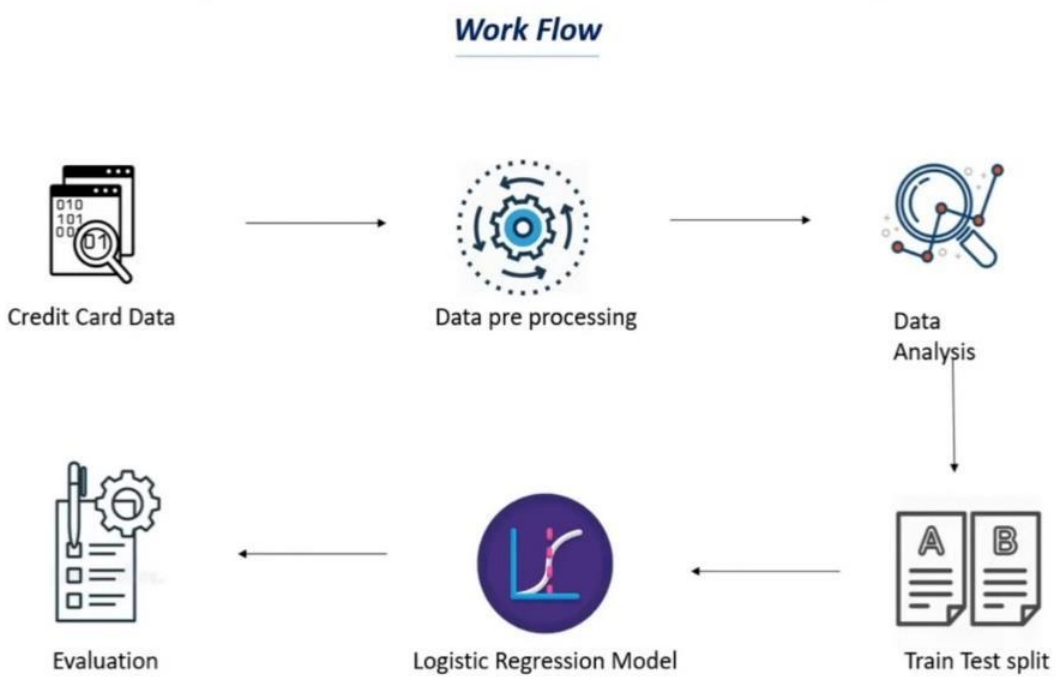


Figure: Work Flow of System

PURPOSE OF THE PROJECT

We propose a Machine learning model to detect fraudulent credit card activities in online financial transactions.

4.1 Overview

We propose a Machine learning model to detect fraudulent credit card activities in online financial transactions.Analyzing fake transactions manually is impracticable due to vast amounts of data and its complexity. However, adequately given informative features, could make it is possible using Machine Learning. This hypothesis will be explored in the project.
To classify fraudulent and legitimate credit card transaction by supervised learning Algorithm such as Logistic Regression . To help us to get awareness about the fraudulent and without loss of any financially.

4.2 Dataset Description

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.

Proposed Method

The proposed techniques emphasizes on detecting Credit Card Fraudulent transactions whether it is a genuine/nonfraud or a fraud transaction and the approaches used to separate fraud and non-fraud are KNN, Decision Tree, Logistic regression, Random forest and Finally we will detect credit card frauds.

The system architecture has following steps:

1. Import of Necessary Packages

2.Read the Dataset

3.Exploratory Data Analysis i.e. finding null values, duplicate values etc.

4.Selecting Features (X) and the Target (y) columns

5.Train Test Split will split the whole dataset into train and test data

6.Build the model i.e. Training the model

7.Test the model i.e. Model prediction

8. Evaluation of the system i.e. Accuracy score, F1- score etc.

Importing the Dependencies

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
# loading the dataset to a Pandas DataFrame
credit_card_data = pd.read_csv('creditcard.csv')

# first 5 rows of the dataset
credit_card_data.head()

credit_card_data.tail()

# dataset informations
credit_card_data.info()

# checking the number of missing values in each column
credit_card_data.isnull().sum()

# distribution of legit transactions & fraudulent transactions
credit_card_data['Class'].value_counts()

# separating the data for analysis
legit = credit_card_data[credit_card_data.Class == 0]
fraud = credit_card_data[credit_card_data.Class == 1]

print(legit.shape)
print(fraud.shape)

# statistical measures of the data
legit.Amount.describe()

fraud.Amount.describe()

# compare the values for both transactions
credit_card_data.groupby('Class').mean()
```

Under Sampling

Build a sample dataset containing similar distribution of normal transactions and Fraudulent Transactions

Number of Fraudulent Transactions --> 492

```
legit_sample = legit.sample(n=492)
```

Concatenating two DataFrames

```
new_dataset = pd.concat([legit_sample, fraud], axis=0)
new_dataset.head()

new_dataset.tail()

new_dataset['Class'].value_counts()

new_dataset.groupby('Class').mean()
```

Splitting Dataset into Features and Targets

```
X = new_dataset.drop(columns='Class', axis=1)
Y = new_dataset['Class']

print(X)

print(Y)
```

Split the data into Training data and Testing data

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, random_state=2)

print(X.shape, X_train.shape, X_test.shape)
```

Model Training

Logistic Regression

```
model = LogisticRegression()

# training the Logistic Regression Model with Training Data
model.fit(X_train, Y_train)
```

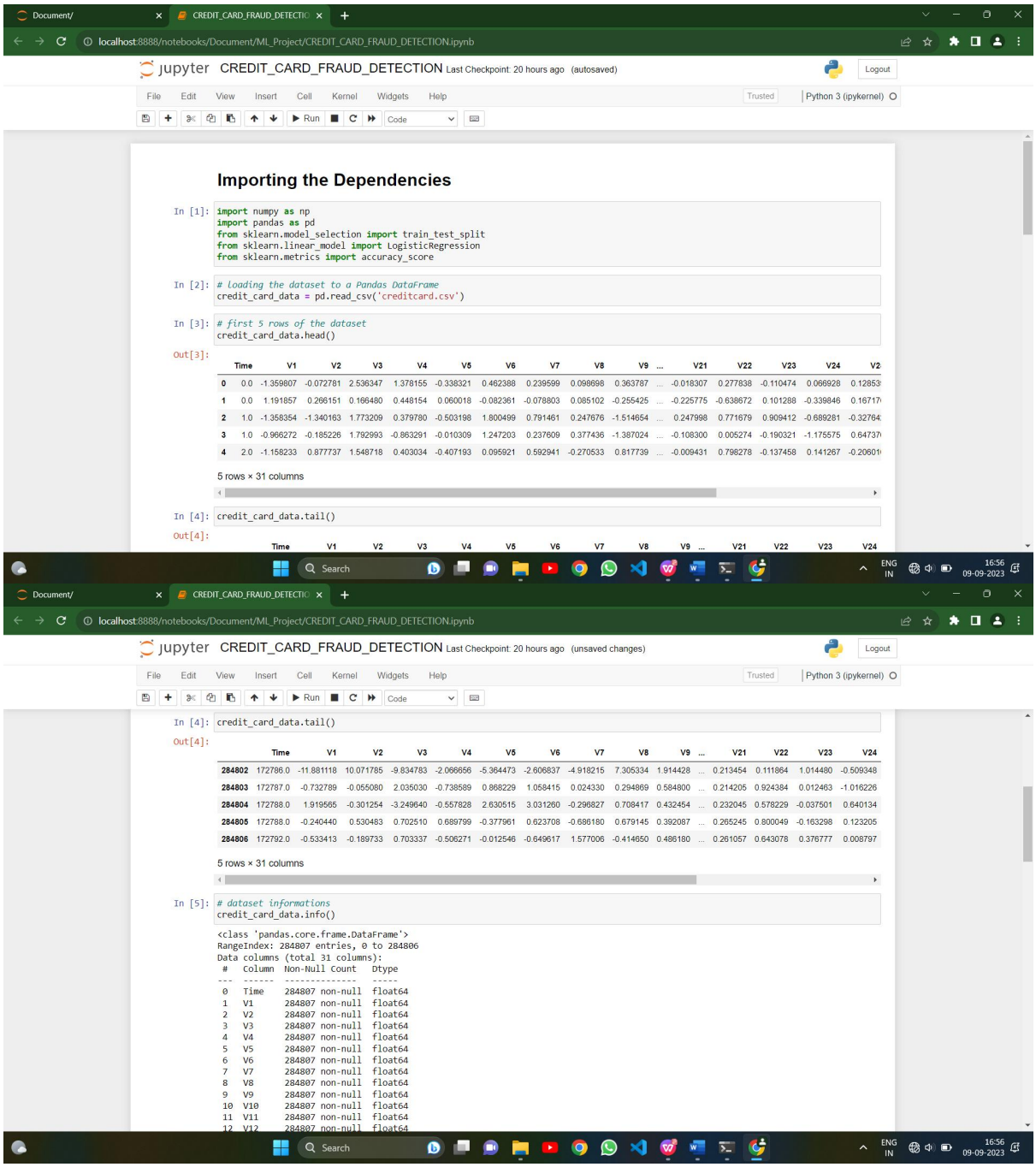
Model Evaluation

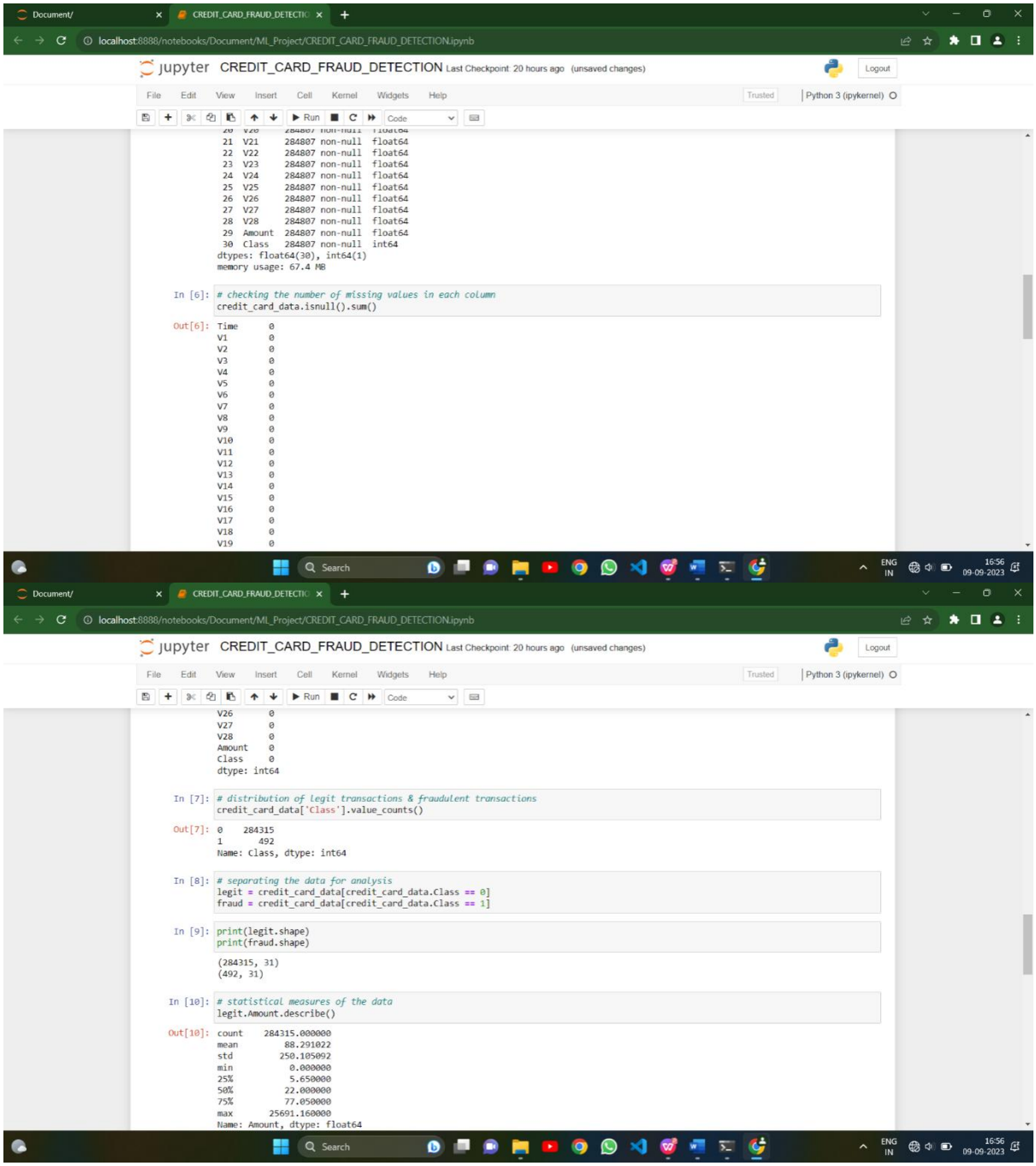
Accuracy Score

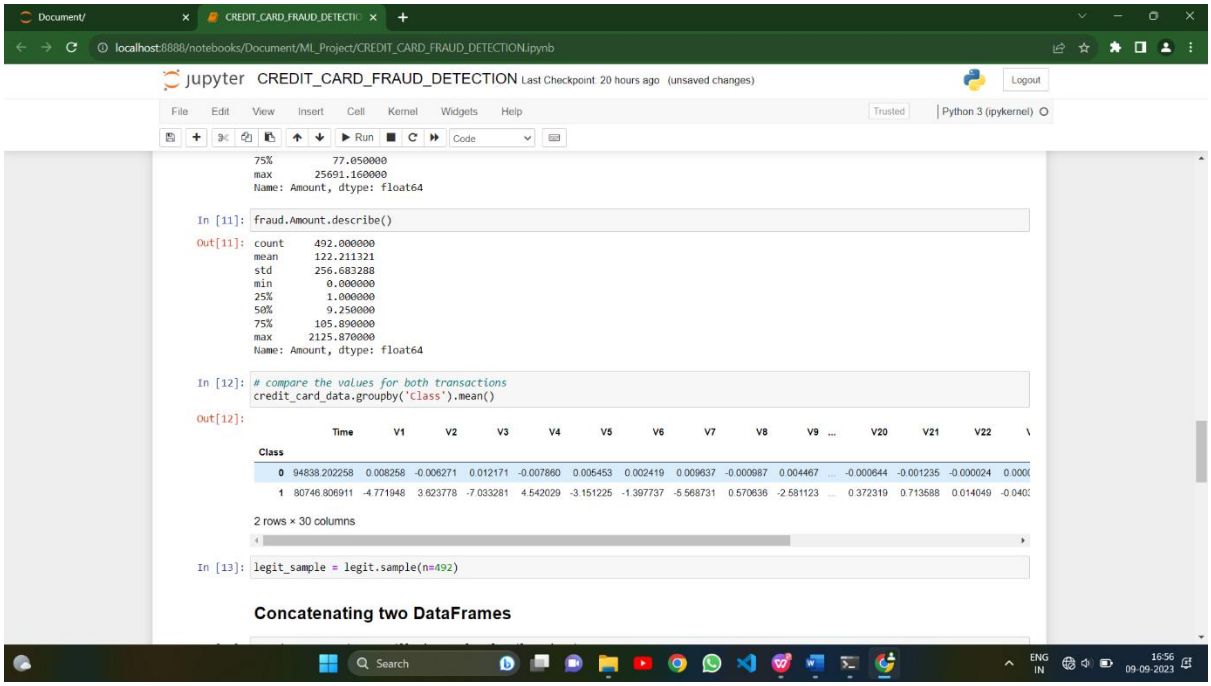
```
# accuracy on training data
X_train_prediction = model.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)

print('Accuracy on Training data : ', training_data_accuracy)
```

```
# accuracy on test data
X_test_prediction = model.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)

print('Accuracy score on Test Data : ', test_data_accuracy)
```

IMPLIMENTATION AND SCREENSHOTS AND IMAGES OF THE RUNNING PROJECTS




jupyter CREDIT_CARD_FRAUD_DETECTION Last Checkpoint: 20 hours ago (unsaved changes)
File Edit View Insert Cell Kernel Widgets Help
Python 3 (ipykernel)
21 V21 284807 non-null float64
22 V22 284807 non-null float64
23 V23 284807 non-null float64
24 V24 284807 non-null float64
25 V25 284807 non-null float64
26 V26 284807 non-null float64
27 V27 284807 non-null float64
28 V28 284807 non-null float64
29 Amount 284807 non-null float64
30 Class 284807 non-null int64
dtypes: float64(30), int64(1)
memory usage: 67.4 MB
In [6]: # checking the number of missing values in each column
credit_card_data.isnull().sum()
Out[6]: Time 0
V1 0
V2 0
V3 0
V4 0
V5 0
V6 0
V7 0
V8 0
V9 0
V10 0
V11 0
V12 0
V13 0
V14 0
V15 0
V16 0
V17 0
V18 0
V19 0
V26 0
V27 0
V28 0
Amount 0
Class 0
dtype: int64
In [7]: # distribution of legit transactions & fraudulent transactions
credit_card_data['Class'].value_counts()
Out[7]: 0 284315
1 492
Name: Class, dtype: int64
In [8]: # separating the data for analysis
legit = credit_card_data[credit_card_data.Class == 0]
fraud = credit_card_data[credit_card_data.Class == 1]
In [9]: print(legit.shape)
print(fraud.shape)
(284315, 31)
(492, 31)
In [10]: # statistical measures of the data
legit.Amount.describe()
Out[10]: count 284315.000000
mean 88.291022
std 250.105092
min 0.000000
25% 5.650000
50% 22.000000
75% 77.050000
max 25691.160000
Name: Amount, dtype: float64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File Edit View Insert Cell Kernel Widgets Help
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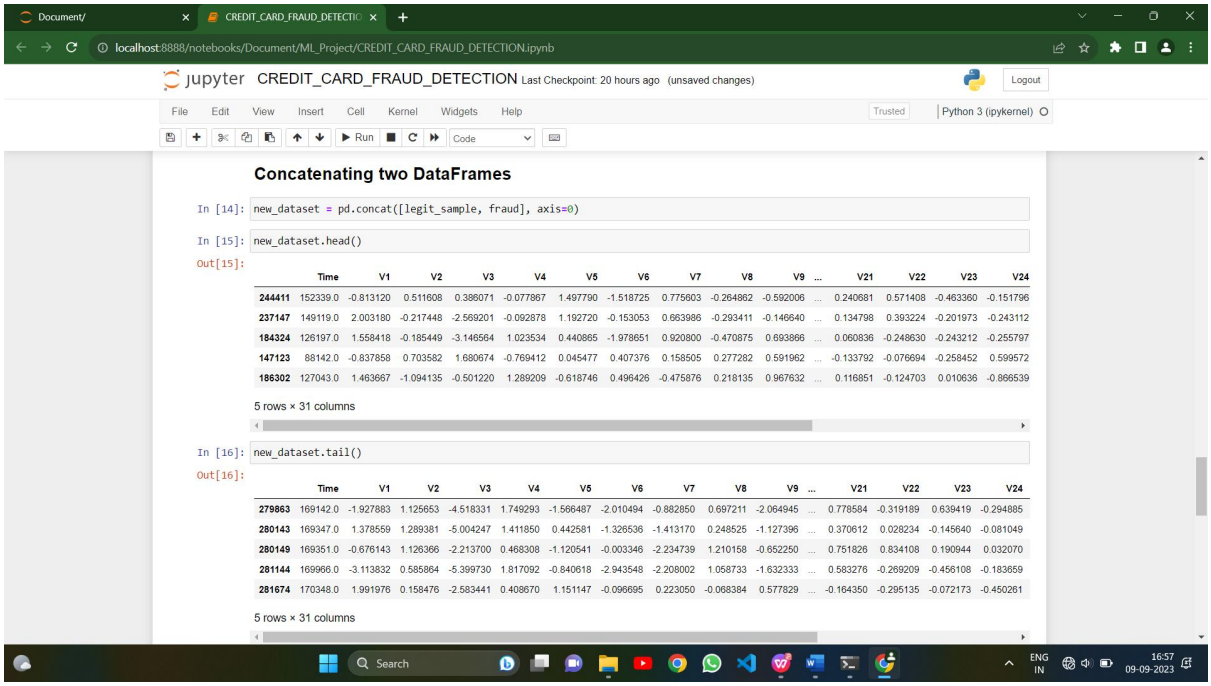
Python 3 (ipykernel)
Run
Code
75% 77.050000
max 25691.160000
Name: Amount, dtype: float64
In [11]: fraud.Amount.describe()
Out[11]: count 492.000000
mean 122.211321
std 256.683288
min 0.000000
25% 1.000000
50% 9.250000
75% 105.890000
max 2125.870000
Name: Amount, dtype: float64
In [12]: # compare the values for both transactions
credit_card_data.groupby('Class').mean()
Out[12]:
Time V1 V2 V3 V4 V5 V6 V7 V8 V9 ... V20 V21 V22
Class
0 94838.202258 0.008258 -0.006271 0.012171 -0.007860 0.005453 0.002419 0.009637 -0.000987 0.004467 ... -0.000644 -0.001235 -0.000024
1 80746.806911 -4.771948 3.623778 -7.033281 4.542029 -3.151225 -1.397737 -5.568731 0.570636 -2.581123 ... 0.372319 0.713588 0.014049
2 rows × 30 columns
In [13]: legit_sample = legit.sample(n=492)
Concatenating two DataFrames
Search
ENG IN
16:56 09-09-2023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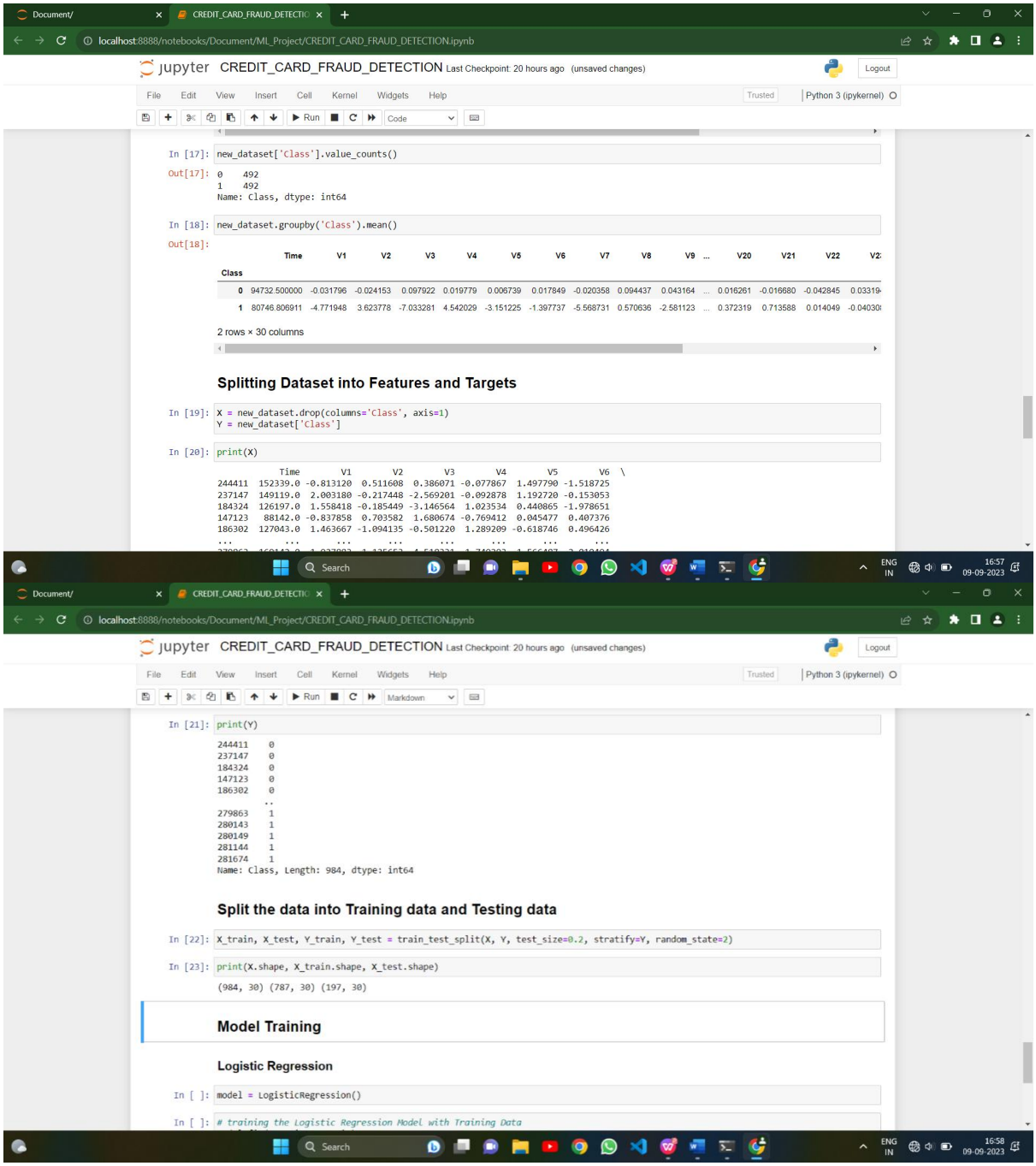
File Edit View Insert Cell Kernel Widgets Help
Trusted
Python 3 (ipykernel)
Run
Code
Concatenating two DataFrames
In [14]: new_dataset = pd.concat([legit_sample, fraud], axis=0)
In [15]: new_dataset.head()
Out[15]:
Time V1 V2 V3 V4 V5 V6 V7 V8 V9 ... V21 V22 V23 V24
244411 152339.0 -0.813120 0.511608 0.386071 -0.077867 1.497790 -1.518725 0.775603 -0.264862 -0.592006 ... 0.240681 0.571408 -0.463360 -0.151796
237147 149119.0 2.003180 -0.217448 -2.569201 -0.092878 1.192720 -0.153053 0.663986 -0.293411 -0.146640 ... 0.134798 0.393224 -0.201973 -0.243112
184324 126197.0 1.558418 -0.185449 -3.146564 1.023534 0.440865 -1.978651 0.920800 -0.470875 0.693866 ... 0.060836 -0.248630 -0.243212 -0.255797
147123 88142.0 -0.837858 0.703582 1.680674 -0.769412 0.045477 0.407376 0.158505 0.277282 0.591962 ... -0.133792 -0.076694 -0.258452 0.599572
186302 127043.0 1.463667 -1.094135 -0.501220 1.289209 -0.618746 0.496426 -0.475876 0.218135 0.967632 ... 0.116851 -0.124703 0.010636 -0.866539
5 rows × 31 columns
In [16]: new_dataset.tail()
Out[16]:
Time V1 V2 V3 V4 V5 V6 V7 V8 V9 ... V21 V22 V23 V24
279863 169142.0 -1.927883 1.125653 -4.518331 1.749293 -1.566487 -2.010494 -0.882850 0.697211 -2.064945 ... 0.778584 -0.319189 0.639419 -0.294885
280143 169347.0 1.378559 1.289381 -5.004247 1.411850 0.442581 -1.326536 -1.413170 0.248525 -1.127396 ... 0.370612 0.028234 -0.145640 -0.081049
280149 169351.0 -0.676143 1.126366 -2.213700 0.468308 -1.120541 -0.003346 -2.234739 1.210158 -0.652250 ... 0.751826 0.834108 0.190944 0.032070
281144 169966.0 -3.113832 0.585864 -5.399730 1.817092 -0.840618 -2.943548 -2.208002 1.058733 -1.632333 ... 0.583276 -0.269209 -0.456108 -0.183659
281674 170348.0 1.991976 0.158476 -2.583441 0.408670 1.151147 -0.096695 0.223050 -0.068384 0.577829 ... -0.164350 -0.295135 -0.072173 -0.450261
5 rows × 31 columns
Search
ENG IN
16:57 09-09-2023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Logout
File Edit View Insert Cell Kernel Widgets Help
Trusted
Python 3 (ipykernel)
Run
Code
In [17]: new_dataset['Class'].value_counts()
Out[17]: 0 492
1 492
Name: Class, dtype: int64
In [18]: new_dataset.groupby('Class').mean()
Out[18]:
Time V1 V2 V3 V4 V5 V6 V7 V8 V9 ... V20 V21 V22 V23
Class
0 94732.500000 -0.031796 -0.024153 0.097922 0.019779 0.006739 0.017849 -0.020358 0.094437 0.043164 ... 0.016261 -0.016680 -0.042845 0.03319
1 80746.806911 -4.771948 3.623778 -7.033281 4.542029 -3.151225 -1.397737 -5.568731 0.570636 -2.581123 ... 0.372319 0.713588 0.014049 -0.04030
2 rows × 30 columns
Splitting Dataset into Features and Targets
In [19]: X = new_dataset.drop(columns='Class', axis=1)
Y = new_dataset['Class']
In [20]: print(X)
Time V1 V2 V3 V4 V5 V6 \
244411 152339.0 -0.813120 0.511608 0.386071 -0.077867 1.497790 -1.518725
237147 149119.0 2.003180 -0.217448 -2.569201 -0.092878 1.192720 -0.153053
184324 126197.0 1.558418 -0.185449 -3.146564 1.023534 0.440865 -1.978651
147123 88142.0 -0.837858 0.703582 1.680674 -0.769412 0.045477 0.407376
186302 127043.0 1.463667 -1.094135 -0.501220 1.289209 -0.618746 0.496426
Search
ENG IN
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File Edit View Insert Cell Kernel Widgets Help
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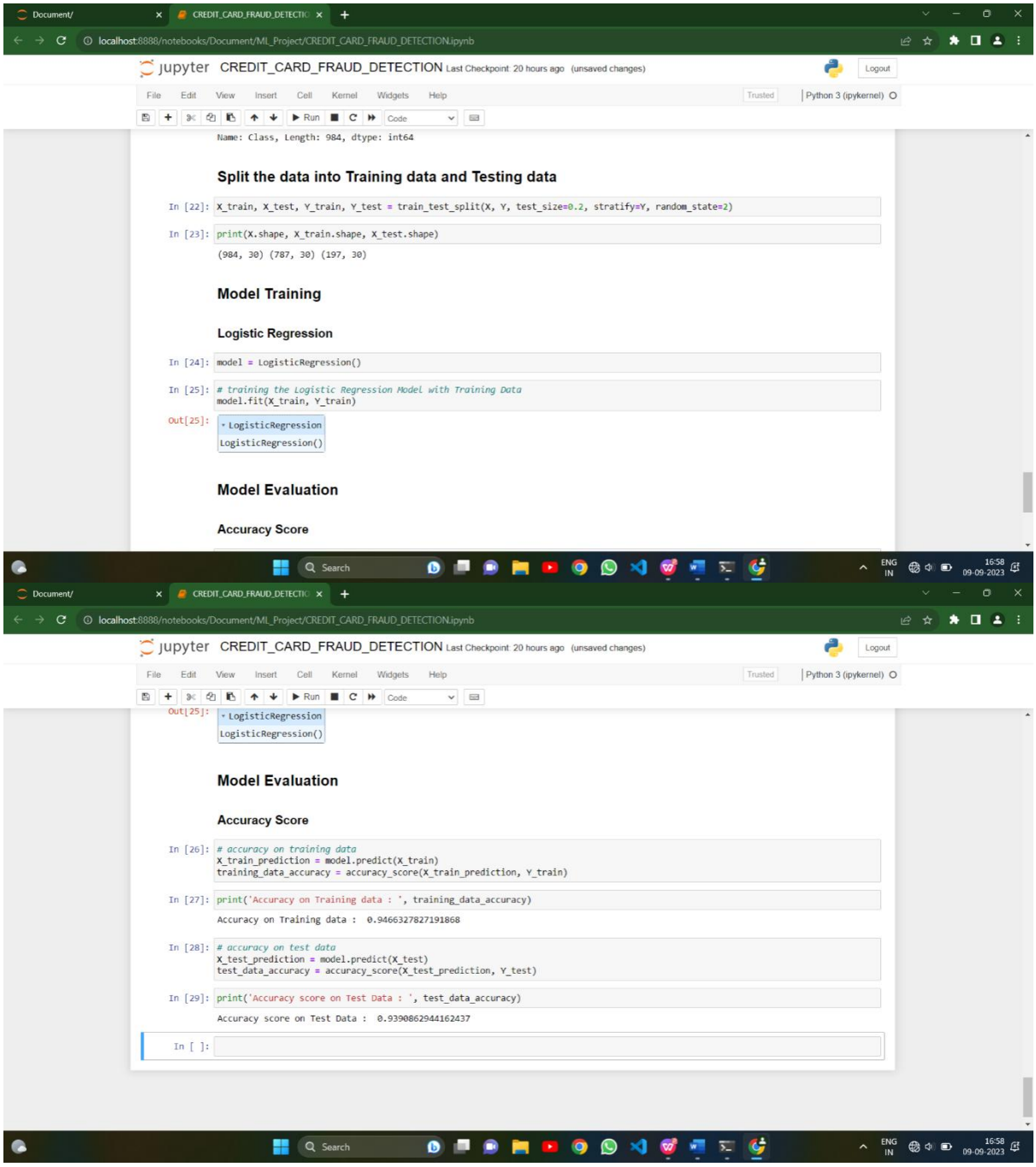
Python 3 (ipykernel)
Run
Markdown
In [21]: print(Y)
244411 0
237147 0
184324 0
147123 0
186302 0
..
279863 1
280143 1
280149 1
281144 1
281674 1
Name: Class, Length: 984, dtype: int64
Split the data into Training data and Testing data
In [22]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, random_state=2)
In [23]: print(X.shape, X_train.shape, X_test.shape)
(984, 30) (787, 30) (197, 30)
Model Training
Logistic Regression
In [ ]: model = LogisticRegression()
In [ ]: # training the Logistic Regression Model with Training Data
Search
ENG IN
16:58
09-09-2023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jupyter CREDIT_CARD_FRAUD_DETECTION Last Checkpoint: 20 hours ago (unsaved changes)
Logout
File Edit View Insert Cell Kernel Widgets Help
Trusted
Python 3 (ipykernel)
Run
Code
Name: Class, Length: 984, dtype: int64
Split the data into Training data and Testing data
In [22]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, stratify=Y, random_state=2)
In [23]: print(X.shape, X_train.shape, X_test.shape)
(984, 30) (787, 30) (197, 30)
Model Training
Logistic Regression
In [24]: model = LogisticRegression()
In [25]: # training the Logistic Regression Model with Training Data
model.fit(X_train, Y_train)
Out[25]: LogisticRegression()
LogisticRegression()
Model Evaluation
Accuracy Score
Search
ENG IN
16:58 09-09-2023
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CREDIT_CARD_FRAUD_DETECTIO
localhost:8888/notebooks/Document/ML_Project/CREDIT_CARD_FRAUD_DETECTION.ipynb
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Logout
File Edit View Insert Cell Kernel Widgets Help
Trusted
Python 3 (ipykernel)
Run
Code
Out[25]: LogisticRegression()
LogisticRegression()
Model Evaluation
Accuracy Score
In [26]: # accuracy on training data
X_train_prediction = model.predict(X_train)
training_data_accuracy = accuracy_score(X_train_prediction, Y_train)
In [27]: print('Accuracy on Training data : ', training_data_accuracy)
Accuracy on Training data : 0.9466327827191868
In [28]: # accuracy on test data
X_test_prediction = model.predict(X_test)
test_data_accuracy = accuracy_score(X_test_prediction, Y_test)
In [29]: print('Accuracy score on Test Data : ', test_data_accuracy)
Accuracy score on Test Data : 0.9390862944162437
In [ ]:
Search
ENG IN
16:58 09-09-2023

[1] Machine Learning Group — ULB, Credit Card Fraud Detection (2018), Kaggle.

[2] R. J. Bolton and D. J. Hand. Unsupervised profiling methods for fraud detection. In conference of Credit Scoring and Credit Connol VII, Edinburgh. UK, Sept 5-7,2001.

[3] Quah, J. T. S., and Sriganesh, M. (2020). Real-time credit card fraud detection using computational intelligence. Expert Systems with Applications, 35(4), 1721-1732.