#### Task 5

# Credit\_Card\_Fraud\_Detection

### Importing the libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

#### Importing the dataset

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```
dataset = pd.read_csv('Credit_card.csv')
```

#### dataset.head(3)

<b>→</b>		Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	• • •	V21	
	0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0.363787		-0.018307	(
	1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0.255425		-0.225775	-(
	2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1.514654		0.247998	C

3 rows × 31 columns

dataset.shape

**→** (171585, 31)

16 V16

dataset.info()

<del>_</del>	<clas< th=""><th>ss 'panda</th><th>s.core</th><th>.frame.Data</th><th>aFrame'&gt;</th></clas<>	ss 'panda	s.core	.frame.Data	aFrame'>
	Range	eIndex: 1	L71585 e	entries, 0	to 171584
	Data	columns	(total	31 columns	5):
	#	Column	Non-Nu	ll Count	Dtype
	0	Time	171585	non-null	float64
	1	V1	171585	non-null	float64
	2	V2	171585	non-null	float64
	3	V3	171585	non-null	float64
	4	V4	171585	non-null	float64
	5	V5	171585	non-null	float64
	6	V6	171585	non-null	float64
	7	V7	171585	non-null	float64
	8	V8	171585	non-null	float64
	9	V9	171585	non-null	float64
	10	V10	171585	non-null	float64
	11	V11	171585	non-null	float64
	12	V12	171585	non-null	float64
	13	V13	171585	non-null	float64
	14	V14	171585	non-null	float64
	15	V15	171584	non-null	float64

171584 non-null float64

17	V17	171584	non-null	float64					
18	V18	171584	non-null	float64					
19	V19	171584	non-null	float64					
20	V20	171584	non-null	float64					
21	V21	171584	non-null	float64					
22	V22	171584	non-null	float64					
23	V23	171584	non-null	float64					
24	V24	171584	non-null	float64					
25	V25	171584	non-null	float64					
26	V26	171584	non-null	float64					
27	V27	171584	non-null	float64					
28	V28	171584	non-null	float64					
29	Amount	171584	non-null	float64					
30	Class	171584	non-null	float64					
dtyp	dtypes: float64(31)								

dtypes: float64(31)
memory usage: 40.6 MB

dataset.isnull().sum()

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```
\overline{2}
                                        0
```

Time	0
V1	0

V2 0

٧3 0

**V**4 0

V5 0

V6 0

**V7** 0

**V**8 0

V9 0

V10 0

V11 0

V12 0

V13 0

V14 0

V15 0

V16 0

V17 0

V18 0

V19 0

V20 0

V21 0

V22 0

V23 0

V24 0

V25 0

V26 0

V27 0

V28 0

Amount 0

Class 0

dtype: int64

from sklearn.preprocessing import StandardScaler sc=StandardScaler()

dataset['Amount']=sc.fit\_transform(pd.DataFrame(dataset['Amount']))

dataset.head()

<del>_</del>		Time	V1	V2	V3	V4	V5	V6	V7	V8	V9	• • •	V21
	0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.098698	0.363787		-0.018307
	1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.085102	-0.255425		-0.225775
	2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.247676	-1.514654		0.247998
	3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.377436	-1.387024		-0.108300
	4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.270533	0.817739		-0.009431

5 rows × 31 columns

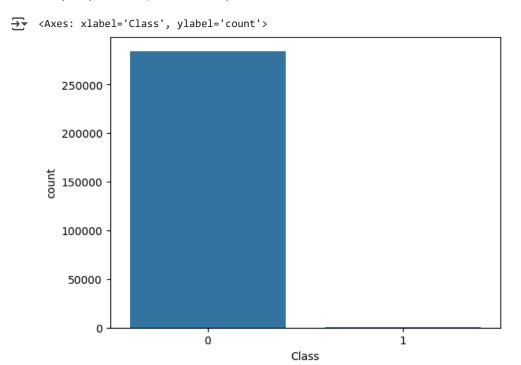
data=dataset.duplicated().any()

data.shape

<del>\_</del> → ()

import seaborn as sns
sns.countplot(x='Class',data=dataset)

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# Splitting the dataset into the Training set and Test set

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.20, random_state = 0)
print(X_train)
₹ [[ 5.31500000e+04 -1.11504743e+00 1.03558276e+00 ... 2.29736936e-01
       -7.05913040e-02 -3.01454177e-01]
      [ 4.00600000e+04 1.22847256e+00 -1.38826483e-01 ... -6.91781060e-02
        1.89006100e-03 -2.50998329e-01]
      [ 1.54865000e+05 1.92614789e+00 -1.98627962e-01 ... 5.27473010e-02
       -3.26215510e-02 -3.25642598e-01]
      [ 7.66160000e+04 1.51260229e+00 -9.49435315e-01 ... -1.41537100e-03
        3.66494400e-03 -2.13696185e-01]
      [ 9.72530000e+04 1.79886333e+00 -1.69979073e+00 ... -3.36155800e-02
       -3.24705130e-02 3.31682753e-01]
      [ 7.48870000e+04 -5.89399721e-01 7.47828393e-01 ... 8.65917860e-02
        1.18083774e-01 -2.89299995e-01]]
print(y_train)
                                             Run cell (Ctrl+Enter)
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<del>→•</del> [000...000]
print(X_test)
₹ [[ 1.25821000e+05 -3.23333572e-01 1.05745525e+00 ... 1.08494430e-01
        1.61139167e-01 -1.93305945e-01]
      [ 1.57235000e+05 -3.49718405e-01 9.32618570e-01 ... 7.68300270e-02
        1.75561960e-01 -3.45313182e-01]
      [ 1.52471000e+05 -1.61471082e+00 -2.40656973e+00 ... 2.86285101e-01
        4.37321868e-01 3.05868817e-02]
      [ 1.37149000e+05 1.12540205e+00 -2.28899827e+00 ... -2.40867448e-01
       -7.49435200e-03 1.97849446e+00]
      [ 1.60893000e+05 2.06485724e+00 2.85198054e-01 ... -1.03162057e-01
       -6.17434560e-02 -3.41435039e-01]
      [ 1.53086000e+05 2.35138178e+00 -1.32522606e+00 ... 7.26919800e-03
       -5.16343390e-02 -2.93258100e-01]]
print(y_test)
<del>→</del> [000...000]
Feature Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
print(X_train)
→ [[-0.87750986 -0.57177938 0.62828369 ... 0.57377804 -0.2196612
       -0.30511629]
```

[-1.15308534 0.62679248 -0.08362249 ... -0.17091545 0.00631403

[ 1.26383162 0.98361214 -0.11987309 ... 0.13283995 -0.10128286

 $[ \ 0.05096267 \ \ 0.91851362 \ -1.02985151 \ \dots \ -0.08231776 \ -0.10081197$ 

-0.25423683]

-0.32950779]

-0.216621511

```
0.33333622]
    [-0.41989457 \ -0.30294167 \ \ 0.45385207 \ \dots \ \ 0.21715744 \ \ 0.36857129
     -0.29286006]]
print(X_test)
₹ [[ 0.65238668 -0.16686457 0.64154241 ... 0.27172397 0.50280513
     -0.19606009]
    [ 1.31372573 -0.18035885  0.56586861 ...  0.19283776  0.54777109
     -0.34934352]
    [ 1.21343225 -0.82732765 -1.45828848 ... 0.7146577
                                               1.36386022
      0.02971246]
    [ 0.89086789  0.57407801 -1.38701874 ... -0.5986488  -0.02294377
      1.99397392]
    -0.34543281]
    -0.2968514 ]]
```

▼ Training the Logistic Regressic Run cell (Ctrl+Enter) cell has not been executed in this session
 | set

#### Predicting a new result

### Predicting the Test set results

```
y_pred = classifier.predict(X_test)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))

[[0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
```

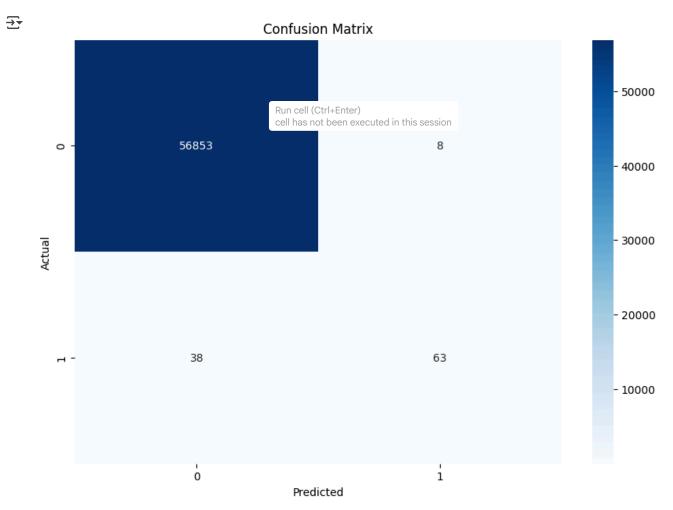
### Making the Confusion Matrix

```
from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)
```

```
[[56853 8]
[ 38 63]]
0.9991924440855307
```

Model has 99.91% Accuracy.

```
plt.figure(figsize=(10,7))
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()
```



```
from sklearn.metrics import accuracy_score
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
from sklearn.metrics import f1_score
accuracy_score(y_test,y_pred)
```

→ 0.9991924440855307

Start coding or generate with AI.

#### Random Forest

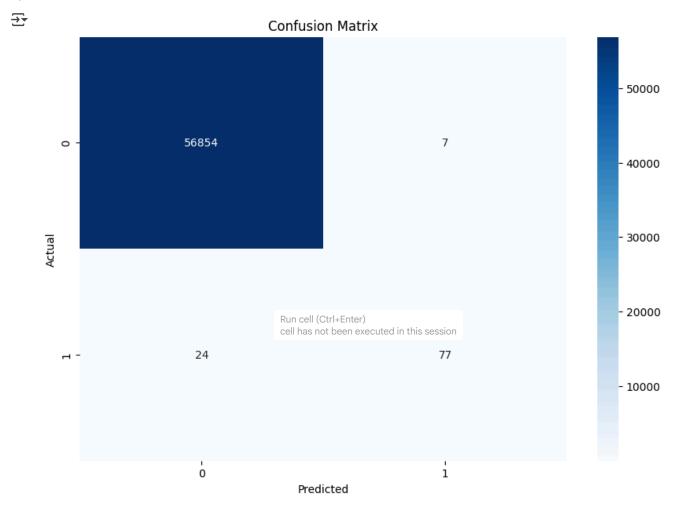
# Training the Random Forest Classification model on the training set

```
from sklearn.ensemble import RandomForestClassifier
classifier=RandomForestClassifier(n_estimators=10,criterion='entropy',random_state=0)
classifier.fit(X_train,y_train)
 →
                                RandomForestClassifier
      RandomForestClassifier(criterion='entropy', n_estimators=10, random_state=0)
y_pred=classifier.predict(X_test)
y_pred
 ⇒ array([0, 0, 0, ..., 0, 0, 0])

→ Predicting the Test set resultance (Ctrl+Enter)

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y_pred = classifier.predict(X_test)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))
 <del>.</del> → [[0 0]
      [0 0]
      [0 0]
      [0 0]
      [0 0]
      [0 0]]
```

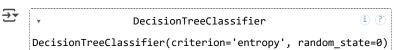
## Making the Confusion Matrix



#### **Decision Tree**

Training The Decision Tree classification Model on the Training set

from sklearn.tree import DecisionTreeClassifier
classifier=DecisionTreeClassifier(criterion='entropy',random\_state=0)
classifier.fit(X\_train,y\_train)



# Predicting a new result

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

→ array([0, 0, 0, ..., 0, 0, 0])

# Predicting the Test set results

```
y_pred = classifier.predict(X_test)
print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),1)),1))

Transport
[[0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
       [0 0]
```

# Making the Confusion Matrix

```
from sklearn.metrics import confusion_matrix, accuracy_score
cm = confusion_matrix(y_test, y_pred)
print(cm)
accuracy_score(y_test, y_pred)

Print(cm)
accuracy_score(y_test, y_pred)

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0.9991573329588147

Model has 99,91% Accuracy

plt.figure(figsize=(10,7))
sns.heatmap(confusion_matrix(y_test,y_pred),annot=True,fmt='d',cmap='Blues')
plt.title('Confusion Matrix')
```