

codsoft-ml1-1

August 5, 2024

TASK-1 CREDIT CARD DETECTION

```
[1]: 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
```

```
[2]: # loading the dataset to a Pandas DataFrame
credit_card_data = pd.read_csv('/content/creditcard (1) (1).csv')
```

```
[3]: credit_card_data.head(5)
```

```
[3]:
```

	Time	V1	V2	V3	V4	V5	V6	V7	\
0	0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	
1	0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	
2	1	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	
3	1	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	
4	2	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	

	V8	V9	...	V21	V22	V23	V24	V25	\
0	0.098698	0.363787	...	-0.018307	0.277838	-0.110474	0.066928	0.128539	
1	0.085102	-0.255425	...	-0.225775	-0.638672	0.101288	-0.339846	0.167170	
2	0.247676	-1.514654	...	0.247998	0.771679	0.909412	-0.689281	-0.327642	
3	0.377436	-1.387024	...	-0.108300	0.005274	-0.190321	-1.175575	0.647376	
4	-0.270533	0.817739	...	-0.009431	0.798278	-0.137458	0.141267	-0.206010	

	V26	V27	V28	Amount	Class
0	-0.189115	0.133558	-0.021053	149.62	0.0
1	0.125895	-0.008983	0.014724	2.69	0.0
2	-0.139097	-0.055353	-0.059752	378.66	0.0
3	-0.221929	0.062723	0.061458	123.50	0.0
4	0.502292	0.219422	0.215153	69.99	0.0

[5 rows x 31 columns]

```
[4]: credit_card_data.shape
```

[4]: (5848, 31)

```
[5]: features = credit_card_data.iloc[:,0:-1]
labels = credit_card_data.iloc[:, -1]
print(features.shape)
print(labels.shape)
```

(5848, 30)

(5848,)

```
[6]: # Handle missing values in the target variable (labels) before splitting
import pandas as pd
from sklearn.impute import SimpleImputer

# Impute missing values in the target variable using the most frequent value
imputer = SimpleImputer(strategy='most_frequent')
labels = imputer.fit_transform(labels.values.reshape(-1, 1)) # Reshape for
    ↪ single feature imputation
labels = pd.Series(labels.flatten()) # Convert back to Series

# Now split the data into training and testing sets
x_train, x_test, y_train, y_test = train_test_split(features, labels,
    ↪ test_size=0.2, random_state=0)
```

```
[7]: # Splitting the data into training and testing sets
x_train,x_test,y_train,y_test = train_test_split(features,labels,test_size=0.
    ↪ 2,random_state=0)
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

(4678, 30)

(1170, 30)

(4678,)

(1170,)

```
[8]: #Building model
from sklearn import tree
decision_tree_model = tree.DecisionTreeClassifier()
decision_tree_model.fit(x_train,y_train)
```

[8]: DecisionTreeClassifier()

```
[9]: # Handle missing values before model prediction
import pandas as pd
from sklearn.impute import SimpleImputer
```

```
# Impute missing values using the mean strategy
imputer = SimpleImputer(strategy='mean')
x_train = imputer.fit_transform(x_train)
x_test = imputer.transform(x_test)

# Now predict using the imputed data
y_pred = decision_tree_model.predict(x_test)
```

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:465: UserWarning: X does not have valid feature names, but DecisionTreeClassifier was fitted with feature names

```
warnings.warn(
```

```
[10]: #Accuracy
from sklearn.metrics import accuracy_score
round(accuracy_score(y_test,y_pred)*100,2)
```

[10]: 99.83

```
[11]: #Classification report
from sklearn.metrics import classification_report
print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
0.0	1.00	1.00	1.00	1169
1.0	0.00	0.00	0.00	1
accuracy			1.00	1170
macro avg	0.50	0.50	0.50	1170
weighted avg	1.00	1.00	1.00	1170

```
[12]: #Confusion matrix
from sklearn.metrics import confusion_matrix
print(confusion_matrix(y_test,y_pred))
```

```
[[1168  1]
 [  1  0]]
```