import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
from sklearn.ensemble import IsolationForest
from scipy.stats import multivariate\_normal
from sklearn.metrics import f1\_score
import io
import pandas as pd
from sklearn.model\_selection import train\_test\_split
from sklearn.metrics import f1\_score,roc\_auc\_score,roc\_curve,precision\_recall\_curve,
from google.colab import drive
drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call dr

!ls'/content/drive/My Drive/fraudData'

/bin/bash: line 1: ls/content/drive/My Drive/fraudData: No such file or director

creditcardDF=pd.read\_csv('/content/creditcard.csv')
creditcardDF.head()

	Time	V1	V2	V3	V4	V5	V6	V7	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.0
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.0
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.2
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.0
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.2

 $5 \text{ rows} \times 31 \text{ columns}$ 

creditcardDF['Class'].value\_counts()

Class

0 284315

Name: count, dtype: int64

492

creditcardDF.shape

(284807, 31)

## creditcardDF.isna().sum()

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
V20	0
V21	0
V22	0
V23	0
V24	0
V25	0
V26	0
V27	0
V28	0
Amount	0
Class	0
dtype:	int64

sns.distplot(creditcardDF['Time'])

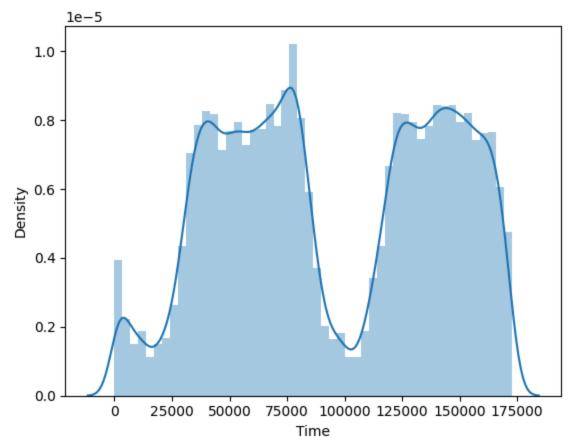
<ipython-input-21-16b1e7f6f4a2>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <a href="https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751">https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751</a>

sns.distplot(creditcardDF['Time'])
<Axes: xlabel='Time', ylabel='Density'>



sns.distplot(creditcardDF['Amount'])

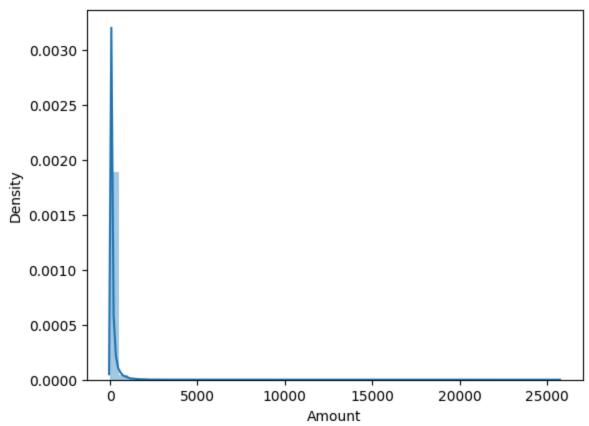
<ipython-input-22-8bcfe78dec34>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <a href="https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751">https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751</a>

```
sns.distplot(creditcardDF['Amount'])
<Axes: xlabel='Amount', ylabel='Density'>
```



```
creditcardDF['Amount'] = np.log(creditcardDF['Amount']+1)
creditcardDF['Time'] = np.log(creditcardDF['Time']+1)
normal = creditcardDF[creditcardDF['Class']==0]
anomaly = creditcardDF[creditcardDF['Class']==1]
train, small_normal = train_test_split(normal, test_size=0.2, random_state=0)
normal_valid, normal_test = train_test_split(small_normal, test_size=0.5, random_state
anomaly_valid, anomaly_test = train_test_split(anomaly, test_size=0.5, random_state=
validation = pd.concat([normal_valid,anomaly_valid])
test = pd.concat([normal_test,anomaly_test])
print(validation.shape)
print(test.shape)
train = train.drop(columns = ['Class']).reset_index(drop=True)
print(train.shape)
```

```
(28677, 31)
  (28678, 31)
  (227452, 30)

featureNames = list(train.columns.values)
valFeatures = validation[featureNames].reset_index(drop= True)
testFeatures = test[featureNames].reset_index(drop= True)

valLabel = validation['Class']
testLabel = test['Class']
```

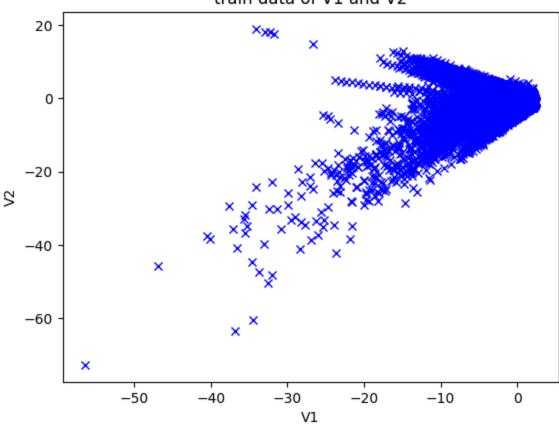
### valFeatures.head()

	Time	V1	V2	V3	V4	V5	V6	V.
0	11.827043	-0.248023	1.259502	-0.993999	-1.587788	1.913462	-0.630270	1.958852
1	10.809566	-1.614505	-0.970137	1.730517	-1.715497	-0.869271	-0.171355	1.216768
2	11.340380	1.106176	0.148096	0.424489	1.282916	-0.080275	0.146526	-0.007108
3	11.321208	-1.791995	1.102738	0.324217	1.082267	-0.303348	-1.050303	0.066270
4	11.956784	1.924286	0.324362	-0.734639	3.370481	0.783552	1.224944	-0.29888

5 rows × 30 columns

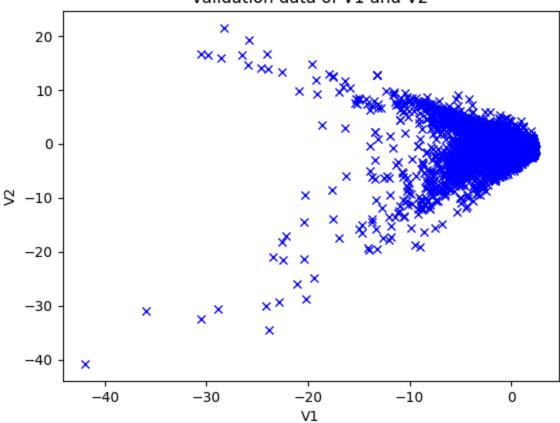
```
validation['Class'].value_counts()
    Class
         28431
            246
    Name: count, dtype: int64
test['Class'].value_counts()
    Class
         28432
           246
    Name: count, dtype: int64
plt.figure()
plt.title("train data of V1 and V2")
plt.xlabel("V1")
plt.ylabel("V2")
plt.plot(train.iloc[:, 1],train.iloc[:,2],"bx")
plt.show()
```

## train data of V1 and V2



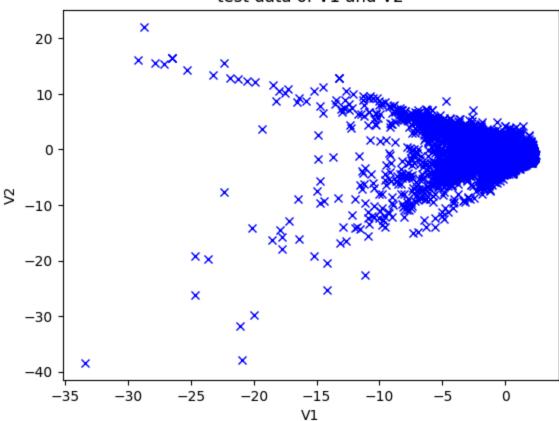
```
plt.figure()
plt.title("validation data of V1 and V2")
plt.xlabel("V1")
plt.ylabel("V2")
plt.plot(validation.iloc[:, 1],validation.iloc[:,2],"bx")
plt.show()
```

# validation data of V1 and V2



```
plt.figure()
plt.title("test data of V1 and V2")
plt.xlabel("V1")
plt.ylabel("V2")
plt.plot(test.iloc[:, 1],test.iloc[:,2],"bx")
plt.show()
```

## test data of V1 and V2



```
np.arange(1,20,2)
    array([ 1,  3,  5,  7,  9,  11,  13,  15,  17,  19])

def estimateGaussian(dataset):
    mu = np.mean(dataset,axis = 0)
    sigma = np.cov(dataset.T)
    return mu,sigma

np.linspace(1,21,10,endpoint = False)
    array([ 1.,  3.,  5.,  7.,  9.,  11.,  13.,  15.,  17.,  19.])

mu,sigma = estimateGaussian(train)
model = multivariate_normal(mean = mu,cov = sigma,allow_singular = True)
```

```
pdfVal = model.pdf(valFeatures)
print(max(pdfVal))
print(min(pdfVal))
p_val = model.logpdf(valFeatures)
print(max(p val))
print(min(p_val))
    3.936022689245927e-12
    0.0
    -26,260850372210967
    -7554.270217704669
p = model.logpdf(train)
print(p.shape)
print((p_val.shape))
    (227452,)
    (28677,)
print(p_val)
print(p_val < -500)
                                     -27.79402451 ... -5175.93656039
    [ -31,28574735 -34,94205051
     -4545.5057626 -29.7152192 ]
    [False False ... True True False]
[[1],[2],[3]]
    [[1], [2], [3]]
scores = []
p val = model.logpdf(valFeatures)
thresholds = np.linspace(min(p val), max(p val), 200)
for threshold in thresholds:
  y_pred = (p_val<threshold).astype(int)</pre>
  scores.append([recall_score(valLabel, y_pred),
                precision score(valLabel, y pred),
                f1_score(valLabel, y_pred, average = "binary")])
scores = np.array(scores)
maxIndex = scores[...,2].ravel().argmax()#maxIndex of the 3rd column (f1 score) #193
bestThreshold = thresholds[maxIndex]
print(scores.shape)#each row is a pair of (recall, precision, f1) corresponding to a
    /usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344:
      _warn_prf(average, modifier, msg_start, len(result))
    (200, 3)
```

print(scores)

```
Outlier detection and fraud data01 - Colaboratory
                               [\0556\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\010100 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01010 \0066\01000 \0066\01000 \0066\01000 \0066\01000 \0066\01000 \0066
                               [0.82113821 0.45190157 0.58297258]
                               [0.84146341 0.34214876 0.48648649]
                               [0.85365854 0.19337017 0.31531532]
                                                                                                0.0085786 0.01701127]]
                               [1.
print(maxIndex)
print(bestThreshold)
                         193
                         -253.23600717620593
np.mean(train.iloc[:,1])
                         0.00524675542006222
mu[1]
                         0.00524675542006222
print(mu)
                        Time
                                                                                11,252384
                        V1
                                                                                     0.005247
                        ٧2
                                                                               -0.005416
                        V3
                                                                                     0.010293
                        ٧4
                                                                                -0.008144
                        ۷5
                                                                                     0.004281
                        ۷6
                                                                                     0.001813
                        ٧7
                                                                                     0.010354
                        ٧8
                                                                               -0.001103
                        ۷9
                                                                                     0.006351
                        V10
                                                                                     0.009573
                        V11
                                                                               -0.007736
```

dtype: float64

Amount

V12

V13

V14

V15

V16

V17

V18

V19

V20

V21

V22

V23

V24

V25

V26

V27

V28

0.009943

0.001084

0.010816

0.001082

0.007216

0.012364

0.003412

-0.001811

-0.001092

-0.001302

-0.000354

0.000209

0.000288

0.000375

0.000457

-0.000509

-0.000119

3.152259

```
y test pred raw = model.logpdf(testFeatures)
y pred test = y test pred raw < bestThreshold
f1 score(testLabel, y pred test, average = "binary")
    0.7401574803149606
y pred test
    array([False, False, False, ..., True, False, True])
predoutliersTest = np.asarray(np.where(y_pred_test))
len(predoutliersTest[0])
    262
predoutliersTest
                                                        1462.
    array([[
             248.
                      437.
                             605.
                                   1007,
                                          1353.
                                                 1451.
                                                               1546.
                                                                       1988.
                     3674,
                            3928,
                                   4216,
                                          4928,
                                                 5144,
                                                        5846,
                                                               5975.
                                                                       6022.
              2461,
              6682,
                     6706,
                            6858,
                                   7017,
                                          7138,
                                                 8267,
                                                        8452,
                                                               8611,
                                                                       8677.
             8936,
                           9207,
                                   9443,
                                          9807, 9988, 10263, 10391, 10657,
                    8996.
             11224, 12205, 13539, 13935, 14050, 14573, 14579, 14802, 14869,
            15740, 16061, 16888, 17322, 17663, 19352, 19902, 20680, 20800,
            21748, 22366, 22552, 22859, 23217, 23456, 23742, 24639, 24819,
            25654, 25678, 26035, 27282, 27293, 27314, 27587, 27723, 28117,
            28178, 28396, 28432, 28433, 28434, 28435, 28436, 28437, 28438,
            28440, 28443, 28444, 28445, 28446, 28447, 28449, 28450, 28453,
            28454, 28455, 28456, 28457, 28458, 28459, 28460, 28461, 28462,
            28463, 28464, 28465, 28466, 28467, 28468, 28469, 28470, 28471,
            28472, 28473, 28475, 28479, 28480, 28481, 28482, 28483, 28484,
            28486, 28487, 28490, 28492, 28493, 28494, 28496, 28497, 28498,
            28499, 28500, 28501, 28502, 28503, 28505, 28506, 28507, 28508,
            28510, 28511, 28512, 28513, 28517, 28521, 28523, 28525, 28526,
            28527, 28528, 28529, 28530, 28531, 28532, 28536, 28538, 28539,
            28540, 28542, 28543, 28544, 28546, 28547, 28549, 28550, 28551,
            28552, 28553, 28554, 28555, 28556, 28558, 28559, 28560, 28561,
            28562, 28564, 28565, 28566, 28567, 28568, 28570, 28572, 28574,
            28575, 28576, 28577, 28578, 28579, 28580, 28581, 28583, 28584,
            28585, 28586, 28588, 28589, 28591, 28592, 28594, 28596, 28598,
            28599, 28600, 28601, 28602, 28603, 28604, 28605, 28606, 28607,
            28609, 28610, 28612, 28615, 28617, 28618, 28619, 28620, 28621,
            28622, 28623, 28625, 28626, 28628, 28629, 28630, 28631, 28632,
            28633, 28636, 28637, 28638, 28639, 28640, 28641, 28642, 28643,
            28645, 28646, 28647, 28648, 28649, 28650, 28651, 28652, 28653,
            28654, 28656, 28657, 28658, 28659, 28660, 28661, 28662, 28663,
            28664, 28666, 28669, 28670, 28671, 28672, 28673, 28674, 28675,
            28677]])
```

```
plt.figure()
plt.title("test_data with outlier flaged red")
plt.xlabel("V2")
plt.ylabel("V3")
plt.plot(testFeatures.iloc[:, 2],testFeatures.iloc[:,3],"bx")
plt.plot(testFeatures.iloc[predoutliersTest[0],1],testFeatures.iloc[predoutliersTest
plt.show()
```

