8/28/23, 12:17 PM Credit Card fraud

```
import numpy as np
In [11]:
          import pandas as pd
          from sklearn.model_selection import train_test_split
          from sklearn.linear_model import LogisticRegression
          from sklearn.metrics import accuracy score
In [12]:
          credited card df = pd.read csv(r'C:\Users\Admin\OneDrive\Desktop\creditcard.csv')
          credited_card_df.head()
Out[12]:
                       V1
                                V2
                                         V3
                                                   V4
                                                             V5
                                                                       V6
                                                                                V7
                                                                                          V8
                                                                                                    V9
              X
          0 0.0 -1.359807 -0.072781 2.536347
                                              1.378155 -0.338321
                                                                 0.462388
                                                                           0.239599
                                                                                     0.098698
                                                                                               0.363787
          1 0.0 1.191857
                           0.266151 0.166480
                                                                -0.082361
                                                                           -0.078803
                                                                                     0.085102 -0.255425
                                              0.448154
                                                        0.060018
          2 1.0 -1.358354 -1.340163 1.773209
                                              0.379780 -0.503198
                                                                 1.800499
                                                                           0.791461
                                                                                     0.247676 -1.514654
          3 1.0 -0.966272 -0.185226 1.792993
                                             -0.863291 -0.010309
                                                                                     0.377436 -1.387024
                                                                 1.247203
                                                                           0.237609
          4 2.0 -1.158233 0.877737 1.548718
                                             0.403034 -0.407193
                                                                 0.095921
                                                                           0.592941 -0.270533
                                                                                               0.817739
         5 rows × 31 columns
          credited_card_df.info()
In [13]:
```

```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 284809 entries, 0 to 284808
         Data columns (total 31 columns):
              Column
                      Non-Null Count
                                        Dtype
                       _____
                      284805 non-null float64
          0
              Х
          1
              ٧1
                      284809 non-null
                                       float64
          2
              V2
                      284807 non-null
                                        float64
          3
              V3
                      284807 non-null
                                        float64
          4
              V4
                                       float64
                      284807 non-null
          5
              V5
                      284807 non-null float64
          6
              ۷6
                      284807 non-null
                                       float64
          7
              V7
                      284807 non-null float64
          8
              V8
                      284807 non-null
                                       float64
          9
              ۷9
                      284807 non-null
                                        float64
          10
              V10
                      284807 non-null
                                       float64
          11
              V11
                      284807 non-null
                                       float64
          12
              V12
                      284807 non-null
                                       float64
          13
              V13
                      284807 non-null float64
          14
              V14
                      284807 non-null
                                       float64
                      284807 non-null
                                        float64
          15
              V15
          16
              V16
                      284807 non-null
                                       float64
          17
              V17
                      284807 non-null
                                       float64
          18
              V18
                      284807 non-null
                                       float64
          19
              V19
                      284807 non-null
                                       float64
                      284807 non-null float64
          20
              V20
                                       float64
          21
              V21
                      284807 non-null
          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
                                       float64
              V27
                      284807 non-null
          28
              V28
                      284807 non-null
                                       float64
          29
              Amount 284807 non-null
                                       float64
          30
              Class
                       284807 non-null float64
         dtypes: float64(31)
         memory usage: 67.4 MB
         credited_card_df.shape
In [14]:
         (284809, 31)
Out[14]:
         credited_card_df['Class'].value_counts()
In [15]:
         0.0
                284315
Out[15]:
         1.0
                   492
         Name: Class, dtype: int64
         credited card df.describe()
In [16]:
```

23, 12:17 PM	Credit Card fraud												
Out[16]:		х		V1	V2		V3	V4	•	V5			
	count	284805.000000	284809.00	0000	2.848070e+05	2.848070	e+05 2.8	48070e+05	2.848070e+	05 2.			
	mean	94813.765629	0.75	9751 -	-8.252296e-13	-9.637438	Be-13 8.3	316234e-13	1.592013e-	13 4			
	std	47488.287552	288.67	5554	1.651309e+00	1.516255	e+00 1.4	15869e+00	1.380247e+	00 1.			
	min	0.000000	-56.40	7510 -	7.271573e+01	-4.832559	e+01 -5.6	83171e+00	-1.137433e+	02 -2.			
	25%	54201.000000	-0.92	0372 -	-5.985499e-01	-8.903648	3e-01 -8.4	186401e-01	-6.915971e-	01 -7			
	50%	84691.000000	0.01	8109	6.548556e-02	1.798463	Be-01 -1.9	984653e-02	-5.433583e-	02 -2			
	75%	139321.000000	1.31	5648	8.037239e-01	1.027196	e+00 7.4	133413e-01	6.119264e-	01 3			
	max	172792.000000	792.000000 120883.000000		2.205773e+01	9.382558e+00 1.6		87534e+01	3.480167e+	01 7			
	8 rows	× 31 columns											
4										>			
In [17]:	<pre>legit= credited_card_df[credited_card_df.Class==0] fraud= credited_card_df[credited_card_df['Class']==1]</pre>												
In [18]:	legit												
Out[18]:		x	V1	V2	2 V3	V4	V5	V6	V7				
	(0.0 -1	.359807 -	0.072781	1 2.536347	1.378155	-0.338321	0.462388	0.239599	0.0986			
	1	0.0 1	.191857	0.26615	1 0.166480	0.448154	0.060018	-0.082361	-0.078803	0.0851			
	2	2 1.0 -1	.358354 -	1.340163	3 1.773209	0.379780	-0.503198	1.800499	0.791461	0.2476			
	3	3 1.0 -0).966272 -	0.185226	6 1.792993	-0.863291	-0.010309	1.247203	0.237609	0.3774			
	4	2.0 -1	.158233	0.877737	7 1.548718	0.403034	-0.407193	0.095921	0.592941	-0.2705			

	x	V1	V2	V3	V4	V5	V6	V7	
0	0.0	-1.359807	-0.072781	2.536347	1.378155	-0.338321	0.462388	0.239599	0.0986
1	0.0	1.191857	0.266151	0.166480	0.448154	0.060018	-0.082361	-0.078803	0.0851
2	1.0	-1.358354	-1.340163	1.773209	0.379780	-0.503198	1.800499	0.791461	0.2476
3	1.0	-0.966272	-0.185226	1.792993	-0.863291	-0.010309	1.247203	0.237609	0.3774
4	2.0	-1.158233	0.877737	1.548718	0.403034	-0.407193	0.095921	0.592941	-0.2705
•••									
284804	172786.0	-11.881118	10.071785	-9.834783	-2.066656	-5.364473	-2.606837	-4.918215	7.3053
284805	172787.0	-0.732789	-0.055080	2.035030	-0.738589	0.868229	1.058415	0.024330	0.2948
284806	172788.0	1.919565	-0.301254	-3.249640	-0.557828	2.630515	3.031260	-0.296827	0.7084
284807	172788.0	-0.240440	0.530483	0.702510	0.689799	-0.377961	0.623708	-0.686180	0.6791
284808	172792.0	-0.533413	-0.189733	0.703337	-0.506271	-0.012546	-0.649617	1.577006	-0.4146

284315 rows × 31 columns

In [19]: fraud 8/28/23, 12:17 PM Credit Card fraud

X

Out[19]:

V1

V2

V3

V4

V5

V6

V7

V8

```
541
                        406.0 -2.312227
                                          1.951992
                                                   -1.609851
                                                              3.997906
                                                                        -0.522188
                                                                                                         1.391657
                                                                                  -1.426545
                                                                                             -2.537387
                623
                        472.0
                              -3.043541
                                         -3.157307
                                                     1.088463
                                                              2.288644
                                                                         1.359805
                                                                                   -1.064823
                                                                                              0.325574
                                                                                                        -0.067794
               4920
                       4462.0
                               -2.303350
                                                    -0.359745
                                                              2.330243
                                                                        -0.821628
                                                                                   -0.075788
                                                                                                        -0.399147
                                          1.759247
                                                                                              0.562320
               6108
                       6986.0
                               -4.397974
                                          1.358367
                                                    -2.592844
                                                              2.679787
                                                                        -1.128131
                                                                                   -1.706536
                                                                                             -3.496197
                                                                                                        -0.248778
               6329
                       7519.0
                                1.234235
                                          3.019740
                                                    -4.304597
                                                              4.732795
                                                                         3.624201
                                                                                   -1.357746
                                                                                              1.713445
                                                                                                        -0.496358
                     169142.0
                               -1.927883
                                                                                   -2.010494
             279865
                                          1.125653
                                                    -4.518331
                                                              1.749293
                                                                        -1.566487
                                                                                             -0.882850
                                                                                                         0.697211
                                1.378559
             280145
                     169347.0
                                          1.289381
                                                    -5.004247
                                                              1.411850
                                                                         0.442581
                                                                                   -1.326536
                                                                                             -1.413170
                                                                                                         0.248525
             280151
                     169351.0
                               -0.676143
                                                              0.468308
                                                                        -1.120541
                                                                                   -0.003346
                                                                                                         1.210158
                                          1.126366
                                                    -2.213700
                                                                                             -2.234739
             281146
                    169966.0
                               -3.113832
                                          0.585864
                                                    -5.399730
                                                              1.817092
                                                                         -0.840618
                                                                                   -2.943548
                                                                                              -2.208002
                                                                                                         1.058733
             281676 170348.0
                                1.991976
                                          0.158476
                                                   -2.583441 0.408670
                                                                         1.151147
                                                                                  -0.096695
                                                                                              0.223050
                                                                                                        -0.068384
            492 rows × 31 columns
 In [20]:
             fraud.shape
             (492, 31)
 Out[20]:
 In [21]:
             legit.shape
             (284315, 31)
 Out[21]:
             legit sample = legit.sample(n=492)
 In [22]:
             credited card df = pd.concat([legit sample,fraud],axis=0)
             credited card df['Class'].value counts()
 In [23]:
            0.0
                     492
 Out[23]:
            1.0
                     492
            Name: Class, dtype: int64
             credited_card_df.groupby('Class').mean()
 In [24]:
 Out[24]:
                                       ۷1
                                                 V2
                                                           V3
                                                                      V4
                                                                                 V5
                                                                                           V6
                                                                                                     V7
                              X
             Class
              0.0
                   95917.508130 -0.157794 0.103825
                                                      0.075746
                                                               -0.066376
                                                                           0.002951
                                                                                      0.041324 -0.069352
                                                                                                          -0.0624
                   80746.806911 -4.771948 3.623778 -7.033281
                                                                 4.542029
                                                                          -3.151225 -1.397737 -5.568731
                                                                                                           0.5706
            2 rows × 30 columns
4
            x = credited card df.drop('Class',axis=1)
             y = credited_card_df['Class']
```

8/28/23, 12:17 PM Credit Card fraud

```
In [26]:
         x.shape
          (984, 30)
Out[26]:
In [27]:
         y.shape
          (984,)
Out[27]:
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,stratify=y,random_statest)
In [30]:
         model = LogisticRegression()
In [34]:
          model.fit(x_train,y_train)
         ypred = model.predict(x_test)
         accuracy_score(ypred,y_test)
In [35]:
         0.9187817258883249
Out[35]:
In [ ]:
```