```
In [2]:
         import tensorflow as tf
          import pandas as pd
          import numpy as np
In [3]: | df = pd.read_csv('Desktop\creditcard.csv')
In [4]: | df.head()
Out[4]:
                         V1
                                   V2
                                             V3
                                                                 V5
                                                                           V6
                                                                                     V7
                                                                                               ۷ŧ
             Time
                                                       V4
          0
               0.0 -1.359807
                             -0.072781
                                       2.536347
                                                  1.378155 -0.338321
                                                                     0.462388
                                                                               0.239599
                                                                                          0.098698
          1
                   1.191857
                              0.266151 0.166480
                                                 0.448154
                                                           0.060018
                                                                     -0.082361
               0.0
                                                                               -0.078803
                                                                                          0.085102
          2
               1.0 -1.358354 -1.340163 1.773209
                                                 0.379780 -0.503198
                                                                      1.800499
                                                                               0.791461
                                                                                          0.247676
          3
               1.0 -0.966272 -0.185226 1.792993
                                                 -0.863291
                                                          -0.010309
                                                                     1.247203
                                                                               0.237609
                                                                                          0.377436
               2.0 -1.158233
                            0.877737 1.548718
                                                 0.403034 -0.407193
                                                                     0.095921
          4
                                                                               0.592941
                                                                                         -0.270533
          5 rows × 31 columns
In [5]:
         df = df.drop(['Time', 'Class'], axis = 1)
         df.shape
Out[5]: (284807, 29)
In [6]:
         from sklearn.preprocessing import StandardScaler
          scaler = StandardScaler()
          df.iloc[:, :-1] = scaler.fit_transform(df.iloc[:, :-1])
In [7]: df.head()
Out[7]:
                   V1
                             V2
                                       V3
                                                 V4
                                                           V5
                                                                     V6
                                                                               V7
                                                                                         V8
                                                               0.347068
            -0.694242
                       -0.044075
                                                                          0.193679
                                                                                    0.082637
                                                                                              0.33
                                 1.672773
                                           0.973366
                                                     -0.245117
             0.608496
                        0.161176 0.109797
                                           0.316523
                                                     0.043483
                                                               -0.061820
                                                                         -0.063700
                                                                                   0.071253
                                                                                            -0.23
                                                                         0.639776
            -0.693500
                       -0.811578
                                 1.169468
                                           0.268231
                                                     -0.364572
                                                                1.351454
                                                                                    0.207373 -1.37
             -0.493325
                       -0.112169
                                1.182516
                                          -0.609727
                                                    -0.007469
                                                               0.936150
                                                                          0.192071
                                                                                    0.316018 -1.26
             -0.591330
                        0.531541 1.021412
                                           0.284655
                                                    -0.295015
                                                               0.071999
                                                                         0.479302
                                                                                   -0.226510
                                                                                             0.74
          5 rows × 29 columns
```

```
In [8]: from sklearn.model_selection import train_test_split
        x_train, x_test = train_test_split(df, test_size=0.2)
        print("x_train shape : ", x_train.shape)
        print("x_test shape : ", x_test.shape)
        x_train shape : (227845, 29)
        x_test shape : (56962, 29)
In [9]: from keras.models import Model, Sequential
        from tensorflow.keras.layers import Dense, Dropout
        from tensorflow.keras import layers, models
        encoder = tf.keras.models.Sequential([
            layers.Input(shape=(x_train.shape[1],)),
            layers.Dense(20, activation='relu'),
            layers.Dense(20, activation='relu'),
            layers.Dense(10, activation='relu')
        ])
        decoder = tf.keras.models.Sequential([
            layers.Input(shape=(10,)),
            layers.Dense(20, activation='relu'),
            layers.Dense(20, activation='relu'),
            layers.Dense(x_train.shape[1], activation='linear')
        ])
        model = tf.keras.models.Sequential([
            encoder,
            decoder,
        ])
```

```
In [10]: model.compile(optimizer='adam', loss ='mean_squared_error')
```

```
In [11]: history = model.fit(
           x_train,
           x_train,
           validation_data=(x_test,x_test),
           epochs=10,
           batch_size = 500,
           shuffle=True
        )
        Epoch 1/10
        456/456 [============== ] - 4s 6ms/step - loss: 223.8373 -
        val_loss: 1.0004
        Epoch 2/10
        456/456 [============== ] - 2s 5ms/step - loss: 0.8491 - v
        al loss: 0.8041
        Epoch 3/10
        456/456 [=============== ] - 2s 5ms/step - loss: 0.7655 - v
        al_loss: 0.7357
        Epoch 4/10
        456/456 [============== ] - 2s 5ms/step - loss: 0.6828 - v
        al loss: 0.7777
        Epoch 5/10
        456/456 [============= ] - 2s 4ms/step - loss: 0.7826 - v
        al_loss: 0.6901
        Epoch 6/10
        456/456 [============== ] - 2s 5ms/step - loss: 0.6614 - v
        al loss: 0.6123
        Epoch 7/10
        456/456 [============= ] - 2s 4ms/step - loss: 0.7350 - v
        al_loss: 0.6275
        Epoch 8/10
        456/456 [============= ] - 1s 3ms/step - loss: 0.6138 - v
        al loss: 0.6003
        Epoch 9/10
        456/456 [============== ] - 1s 3ms/step - loss: 1.0103 - v
        al loss: 0.8047
        Epoch 10/10
        456/456 [============= ] - 2s 5ms/step - loss: 0.6182 - v
        al loss: 0.5644
In [ ]:
In [13]: predictions = model.predict(x_test)
        1781/1781 [========== ] - 3s 2ms/step
```

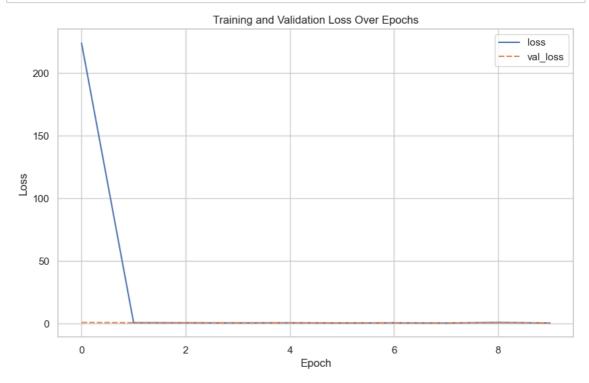
```
mse = np.mean(np.power(x_test - predictions, 2), axis=1)
In [14]:
         mse
Out[14]: 49994
                   0.253195
         194137
                   0.421602
                   0.387397
         80151
                   0.305153
         163520
         180985
                   0.347469
         221778
                   0.363021
         3480
                  0.230793
         207647
                  0.843475
                   0.244909
         138889
         72395
                   0.161340
         Length: 56962, dtype: float64
In [15]: threshold = np.percentile(mse, 95)
         threshold
Out[15]: 1.1053713304779724
In [16]: | anomalies = mse > threshold
In [17]: | num_anomalies = np.sum(anomalies)
         print(f"Number of Anomalies: {num_anomalies}")
```

Number of Anomalies: 2849

```
In [25]: import seaborn as sns
import matplotlib.pyplot as plt

sns.set(style="whitegrid")
plt.figure(figsize=(10, 6))
loss_plot = sns.lineplot(data=history.history)

# Set plot labels and title
loss_plot.set(xlabel='Epoch', ylabel='Loss', title='Training and Validation
plt.show()
```



```
In [ ]:
```