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In [1]: import pandas as pd
import numpy as np
import pickle
import matplotlib.pyplot as plt
import tensorflow as tf
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import confusion_matrix
from keras.models import Model, load_model, Sequential
from keras.layers import Input, Dense
from keras.callbacks import ModelCheckpoint, TensorBoard

In [33]: %matplotlib inline
sns.set(style='whitegrid')

In [34]: df=pd.read_csv('C:/sameer/creditcard1.csv')

In [35]: df = df.drop(['Time'], axis=1)

In [36]: df['Amount'] = StandardScaler().fit_transform(df['Amount'].values.reshape(-1, 1))

In [37]: df_fraud = df[df['Class']==1]
df_normal = df[df['Class']==0]
df_normal = df_normal.sample(frac = 1.0).reset_index(drop = True)
df_normal_1 = df_normal.iloc[:int(df_normal.shape[0]*0.8),:]
df_normal_2 = df_normal.iloc[:int(df_normal.shape[0]*0.8),:]

In [38]: x_test = pd.concat([df_fraud, df_normal_2], axis = 0)
x_test = x_test.sample(frac = 1.0).reset_index(drop = True)

#Separate in input and target variables
x_train = df_normal_1[df_normal_1['Class']== 0]
x_train = x_train.drop(['Class'], axis=1)

y_test = x_test['Class']
x_test = x_test.drop(['Class'], axis=1)

In [39]: #Build the Neural Network
input_dim = x_train.shape[1]
encoding_dim = 14

model = Sequential()
model.add(Dense(29, input_dim = input_dim, activation="relu"))
model.add(Dense(14, activation="relu"))
model.add(Dense(7, activation="relu"))
model.add(Dense(14, activation="relu"))
model.add(Dense(input_dim, activation="sigmoid"))
model.compile(optimizer='adam', loss='mean_squared_error', metrics=['accuracy'])

In [40]: model.summary()
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Model: "sequential_2"

Layer (type)	Output Shape	Param #
dense_9 (Dense)	(None, 29)	870
dense_10 (Dense)	(None, 14)	420
dense_11 (Dense)	(None, 7)	105
dense_12 (Dense)	(None, 14)	112
dense_13 (Dense)	(None, 29)	435

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 Total params: 1942 (7.59 KB)
 Trainable params: 1942 (7.59 KB)
 Non-trainable params: 0 (0.00 Byte)

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In [41]: #Fit the autoencoder and check loss for train and test
checkpointer = ModelCheckpoint(filepath="nae.h5", verbose=0, save_best_only=True)
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In [42]: #Save history to plot Learning curves
history = model.fit(x_train, x_train,
epochs=10,
batch_size=32,
shuffle=True,
validation_data=(x_test, x_test),
verbose=1,
callbacks=[checkpointer]).history

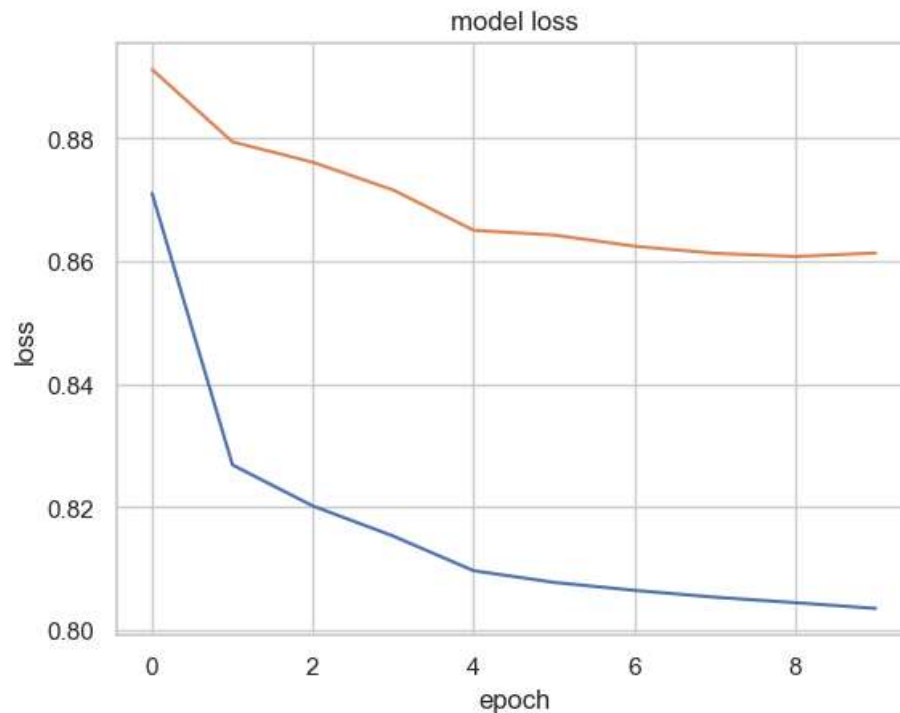
autoencoder = load_model('nae.h5')

Epoch 1/10
7108/7108 [=====] - 62s 8ms/step - loss: 0.8710 - accuracy: 0.5202 - val_loss: 0.8912
- val_accuracy: 0.5530
Epoch 2/10
21/7108 [.....] - ETA: 37s - loss: 0.9002 - accuracy: 0.5729
C:\Users\samir\anaconda3\Lib\site-packages\keras\src\engine\training.py:3079: UserWarning: You are saving your
model as an HDF5 file via `model.save()`. This file format is considered legacy. We recommend using instead the
native Keras format, e.g. `model.save('my_model.keras')`.
  saving_api.save_model(
7108/7108 [=====] - 52s 7ms/step - loss: 0.8269 - accuracy: 0.5632 - val_loss: 0.8794
- val_accuracy: 0.5690
Epoch 3/10
7108/7108 [=====] - 52s 7ms/step - loss: 0.8202 - accuracy: 0.5681 - val_loss: 0.8761
- val_accuracy: 0.5617
Epoch 4/10
7108/7108 [=====] - 53s 7ms/step - loss: 0.8153 - accuracy: 0.5779 - val_loss: 0.8716
- val_accuracy: 0.5863
Epoch 5/10
7108/7108 [=====] - 55s 8ms/step - loss: 0.8097 - accuracy: 0.5748 - val_loss: 0.8650
- val_accuracy: 0.5739
Epoch 6/10
7108/7108 [=====] - 52s 7ms/step - loss: 0.8078 - accuracy: 0.5772 - val_loss: 0.8643
- val_accuracy: 0.5822
Epoch 7/10
7108/7108 [=====] - 49s 7ms/step - loss: 0.8065 - accuracy: 0.5771 - val_loss: 0.8625
- val_accuracy: 0.5845
Epoch 8/10
7108/7108 [=====] - 50s 7ms/step - loss: 0.8054 - accuracy: 0.5778 - val_loss: 0.8613
- val_accuracy: 0.5803
Epoch 9/10
7108/7108 [=====] - 50s 7ms/step - loss: 0.8045 - accuracy: 0.5765 - val_loss: 0.8608
- val_accuracy: 0.5747
Epoch 10/10
7108/7108 [=====] - 52s 7ms/step - loss: 0.8035 - accuracy: 0.5711 - val_loss: 0.8614
- val_accuracy: 0.5717
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In [43]: #Plot losses
plt.plot(history['loss'])
plt.plot(history['val_loss'])
plt.title('model loss')
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plt.ylabel('loss')
plt.xlabel('epoch')
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Out[43]: Text(0.5, 0, 'epoch')



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In [44]: #Predict on test set
predictions = model.predict(x_test)

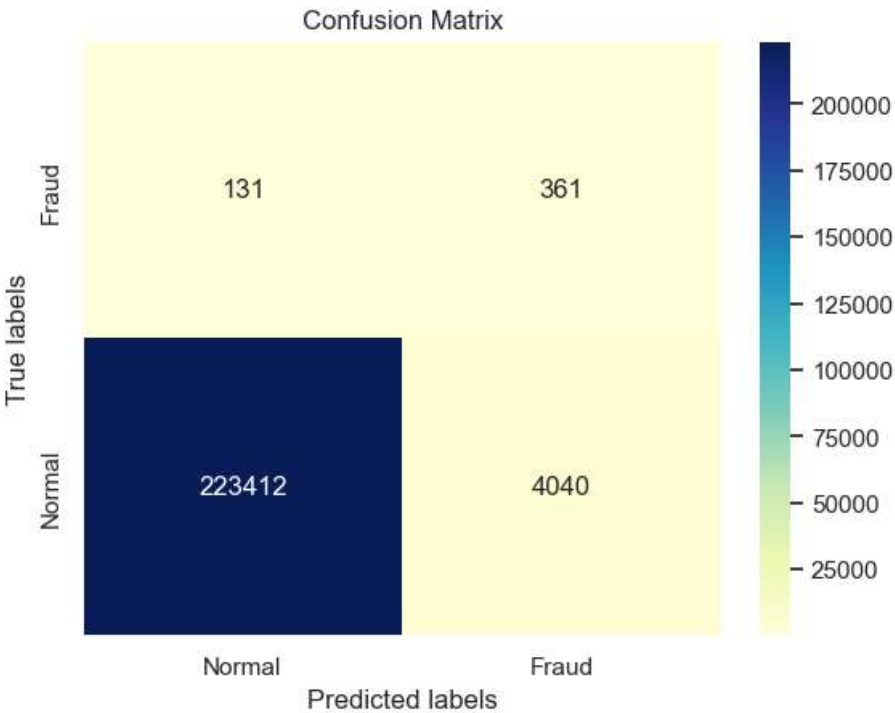
mse = np.mean(np.power(x_test - predictions, 2), axis=1)
error_df = pd.DataFrame({'mse': mse, 'fraud': y_test})

7124/7124 [=====] - 22s 3ms/step
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In [45]: #Set an error threshold above which a transaction is considered fraud
threshold = 4.5
error_df['pred_01'] = [1 if e > threshold else 0 for e in error_df['mse'].values]
conf_mat = confusion_matrix(error_df['fraud'], error_df['pred_01'])
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In [46]: #Print confusion matrix for the given threshold
ax=plt.subplot()
sns.heatmap(conf_mat, annot=True, fmt="g", cmap="YlGnBu")
ax.set_xlabel('Predicted labels');ax.set_ylabel('True labels');
ax.set_title('Confusion Matrix');
ax.set_ylim([0,2])
ax.xaxis.set_ticklabels(["Normal", "Fraud"]); ax.yaxis.set_ticklabels(["Normal", "Fraud"])
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Out[46]: [Text(0, 0.5, 'Normal'), Text(0, 1.5, 'Fraud')]



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