

Importing Libraries

```
import sklearn
import pandas as pd
import numpy as np
from tensorflow import keras
import tensorflow as tf
import matplotlib.pyplot as plt
```

📌 The default version of TensorFlow in Colab will soon switch to TensorFlow 2.x.
We recommend you [upgrade](#) now or ensure your notebook will continue to use TensorFlow 1.x via the %tensorflow1 magic.

```
from google.colab import files
uploaded = files.upload()
```

📌 Upload widget is only available when the cell has been executed in a notebook. Saving creditcard.csv to creditcard.csv

```
import io
#Basic Dataset characteristics - initial EDA
cc_df = pd.read_csv(io.BytesIO(uploaded["creditcard.csv"]))
cc_df.shape
```

📌 (284807, 31)

```
cc_df.head()
```

📌

| | Time | V1 | V2 | V3 | V4 | V5 | V6 | V7 | V8 |
|---|------|-----------|-----------|----------|-----------|-----------|-----------|-----------|-----------|
| 0 | 0.0 | -1.359807 | -0.072781 | 2.536347 | 1.378155 | -0.338321 | 0.462388 | 0.239599 | 0.098691 |
| 1 | 0.0 | 1.191857 | 0.266151 | 0.166480 | 0.448154 | 0.060018 | -0.082361 | -0.078803 | 0.085106 |
| 2 | 1.0 | -1.358354 | -1.340163 | 1.773209 | 0.379780 | -0.503198 | 1.800499 | 0.791461 | 0.247671 |
| 3 | 1.0 | -0.966272 | -0.185226 | 1.792993 | -0.863291 | -0.010309 | 1.247203 | 0.237609 | 0.377436 |
| 4 | 2.0 | -1.158233 | 0.877737 | 1.548718 | 0.403034 | -0.407193 | 0.095921 | 0.592941 | -0.270533 |

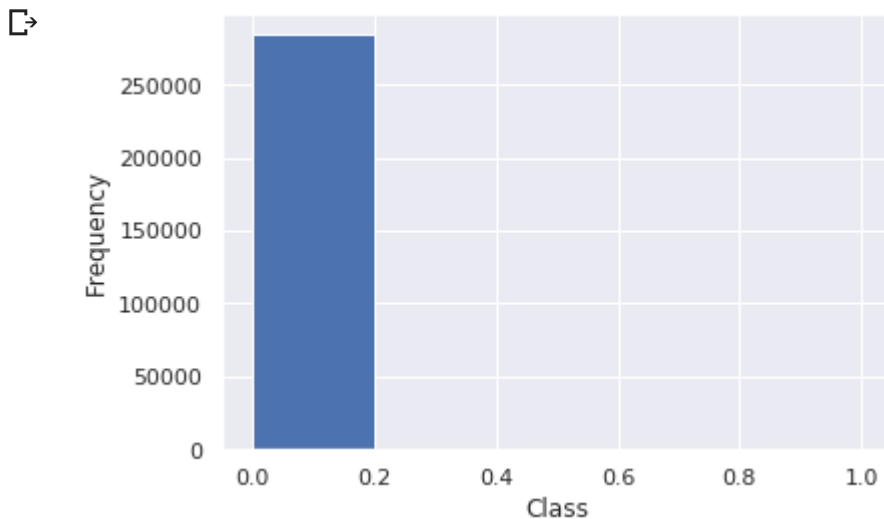
```
cc_df.tail()
```

📌

| | Time | V1 | V2 | V3 | V4 | V5 | V6 | V |
|---------------|----------|------------|-----------|-----------|-----------|-----------|-----------|----------|
| 284802 | 172786.0 | -11.881118 | 10.071785 | -9.834783 | -2.066656 | -5.364473 | -2.606837 | -4.91821 |
| 284803 | 172787.0 | -0.732789 | -0.055080 | 2.035030 | -0.738589 | 0.868229 | 1.058415 | 0.02433 |
| 284804 | 172788.0 | 1.919565 | -0.301254 | -3.249640 | -0.557828 | 2.630515 | 3.031260 | -0.29682 |
| 284805 | 172788.0 | -0.240440 | 0.530483 | 0.702510 | 0.689799 | -0.377961 | 0.623708 | -0.68618 |
| 284806 | 172792.0 | -0.533413 | -0.189733 | 0.703337 | -0.506271 | -0.012546 | -0.649617 | 1.57700 |

```
X = cc_df.iloc[:,0:30]
y = cc_df[['Class']]
```

```
#Frequency distribution of fraudulent and non-fraudulent txns
import seaborn as sns
sns.set()
plt.hist(cc_df['Class'],bins=5)
plt.xlabel('Class')
plt.ylabel('Frequency')
plt.show()
```



```
print(X.head(),"\n",X.tail())
```



| | Time | V1 | V2 | V3 | ... | V26 | V27 | V28 | Amount |
|---|------|-----------|-----------|----------|-----|-----------|-----------|-----------|--------|
| 0 | 0.0 | -1.359807 | -0.072781 | 2.536347 | ... | -0.189115 | 0.133558 | -0.021053 | 149.62 |
| 1 | 0.0 | 1.191857 | 0.266151 | 0.166480 | ... | 0.125895 | -0.008983 | 0.014724 | 2.69 |
| 2 | 1.0 | -1.358354 | -1.340163 | 1.773209 | ... | -0.139097 | -0.055353 | -0.059752 | 378.66 |
| 3 | 1.0 | -0.966272 | -0.185226 | 1.792993 | ... | -0.221929 | 0.062723 | 0.061458 | 123.50 |
| 4 | 2.0 | -1.158233 | 0.877737 | 1.548718 | ... | 0.502292 | 0.219422 | 0.215153 | 69.99 |

[5 rows x 30 columns]

| | Time | V1 | V2 | ... | V27 | V28 | Amount |
|--------|----------|------------|-----------|-----|-----------|-----------|--------|
| 284802 | 172786.0 | -11.881118 | 10.071785 | ... | 0.943651 | 0.823731 | 0.77 |
| 284803 | 172787.0 | -0.732789 | -0.055080 | ... | 0.068472 | -0.053527 | 24.79 |
| 284804 | 172788.0 | 1.919565 | -0.301254 | ... | 0.004455 | -0.026561 | 67.88 |
| 284805 | 172788.0 | -0.240440 | 0.530483 | ... | 0.108821 | 0.104533 | 10.00 |
| 284806 | 172792.0 | -0.533413 | -0.189733 | ... | -0.002415 | 0.013649 | 217.00 |

[5 rows x 30 columns]

```
print(y.head(),"\n",y.tail())
```

```

[ ]> Class
0      0
1      0
2      0
3      0
4      0

      Class
284802    0
284803    0
284804    0
284805    0
284806    0

```

```
#Computing number of fraudulent and authentic transactions in the dataset
```

```
fraud_Txn = cc_df.loc[cc_df['Class'] == 1]
```

```
nonfraud_Txn = cc_df.loc[cc_df['Class'] == 0]
```

```
print("No of fraudulent transactions: " + str(len(fraud_Txn)))
```

```
print("No of non-fraudulent transactions: " + str(len(nonfraud_Txn)))
```

```

[ ]> No of fraudulent transactions: 492
      No of non-fraudulent transactions: 284315

```

```
from sklearn.model_selection import train_test_split
```

```
#Train test split
```

```
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size = 0.33, random_state = 42,
```

```

#Implementing NN - with optimizer adam and loss fn = binary_crossentropy. These 2 combinator
model = keras.Sequential()

```

```

model.add(keras.layers.Dense(30, input_dim=30, activation='relu')) # kernel_initializer='
model.add(keras.layers.Dense(1, activation='sigmoid')) # kernel_initializer='
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
model.summary()

```

⏏ WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/op
Instructions for updating:
If using Keras pass *_constraint arguments to layers.
WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/op
Instructions for updating:
Use tf.where in 2.0, which has the same broadcast rule as np.where
Model: "sequential"

| Layer (type) | Output Shape | Param # |
|-------------------------|--------------|---------|
| dense (Dense) | (None, 30) | 930 |
| dense_1 (Dense) | (None, 1) | 31 |
| Total params: 961 | | |
| Trainable params: 961 | | |
| Non-trainable params: 0 | | |

```
model.fit(X_train,y_train,epochs = 10)
```

⏏ Train on 190820 samples
Epoch 1/10
190820/190820 [=====] - 9s 46us/sample - loss: 6.9182 - acc: 0.
Epoch 2/10
190820/190820 [=====] - 9s 49us/sample - loss: 5.4182 - acc: 0.
Epoch 3/10
190820/190820 [=====] - 8s 43us/sample - loss: 4.2953 - acc: 0.
Epoch 4/10
190820/190820 [=====] - 8s 42us/sample - loss: 4.3614 - acc: 0.
Epoch 5/10
190820/190820 [=====] - 8s 42us/sample - loss: 3.6643 - acc: 0.
Epoch 6/10
190820/190820 [=====] - 8s 42us/sample - loss: 2.7140 - acc: 0.
Epoch 7/10
190820/190820 [=====] - 8s 44us/sample - loss: 2.0495 - acc: 0.
Epoch 8/10
190820/190820 [=====] - 8s 43us/sample - loss: 1.4380 - acc: 0.
Epoch 9/10
190820/190820 [=====] - 8s 43us/sample - loss: 1.2719 - acc: 0.
Epoch 10/10
190820/190820 [=====] - 8s 42us/sample - loss: 1.0775 - acc: 0.
<tensorflow.python.keras.callbacks.History at 0x7f83a68decf8>

```

from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix

```

```
#Predict test set
y_predicted = model.predict(X_test.as_matrix()).T[0].astype(int)
#Printing the confusion matrix
cm = confusion_matrix(y_test, y_predicted)
print("Confusion matrix:\n%s" % cm)
```

```
↳ /usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:1: FutureWarning: Method .as_matrix() is deprecated, use .toarray() instead.
    """Entry point for launching an IPython kernel.
```

```
Confusion matrix:
```

```
[[93816    9]
 [ 141    21]]
```

```
a = 93816+21
b = 93816+9+141+21
accuracy_score = a/b
precision = 93816/(93816+141)
recall = 93816/(93816+9)
#Computing the f1 score as the harmonic mean between precision and recall
f1_score = 2*((precision*recall)/(precision+recall))
#Print Accuracy and F1 metrics
print("Accuracy: " + str(accuracy_score))
print("F1 Score: " + str(f1_score))
print(classification_report(y_test,y_predicted))
```

```
↳ Accuracy: 0.9984040346005298
   F1 Score: 0.9992012013931048
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 1.00 | 1.00 | 1.00 | 93825 |
| 1 | 0.70 | 0.13 | 0.22 | 162 |
| accuracy | | | 1.00 | 93987 |
| macro avg | 0.85 | 0.56 | 0.61 | 93987 |
| weighted avg | 1.00 | 1.00 | 1.00 | 93987 |

