```
import pandas as pd
import seaborn as sns
```

Read The DataSet

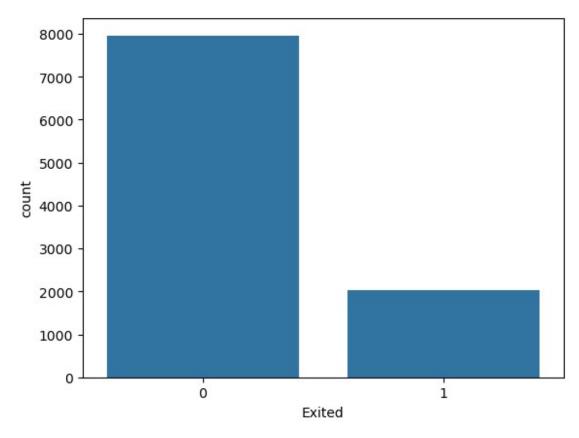
```
df = pd.read csv('Churn Modelling.csv')
df.shape
(10000, 14)
df.columns
Index(['RowNumber', 'CustomerId', 'Surname', 'CreditScore',
'Geography',
       'Gender', 'Age', 'Tenure', 'Balance', 'NumOfProducts',
'HasCrCard',
       'IsActiveMember', 'EstimatedSalary', 'Exited'],
      dtype='object')
df
      RowNumber CustomerId
                                Surname CreditScore Geography
                                                                 Gender
Age
                    15634602
                               Hargrave
                                                  619
                                                         France
                                                                 Female
0
42
              2
1
                    15647311
                                   Hill
                                                  608
                                                          Spain
                                                                 Female
41
              3
2
                    15619304
                                   Onio
                                                  502
                                                         France
                                                                 Female
42
3
                    15701354
                                   Boni
                                                  699
                                                         France
                                                                 Female
39
              5
                    15737888
                               Mitchell
                                                  850
                                                          Spain
                                                                 Female
4
43
. . .
9995
           9996
                    15606229
                               Obijiaku
                                                  771
                                                         France
                                                                   Male
39
9996
           9997
                    15569892
                              Johnstone
                                                  516
                                                         France
                                                                   Male
35
9997
           9998
                    15584532
                                    Liu
                                                  709
                                                         France
                                                                 Female
36
9998
           9999
                    15682355
                              Sabbatini
                                                  772
                                                        Germany
                                                                   Male
42
9999
          10000
                    15628319
                                 Walker
                                                  792
                                                         France
                                                                 Female
28
      Tenure
                Balance
                          NumOfProducts
                                         HasCrCard IsActiveMember \
0
                    0.00
```

9995	8 1 2 1 5 10 7 3	83807.86 59660.80 0.00 25510.82 0.00 57369.61 0.00 75075.31 30142.79		1 3 2 1 	0 1 0 1 1 1 0 1	1 0 0 1 0 1 1 0	
Esti 0 1 2 3 4 9995 9996 9997 9998 9999	101 112 113 93 79 96 101 42 92 38	348.88 542.58 931.57 826.63 084.10 270.64 699.77 085.58 888.52 190.78	ited 1 0 1 0 0 0 0 1 1 0				
<pre>df.head() RowNumb</pre>	er (ustomerId	Surname	CreditScore	Geography	Gender	Age
0	1	15634602	Hargrave	619		Female	42
1	2	15647311	Hill	608		Female	41
2	3	15619304	Onio	502		Female	42
3	4	15701354	Boni	699		Female	39
4	5	15737888	Mitchell	850	Spain	Female	43
	838 1596 1255	0.00 07.86 60.80 0.00 10.82 ary Exite	00fProducts 1 1 3 2 1	1 0 1	IsActiveMe	ember \ 1	

```
2 113931.57 1
3 93826.63 0
4 79084.10 0
```

Distinguish The Feature And Target Set

```
# input data/ Features Set
x = df[['CreditScore', 'Age', 'Tenure', 'Balance', 'NumOfProducts',
      'HasCrCard', 'IsActiveMember', 'EstimatedSalary']]
# output data/ Target Set
y = df['Exited']
      CreditScore
                    Age
                         Tenure
                                    Balance
                                              NumOfProducts
                                                              HasCrCard \
0
               619
                     42
                               2
                                        0.00
1
                               1
                                   83807.86
                                                           1
               608
                     41
                                                                       0
2
               502
                     42
                               8 159660.80
                                                           3
                                                                       1
3
                                                           2
               699
                     39
                               1
                                        0.00
4
               850
                     43
                               2 125510.82
                                                           1
                                                                       1
9995
               771
                     39
                                        0.00
                                                           2
                               5
                                                                       1
9996
               516
                     35
                              10
                                   57369.61
                                                           1
                                                                       1
9997
               709
                     36
                               7
                                        0.00
                                                           1
                                                                       0
                                                           2
                                                                       1
9998
               772
                     42
                                   75075.31
9999
               792
                     28
                                  130142.79
      IsActiveMember EstimatedSalary
0
                    1
                              101348.88
1
                    1
                              112542.58
2
                    0
                              113931.57
3
                    0
                               93826.63
4
                    1
                               79084.10
                               96270.64
9995
                    0
9996
                    1
                              101699.77
9997
                    1
                               42085.58
                    0
                               92888.52
9998
9999
                               38190.78
[10000 \text{ rows } \times 8 \text{ columns}]
sns.countplot(x=y)
<Axes: xlabel='Exited', ylabel='count'>
```



```
y.value_counts()

Exited
0 7963
1 2037
Name: count, dtype: int64
```

Normalize The Train And Test Data

Cross Validation - Train Test Data

```
# Cross Validation - Train Test Data
from sklearn.model_selection import train_test_split

x_train, x_test, y_train, y_test = train_test_split(x_scaled, y, random_state=0, test_size=0.25)

x.shape
(10000, 8)

x_test.shape
(2500, 8)

x_train.shape
(7500, 8)
```

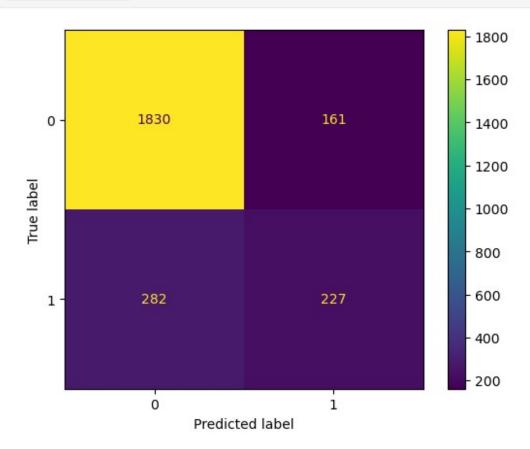
Initialize and build the model

```
# Initialize and build the model
from sklearn.neural_network import MLPClassifier
ann = MLPClassifier(hidden_layer_sizes=(100,100,100), random_state=0,
max_iter=100, activation='relu')
ann.fit(x_train, y_train)
C:\Users\Shubham\anaconda3\Lib\site-packages\sklearn\neural_network\
    multilayer_perceptron.py:691: ConvergenceWarning: Stochastic
Optimizer: Maximum iterations (100) reached and the optimization
hasn't converged yet.
    warnings.warn(
MLPClassifier(hidden_layer_sizes=(100, 100, 100), max_iter=100,
random_state=0)
```

Print Accuracy Score And Confusion Matrics

ConfusionMatrixDisplay.from_predictions(y_test, y_pred)

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x1e15a2fc6e0>



```
accuracy_score(y_test, y_pred)
0.8228
```

imbalenced Classifier is this

<pre>print(classification_report(y_test, y_pred))</pre>							
	precision	recall	f1-score	support			
0 1	0.87 0.59	0.92 0.45	0.89 0.51	1991 509			
accuracy macro avg weighted avg	0.73 0.81	0.68 0.82	0.82 0.70 0.81	2500 2500 2500			

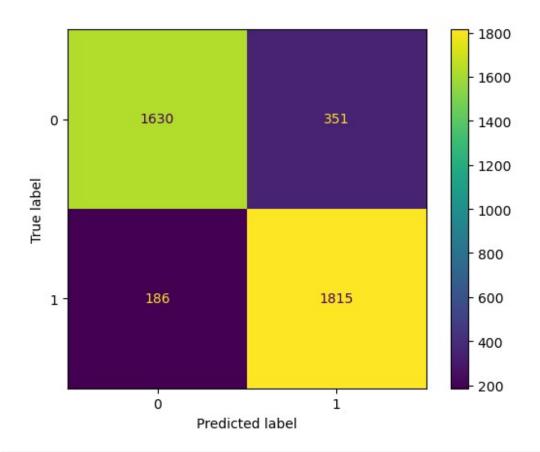
to balace the dataset

```
!pip install imbalanced-learn
Requirement already satisfied: imbalanced-learn in c:\users\shubham\
anaconda3\lib\site-packages (0.12.3)
Requirement already satisfied: numpy>=1.17.3 in c:\users\shubham\
anaconda3\lib\site-packages (from imbalanced-learn) (1.26.4)
Requirement already satisfied: scipy>=1.5.0 in c:\users\shubham\
anaconda3\lib\site-packages (from imbalanced-learn) (1.13.1)
Requirement already satisfied: scikit-learn>=1.0.2 in c:\users\
shubham\anaconda3\lib\site-packages (from imbalanced-learn) (1.4.2)
Requirement already satisfied: joblib>=1.1.1 in c:\users\shubham\
anaconda3\lib\site-packages (from imbalanced-learn) (1.4.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\
shubham\anaconda3\lib\site-packages (from imbalanced-learn) (2.2.0)
from imblearn.over sampling import RandomOverSampler
ros = RandomOverSampler(random state = 0)
x res, y res = ros.fit resample(x,y)
y_res.value_counts()
Exited
     7963
     7963
Name: count, dtype: int64
```

This above data is balanced using Over Sampler techniques

```
# Normalize
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
x scaled = scaler.fit transform(x res)
x scaled
array([[-0.29877723,
                     0.08418894, -1.01840607, ..., 0.6512495,
        1.08223556,
                    0.00817382],
       [-0.4103938, -0.01032629, -1.36135608, ..., -1.53550983,
         1.08223556, 0.20261687],
       [-1.48597169, 0.08418894, 1.03929402, ..., 0.6512495,
        -0.92401325, 0.22674468],
       [-0.84671313, 1.02934128, 0.01044398, ..., 0.6512495,
       -0.92401325, 1.28878188],
       [-0.96847667, 0.65128034, -0.67545605, ..., -1.53550983,
         1.08223556, -1.21851316],
       [-1.5874413 , 0.74579558, 1.03929402, ..., 0.6512495 ,
        1.08223556, 1.42417217]])
# Cross Validation - Train Test Data
from sklearn.model selection import train test split
x train, x test, y train, y test = train test split(x scaled, y res,
random state=0, test size=0.25)
x res.shape
(15926, 8)
x test.shape
(3982, 8)
x train.shape
(11944, 8)
# Initialize and build the model
from sklearn.neural network import MLPClassifier
ann = MLPClassifier(hidden layer_sizes=(100,100,100), random_state=0,
max iter=100, activation='relu')
ann.fit(x train, y train)
```

```
C:\Users\Shubham\anaconda3\Lib\site-packages\sklearn\neural network\
multilayer perceptron.py:691: ConvergenceWarning: Stochastic
Optimizer: Maximum iterations (100) reached and the optimization
hasn't converged yet.
 warnings.warn(
MLPClassifier(hidden layer sizes=(100, 100, 100), max iter=100,
random state=0)
y pred = ann.predict(x test)
y pred
array([1, 1, 1, ..., 1, 1, 1], dtype=int64)
from sklearn.metrics import ConfusionMatrixDisplay,
classification_report, accuracy score
y test.value counts()
Exited
     2001
     1981
Name: count, dtype: int64
ConfusionMatrixDisplay.from predictions(y test, y pred)
<sklearn.metrics. plot.confusion matrix.ConfusionMatrixDisplay at</pre>
0x1e15a48b260>
```



accuracy_score(y_test, y_pred)
0.865143144148669

balenced Classifier is this

<pre>print(classification_report(y_test, y_pred))</pre>						
	precision	recall	f1-score	support		
0 1	0.90 0.84	0.82 0.91	0.86 0.87	1981 2001		
accuracy macro avg weighted avg	0.87 0.87	0.86 0.87	0.87 0.86 0.86	3982 3982 3982		