



MACHINE LEARNING ASSIGNMENT # 03

Binary Classification



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Subject: Machine Learning (MS DS)

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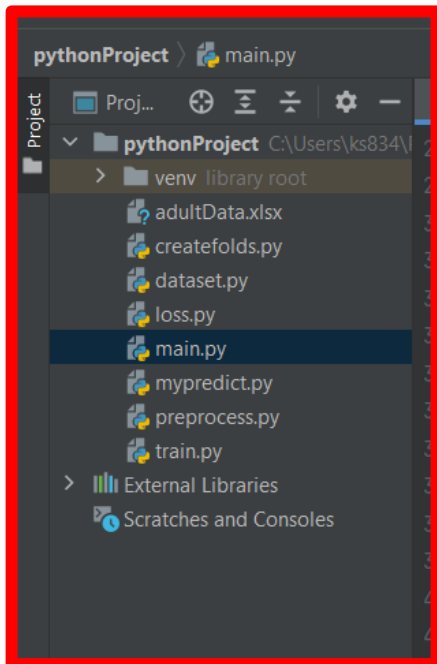
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2 ASSIGNMENT QUESTION:

Use an IDE with a somewhat similar directory/file structure as given below. Try different possibilities to improve accuracy as much as possible. Report accuracies, precision, recall, f1-scores for 5-fold and 10-fold cross validation. Its a binary class classification problem with labels as individuals earning above 50K per year or below.

3 DIRECTORY/FILE STRUCTURE (IDE PYCHARM)



4 ABOUT DATA (DISCUSSION):

1. Training and testing Data are separate. Note: I have set 0 for $\leq 50k$ and 1 for $> 50k$ in excel file before loading into this project.

5 STEPS (DISCUSSION):

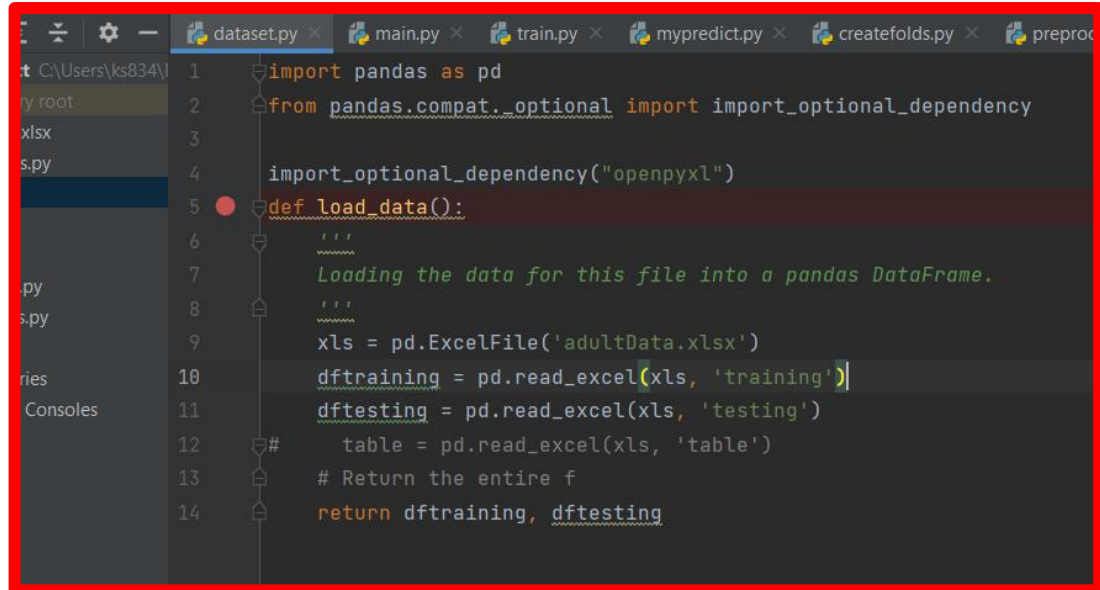
It's to be noted that we have defined a function for each step and we have made the generalized code.

5.1 IMPORTING LIBRARIES

In this step we just imported the all libraries which is being used in assignment

5.2 LOADING THE DATA (DATASET.PY)

Here we just imported / read our xlsx file.

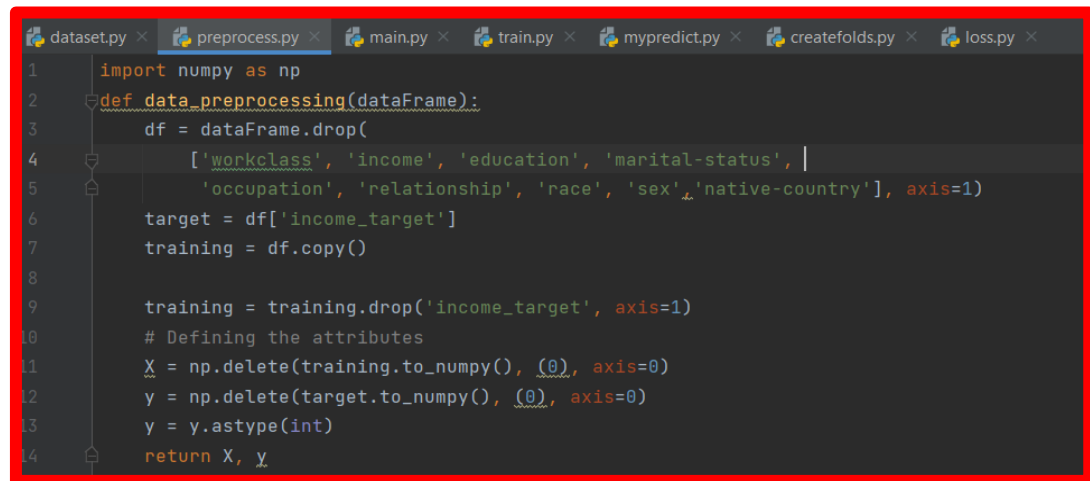


```

1 import pandas as pd
2 from pandas.compat.optional import import_optional_dependency
3
4 import_optional_dependency("openpyxl")
5
6 def load_data():
7     """
8     Loading the data for this file into a pandas DataFrame.
9     """
10    xls = pd.ExcelFile('adultData.xlsx')
11    dftraining = pd.read_excel(xls, 'training')
12    dftesting = pd.read_excel(xls, 'testing')
13    # table = pd.read_excel(xls, 'table')
14    # Return the entire f
15    return dftraining, dftesting
  
```

5.3 DATA PREPROCESSING (PREPROCESSING.PY)

In Data Preprocessing we just convert our data into NumPy Array and then we have just changed our data label is 0 for $\leq 50k$ and 1 for $> 50k$ respectively for smooth classification and evaluation.

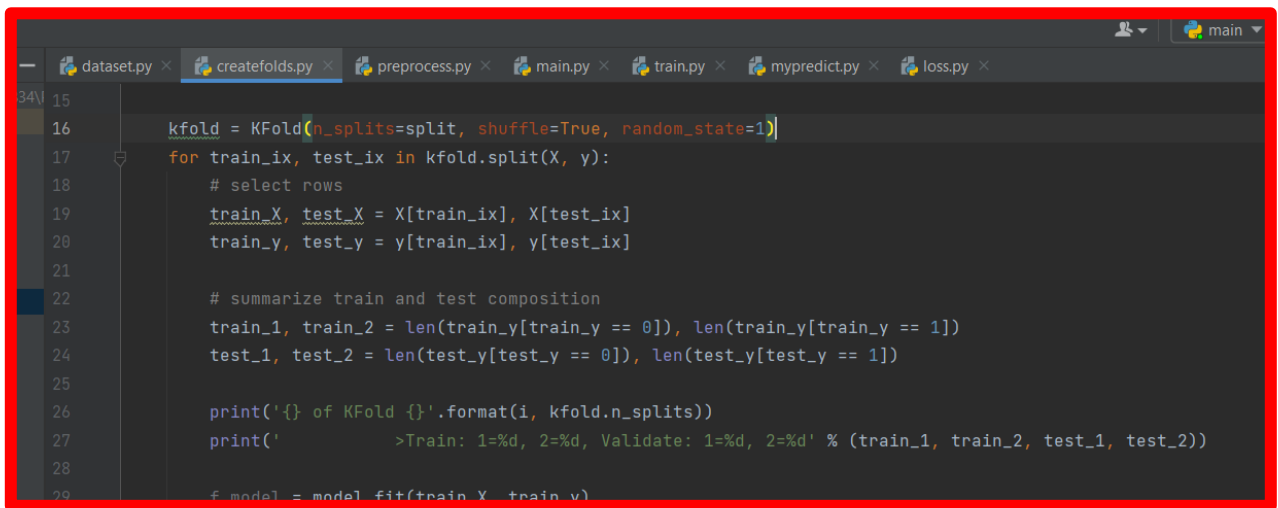


```

1 import numpy as np
2
3 def data_preprocessing(dataFrame):
4     df = dataFrame.drop(
5         ['workclass', 'income', 'education', 'marital-status',
6          'occupation', 'relationship', 'race', 'sex', 'native-country'], axis=1)
7     target = df['income_target']
8     training = df.copy()
9
10    training = training.drop('income_target', axis=1)
11    # Defining the attributes
12    X = np.delete(training.to_numpy(), (0), axis=0)
13    y = np.delete(target.to_numpy(), (0), axis=0)
14    y = y.astype(int)
15    return X, y
  
```

5.4 CREATE FOLDS (CREATESFOLDS.PY)

Here we have defined our training and testing generalized function/file for classifier.



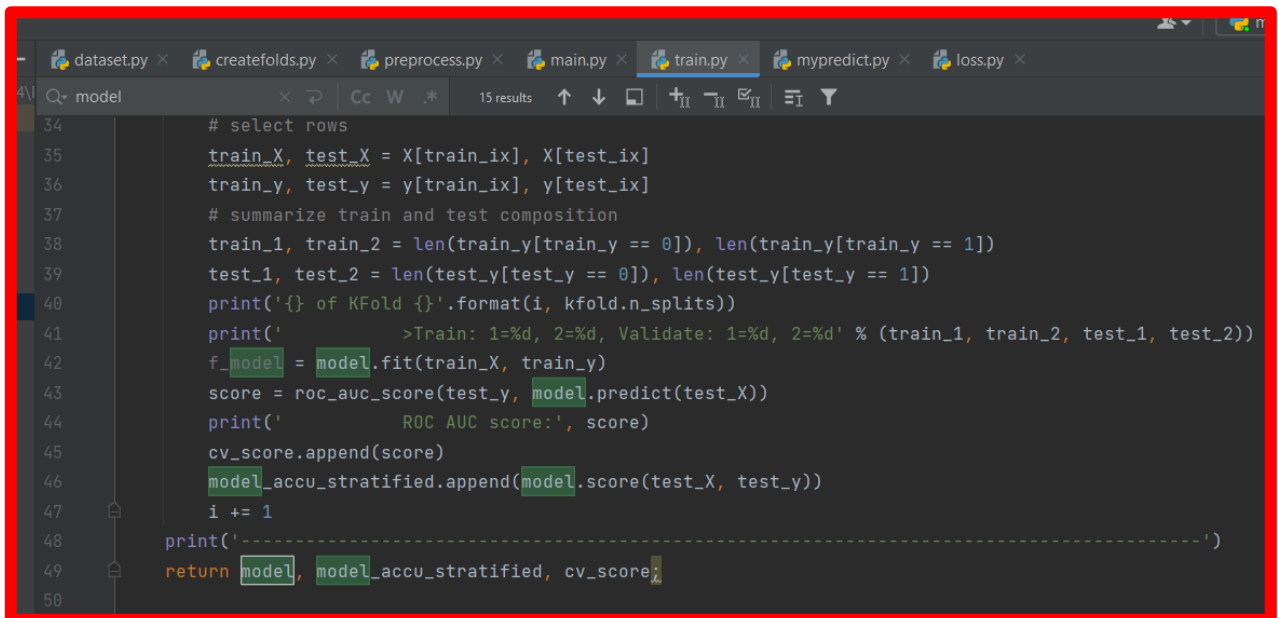
```

15
16 kfold = KFold(n_splits=split, shuffle=True, random_state=1)
17 for train_ix, test_ix in kfold.split(X, y):
18     # select rows
19     train_X, test_X = X[train_ix], X[test_ix]
20     train_y, test_y = y[train_ix], y[test_ix]
21
22     # summarize train and test composition
23     train_1, train_2 = len(train_y[train_y == 0]), len(train_y[train_y == 1])
24     test_1, test_2 = len(test_y[test_y == 0]), len(test_y[test_y == 1])
25
26     print('{} of KFold {}'.format(i, kfold.n_splits))
27     print('
>Train: 1=%d, 2=%d, Validate: 1=%d, 2=%d' % (train_1, train_2, test_1, test_2))
28
29     f_model = model.fit(train_X, train_y)

```

5.5 TRAINING (TRAIN.PY)

Here we have defined our training and testing generalized function for classifier.



```

34 # select rows
35 train_X, test_X = X[train_ix], X[test_ix]
36 train_y, test_y = y[train_ix], y[test_ix]
37 # summarize train and test composition
38 train_1, train_2 = len(train_y[train_y == 0]), len(train_y[train_y == 1])
39 test_1, test_2 = len(test_y[test_y == 0]), len(test_y[test_y == 1])
40 print('{} of KFold {}'.format(i, kfold.n_splits))
41 print('
>Train: 1=%d, 2=%d, Validate: 1=%d, 2=%d' % (train_1, train_2, test_1, test_2))
42 f_model = model.fit(train_X, train_y)
43 score = roc_auc_score(test_y, model.predict(test_X))
44 print('
ROC AUC score:', score)
45 cv_score.append(score)
46 model_accu_stratified.append(model.score(test_X, test_y))
47 i += 1
48 print('-----')
49 return model, model_accu_stratified, cv_score
50

```

5.6 TESTING (MYPREDICT.PY)

Here we have defined testing function for classifier in my predict file. Here we have a function for evaluation report

```

8 import seaborn as sns
9
10 def Evaluation(model: object, model_accu_stratified: object, cv_score: object,
11               test_X: object, test_y: object, classifierName: object) -> object:
12     print('Possible accuracy are :', (round(model_accu_stratified[0], 5)))
13     print('\nMaximum Accuracy That can be obtained from this model is:',
14           max(model_accu_stratified) * 100, '%')
15     print('\nMinimum Accuracy:',
16           min(model_accu_stratified) * 100, '%')
17     print('\nOverall (Mean) Accuracy:',
18           mean(model_accu_stratified) * 100, '%')
19     # print('\nStandard Deviation is:', stdev(model_accu_stratified))
20     print('-----')
21     print('Cv: ', cv_score, '\nMean cv Score :', np.mean(cv_score))
22     print('-----')
23     print("\n          Confusion Matrix on tested data\n")
24     cm = confusion_matrix(test_y, model.predict(test_X))
25     print('-----')
26     tp, fn, fp, tn = confusion_matrix(test_y, model.predict(test_X)).reshape(-1)
27     # classification report for precision, recall f1-score and accuracy
28     matrix = classification_report(test_y, model.predict(test_X))
29     print('Classification report : \n', matrix)
30     print('-----')
31     plt.figure(figsize=(9, 9))
32     sns.heatmap(cm, annot=True, fmt=".3f", linewidths=.5, square=True, cmap='Blues_r')
33     plt.ylabel('Actual Label')
34     plt.xlabel('Predicted label')

```

5.7 DRIVER CLASS (__MAIN__.PY)

Finally, here the driver file from where we will drive the whole code.

```

dataset.py × createfolds.py × preprocess.py × main.py × train.py × mypredict.py × loss.py ×
# Press the green button in the gutter to run the script.
from loss import my_custom_loss_func

if __name__ == '__main__':
    print('Stat')
    # read data
    print('Load data')
    df1, df2 = dataset.load_data()
    # DataPreprocessing
    print('Data Preprocessing')
    train_X, train_y = preprocess.data_preprocessing(df1)
    test_X, test_y = preprocess.data_preprocessing(df2)
    print('Training')
    # Create classifier object.
    # Create classifier object.
    Ada = AdaBoostClassifier(n_estimators=50, learning_rate=1.0, algorithm='SAMME.R')
    print("***** 5 k fold *****")

    model1, model_accu_stratified, cv_score = train.ApplyClassifier(Ada, train_X, train_y, 5)
    print("Accuracy with 5k fold ", model1.score(test_X, test_y))
    mypredict.Evaluation(model1, model_accu_stratified, cv_score, test_X, test_y, type(model1).__name__)
    print("***** 10 k fold *****")
    model2, model_accu_stratified, cv_score = train.ApplyClassifier(Ada, train_X, train_y, 10)
    print("Accuracy with 5k fold ", model2.score(test_X, test_y))
    mypredict.Evaluation(model2, model_accu_stratified, cv_score, test_X, test_y, type(model2).__name__)

    # print ("Loss: with 5k fold ", my_custom_loss_func(test_y, Ada5.predict(test_X) ))

```

6 ADA BOOST CLASSIFIER (TESTING):

Classifier Name: AdaBoostClassifier

***** 5 k fold *****

1 of KFold 5

>Train: 1=19729, 2=6319, Validate: 1=4990, 2=1522
 ROC AUC score: 0.6988561880660138

2 of KFold 5

>Train: 1=19798, 2=6250, Validate: 1=4921, 2=1591
 ROC AUC score: 0.7092361639485262

3 of KFold 5

>Train: 1=19806, 2=6242, Validate: 1=4913, 2=1599
 ROC AUC score: 0.6917904369041968

4 of KFold 5

>Train: 1=19766, 2=6282, Validate: 1=4953, 2=1559
 ROC AUC score: 0.6915442620543306

5 of KFold 5

>Train: 1=19777, 2=6271, Validate: 1=4942, 2=1570
 ROC AUC score: 0.6980985804761991

Accuracy with 5k fold 0.8345823095823096

Possible accuracy is: 0.84045

Maximum Accuracy can be obtained from this model is: 84.0448402948403 %

Minimum Accuracy: 82.83169533169533 %

Overall (Mean) Accuracy: 83.41523341523342 %

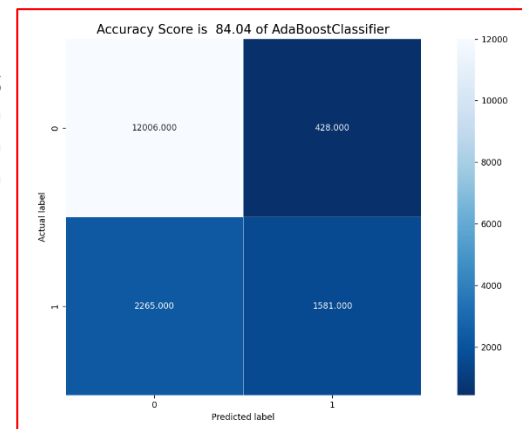
Cv: [0.6988561880660138, 0.7092361639485262, 0.6917904369041968, 0.6915442620543306, 0.6980985804761991]

Mean cv Score: 0.6979051262898534

Confusion Matrix on tested data

Classification report:

	precision	recall	f1-score	support
0	0.84	0.97	0.90	12434
1	0.79	0.41	0.54	3846
accuracy			0.83	16280
macro avg	0.81	0.69	0.72	16280
weighted avg	0.83	0.83	0.81	16280



Machine Learning (Binary Classification Assignment)

Classifier Name: AdaBoostClassifier

******* 10 k fold *******

1 of KFold 10

>Train: 1=22212, 2=7092, Validate: 1=2507, 2=749
ROC AUC score: 0.6906576139546253

2 of KFold 10

>Train: 1=22236, 2=7068, Validate: 1=2483, 2=773
ROC AUC score: 0.7040996499352128

3 of KFold 10

>Train: 1=22256, 2=7048, Validate: 1=2463, 2=793
ROC AUC score: 0.6979109227666564

4 of KFold 10

>Train: 1=22261, 2=7043, Validate: 1=2458, 2=798
ROC AUC score: 0.6849033690817768

5 of KFold 10

>Train: 1=22264, 2=7040, Validate: 1=2455, 2=801
ROC AUC score: 0.7183815546249469

6 of KFold 10

>Train: 1=22261, 2=7043, Validate: 1=2458, 2=798
ROC AUC score: 0.6851067864943072

7 of KFold 10

>Train: 1=22228, 2=7076, Validate: 1=2491, 2=765
ROC AUC score: 0.695922576176195

8 of KFold 10

>Train: 1=22257, 2=7047, Validate: 1=2462, 2=794
ROC AUC score: 0.6895102791652257

9 of KFold 10

>Train: 1=22260, 2=7044, Validate: 1=2459, 2=797
ROC AUC score: 0.6915035184299807

10 of KFold 10

>Train: 1=22236, 2=7068, Validate: 1=2483, 2=773
ROC AUC score: 0.7047037578691636

Accuracy with 10k fold 0.8347051597051597

Possible accuracy are : 0.83937

Maximum Accuracy can be obtained from this model is: 84.24447174447175 %

Minimum Accuracy: 82.7088452088452 %

Overall(Mean) Accuracy: 83.49201474201475 %

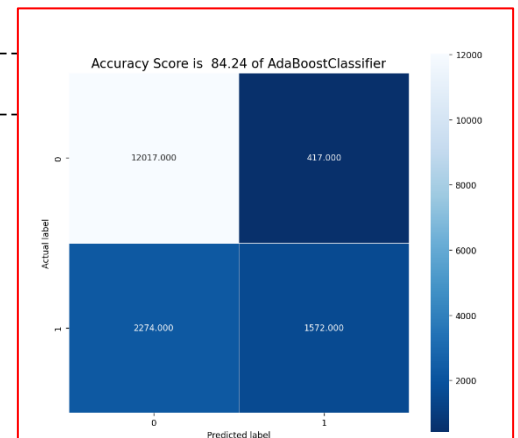
Cv: [0.6906576139546253, 0.7040996499352128, 0.6979109227666564, 0.6849033690817768, 0.7183815546249469, 0.6851067864943072, 0.695922576176195, 0.6895102791652257, 0.6915035184299807, 0.7047037578691636]

Mean cv Score : 0.696270002849809

Confusion Matrix on tested data

Classification report:

	precision	recall	f1-score	support
0	0.84	0.97	0.90	12434
1	0.79	0.41	0.54	3846
accuracy			0.83	16280
macro avg	0.82	0.69	0.72	16280
weighted avg	0.83	0.83	0.81	16280



7 LOGISTIC REGRESSION (TESTING):

Classifier Name: LogisticRegression

***** 5 k fold *****

1 of KFold 5

>Train: 1=19729, 2=6319, Validate: 1=4990, 2=1522
 ROC AUC score: 0.6142044930860406

2 of KFold 5

>Train: 1=19798, 2=6250, Validate: 1=4921, 2=1591
 ROC AUC score: 0.617316389654211

3 of KFold 5

>Train: 1=19806, 2=6242, Validate: 1=4913, 2=1599
 ROC AUC score: 0.6127285817629505

4 of KFold 5

>Train: 1=19766, 2=6282, Validate: 1=4953, 2=1559
 ROC AUC score: 0.6150461548303896

5 of KFold 5

>Train: 1=19777, 2=6271, Validate: 1=4942, 2=1570
 ROC AUC score: 0.6114479555196973

 Accuracy with 5k fold 0.802027027027027

Possible accuracy are : 0.80344

Maximum Accuracy That can be obtained from this model is: 80.34398034398035 %

Minimum Accuracy: 79.43796068796068 %

Overall(Mean) Accuracy: 79.71130221130221 %

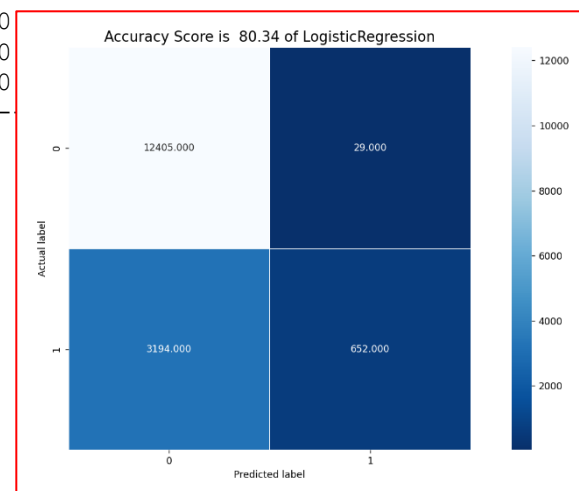
 Cv: [0.6142044930860406, 0.617316389654211, 0.6127285817629505, 0.6150461548303896, 0.6114479555196973]

Mean cv Score : 0.6141487149706577

 Confusion Matrix on tested data

 Classification report :

	precision	recall	f1-score	support
0	0.80	1.00	0.89	12434
1	0.96	0.17	0.29	3846
accuracy			0.80	16280
macro avg	0.88	0.58	0.59	16280
weighted avg	0.83	0.80	0.74	16280



Machine Learning (Binary Classification Assignment)

Classifier Name: LogisticRegression

******* 10 k fold *******

1 of KFold 10

>Train: 1=22212, 2=7092, Validate: 1=2507, 2=749
ROC AUC score: 0.5859039282798552

2 of KFold 10

>Train: 1=22236, 2=7068, Validate: 1=2483, 2=773
ROC AUC score: 0.6254731397305038

3 of KFold 10

>Train: 1=22256, 2=7048, Validate: 1=2463, 2=793
ROC AUC score: 0.6132186370899655

4 of KFold 10

>Train: 1=22261, 2=7043, Validate: 1=2458, 2=798
ROC AUC score: 0.6198949366907912

5 of KFold 10

>Train: 1=22264, 2=7040, Validate: 1=2455, 2=801
ROC AUC score: 0.624957601368961

6 of KFold 10

>Train: 1=22261, 2=7043, Validate: 1=2458, 2=798
ROC AUC score: 0.602774226045178

7 of KFold 10

>Train: 1=22228, 2=7076, Validate: 1=2491, 2=765
ROC AUC score: 0.615721171380368

8 of KFold 10

>Train: 1=22257, 2=7047, Validate: 1=2462, 2=794
ROC AUC score: 0.6085404956343986

9 of KFold 10

>Train: 1=22260, 2=7044, Validate: 1=2459, 2=797
ROC AUC score: 0.6050676515175094

10 of KFold 10

>Train: 1=22236, 2=7068, Validate: 1=2483, 2=773
ROC AUC score: 0.6223121886004651

Accuracy with 10k fold 0.8020884520884521

Possible accuracy are: 0.79484

Maximum Accuracy can be obtained from this model is: 80.37469287469288 %

Minimum Accuracy: 78.83906633906635 %

Overall (Mean) Accuracy: 79.60995085995086 %

Cv: [0.5859039282798552, 0.6254731397305038, 0.6132186370899655, 0.6198949366907912, 0.624957601368961, 0.602774226045178, 0.615721171380368, 0.6085404956343986, 0.6050676515175094, 0.6223121886004651]

Mean cv Score: 0.6123863976337995

Confusion Matrix on tested data

Classification report:

	precision	recall	f1-score	support
0	0.80	1.00	0.89	12434
1	0.95	0.17	0.29	3846
accuracy			0.80	16280
macro avg	0.87	0.58	0.59	16280
weighted avg	0.83	0.80	0.74	16280

