Implementation of Logistic Regression Model to Predict the Placement Status of Student

Aim:

To write a program to implement the Logistic Regression Model to Predict the Placement Status of Student.

'Equipments Required:

- 1. Hardware PCs
- 2. Anaconda Python 3.7 Installation / Moodle-Code Runner

² Algorithm

- 1. Import the standard libraries.
- 2. Upload the dataset and check for any null or duplicated values using .isnull() and .duplicated() function respectively.
- 3. Import LabelEncoder and encode the dataset.
- 4. Import LogisticRegression from sklearn and apply the model on the dataset.
- 5. Predict the values of array.
- 6. Calculate the accuracy, confusion and classification report by importing the required modules from sklearn.
- 7. Apply new unknown values

Program:

```
/*
Program to implement the Logistic Regression Model to Predict the Placement Status of
Student.
Developed by: SANJAI A
RegisterNumber: 212220040142
*/
import pandas as pd
data=pd.read_csv("Placement_Data.csv")
data.head()
data1=data1.copy()
data1=data1.drop(["sl_no","salary"],axis=1)
data1.head()
data1=data1.drop(["sl_no","salary"],axis=1)
data1.head()
data1=data1.drop(["sl_no","salary"],axis=1)
data1.head()
data1.head()
data1.inull().sum()
```

```
data1.head()
data.duplicated().sum()
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
data1["gender"]=le.fit_transform(data1["gender"])
data1["ssc_b"]=le.fit_transform(data1["ssc_b"])
data1["hsc_b"]=le.fit_transform(data1["hsc_b"])
data1["hsc_s"]=le.fit_transform(data1["hsc s"])
data1["degree_t"]=le.fit_transform(data1["degree_t"])
data1["workex"]=le.fit_transform(data1["workex"])
data1["specialisation"]=le.fit_transform(data1["specialisation"])
data1["status"]=le.fit_transform(data1["status"])
print (data1)
x=data1.iloc[:,:-1]
y=data1["status"]
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
from sklearn.linear_model import LogisticRegression
c=LogisticRegression(solver="liblinear")
c.fit(x_train,y_train)
y_pred=c.predict(x_test)
print(y_pred)
from sklearn.metrics import accuracy_score
acur=accuracy_score(y_test,y_pred)
acur
from sklearn.metrics import confusion_matrix
con=confusion_matrix(y_test,y_pred)
print(con)
from sklearn.metrics import classification_report
class_report=classification_report(y_test,y_pred)
print(class_report)
print(c.predict([[1,80,1,90,1,1,90,1,0,85,1,85]]))
```

[°]Output:

```
gender ssc_p ssc_b hsc_p hsc_b hsc_s degree_p degree_t workex \
                                                     2
       1 67.00 1 91.00 1 1 58.00
                                                                  0
                                        2
          1 79.33
                      0 78.33
                                   1
                                              77.48
                                                          2
                                                                  1
₽
          1 65.00
                      0 68.00
                                        0 64.00
                       0 52.00
                                   0
                                        2 52.00
   3
           1 56.00
                                                          2
                                                                  0
                                        1
                                  0
                                            73.30
                                                          0
           1 85.80
                       0 73.60
   4
                                                                  0
                    1 82.00
1 60
   . .
          . . .
               . . .
                                  ...
                                                . . .
                                                         . . .
                                                                . . .
                      1 82.00 1 1 77.60
1 60.00 1 2 72.00
1 67.00 1 1 73.00
1 66.00 1 1 58.00
0 58.00 1 2 53.00
   210
           1 80.60
                                                          0
                                                                  0
   211
          1 58.00
                                                          2
                                                                  0
   212
          1 67.00
                                                          0
                                                                 1
   213
          0 74.00
                                                          0
                                                           0
   214
          1 62.00
                                                                 0
       etest_p specialisation mba_p status
                1 58.80
   0
         55.0
   1
          86.5
                         0 66.28
   2
         75.0
                         0 57.80
   3
        66.0
                         1 59.43
        96.8
                         0 55.50
                                      1
          . . .
                        . . .
                              . . .
                         0 74.49
   210
          91.0
         74.0
   211
                         0 53.62
                                       1
                         0 69.72
                                      1
   212
         59.0
   213
         70.0
                         1 60.23
                                      1
        89.0
                          1 60.22
   214
   [215 rows x 13 columns]
   [0\ 1\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1
    111001]
   [[11 5]
  [ 3 24]]
              precision
                         recall f1-score
                                            support
           0
                   0.79
                            0.69
                                     0.73
                                                 16
           1
                   0.83
                            0.89
                                     0.86
                                                 27
     accuracy
                                     0.81
                                                43
```

[']Result:

macro avg

weighted avg

0.81

0.81

0.79

0.81

Thus the program to implement the Logistic Regression Model to Predict the Placement Status of Student is written and verified using python programming.

0.80

0.81

43

43