

## PRACTICAL NO 5

### Code :

```
import pandas as pd
import numpy as np
import statsmodels.api as sm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix,
classification_report

# Load the dataset
url = "https://stats.idre.ucla.edu/stat/data/binary.csv"
admissions_data = pd.read_csv(url)

# Preprocess the data
# For simplicity, we'll assume there are no missing values and the data is
clean
# Define predictor (X) and target (y) variables
X = admissions_data[['gre', 'gpa', 'rank']]
y = admissions_data['admit']

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,
random_state=42)

# Perform logistic regression
# Add intercept term
X_train = sm.add_constant(X_train)
X_test = sm.add_constant(X_test)

# Fit logistic regression model
logit_model = sm.Logit(y_train, X_train)
result = logit_model.fit()

# Print summary of the model
print(result.summary())

# Evaluate the model
# Predict on test set
y_pred = result.predict(X_test)
y_pred_class = np.where(y_pred > 0.5, 1, 0) # Convert predicted
probabilities to binary classes

# Model accuracy
accuracy = accuracy_score(y_test, y_pred_class)
print("\nAccuracy:", accuracy)

# Confusion matrix
conf_matrix = confusion_matrix(y_test, y_pred_class)
print("\nConfusion Matrix:")
print(conf_matrix)

# Classification report
class_report = classification_report(y_test, y_pred_class)
print("\nClassification Report:")
print(class_report)
```

Output :

Current function value: 0.577245

Iterations 6

Logit Regression Results

Dep. Variable:	admit	No. Observations:	280
Model:	Logit	Df Residuals:	276
Method:	MLE	Df Model:	3
Date:	Tue, 13 Feb 2024	Pseudo R-squ.:	0.07268
Time:	08:47:20	Log-Likelihood:	-161.63
converged:	True	LL-Null:	-174.30
Covariance Type:	nonrobust	LLR p-value:	1.314e-05

	coef	std err	z	P> z	[0.025	0.975]
const	-3.4637	1.368	-2.532	0.011	-6.145	-0.782
gre	0.0016	0.001	1.218	0.223	-0.001	0.004
gpa	0.8822	0.390	2.265	0.024	0.119	1.646
rank	-0.5205	0.151	-3.457	0.001	-0.816	-0.225

Accuracy: 0.725

Confusion Matrix:

```
[[78 3]
 [30 9]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.72	0.96	0.83	81
1	0.75	0.23	0.35	39
accuracy			0.73	120
macro avg	0.74	0.60	0.59	120
weighted avg	0.73	0.72	0.67	120