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In [1]:
                                                                                         M
import pandas as pd
import numpy as np
from sklearn import *
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import accuracy score, precision score, recall score
from sklearn.metrics import f1_score
In [2]:
                                                                                         M
df = pd.read_csv("titanic_train.csv")
In [3]:
                                                                                         M
df.dtypes
Out[3]:
passenger_id
                  int64
pclass
                  int64
name
                 object
                 object
sex
age
                float64
                  int64
sibsp
parch
                  int64
                 object
ticket
fare
                float64
                 object
cabin
embarked
                 object
                 object
boat
body
                float64
home.dest
                 object
                  int64
survived
dtype: object
                                                                                         M
In [4]:
label_encoder = LabelEncoder()
df['sex'] = label_encoder.fit_transform(df['sex'])
df['embarked'] = label_encoder.fit_transform(df['embarked'].fillna('Unknown'))
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In [5]:
# Fill in missing values with median values
df['age'] = df['age'].fillna(df['age'].median())
df['fare'] = df['fare'].fillna(df['fare'].median())
# Display the preprocessed DataFrame
print(df.head())
   passenger_id pclass
                                                                        nam
e
           1216
                       3
0
                                                          Smyth, Miss. Juli
а
                       3
                                                             Cacic, Mr. Luk
            699
1
а
                          Van Impe, Mrs. Jean Baptiste (Rosalie Paula G
2
           1267
0...
                       2
                                      Hocking, Mrs. Elizabeth (Eliza Need
3
            449
s)
4
            576
                       2
                                                             Veal, Mr. Jame
S
                                        fare cabin
                                                    embarked boat
   sex
         age
              sibsp
                     parch ticket
                                                                    body
       28.0
                                      7.7333
0
                  0
                          0
                            335432
                                               NaN
                                                            1
                                                                13
                                                                     NaN
       38.0
                                      8.6625
                                                            2
                                                              NaN
                                                                     NaN
     1
                  0
                            315089
                                               NaN
1
                          0
2
     0
        30.0
                  1
                          1
                             345773
                                     24.1500
                                               NaN
                                                            2
                                                               NaN
                                                                     NaN
3
       54.0
                  1
                              29105
                                     23.0000
                                               NaN
                                                            2
                                                                 4
                                                                     NaN
     0
                          3
4
     1 40.0
                          0
                              28221 13.0000
                                               NaN
                                                            2
                                                              NaN
                                                                     NaN
                  home.dest survived
0
                        NaN
                                     1
1
                     Croatia
                                     0
2
                        NaN
                                     0
3
       Cornwall / Akron, OH
                                     1
   Barre, Co Washington, VT
In [6]:
                                                                                         M
scaler = StandardScaler()
numerical_cols = ['age', 'fare']
df[numerical cols] = scaler.fit transform(df[numerical cols])
In [7]:
                                                                                         H
features = ['pclass', 'sex', 'age', 'sibsp', 'parch', 'fare', 'embarked']
target = 'survived'
X_train, X_test, y_train, y_test = train_test_split(df[features], df[target])
# Display the shapes of the resulting data splits
print('X_train shape:', X_train.shape)
print('y_train shape:', y_train.shape)
print('X_test shape:', X_test.shape)
print('y_test shape:', y_test.shape)
X_train shape: (637, 7)
y_train shape: (637,)
X_test shape: (213, 7)
y_test shape: (213,)
```

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In [8]:
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mlp = MLPClassifier(hidden_layer_sizes=(10,), max_iter=1000, random_state=42)
mlp.fit(X_train, y_train)
# Print the accuracy of the classifier on the training data
train_accuracy = mlp.score(X_train, y_train)
print('Training accuracy:', train_accuracy)
Training accuracy: 0.8288854003139717
In [9]:
                                                                                       M
y_pred = mlp.predict(X_test)
# calculate evaluation metrics
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)
# print the results
print("Accuracy: {:.3f}".format(accuracy))
print("Precision: {:.3f}".format(precision))
print("Recall: {:.3f}".format(recall))
Accuracy: 0.808
```

Precision: 0.828 Recall: 0.639

In []: