```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
# Load dataset
df = pd.read_csv("/content/heart.csv")
# Preview data
print(df.head())
print(df.info())
       sex cp trestbps chol fbs restecg thalach exang oldpeak slope
                                0
         1
                    125
                          212
                                                               1.0
   53
                          203
                                                155
                                                               3.1
1
         1
            0
                    140
                                1
                                         0
                                                        1
                                                                       0
                    145
2
   70
         1 0
                         174
                                0
                                         1
                                                125
                                                        1
                                                               2.6
                                                                       0
   61
         1
            0
                    148
                          203
                                0
                                                161
                                                        0
                                                               0.0
                                                                       2
4
         0 0
                    138 294
                                                106
                                                               1.9
   62
                                1
                                         1
                                                        0
                                                                       1
   ca thal
            target
  2
1
   0
         3
                0
   0
         3
                0
                 0
   3
        2
                0
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
Data columns (total 14 columns):
# Column Non-Null Count Dtype
0
              1025 non-null
    age
             1025 non-null
1
                            int64
    sex
 2
    ср
              1025 non-null
                            int64
    trestbps 1025 non-null
                             int64
    chol
             1025 non-null
                            int64
5
    fbs
              1025 non-null
                            int64
    restecg
             1025 non-null
                             int64
    thalach 1025 non-null
                            int64
              1025 non-null
 8
    exang
                             int64
 9
    oldpeak
             1025 non-null
                             float64
 10 slope
              1025 non-null
                             int64
              1025 non-null
 11 ca
                             int64
 12 thal
             1025 non-null
                             int64
 13 target
             1025 non-null
                             int64
dtypes: float64(1), int64(13)
memory usage: 112.2 KB
None
print(df.isnull().sum()) # Check missing values
```

```
print(df.isnull().sum())  # Check missing values
print(df.describe())  # Statistical summary

# Count of target classes
sns.countplot(x="target", data=df)
plt.show()
```

```
age
            0
sex
ср
            0
trestbps
            0
chol
fbs
            0
restecg
            0
thalach
            0
exang
oldpeak
            0
slope
            0
ca
thal
            0
target
dtype: int64
                                                  trestbps
                                                                   chol
               age
                             sex
                                           ср
                    1025.000000 1025.000000 1025.000000 1025.00000
count 1025.000000
         54.434146
                       0.695610
                                     0.942439
                                                131.611707
                                                              246.00000
std
          9.072290
                       0.460373
                                     1.029641
                                                 17.516718
                                                               51.59251
                       0.000000
                                     0.000000
                                                 94.000000
min
         29.000000
                                                              126,00000
25%
         48.000000
                       0.000000
                                     0.000000
                                                 120.000000
                                                              211.00000
50%
         56.000000
                       1.000000
                                     1.000000
                                                130.000000
                                                              240.00000
         61.000000
                       1.000000
                                     2.000000
                                                140.000000
                                                              275.00000
75%
max
         77.000000
                       1.000000
                                     3.000000
                                                200.000000
                                                              564.00000
               fbs
                                      thalach
                                                                 oldpeak
                         restecg
                                                      exang
      1025.000000
                    1025.000000
                                  1025.000000
                                               1025.000000
                                                             1025.000000
count
          0.149268
                       0.529756
                                   149.114146
                                                  0.336585
                                                                1.071512
std
          0.356527
                       0.527878
                                    23.005724
                                                  0.472772
                                                                1.175053
          0.000000
                                                  0.000000
                                                                0.000000
min
                       0.000000
                                    71.000000
25%
          0.000000
                       0.000000
                                   132.000000
                                                  0.000000
                                                                0.000000
50%
          0.000000
                        1.000000
                                   152.000000
                                                   0.000000
                                                                0.800000
                                                                1.800000
75%
          0.000000
                       1.000000
                                   166.000000
                                                  1.000000
                                                                6.200000
max
          1.000000
                       2.000000
                                   202.000000
                                                  1.000000
             slope
                                         thal
                                                     target
                              ca
count 1025.000000
                    1025.000000
                                 1025.000000 1025.000000
mean
          1.385366
                       0.754146
                                     2.323902
                                                  0.513171
          0.617755
                       1.030798
                                     0.620660
                                                   0.500070
std
                       0.000000
                                     0.000000
                                                  0.000000
          0.000000
min
25%
          1.000000
                       0.000000
                                     2.000000
                                                  0.000000
50%
          1.000000
                        0.000000
                                     2.000000
                                                   1.000000
75%
          2.000000
                       1.000000
                                     3.000000
                                                   1.000000
          2.000000
                       4.000000
                                     3.000000
                                                  1.000000
max
    500
    400
 count count
    200
    100
      0
                                       target
```

```
X = df.drop("target", axis=1)  # Features
y = df["target"]  # Target
```

```
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42, stratify=y
)
```

```
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
```

```
X_test = scaler.transform(X_test)
```

```
model = LogisticRegression()
model.fit(X_train, y_train)
```

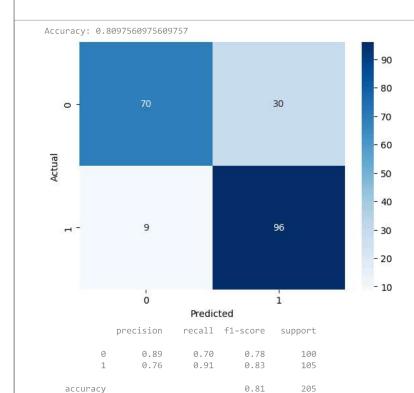
v LogisticRegression ① ??
LogisticRegression()

```
y_pred = model.predict(X_test)
```

```
print("Accuracy:", accuracy_score(y_test, y_pred))

# Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.show()

# Classification Report
print(classification_report(y_test, y_pred))
```



0.81

0.81

0.81

0.81

205

0.82

0.82

macro avg

weighted avg

```
# Example input (values must match dataset features order)

new_data = np.array([[63,1,3,145,233,1,0,150,0,2.3,0,0,1]])

# Scale input

new_data_scaled = scaler.transform(new_data)

# Prediction

prediction = model.predict(new_data_scaled)

print("Prediction (1=Heart Disease, 0=No Disease):", prediction)

Prediction (1=Heart Disease, 0=No Disease): [1]

/usr/local/lib/python3.12/dist-packages/sklearn/utils/validation.py:2739: UserWarning: X does not have valid feature names, but Sta warnings.warn(
```