```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, f1_score
# Load Dataset
# which is a classic dataset used for classification tasks
iris = load_iris()
X = iris.data
y = iris.target
just to view the dataset
import pandas as pd
# Convert the data to a pandas DataFrame for easier viewing
df = pd.DataFrame(data=iris.data, columns=iris.feature_names)
# Add the target column to the DataFrame
df['target'] = iris.target
# Display the first few rows of the DataFrame
print(df.head())
       sepal length (cm)
                           sepal width (cm) petal length (cm)
                                                                 petal width (cm) \
\Box
                      5.1
                                        3.5
                                                            1.4
                                                                              0.2
                                        3.0
                                                            1.4
                                                                              0.2
    2
                      4.7
                                        3.2
                                                            1.3
                                                                              0.2
    3
                      4.6
                                        3.1
                                                            1.5
                                                                              0.2
    4
                      5.0
                                        3.6
                                                            1.4
                                                                              0.2
        target
    0
             0
             0
    1
    2
             0
    3
             0
     4
             0
#Split Dataset
X_train, X_test, y_train, y_test = train_test_split(iris.data, iris.target, test_size=0.2, random_state=42)
# Train Logistic Regression Model
logistic_model = LogisticRegression(max_iter=1000)
logistic_model.fit(X_train, y_train)
     ▶ LogisticRegression
# Evaluate the Model
y_pred = logistic_model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred, average='macro')
print("Accuracy:", accuracy)
print("F1-Score:", f1)
    Accuracy: 1.0
    F1-Score: 1.0
```

NEW DATASET

from sklearn.datasets import load_breast_cancer

just for viewing the dataset

```
import pandas as pd
# Load the Breast Cancer Wisconsin dataset
breast_cancer = load_breast_cancer()
# Convert the data to a pandas DataFrame for easier viewing
df = pd.DataFrame(data=breast_cancer.data, columns=breast_cancer.feature_names)
# Add the target column to the DataFrame
df['target'] = breast_cancer.target
# Display the first few rows of the DataFrame
print(df.head(8))
        mean radius mean texture mean perimeter
                                                    mean area mean smoothness
    0
              17.99
                            10.38
                                            122.80
                                                       1001.0
                                                                        0.11840
                            17.77
                                            132.90
                                                       1326.0
                                                                        0.08474
    1
              20.57
              19.69
                            21.25
                                            130.00
                                                       1203.0
                                                                       0.10960
    2
    3
              11.42
                            20.38
                                             77.58
                                                        386.1
                                                                        0.14250
    4
              20.29
                            14.34
                                            135.10
                                                       1297.0
                                                                        0.10030
              12.45
                            15.70
                                            82.57
                                                        477.1
                                                                       0.12780
    5
                            19.98
    6
              18.25
                                            119.60
                                                       1040.0
                                                                        0.09463
     7
              13.71
                            20.83
                                             90.20
                                                        577.9
                                                                        0.11890
        mean compactness mean concavity mean concave points mean symmetry
    0
                 0.27760
                                 0.30010
                                                       0.14710
                                                                        0.2419
                 0.07864
                                 0.08690
                                                       0.07017
                                                                        0.1812
    1
    2
                 0.15990
                                 0.19740
                                                       0.12790
                                                                        0.2069
    3
                 0.28390
                                 0.24140
                                                       0.10520
                                                                       0.2597
     4
                 0.13280
                                 0.19800
                                                       0.10430
                                                                        0.1809
    5
                 0.17000
                                 0.15780
                                                       0.08089
                                                                        0.2087
                 0.10900
                                 0.11270
                                                       0.07400
                                                                       0.1794
    6
     7
                 0.16450
                                 0.09366
                                                       0.05985
                                                                       0.2196
        mean fractal dimension ... worst texture worst perimeter worst area
                       0.07871 ...
    0
                                              17.33
                                                              184.60
                                                                           2019.0
                       0.05667
                                              23.41
                                                              158.80
                                                                           1956.0
    1
                                . . .
    2
                       0.05999
                                              25.53
                                                              152.50
                                                                           1709.0
                               . . .
    3
                       0.09744
                                              26.50
                                                               98.87
                                                                            567.7
     4
                       0.05883
                                . . .
                                              16.67
                                                               152.20
                                                                           1575.0
                                              23.75
                                                              103.40
    5
                       0.07613
                                                                           741.6
                               . . .
    6
                       0.05742
                                              27.66
                                                              153.20
                                                                           1606.0
                                . . .
     7
                       0.07451
                                              28.14
                                                              110.60
                                                                            897.0
        worst smoothness worst compactness worst concavity worst concave points
    0
                                      0.6656
                                                       0.7119
                  0.1622
                                                                              0.2654
                  0.1238
                                      0.1866
                                                       0.2416
                                                                              0.1860
    2
                  0.1444
                                      0.4245
                                                       0.4504
                                                                              0.2430
                  0.2098
                                     0.8663
                                                       0.6869
                                                                              0.2575
    3
                                     0.2050
                                                       0.4000
    4
                  0.1374
                                                                              0.1625
    5
                  0.1791
                                      0.5249
                                                       0.5355
                                                                              0.1741
                  0.1442
                                      0.2576
                                                       0.3784
                                                                              0.1932
    6
                  0.1654
                                     0.3682
                                                       0.2678
                                                                              0.1556
        worst symmetry worst fractal dimension target
    0
                0.4601
                                         0.11890
                                                       0
    1
                0.2750
                                         0.08902
                                                       0
    2
                0.3613
                                         0.08758
                                                       0
    3
                0.6638
                                         0.17300
                                                       0
    4
                0.2364
                                         0.07678
                                                       0
    5
                0.3985
                                         0.12440
                                                       0
                0.3063
                                         0.08368
                                                       0
                0.3196
                                                       0
                                         0.11510
     [8 rows x 31 columns]
```

```
breast_cancer = load_breast_cancer()
X = breast_cancer.data
y = breast_cancer.target
# Split Dataset
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Train Logistic Regression Model
logistic_model = LogisticRegression(max_iter=1000) # Increase max_iter if needed
logistic_model.fit(X_train, y_train)
# Evaluate the Model
y_pred = logistic_model.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)
print("Accuracy:", accuracy)
print("F1-Score:", f1)
    Accuracy: 0.956140350877193
    F1-Score: 0.9655172413793103
     /usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to converge
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
     Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
      n_iter_i = _check_optimize_result(
```