

In [2]: `import pandas as pd`

In [3]: `import numpy as np`

In [4]: `import matplotlib.pyplot as plt`

GENERATE DATASET

In [6]: `from sklearn.datasets import make_classification`

In [8]: `#without coefficient of underline model
x,y=make_classification(n_samples=1000,n_features=5,n_clusters_per_class=1,n_classes=2,random_state=2529)`

Get the first Five Rows of Target Variable(y) and Features(x)

In [9]: `x[0:5]`

Out[9]: `array([[1.54701705, 0.84770596, -0.41725021, -0.62356778, -0.19388577],
 [0.80633556, 0.40985594, -0.45641095, -0.3052022 , 0.50935923],
 [0.94390268, 0.70041038, 1.11385452, -0.49394417, 1.42305455],
 [1.92091517, 0.95815739, -1.2235022 , -0.71578154, 0.66588981],
 [1.45270369, 0.69035375, -1.18119669, -0.52009219, -0.22745417]])`

In [10]: `y[0:5]`

Out[10]: `array([0, 0, 1, 0, 0])`

Get Shape of DataFrame

In [11]: `x.shape,y.shape`

Out[11]: `((1000, 5), (1000,))`

Get Train Test Split

In [43]: `from sklearn.model_selection import train_test_split`

In [44]: `x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=2529)`

In [45]: `x_train.shape,x_test.shape,y_train.shape,y_test.shape`

Out[45]: `((700, 5), (300, 5), (700,), (300,))`

Get Logistic Regression Classification Model Train

In [46]: `from sklearn.linear_model import LogisticRegression`

In [47]: `model=LogisticRegression()`

In [48]: `model.fit(x_train,y_train)`

Out[48]: `LogisticRegression()`

Get Model Prediction

In [49]: `y_pred=model.predict(x_test)`

In [50]: `y_pred.shape`

Out[50]: `(300,)`

In [51]: `y_pred`

Out[51]: `array([[1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1,
 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0,
 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0,
 1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0,
 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0,
 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1,
 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1,
 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0,
 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1,
 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 0,
 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0,
 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1])`

Get Model Evaluation

In [52]: `from sklearn.metrics import accuracy_score,confusion_matrix,classification_report`

In [53]: `accuracy_score(y_test,y_pred)`

Out[53]: `0.99`

In [54]: `confusion_matrix(y_test,y_pred)`

Out[54]: `array([[156, 1],
 [2, 141]], dtype=int64)`

In [28]: `print(classification_report(y_test,y_pred))`

	precision	recall	f1-score	support
0	0.99	0.99	0.99	157
1	0.99	0.99	0.99	143
accuracy			0.99	300
macro avg	0.99	0.99	0.99	300
weighted avg	0.99	0.99	0.99	300

Hyperparameter Tunning: Grid Search

In [59]: `from sklearn.model_selection import GridSearchCV
parameters={'penalty':['l1','l2'],'C':[0.001,.009,0.01,.09,1,5,10,25],'solver':['liblinear']}
gridsearch=GridSearchCV(LogisticRegression(),parameters)
gridsearch.fit(x_train,y_train)`

Out[59]: `GridSearchCV(estimator=LogisticRegression(),
 param_grid={'C': [0.001, 0.009, 0.01, 0.09, 1, 5, 10, 25],
 'penalty': ['l1', 'l2'], 'solver': ['liblinear']})`

In [60]: `gridsearch.best_params_`

Out[60]: `{'C': 1, 'penalty': 'l1', 'solver': 'liblinear'}`

In [61]: `gridsearch.best_score_`

Out[61]: `0.9914285714285714`

In [62]: `gridsearch.best_estimator_`

Out[62]: `LogisticRegression(C=1, penalty='l1', solver='liblinear')`

In [63]: `gridsearch.best_index_`

Out[63]: `8`

In [64]: `y_pred_grid=gridsearch.predict(x_test)`

In [65]: `confusion_matrix(y_test,y_pred_grid)`

Out[65]: `array([[156, 1],
 [2, 141]], dtype=int64)`

In [67]: `print(classification_report(y_test,y_pred_grid))`

	precision	recall	f1-score	support
0	0.99	0.99	0.99	157
1	0.99	0.99	0.99	143
accuracy			0.99	300
macro avg	0.99	0.99	0.99	300
weighted avg	0.99	0.99	0.99	300