# Logistic regression Classifier with Arificial Generated

## **Import Library**

```
In [1]:
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

### **Generate Dataset**

```
In [2]:
```

```
from sklearn.datasets import make_classification
In [3]:
```

```
# without coeficient of underline model
X,y = make_classification(n_samples=1000,n_features=5,n_clusters_per_class=1,n_classes=2,ra
```

# Get the First Five Rows of Target Variable(y) and Features(X)

## **Get Shape of Dataframe**

```
In [6]:
```

X.shape,y.shape

#### Out[6]:

((1000, 5), (1000,))

# **Get Train Test Split**

### In [7]:

from sklearn.model\_selection import train\_test\_split

### In [8]:

X\_train,X\_test,y\_train,y\_test = train\_test\_split(X,y,test\_size=0.3,random\_state=2529)

```
In [9]:

X_train.shape,X_test.shape,y_train,y_test.shape
Out[9]:
```

```
((700, 5),
 (300, 5),
array([1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1,
       1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0,
       1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1,
       1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1,
             0, 1, 0,
                     1, 0, 1, 0, 0, 0, 1, 1,
                                             0,
                                                0, 1,
                                                      1,
                                                         1,
       0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1,
       1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1,
             0, 0,
                   1,
                      0, 1,
                            1, 0, 1, 1, 0, 0,
                                             1,
                                                      1,
                                                         0, 0,
                                                1,
                                                   0,
                                                               0,
       1, 0, 0, 1, 0,
                     0, 1, 0,
                              0, 1, 1, 1, 1, 0, 1, 1,
                                                      0, 1,
                                                           1,
       0, 0, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1,
       1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1,
       0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1,
                                                         1, 0,
                                                   0,
                                                      1,
                                                              1,
       1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1,
                                                           0,
       1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1,
       0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
                                                      1, 0, 0, 1,
             0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
                                                      1,
                                                         1,
                                                            0,
       0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0,
                                                              0, 1, 1,
       1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1,
                     1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
       1, 0, 1, 1, 0,
             1, 1,
                  0,
                     0, 1,
                           0, 1, 1, 0, 0, 1,
                                            1, 1, 0,
                                                      0,
                                                        0,
       0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1,
       0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1,
       1, 1, 1,
                   0,
                     1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0,
                                                         1, 1, 0,
                1,
       0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0,
       0, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1,
       1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0,
                0, 0, 1,
                         1, 1, 0, 0, 0, 0, 0, 1, 0,
                                                   0,
                                                      0,
                                                         0,
                                                           1,
       0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1,
       1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1,
       1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1,
       1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
       (300,))
```

## Get Logistic Regression classification Model Train

```
LogisticRegression(penalty='l2', dual = False, tol=0.0001,C=1.0,fit_intercept
=True,intercept_scaling=1,class_weight=None,random_state=None,solver='lbfgs',max_iter=100,multi_class_logistic.

In [10]:

from sklearn.linear_model import LogisticRegression

In [11]:

model = LogisticRegression()
```

```
In [12]:
model.fit(X_train,y_train)
Out[12]:
LogisticRegression()
Get Model Prediction
In [13]:
y_pred = model.predict(X_test)
In [14]:
y_pred.shape
Out[14]:
(300,)
In [15]:
y_pred
Out[15]:
array([1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 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, 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, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 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, 1, 0, 0, 1, 1, 0, 1, 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,
```

### **Get Model Evaluation**

0.99

1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1])

```
In [16]:
from sklearn.metrics import accuracy_score, confusion_matrix,classification_report

In [17]:
accuracy_score(y_test,y_pred)
Out[17]:
```

```
In [18]:
confusion_matrix(y_test,y_pred)
Out[18]:
array([[156, 1],
       [ 2, 141]], dtype=int64)
In [19]:
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
                                      0.99
                                                 300
   accuracy
                  0.99
                            0.99
                                      0.99
                                                 300
   macro avg
weighted avg
                  0.99
                            0.99
                                      0.99
                                                 300
Hyperparameter Tunning: Grid Search
GridSearchCV(estimator,
param_grid,scoring=None,n_jobs=None,refit=True,cv=None,verbose=0,pre_dispatch='2n_jobs',error_sco
```

```
In [20]:
from sklearn.model_selection import GridSearchCV
parameters = {'penalty':['11','12'],'C':[0.001,.009,.09,1,5,10,25],'solver':['liblinear']}
gridsearch = GridSearchCV(LogisticRegression(),parameters)
gridsearch.fit(X_train,y_train)
Out[20]:
GridSearchCV(estimator=LogisticRegression(),
             param_grid={'C': [0.001, 0.009, 0.09, 1, 5, 10, 25],
                          'penalty': ['l1', 'l2'], 'solver': ['liblinear']})
In [21]:
gridsearch.best_params_
Out[21]:
{'C': 1, 'penalty': 'l1', 'solver': 'liblinear'}
In [22]:
gridsearch.best_estimator_
Out[22]:
LogisticRegression(C=1, penalty='l1', solver='liblinear')
```

```
In [23]:
gridsearch.best_index_
Out[23]:
6
In [24]:
y_pred_grid = gridsearch.predict(X_test)
In [25]:
confusion_matrix(y_test,y_pred_grid)
Out[25]:
array([[156, 1],
       [ 2, 141]], dtype=int64)
In [26]:
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
                                       0.99
                                                   300
    accuracy
   macro avg
                   0.99
                             0.99
                                       0.99
                                                   300
                             0.99
                                       0.99
                                                   300
weighted avg
                   0.99
```

In [ ]: