

Mini Project 2: Classification Dummy Data

▼ Step 1: Import Library

```
import pandas as pd
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

▼ Step 2: Generating Dataset

```
from sklearn.datasets import make_classification
```

```
X, y = make_classification(n_samples=1000, n_features=5, n_clusters_per_class=1, n_classes=2, random_state=2529)
```

```
X[0:5]
```

```
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]])
```

```
y[0:5]
```

```
array([0, 0, 1, 0, 0])
```

```
X.shape, y.shape
```

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

▼ Step 3: Splitting Data

```
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3 , random_state=2529)
```

```
X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

```
((700, 5), (300, 5), (700,), (300,))
```

▼ Step 4: Creating Model

```
from sklearn.linear_model import LogisticRegression
```

```
model = LogisticRegression()
```

▼ Step 5: Training Model

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

```
LogisticRegression()
```

▼ Step 6: Predicting Model

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

```
y_pred.shape
```

```
(300,)
```

```
y_pred
```

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

▼ Step 7: Accuracy

```
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```

```
accuracy_score(y_test,y_pred)
```

```
0.99
```

```
confusion_matrix(y_test,y_pred)
```

```
array([[156,  1],
       [ 2, 141]])
```

```
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

▼ Step 8: Hyperparameter Tuning

```
from sklearn.model_selection import GridSearchCV
parameters = {'penalty': ['l1', 'l2'], 'C':[0.001,.009,0.01,.09,1,5,10,25], 'solver' : ['liblinear']}
gs = GridSearchCV(LogisticRegression(), parameters)
gs.fit(X_train, y_train)
GridSearchCV(estimator=LogisticRegression(),
              param_grid={'C': [0.001, 0.009, 0.01, 0.09, 1, 5, 10, 25],
                           'penalty': ['l1', 'l2'], 'solver': ['liblinear']})

GridSearchCV(estimator=LogisticRegression(),
              param_grid={'C': [0.001, 0.009, 0.01, 0.09, 1, 5, 10, 25],
```

```
'penalty': ['l1', 'l2'], 'solver': ['liblinear']})
```

```
gs.best_params_
```

```
{'C': 1, 'penalty': 'l1', 'solver': 'liblinear'}
```

```
gs.best_score_
```

```
0.9914285714285714
```

```
gs.best_estimator_
```

```
LogisticRegression(C=1, penalty='l1', solver='liblinear')
```

```
gs.best_index_
```

```
8
```

▼ Step 9: Re-Prediction and Re-Evaluation

```
y_pred_grid = gs.predict(X_test)
```

```
confusion_matrix(y_test,y_pred_grid)
```

```
array([[156,  1],  
       [ 2, 141]])
```

```
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

Link of the same:

<https://colab.research.google.com/drive/1gPcSrnlSTVKDY8hibBGazSbUeJJxL2yZ?usp=sharing>

