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1.Decision Tree Classification with Artificial Generated Dataset
         Import Library
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
         Generate Dataset
          from sklearn.datasets import make_classification
 In [5]:
          #without coeficient 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 [6]:
          x[0:5]
         array([[ 1.54701705, 0.84770596, -0.41725021, -0.62356778, -0.19388577],
 Out[6]:
                   0.80633556, \quad 0.40985594, \quad -0.45641095, \quad -0.3052022 \quad , \quad 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])
         Get Shape of DataFrame
          x.shape, y.shape
         ((1000, 5), (1000,))
 Out[8]:
         Get Train Test Split
          from sklearn.model_selection import train_test_split
In [10]:
          x_train, x_test, y_train, y_test=train_test_split(x, y, test_size=0.3, random_state=2529)
In [11]:
          x_train.shape, x_test.shape, y_train.shape, y_test.shape
          ((700, 5), (300, 5), (700,), (300,))
Out[11]:
         Get Decision Tree Model Train
In [12]:
          from sklearn.tree import DecisionTreeClassifier
In [13]:
          model=DecisionTreeClassifier()
          model.fit(x_train,y_train)
         DecisionTreeClassifier()
Out[14]:
         Get Model Prediction
          y_pred=model.predict(x_test)
          y_pred.shape
         (300,)
Out[16]:
In [17]:
          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,
Out[17]:
                 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, 1, 1, 1, 0, 0, 1, 0, 0, 0,
                1, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0,
                 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1,
                 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0,
                1, 0, 1, 0, 1, 1, 0, 1, 0, 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,
                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, 1, 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,
                1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1])
         Get Model Evaluation
In [19]:
          from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
In [20]:
          accuracy_score(y_test,y_pred)
         0.9833333333333333
In [21]:
          confusion_matrix(y_test,y_pred)
          array([[155, 2],
Out[21]:
                 [ 3, 140]], dtype=int64)
          print(classification_report(y_test,y_pred))
                                     recall f1-score support
                        precision
                                       0.99
                                                  0.98
                                                             157
                     0
                             0.98
                     1
                             0.99
                                       0.98
                                                  0.98
                                                             143
                                                  0.98
                                                             300
              accuracy
                             0.98
                                       0.98
                                                  0.98
                                                             300
             macro avg
          weighted avg
                             0.98
                                       0.98
                                                  0.98
                                                             300
         Hyperparameter Tunning: Grid Search
In [24]:
          from sklearn.model_selection import GridSearchCV
          parameters={'criterion':['gini', 'entropy'], 'max_depth':[2,3,4,5,6,7,8,9,10,11,12,15,20,30,40,50,70,90,120,150]}
          gridsearch=GridSearchCV(DecisionTreeClassifier(), parameters)
          gridsearch.fit(x_train,y_train)
          GridSearchCV(estimator=DecisionTreeClassifier(),
Out[24]:
                       param_grid={'criterion': ['gini', 'entropy'],
                                    'max_depth': [2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 15,
                                                  20, 30, 40, 50, 70, 90, 120, 150]})
          gridsearch.best_params_
          {'criterion': 'gini', 'max_depth': 3}
Out[25]
In [26]:
          gridsearch.best_score_
         0.9885714285714287
In [27]:
          gridsearch.best_estimator_
          DecisionTreeClassifier(max_depth=3)
Out[27]:
In [28]:
          gridsearch.best_index_
Out[28]:
In [29]:
          y_pred_grid=gridsearch.predict(x_test)
In [30]:
          confusion_matrix(y_test,y_pred_grid)
          array([[155, 2],
Out[30]:
                 [ 1, 142]], dtype=int64)
In [32]:
          print(classification_report(y_test,y_pred_grid))
                        precision
                                     recall f1-score
                                                         support
```

0

1

accuracy macro avq

weighted avg

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

0.99

157

143

300

300

300