```
In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
 In [2]:
         import warnings
         warnings.filterwarnings('ignore')
 In [3]: | from sklearn.datasets import make_classification
 In [4]: X,y=make_classification(n_samples=1000,n_features=5,n_clusters_per_class=1,n_class=1)
 In [5]: X[0:5]
 Out[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]])
 In [6]: y[0:5]
 Out[6]: array([0, 0, 1, 0, 0])
 In [7]: X.shape,y.shape
 Out[7]: ((1000, 5), (1000,))
 In [8]: #get train test split
 In [9]: 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=251
In [11]: X_train.shape,X_test.shape,y_train.shape,y_test.shape
Out[11]: ((700, 5), (300, 5), (700,), (300,))
In [12]: from sklearn.tree import DecisionTreeClassifier
In [13]: model=DecisionTreeClassifier()
```

```
In [14]: model.fit(X train,y train)
Out[14]: DecisionTreeClassifier()
In [15]: #get model predictions
In [16]: y pred=model.predict(X test)
In [17]: y_pred.shape
Out[17]: (300,)
In [18]: y_pred
Out[18]: 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, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 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, 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])
In [19]: from sklearn.metrics import confusion matrix, classification report, accuracy score
In [20]: |confusion matrix(y pred,y test)
Out[20]: array([[155,
                        3],
                [ 2, 140]], dtype=int64)
In [21]: | accuracy_score(y_pred,y_test)
Out[21]: 0.9833333333333333
```

```
In [22]: print(classification report(y pred,y test))
                       precision
                                     recall f1-score
                                                        support
                    0
                            0.99
                                       0.98
                                                 0.98
                                                            158
                    1
                            0.98
                                       0.99
                                                 0.98
                                                            142
             accuracy
                                                 0.98
                                                            300
            macro avg
                            0.98
                                       0.98
                                                 0.98
                                                            300
         weighted avg
                            0.98
                                       0.98
                                                 0.98
                                                            300
In [23]: r2_score(y_pred,y_test)
Out[23]: 0.933143162774113
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]
In [25]:
         gridsearch=GridSearchCV(DecisionTreeClassifier(),parameters)
         gridsearch.fit(X_train,y_train)
Out[25]: GridSearchCV(estimator=DecisionTreeClassifier(),
                      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]})
In [26]: |gridsearch.best_params_
Out[26]: {'criterion': 'entropy', 'max_depth': 3}
In [27]: gridsearch.best_score_
Out[27]: 0.9885714285714287
In [28]: gridsearch.best_estimator_
Out[28]: DecisionTreeClassifier(criterion='entropy', max_depth=3)
In [29]: gridsearch.best_index_
Out[29]: 21
In [30]: y_pred_grid=gridsearch.predict(X_test)
```

```
In [31]: |confusion_matrix(y_test,y_pred_grid)
Out[31]: array([[156,
                         1],
                [ 1, 142]], dtype=int64)
In [32]: 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
         #https://www.youtube.com/watch?v=w4frwjt8uCo
In [33]:
 In [ ]:
 In [ ]:
 In [ ]:
         P4.b 2. Random Forest Classification with Artificial Generated Dataset
 In [ ]:
         import pandas as pd
In [34]:
         import numpy as np
         import matplotlib.pyplot as plt
In [35]: | from sklearn.datasets import make_classification
In [36]: X,y=make classification(n samples=1000,n features=5,n clusters per class=1,n class
In [37]: X[0:5]
Out[37]: 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],
                               0.69035375, -1.18119669, -0.52009219, -0.22745417]])
                [ 1.45270369,
```

```
In [38]: y[0:5]
Out[38]: array([0, 0, 1, 0, 0])
In [39]: |X.shape,y.shape
Out[39]: ((1000, 5), (1000,))
In [40]: | from sklearn.model_selection import train_test_split
In [41]: |X_train,X_test,y_train,y_test=train_test_split(X,y,train_size=0.7,random_state=25
In [42]: | X_train.shape, X_test.shape, y_train.shape, y_test.shape
Out[42]: ((700, 5), (300, 5), (700,), (300,))
In [43]: | from sklearn.ensemble import RandomForestClassifier
In [44]: |model=RandomForestClassifier()
In [45]: |model.fit(X_train,y_train)
Out[45]: RandomForestClassifier()
In [46]: y pred=model.predict(X test)
In [47]: y_pred.shape
Out[47]: (300,)
In [48]: y_pred
Out[48]: 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,
                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, 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])
```

```
In [49]: from sklearn.metrics import accuracy score, confusion matrix, classification report
In [50]: | accuracy_score(y_test,y_pred)
Out[50]: 0.99
In [51]: |confusion_matrix(y_test,y_pred)
Out[51]: array([[156,
                         1],
                 [ 2, 141]], dtype=int64)
In [52]: 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
            macro avg
                             0.99
                                       0.99
                                                 0.99
                                                             300
         weighted avg
                             0.99
                                       0.99
                                                 0.99
                                                             300
In [56]: from sklearn.model selection import GridSearchCV
         parameters={'n estimators':[10,20,30,100,200,500],'max features':['auto','sqrt']
         gridsearch=GridSearchCV(RandomForestClassifier(),parameters)
         gridsearch.fit(X_train,y_train)
Out[56]: GridSearchCV(estimator=RandomForestClassifier(),
                       param_grid={'bootstrap': [True, False],
                                   'max_features': ['auto', 'sqrt'],
                                   'min_samples_split': [4, 8],
                                   'n_estimators': [10, 20, 30, 100, 200, 500]})
In [54]:
         gridsearch.best_params_
Out[54]: {'bootstrap': True,
           'max features': 'sqrt',
           'min_samples_split': 4,
           'n_estimators': 10}
In [55]: gridsearch.best_estimator_
Out[55]: RandomForestClassifier(max_features='sqrt', min_samples_split=4,
                                 n_estimators=10)
```

```
In [57]: gridsearch.best_index_
Out[57]: 25
In [58]: y_pred_grid=gridsearch.predict(X_test)
         confusion_matrix(y_test,y_pred_grid)
In [64]:
Out[64]: array([[156,
                 [ 2, 141]])
         print(classification_report(y_test,y_pred_grid))
In [65]:
                        precision
                                      recall f1-score
                                                         support
                     0
                             0.99
                                        0.99
                                                  0.99
                                                              157
                             0.99
                                        0.99
                                                              143
                     1
                                                  0.99
                                                              300
              accuracy
                                                  0.99
                             0.99
                                        0.99
                                                  0.99
                                                              300
            macro avg
         weighted avg
                             0.99
                                        0.99
                                                  0.99
                                                              300
         print(classification report(y pred,y pred grid))
In [66]:
                        precision
                                      recall f1-score
                                                         support
                     0
                             1.00
                                        1.00
                                                  1.00
                                                              158
                     1
                             1.00
                                        1.00
                                                  1.00
                                                              142
                                                  1.00
                                                              300
              accuracy
             macro avg
                             1.00
                                        1.00
                                                  1.00
                                                              300
         weighted avg
                             1.00
                                        1.00
                                                  1.00
                                                              300
```