```
2.Random Forest Classification with Artificial Generated Dataset
          #Import Library
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
 In [3]:
          #Generate Dataset
          from sklearn.datasets import make_classification
          #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 [5]:
          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])
         Get Shape of DataFrame
          x.shape, y.shape
         ((1000, 5), (1000,))
 Out[7]:
         Get Train Test Split
          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,))
Out[10]:
         Get Random Forest Classification Model Train
In [12]:
          from sklearn.ensemble import RandomForestClassifier
In [13]:
          model=RandomForestClassifier()
In [14]:
          model.fit(x_train,y_train)
          RandomForestClassifier()
Out[14]:
         Get Model Prediction
          y_pred=model.predict(x_test)
          y_pred.shape
Out[16]:
          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,\ 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,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,\ 1,\ 0,\ 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, 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, 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)
Out[20]:
In [21]:
          confusion_matrix(y_test,y_pred)
         array([[156, 1],
Out[21]:
                 [ 2, 141]], dtype=int64)
In [22]:
          print(classification_report(y_test,y_pred))
                                     recall f1-score
                        precision
                     0
                                       0.99
                                                  0.99
                                                             157
                             0.99
                             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
          from sklearn.model_selection import GridSearchCV
          parameters={'n_estimators':[10,20,30,100,200,500], 'max_features':['auto', 'sqrt'], 'min_samples_split':[4, 8], 'bootstrap':[True, False]}
          gridsearch=GridSearchCV(RandomForestClassifier(), parameters)
          gridsearch.fit(x_train,y_train)
          GridSearchCV(estimator=RandomForestClassifier(),
Out[28]:
                       param_grid={'bootstrap': [True, False],
                                    'max_features': ['auto', 'sqrt'],
                                    'min_samples_split': [4, 8],
                                   'n_estimators': [10, 20, 30, 100, 200, 500]})
In [29]:
          gridsearch.best_params_
          {'bootstrap': False,
Out[29]:
           'max_features': 'auto',
           'min_samples_split': 8,
           'n_estimators': 10}
In [30]:
          gridsearch.best_score_
          0.9885714285714287
Out[30]
In [31]:
          gridsearch.best_estimator_
          RandomForestClassifier(bootstrap=False, min_samples_split=8, n_estimators=10)
Out[31]:
In [32]:
          gridsearch.best_index_
Out[32]:
In [33]:
          y_pred_grid=gridsearch.predict(x_test)
In [34]:
          confusion_matrix(y_test,y_pred_grid)
          array([[156, 1],
Out[34]:
                 [ 2, 141]], dtype=int64)
In [35]:
          print(classification_report(y_test,y_pred_grid))
                        precision
                                     recall f1-score support
                                       0.99
                                                  0.99
                                                             157
                     0
                             0.99
                             0.99
                                       0.99
                                                  0.99
                                                             143
                                                  0.99
                                                             300
             accuracy
             macro avg
                             0.99
                                       0.99
                                                  0.99
                                                             300
```

weighted avg

In []:

0.99

0.99

300