## IMPLEMENTATION OF DECISION TREE USING ID3 ALGORITHM

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In [5]:
         # Implementation of decision tree classifier
          # Importing the required packages
         import numpy as np
         import pandas as pd
         from sklearn.metrics import confusion matrix
         from sklearn.model selection import train test split
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.tree import export graphviz
         from sklearn.metrics import accuracy score
         from sklearn.metrics import classification report
 In [6]:
         def importdata():
             data = pd.read csv('Sample Data.csv')
             # Printing the dataset shape
             print ("Dataset Length: ", len(data))
             print ("Dataset Shape: ", data.shape)
             # Printing the dataset obseravtions
             print ("Dataset: ", data.head())
             return data
 In [7]:
         # Function to split the dataset
         def splitdataset(data):
            # Separating the target variable
             X = data.iloc[:, 1:3].values
             Y = data.iloc[:, 4].values
             # Splitting the dataset into train and test
             X train, X test, y train, y test = train test split(
             X, Y, test size = 0.25, random state = 100)
             return X, Y, X train, X test, y train, y test
 In [8]:
         # Function to perform training with giniIndex.
         def train using gini(X train, X test, y train):
             # Creating the classifier object
             clf gini = DecisionTreeClassifier(criterion = "gini",
             random state = 100, max depth=4, min samples leaf=2)
             # Performing training
             clf_gini.fit(X_train, y_train)
             return clf gini
 In [9]:
         # Function to perform training with entropy.
         def train using entropy(X train, X test, y train):
             # Decision tree with entropy
             clf entropy = DecisionTreeClassifier(
             criterion = "entropy", random state = 100,
             \max depth = 4, \min samples leaf = 2)
             # Performing training
             clf entropy.fit(X train, y train)
             return clf_entropy
In [10]:
         # Function to make predictions
         def prediction(X test, clf object):
             # Predicton on test with giniIndex
             y pred = clf object.predict(X test)
             print("Predicted values:")
             print(y pred)
             return y pred
In [11]:
         # Function to calculate accuracy
         def cal_accuracy(y_test, y_pred):
            print("Confusion Matrix: ",
             confusion_matrix(y_test, y_pred))
            print ("Accuracy : ",
             accuracy_score(y_test,y_pred)*100)
             print("Report : ",
             classification_report(y_test, y_pred))
In [12]:
         # Driver code
         def main():
             # Building Phase
             data = importdata()
             ## data.info()
             data['Gender'], = pd.factorize(data['Gender'])
             data['PurchasedTheProduct'], class names = pd.factorize(data['PurchasedTheProduct'])
             ## data.info()
             Xf = data.iloc[:,1:-1]
             X, Y, X train, X test, y train, y test = splitdataset(data)
             clf gini = train using gini(X train, X test, y train)
             clf_entropy = train_using_entropy(X_train, X_test, y_train)
             # Operational Phase
             print("Results Using Gini Index:")
             # Prediction using gini
             y_pred_gini = prediction(X test, clf gini)
             cal_accuracy(y_test, y_pred_gini)
             print("Results Using Entropy:")
             # Prediction using entropy
             y_pred_entropy = prediction(X_test, clf_entropy)
             cal_accuracy(y_test, y_pred_entropy)
In [13]:
         # Calling main function
         if name ==" main ":
             main()
        Dataset Length: 400
        Dataset Shape: (400, 5)
        Dataset: User ID Gender Age Salary PurchasedTheProduct
        0 15624510 Male 19 19000
                                          0
        1 15810944 Male 35 20000
                                                           0
        2 15668575 Female 26 43000
                                                           0
        3 15603246 Female 27 57000
                                                           0
        4 15804002 Male 19
                                  76000
        Results Using Gini Index:
        Predicted values:
        [0\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 0
         Confusion Matrix: [[62 3]
         [12 23]]
        Accuracy: 85.0
                              precision recall f1-score
        Report :
                                                              support
                      0.84 0.95
                                          0.89
                   0
                                                    65
                          0.88
                                             0.75
                                   0.66
                                                    100
                                             0.85
            accuracy

      0.86
      0.81
      0.82

      0.85
      0.85
      0.84

           macro avg
        weighted avg
        Results Using Entropy:
        Predicted values:
        [0\ 1\ 0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 0
         Confusion Matrix: [[62 3]
         [12 23]]
        Accuracy: 85.0
        Report :
                              precision recall f1-score support
                   0
                      0.84 0.95
                                            0.89

      accuracy
      0.85
      100

      macro avg
      0.86
      0.81
      0.82
      100

      eighted avg
      0.85
      0.85
      0.84
      100

        weighted avg
```