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In [18]:
          # importing libraries
          import numpy as nm
          import matplotlib.pyplot as mtp
          import pandas as pd
          from sklearn import metrics
          #importing datasets
          data_set= pd.read_csv('user_data.csv')
          #Extracting Independent and dependent Variable
          x= data_set.iloc[:, [2,3]].values
          y= data_set.iloc[:, 4].values
          # Splitting the dataset into training and test set.
          from sklearn.model_selection import train_test_split
          x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 0.25, random_state=0)
          #feature Scaling
          from sklearn.preprocessing import StandardScaler
          st_x= StandardScaler()
          x_train= st_x.fit_transform(x_train)
          x_test= st_x.transform(x_test)
In [19]:
          data_set
               User ID Gender Age EstimatedSalary Purchased
Out[19]:
            0 15624510
                               19
                                           19000
                                                        0
                         Male
           1 15810944
                               35
                                           20000
                                                        0
                         Male
           2 15668575 Female
                               26
                                           43000
                                                        0
           3 15603246 Female
                                           57000
                                                        0
                               27
            4 15804002
                         Male
                               19
                                           76000
                                                        0
          395 15691863 Female
                                           41000
                                                        1
          396 15706071
                                           23000
                                                        1
                         Male
                               51
          397 15654296 Female
                                           20000
                                                        1
          398 15755018
                                           33000
                                                        0
                               36
                         Male
          399 15594041 Female
                                           36000
                                                        1
         400 rows × 5 columns
In [20]:
          #Fitting Decision Tree classifier to the training set random forest
          from sklearn.ensemble import RandomForestClassifier
          classifier= RandomForestClassifier(n_estimators= 10, criterion="entropy")
          classifier.fit(x_train, y_train)
         RandomForestClassifier(criterion='entropy', n_estimators=10)
Out[20]:
In [21]:
          #Predicting the test set result
          y_pred= classifier.predict(x_test)
In [22]:
          y_pred
Out[22]: array([0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
                 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1,
                0,\ 0,\ 0,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 0,\ 1,
                1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1], dtype=int64)
In [23]:
          #Now we will create the confusion matrix to determine the correct and incorrect predictions.
          #Creating the Confusion matrix
          from sklearn.metrics import confusion_matrix
          cm= confusion_matrix(y_test, y_pred)
In [24]:
          cm
Out[24]: array([[65, 3],
                 [ 4, 28]], dtype=int64)
In [25]:
          #Visualizing the training Set result
          from matplotlib.colors import ListedColormap
          x_{set}, y_{set} = x_{train}, y_{train}
          x1, x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),
          nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))
          mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape),
          alpha = 0.75, cmap = ListedColormap(('purple', 'green')))
          mtp.xlim(x1.min(), x1.max())
          mtp.ylim(x2.min(), x2.max())
          for i, j in enumerate(nm.unique(y_set)):
              mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],
                  c = ListedColormap(('purple', 'green'))(i), label = j)
          mtp.title('Random Forest Algorithm (Training set)')
          mtp.xlabel('Age')
          mtp.ylabel('Estimated Salary')
          mtp.legend()
          mtp.show()
          *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *
          x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point
          *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *
         x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point
         S.
                     Random Forest Algorithm (Training set)
             3
             2
          Estimated Salary
             0
            -1
            -2
                            -1
                                    Ó
                                    Age
In [26]:
          #Visualizing the test set result
          from matplotlib.colors import ListedColormap
          x_{set}, y_{set} = x_{test}, y_{test}
          x1, x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),
          nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))
          mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape),
          alpha = 0.75, cmap = ListedColormap(('purple', 'green' )))
          mtp.xlim(x1.min(), x1.max())
          mtp.ylim(x2.min(), x2.max())
          for i, j in enumerate(nm.unique(y_set)):
              mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],
                  c = ListedColormap(('purple', 'green'))(i), label = j)
          mtp.title('Random Forest Algorithm(Test set)')
          mtp.xlabel('Age')
          mtp.ylabel('Estimated Salary')
          mtp.legend()
          mtp.show()
         *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *
         x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point
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         x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point
                       Random Forest Algorithm(Test set)
             3
             2
          Estimated Salary
             1
            ^{-1}
            -2
                     -2
                            -1
                                    0
                                    Age
In [30]:
          accuracy = metrics.accuracy_score(y_test,y_pred)
          report = metrics.classification_report(y_test,y_pred)
          cm = metrics.confusion_matrix(y_test,y_pred)
          print("Classification report:")
          print("Accuracy: ", accuracy)
          print(report)
          print("Confusion matrix:")
          print(cm)
          Classification report:
         Accuracy: 0.93
                        precision
                                      recall f1-score
                                                         support
                     0
                             0.94
                                       0.96
                                                  0.95
                                                              68
                     1
                             0.90
                                       0.88
                                                  0.89
                                                              32
             accuracy
                                                  0.93
                                                             100
             macro avg
                             0.92
                                       0.92
                                                  0.92
                                                             100
         weighted avg
                             0.93
                                       0.93
                                                  0.93
                                                             100
          Confusion matrix:
          [[65 3]
          [ 4 28]]
 In [ ]:
 In [ ]:
```