INTRODUCTION TO MACHINE LEARNING

LAB ASSIGNMENT – 4

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DECISION TREE CLASSIFICATION:

CODE:

```
import pandas as pd
from sklearn.tree import DecisionTreeClassifier
from sklearn.model selection import train test split
from sklearn.metrics import accuracy score
from sklearn.ensemble import BaggingClassifier
from sklearn.ensemble import AdaBoostClassifier
from sklearn.metrics import classification report, confusion matrix
company=pd.read csv("Carseats.csv")
company.head()
company.dtypes
company['High'] = company.Sales.map(lambda x: 1 if x>8 else 0)
company['ShelveLoc']=company['ShelveLoc'].astype('category')
company['Urban']=company['Urban'].astype('category')
company['US'] = company['US'].astype('category')
company.dtypes
company.head()
company['ShelveLoc'] = company['ShelveLoc'].cat.codes
company['Urban']=company['Urban'].cat.codes
company['US']=company['US'].cat.codes
company.tail()
feature cols=['CompPrice','Income','Advertising','Population','Price','She
lveLoc','Age','Education','Urban','US']
x = company[feature cols]
y = company.High
print(x)
print(y)
x train, x test, y train, y test= train test split(x,y, test size=0.2, random
state=0)
dcmodel = BaggingClassifier(DecisionTreeClassifier(max depth = 6), random
state=0)
dcmodel = AdaBoostClassifier(DecisionTreeClassifier(max depth = 6), rando
m state=0)
```

```
dcmodel = dcmodel.fit(x_train,y_train)
y_predict = dcmodel.predict(x_test)
print("Accuracy : ", accuracy_score(y_test,y_predict)*100 )
print(confusion_matrix(y_test,y_predict))
print(classification report(y test,y predict))
```

```
import pandas as pd
 from sklearn.tree import DecisionTreeClassifier
 from sklearn.model_selection import train_test_split
 from sklearn.metrics import accuracy score
 from sklearn.ensemble import BaggingClassifier
 from sklearn.ensemble import AdaBoostClassifier
 from sklearn.metrics import classification_report, confusion_matrix
 company=pd.read_csv("Carseats.csv")
 company.head()
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 company['ShelveLoc']=company['ShelveLoc'].astype('category')
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 company.dtypes
 company.head()
 company['ShelveLoc']=company['ShelveLoc'].cat.codes
 company['Urban']=company['Urban'].cat.codes
 company['US']=company['US'].cat.codes
 company.tail()
 feature cols=['CompPrice','Income','Advertising','Population','Price','ShelveLoc','Age','Education','Urban','US']
 x = company[feature cols]
 y = company.High
 print(x)
 print(y)
 x train,x test,y train,y test= train test split(x,y, test size=0.2,random state=0)
 dcmodel = BaggingClassifier(DecisionTreeClassifier(max_depth = 6), random_state=0)
 dcmodel = AdaBoostClassifier(DecisionTreeClassifier(max_depth = 6), random_state=0)
 dcmodel = dcmodel.fit(x_train,y_train)
y_predict = dcmodel.predict(x_test)
 print("Accuracy : ", accuracy_score(y_test,y_predict)*100 )
 print(confusion_matrix(y_test,y_predict))
 print(classification_report(y_test,y_predict))
```

OUTPUT:

```
CompPrice Income Advertising Population Price ShelveLoc Age \
                73
                    11
                                  276
                                       120
□ 1
                48
                                   260
                                       83
          111
                          16
                                                 1 65
          113
                35
                          10
                                  269
                                       80
                          4
          117
                100
                                  466 97
   4
          141
                64
                          3
                                  340 128
          . . .
                ...
                          ...
                                   ... ...
                                                 ... ...
   395
          138
              108
                          17
                                  203 128
                23
   396
          139
                          3
                                   37 120
                                                 2 55
   397
          162
                          12
                                  368 159
                                                 2 40
                26
   398
          100
                79
                          7
                                  284
                                        95
                                                 0 50
                          0
   399
          134
                37
                                   27 120
                                                 1 49
      Education Urban US
       17
   0
               1 1
  1
           10
                 1 1
           12
                1 1
   2
   3
           14
                1 1
   4
           13
                 1 0
           ...
   395
           14
                1 1
                 0
   396
           11
                    1
                 1 1
   397
           18
   398
           12
                 1 1
   399
           16
   [400 rows x 10 columns]
   0
      1
   1
        1
   2
        1
   3
        0
   4
       0
   395
       1
   396 0
   397
      0
   398 0
   399
      1
   Name: High, Length: 400, dtype: int64
   Accuracy: 72.5
   [[34 9]
   [13 24]]
             precision recall f1-score support
        0
              0.72
                      0.79
                             0.76
                                      43
              0.73
                      0.65
                             0.69
                                      37
   accuracy
                             0.73
                                      80
              0.73
                      0.72
   macro avg
                             0.72
                                      80
weighted avg
              0.73
                      0.72
                             0.72
                                      80
```

REGRESSION:

CODE:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
dataset = pd.read csv('Company data.csv')
X = dataset['Temperature'].values
y = dataset['Revenue'].values
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.05
# Fitting Decision Tree Regression to the dataset
from sklearn.tree import DecisionTreeRegressor
regressor = DecisionTreeRegressor()
regressor.fit(X train.reshape(-1,1), y train.reshape(-1,1))
y pred = regressor.predict(X test.reshape(-1,1))
y pred
df = pd.DataFrame({'Real Values':y test.reshape(-
1), 'Predicted Values':y pred.reshape(-1)})
df
# Visualising the Decision Tree Regression Results
X \text{ grid} = \text{np.arange}(\min(X), \max(X), 0.01)
X grid = X grid.reshape((len(X grid), 1))
plt.scatter(X_test, y_test, color = 'red')
plt.scatter(X_test, y_pred, color = 'green')
plt.title('Decision Tree Regression')
plt.xlabel('Temperature')
plt.ylabel('Revenue')
plt.show()
plt.plot(X_grid, regressor.predict(X_grid), color = 'black')
plt.title('Decision Tree Regression')
plt.xlabel('Temperature')
plt.ylabel('Revenue')
plt.show()
```

```
[ ] import numpy as np
     import matplotlib.pyplot as plt
     import pandas as pd
     dataset = pd.read_csv('Company_data.csv')
     X = dataset['Temperature'].values
     y = dataset['Revenue'].values
[ ] from sklearn.model_selection import train_test_split
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.05)
     # Fitting Decision Tree Regression to the dataset
     from sklearn.tree import DecisionTreeRegressor
     regressor = DecisionTreeRegressor()
     regressor.fit(X_train.reshape(-1,1), y_train.reshape(-1,1))
     DecisionTreeRegressor()
y_pred = regressor.predict(X_test.reshape(-1,1))
     y_pred
rray([800.2024937, 625.1901215, 612.1539491, 546.6938576, 774.1080813,
            145.6253019, 118.8121496, 641.0253891, 807.5412872, 696.7166402,
            665.6726764, 586.150568 , 534.7990284, 321.7500343, 402.3984607,
            646.2669458, 454.1892673, 246.7871609, 653.9867356, 827.6848313,
            682.7528689, 493.1154676, 594.8048712, 905.4776043, 537.6648006])
[ ] df = pd.DataFrame({'Real Values':y_test.reshape(-1), 'Predicted Values':y_pred.reshape(-1)})
   # Visualising the Decision Tree Regression Results
    X_{grid} = np.arange(min(X), max(X), 0.01)
    X_grid = X_grid.reshape((len(X_grid), 1))
    plt.scatter(X_test, y_test, color = 'red')
    plt.scatter(X_test, y_pred, color = 'green')
    plt.title('Decision Tree Regression')
    plt.xlabel('Temperature')
    plt.ylabel('Revenue')
    plt.show()
    plt.plot(X_grid, regressor.predict(X_grid), color = 'black')
    plt.title('Decision Tree Regression')
    plt.xlabel('Temperature')
    plt.ylabel('Revenue')
    plt.show()
```

OUTPUT:

0		Real Values	Predicted Values
D	0	828.296077	800.202494
	1	570.577875	625.190122
	2	625.846421	612.153949
	3	524.236115	546.693858
	4	750.444733	774.108081
	5	242.509855	145.625302
	6	131.657017	118.812150
	7	662.558990	641.025389
	8	809.672053	807.541287
	9	706.364904	696.716640
	10	628.453211	665.672676
	11	588.527551	586.150568
	12	531.742485	534.799028
	13	219.303993	321.750034
	14	467.446707	402.398461
	15	574.423310	646.266946
	16	473.604335	454.189267
	17	242.236208	246.787161
	18	654.129377	653.986736
	19	824.954357	827.684831
	20	618.235765	682.752869
	21	489.569090	493.115468
	22	540.977511	594.804871
	23	898.805423	905.477604
	24	545.903929	537.664801

