DATA WAREHOUSING AND DATA MINING MODEL LAB

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1.List the attributes and its type in a word Doc:

My dataset is fully Numeric.

2. Build decision tree classifier with Entropy criteria. Perform Prediction for test dataset using Entropy and print the results in the form of confusion matrix, accuracy and classification report. visualize the decision tree:

```
In [1]: import numpy as np
        import pandas as pd
        from sklearn.metrics import confusion_matrix
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.metrics import accuracy_score
        from sklearn.metrics import classification_report
        from sklearn.model_selection import train_test_split
        import warnings
        warnings.filterwarnings("ignore")
In [2]: def importdata():
           balance_data = pd.read_csv('balance-scale.csv')
            print ("Dataset Length: ", len(balance_data))
print ("Dataset Shape: ", balance_data.shape)
            print ("Dataset: ",balance_data.head())
            return balance_data
In [3]: def splitdataset(balance_data):
            # Separating the target variable
            X = balance_data.values[:, 1:5]
            Y = balance_data.values[:, 0]
            # Splitting the dataset into train and test
            X_train, X_test, y_train, y_test = train_test_split(
            X, Y, test_size = 0.3, random_state = 100)
            return X, Y, X_train, X_test, y_train, y_test
In [4]: def tarin_using_entropy(X_train, X_test, y_train):
            clf_entropy = DecisionTreeClassifier(
                    criterion = "entropy", random_state = 100,
                     max_depth = 3, min_samples_leaf = 5)
            clf_entropy.fit(X_train, y_train)
            return clf_entropy
```

```
In [5]: def prediction(X_test, clf_object):
                                        y_pred = clf_object.predict(X_test)
                                        print("Predicted values:")
                                        print(y_pred)
                                        return y_pred
              In [6]: 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 [7]: def main():
                                        data = importdata()
                                        X, Y, X_train, X_test, y_train, y_test = splitdataset(data)
                                        clf_entropy = tarin_using_entropy(X_train, X_test, y_train)
                                        print("Results Using Entropy:")
                                        y_pred_entropy = prediction(X_test, clf_entropy)
cal_accuracy(y_test, y_pred_entropy)
Jupyter Untitled Last Checkpoint: a few seconds ago (autosaved)
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v
           In [8]: main()
                         Dataset Length: 767
                         Dataset Length: 767
Dataset Shape: (767, 9)
Dataset; 6 148 72 35 0 33.6 0.627 50 1
0 1 85 66 29 0 26.6 0.351 31 0
1 8 183 64 0 0 23.3 0.672 32 1
2 1 89 66 23 94 28.1 0.167 21 0
3 0 137 40 35 168 43.1 2.288 33 1
4 5 116 74 0 0 25.6 0.201 30 0
                         Results Using Entropy:
Predicted values:
                         Predicted values:
[1. 1. 1. 1. 4. 0. 0. 1. 3. 0. 0. 0. 1. 0. 0. 0. 1. 0. 0. 5. 0. 1. 1. 0. 0. 0. 0. 5. 1. 0. 0. 0. 1. 3. 0. 0. 0. 1. 1. 0. 0. 0. 0. 3. 1. 1. 0. 0. 1. 5. 0. 0. 1. 1. 4. 1. 0. 1. 1. 0. 0. 3. 1. 1. 0. 0. 1. 5. 0. 0. 1. 1. 4. 3. 0. 0. 1. 0. 4. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 4. 0. 1. 0. 1. 1. 1. 5. 0. 1. 0. 4. 1. 1. 5. 1. 0. 3. 1. 1. 0. 0. 0. 0. 0. 0. 4. 0. 1. 0. 1. 1. 1. 3. 1. 1. 3. 1. 1. 0. 0. 1. 0. 1. 1. 1. 1. 1. 1. 0. 1. 0. 5. 1. 0. 5. 3. 1. 3. 0. 1. 0. 4. 0. 4. 1. 5. 1. 0. 0. 1. 1. 1. 1. 1. 0. 1. 0. 5. 1. 0. 5. 3. 1. 3. 0. 1. 0. 4. 0. 4. 1. 5. 1. 0. 0. 1. 1. 0. 0. 1. 0. 0. 1. 1. 1. 0. 0. 0. 1. 1. 0. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 0. 1. 1. 0. 0. 1. 0. 1. 0. 0. 1. 1. 0. 0. 1. 1. 0. 0. 0. 1. 1. 0. 0. 1. 0. 1. 0. 0. 1. 0. 1. 0. 0. 1. 0. 1. 0. 0. 1. 0. 1. 0. 0. 0. 0. 0. 3. 0. 3. 0. 3. 0. 1. 0. 1. 1. 1. 5. 0. 1. 1. 0. 1. 0. 1. 3. 1. 3. 3. 3. 1. 1. 0. 0. 1. 5. 1. 3. 4. 0. 0. 0.]

Confusion Matrix: [[22 15 0 3 3 1 0 0 0 0 0 0 0 0 0 0 0]
                         1. 5. 5. 1. 1. 0. 0. 1. 5. 1. 3. 4. 0. 0. 0.]

Confusion Matrix: [[22 15 0 3 3 1 0 0 0 0 0 0 0 0 0 0]

[17 21 0 4 0 2 0 0 0 0 0 0 0 0 0 0]

[6 13 0 0 1 0 0 0 0 0 0 0 0 0 0 0]

[10 8 0 2 2 1 0 0 0 0 0 0 0 0 0 0]

[7 2 0 1 2 1 0 0 0 0 0 0 0 0 0 0]
                              9 6
7 2
4 6
                         0 1 3
                                                         1
                                                              0 0
                                                                                                         0]
                          Report :
                                                                    precision
                                                                                          recall f1-score support
                                                                             0.50
                                                                                              0.30
                                          0.0
                                                           0.21
                                                          0.25
0.00
0.12
                                                                            0.48
0.00
0.09
                                         1.0
                                                                                              0.33
                                                                                                                    20
23
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                                          3.0
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13
15
                                          4.0
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12
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                                                                                               0.00
                                 accuracy
                                                                                              0.20
                                                                                                                  231
                         macro avg
weighted avg
                                                           0.05
                                                                             0.08
```