## 50-day50-decision-tree-iris

## November 22, 2023

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Day50 Decision Tree(Iris)
                                                        By: Loga Aswin
[2]: # Import Libraries
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
[3]: # Load Dataset
     df = pd.read_csv('/content/IRIS..csv')
    Exploratory Data Analysis (EDA):
[4]: df.head()
[4]:
       sepal_length sepal_width petal_length petal_width
                                                                  species
                 5.1
     0
                              3.5
                                            1.4
                                                         0.2 Iris-setosa
     1
                 4.9
                              3.0
                                            1.4
                                                         0.2 Iris-setosa
     2
                 4.7
                              3.2
                                            1.3
                                                         0.2 Iris-setosa
     3
                 4.6
                                            1.5
                              3.1
                                                         0.2 Iris-setosa
                 5.0
                              3.6
                                            1.4
                                                         0.2 Iris-setosa
[5]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 150 entries, 0 to 149
    Data columns (total 5 columns):
         Column
                       Non-Null Count Dtype
         -----
                       _____
                                       ____
                                       float64
     0
         sepal_length 150 non-null
         sepal_width
                       150 non-null
                                       float64
     1
     2
         petal_length 150 non-null
                                       float64
     3
         petal_width
                       150 non-null
                                       float64
         species
                                       object
                       150 non-null
    dtypes: float64(4), object(1)
    memory usage: 6.0+ KB
[6]: df.shape
[6]: (150, 5)
```

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[7]: df.isnull().sum()
 [7]: sepal_length
                      0
      sepal_width
                      0
      petal_length
      petal_width
                      0
      species
      dtype: int64
 [8]: df.describe()
 [8]:
             sepal_length
                           sepal_width petal_length petal_width
               150.000000
                            150.000000
                                           150.000000
                                                        150.000000
      count
                 5.843333
                              3.054000
                                             3.758667
                                                          1.198667
      mean
      std
                 0.828066
                              0.433594
                                             1.764420
                                                          0.763161
     min
                 4.300000
                              2,000000
                                             1.000000
                                                          0.100000
      25%
                 5.100000
                              2.800000
                                             1.600000
                                                          0.300000
      50%
                 5.800000
                              3.000000
                                             4.350000
                                                          1.300000
      75%
                 6.400000
                              3.300000
                                             5.100000
                                                          1.800000
                 7.900000
     max
                              4.400000
                                             6.900000
                                                          2.500000
[12]: #Seprated into dependent and independent variable
      X = df.drop("species",axis=1)
      y = df["species"]
[11]: #Splitting into training and testing
      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=_
       →1)
     Model Training:
[15]: from sklearn.tree import DecisionTreeClassifier
      classifier = DecisionTreeClassifier()
      classifier.fit(X_train, y_train)
[15]: DecisionTreeClassifier()
     Predict Output:
[17]: | y_pred = classifier.predict(X_test)
     Model Evaluation Metrics:
[18]: from sklearn.metrics import classification_report, accuracy_score,__
       ⇔confusion_matrix
```

```
report = classification_report(y_test, y_pred)
       print(report)
                          precision
                                         recall f1-score
                                                               support
           Iris-setosa
                                1.00
                                           1.00
                                                       1.00
                                                                     14
                                0.94
                                           0.94
                                                       0.94
      Iris-versicolor
                                                                     18
       Iris-virginica
                                0.92
                                           0.92
                                                       0.92
                                                                     13
                                                       0.96
                                                                     45
              accuracy
             macro avg
                                0.96
                                           0.96
                                                       0.96
                                                                     45
         weighted avg
                                0.96
                                           0.96
                                                       0.96
                                                                     45
[19]: matrix = confusion matrix(y test, y pred)
       print(matrix)
      [[14 0 0]
       [ 0 17 1]
       [ 0 1 12]]
[30]: accuracy = accuracy_score(y_test, y_pred)
       print((accuracy)*100)
      95.55555555556
      ** Plot Decision Tree :**
[31]: from sklearn import tree
       plt.figure(figsize=(15, 10))
       tree.plot_tree(model, filled=True)
[31]: [Text(0.4, 0.9, 'x[3] \le 0.8 \text{ ngini} = 0.665 \text{ nsamples} = 105 \text{ nvalue} = [36, 32, 32]
      37]'),
       Text(0.3, 0.7, 'gini = 0.0 \setminus samples = 36 \setminus value = [36, 0, 0]'),
        Text(0.5, 0.7, 'x[3] \le 1.65 \text{ ngini} = 0.497 \text{ nsamples} = 69 \text{ nvalue} = [0, 32, 1.65]
       37]'),
        Text(0.2, 0.5, 'x[2] \le 5.0 = 0.161 = 0.161 = 34 = [0, 31, 3]'),
        Text(0.1, 0.3, 'gini = 0.0 \setminus samples = 30 \setminus value = [0, 30, 0]'),
        Text(0.3, 0.3, 'x[0] \le 6.05 \cdot ngini = 0.375 \cdot nsamples = 4 \cdot nvalue = [0, 1, 3]'),
        Text(0.2, 0.1, 'gini = 0.0 \setminus samples = 1 \setminus value = [0, 1, 0]'),
        Text(0.4, 0.1, 'gini = 0.0 \setminus samples = 3 \setminus value = [0, 0, 3]'),
        Text(0.8, 0.5, 'x[2] \le 4.85 \cdot ngini = 0.056 \cdot nsamples = 35 \cdot nvalue = [0, 1, 34]'),
        Text(0.7, 0.3, 'x[1] \le 3.1 \text{ ngini} = 0.375 \text{ nsamples} = 4 \text{ nvalue} = [0, 1, 3]'),
        Text(0.6, 0.1, 'gini = 0.0 \setminus samples = 3 \setminus value = [0, 0, 3]'),
        Text(0.8, 0.1, 'gini = 0.0 \land samples = 1 \land value = [0, 1, 0]'),
```

 $Text(0.9, 0.3, 'gini = 0.0 \setminus samples = 31 \setminus value = [0, 0, 31]')]$ 

