- 1 import pandas as pd
- 2 import matplotlib.pyplot as plt
- 3 import numpy as np
- 4
- 1 import pandas as pd
- 2 from sklearn.tree import DecisionTreeClassifier
- from sklearn.model_selection import train_test_split
- 4 from sklearn import metrics
- 5 from sklearn import tree
- 6 from sklearn.metrics import confusion_matrix
- 7 from sklearn.metrics import accuracy_score
- 8 from sklearn.tree import plot_tree
- 9 from sklearn.metrics import classification_report
- 10 import graphviz
- 1 train=pd.read_csv("/content/iris.csv")
- 2 train.head()

С

| | sepallength | sepalwidth | petallength | petalwidth | class1 |
|---|-------------|------------|-------------|------------|-------------|
| 0 | 5.1 | 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 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |
| | | | | | |

1 train.describe()

| | sepallength | sepalwidth | petallength | petalwidth |
|-------|-------------|------------|-------------|------------|
| count | 150.000000 | 150.000000 | 150.000000 | 150.000000 |
| mean | 5.843333 | 3.054000 | 3.758667 | 1.198667 |
| 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 |
| max | 7.900000 | 4.400000 | 6.900000 | 2.500000 |
| | | | | |

- 1 from sklearn import preprocessing
- 2
- 3 # label_encoder object knows how to understand word labels.
- 4 label_encoder = preprocessing.LabelEncoder()

1 a = np.array(train)

```
5
6 # Encode labels in column 'species'.
7 train['class1']= label_encoder.fit_transform(train['class1'])
8
9 train['class1'].unique()
10 train.head()
```

```
sepallength sepalwidth petallength petalwidth class1
0
             5.1
                           3.5
                                          1.4
                                                        0.2
                                                                    0
             4.9
                           3.0
                                          1.4
                                                        0.2
1
                                                                    0
2
             4.7
                           3.2
                                          1.3
                                                        0.2
                                                                    0
3
             4.6
                           3.1
                                          1.5
                                                        0.2
                                                                    0
             5.0
4
                           3.6
                                          1.4
                                                        0.2
                                                                    \cap
```

```
2 X=train.drop(columns=['class1'])
 3 Y=train['class1']
 1 from sklearn.svm import SVC
 2 X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size = 0.9, random_state =1)
3 model = SVC(kernel='linear')
4 model=model.fit(X_train, y_train)
5 y_pred = model.predict(X_test)
6 print("Predicted values:")
 7 print(y pred)
8 print("Confusion Matrix: ",confusion_matrix(y_test, y_pred))
9 print("Accuracy : ",accuracy_score(y_test,y_pred)*100)
10 print("Report : ",classification_report(y_test, y_pred))
11 # fitting x samples and y classes
12 model.fit(X, Y)
13
14 model.predict([[1,1,1,1]])
15
```

```
Predicted values:
[0\ 1\ 1\ 0\ 2\ 1\ 2\ 0\ 0\ 2\ 1\ 0\ 2\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 0\ 2\ 1\ 0\ 0\ 1\ 2\ 1\ 2\ 1\ 2\ 2\ 0\ 1
 \begin{smallmatrix} 0 & 1 & 2 & 2 & 0 & 1 & 2 & 1 & 2 & 0 & 0 & 0 & 1 & 0 & 0 & 2 & 2 & 2 & 2 & 1 & 1 & 2 & 1 & 0 & 2 & 1 & 0 & 2 & 2 & 2 & 1 & 1 & 2 & 2 & 0 \\ \end{smallmatrix}
 200202021020101100101101
Confusion Matrix: [[47 0 0]
 [ 0 42 2]
 [ 0 4 40]]
Accuracy: 95.555555555556
Report:
                         precision
                                     recall f1-score
                                                           support
           0
                   1.00
                              1.00
                                         1.00
                                                      47
                    0.91
                                                      44
           1
                              0.95
                                         0.93
           2
                    0.95
                              0.91
                                         0.93
                                                      44
                                         0.96
                                                     135
    accuracy
                    0.96
                              0.95
                                         0.95
                                                     135
   macro avg
```

0.96

135

0.96

0.96

weighted avg

