

prediction using Decision Tree Algorithm

Create the Decision Tree classifier and visualize it grahically,the purpose is if we feed any new data to this classifier.it would be able to preadict the right classs accordingly

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Link for dataset:<https://bit.ly/3kXTdox>

Importing the Dependencies

```
In [5]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

In [9]: df = pd.read_csv("C:/Users/sakshi itnare/Downloads/Iris.csv")

In [10]: df.head()

Out[10]:
   sepal_length  sepal_width  petal_length  petal_width  species
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

In [11]: df.shape
Out[11]: (150, 5)

In [12]: df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   sepal_length    150 non-null    float64
1   sepal_width     150 non-null    float64
2   petal_length    150 non-null    float64
3   petal_width     150 non-null    float64
4   species         150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB

In [13]: df.isna().sum()
Out[13]:
sepal_length    0
sepal_width     0
petal_length    0
petal_width     0
species         0
dtype: int64

In [15]: df.describe()

Out[15]:
   sepal_length  sepal_width  petal_length  petal_width
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
```

Visualizing Data

```
In [17]: #plotting distribution of data
plt.figure(figsize=(10,5))
sns.scatterplot(data=df, s=100,alpha=0.7)
plt.grid()
plt.show()

In [18]: #Extracting independent & dependent variable
x = df.iloc[:, :-1]
y = df.iloc[:, -1]

In [19]: #train test splitting
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30,random_state=0)
```

Decision Tree Algorithm

```
In [21]: from sklearn.tree import DecisionTreeClassifier
classifier = DecisionTreeClassifier(criterion="entropy", max_depth = 4)
classifier.fit(x_train,y_train)

Out[21]: DecisionTreeClassifier(criterion='entropy', max_depth=4)

In [22]: y_pred = classifier.predict(x_test)
y_pred

Out[22]: array(['Iris-virginica', 'Iris-versicolor', 'Iris-setosa',
      'Iris-virginica', 'Iris-setosa', 'Iris-virginica', 'Iris-setosa',
      'Iris-versicolor', 'Iris-versicolor', 'Iris-versicolor',
      'Iris-virginica', 'Iris-versicolor', 'Iris-versicolor',
      'Iris-versicolor', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
      'Iris-virginica', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
      'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-versicolor',
      'Iris-versicolor', 'Iris-setosa', 'Iris-virginica',
      'Iris-versicolor', 'Iris-setosa', 'Iris-virginica',
      'Iris-virginica', 'Iris-versicolor', 'Iris-setosa',
      'Iris-virginica', 'Iris-versicolor', 'Iris-versicolor',
      'Iris-virginica', 'Iris-setosa', 'Iris-virginica', 'Iris-setosa',
      'Iris-setosa'], dtype=object)

In [23]: from sklearn import metrics
metrics.accuracy_score(y_test,y_pred)

Out[23]: 0.9777777777777777

In [24]: from sklearn.metrics import confusion_matrix
matrix=confusion_matrix(y_test,y_pred)
matrix

Out[24]: array([[16,  0,  0],
      [ 0, 17,  1],
      [ 0,  0, 11]], dtype=int64)
```

Text representation of the Decision Tree

```
In [ ]: import matplotlib.image as mpimg
from sklearn import tree
from sklearn.tree import export_grapviz

In [9]: text_rep = tree.export_text
print(text_rep)

<function export_text at 0x0000021DFDCB2940>

In [19]: from io import StringIO
import matplotlib.image as mpimg
from sklearn import tree
from sklearn.tree import export_graphviz

In [21]: text_rep = tree.export_text
print(text_rep)

<function export_text at 0x0000021DFDCB2940>
```