

Chapter 6: DecisionTreeClassifier

```
In [1]: # from google.colab import drive
# drive.mount("/content/gdrive", force_remount=True)
# %cd '/content/gdrive/My Drive/LDS6_MachineLearning/practice/Chapter6_Decision_
```

```
In [ ]: from sklearn.tree import DecisionTreeClassifier
from sklearn import datasets
from IPython.display import Image
from sklearn import tree
import pydotplus
import pandas as pd
```

```
In [ ]: iris = pd.read_excel("Iris.xls")
iris.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   sepallength     150 non-null    float64
1   sepalwidth      150 non-null    float64
2   petallength     150 non-null    float64
3   petalwidth      150 non-null    float64
4   iris            150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
In [ ]: X = iris[['sepallength', 'sepalwidth', 'petallength', 'petalwidth']] # numeric
y = iris['iris'] # Loai hoa
```

```
In [ ]: X.head()
```

```
Out[5]:
```

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

```
In [ ]: y[:5]
```

```
Out[6]: 0    Iris-setosa
        1    Iris-setosa
        2    Iris-setosa
        3    Iris-setosa
        4    Iris-setosa
        Name: iris, dtype: object
```

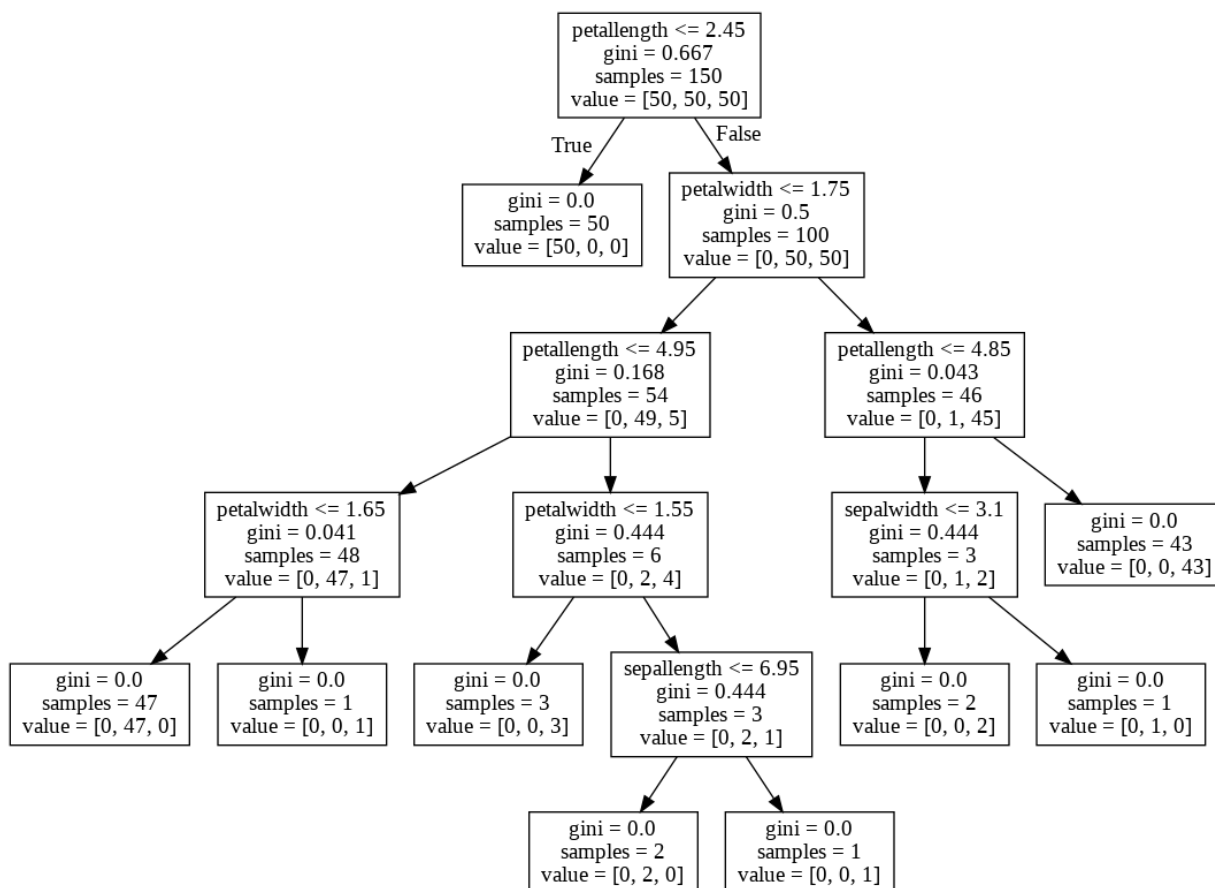
```
In [ ]: # https://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeClassifier.html
        # co the thay the bang criterion='entropy' tuy bai toan
        clf = DecisionTreeClassifier()
        model = clf.fit(X, y)
```

Note:

- Gini is intended for continuous attributes and Entropy is for attributes that occur in classes.
- Gini is to minimize misclassification
- Entropy is for exploratory analysis
- Entropy is a little slower to compute

```
In [ ]: dot_data = tree.export_graphviz(clf, out_file=None,
                                         feature_names=X.columns
                                         )
        graph = pydotplus.graph_from_dot_data(dot_data)
        Image(graph.create_png())
```

Out[8]:



```
In [ ]: # Save Decision Tree Image To File
# Create PDF
graph.write_pdf(path + "iris_1510.pdf")
# Create PNG
graph.write_png(path + "iris_1510.png")
```

Out[9]: True

```
In [ ]: # Link: https://chrisalbon.com/machine\_learning/trees\_and\_forests/visualize\_a\_dec
```

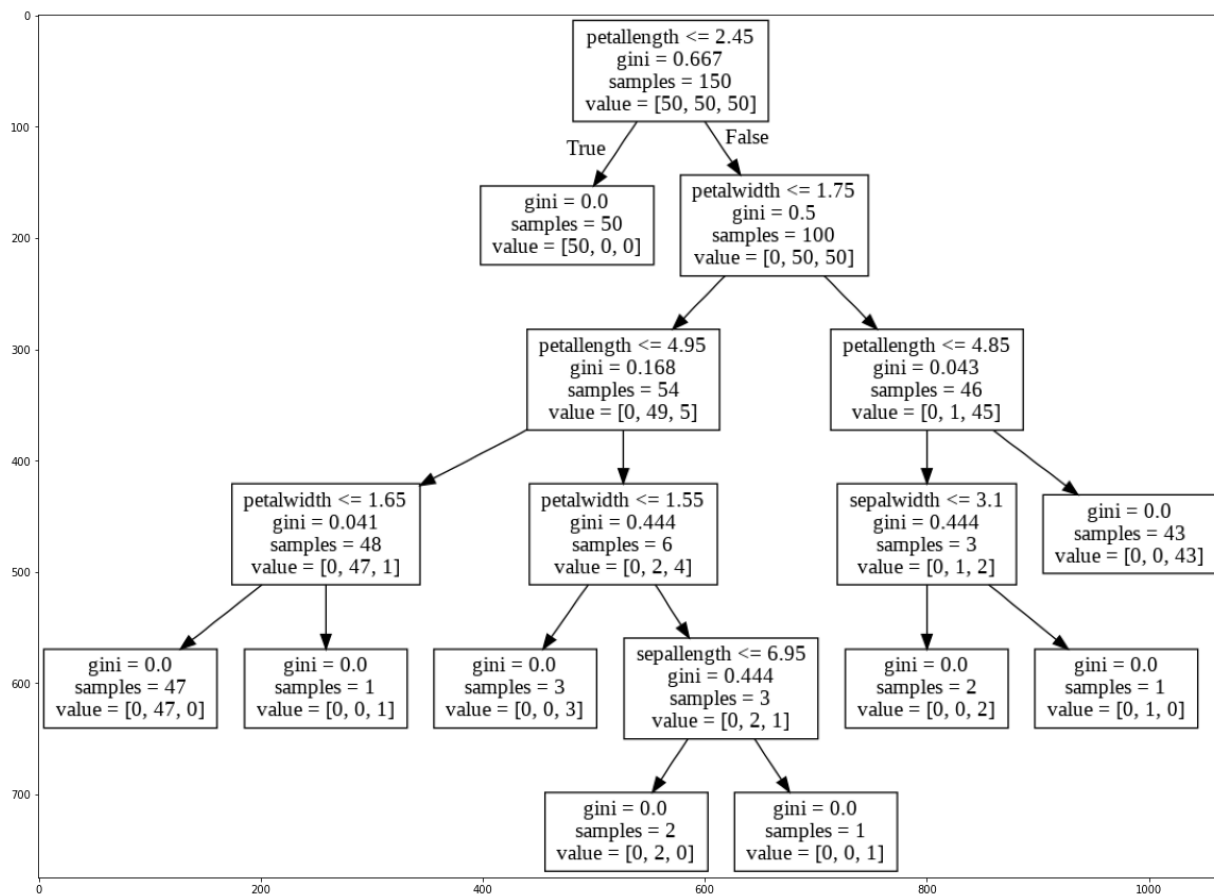
```
In [ ]: # Another solution to save and show decision tree
# Save to file
with open(path + "iris_1510.txt", "w") as f:
    f = tree.export_graphviz(clf, out_file=f,
                             feature_names=X.columns,
                             class_names=iris.iris)

    print("Đã xuất file")
```

Đã xuất file

```
In [ ]: # dùng để mở và xem kết quả
# https://dreampuf.github.io/GraphvizOnline/
# hoặc http://viz-js.com/
import matplotlib.pyplot as plt
import imageio
photo_data = imageio.imread("iris_1510.png")
plt.figure(figsize = (20, 20))
plt.imshow(photo_data)
```

Out[12]: <matplotlib.image.AxesImage at 0x7fcff81fed30>



```
In [ ]: model.score(X, y)
```

Out[13]: 1.0

```
In [ ]: # Make new prediction
import numpy as np
X_new = np.array([[4.7, 3.2, 1.3, 0.2],
                  [6.6, 3. , 4.4, 1.4],
                  [5.9, 3. , 5.1, 1.8]])
```

```
In [ ]: yhat_new = model.predict(X_new)
yhat_new
```

Out[15]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)