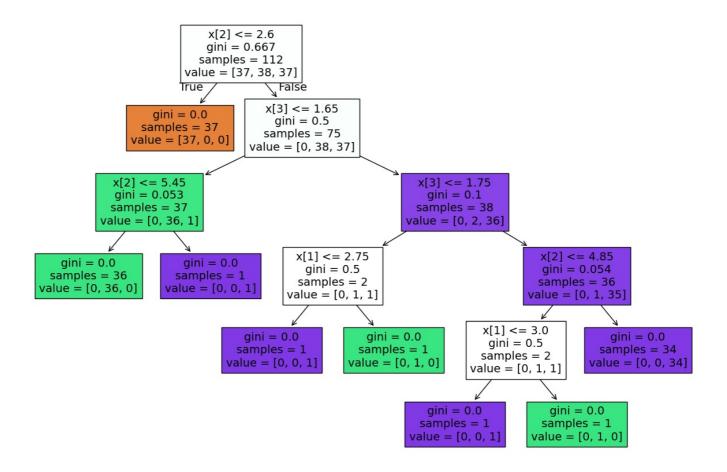
Implement Decision Tree Algorithm

```
In [1]: import pandas as pd
                                          import matplotlib.pyplot as plt
                                          %matplotlib inline
     In [4]: from sklearn.datasets import load iris
     In [6]: iris=load iris()
     In [8]: from sklearn.model selection import train test split
                                          x train,x test,y train,y test=train test split(iris.data,iris.target,random state=100,stratify=iris.target,test
 In [10]: from sklearn.metrics import accuracy_score
In [12]: from sklearn.tree import DecisionTreeClassifier
In [14]: classifier=DecisionTreeClassifier()
In [16]: from sklearn.tree import DecisionTreeClassifier
In [18]: classifier=DecisionTreeClassifier()
In [20]: classifier.fit(x train,y train)
Out[20]: -
                                                           DecisionTreeClassifier 🔍 🧷
                                          DecisionTreeClassifier()
In [22]: pred=classifier.predict(x test)
                                          accuracy_score(y_test,pred)
Out[22]: 0.9473684210526315
In [24]: from sklearn import tree
                                          plt.figure(figsize=(15,10))
                                          tree.plot_tree(classifier,filled=True)
Out[24]: [Text(0.31818181818181818182, 0.916666666666666, 'x[2] <= 2.6 \nqini = 0.667 \nsamples = 112 \nvalue = [37, 38, 37]')
                                               Text(0.22727272727272727, 0.75, 'gini = 0.0 \nsamples = 37 \nvalue = [37, 0, 0]'),
                                                Text(0.27272727272727, 0.83333333333333, 'True '),
                                                Text(0.4090909090909091, 0.75, 'x[3] \le 1.65 = 0.5 = 0.5 \le 75 = 0.5 \le 1.65 \le 1
                                              Text(0.363636363636365, 0.833333333333333, 'False'),
Text(0.181818181818182, 0.583333333333333, 'False'),
Text(0.09090909090909090, 0.416666666666667, 'gini = 0.0\nsamples = 36\nvalue = [0, 36, 0]'),
Text(0.27272727272727, 0.4166666666666667, 'gini = 0.0\nsamples = 1\nvalue = [0, 0, 1]'),
Text(0.63636363636364, 0.5833333333333, 'x[3] <= 1.75\ngini = 0.1\nsamples = 38\nvalue = [0, 2, 36]'),
                                                Text(0.4545454545454545453, 0.416666666666666667, 'x[1] \le 2.75 \cdot ngini = 0.5 \cdot nsamples = 2 \cdot nvalue = [0, 1, 1]'),
                                                Text(0.36363636363636365, 0.25, 'gini = 0.0 \nsamples = 1 \nvalue = [0, 0, 1]'), Text(0.54545454545454545, 0.25, 'gini = 0.0 \nsamples = 1 \nvalue = [0, 1, 0]'),
                                                Text(0.8181818181818182, \ 0.4166666666666667, \ 'x[2] <= 4.85 \\ \ | ini = 0.054 \\ \ | ini = 36 \\ \ | ini = 3
                                                Text(0.72727272727273, 0.25, 'x[1] <= 3.0 \\ line = 0.5 \\ line = 2 \\ line = [0, 1, 1]'), \\ line = (0, 1, 1]'), \\ line = (0, 1, 1)', \\ 
                                                Text(0.8181818181818182, 0.083333333333333333, 'gini = 0.0 = 1 = [0, 1, 0]'),
                                                Text(0.90909090909091, 0.25, 'gini = 0.0 \nsamples = 34 \nvalue = [0, 0, 34]')
In [26]: plt.show()
```



Conclusion: Decision tree learned how to identify iris flowers based on their features like petal length and petal width. Most of leaf node are pure and model got up to 92% accuracy

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js