

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: df = pd.read_csv("E:\\Data Science\\8.Machine Learning Algorithms\\2.Classification\\penguins_size.csv")
```

```
In [3]: df.head()
```

Out[3]:

	species	island	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g
0	Adelie	Torgersen	39.1	18.7	181.0	3750.0
1	Adelie	Torgersen	39.5	17.4	186.0	3800.0
2	Adelie	Torgersen	40.3	18.0	195.0	3250.0
3	Adelie	Torgersen	NaN	NaN	NaN	NaN
4	Adelie	Torgersen	36.7	19.3	193.0	3450.0

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 344 entries, 0 to 343
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   species               344 non-null    object
1   island                344 non-null    object
2   culmen_length_mm      342 non-null    float64
3   culmen_depth_mm       342 non-null    float64
4   flipper_length_mm     342 non-null    float64
5   body_mass_g           342 non-null    float64
6   sex                   334 non-null    object
dtypes: float64(4), object(3)
memory usage: 18.9+ KB
```

```
In [5]: df.shape
```

Out[5]: (344, 7)

```
In [6]: df.isnull().sum()
```

```
Out[6]: species          0
island                  0
culmen_length_mm        2
culmen_depth_mm         2
flipper_length_mm        2
body_mass_g             2
sex                     10
dtype: int64
```

```
In [7]: df = df.dropna()
```

```
In [8]: df.shape
```

```
Out[8]: (334, 7)
```

```
In [9]: df.head()
```

```
Out[9]:
```

	species	island	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g
0	Adelie	Torgersen	39.1	18.7	181.0	3750.0
1	Adelie	Torgersen	39.5	17.4	186.0	3800.0
2	Adelie	Torgersen	40.3	18.0	195.0	3250.0
4	Adelie	Torgersen	36.7	19.3	193.0	3450.0
5	Adelie	Torgersen	39.3	20.6	190.0	3650.0



```
In [10]: df.describe(include="all")
```

```
Out[10]:
```

	species	island	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g
count	334	334	334.000000	334.000000	334.000000	334.000000
unique	3	3	NaN	NaN	NaN	NaN
top	Adelie	Biscoe	NaN	NaN	NaN	NaN
freq	146	164	NaN	NaN	NaN	NaN
mean	NaN	NaN	43.994311	17.160479	201.014970	4209.05688
std	NaN	NaN	5.460521	1.967909	14.022175	804.83612
min	NaN	NaN	32.100000	13.100000	172.000000	2700.00000
25%	NaN	NaN	39.500000	15.600000	190.000000	3550.00000
50%	NaN	NaN	44.500000	17.300000	197.000000	4050.00000
75%	NaN	NaN	48.575000	18.700000	213.000000	4793.75000
max	NaN	NaN	59.600000	21.500000	231.000000	6300.00000



```
In [11]: df["sex"].unique()
```

```
Out[11]: array(['MALE', 'FEMALE', '.'], dtype=object)
```

```
In [14]: df = df[df['sex'] != '.']
```

```
In [15]: df.shape
```

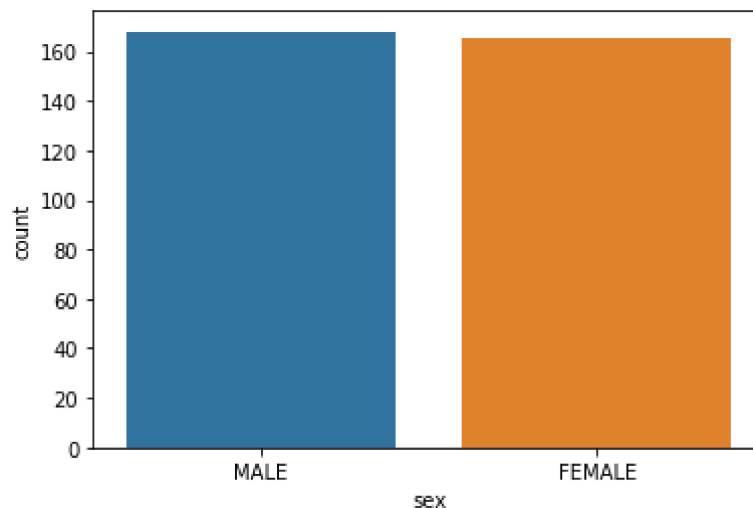
```
Out[15]: (333, 7)
```

```
In [39]: df.nunique()
```

```
Out[39]: species          3
island          3
culmen_length_mm  163
culmen_depth_mm   79
flipper_length_mm  54
body_mass_g      93
sex              2
dtype: int64
```

```
In [16]: sns.countplot(data = df,x = "sex")
Male,Female = df["sex"].value_counts()
print("Number of Male:",Male)
print("Number of Female:",Female)
plt.show()
```

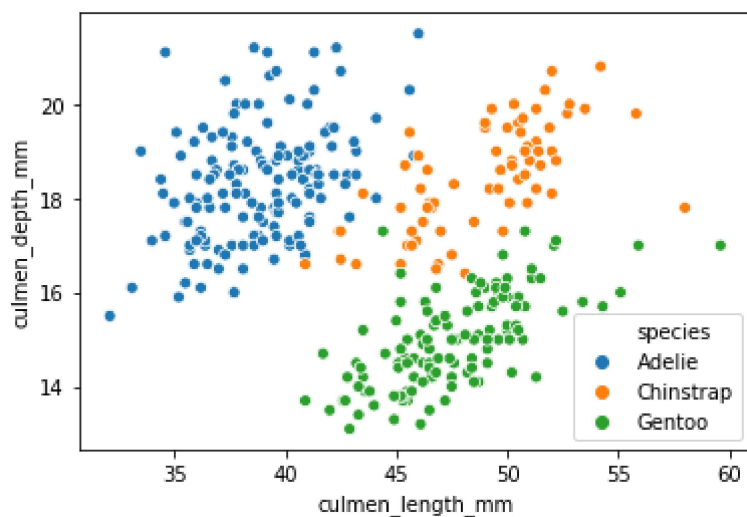
```
Number of Male: 168
Number of Female: 165
```



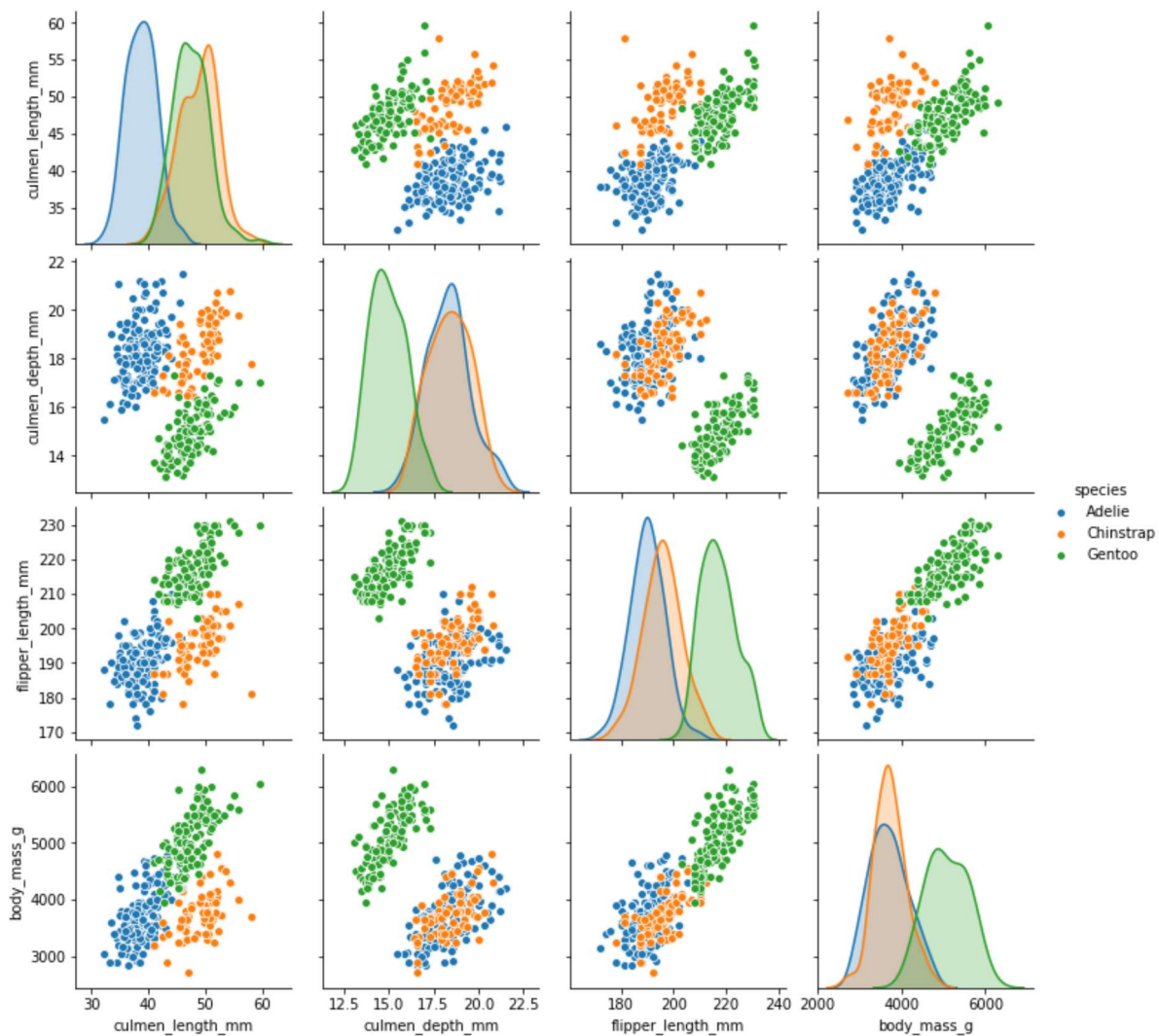
```
In [17]: df["island"].unique()
```

```
Out[17]: array(['Torgersen', 'Biscoe', 'Dream'], dtype=object)
```

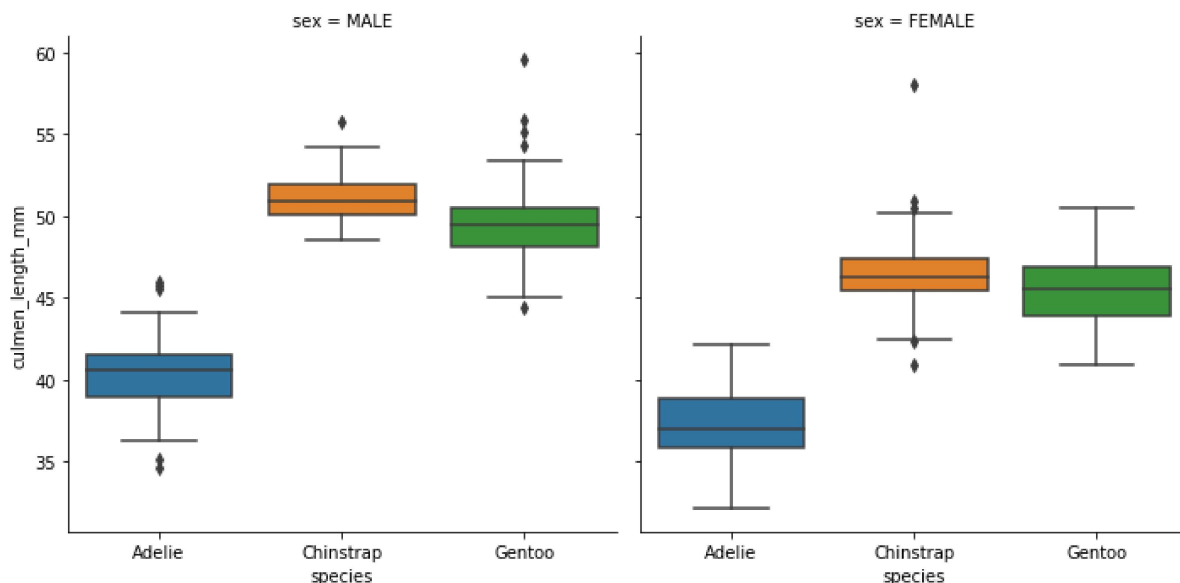
```
In [18]: sns.scatterplot(x="culmen_length_mm",y="culmen_depth_mm",data = df, hue= "species")
plt.show()
```



```
In [19]: sns.pairplot(data=df,hue="species")
plt.show()
```



```
In [20]: sns.catplot(data=df,x="species",y="culmen_length_mm",kind="box",col = "sex")
plt.show()
```



```
In [21]: pd.get_dummies(df.drop('species',axis=1),drop_first=True)
```

Out[21]:

	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g	island_Dream	isl
0	39.1	18.7	181.0	3750.0	0	
1	39.5	17.4	186.0	3800.0	0	
2	40.3	18.0	195.0	3250.0	0	
4	36.7	19.3	193.0	3450.0	0	
5	39.3	20.6	190.0	3650.0	0	
...	...	...	...	...	...	...
338	47.2	13.7	214.0	4925.0	0	
340	46.8	14.3	215.0	4850.0	0	
341	50.4	15.7	222.0	5750.0	0	
342	45.2	14.8	212.0	5200.0	0	
343	49.9	16.1	213.0	5400.0	0	

333 rows × 7 columns



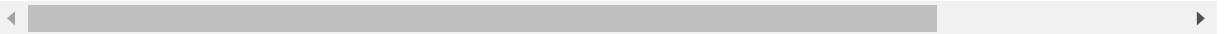
```
In [22]: X = pd.get_dummies(df.drop("species",axis=1),drop_first=True)
y = df["species"]
```

In [23]: X

Out[23]:

	culmen_length_mm	culmen_depth_mm	flipper_length_mm	body_mass_g	island_Dream	isl
0	39.1	18.7	181.0	3750.0	0	
1	39.5	17.4	186.0	3800.0	0	
2	40.3	18.0	195.0	3250.0	0	
4	36.7	19.3	193.0	3450.0	0	
5	39.3	20.6	190.0	3650.0	0	
...	...	...	...	...	...	...
338	47.2	13.7	214.0	4925.0	0	
340	46.8	14.3	215.0	4850.0	0	
341	50.4	15.7	222.0	5750.0	0	
342	45.2	14.8	212.0	5200.0	0	
343	49.9	16.1	213.0	5400.0	0	

333 rows × 7 columns



In [24]: y

```
Out[24]: 0    Adelie
1    Adelie
2    Adelie
4    Adelie
5    Adelie
...
338  Gentoo
340  Gentoo
341  Gentoo
342  Gentoo
343  Gentoo
Name: species, Length: 333, dtype: object
```

```
In [25]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=101)
```

```
In [26]: from sklearn.tree import DecisionTreeClassifier
model = DecisionTreeClassifier()
model.fit(X_train,y_train)
```

```
Out[26]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='gini',
                                max_depth=None, max_features=None, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, presort='deprecated',
                                random_state=None, splitter='best')
```

```
In [27]: base_pred = model.predict(X_test)
```

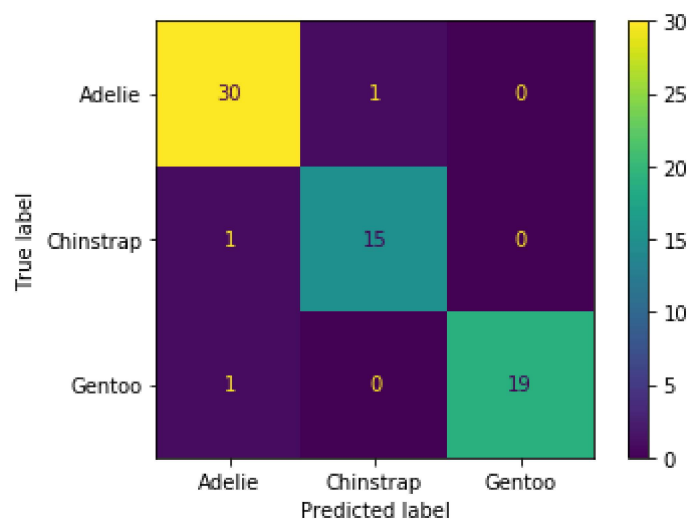
```
In [30]: from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
print("Accuracy Score:", accuracy_score(y_test, base_pred))
```

Accuracy Score: 0.9552238805970149

```
In [32]: print("Confusion Matrix\n", confusion_matrix(y_test, base_pred))
```

Confusion Matrix  
[[30 1 0]  
[ 1 15 0]  
[ 1 0 19]]

```
In [35]: from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(model, X_test, y_test)
plt.show()
```



```
In [37]: print(classification_report(y_test, base_pred))
```

	precision	recall	f1-score	support
Adelie	0.94	0.97	0.95	31
Chinstrap	0.94	0.94	0.94	16
Gentoo	1.00	0.95	0.97	20
accuracy			0.96	67
macro avg	0.96	0.95	0.95	67
weighted avg	0.96	0.96	0.96	67

```
In [40]: model.feature_importances_
```

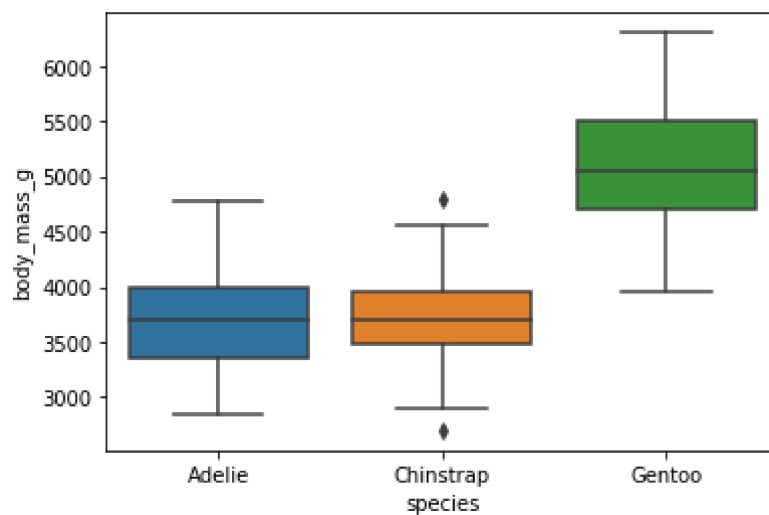
```
Out[40]: array([0.35128085, 0.07088022, 0.54456291, 0.          , 0.03327601,
                0.          , 0.          ])
```

```
In [41]: pd.DataFrame(index= X.columns, data=model.feature_importances_,columns=["Feature Importaces"])
```

Out[41]:

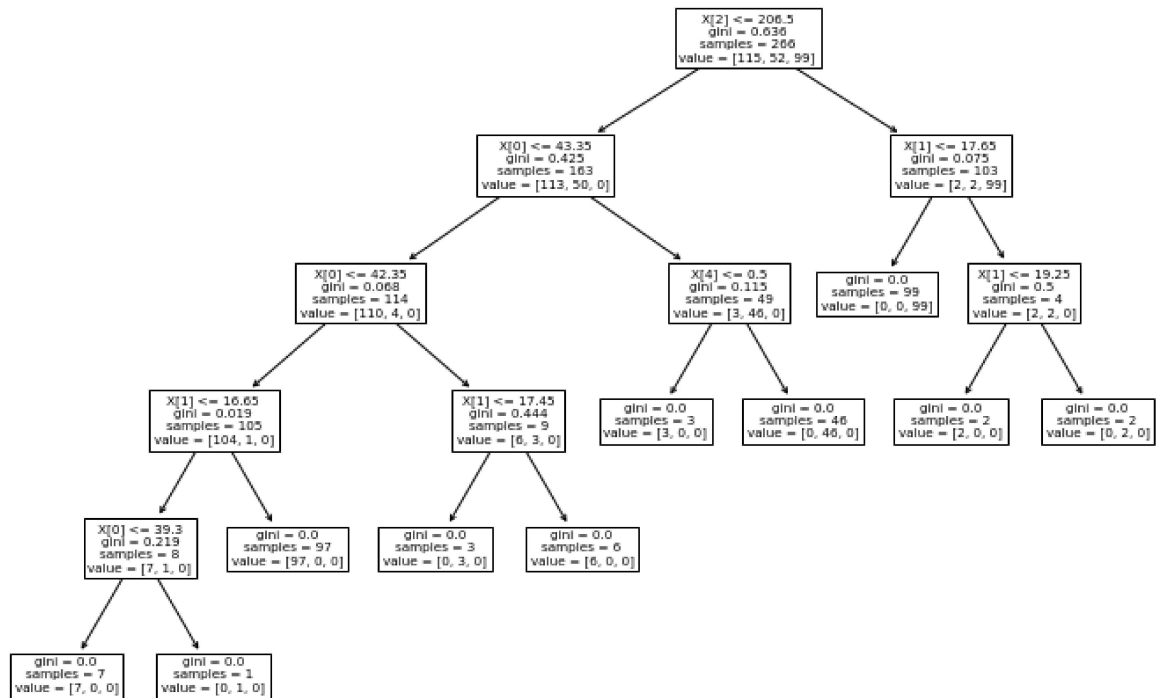
Feature Importaces	
culmen_length_mm	0.351281
culmen_depth_mm	0.070880
flipper_length_mm	0.544563
body_mass_g	0.000000
island_Dream	0.033276
island_Torgersen	0.000000
sex_MALE	0.000000

```
In [43]: sns.boxplot(x="species",y='body_mass_g',data=df)  
plt.show()
```

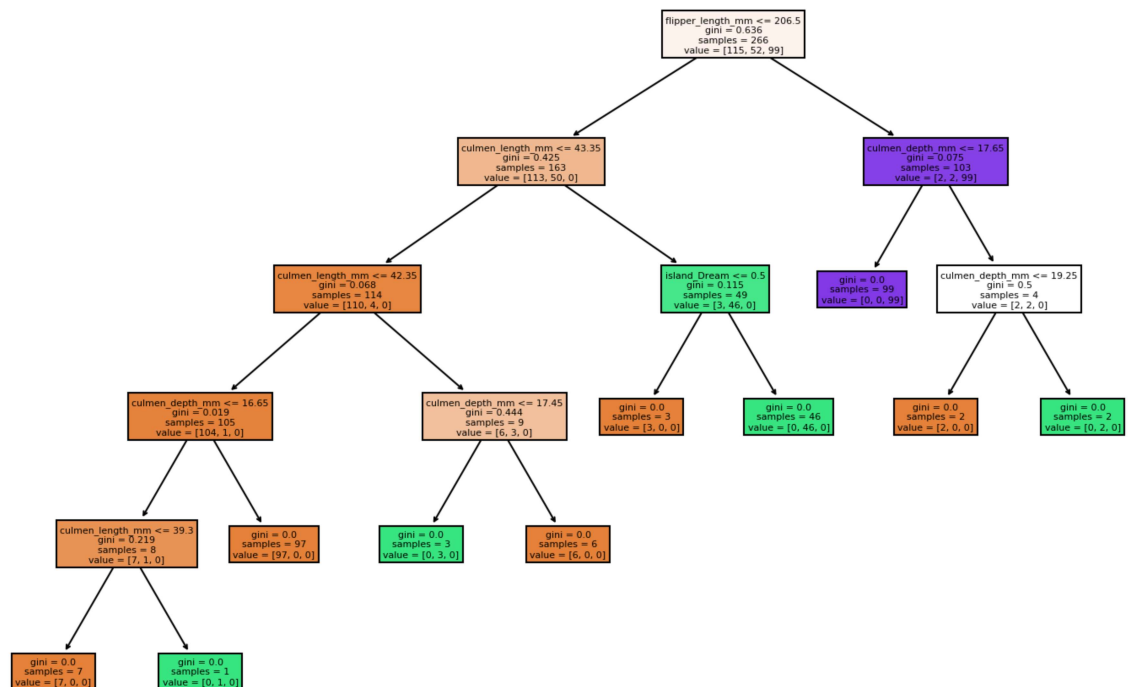




```
In [46]: from sklearn.tree import plot_tree
plt.figure(figsize=(12,8))
plot_tree(model)
plt.show()
```



```
In [48]: plt.figure(figsize=(12,8),dpi=150)
plot_tree(model,filled=True,feature_names=X.columns)
plt.show()
```



```
In [51]: def report_model(model):  
        model_preds = model.predict(X_test)  
        print(classification_report(y_test,model_preds))  
        print('\n')  
        plt.figure(figsize=(12,8),dpi=150)  
        plot_tree(model,filled=True,feature_names = X.columns)
```

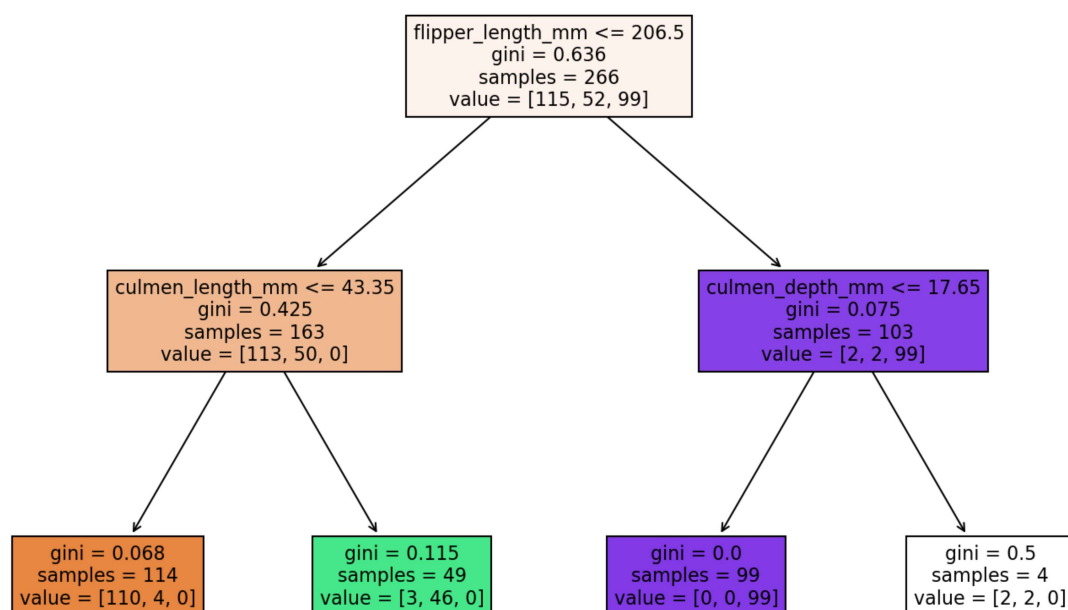
```
In [52]: #Understanding Hyperparameters in DT
```

```
In [57]: pruned_tree = DecisionTreeClassifier(max_depth=2)  
        pruned_tree.fit(X_train,y_train)
```

```
Out[57]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='gini',  
                                max_depth=2, max_features=None, max_leaf_nodes=None,  
                                min_impurity_decrease=0.0, min_impurity_split=None,  
                                min_samples_leaf=1, min_samples_split=2,  
                                min_weight_fraction_leaf=0.0, presort='deprecated',  
                                random_state=None, splitter='best')
```

```
In [58]: report_model(pruned_tree)
```

	precision	recall	f1-score	support
Adelie	0.88	0.97	0.92	31
Chinstrap	0.86	0.75	0.80	16
Gentoo	1.00	0.95	0.97	20
accuracy			0.91	67
macro avg	0.91	0.89	0.90	67
weighted avg	0.91	0.91	0.91	67

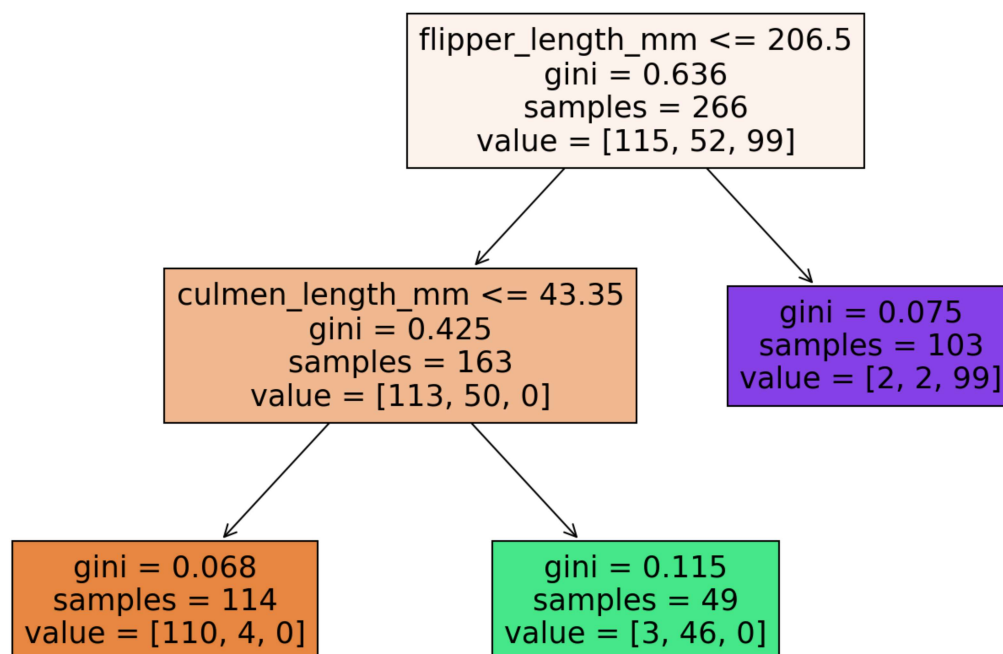


```
In [59]: pruned_tree = DecisionTreeClassifier(max_leaf_nodes=3)
pruned_tree.fit(X_train,y_train)
```

```
Out[59]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='gini',
max_depth=None, max_features=None, max_leaf_nodes=3,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, presort='deprecated',
random_state=None, splitter='best')
```

```
In [60]: report_model(pruned_tree)
```

	precision	recall	f1-score	support
Adelie	0.97	0.97	0.97	31
Chinstrap	0.86	0.75	0.80	16
Gentoo	0.86	0.95	0.90	20
accuracy			0.91	67
macro avg	0.90	0.89	0.89	67
weighted avg	0.91	0.91	0.91	67



```
In [62]: entropy_tree = DecisionTreeClassifier(criterion = "entropy")
entropy_tree.fit(X_train,y_train)
```

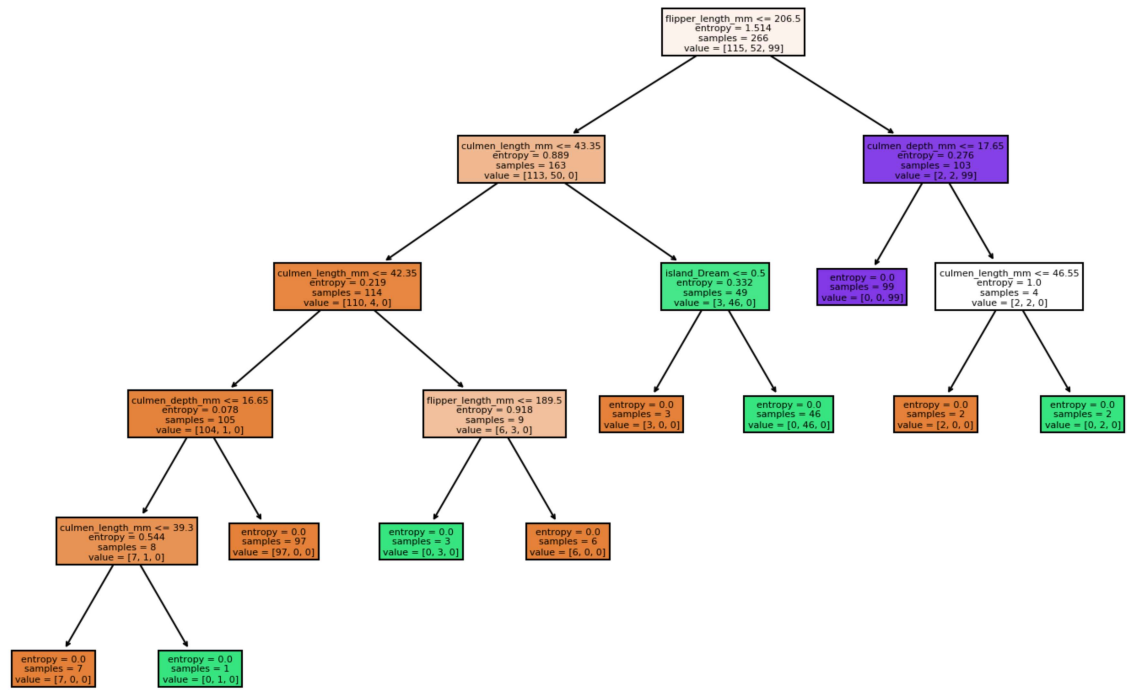
```
Out[62]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='entropy',
                                max_depth=None, max_features=None, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, presort='deprecated',
                                random_state=None, splitter='best')
```

```
In [63]: DecisionTreeClassifier(criterion = "entropy")
```

```
Out[63]: DecisionTreeClassifier(ccp_alpha=0.0, class_weight=None, criterion='entropy',
                                max_depth=None, max_features=None, max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, presort='deprecated',
                                random_state=None, splitter='best')
```

```
In [65]: report_model(entropy_tree)
```

	precision	recall	f1-score	support
Adelie	0.97	0.97	0.97	31
Chinstrap	0.94	1.00	0.97	16
Gentoo	1.00	0.95	0.97	20
accuracy			0.97	67
macro avg	0.97	0.97	0.97	67
weighted avg	0.97	0.97	0.97	67



```
In [ ]:
```