

```
In [2]: import pandas as pd
import numpy as np
df=pd.read_csv('IRIS.csv')
```

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

```
Out[3]:
```

	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 [4]: df.tail()
```

```
Out[4]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

```
In [5]: 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 [6]: df.describe()
```

```
Out[6]:
```

	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

```
In [7]: df.isnull().sum()
```

```
Out[7]: sepal_length    0
sepal_width          0
petal_length         0
petal_width          0
species              0
dtype: int64
```

```
In [8]: import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [9]: df['species']=df['species'].astype('category')
df['species']=df['species'].cat.codes
```

```
In [10]: df.head()
```

Out[10]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

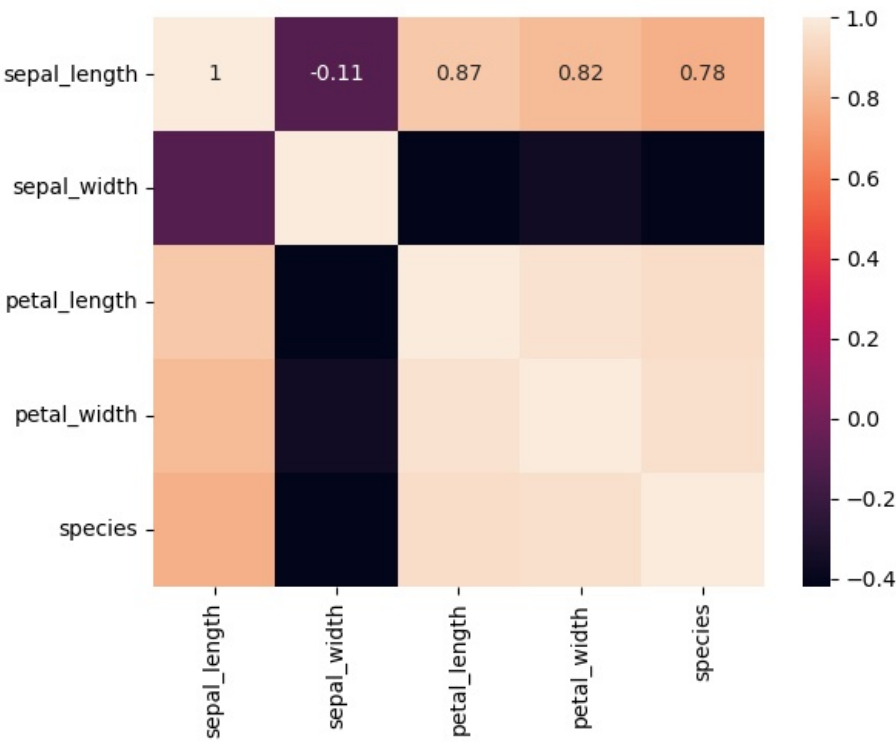
In [11]: df.tail()

Out[11]:

	sepal_length	sepal_width	petal_length	petal_width	species
145	6.7	3.0	5.2	2.3	2
146	6.3	2.5	5.0	1.9	2
147	6.5	3.0	5.2	2.0	2
148	6.2	3.4	5.4	2.3	2
149	5.9	3.0	5.1	1.8	2

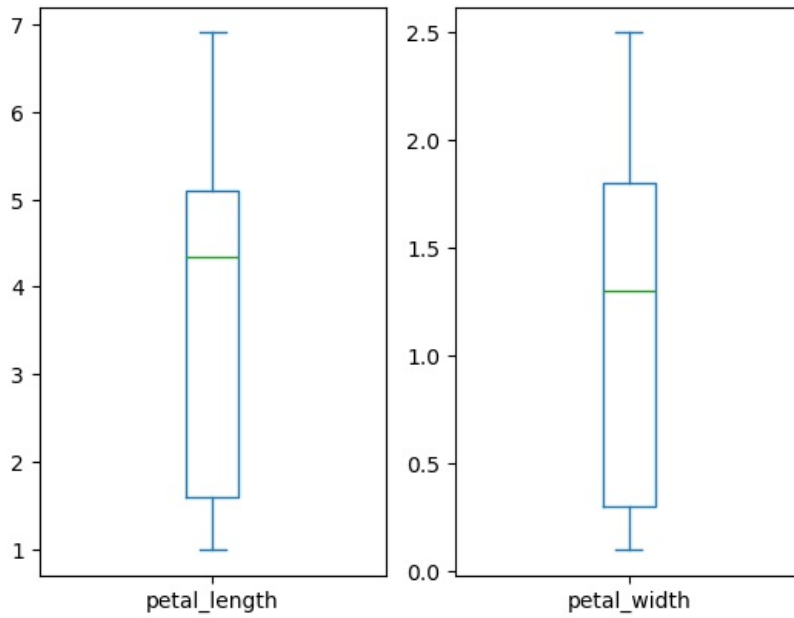
In [12]: correaltion_matrix=df.corr()
sns.heatmap(data=correaltion_matrix,annot=True)

Out[12]: <Axes: >



In [13]: plt.subplot(1,2,1)
df['petal_length'].plot.box()
plt.subplot(1,2,2)
df['petal_width'].plot.box()

Out[13]: <Axes: >



```
In [14]: x=df.loc[:,['petal_length','petal_width']]
         y=df['species']
```

```
In [15]: 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)
```

```
In [16]: from sklearn.naive_bayes import GaussianNB
```

```
In [17]: model=GaussianNB()
         model.fit(x_train,y_train)
```

```
Out[17]: ▼ GaussianNB
         GaussianNB()
```

```
In [18]: y_pred=model.predict(x_test)
```

```
In [19]: y_test.head()
```

```
Out[19]: 134    2
          8     0
          17    0
          106   2
          116   2
          Name: species, dtype: int8
```

```
In [20]: y_pred
```

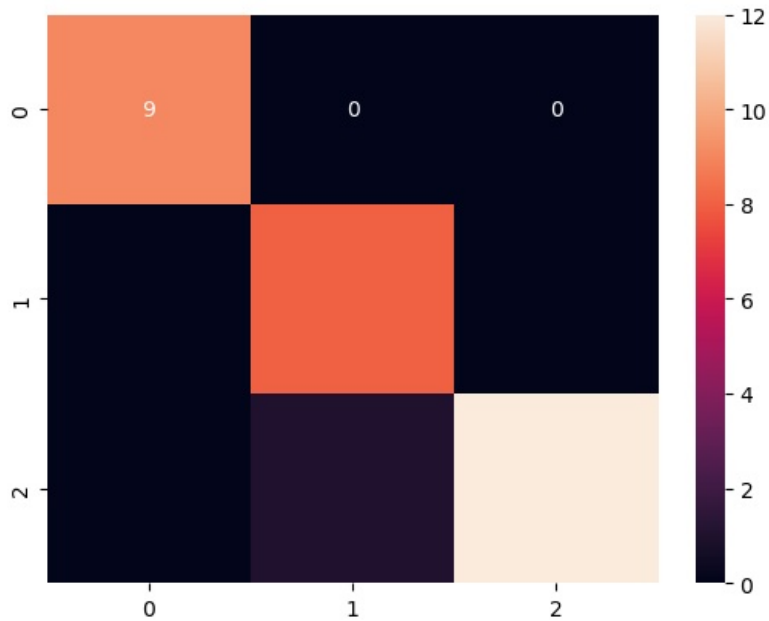
```
Out[20]: array([2, 0, 0, 1, 2, 2, 1, 2, 2, 1, 2, 2, 0, 1, 1, 1, 0, 2, 2, 0, 0, 2,
                1, 2, 0, 0, 2, 1, 0, 1], dtype=int8)
```

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In [ ]: p
```

```
In [22]: from sklearn.metrics import confusion_matrix
         confu_matrix=confusion_matrix(y_test,y_pred)
         print(confu_matrix)
```

```
[[ 9  0  0]
 [ 0  8  0]
 [ 0  1 12]]
```

```
In [25]: dataplot=sns.heatmap(data=confu_matrix,annot=True)
plt.show()
```



```
In [41]: accuracy=(9+8+12)/(9+8+12+1)
print('Accuracy based on the confusion matrix is :',(accuracy*100),' %')
```

Accuracy based on the confusion matrix is : 96.66666666666667 %

```
In [38]: from sklearn.metrics import mean_squared_error
mse=mean_squared_error(y_test,y_pred)
print("Mean square error is :",(mse*100).round(2),' %')
```

Mean square error is : 3.33 %

```
In [39]: rmse=np.sqrt(mse)
print("Root mean square error is :",(rmse*100).round(2),' %')
```

Root mean square error is : 18.26 %

```
In [43]: r2_score=model.score(x_test,y_test)
print("R2 score is :",r2_score)
```

R2 score is : 0.9666666666666667

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