assignment-6

April 8, 2024

import libraries

[2]:
```
import numpy as np
import pandas as pd
import seaborn as sns
```

import dataset

[3]:
```
df=sns.load_dataset('iris')
```

understand the dataset

[4]:
```
df.shape
```

[4]:
```
(150, 5)
```

[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
```

[6]:
```
df.head()
```

[6]:

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

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

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

[8]:
```
df.describe()
```

[8]:

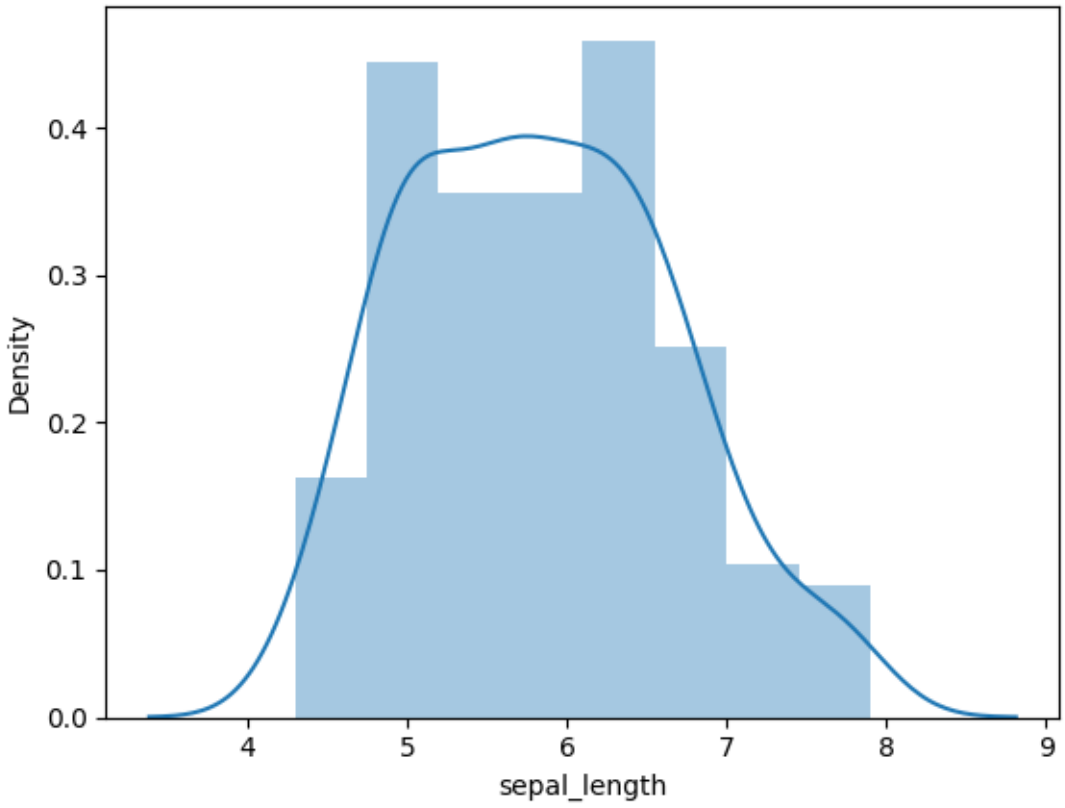
	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.057333	3.758000	1.199333
std	0.828066	0.435866	1.765298	0.762238
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

exploratory data analysis

[11]:
```
sns.distplot(df['sepal_length'])
```

```
<ipython-input-11-89138501e731>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

  sns.distplot(df['sepal_length'])
```

[11]:
```
<Axes: xlabel='sepal_length', ylabel='Density'>
```



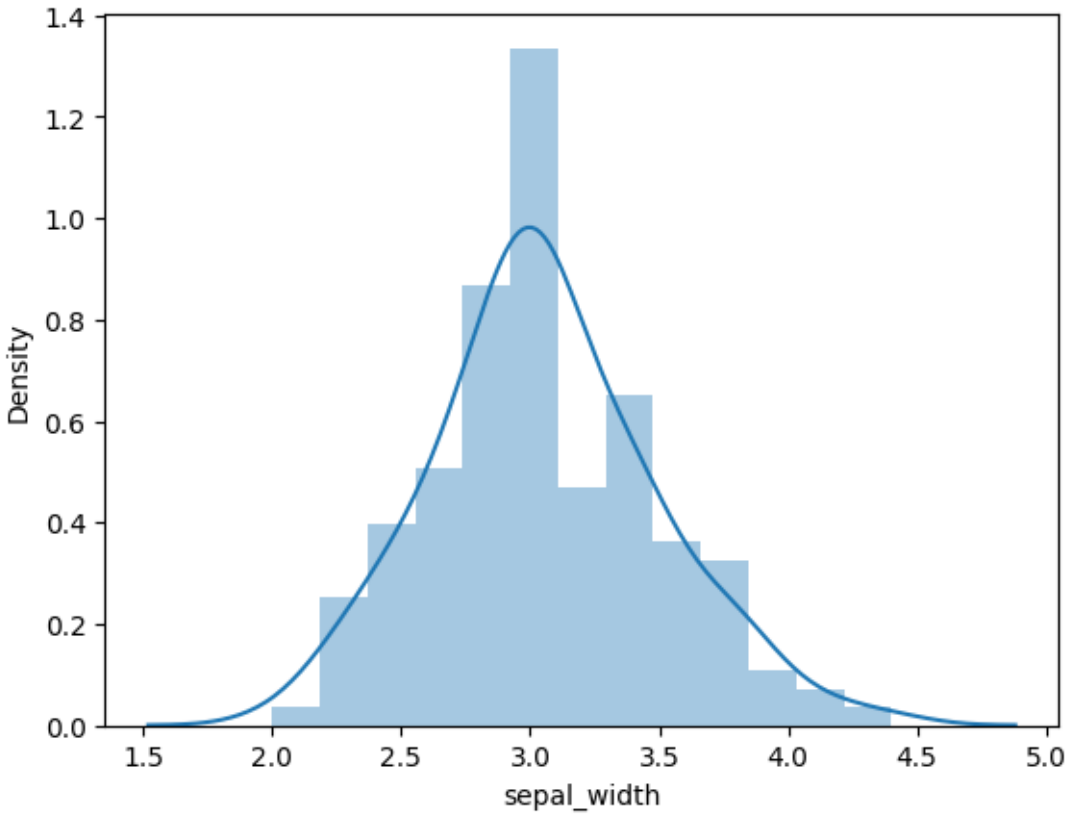
[12]:
```
df['sepal_length'].skew()
```

[12]: 0.3149109566369728

[13]:
```
sns.distplot(df['sepal_width'])
```

```
<ipython-input-13-6c237bf4ae06>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

  sns.distplot(df['sepal_width'])
```

[13]: <Axes: xlabel='sepal_width', ylabel='Density'>



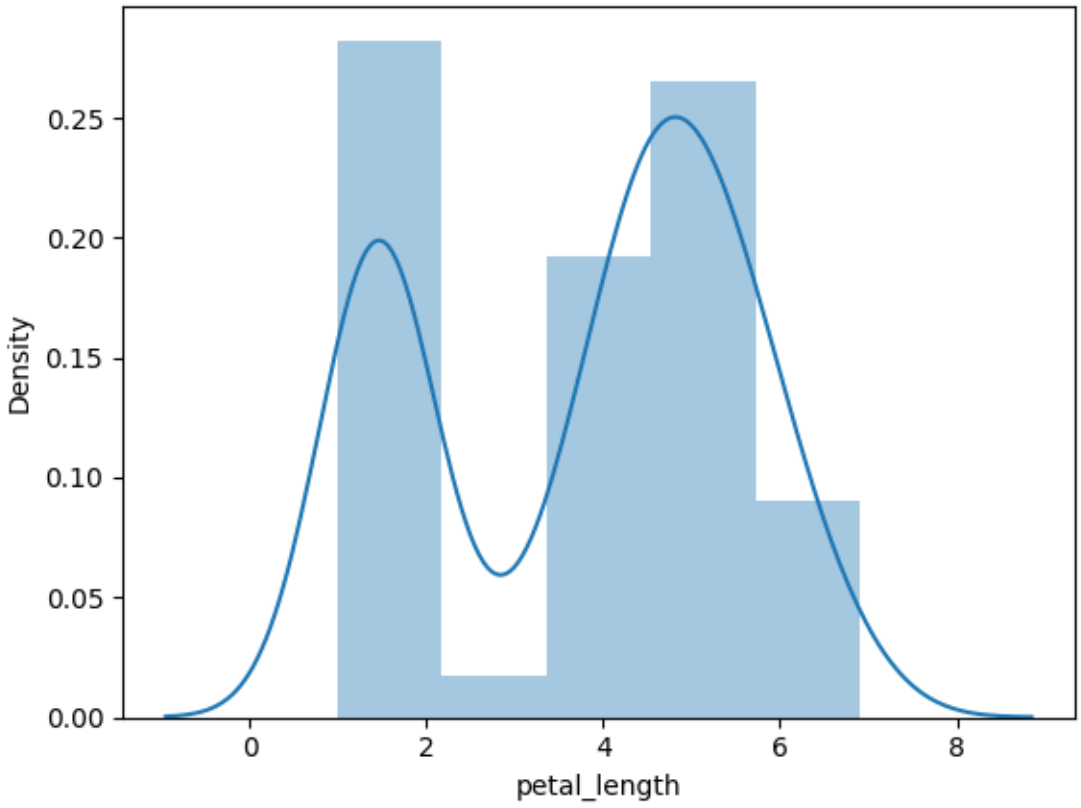
[14]:
```
df['sepal_width'].skew()
```

[14]: 0.31896566471359966

[15]:
```
sns.distplot(df['petal_length'])
```

```
<ipython-input-15-e0ee5e61121d>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

  sns.distplot(df['petal_length'])
```

[15]: <Axes: xlabel='petal_length', ylabel='Density'>



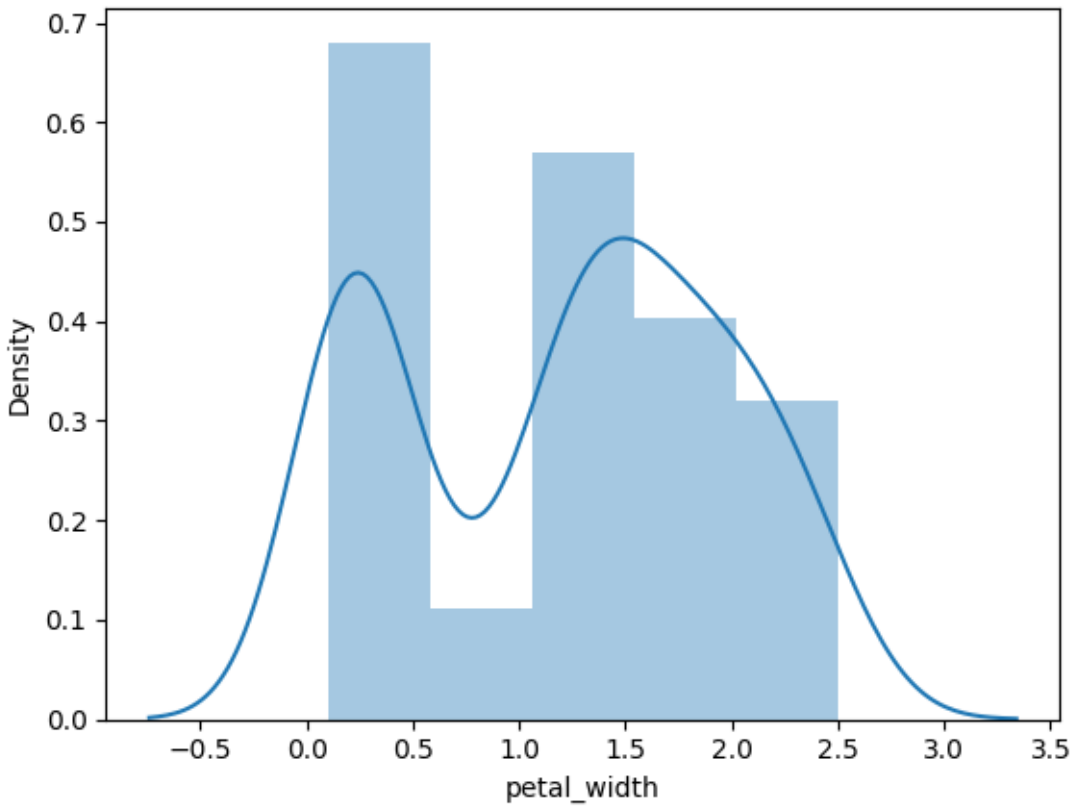
[16]:
```
df['petal_length'].skew()
```

[16]:
```
-0.27488417975101276
```

[17]:
```
sns.distplot(df['petal_width'])
```

```
<ipython-input-17-2b0be1ef4ca2>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

  sns.distplot(df['petal_width'])
```

[17]:
```
<Axes: xlabel='petal_width', ylabel='Density'>
```



[18]:
```
df['petal_width'].skew()
```

[18]:
```
-0.10296674764898116
```

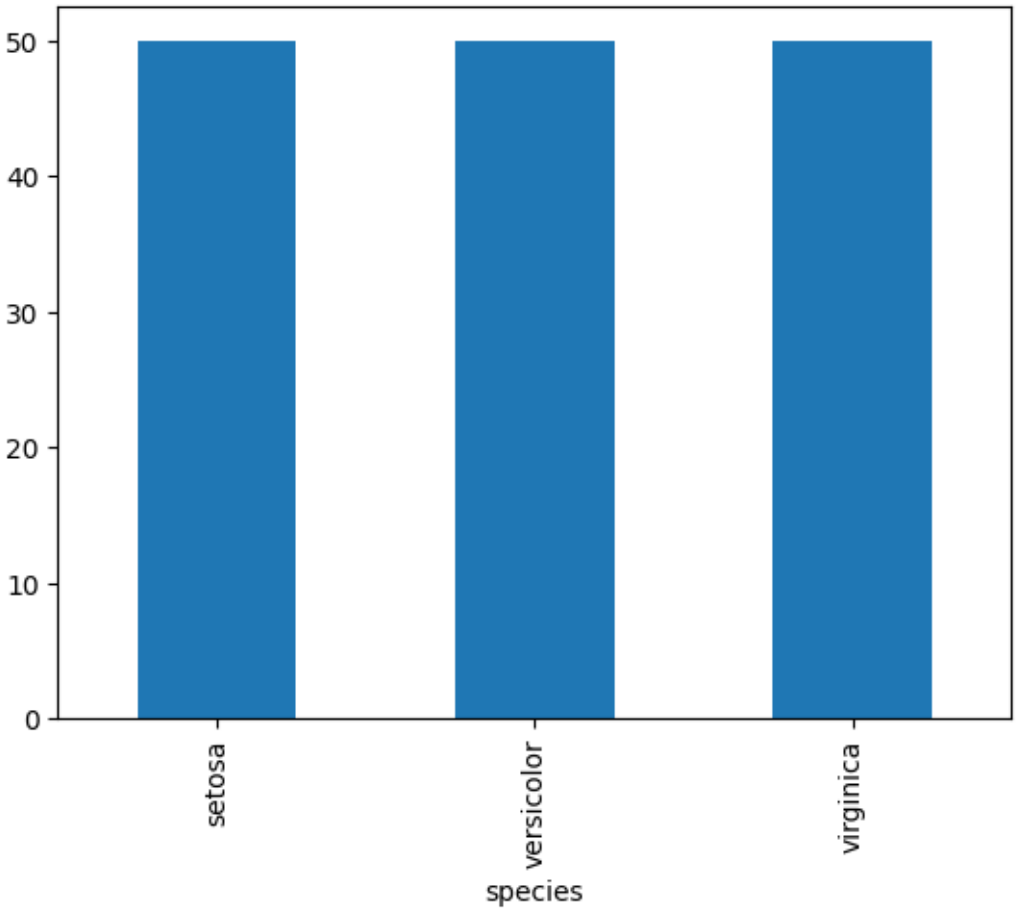
[19]:
```
#all features are normally distributed(skewness value between -0.5 and 0.5)
```

[21]:
```
df['species'].value_counts()
```

[21]:
```
species
setosa        50
versicolor    50
virginica     50
Name: count, dtype: int64
```

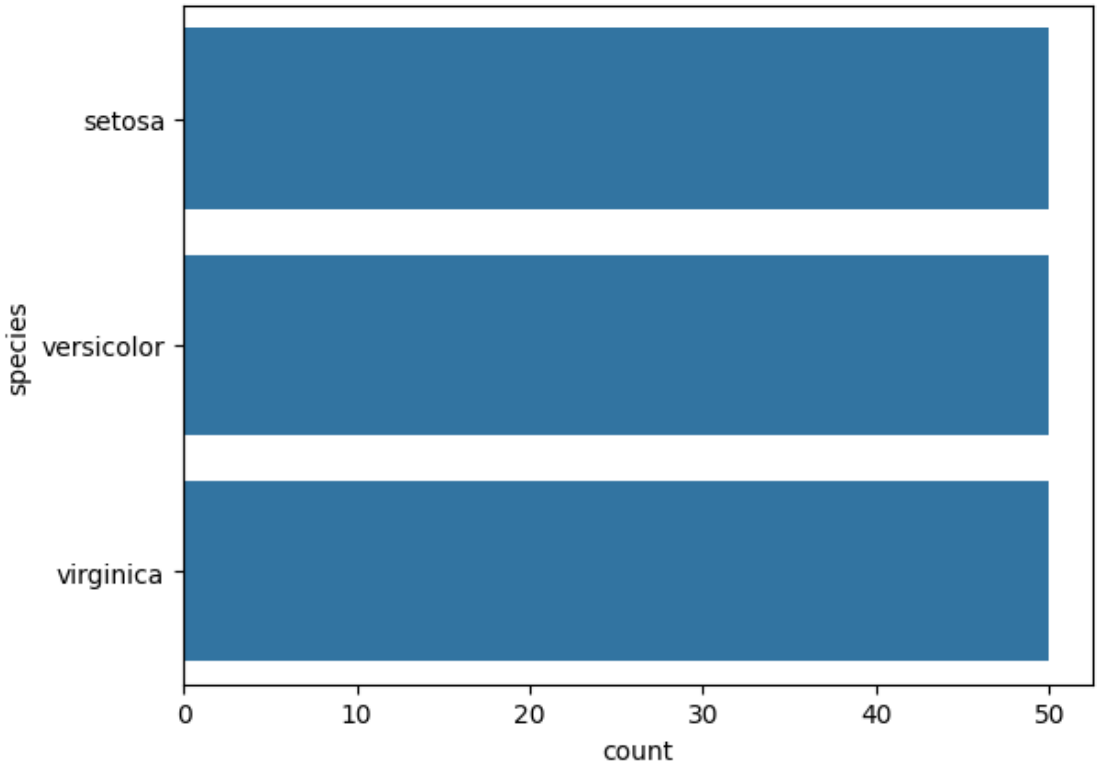
[22]:
```
df['species'].value_counts().plot(kind='bar')
```

[22]:
```
<Axes: xlabel='species'>
```



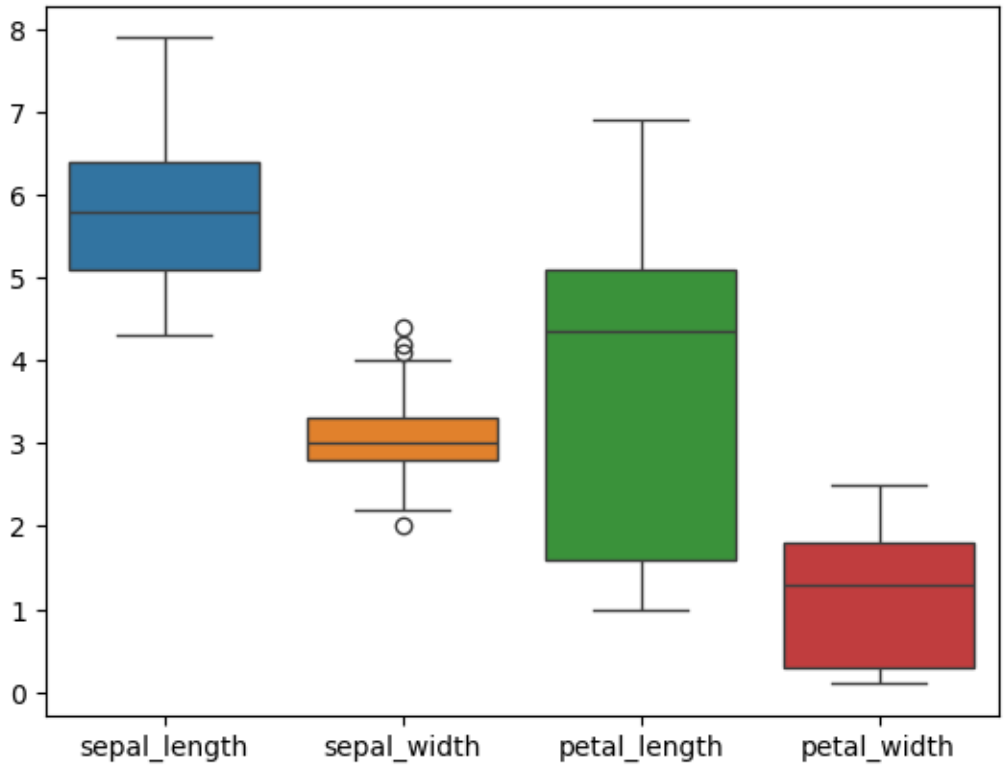
[26]:
```
sns.countplot(df['species'])
```

[26]:
```
<Axes: xlabel='count', ylabel='species'>
```



[29]:
```
sns.boxplot(df)
```

[29]: <Axes: >



separate the dataset into x and y where y is the target variable

[30]:
```
x=df.iloc[:,:-1]
```

[31]:
```
x
```

[31]:

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

[150 rows x 4 columns]

[32]:
```
y=df.iloc[:,-1]
```

[33]:
```
y
```

[33]:
```
0         setosa
1         setosa
2         setosa
3         setosa
4         setosa
          ...
145    virginica
146    virginica
147    virginica
148    virginica
149    virginica
Name: species, Length: 150, dtype: object
```

splitting dataset into training and testing

[34]:
```
from sklearn.model_selection import train_test_split
```

[35]:
```
x_train, x_test, y_train, y_test=train_test_split(x,y,train_size=0.
↪8,random_state=0)
```

[36]:
```
x_train.shape
```

[36]:
```
(120, 4)
```

[37]:
```
x_test.shape
```

[37]:
```
(30, 4)
```

[38]:
```
y_train.shape
```

[38]:
```
(120,)
```

[39]:
```
y_test.shape
```

[39]:
```
(30,)
```

model building

[41]:
```
#build naive bayes classifier model
```

[40]:
```
from sklearn.naive_bayes import GaussianNB
```

[45]:
```
classifier=GaussianNB()
```

[47]:
```
#train the model
classifier.fit(x_train,y_train)
```

[47]:
```
GaussianNB()
```

[48]:
```
#testing the model
y_pred=classifier.predict(x_test)
```

[52]:
```
result=pd.DataFrame({'actual':y_test,'prediction':y_pred})
```

[53]:
```
result
```

[53]:

	actual	prediction
114	virginica	virginica
62	versicolor	versicolor
33	setosa	setosa
107	virginica	virginica
7	setosa	setosa
100	virginica	virginica
40	setosa	setosa
86	versicolor	versicolor
76	versicolor	versicolor
71	versicolor	versicolor
134	virginica	versicolor
51	versicolor	versicolor
73	versicolor	versicolor
54	versicolor	versicolor
63	versicolor	versicolor
37	setosa	setosa
78	versicolor	versicolor
90	versicolor	versicolor
45	setosa	setosa
16	setosa	setosa
121	virginica	virginica
66	versicolor	versicolor
24	setosa	setosa
8	setosa	setosa
126	virginica	virginica
22	setosa	setosa
44	setosa	setosa
97	versicolor	versicolor
93	versicolor	versicolor
26	setosa	setosa

evaluate the model

[59]:
```
from sklearn.metrics import␣
↪confusion_matrix,accuracy_score,precision_score,recall_score,fbeta_score,classification_rep
```

[61]:
```
cm=confusion_matrix(y_test,y_pred)
```

[62]:
```
cm
```

[62]:
```
array([[11,  0,  0],
       [ 0, 13,  0],
       [ 0,  1,  5]])
```

[63]:
```
cm=confusion_matrix(y_pred,y_test)
```

[64]:
```
cm
```

[64]:
```
array([[11,  0,  0],
       [ 0, 13,  1],
       [ 0,  0,  5]])
```

[65]:
```
accuracy_score(y_test,y_pred)
```

[65]:
```
0.9666666666666667
```

[81]:
```
precision_score(y_test,y_pred,average='micro')
```

[81]:
```
0.9666666666666667
```

[80]:
```
recall_score(y_test,y_pred,average='micro')
```

[80]:
```
0.9666666666666667
```

[79]:
```
fbeta_score(y_test,y_pred,beta=0.5,average='micro')
```

[79]:
```
0.9666666666666666
```

[82]:
```
print(classification_report(y_test,y_pred))
```

```
              precision    recall  f1-score   support

      setosa       1.00      1.00      1.00        11
  versicolor       0.93      1.00      0.96        13
   virginica       1.00      0.83      0.91         6

    accuracy                           0.97        30
   macro avg       0.98      0.94      0.96        30
weighted avg       0.97      0.97      0.97        30
```

[ ]: