Name : Suraj Sawant

Roll No: TEB38

Practical No: 3 (B)

In [1]:	import pandas as pd
In [2]:	from sklearn import datasets
In [4]:	<pre>iris=datasets.load_iris()</pre>
In [5]:	iris

```
Out[5]:
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```
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     [5.9, 3., 5.1, 1.8]]),
  1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
     'frame': None,
'target names': array(['setosa', 'versicolor', 'virginica'], dtype='<U10'),
'DESCR': '.. _iris_dataset:\n\nIris plants dataset\n -------\n\n**Data
Set Characteristics:**\n\n :Number of Instances: 150 (50 in each of three classe s)\n
```

1

2

4.9

4.7

3.0

3.2

1.4

1.3

```
:Number of Attributes: 4 numeric, predictive attributes and the class\n :Attribute
        Information:\n - sepal length in cm\n - sepal width in cm
                 - petal length in cm\n

    petal width in cm\n

                                                                          - class:\n
        - Iris-Setosa\n
                                     - Iris-Versicolour\n
                                                                        - Iris-Virginica
                              :Summary Statistics:\n\n
                                                          -----
                         \n
        = ==== =======\n Min Max Mean SD Class Cor relation\n =========
        0.7826\n sepal width: 2.0 4.4 3.05 0.43 -0.4194\n petal length: 1.0 6.9 3.76 1.76
        0.9490 (high!)\n petal width: 0.1 2.5 1.20 0.76 0.9565 (high!)\n ===============
        ==== ============\n\n :Missing Attribute Values: None\n :Class
        Distribution: 33.3% for each of 3 classes.\n :Creator: R.A. Fisher\n
       Michael Marshall (MARSHALL%PLU@io.arc.nasa.gov)\n :Date: July, 1988\n\nTh e famous
        Iris database, first used by Sir R.A. Fisher. The dataset is taken\nfrom F isher\'s
        paper. Note that it\'s the same as in R, but not as in the UCI\nMachine Le arning
        Repository, which has two wrong data points.\n\nThis is perhaps the best kno wn
        database to be found in the\npattern recognition literature. Fisher\'s paper is a
        classic in the field and\nis referenced frequently to this day. (See Duda & Hart ,
        for example.) The \ndata set contains 3 classes of 50 instances each, where each class
        refers to a\ntype of iris plant. One class is linearly separable from the ot her 2;
        the\nlatter are NOT linearly separable from each other.\n\n.. topic:: Refere nces\n\n
        - Fisher, R.A. "The use of multiple measurements in taxonomic problem s"\n Annual
        Eugenics, 7, Part II, 179-188 (1936); also in "Contributions to\n Mathematical
        Statistics" (John Wiley, NY, 1950).\n - Duda, R.O., & Hart, P.E. (19 73) Pattern
        Classification and Scene Analysis.\n (Q327.D83) John Wiley & Sons. ISBN 0-471-22361-
        1. See page 218.\n - Dasarathy, B.V. (1980) "Nosing Around the Neighborhood: A New
        System\n Structure and Classification Rule for Recognition in Partially Exposed\n
        Environments". IEEE Transactions on Pattern Analysis and Machine\n Intelligence,
        Vol. PAMI-2, No. 1, 67-71.\n - Gates, G.W. (1972) " The Reduced Nearest Neighbor
        Rule". IEEE Transactions\n on Information Theory,
        May 1972, 431-433.\n - See also: 1988 MLC Proceedings, 54-64. Cheeseman et al"s
        AUTOCLASS II\n conceptual clustering system finds 3 classes in the data.\n
        Many, many more ...',
         'feature names': ['sepal length (cm)',
          'sepal width (cm)',
          'petal length (cm)',
          'petal width (cm)'],
         'filename': 'iris.csv',
         'data module': 'sklearn.datasets.data'}
In [6]: df=pd.DataFrame(iris['data'],columns=iris['feature names'])
Ιn
       df
[7]:
Out[7]:
             sepal length (cm) sepal width (cm)
                                          petal length (cm) petal width (cm)
          0
                       5.1
                             3.5
                                    1.4
                                          0.2
```

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0.2

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	× 4 co	olumns			
[8]:	df.descri	be()					
[8]:		epal length (cm)	sepa	l width (cm)	peta	l length (cm) p	etal width (cm)
	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
10]:	df['speci	es']=iris['tar	rget']			
15]: [16]:							
[16]: pal	df						
(cm)	(cm) petal le	ength (cm) petal v	width	(cm)	S	pecies	
	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	
	145	6.7	3.0	5.2	2.3 Virg	inica	
	146	6.3	2.5	5.0	1.9 Virg	inica	
	147	6.5	3.0	5.2	2.0 Virg	inica	
	148	6.2	3.4	5.4	2.3 Virg	inica	
	149	5.9	3.0	5.1	1.8 Virg	inica	
	150	rows	× 5 colu	mns			
In [17]:	df.describe()						

Out[17]:	sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)

	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

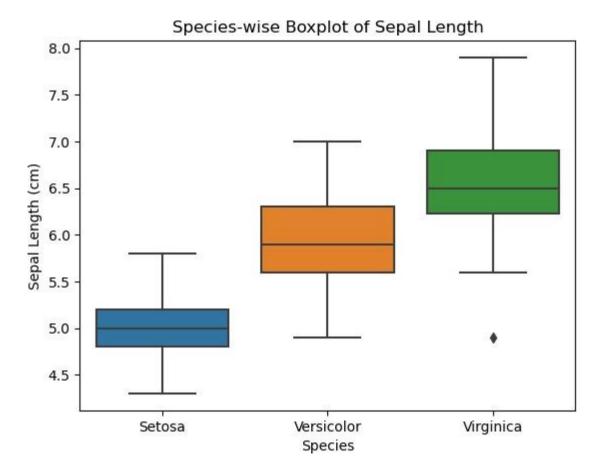
```
In [20]:
In [21]:
setosa_df=df[df['species']=='Setosa']
Out[21]:
sepal
setosa_df.describe()
```

length (cm) sepal width (cm) petal length (cm) petal width (cm)

	count	50.00000	50.000000	50.000000	50.000000
mean	_	5.00600	3.428000	1.462000	0.246000
	std	0.35249	0.379064	0.173664	0.105386
	min	4.30000	2.300000	1.000000	0.100000
	25%	4.80000	3.200000	1.400000	0.200000
	50%	5.00000	3.400000	1.500000	0.200000
	75%	5.20000	3.675000	1.575000	0.300000
	max	5.80000	4.400000	1.900000	0.600000
ı [22]:	versicolo	r_df=df[df['spec	ies']=='Versico]	lor']	
[24]:	versicolo	r_df.describe()			
Out[24]: pal length		vidth (cm) petal leng	gth (cm) petal widt	h (cm)	
	count	50.000000	50.000000	50.000000	50.000000
nean		5.936000	2.770000	4.260000	1.326000
	std	0.516171	0.313798	0.469911	0.197753
	min	4.900000	2.000000	3.000000	1.000000
	25%	5.600000	2.525000	4.000000	1.200000
	50%	5.900000	2.800000	4.350000	1.300000
	75%	6.300000	3.000000	4.600000	1.500000
	max	7.000000	3.400000	5.100000	1.800000
[25]: [26]:	virginica	_df=df[df['speci	es']=='Virginica	a']	
t[26]:	virginica	_df.describe()			
sepal gth (cm)	sepal width	(cm) petal length (cı	m) petal width (cm)		
	count	50.00000	50.000000	50.000000	50.00000
mean		6.58800	2.974000	5.552000	2.02600

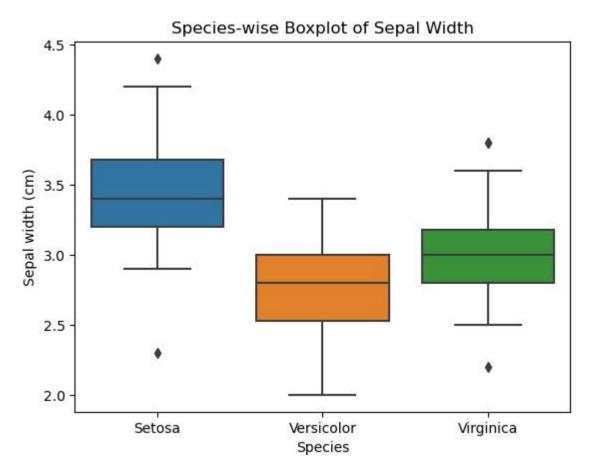
std								
		0.63588	0.322497	0.551895	0.27465			
min		4.90000	2.200000	4.500000	1.40000			
	25%	6.22500	2.800000	5.100000	1.80000			
	50%	6.50000	3.000000	5.550000	2.00000			
	75%	6.90000	3.175000	5.875000	2.30000			
max		7.90000	3.800000	6.900000	2.50000			
In [27]:	<pre>import seaborn as sns import matplotlib.pyplot as plt # Create the boxplot sns.boxplot(x='species', y='sepal length (cm)', data=df) plt.xlabel("Species") plt.ylabel("Sepal Length (cm)") plt.title("Species-wise Boxplot of Sepal Length")</pre>							

```
Out[27]: Text(0.5, 1.0, 'Species-wise Boxplot of Sepal Length')
```



```
In [28]: sns.boxplot(x='species', y='sepal width (cm)', data=df)
    plt.xlabel("Species") plt.ylabel("Sepal width (cm)")
    plt.title("Species-wise Boxplot of Sepal Width")
```

Out[28]: Text(0.5, 1.0, 'Species-wise Boxplot of Sepal Width')



```
In [29]: df.shape
Out[29]: (150, 5)
In [30]: df.mean()
```

C:\Users\Admin\AppData\Local\Temp\ipykernel_7276\3698961737.py:1: FutureWarning: The default value of numeric_only in DataFrame.mean is deprecated. In a future version, it will default to False. In addition, specifying 'numeric_only=None' is deprecated.

Select only valid columns or specify the value of numeric_only to silence this warning.

Out[31]: sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)

		•	
S	pe	cies	

Setosa	5.006	3.428	1.462	0.246
Versicolor	5.936	2.770	4.260	1.326
Virginica	6.588	2.974	5.552	2.026

In [32]: df.median()

C:\Users\Admin\AppData\Local\Temp\ipykernel_7276\530051474.py:1: FutureWarning: The default value of numeric_only in DataFrame.median is deprecated. In a future versio n, it will default to False. In addition, specifying 'numeric_only=None' is depreca ted.

Select only valid columns or specify the value of numeric_only to silence this warning. df.median()

sepal length (cm) 5.80

Out[32]: sepal width (cm) 3.00 petal length (cm) 4.35 petal width (cm) 1.30

dtype: float64

species

df.groupby('species').median()

In [33]:

Out[33]:

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)

Setosa	5.0	3.4	1.50	0.2
Versicolor	5.9	2.8	4.35	1.3
Virginica	6.5	3.0	5.55	2.0

In [34]: df.groupby('species').count()

Out[34]: sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)

species

Setosa	50	50	50	50
Versicolor	50	50	50	50

Virginica 50 50

In [35]: df.species.mode()

Out[35]: 0 Setosa

1 Versicolor2 Virginica

Name: species, dtype: object

```
In [38]: df.quantile(0.25)
```

C:\Users\Admin\AppData\Local\Temp\ipykernel_7276\3656653379.py:1: FutureWarning: The default value of numeric_only in DataFrame.quantile is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of

50

50

```
numeric_only to silence this warning.
df.quantile(0.25)
```

```
Out[38]: sepal length (cm) 5.1 sepal width (cm) 2.8 petal length (cm) 1.6 petal width (cm) 0.3 Name: 0.25, dtype: float64
```

```
In [39]: df.quantile(0.5)
```

C:\Users\Admin\AppData\Local\Temp\ipykernel_7276\1793688606.py:1: FutureWarning: The default value of numeric_only in DataFrame.quantile is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of

```
numeric_only to silence this warning.
df.quantile(0.5)
```

```
Out[39]: sepal length (cm) 5.80 sepal width (cm) 3.00 petal length (cm) 4.35 petal width (cm) 1.30
```

Name: 0.5, dtype: float64

```
In [40]: df.std()
```

C:\Users\Admin\AppData\Local\Temp\ipykernel_7276\3390915376.py:1: FutureWarning: The default value of numeric_only in DataFrame.std is deprecated. In a future version

, it will default to False. In addition, specifying 'numeric_only=None' is deprecat ed. Select only valid columns or specify the value of numeric_only to silence this warning.

```
df.std()
```

prac3-B http://localhost:8888/nbconvert/html/Desktop/aashrya/prac3-B.ipynb?d...