

Iris Flowers Classification ML Project :

Beginner Task -1 The iris flowers dataset contains numeric attributes, and it is perfect for beginners to learn about supervised ML algorithms, mainly how to load and handle data. Also, since this is a small dataset, it can easily fit in memory without requiring special transformations or scaling capabilities.

Datasetlink : <http://archive.ics.uci.edu/ml/datasets/Iris>

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Required Libraries:

```
import numpy as np
import pandas as pd
import matplotlib as plt
from sklearn.model_selection import train_test_split
from sklearn import tree
from sklearn.cluster import KMeans
from sklearn import metrics
```

Reading Dataset:

```
header=["sepal length","sepal width","petal length","petal
width","class"]
data=pd.read_csv("http://archive.ics.uci.edu/ml/machine-learning-
databases/iris/iris.data",header= None,names = header)
print(data)
```

	sepal length	sepal width	petal length	petal width	
class					
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					
..	
...					
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
```

```
[150 rows x 5 columns]
```

```
Exploring Dataset:
```

```
data.index
```

```
RangeIndex(start=0, stop=150, step=1)
```

```
data.columns
```

```
Index(['sepal length', 'sepal width', 'petal length', 'petal width',  
'class'], dtype='object')
```

```
data.size
```

```
750
```

```
data.shape
```

```
(150, 5)
```

```
data.head(10)
```

	sepal length	sepal width	petal length	petal width	class
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
5	5.4	3.9	1.7	0.4	Iris-setosa
6	4.6	3.4	1.4	0.3	Iris-setosa
7	5.0	3.4	1.5	0.2	Iris-setosa
8	4.4	2.9	1.4	0.2	Iris-setosa
9	4.9	3.1	1.5	0.1	Iris-setosa

```
data.tail(10)
```

	sepal length	sepal width	petal length	petal width	class
140	6.7	3.1	5.6	2.4	Iris-
virginica					
141	6.9	3.1	5.1	2.3	Iris-
virginica					
142	5.8	2.7	5.1	1.9	Iris-
virginica					
143	6.8	3.2	5.9	2.3	Iris-
virginica					
144	6.7	3.3	5.7	2.5	Iris-
virginica					
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

```

```
data.info
```

```

<bound method DataFrame.info of
length  petal  width      class      sepal length  sepal width  petal
0      5.1      3.5      Iris-
setosa
1      4.9      3.0      Iris-
setosa
2      4.7      3.2      Iris-
setosa
3      4.6      3.1      Iris-
setosa
4      5.0      3.6      Iris-
setosa
..      ...      ...      ...
...
145     6.7      3.0      Iris-
virginica
146     6.3      2.5      Iris-
virginica
147     6.5      3.0      Iris-
virginica
148     6.2      3.4      Iris-
virginica
149     5.9      3.0      Iris-
virginica

```

```
[150 rows x 5 columns]>
```

```
data.describe()
```

```

count      sepal length  sepal width  petal length  petal width
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

```

Checking the Null values :

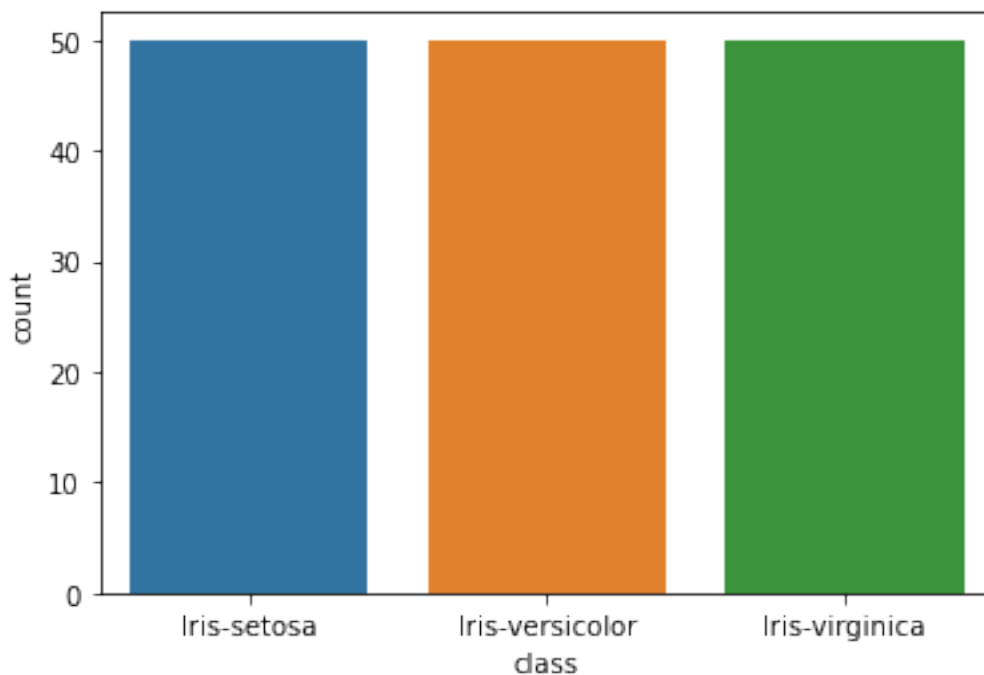
```
data.isnull().sum()
```

```
sepal length    0
sepal width     0
petal length    0
petal width     0
class           0
dtype: int64
```

Visualization of dataset :

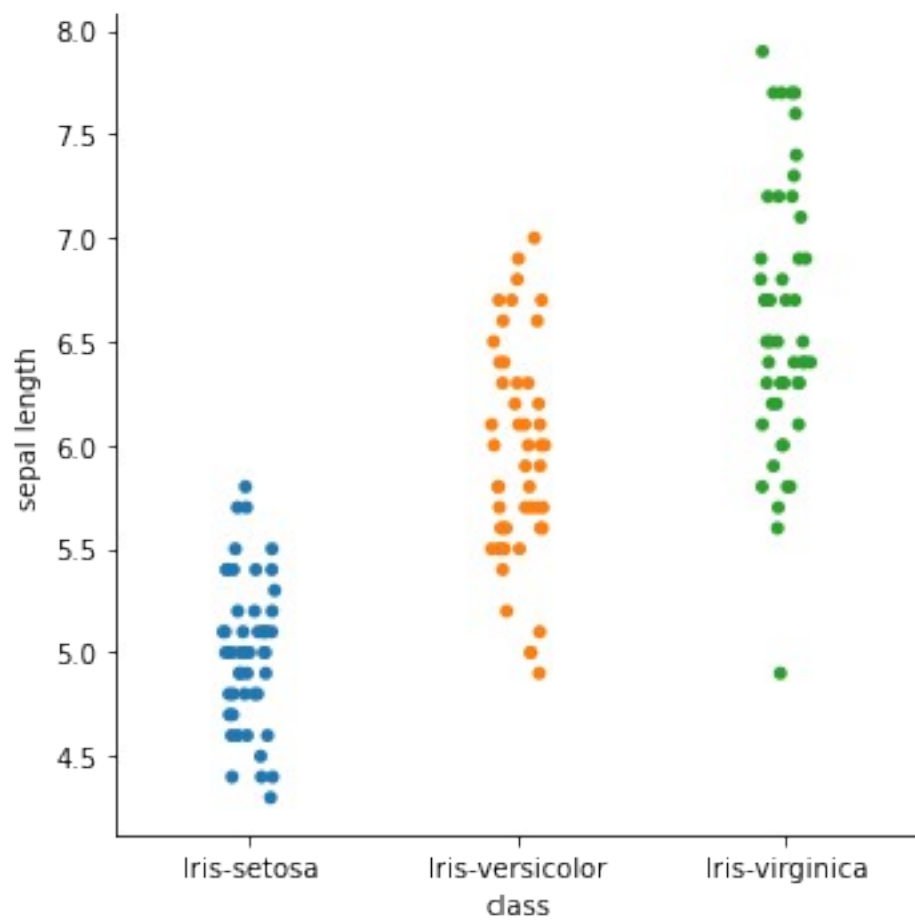
```
import matplotlib.pyplot as plt
import seaborn as sns
sns.countplot(data['class'],label="Size")
plt.show()
```

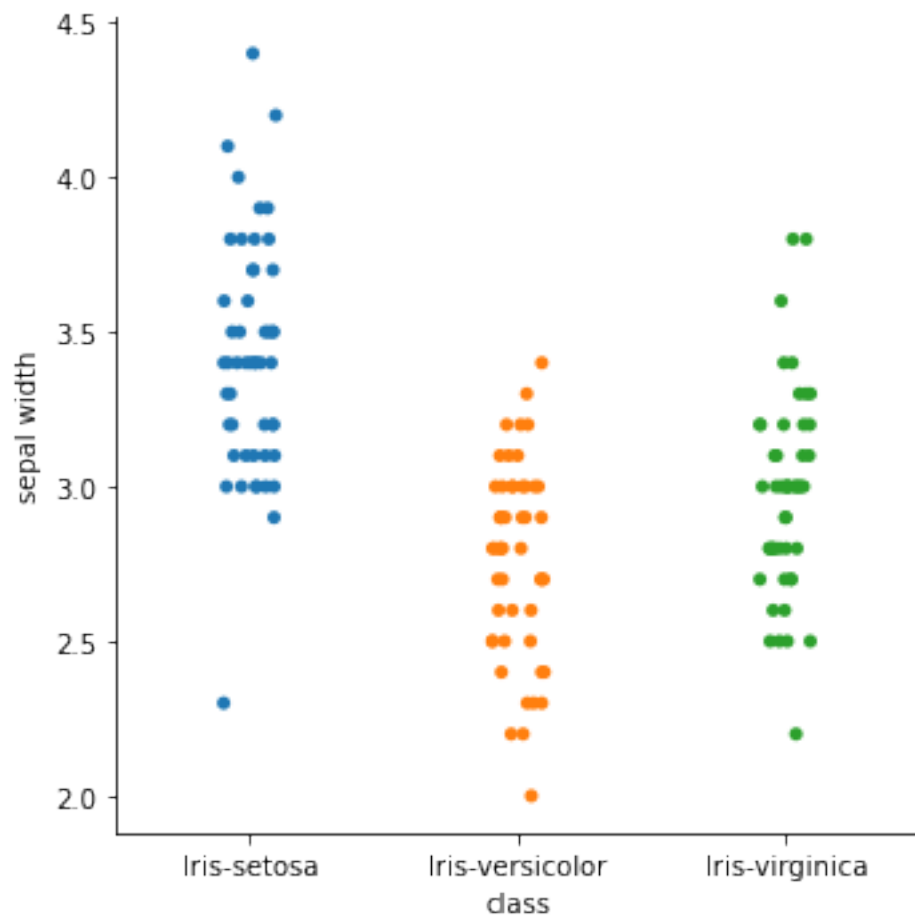
```
C:\Users\Sharayu\OneDrive\jupyter\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```

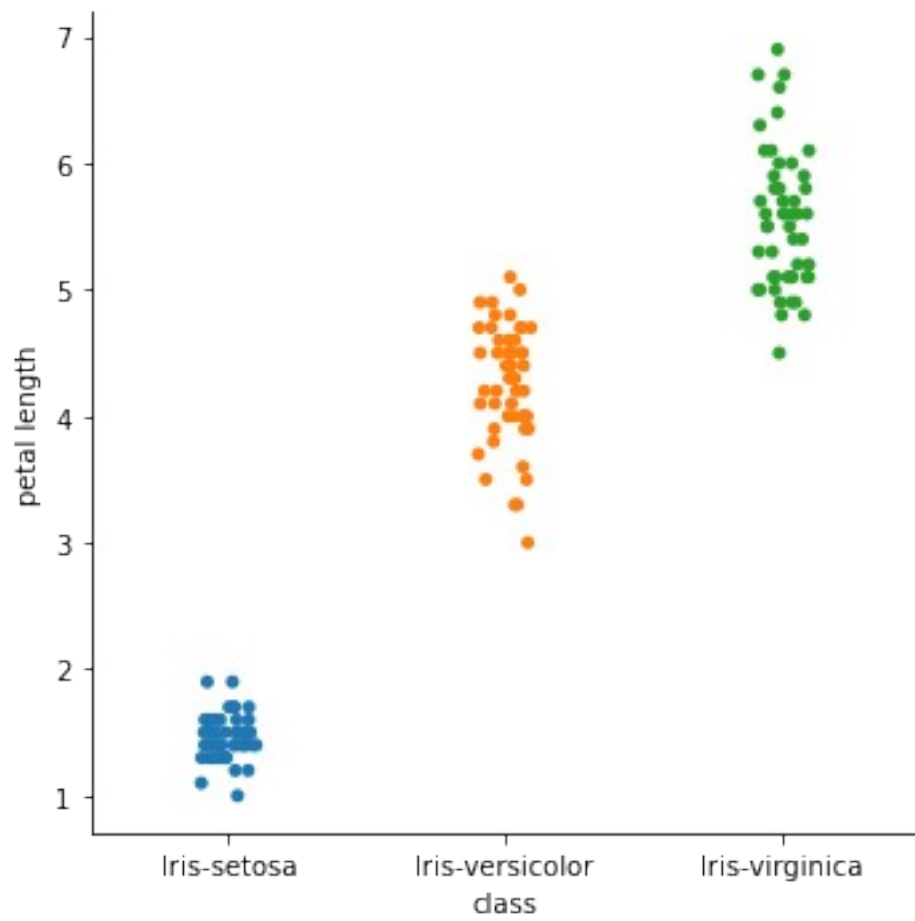


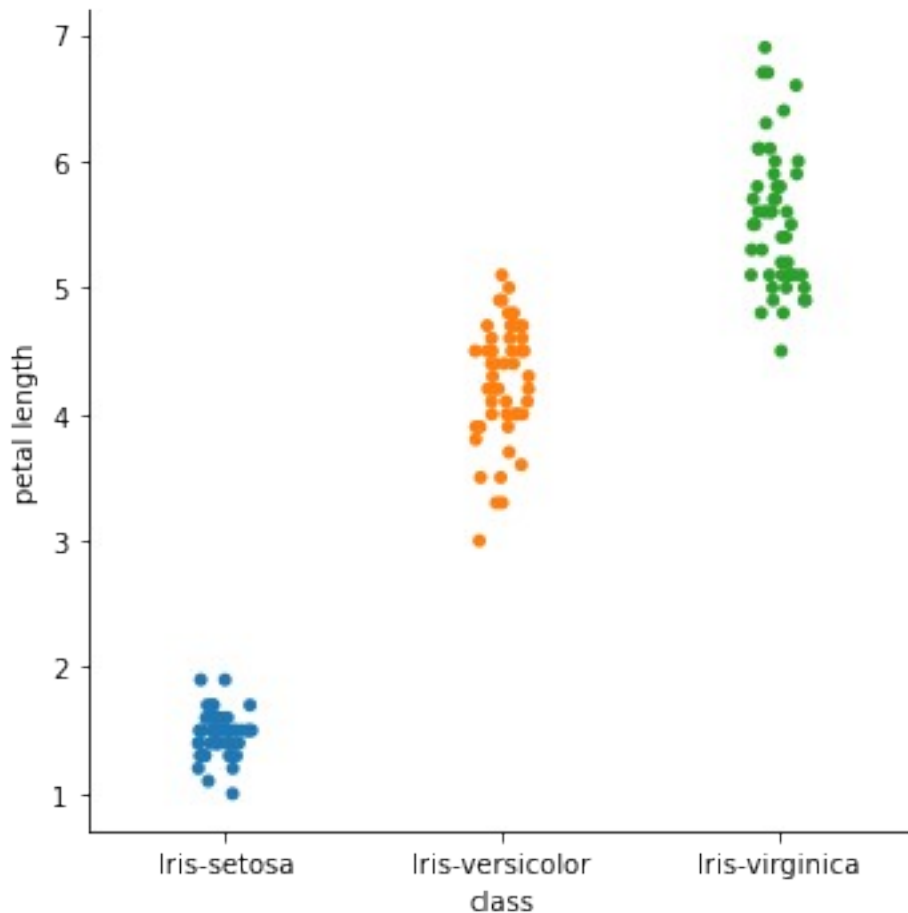
```
sns.catplot(x='class',y='sepal length',data=data)
sns.catplot(x='class',y='sepal width',data=data)
sns.catplot(x='class',y='petal length',data=data)
sns.catplot(x='class',y='petal length',data=data)
```

<seaborn.axisgrid.FacetGrid at 0x21d4b081850>









Knowing the Number of optimum Cluster :

```
from sklearn.cluster import KMeans
ic=[]
for i in range(1,10):
    kmeans =KMeans(n_jobs=-1,n_clusters = i,init = 'k-means++')
    kmeans.fit(data.iloc[:,[0,1,2,3]])
    ic.append(kmeans.inertia_)
```

C:\Users\Sharayu\OneDrive\jupyter\lib\site-packages\sklearn\cluster_kmeans.py:792: FutureWarning: 'n_jobs' was deprecated in version 0.23 and will be removed in 1.0 (renaming of 0.25).

warnings.warn("'n_jobs' was deprecated in version 0.23 and will be"
C:\Users\Sharayu\OneDrive\jupyter\lib\site-packages\sklearn\cluster_kmeans.py:881: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(
C:\Users\Sharayu\OneDrive\jupyter\lib\site-packages\sklearn\cluster_kmeans.py:792: FutureWarning: 'n_jobs' was deprecated in version 0.23 and will be removed in 1.0 (renaming of 0.25).


```
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and will be removed in 1.0 (renaming of 0.25).
```

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```

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_kmeans.py:792: FutureWarning: 'n_jobs' was deprecated in version 0.23
and will be removed in 1.0 (renaming of 0.25).
```

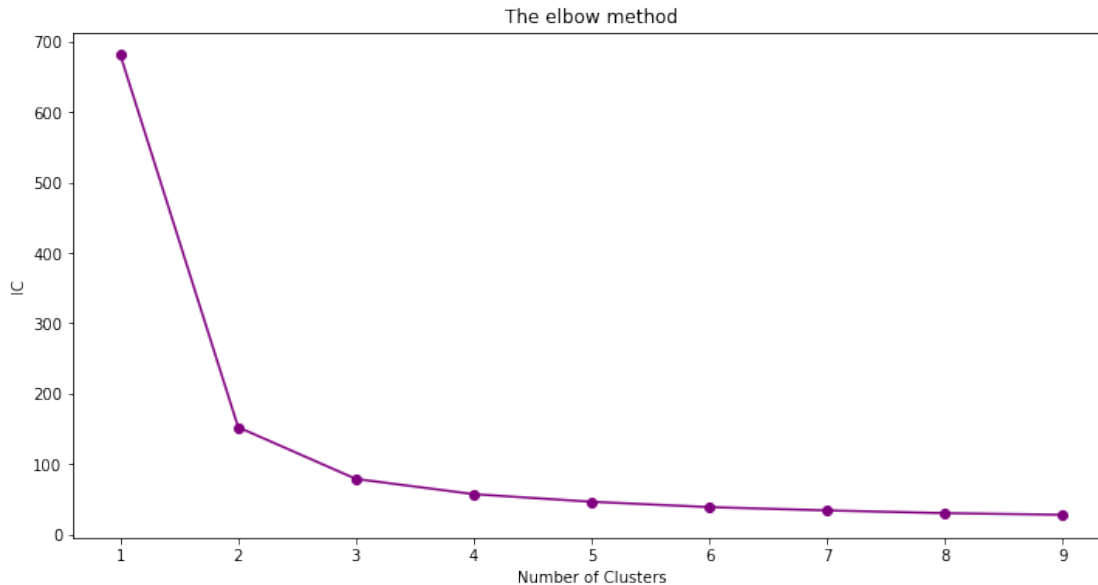
```
warnings.warn("'n_jobs' was deprecated in version 0.23 and will be"
C:\Users\Sharayu\OneDrive\jupyter\lib\site-packages\sklearn\cluster\
_kmeans.py:792: FutureWarning: 'n_jobs' was deprecated in version 0.23
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```

```
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_kmeans.py:792: FutureWarning: 'n_jobs' was deprecated in version 0.23
and will be removed in 1.0 (renaming of 0.25).
```

```
warnings.warn("'n_jobs' was deprecated in version 0.23 and will be"
```

```
df=pd.DataFrame({'Cluster':range(1,10),'ic':ic})
plt.figure(figsize=(12,6))
plt.plot(df['Cluster'],df['ic'],marker='o',color='purple')
plt.xlabel('Number of Clusters')
plt.ylabel('IC')
plt.title('The elbow method')
plt.show()
```

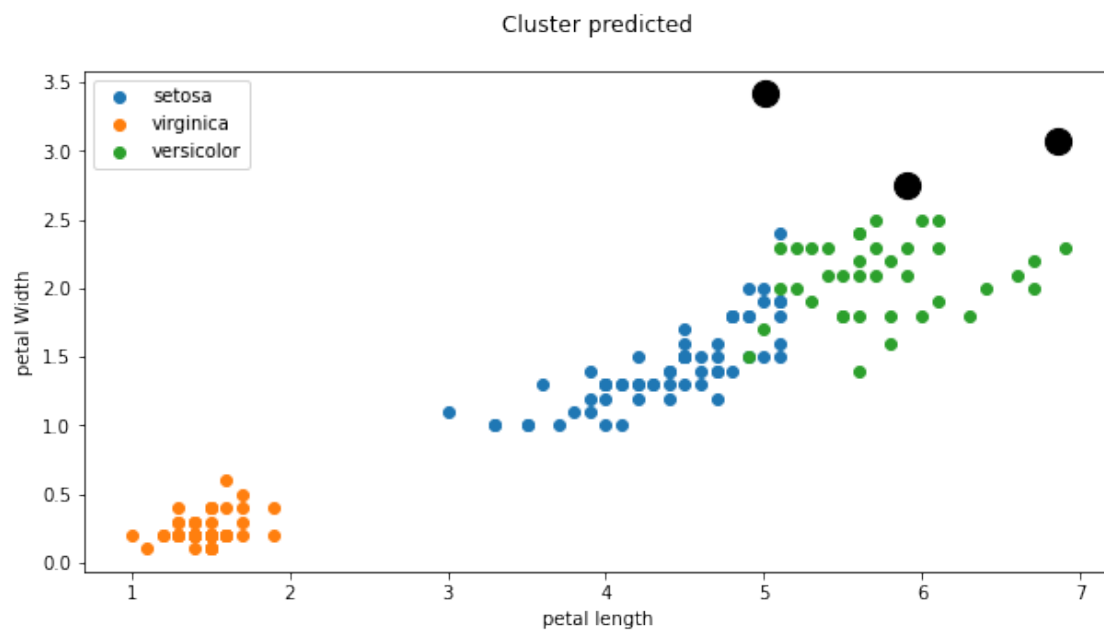
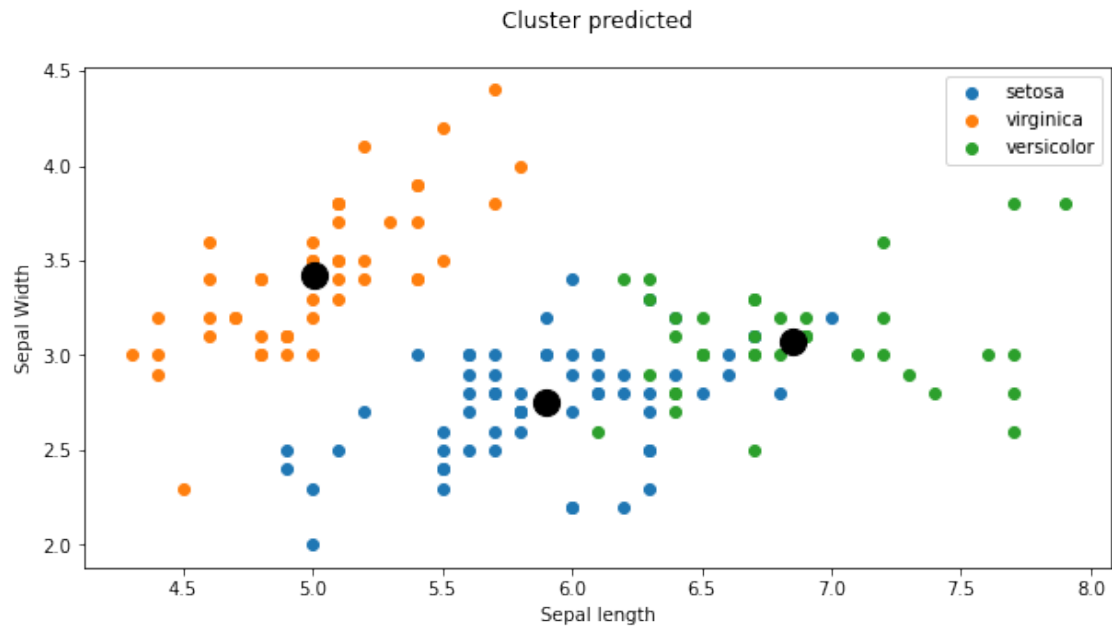


Visualizing Centroids on the Clusters of columns sepalLength and sepalWidth :

```
kmeans=KMeans(n_clusters=3, init='k-means++')
y_kmeans=kmeans.fit_predict(data.iloc[:,[0,1,2,3]].values)

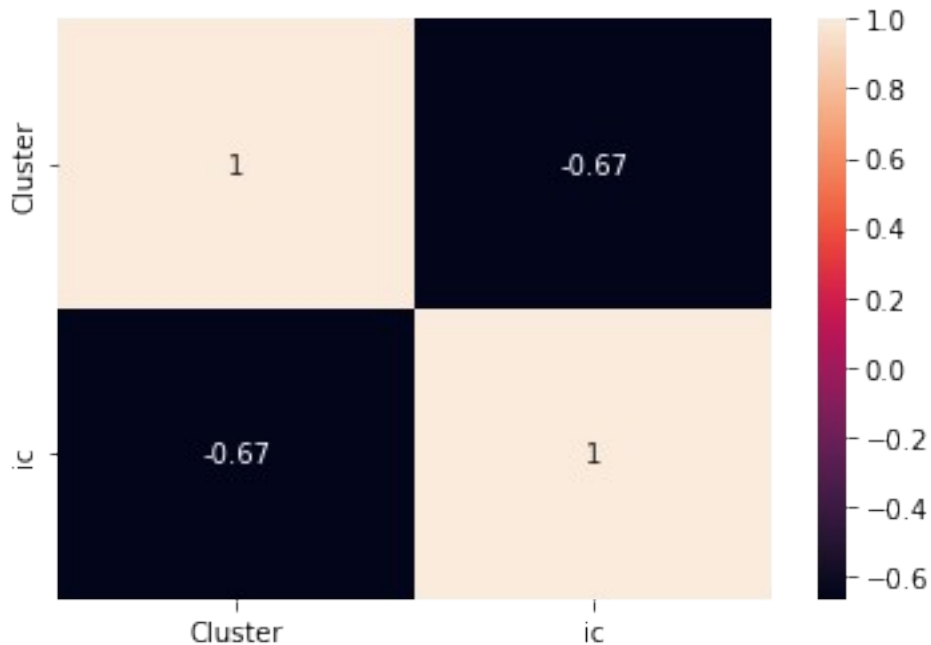
x= data.iloc[:,[0,1,2,3]].values
plt.figure(figsize=(10,5))
plt.scatter(x[y_kmeans==0,0],x[y_kmeans==0,1],label='setosa')
plt.scatter(x[y_kmeans==1,0],x[y_kmeans==1,1],label='virginica')
plt.scatter(x[y_kmeans==2,0],x[y_kmeans==2,1],label='versicolor')
plt.legend()
plt.title('Cluster predicted\n')
plt.xlabel('Sepal length')
plt.ylabel('Sepal Width')
plt.scatter(kmeans.cluster_centers_[0],kmeans.cluster_centers_[1],
s=200,c='black',label='Centroids')
plt.show()

plt.figure(figsize=(10,5))
plt.scatter(x[y_kmeans==0,2],x[y_kmeans==0,3],label='setosa')
plt.scatter(x[y_kmeans==1,2],x[y_kmeans==1,3],label='virginica')
plt.scatter(x[y_kmeans==2,2],x[y_kmeans==2,3],label='versicolor')
plt.legend()
plt.title('Cluster predicted\n')
plt.xlabel('petal length')
plt.ylabel('petal Width')
plt.scatter(kmeans.cluster_centers_[0],kmeans.cluster_centers_[1],
s=200,c='black',label='Centroids')
plt.show()
```



Correlation Matrix :

```
sns.heatmap(df.corr(),annot= True)
plt.show()
```



Gaussians Naive Bayes Accuracy check :

```
x_data=data.reindex(columns=["sepal length","sepal width","petal
length","petal width"])
x_train,x_test,y_train,y_test=train_test_split(x_data,
data["class"],test_size=0.3,random_state=42)

from sklearn.naive_bayes import GaussianNB
gnb = GaussianNB()
gnb.fit(x_train, y_train)
print('Accuracy of Gaussians naive bayes classifiers on training set:
{:.2f}'.format(gnb.score(x_train,y_train)))
print('Accuracy of Gaussians naive bayes classifiers on test set:
{:.2f}'.format(gnb.score(x_test,y_test)))
```

Accuracy of Gaussians naive bayes classifiers on training set: 0.94
Accuracy of Gaussians naive bayes classifiers on test set: 0.98