

IMPORTING THE LIBRARIES

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
from sklearn import datasets
```

Loading the data set and understanding the data

```
iris = datasets.load_iris()
```

```
df = pd.DataFrame(iris.data, columns = iris.feature_names)
```

```
df
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
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]
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 150 entries, 0 to 149
```

```
Data columns (total 4 columns):
```

#	Column	Non-Null Count	Dtype
0	sepal length (cm)	150 non-null	float64
1	sepal width (cm)	150 non-null	float64

```
2    petal length (cm)    150 non-null    float64
3    petal width (cm)     150 non-null    float64
dtypes: float64(4)
memory usage: 4.8 KB
```

Checking for null values

```
df.isnull().sum()
```

```
sepal length (cm)    0
sepal width (cm)     0
petal length (cm)    0
petal width (cm)     0
dtype: int64
```

No null values

Selecting our features: We choose to form clusters based on two features.

1. Petal length

2. Petal width

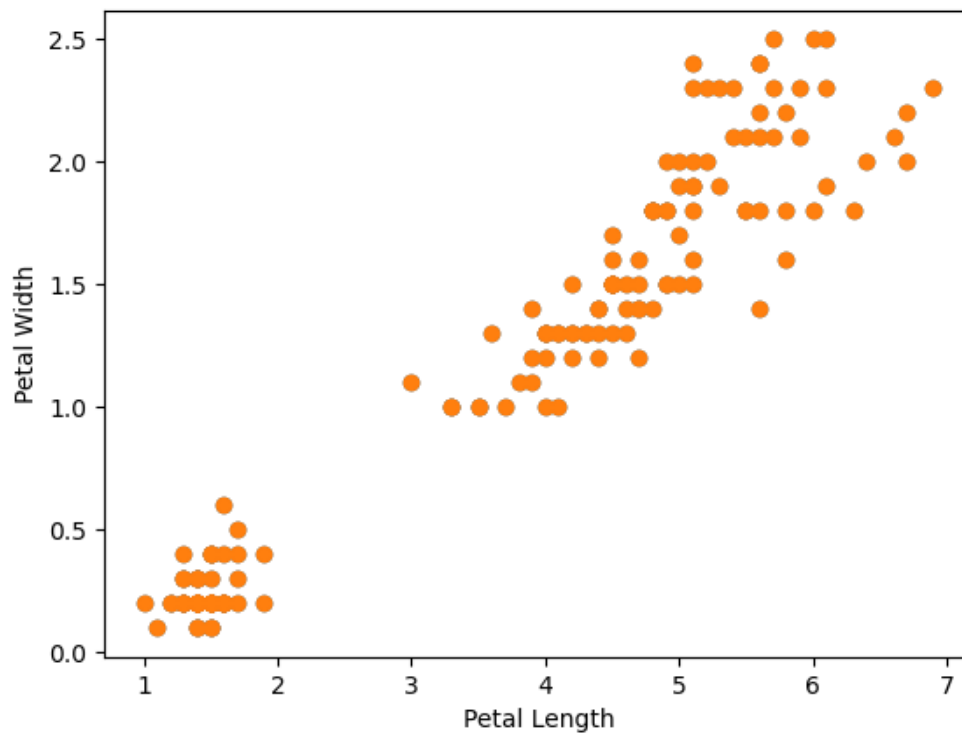
```
X = df.iloc[:,2:4]
```

X

	petal length (cm)	petal width (cm)
0	1.4	0.2
1	1.4	0.2
2	1.3	0.2
3	1.5	0.2
4	1.4	0.2
...
145	5.2	2.3
146	5.0	1.9
147	5.2	2.0
148	5.4	2.3
149	5.1	1.8

[150 rows x 2 columns]

```
plt.scatter(X['petal length (cm)'],X['petal width (cm)'])
plt.xlabel('Petal Length')
plt.ylabel('Petal Width')
plt.show()
```



FINDING NUMBER OF CLUSTERS: ELBOW METHOD

```
from sklearn.cluster import KMeans
```

```
sse = []
```

```
for i in range(1,11):
    kms = KMeans(n_clusters = i, init = 'random', random_state = 43)
    kms.fit(X)
    sse.append(kms.inertia_)
```

```
sse1 = pd.DataFrame(sse, range(1,11))
```

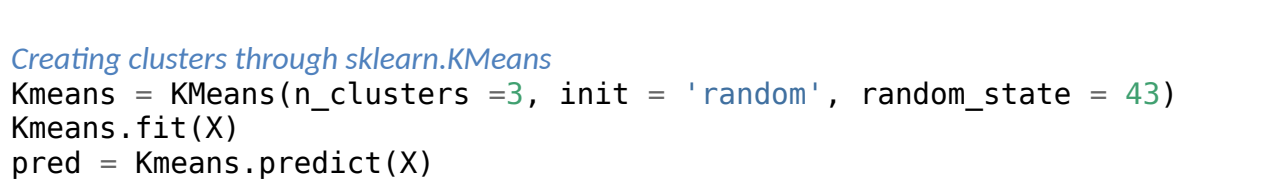
```
sse1
```

```

      0
1  550.895333
2   86.390220
3   31.371359
4   19.483001
5   14.171086
6   11.447138
7    9.355972
8    8.012587
9    6.916963
10   6.483080
```

A line graph showing the relationship between the Number of clusters (X-axis) and the SSE/Inertia/Cluster_Score (Y-axis). The X-axis ranges from 1 to 10, and the Y-axis ranges from 0 to 500. The score starts at approximately 550 for 1 cluster, drops sharply to about 85 for 2 clusters, and then continues to decrease more gradually, reaching approximately 10 for 10 clusters.

Number of clusters	SSE/Inertia/Cluster_Score
1	550
2	85
3	35
4	25
5	20
6	15
7	12
8	10
9	10
10	10



```
Kmeans = KMeans(n_clusters = 3, init = 'random', random_state = 43)
Kmeans.fit(X)
pred = Kmeans.predict(X)
```

pred

```
array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 2, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 2, 2, 2, 2, 2, 0, 2, 2,
       2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2,
```

```
2,
      2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

```
CC = Kmeans.cluster_centers_  
CC
```

```
array([[4.26923077, 1.34230769],  
       [1.462      , 0.246      ],  
       [5.59583333, 2.0375      ]])
```

VISUALIZING THE CLUSTERS

```
pred1 = X[pred == 0]  
pred2 = X[pred == 1]  
pred3 = X[pred == 2]
```

```
plt.scatter(pred1.iloc[:,0],pred1.iloc[:,1],c = 'black', label =  
'cluster1')  
plt.scatter(pred2.iloc[:,0],pred2.iloc[:,1],c = 'pink', label =  
'cluster2')  
plt.scatter(pred3.iloc[:,0],pred3.iloc[:,1],c = 'cyan', label =  
'cluster3')  
plt.xlabel('Petal Length')  
plt.ylabel('Petal Width')  
plt.legend()
```

```
plt.scatter(CC[:,0],CC[:,1],c='red',s = 100, marker = '*')  
plt.show()
```

