

TASK 2:Prediction using Unsupervised ML

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KMeans_Clustering

Understanding the data

```
In [172...
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from math import sqrt
from sklearn import datasets
```

```
In [173...
data = pd.read_csv(r"C:\Users\Rajkumar\Downloads\Iris.csv")
data.shape
```

Out[173... (150, 6)

```
In [174...
data.head(10)
```

Out[174...

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|---|----|---------------|--------------|---------------|--------------|-------------|
| 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |
| 5 | 6 | 5.4 | 3.9 | 1.7 | 0.4 | Iris-setosa |
| 6 | 7 | 4.6 | 3.4 | 1.4 | 0.3 | Iris-setosa |
| 7 | 8 | 5.0 | 3.4 | 1.5 | 0.2 | Iris-setosa |
| 8 | 9 | 4.4 | 2.9 | 1.4 | 0.2 | Iris-setosa |
| 9 | 10 | 4.9 | 3.1 | 1.5 | 0.1 | Iris-setosa |

```
In [175...
data.describe()
```

Out[175...

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm |
|-------|------------|---------------|--------------|---------------|--------------|
| count | 150.000000 | 150.000000 | 150.000000 | 150.000000 | 150.000000 |
| mean | 75.500000 | 5.843333 | 3.054000 | 3.758667 | 1.198667 |
| std | 43.445368 | 0.828066 | 0.433594 | 1.764420 | 0.763161 |
| min | 1.000000 | 4.300000 | 2.000000 | 1.000000 | 0.100000 |
| 25% | 38.250000 | 5.100000 | 2.800000 | 1.600000 | 0.300000 |
| 50% | 75.500000 | 5.800000 | 3.000000 | 4.350000 | 1.300000 |
| 75% | 112.750000 | 6.400000 | 3.300000 | 5.100000 | 1.800000 |
| max | 150.000000 | 7.900000 | 4.400000 | 6.900000 | 2.500000 |

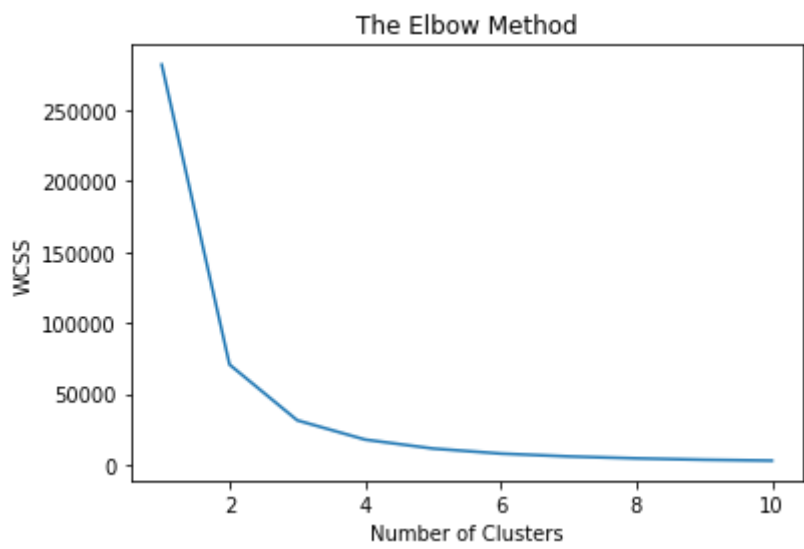
```
In [176...
data
```

Out[176...

| | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species |
|-----|-----|---------------|--------------|---------------|--------------|----------------|
| 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa |
| 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa |
| 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa |
| 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa |
| 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa |
| ... | ... | ... | ... | ... | ... | ... |
| 145 | 146 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica |
| 146 | 147 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica |
| 147 | 148 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica |
| 148 | 149 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica |
| 149 | 150 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica |

150 rows × 6 columns

```
In [177...
x=data.iloc[:,[0,1,2,3]].values
from sklearn.cluster import KMeans
list=[]
for i in range(1,11):
    kmeans=KMeans(n_clusters=i, init='k-means++', max_iter=300, n_init=10, random_state=0)
    kmeans.fit(x)
    list.append(kmeans.inertia_)
plt.plot(range(1,11),list)
plt.title("The Elbow Method")
plt.xlabel("Number of Clusters")
plt.ylabel("WCSS")
plt.show()
```



Creating the k-means classifier

```
In [178...
kmeans=KMeans(n_clusters=3, init='k-means++', max_iter=300, n_init=10, random_state=0)
y_kmeans=kmeans.fit_predict(x)
```

Visualizing the Clusters on first 2 columns

```
In [179...
plt.scatter(x[y_kmeans==0,0], x[y_kmeans==0,1], s=100, c='pink', label='Iris-setosa')
plt.scatter(x[y_kmeans==1,0], x[y_kmeans==1,1], s=100, c='orange', label='Iris-versicolor')
plt.scatter(x[y_kmeans==2,0], x[y_kmeans==2,1], s=100, c='yellow', label='Iris-verginica')

#plotting centriods of clusters
plt.scatter(kmeans.cluster_centers[:,0], kmeans.cluster_centers[:,1], s=100, c='black', label='Ce
plt.legend()
```

Out[179... <matplotlib.legend.Legend at 0x2871fe5ddc0>

