

▼ To Explore Unsupervised Machine Learning:K-Means Clustering

In this task,we will use the iris dataset,to predict the optimum number of cluster

Importing Libraries

```
import numpy as np
import pandas as pd
from sklearn import datasets
import matplotlib.pyplot as plt
```

▼ Loading Iris Data

```
iris = datasets.load_iris()
iris_df = pd.DataFrame(iris.data, columns = iris.feature_names)
iris_df.head(10)
```

	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
5	5.4	3.9	1.7	0.4
6	4.6	3.4	1.4	0.3
7	5.0	3.4	1.5	0.2
8	4.4	2.9	1.4	0.2
9	4.9	3.1	1.5	0.1

```
iris_df.shape
```

```
(150, 4)
```

```
iris_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 4 columns):
#   Column              Non-Null Count  Dtype
#   ...
#   sepal length (cm)    150 non-null    float64
#   sepal width (cm)     150 non-null    float64
#   petal length (cm)    150 non-null    float64
#   petal width (cm)     150 non-null    float64
```

```

---
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

```

```

# Finding the optimum number of clusters for k-means classification
x = iris_df.iloc[:, [0, 1, 2, 3]].values
from sklearn.cluster import KMeans
wcss = []

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)
    wcss.append(kmeans.inertia_)

# Plotting the results onto a line graph,
# `allowing us to observe 'The elbow'
plt.plot(range(1, 11), wcss)
plt.title('The elbow method')
plt.xlabel('Number of clusters')
plt.ylabel('WCSS')
plt.show()

```

```

# Applying kmeans to the dataset / Creating the kmeans classifier
kmeans = KMeans(n_clusters = 3, init = 'k-means++', max_iter = 300, n_init = 10, random_state = 0)
y_kmeans = kmeans.fit_predict(x)

```

```

# Visualising the clusters - On the first two columns
plt.scatter(x[y_kmeans == 0, 0], x[y_kmeans == 0, 1], s = 100, c = 'red', label = 'Iris setosa')
plt.scatter(x[y_kmeans == 1, 0], x[y_kmeans == 1, 1], s = 100, c = 'blue', label = 'Iris versicolour')
plt.scatter(x[y_kmeans == 2, 0], x[y_kmeans == 2, 1], s = 100, c = 'black', label = 'Iris virginica')

# Plotting the centroids of the clusters
plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:, 1], s = 300, c = 'yellow', label = 'Centroids')

```

```
s = 100, c = 'yellow', label = 'Centroids')  
plt.title("K-Means")  
plt.legend()  
plt.show()
```



This concludes the K-Means Workshop.

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