

TASK 1: PREDICTION USING UNSUPERVISED LEARNING

we shall use Iris dataset for the given task # hope you like it

```
In [1]: #Step1.0: Import the Library  
import numpy as np  
import matplotlib.pyplot as plt  
import pandas as pd  
from sklearn import datasets
```

```
In [2]: #Step1.1: Load the Iris dataset  
#make sure to have the dataset in your pc  
  
iris = datasets.load_iris()  
iris_df = pd.DataFrame(iris.data, columns = iris.feature_names)  
iris_df.head() # See the first 5 rows
```

```
Out[2]:
```

	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

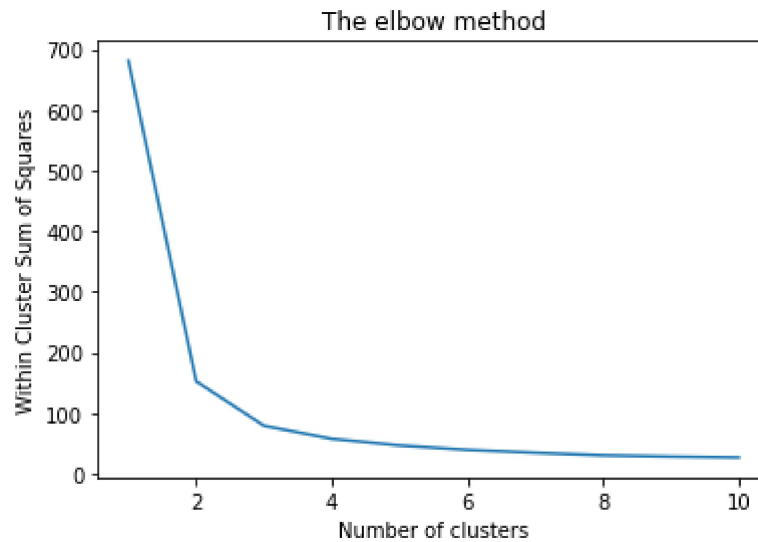
Step 2

Finding the optimum number of clusters using K-means clustering algorithm

```
In [3]: 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_)
```

```
In [4]: #here, we want to observe "The Elbow"
        #Plot results into line graph

        plt.plot(range(1, 11), wcss)
        plt.title('The elbow method')
        plt.xlabel('Number of clusters')
        plt.ylabel('Within Cluster Sum of Squares')
        plt.show()
```

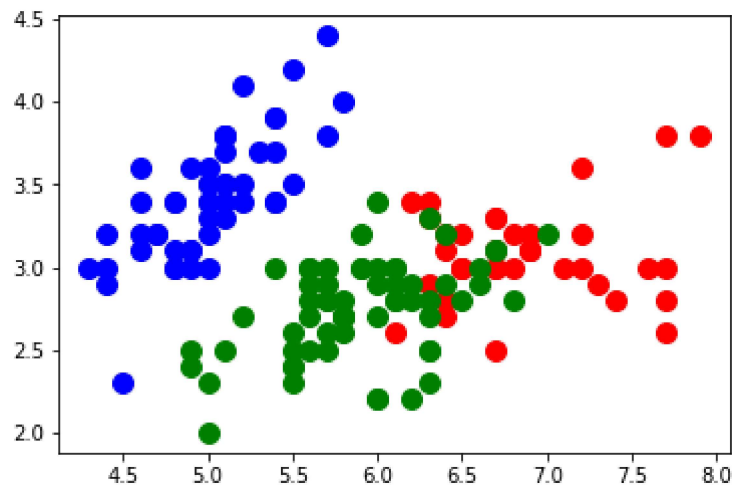


We choose the number of clusters as 3.

```
In [5]: # Applying kmeans to the dataset and 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)
```

```
In [6]: # 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 = 'Setosa')
plt.scatter(x[y_kmeans == 1, 0], x[y_kmeans == 1, 1],
            s = 100, c = 'blue', label = 'Versicolour')
plt.scatter(x[y_kmeans == 2, 0], x[y_kmeans == 2, 1],
            s = 100, c = 'green', label = 'Virginica')
```

Out[6]: <matplotlib.collections.PathCollection at 0x141342d67c0>

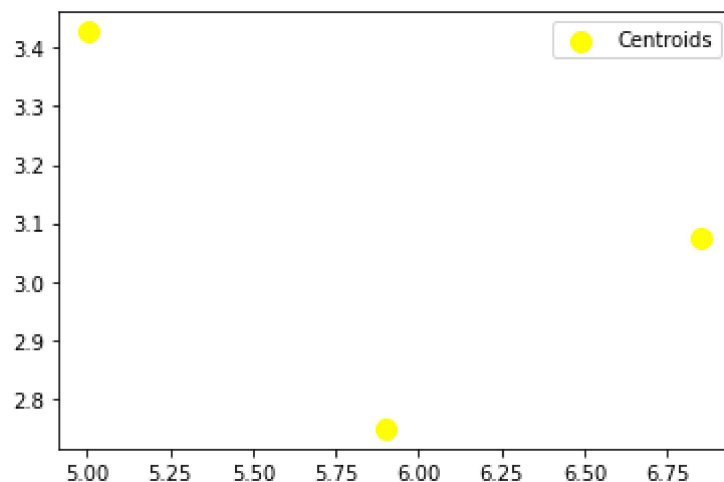


Plotting the centroids of the clusters

```
In [7]: plt.scatter(kmeans.cluster_centers[:, 0], kmeans.cluster_centers[:,1],
                    s = 100, c = 'yellow', label = 'Centroids')

plt.legend()
```

Out[7]: <matplotlib.legend.Legend at 0x141342ae220>

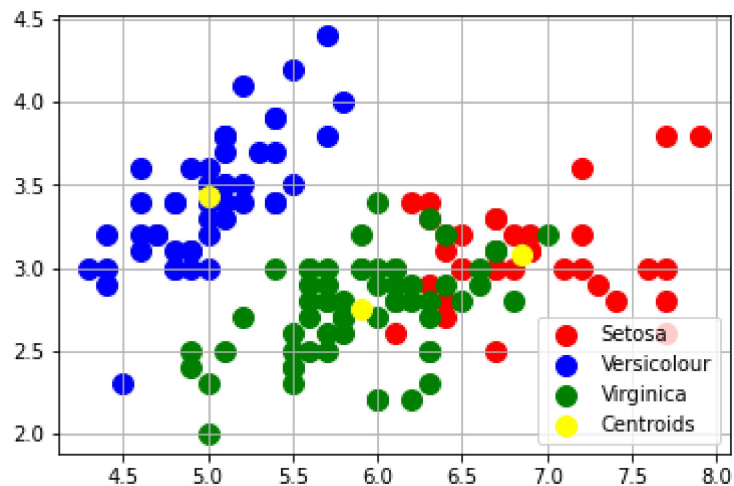


Putting it together

```
In [9]: plt.scatter(x[y_kmeans == 0, 0], x[y_kmeans == 0, 1],
                    s = 100, c = 'red', label = 'Setosa')
plt.scatter(x[y_kmeans == 1, 0], x[y_kmeans == 1, 1],
            s = 100, c = 'blue', label = 'Versicolour')
plt.scatter(x[y_kmeans == 2, 0], x[y_kmeans == 2, 1],
            s = 100, c = 'green', label = 'Virginica')

plt.scatter(kmeans.cluster_centers_[0, 0], kmeans.cluster_centers_[0, 1],
            s = 100, c = 'yellow', label = 'Centroids')

plt.legend()
plt.grid()
```



Finally we got the visual output along with the clusters

Hope you like it

Any problems or issues, just ask

THANK YOU