

```
In [4]: import pandas as pd
        from sklearn.cluster import KMeans
```

```
In [5]: df=pd.read_csv("f:/dataset/fruits_cluster.csv")
        X=df.iloc[:,:].values
```

```
In [6]: model=KMeans(n_clusters=3,max_iter=50,n_init=5)
        model.fit(X)
        print(model.labels_)
```

```
C:\Users\Ducat\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
KMeans is known to have a memory leak on Windows with MKL, when there are less chunks th
an an available threads. You can avoid it by setting the environment variable OMP_NUM_THREA
DS=1.
```

```
warnings.warn(
[0 1 2 2 1 0 1 0 1 1 0 1 0 0 2 2 0 0 2 2 1 1])
```

```
In [7]: model=KMeans(n_clusters=2,max_iter=50,n_init=5)
        model.fit(X)
        print(model.labels_)
```

```
C:\Users\Ducat\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning:
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```

```
warnings.warn(
[0 1 1 1 0 0 1 0 1 0 0 0 0 0 1 1 0 0 1 1 0 0])
```

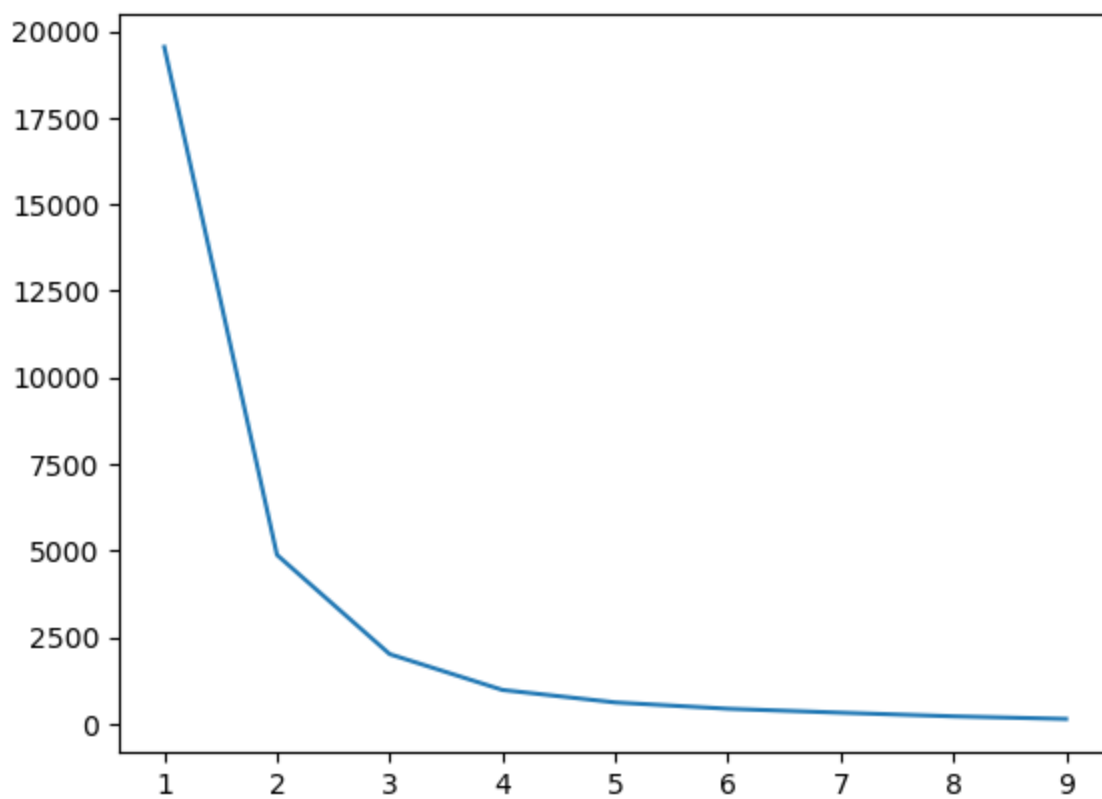
```
In [ ]: #how to select number of clusters in KMeans
        >elbow technique can be used in select num of clusters
        >it is a graphical technique where we plot line chart b/w num_of_clusters & wsse scores
```

```
In [9]: import warnings
        warnings.filterwarnings('ignore')
```

```
In [17]: from sklearn.metrics import silhouette_score,silhouette_samples
```

```
In [10]: wsse=[]
        for i in range(1,10):
            model=KMeans(n_clusters=i,max_iter=50,n_init=5)
            model.fit(X)
            wsse.append(model.inertia_)
```

```
In [12]: import matplotlib.pyplot as plt
        plt.plot(range(1,10),wsse)
        plt.show()
```



```
In [14]: model=KMeans(n_clusters=2,max_iter=50,n_init=5)
model.fit(X)
print(silhouette_score(X,model.labels_))
```

0.6047687920287002

```
In [15]: model=KMeans(n_clusters=3,max_iter=50,n_init=5)
model.fit(X)
print(silhouette_score(X,model.labels_))
```

0.5562767664418119

```
In [16]: model=KMeans(n_clusters=4,max_iter=50,n_init=5)
model.fit(X)
print(silhouette_score(X,model.labels_))
```

0.5952186571525124

```
In [18]: model=KMeans(n_clusters=2,max_iter=50,n_init=5)
model.fit(X)
print(silhouette_samples(X,model.labels_))
```

```
[0.756896  0.08910924 0.7106429  0.73673795 0.37057472 0.68911747
 0.48795214 0.74580381 0.09269296 0.32304302 0.68826806 0.54228985
 0.73426441 0.72721378 0.74553973 0.67113401 0.75763073 0.6883263
 0.66879425 0.74543725 0.66687718 0.66656766]
```

```
In [20]: model.labels_
```

```
Out[20]: array([0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0])
```

```
In [ ]:
```