

```
In [1]: 1 import pandas as pd
2 import numpy as np
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5 %matplotlib inline
```

C:\Users\admin\anaconda3\lib\site-packages\scipy\\_\_init\_\_.py:146: UserWarning: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (detected version 1.23.5  
 warnings.warn(f"A NumPy version >={np\_minversion} and <{np\_maxversion}")

```
In [5]: 1 data = pd.read_csv('Mall_Customers.csv')
2 data.head()
```

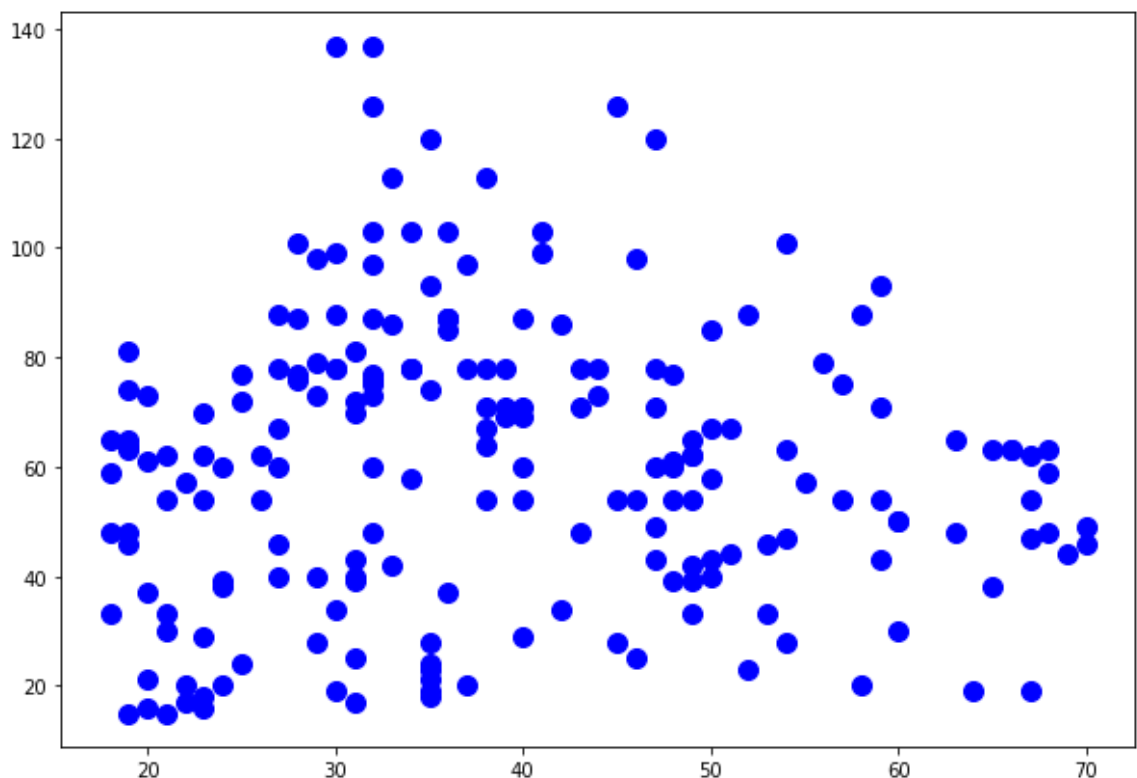
```
Out[5]:
```

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

```
In [4]: 1 datasubset = data.loc[:,["Age", "Annual Income (k$)"]]
```

```
In [6]: 1 #Scatter diagram
2 plt.figure(figsize = (10,7))
3 plt.scatter(datasubset[['Age']], datasubset[['Annual Income (k$)']], s
```

```
Out[6]: <matplotlib.collections.PathCollection at 0x1d6704e7820>
```

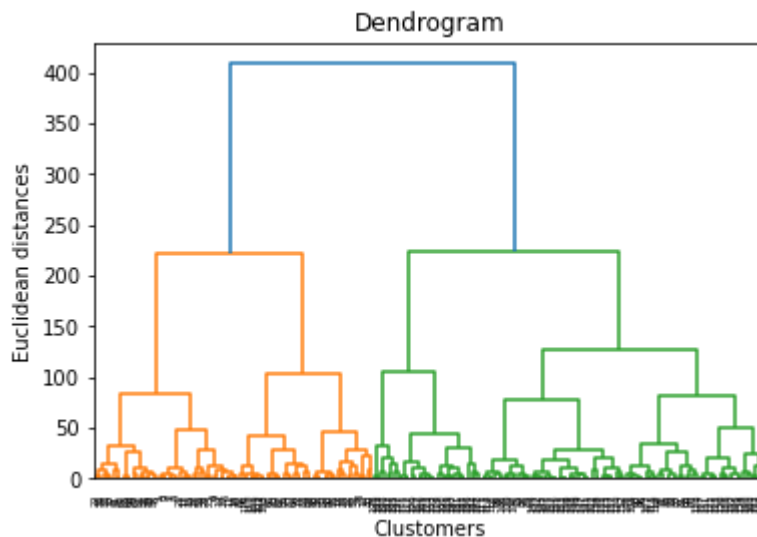


```
In [9]: 1 import scipy.cluster.hierarchy as sch
        2 plt.figure(figsize = (10,7))
```

Out[9]: <Figure size 720x504 with 0 Axes>

<Figure size 720x504 with 0 Axes>

```
In [11]: 1 dendrogram = sch.dendrogram(sch.linkage(datasubset, method = 'ward'))
2 plt.title('Dendrogram')
3 plt.xlabel('Clustomers')
4 plt.ylabel('Euclidean distances')
5 plt.show()
```



```
In [12]: 1 #check for largest distance vertically without crossing any horizontal
2 from sklearn.cluster import AgglomerativeClustering
3 cluster = AgglomerativeClustering(n_clusters = 2, affinity = 'euclidean'
4 cluster.fit_predict(datasubset)
```

```
Out[12]: 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, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,  
                0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0,  
                0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1,  
                1, 0, 0, 0, 0, 0, 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, 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, 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], dtype=int64)
```

```
In [13]: 1 cl = cluster.fit_predict(datasubset)
```

```
In [14]: 1 from sklearn.metrics import silhouette_score
          2 silhouette_score(datasubset,cl)
```

Out[14]: 0.4104652474372429

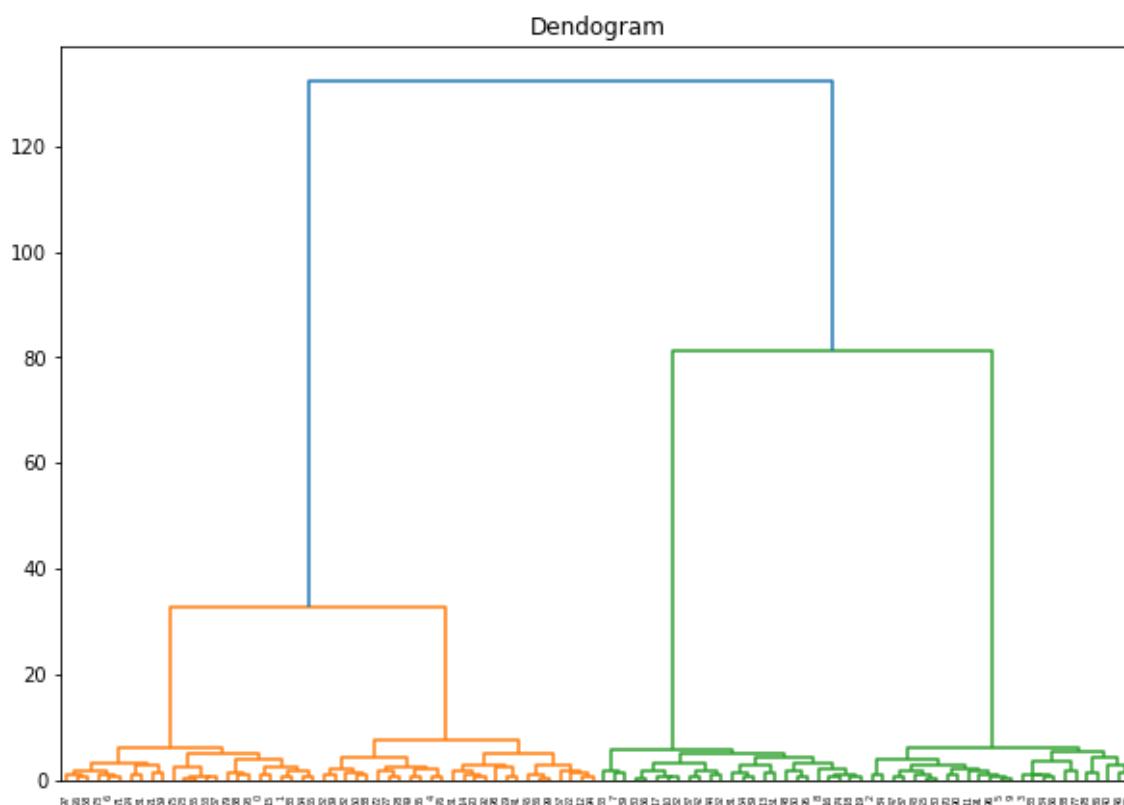
```
In [16]: 1 from sklearn.cluster import AgglomerativeClustering
2 import scipy.cluster.hierarchy as sch
3 from sklearn.datasets import make_blobs
```

```
In [17]: 1 #Generate a random dataset with 100 samples and 3 features
2 x, y = make_blobs(n_samples = 100, centers = 4, n_features = 3, random_
```

```
In [19]: 1 #Perform the dendrogram
2 clustering = AgglomerativeClustering(n_clusters = 3, affinity = 'euclid
3 clustering.fit(x)
```

Out[19]: AgglomerativeClustering(n\_clusters=3)

```
In [21]: 1 #plot the dendrogram
2 plt.figure(figsize = (10,7))
3 plt.title('Dendrogram')
4 dendrogram = sch.dendrogram(sch.linkage(x, method = 'ward'))
5 plt.show()
```



```
In [ ]: 1
```