

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
```

```
In [2]: df = pd.read_csv('mall.csv')
```

```
In [3]: df.shape
```

```
Out[3]: (200, 5)
```

```
In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
 #   Column                Non-Null Count  Dtype  -
--  -
CustomerID            200 non-null    int64
1  Genre                200 non-null    object
2  Age                  200 non-null    int64
3  Annual Income (k$)    200 non-null    int64  4  Spending Score (1-100)
    200 non-null    int64 dtypes: int64(4), object(1) memory usage: 7.9+ KB
```

```
In [9]: X=df.iloc[:,[3,4]].values
```

```
In [11]: X
```

```
Out[11]: array([[ 15,  39],
 [ 15,  81],
 [ 16,   6],
 [ 16,  77],
 [ 17,  40],
 [ 17,  76],
 [ 18,   6],
 [ 18,  94],
 [ 19,   3],
 [ 19,  72],
 [ 19,  14],
 [ 19,  99],
 [ 20,  15],
 [ 20,  77],
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 [ 20,  79],
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 [ 21,  66],
 [ 23,  29],
 [ 23,  98],
 [ 24,  35],
 [ 24,  73],
 [ 25,   5],
 [ 25,  73],
 [ 28,  14],
 [ 28,  82],
 [ 28,  32],
 [ 28,  61],
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 [ 29,  87],
 [ 30,   4],
 [ 30,  73],
 [ 33,   4],
 [ 33,  92],
 [ 33,  14],
 [ 33,  81],
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 [ 34,  73],
 [ 37,  26],
 [ 37,  75],
 [ 38,  35],
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 [ 40,  42],
 [ 42,  52],
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```

```
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```

```
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[ 97, 86],
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[103, 17],
[103, 85],
[103, 23],
[103, 69],
[113, 8],
[113, 91],
[120, 16],
[120, 79],
[126, 28],
[126, 74],
[137, 18],
[137, 83]], dtype=int64)
```

```
In [13]: from sklearn.cluster import KMeans
# wcss : Within Cluster Sum of Squares : sum of squared difference between insta #
# Elbow method for deciding optimum number of clusters, which minimizes value of
wcss=[] # contain wcss value for various number of clusters

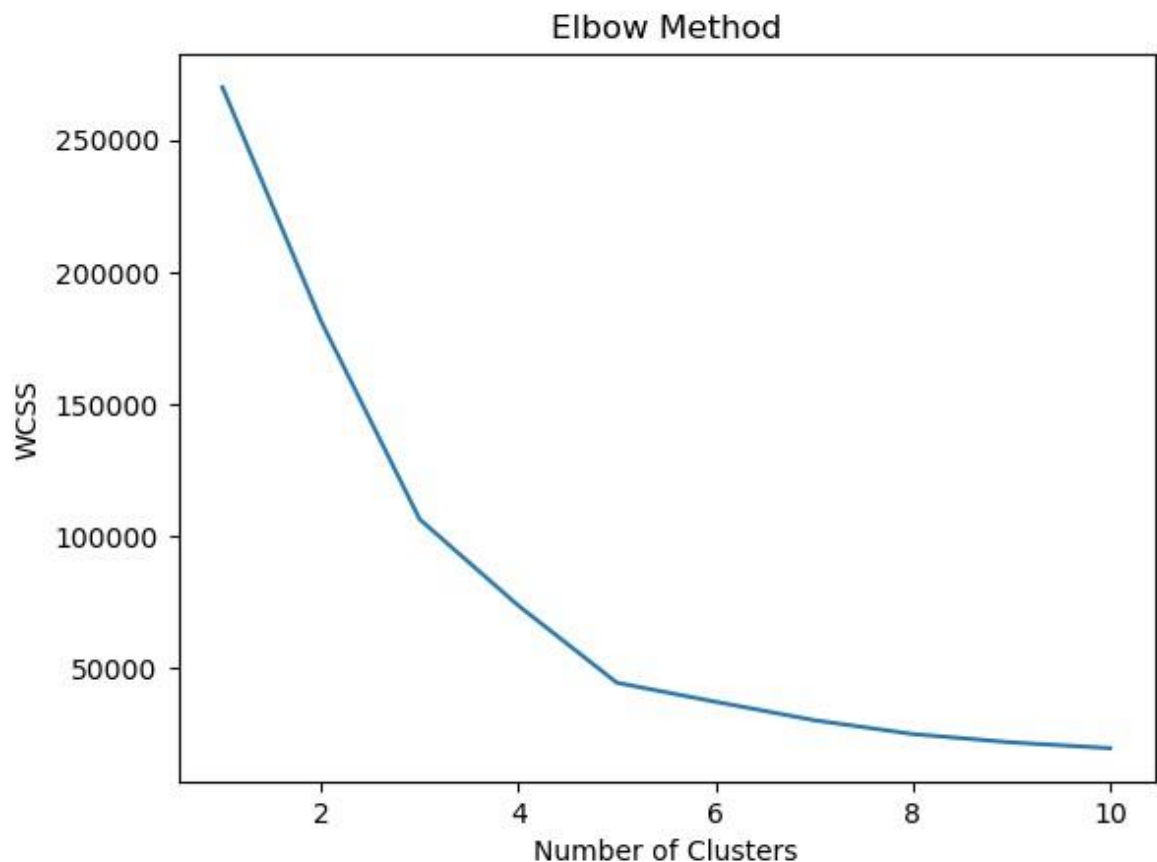
for i in range(1,11):
    kmeans=KMeans(n_clusters=i, init='k-means++',max_iter=300,random_state=42)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)

plt.plot(range(1,11),wcss) plt.title("Elbow
Method") plt.xlabel('Number of Clusters')
plt.ylabel('WCSS')
```

[illegible]

```
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Future
Warning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there are
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Future
Warning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there are
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1. warnings.warn(
```

Out[13]: Text(0, 0.5, 'WCSS')



```
In [14]: kmeans=KMeans(n_clusters=5,init='k-means++',max_iter=300,random_state=42)
y_kmeans=kmeans.fit_predict(X)
```

C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:870: FutureWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1382: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1. warnings.warn(

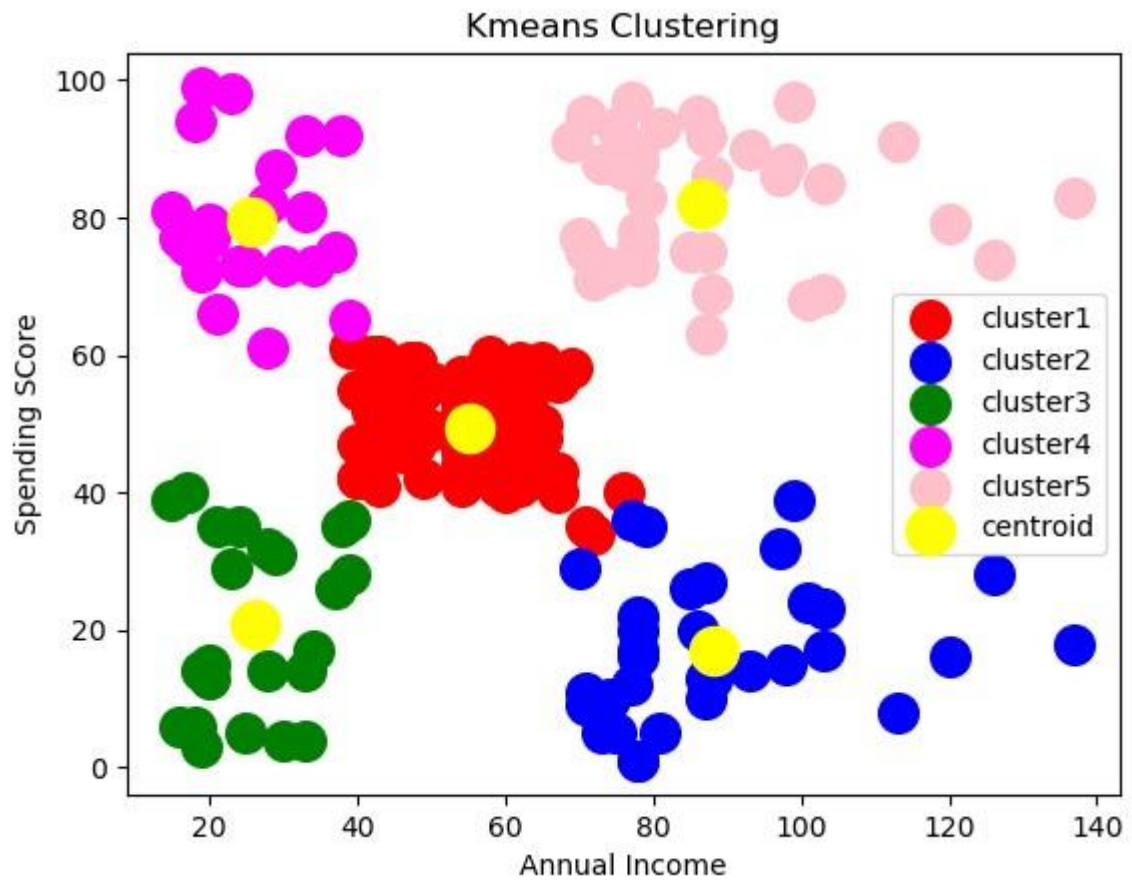
```
In [15]: y_kmeans
```

```
Out[15]: array([2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3, 2, 3,
                2, 3, 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, 4, 1, 4, 0, 4, 1, 4, 1, 4,
                0, 4, 1, 4, 1, 4, 1, 4, 1, 4, 0, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4,
                1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4,
                1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4, 1, 4,
                4])
```

```
In [16]: y_kmeans.shape
```

```
Out[16]: (200,)
```

```
In [17]: plt.scatter(X[y_kmeans==0,0],X[y_kmeans==0,1],s=200,c='red',label='cluster1')
plt.scatter(X[y_kmeans==1,0],X[y_kmeans==1,1],s=200,c='blue',label='cluster2')
plt.scatter(X[y_kmeans==2,0],X[y_kmeans==2,1],s=200,c='green',label='cluster3')
plt.scatter(X[y_kmeans==3,0],X[y_kmeans==3,1],s=200,c='magenta',label='cluster4')
plt.scatter(X[y_kmeans==4,0],X[y_kmeans==4,1],s=200,c='pink',label='cluster5')
plt.scatter(kmeans.cluster_centers_[0,0],kmeans.cluster_centers_[0,1],s=300,c='y')
plt.title("Kmeans Clustering") plt.xlabel('Annual Income') plt.ylabel('Spending
Score') plt.legend() plt.show()
```

```
In [18]: import matplotlib.cm as cm from sklearn.metrics import  
silhouette_samples, silhouette_score import numpy as np
```

```
In [19]:
```

```

range_n_clusters = [2, 3, 4, 5, 6]

h for n_clusters in range_n_clusters:
    # Initialize the clusterer with n_clusters value and a random generator
    # seed of 10 for reproducibility.    clusterer =
    KMeans(n_clusters=n_clusters, random_state=10)
    cluster_labels = clusterer.fit_predict(X)

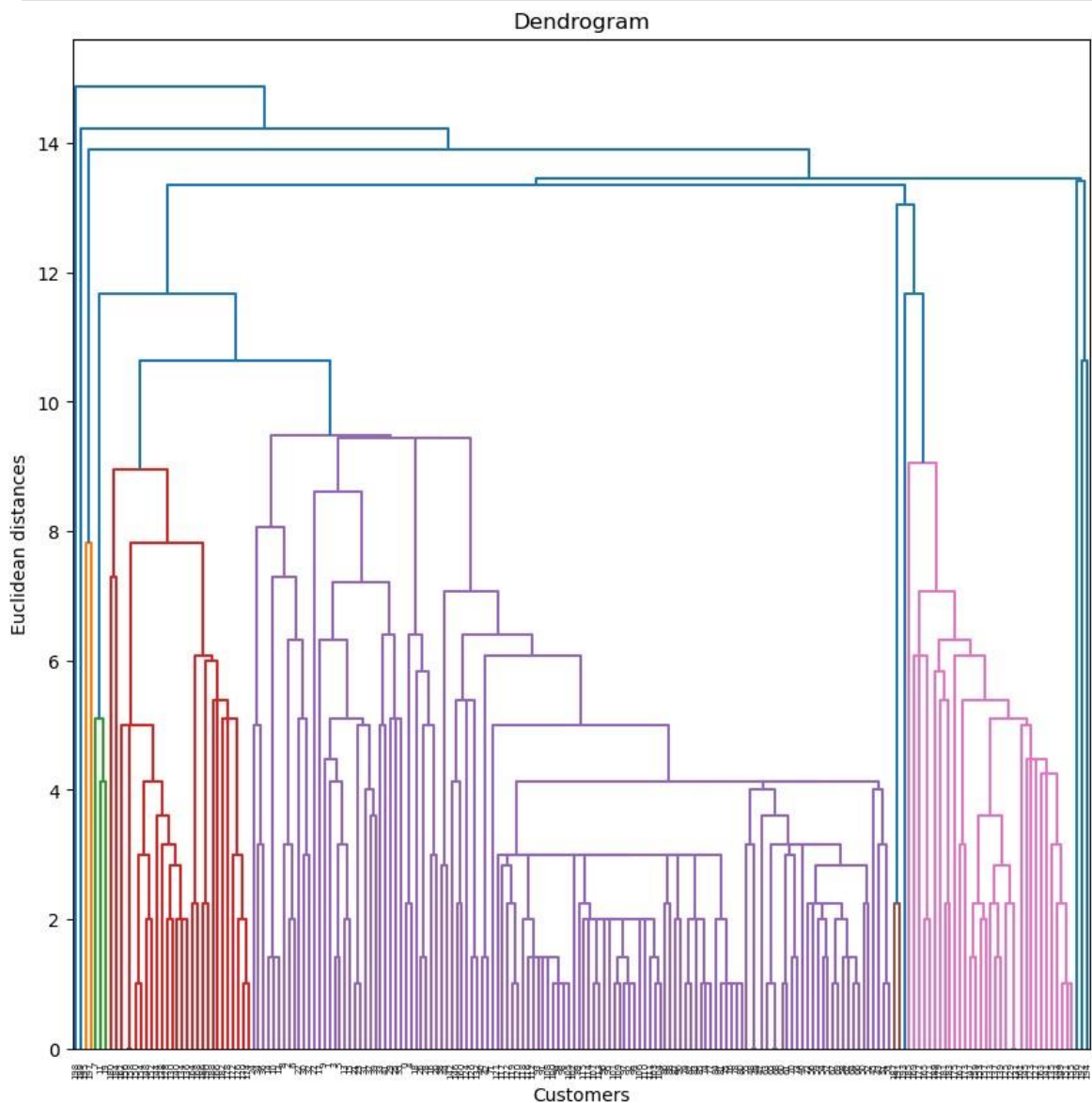
    # The silhouette_score gives the average value for all the samples.
    # This gives a perspective into the density and separation of the formed clu
    silhouette_avg = silhouette_score(X, cluster_labels)    print("For n_clusters
    =",n_clusters,"The average silhouette_score is :",sil

C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Futur
eWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning    warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there ar e
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
    warnings.warn(
For n_clusters = 2 The average silhouette_score is : 0.2968969162503008
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870:
Futur eWarning: The default value of `n_init` will change from 10 to 'auto'
in 1.4. Set the value of `n_init` explicitly to suppress the warning
warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there ar e
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
    warnings.warn(
For n_clusters = 3 The average silhouette_score is : 0.46761358158775435
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Futur
eWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning    warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there ar e
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
    warnings.warn(
For n_clusters = 4 The average silhouette_score is : 0.4931963109249047
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Futur
eWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning    warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there ar e
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
    warnings.warn(
For n_clusters = 5 The average silhouette_score is : 0.553931997444648
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:870: Futur
eWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set
the value of `n_init` explicitly to suppress the warning    warnings.warn(
C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster\_kmeans.py:1382: User
Warning: KMeans is known to have a memory leak on Windows with MKL, when there ar e
less chunks than available threads. You can avoid it by setting the environment
variable OMP_NUM_THREADS=1.
    warnings.warn(

```

For n_clusters = 6 The average silhouette_score is : 0.5376203956398481 In

```
[20]: import scipy.cluster.hierarchy as sch
plt.figure(figsize=(10,10)) dendrogram =
sch.dendrogram(sch.linkage(X, method = 'single'))
plt.title('Dendrogram') plt.xlabel('Customers')
plt.ylabel('Euclidean distances') plt.show()
```



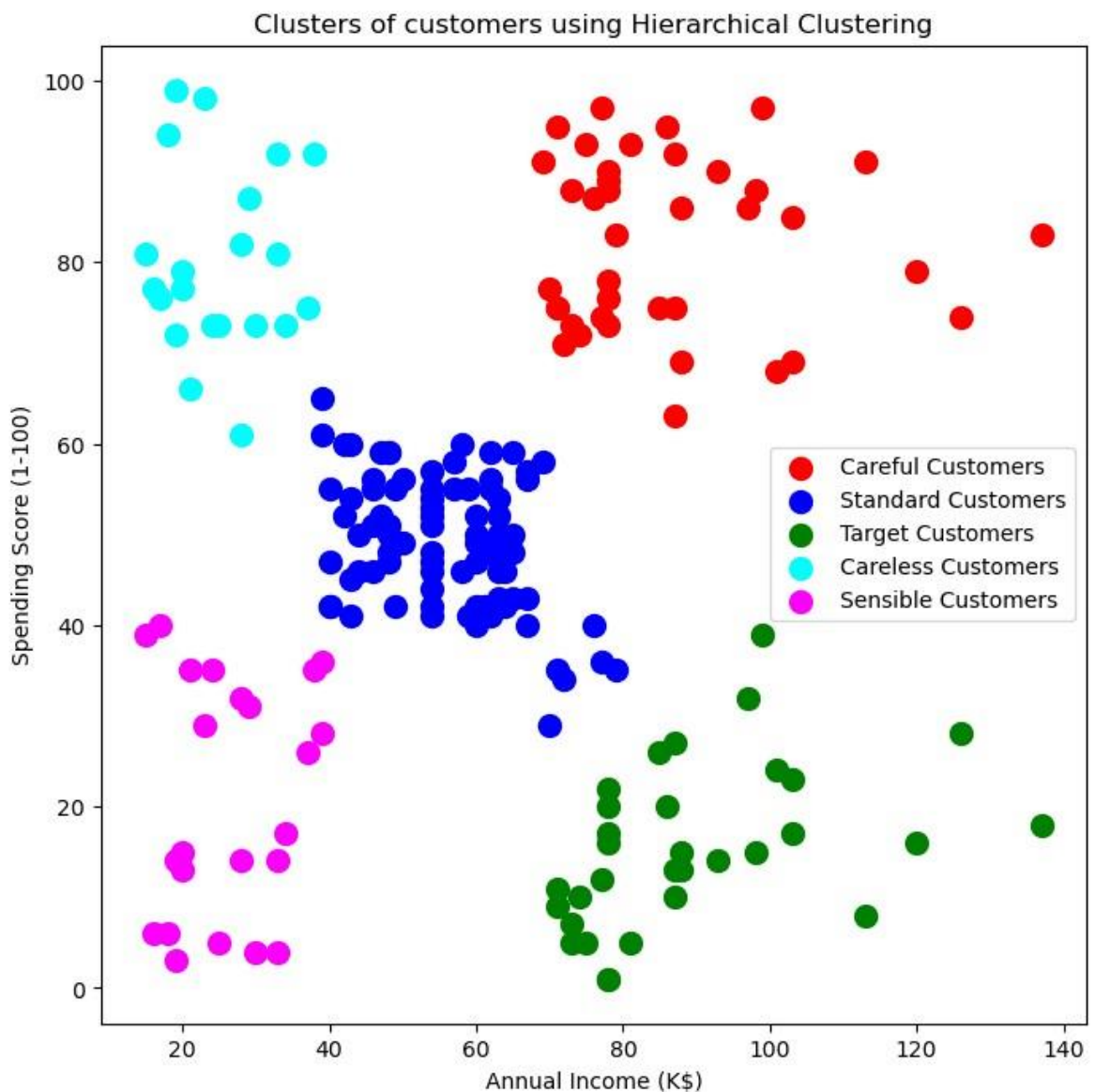
```
In [21]: from sklearn.cluster import AgglomerativeClustering hc =
AgglomerativeClustering(n_clusters = 5, affinity = 'euclidean', linkage = 'y_hc =
hc.fit_predict(X)
```

C:\Users\PARTH\anaconda3\Lib\site-packages\sklearn\cluster_agglomerative.py:98

3: FutureWarning: Attribute `affinity` was deprecated in version 1.2 and will be removed in 1.4. Use `metric` instead warnings.warn(

In [22]: *# Visualising the clusters*

```
plt.figure(figsize=(8,8)) plt.scatter(X[y_hc == 0, 0], X[y_hc == 0, 1], s = 100,
c = 'red', label = 'Careful Customers') plt.scatter(X[y_hc == 1, 0], X[y_hc == 1, 1], s = 100, c =
'blue', label = 'Standard Customers') plt.scatter(X[y_hc == 2, 0], X[y_hc == 2, 1], s = 100, c =
'green', label = 'Target Customers') plt.scatter(X[y_hc == 3, 0], X[y_hc == 3, 1], s = 100, c =
'cyan', label = 'Careless Customers') plt.scatter(X[y_hc == 4, 0], X[y_hc == 4, 1], s = 100, c =
'magenta', label = 'Sensible Customers') plt.title('Clusters of customers using Hierarchical
Clustering') plt.xlabel('Annual Income (K$)') plt.ylabel('Spending Score (1-100)') plt.legend() plt.show()
```



In []: In []:

