```
import pandas as pd
for filename in
                             print(os.path.join(dirname,
          filenames:
          filename))
 In [3]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          df = pd.read_csv(r'C:\Users\admin\Downloads\archive (1)\Mall_Customers.csv' )
 In [8]:
 In
          df
 [9]:
              CustomerID Genre Age Annual Income (k$) Spending Score (1-100)
 Out[9]:
           0
                      1
                           Male
                                 19
                                                  15
                                                                       39
           1
                      2
                                 21
                                                                       81
                                                  15
                           Male
           2
                      3 Female
                                 20
                                                                        6
                                                   16
           3
                                                                       77
                      4 Female
                                 23
                                                   16
                      5 Female
                                 31
                                                  17
                                                                       40
         195
                                                                       79
                                 35
                                                  120
                     196
                        Female
         196
                                                  126
                                                                       28
                     197 Female
                                 45
         197
                     198
                           Male
                                                  126
                                                                       74
                                 32
         198
                     199
                                 32
                                                  137
                           Male
                                                                       18
         199
                    200
                                 30
                                                  137
                                                                       83
                      Male
         200
                    rows × 5
                    columns
In [10]:
          x = df.iloc[:,3:]
Out[10]:
              Annual Income (k$) Spending Score (1-100)
           0
                            15
                                 39
           1
                            15
                                 81
           2
                            16
                                 6
           3
                            16
                                 77
           4
                            17
                                 40
         195
                                 79
                           120
         196
                           126
                                  28
         197
                           126
                                 74
                           137
         199
                                 83
                          137
         200
                          rows × 2 columns
In [11]:
          plt.title('Unclustered Data')
          sns.scatterplot(x=x['Annual Income (k$)'],y=x['Spending Score (1-100)'])
```

<AxesSubplot:title={'center':'Unclustered Data'}, xlabel='Annual Income (k\$)', ylabel='Spending Score (1-100)'> Out[11]:

In [1]: **import** numpy as np

```
from sklearn.cluster import KMeans, AgglomerativeClustering
In [12]:
                       km = KMeans(n_clusters=4)
In [13]:
                      km.fit_predict(x)
In [14]:
                     C:\Users\admin\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on W
                     indows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_TH
                         warnings.warn( array([3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3,
                     0, 3, 0, 3, 0,
Out[14]:
                                   3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 0, 3, 3,
                                   3, 3, 3, 3, 3, 3, 3, 3, 3, 3, 3, 3, 2, 1, 2, 1, 2, 1, 2, 1, 2,
                                   1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2,
                     1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2,
                     2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2, 1, 2,
                                   1, 2])
In [17]:
                       #sse
                       km.inertia_
Out[17]: 12152.549592074593
In [18]: sse =[] for k in range(1,16):
                       km = KMeans(n_clusters=k)
                       km.fit_predict(x)
                       sse.append(km.inertia_)
                      sse
In [19]:
Out[19]: [269981.28,
                       184609.98434090617,
                       106348.37306211118,
                       73679.78903948836,
                       44448.45544793371,
                     38797.9027638142,
                       31644.31903792021,
                       25354.360937251156,
                       23543.475418928974,
                       20614.671734467087,
                       18283.650928500927,
                       16429.94737981317,
                       14416.328255078253,
                       14011.958208458209,
                       12746.851332869057]
plt.ylabel('SSE')
```

```
Text(0, 0.5, 'SSE')
```

```
[21]:
           from sklearn.metrics import silhouette_score
In [22]:
          silh =[] for k in
          range(2,16):
              km = KMeans(n_clusters=k)
                                             labels
          = km.fit_predict(x)
                                  score =
          silhouette_score(x, labels)
          silh.append(score)
          silh
In [23]:
Out[23]: [0.2968969162503008,
          0.46761358158775435,
          0.4931963109249047,
          0.44647974211285657,
         0.4561972992633143,
          0.45425910793220675,
           0.4345270427308732,
           0.4571957283008456,
           0.4284833830885232,
          0.42483696591624037,
          0.4148939785943696,
           0.41870926222608434,
          0.42004018597903964,
          0.39079438087316754]
In [24]: sns.lineplot(range(2,16),y = silh)
```

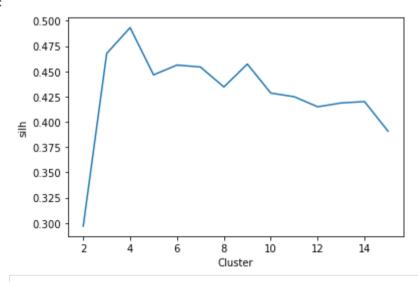
C:\Users\admin\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg:
x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
 warnings.warn(

Text(0, 0.5, 'silh')

plt.xlabel('Cluster')
plt.ylabel('silh')

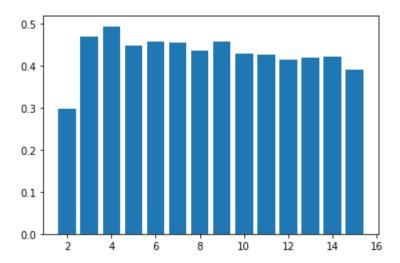
Out[24]:

In



```
plt.bar(range(2,16,1),silh)
In [25]:
```

<BarContainer object of 14 artists >



```
km = KMeans(n_clusters=5,random_state=1)
```

Out[25]:

```
In [26]:
```

Out[28]:

In [29]:

```
labels = km.fit_predict(x)
In [27]:
```

C:\Users\admin\anaconda3\lib\site-packages\sklearn\cluster_kmeans.py:1429: UserWarning: KMeans is known to have a memory leak on W indows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_TH READS=1.

warnings.warn(

```
km.labels_
In [28]:
array([4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4, 2, 4,
```

```
1, 3])
cent = km.cluster_centers_
```

```
In [30]: plt.title('Clustered Data')
sns.scatterplot(x=x['Annual Income (k$)'],y=x['Spending Score (1-
100)'],c=labels ) sns.scatterplot(cent[:,0],cent[:,1], s=200, color='red')
```

C:\Users\admin\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
<AxesSubplot:title={'center':'Clustered Data'}, xlabel='Annual Income (k$)', ylabel='Spending Score (1-100)'> Out[30]:
```

```
100
                              80
                      Spending Score (1-100)
                              60
                              40
                             20
                                0
                                                           40
                                                                                                          100
                                                                                                                                           140
                                           20
                                                                                           80
                                                                                                                           120
                                                                          Annual Income (k$)
                        df[labels==0]
                                 CustomerID
                                                            Genre Age Annual Income (k$) Spending Score (1-100)
                         43
                                                         Female
                                                                             31
                                                                                                                                                                    61
Out[31]:
                                                                                                                                                                    55
                         46
                                                  47
                                                              Female 50
                                                                                                                     40
                         47
                                                                                                                                                                    47
                                                  48
                                                              Female 27
                                                                                                                     40
                         48
                                                                                                                                                                    42
                                                  49
                                                              Female 29
                                                                                                                     40
                         49
                                                                                                                     40
                                                                                                                                                                    42
                                                  50
                                                              Female 31
                                 CustomerID
                                                            Genre Age Annual Income (k$) Spending Score (1-100)
                      121
                                                122
                                                          Female
                                                                              38
                                                                                                                     67
                                                                                                                                                                    40
                      122
                                                123
                                                          Female
                                                                              40
                                                                                                                     69
                                                                                                                                                                    58
                      126
                                                127
                                                               Male
                                                                              43
                                                                                                                     71
                                                                                                                                                                    35
                      132
                                                133
                                                          Female
                                                                              25
                                                                                                                     72
                                                                                                                                                                    34
                                                                                                                                                                    40
                                                                                                                     76
                      142
                                                143
                                                        Female
                                                                              28
                    81 rows × 5 columns
                       agl = AgglomerativeClustering (n_clusters=5)
In [32]:
                       alabels = agl.fit_predict(x)
In [33]:
                       alabels
In [34]:
                      array([4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3,
Out[34]:
                                     4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 1,
                                     1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 2, 1, 2, 0, 2, 0, 2,
                                     1, 2, 0, 2, 0, 2, 0, 2, 0, 2, 1, 2, 0, 2, 1, 2, 0, 2, 0, 2, 0, 2,
                                     0, 2, 0, 2, 0, 2, 1, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2,
                                     0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2, 0, 2,
                                     0, 2], dtype=int64)
In [35]:
                        plt.figure(figsize=(16,9)) plt.subplot(1,2,1)
                        plt.title('Agglomerative')
                        sns.scatterplot(x=x['Annual Income (k$)'],y=x['Spending Score (1-100)'], c= alabels)
                        plt.subplot(1,2,2)
                        plt.title('KMEANS')
                        sns.scatterplot(x=x['Annual Income (k$)'],y=x['Spending Score (1-100)'],c=labels ) sns.scatterplot(cent[:,0],cent[:,1],
                        s=200, color='red')
```

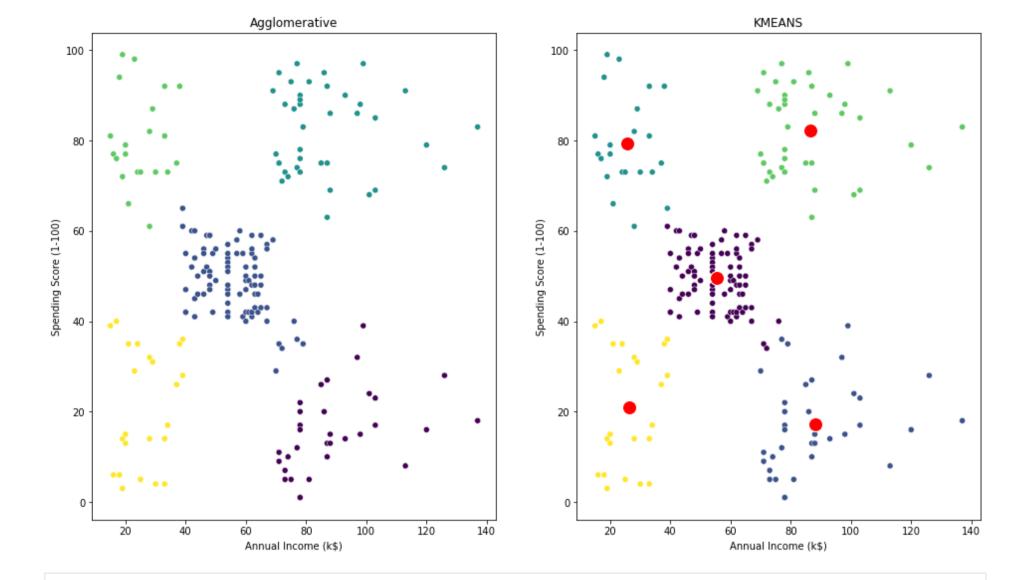
In

[31]:

Clustered Data

C:\Users\admin\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x,
y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will
result in an error or misinterpretation.
warnings.warn(

<AxesSubplot:title={'center':'KMEANS'}, xlabel='Annual Income (k\$)', ylabel='Spending Score (1-100)'> Out[35]:



In []: