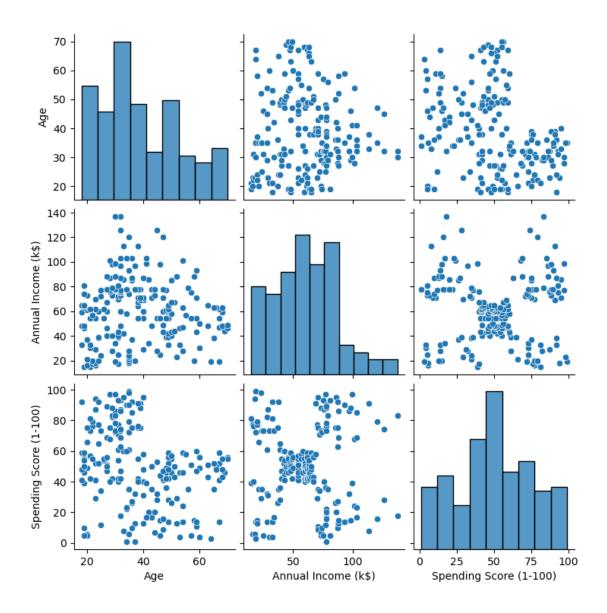
week9

September 25, 2024

```
[7]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
      import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
      import seaborn as sns
 [8]: data = pd.read_csv('C:/Users/HP/Downloads/Mall_Customers.csv')
      print(data.shape)
     (200, 5)
 [9]: data = data.drop('CustomerID',axis=1)
      data
 [9]:
                                             Spending Score (1-100)
            Genre
                   Age
                        Annual Income (k$)
             Male
                    19
                                                                   39
      0
                                          15
      1
             Male
                    21
                                          15
                                                                  81
      2
           Female
                     20
                                         16
                                                                   6
      3
           Female
                    23
                                         16
                                                                  77
           Female
                    31
                                         17
                                                                  40
                                                                  79
                                        120
      195 Female
                    35
      196 Female
                                        126
                                                                  28
                    45
                                                                  74
      197
             Male
                    32
                                        126
      198
             Male
                                        137
                                                                   18
      199
             Male
                    30
                                        137
                                                                  83
      [200 rows x 4 columns]
[10]: data.head()
                                           Spending Score (1-100)
[10]:
                Age
                     Annual Income (k$)
          Genre
           Male
      0
                  19
                                       15
                                                                 39
      1
           Male
                  21
                                       15
                                                                 81
      2 Female
                  20
                                       16
                                                                  6
```

```
16
      3 Female
                   23
                                                                 77
      4 Female
                                        17
                                                                 40
                   31
[11]: data.describe()
[11]:
                          Annual Income (k$)
                                               Spending Score (1-100)
                     Age
                                   200.000000
                                                            200.000000
             200.000000
      count
                                    60.560000
      mean
              38.850000
                                                             50.200000
      std
              13.969007
                                    26.264721
                                                             25.823522
      min
              18.000000
                                    15.000000
                                                              1.000000
      25%
              28.750000
                                    41.500000
                                                             34.750000
      50%
              36.000000
                                    61.500000
                                                             50.000000
                                    78.000000
                                                             73.000000
      75%
              49.000000
      max
              70.000000
                                   137.000000
                                                             99.000000
[12]: from sklearn.preprocessing import LabelEncoder
      label_encoder = LabelEncoder()
      #data['Genre'] = label_encoder.fit_transform(data['Gnere'])
      data
[12]:
            Genre
                         Annual Income (k$)
                                              Spending Score (1-100)
                    Age
      0
             Male
                     19
                                          15
                                                                   39
      1
             Male
                     21
                                          15
                                                                    81
      2
           Female
                     20
                                          16
                                                                    6
      3
           Female
                     23
                                          16
                                                                    77
      4
           Female
                     31
                                          17
                                                                    40
      . .
                                         120
                                                                   79
      195 Female
                     35
      196 Female
                     45
                                         126
                                                                   28
      197
             Male
                     32
                                         126
                                                                   74
             Male
      198
                     32
                                         137
                                                                    18
      199
             Male
                     30
                                         137
                                                                    83
      [200 rows x 4 columns]
[13]: sns.pairplot(data)
     C:\Users\HP\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning:
     The figure layout has changed to tight
       self._figure.tight_layout(*args, **kwargs)
```

[13]: <seaborn.axisgrid.PairGrid at 0x260f4823510>



```
[14]: x = data.iloc[:, 2:4].values
print(x.shape)
x
(200, 2)
[14]: array([[ 15, 39],
```

- [18, 94],
- [19, 3],
- [19, 72],
- [19, 14],
- [19, 99],
- [20, 15],
- [20, 77],
- [20, 13],
- [20, 79],
- [21, 35],
- [21, 66],
- [23, 29],
- [23, 98],
- [24, 35],
- [24, 73],
- [25, 5],
- [25, 73],
- [28, 14],
- [00 00]
- [28, 82],
- [28, 32],
- [28, 61],
- [29, 31],
- [29, 87],
- [30, 4],
- [30, 73],
- [33, 4],
- [33, 92],
- [33, 14],
- [33, 81],
- [34, 17],
- [34, 73],
- [37, 26],
- [37, 75],
- [38, 35],
- [38, 92],
- [39, 36],
- [39, 61],
- [39, 28],
- [39, 65],
- [40, 55],
- [40, 47],
- [40, 42],
- [40, 42],
- [42, 52],
- [42, 60],
- [43, 54],
- [43, 60],

- [43, 45],
- [43, 41],
- 50], [44,
- [44, 46],
- [46, 51],
- [46, 46],
- [46, 56],
- [46, 55],
- 52], [47,
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- [48, 59],
- [48, 47],
- [49, 55],
- [49, 42],
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- [54, 52],
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- 51], [54,
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- [57, 58],
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- 40],
- [60, 42],
- [60, 52],
- [60, 47],
- [60, 50],
- [61, 42],
- [61, 49],
- 41], [62,

- [62, 48],
- 59], [62,
- 55], [62,
- [62, 56],
- [62, 42],
- [63, 50],
- [63, 46],
- [63, 43],
- [63, 48],
- [63, 52],
- [63, 54],
- [64, 42],
- 46], [64,
- [65, 48],
- [65, 50],
- 43], [65,
- [65, 59],
- 43], [67,
- [67, 57],
- [67, 56], [67, 40],
- [69, 58],
- [69, 91],
- [70, 29],
- [70, 77],
- [71, 35],
- 95], [71,
- [71, 11],
- [71, 75], [71, 9],
- [71, 75],
- [72, 34],
- [72, 71],
- 5], [73,
- [73, 88],
- [73, 7],
- [73, 73],
- [74, 10],
- [74, 72],
- [75, 5],
- [75, 93],
- [76, 40],
- [76, 87],
- [77, 12],
- [77, 97],
- [77, 36],
- [77, 74],

```
[ 78,
       22],
       90],
[ 78,
       17],
[ 78,
[ 78,
       88],
[ 78,
       20],
[ 78,
       76],
[ 78,
       16],
[ 78,
       89],
[ 78,
        1],
[ 78,
       78],
[ 78,
        1],
[ 78,
       73],
[79,
       35],
[79,
       83],
[ 81,
        5],
       93],
[ 81,
[ 85,
       26],
[ 85,
       75],
[86,
       20],
[ 86,
       95],
[ 87,
       27],
[ 87,
       63],
[ 87,
       13],
[ 87,
       75],
[ 87,
       10],
[ 87,
       92],
       13],
[88,
[88,
       86],
[88,
       15],
[88,
       69],
[ 93,
       14],
[ 93,
       90],
[ 97,
       32],
[ 97,
       86],
       15],
[ 98,
[ 98,
       88],
[ 99,
       39],
[ 99,
       97],
       24],
[101,
       68],
[101,
[103,
       17],
```

[103,

[103,

[103,

[113,

[113,

[120,

85],

23],

69],

8],

91], 16],

```
[120, 79],

[126, 28],

[126, 74],

[137, 18],

[137, 83]], dtype=int64)
```

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

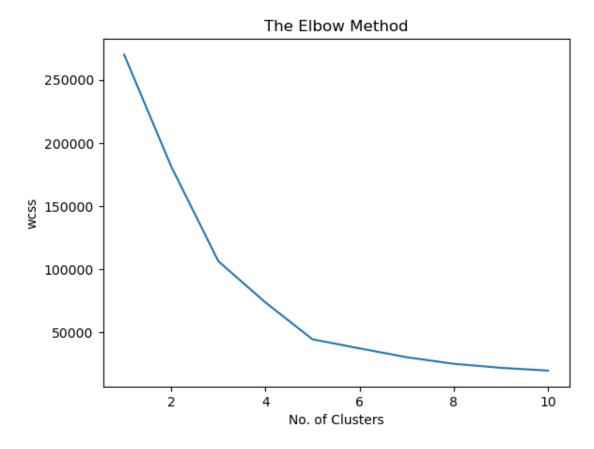
warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

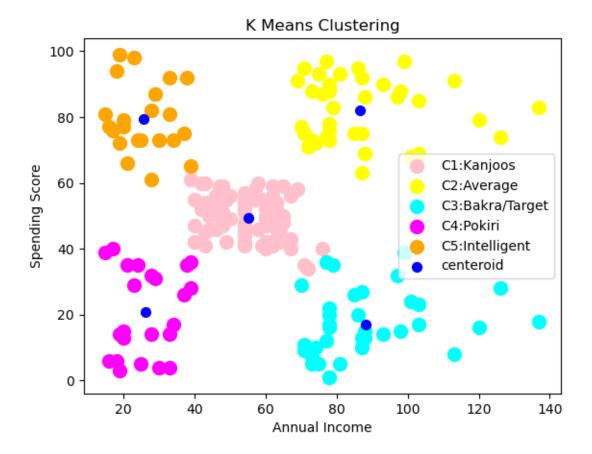
C:\Users\HP\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(



```
[16]: km1 = KMeans(n_clusters = 5, init = 'k-means++', max_iter = 300, n_init = 10,__
             →random_state = 0)
           y_means = km1.fit_predict(x)
          C:\Users\HP\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 than available threads. You can avoid it by setting the
          environment variable OMP_NUM_THREADS=1.
              warnings.warn(
[17]: y_means
[17]: array([3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4,
                        3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 4, 3, 0,
                        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 2, 1, 0, 1, 2, 1, 2, 1,
                        0, 1, 2, 1, 2, 1, 2, 1, 2, 1, 0, 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, 1])
[18]: km1.cluster_centers_
[18]: array([[55.2962963, 49.51851852],
                         [86.53846154, 82.12820513],
                                               , 17.11428571],
                         [26.30434783, 20.91304348],
                         [25.72727273, 79.363636363]])
[19]: plt.scatter(x[y_means == 0, 0], x[y_means == 0, 1], s = 100, c = 'pink', label__
            plt.scatter(x[y_means == 1, 0], x[y_means == 1, 1], s = 100, c = 'yellow', <math>u
              ⇔label = 'C2:Average')
           plt.scatter(x[y_means == 2, 0], x[y_means == 2, 1], s = 100, c = 'cyan', label_u
             plt.scatter(x[y] means == 3, 0], x[y] means == 3, 1], s = 100, c = 'magenta',
             ⇔label = 'C4:Pokiri')
           plt.scatter(x[y_means == 4, 0], x[y_means == 4, 1], s = 100, c = 'orange', u
              ⇔label = 'C5:Intelligent')
           plt.scatter(km1.cluster_centers_[:,0], km1.cluster_centers_[:, 1], s = 50, c = __
             ⇔'blue' , label = 'centeroid')
           plt.title('K Means Clustering')
           plt.xlabel('Annual Income')
           plt.ylabel('Spending Score')
           plt.legend()
```

plt.show()



```
[20]: x[y_means == 1, 0]
[20]: array([ 69,
                  70, 71,
                            71,
                                 71,
                                      72, 73, 73, 74,
                                                          75, 76, 77, 77,
                                           79, 81,
                                                     85,
             78,
                  78,
                       78,
                            78,
                                 78,
                                      78,
                                                          86,
                                                               87, 87,
                                      99, 101, 103, 103, 113, 120, 126, 137],
             88,
                  88,
                       93,
                            97,
                                 98,
            dtype=int64)
[21]: km1.inertia_
[21]: 44448.4554479337
[22]: km1.cluster_centers_
[22]: array([[55.2962963, 49.51851852],
             [86.53846154, 82.12820513],
             [88.2
                        , 17.11428571],
             [26.30434783, 20.91304348],
             [25.72727273, 79.363636363]])
```

```
[23]: x = data.iloc[:, [1, 3]].values
      x.shape
[23]: (200, 2)
[24]: data.columns
[24]: Index(['Genre', 'Age', 'Annual Income (k$)', 'Spending Score (1-100)'],
      dtype='object')
[26]: 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_
       \Rightarrow= 10, random state = 0)
          kmeans.fit(x)
          wcss.append(kmeans.inertia_)
      plt.rcParams['figure.figsize'] = (7, 5)
      plt.plot(range(1, 11), wcss)
      plt.title('K-Means Clustering(The Elbow Method)', fontsize = 20)
      plt.xlabel('Age')
      plt.ylabel('Count')
      plt.show()
     C:\Users\HP\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 than available threads. You can avoid it by setting the
     environment variable OMP_NUM_THREADS=1.
       warnings.warn(
     C:\Users\HP\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 than available threads. You can avoid it by setting the
     environment variable OMP_NUM_THREADS=1.
       warnings.warn(
     C:\Users\HP\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 than available threads. You can avoid it by setting the
     environment variable OMP_NUM_THREADS=1.
       warnings.warn(
     C:\Users\HP\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 than available threads. You can avoid it by setting the
     environment variable OMP_NUM_THREADS=1.
       warnings.warn(
     C:\Users\HP\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 than available threads. You can avoid it by setting the
     environment variable OMP_NUM_THREADS=1.
```

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

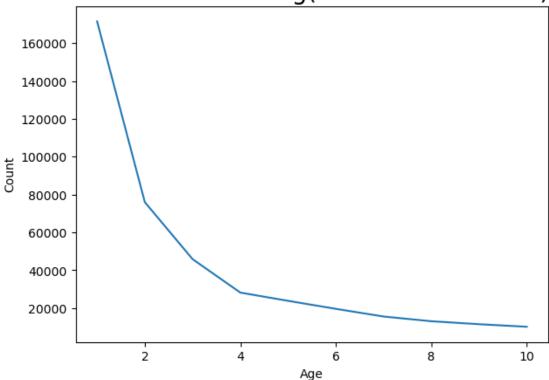
C:\Users\HP\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(

K-Means Clustering(The Elbow Method)



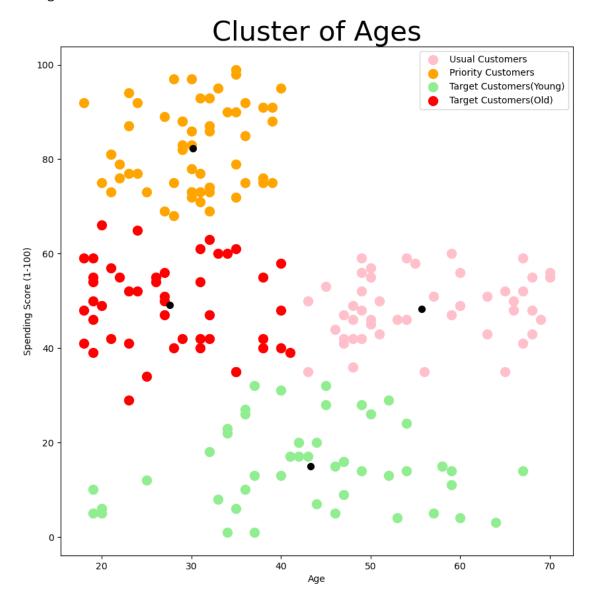
```
[25]: km2 = KMeans(n_clusters = 4, init = 'k-means++', max_iter = 300, n_init = 10,
      →random_state = 0)
     ymeans = km2.fit_predict(x)
     plt.rcParams['figure.figsize'] = (10, 10)
     plt.title('Cluster of Ages', fontsize = 30)
     plt.scatter(x[ymeans == 0, 0], x[ymeans == 0, 1], s = 100, c = 'pink', label =
       plt.scatter(x[ymeans == 1, 0], x[ymeans == 1, 1], s = 100, c = 'orange', label
       →= 'Priority Customers')
     plt.scatter(x[ymeans == 2, 0], x[ymeans == 2, 1], s = 100, c = 'lightgreen', u
       →label = 'Target Customers(Young)')
     plt.scatter(x[ymeans == 3, 0], x[ymeans == 3, 1], s = 100, c = 'red', label =

¬'Target Customers(Old)')
     plt.scatter(km2.cluster_centers_[:, 0], km2.cluster_centers_[:, 1], s = 50, c = __
      plt.xlabel('Age')
     plt.ylabel('Spending Score (1-100)')
     plt.legend()
     plt.show()
```

C:\Users\HP\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 than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.

warnings.warn(



```
[26]: x = data.iloc[:, [0, 3]].values
x.shape
[26]: (200, 2)
[]:
```