In [27]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

## Out[28]:

|  |     | 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                     |
|  | ••• |            |        |     |                     |                        |
|  | 195 | 196        | Female | 35  | 120                 | 79                     |
|  | 196 | 197        | Female | 45  | 126                 | 28                     |
|  | 197 | 198        | Male   | 32  | 126                 | 74                     |
|  | 198 | 199        | Male   | 32  | 137                 | 18                     |
|  | 199 | 200        | Male   | 30  | 137                 | 83                     |

200 rows × 5 columns

In [29]: dataset.isnull().sum()

Out[29]: CustomerID 0
Gender 0
Age 0
Annual Income (k\$) 0
Spending Score (1-100) 0
dtype: int64

In [30]: dataset.describe()

| Out[30]: | ut[30]: CustomerID      |            | Age        | Annual Income (k\$) | Spending Score (1-100) |  |
|----------|-------------------------|------------|------------|---------------------|------------------------|--|
|          | <b>count</b> 200.000000 |            | 200.000000 | 200.000000          | 200.000000             |  |
|          | mean                    | 100.500000 | 38.850000  | 60.560000           | 50.200000              |  |
|          | std                     | 57.879185  | 13.969007  | 26.264721           | 25.823522              |  |
|          | min                     | 1.000000   | 18.000000  | 15.000000           | 1.000000               |  |
|          | 25%                     | 50.750000  | 28.750000  | 41.500000           | 34.750000              |  |
|          | 50%                     | 100.500000 | 36.000000  | 61.500000           | 50.000000              |  |
|          | <b>75</b> %             | 150.250000 | 49.000000  | 78.000000           | 73.000000              |  |
|          | max                     | 200.000000 | 70.000000  | 137.000000          | 99.000000              |  |

In [31]: dataset.dtypes

Out[31]: CustomerID int64
Gender object
Age int64
Annual Income (k\$) int64
Spending Score (1-100) int64
dtype: object

In [32]: dataset['Gender'].value\_counts()

```
Out[32]: Gender
          Female
                     112
          Male
                      88
          Name: count, dtype: int64
         annual_income=dataset['Annual Income (k$)']
In [33]:
          spending score=dataset['Spending Score (1-100)']
In [34]: sns.scatterplot(x=annual_income,y=spending_score,data=dataset)
Out[34]: <Axes: xlabel='Annual Income (k$)', ylabel='Spending Score (1-100)'>
            100
             80
        Spending Score (1-100)
             60
             40
```

20

0

20

40

120

140

100

80

Annual Income (k\$)

60

```
Out[37]: 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],
               [ 20, 13],
               [ 20, 79],
               [ 21, 35],
               [ 21, 66],
               [ 23, 29],
               [ 23, 98],
               [ 24, 35],
               [ 24, 73],
               [ 25,
                     5],
               [ 25, 73],
               [ 28, 14],
               [ 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],
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- [ 44, 50],
- [ 44, 46],
- [ 46, 51],
- [ 46, 46],
- [ 46, 56],
- [ 46, 55],
- [ 47, 52],
- [ 47, 59],
- [ 48, 51],
- [ 48, 59],
- [ 48, 50],
- [ 48, 48],
- [ 48, 59],
- [ 48, 47],
- [ 49, 55],
- [ 49, 42],
- [ 50, 49],
- [ 50, 56],
- [ 54, 47],
- [ 54, 54],
- [ 54, 53],
- [ 54, 48],
- [ 54, 52],
- [ 54, 42],
- [ 54, 51],

- [ 54, 55],
- [ 54, 41],
- [ 54, 44],
- [ 54, 57],
- [ 54, 46],
- [ 57, 58],
- [ 57, 55],
- [ 58, 60],
- [ 58, 46],
- [ 59, 55],
- [ 59, 41],
- [ 60, 49],
- [ 60, 40],
- [ 60, 42],
- [ 60, 52],
- [ 60, 47],
- [ 60, 50],
- [ 61, 42],
- [ 61, 49],
- [ 62, 41],
- [ 62, 48],
- [ 62, 59],
- [ 62, 55],
- [ 62, 56],
- [ 62, 42],
- [ 63, 50],
- [ 63, 46],
- [ 63, 43],
- [ 63, 48],
- [ 63, 52],
- [ 63, 54],
- [ 64, 42],
- [ 64, 46],
- [ 65, 48],
- [ 65, 50],
- [ 65, 43],
- [ 65, 59],
- [ 67, 43],
- [ 67, 57],
- [ 67, 56],
- [ 67, 40],

```
[ 69, 58],
```

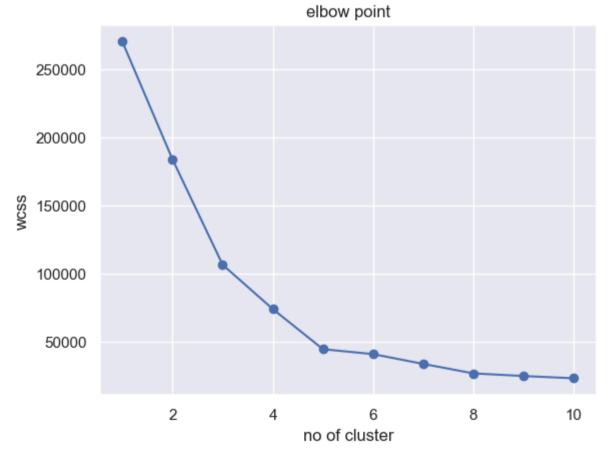
- [ 69, 91],
- [ 70, 29],
- [ 70, 77],
- [ 71, 35],
- [ 71, 95],
- [ 71, 11],
- [ 71, 75],
- [ 71, 9],
- [ 71, 75],
- [ 72, 34],
- [ 72, 71],
- [ 73, 5],
- [ 73, 88],
- [ 73, 7],
- [ 73, 73],
- [ 74, 10],
- [ 74, 72],
- [ 75, 5],
- [ 75, 93],
- [ 76, 40],
- [ 76, 87],
- [ 77, 12],
- [ 77, 97],
- [ 77, 36],
- [ 77, 74],
- [ 78, 22],
- [ 78, 90],
- [ 78, 17],
- [ 78, 88],
- [ 78, 20],
- [ 78, 76],
- [ 78, 16],
- [ 78, 89],
- [ 78, 1],
- [ 78, 78],
- [ 78, 1],
- [ 78, 73],
- [ 79, 35],
- [ 79, 83],
- [ 81, 5],

```
[ 81, 93],
[ 85, 26],
[ 85, 75],
[ 86, 20],
[ 86, 95],
[ 87, 27],
[ 87, 63],
[ 87, 13],
[ 87, 75],
[ 87, 10],
[ 87, 92],
[ 88, 13],
[ 88, 86],
[ 88, 15],
[ 88, 69],
[ 93, 14],
[ 93, 90],
[ 97, 32],
[ 97, 86],
[ 98, 15],
[ 98, 88],
[ 99, 39],
[ 99, 97],
[101, 24],
[101, 68],
[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)
```

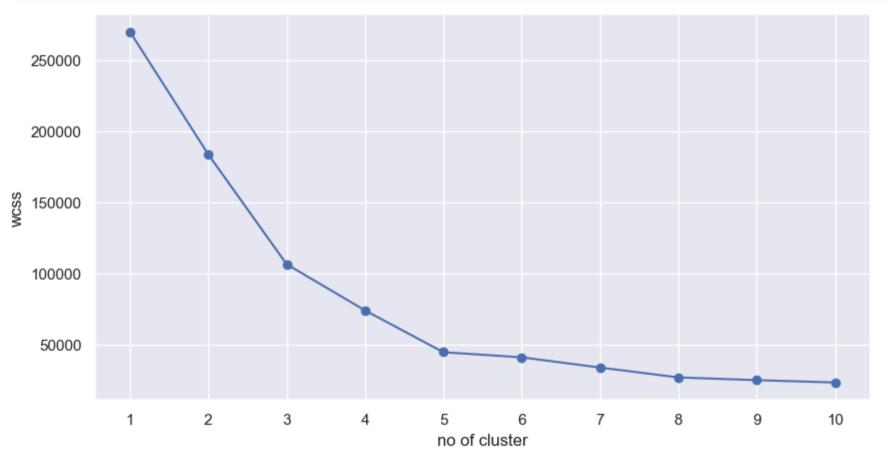
```
In [38]: from sklearn.cluster import KMeans
    wcss=[]
```

```
for i in range(1,11):
    km=KMeans(n_clusters=i,init='k-means++',random_state=42)
    km.fit(x)
    wcss.append(km.inertia_)

In [39]:
    sns.set()
    plt.plot(range(1,11),wcss,marker='o')
    plt.xlabel('no of cluster')
    plt.ylabel('wcss')
    plt.title('elbow point')
Out[39]: Text(0.5, 1.0, 'elbow point')
```

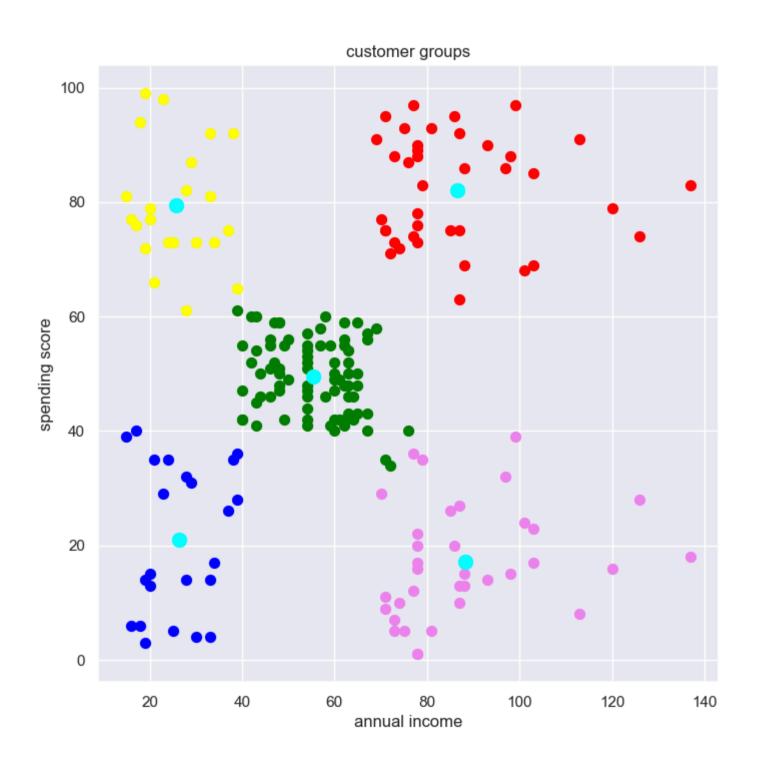


```
In [40]: plt.figure(figsize=(10,5))
    plt.plot([i for i in range(1,11)],wcss,marker='o')
    plt.xlabel('no of cluster')
    plt.xticks([i for i in range(1,11)])
    plt.ylabel('wcss')
    plt.show()
```



In [41]: kmeans=KMeans(n\_clusters=5,init='k-means++',random\_state=42)
 y=kmeans.fit\_predict(x)
 print(y)

```
1 3 1 3 1 3 1 3 1 3 1 3 1 3 1 3 1
In [45]: plt.figure(figsize=(8,8))
     plt.scatter(x[y=0,0], x[y=0,1], s=50, c='green', label='cluster 1')
     plt.scatter(x[y==1,0], x[y==1,1], s=50, c='red', label='cluster 2')
     plt.scatter(x[y==2,0], x[y==2,1], s=50, c='yellow', label='cluster 3')
     plt.scatter(x[y==3,0], x[y==3,1], s=50, c='violet', label='cluster 4')
     plt.scatter(x[y=4,0], x[y=4,1], s=50, c='blue', label='cluster 5')
     plt.scatter(kmeans.cluster centers [:,0], kmeans.cluster centers [:,1], s=100, c='cyan', label='centroids')
     plt.title('customer groups')
     plt.xlabel('annual income')
     plt.ylabel('spending score')
     plt.show()
```



| In [ ]: |  |  |  |
|---------|--|--|--|
| In [ ]: |  |  |  |