```
In [3]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.cluster import KMeans
customer_dataset = pd.read_csv("C:\\Users\\Acer\\OneDrive\\Desktop\\Mall_Customers.csv")
customer_dataset.head()
```

Out[3]:

	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]: customer_dataset.tail()

Out[4]:

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
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

```
In [5]: # shape of dataset
        customer_dataset.shape
        (200, 5)
Out[5]: (200, 5)
In [6]: # information about dataset
        customer dataset.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 200 entries, 0 to 199
        Data columns (total 5 columns):
                                     Non-Null Count Dtype
         #
             Column
             CustomerID
                                     200 non-null
                                                     int64
         0
            Gender
                                     200 non-null
         1
                                                     object
         2 Age
                                     200 non-null
                                                     int64
             Annual Income (k$)
                                     200 non-null
                                                     int64
             Spending Score (1-100) 200 non-null
                                                     int64
        dtypes: int64(4), object(1)
        memory usage: 7.9+ KB
In [7]: # datatypes of columns dataset
        customer dataset.dtypes
Out[7]: CustomerID
                                   int64
                                  object
        Gender
                                   int64
        Age
        Annual Income (k$)
                                   int64
        Spending Score (1-100)
                                   int64
        dtype: object
```

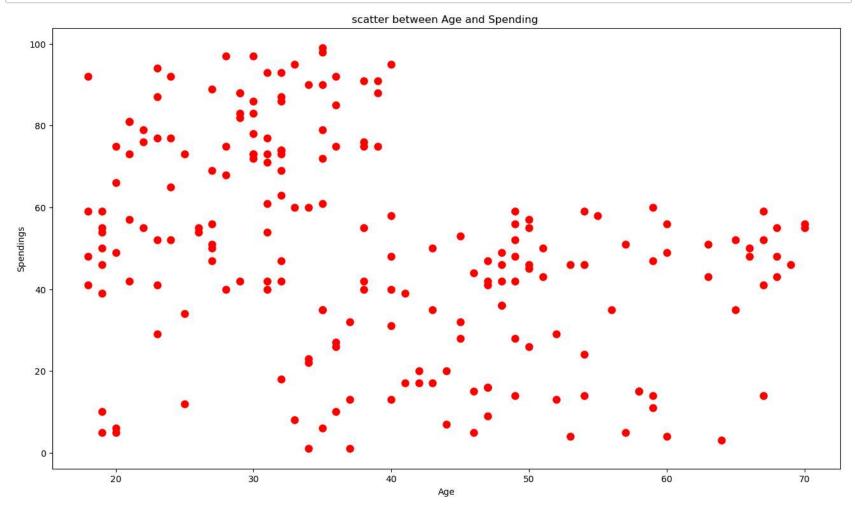
Out[8]:

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	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
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

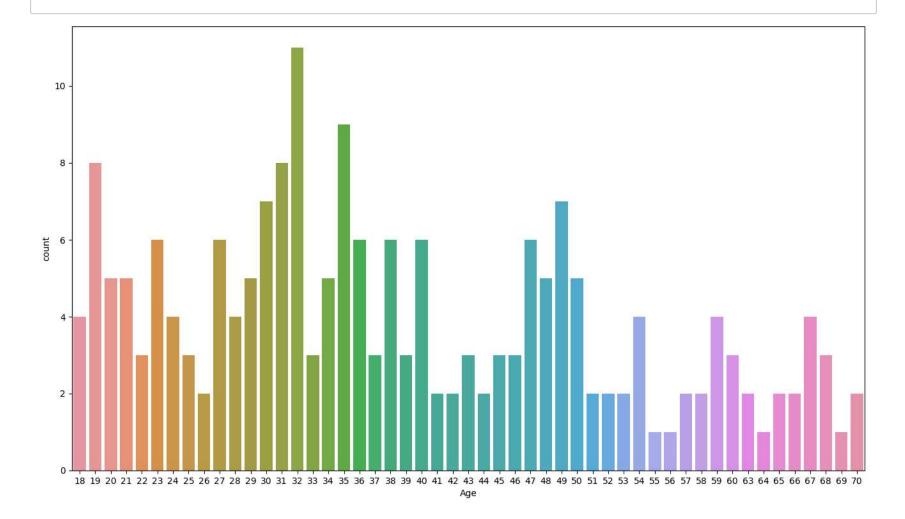
Data Visulalization

```
In [9]: plt.figure(figsize = (16,9))
    plt.scatter(x = customer_dataset.Age, y = customer_dataset['Spending Score (1-100)'], c = "red", linewidth =
        plt.title('scatter between Age and Spending ')
        plt.ylabel("Spendings")
        plt.xlabel("Age")

    plt.show()
```



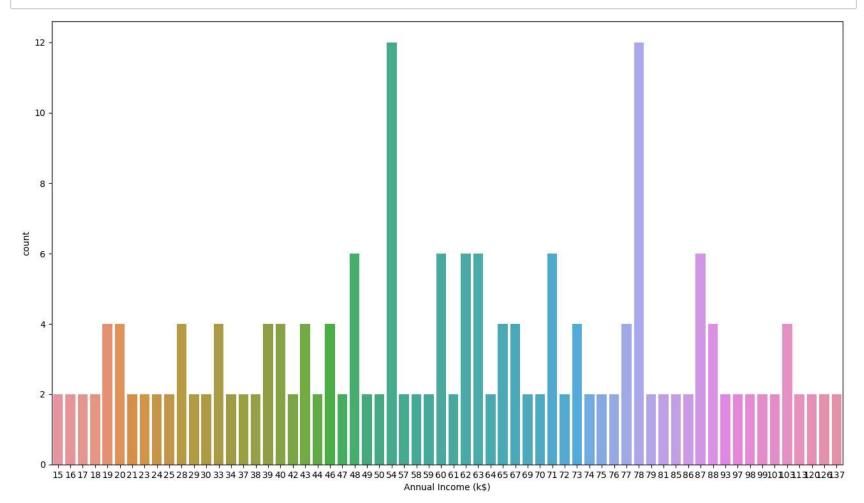
In [16]: plt.figure(figsize=(16,9))
 sns.countplot(x='Age', data=customer_dataset) # This is the plotting command
 plt.show() # This will display the plot



```
In [18]: # Plot configuration
    plt.figure(figsize=(16,9))

# Create the count plot
    sns.countplot(x='Annual Income (k$)', data=customer_dataset)

# Display the plot
    plt.show()
```



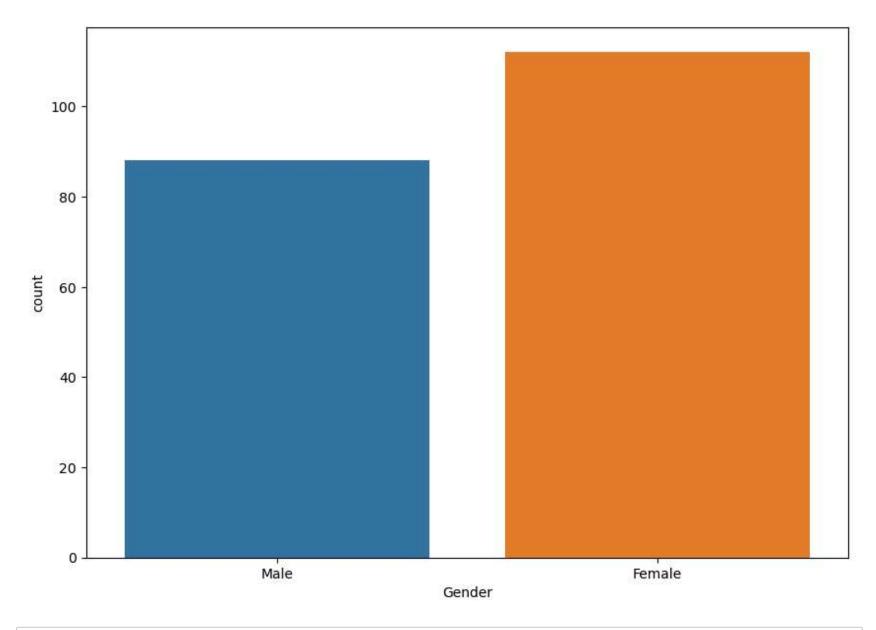
```
In [20]: # Set the figure size
         plt.figure(figsize=(16,9))
         # Create the count plot
         sns.countplot(x='Spending Score (1-100)', data=customer_dataset)
         # Show the plot
         plt.show()
            3 -
            2 -
            1 -
```

Spending Score (1-100)

```
In [21]: # Configure the plot size
    plt.figure(figsize=(10,7))

# Create the count plot with explicit keyword arguments
    sns.countplot(x='Gender', data=customer_dataset)

# Display the plot
    plt.show()
```



Choosing the Annual Income Column and Spending Columns

```
In [22]:
    customer_dataset.head()
```

Out[22]:

	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 [24]: customer_dataset.columns
```

```
In [25]:
         # using iloc for picking 3rd and 4th columns
         X = customer_dataset.iloc[:,[3,4]].values
         print(X)
          [ 2/ 24]
          [ 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]]
```

choosing the number of clusters

WCSS: within Clusters Sum of Squares

```
In [26]: # finding WCSS values for different number of clusters

wcss = []

for i in range(1,11):
    # 1 and 11 will be excluded
    kmeans = KMeans(n_clusters=i, init = 'k-means++', random_state = 42)
    kmeans.fit(X)
    wcss.append(kmeans.inertia_)
```

```
C:\pyth\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: 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
    super()._check_params_vs_input(X, default_n_init=10)
C:\pyth\Lib\site-packages\sklearn\cluster\_kmeans.py:1436: UserWarning: KMeans is known to have a memory lea
k on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the en
vironment variable OMP_NUM_THREADS=1.
    warnings.warn(
C:\pyth\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarning: The default value of `n_init` will
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    warnings.warn(
C:\pyth\Lib\site-packages\sklearn\cluster\_kmeans.py:1412: FutureWarning: The default value of `n_init` will
```

C:\pyth\Lib\site-packages\sklearn\cluster_kmeans.py:1412: FutureWarning: The default value of `n_init` will
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 super(). check params vs input(X, default n init=10)

C:\pyth\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.

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warnings.warn(

C:\pyth\Lib\site-packages\sklearn\cluster_kmeans.py:1412: FutureWarning: The default value of `n_init` will
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warnings.warn(

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C:\pyth\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.

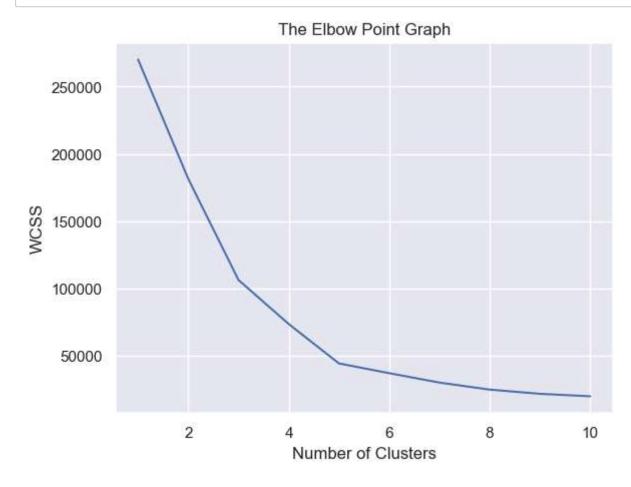
warnings.warn(

C:\pyth\Lib\site-packages\sklearn\cluster_kmeans.py:1412: 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 super(). check params vs input(X, default n init=10)

C:\pyth\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(

In [28]: # plot an elbow graph sns.set() plt.plot(range(1,11), wcss) plt.title("The Elbow Point Graph") plt.xlabel("Number of Clusters") plt.ylabel("WCSS") plt.show()



Optimum number of clusters = 5

Training the K-Means Clustering model

```
In [30]: kmeans = KMeans(n clusters =5, init = 'k-means++', random state =0)
                        # return a label for each data point based on their clusters
                        Y = kmeans.fit predict(X)
                        print(Y)
                        C:\pyth\Lib\site-packages\sklearn\cluster\ kmeans.py:1412: 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
                             super(). check params vs input(X, default n init=10)
                        C:\pyth\Lib\site-packages\sklearn\cluster\ kmeans.py:1436: UserWarning: KMeans is known to have a memory lea
                        k on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the en
                        vironment variable OMP NUM THREADS=1.
                             warnings.warn(
                         \begin{smallmatrix} 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 1 2 1 2 1 2 1 2 1 2 1 2 1 ]
```

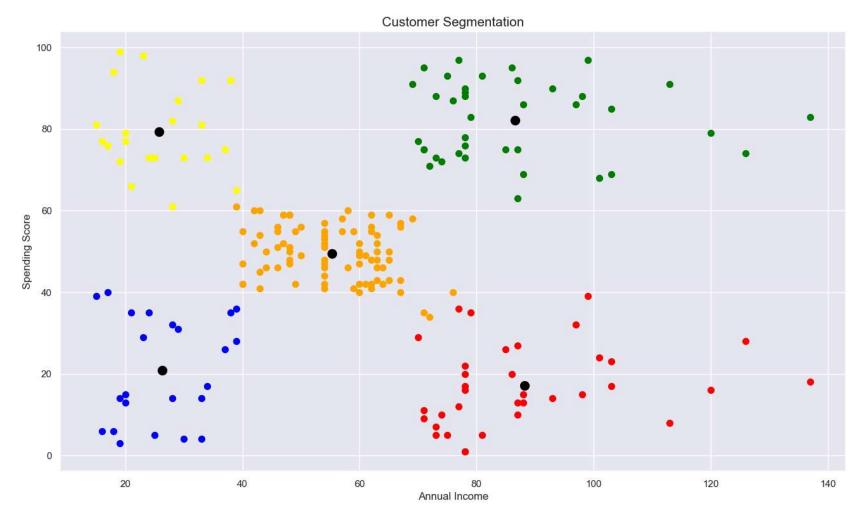
Plotting data in graph

```
In [31]: # plotting all the clusters and their centroids

plt.figure(figsize = (16,9))
plt.title("Customer Segmentation", fontsize = 15)
plt.scatter(X[Y==0,0], X[Y==0,1], s= 50, c = 'orange', label = 'Cluster 1')
plt.scatter(X[Y==1,0], X[Y==1,1], s= 50, c = 'green', label = 'Cluster 1')
plt.scatter(X[Y==2,0], X[Y==2,1], s= 50, c = 'red', label = 'Cluster 2')
plt.scatter(X[Y==3,0], X[Y==3,1], s= 50, c = 'blue', label = 'Cluster 3')
plt.scatter(X[Y==4,0], X[Y==4,1], s= 50, c = 'yellow', label = 'Cluster 4')

# plot the centroids
plt.xlabel("Annual Income")
plt.ylabel("Spending Score")
plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s =100, c='black', label = 'Centroids')
```

Out[31]: <matplotlib.collections.PathCollection at 0x20890cfe610>



In []: