```
In [1]: import numpy as np
          import pandas as pd
          {\color{red}\textbf{import}} \ \texttt{matplotlib.pyplot} \ {\color{red}\textbf{as}} \ \texttt{plt}
          import seaborn as sns
 In [2]: df = pd.read_csv("Mall_Customers.csv")
 In [3]: df.head()
                                        Annual Income (k$) Spending Score (1-100)
 Out[3]:
             CustomerID
                          Gender
                                   Age
          0
                                    19
                                                        15
                       1
                            Male
                                                                               39
          1
                       2
                                    21
                                                        15
                                                                               81
                            Male
          2
                                                        16
                                                                                6
                       3 Female
                                    20
          3
                          Female
                                    23
                                                        16
                                                                               77
                                                                               40
          4
                       5 Female
                                    31
                                                        17
 In [4]: df.shape
 Out[4]:
           (200, 5)
 In [5]:
         df.describe()
 Out[5]:
                  CustomerID
                                     Age Annual Income (k$) Spending Score (1-100)
          count
                  200.000000
                              200.000000
                                                  200.000000
                                                                         200.000000
           mean
                  100.500000
                               38.850000
                                                   60.560000
                                                                          50.200000
                   57.879185
                                                                          25.823522
                               13.969007
                                                   26.264721
             std
            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
                                                                          73.000000
            75%
                  150.250000
                               49.000000
                                                   78.000000
                  200.000000
                               70.000000
                                                  137.000000
                                                                          99.000000
            max
 In [6]: df.dtypes
Out[6]: CustomerID
                                         int64
           Gender
                                         object
                                          int64
           Annual Income (k$)
                                          int64
           Spending Score (1-100)
                                          int64
           dtype: object
 In [7]: df.isnull().sum()
 Out[7]:
          CustomerID
                                         0
                                         0
           Gender
           Age
                                         0
                                         0
           Annual Income (k$)
           Spending Score (1-100)
                                         0
           dtype: int64
 In [8]: df.drop(["CustomerID"],axis=1,inplace =True)
 In [9]: df.head()
 Out[9]:
             Gender Age
                           Annual Income (k$) Spending Score (1-100)
          0
                       19
                Male
                                           15
                                                                  39
          1
                Male
                       21
                                           15
                                                                  81
                                           16
                                                                   6
             Female
                       20
          3
                       23
                                           16
                                                                  77
              Female
                                           17
                                                                  40
              Female
In [10]: plt.figure(1, figsize=(15,6))
          for x in ['Age', 'Annual Income (k$)', 'Spending Score (1-100)']:
               n += 1
```

```
plt.subplot(1, 3, n)

plt.subplots_adjust(hspace=0.5, wspace = 0.5)

sns.distplot(df[x], bins = 20)

plt.title('Distplot of {}'.format(x))

plt.show()
```

C:\Users\srinu\AppData\Local\Temp\ipykernel 6904\1035435271.py:11: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see $\verb|https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751|$

sns.distplot(df[x], bins = 20)

 $\verb|C:\Users\srinu\appData\Local\Temp\ipykernel_6904\1035435271.py: 11: UserWarning: \\$

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For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(df[x], bins = 20)

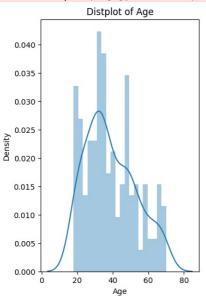
C:\Users\srinu\AppData\Local\Temp\ipykernel 6904\1035435271.py:11: UserWarning:

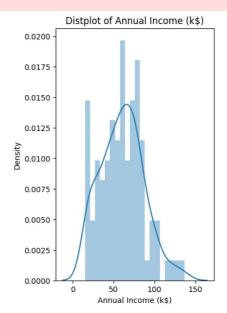
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

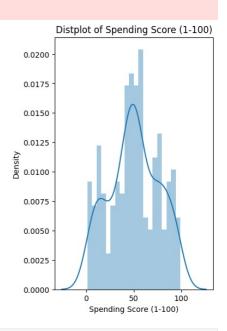
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

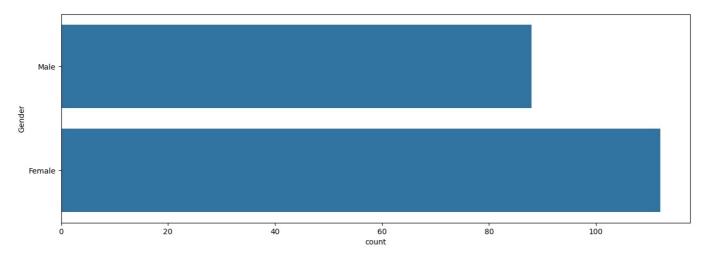
sns.distplot(df[x], bins = 20)







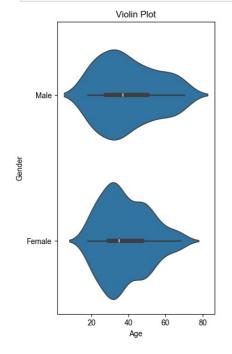
```
In [11]: plt.figure(figsize=(15,5))
    sns.countplot(y = 'Gender', data = df)
    plt.show()
```

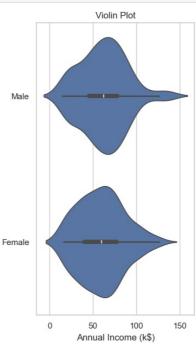


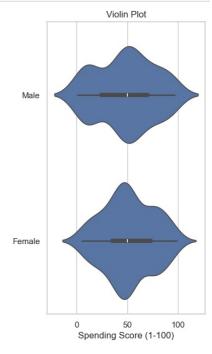
```
In [12]: plt.figure(1,figsize=(15,7))
    n=0

for cols in ['Age', 'Annual Income (k$)', 'Spending Score (1-100)']:
    n+=1
    plt.subplot(1, 3, n)
    sns.set(style="whitegrid")
    plt.subplots_adjust(hspace=0.5, wspace = 0.5)
    sns.violinplot(x = cols, y = 'Gender', data = df)
    plt.ylabel('Gender' if n == 1 else '')
    plt.title('Violin Plot')

plt.show()
```







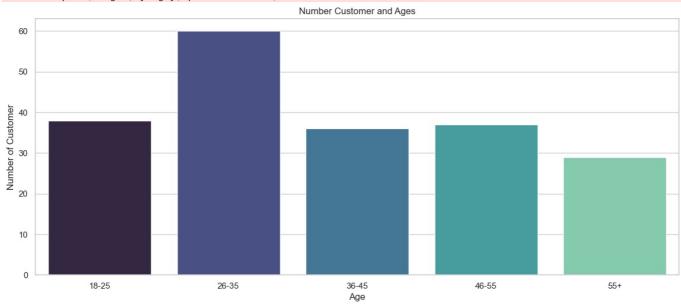
```
In [13]: age_18_25 = df.Age[(df.Age >= 18 ) & (df.Age <= 25 )]
    age_26_35 = df.Age[(df.Age >= 26 ) & (df.Age <= 35 )]
    age_36_45 = df.Age[(df.Age >= 36 ) & (df.Age <= 45 )]
    age_46_55 = df. Age[(df.Age >= 46 ) & (df.Age <= 55 )]
    age_55above = df.Age [(df.Age >= 56 ])
    agex = ["18-25", "26-35", "36-45", "46-55", "55+"]
    agey = [len(age_18_25.values), len(age_26_35.values), len(age_36_45.values), len(age_46_55.values), len(age_55aplt.figure(figsize=(15,6))
```

```
sns.barplot(x=agex, y=agey, palette="mako")
plt.title("Number Customer and Ages")
plt.xlabel("Age")
plt.ylabel("Number of Customer")
plt.show()
```

C:\Users\srinu\AppData\Local\Temp\ipykernel 6904\323071086.py:17: FutureWarning:

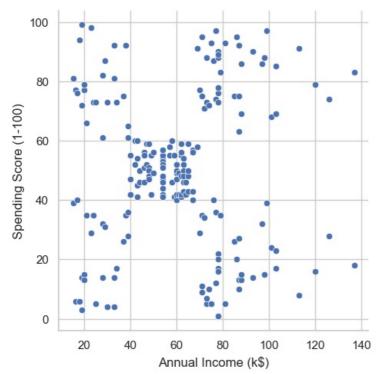
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=agex, y=agey, palette="mako")



In [14]: sns.relplot(x="Annual Income (k\$)" , y = "Spending Score (1-100)" , data= df)

Out[14]: <seaborn.axisgrid.FacetGrid at 0x14ef0610830>



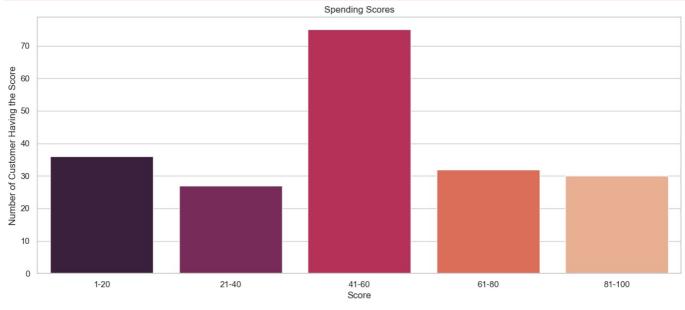
```
In [15]: ss_1_20 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 1) & (df["Spending Score (1-100)"] <= : ss_21_40 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 21) & (df["Spending Score (1-100)"] <= : ss_41_60 = df["Spending Score (1-100)"] [(df ["Spending Score (1-100)"] >= 41) & (df["Spending Score (1-100)"] <= : ss_61_80 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 61) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] [(df["Spending Score (1-100)"] >= 81) & (df["Spending Score (1-100)"] <= : ss_81_100 = df["Spending Score (1-100)"] <= : ss_8
```

```
ssx = ["1-20", "21-40", "41-60", "61-80", "81-100"]
ssy = [len(ss_1_20.values), len(ss_21_40.values), len(ss_41_60.values), len (ss_61_80.values), len (ss_81_100.values), len (ss_61_80.values), len (ss_61_80.value
```

 $\verb|C:\Users\srinu\AppData\Local\Temp\ipykernel_6904\2011223906.py:17: Future Warning: | AppData\Local\Temp\ipykernel_6904\2011223906.py:17: Future Warning: | AppData\Local\Temp\ipykernel_6904\2011233906.py:17: Future Warning: | AppData\Local\Temp\ipykernel_6904\2011233906.py:17: Future Warning: | AppData\Local\Temp\ipykernel_6904\2011233906.py:17: Future Warning$

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=ssx, y=ssy, palette="rocket")



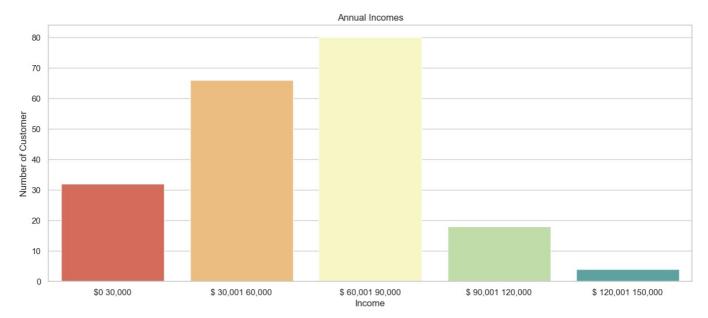
```
In [16]:
    ai10_30 = df["Annual Income (k$)"] [(df ["Annual Income (k$)"] >= 0) & (df ["Annual Income (k$)"] <= 30)]
    ai31_60 = df["Annual Income (k$)"] [(df["Annual Income (k$)"] >= 31) & (df ["Annual Income (k$)"] <= 60)]
    ai61_90 = df["Annual Income (k$)"] [(df["Annual Income (k$)"] >= 61) & (df["Annual Income (k$)"] <= 90)]
    ai91_120 = df["Annual Income (k$)"] [(df["Annual Income (k$)"] >= 91) & (df ["Annual Income (k$)"] <= 120)]
    ai121_150 = df["Annual Income (k$)"] [(df ["Annual Income (k$)"] >= 121) & (df ["Annual Income (k$)"] <= 150)]
    aix = ["$0 30,000", "$ 30,001 60,000", "$ 60,001 90,000", "$ 90,001 120,000", "$ 120,001 150,000"]
    aiy = [len(ai10_30.values), len(ai31_60.values), len (ai61_90.values), len(ai91_120.values), len(ai121_150.value)
    plt.figure(figsize=(15,6))
    sns.barplot(x=aix, y=aiy, palette="Spectral")
    plt.title("Annual Incomes")
    plt.xlabel("Income")
    plt.ylabel("Number of Customer")
    plt.show</pre>
```

 $\verb| C:\Users\srinu\AppData\Local\Temp\ipykernel_6904\4007312437.py:17: Future \verb| Warning: Part | P$

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=aix, y=aiy, palette="Spectral")

Out[16]: <function matplotlib.pyplot.show(close=None, block=None)>



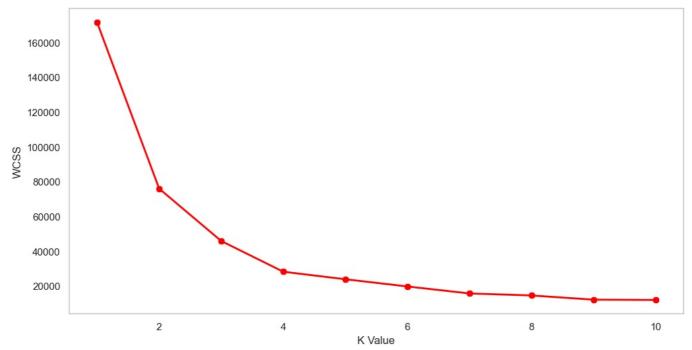
```
In [17]: X1=df.loc[:, ["Age", "Spending Score (1-100)"]].values
    from sklearn.cluster import KMeans

wcss = []
    for k in range(1,11):
        kmeans = KMeans (n_clusters=k, init="k-means++")
        kmeans.fit(X1)
        wcss.append(kmeans.inertia_)

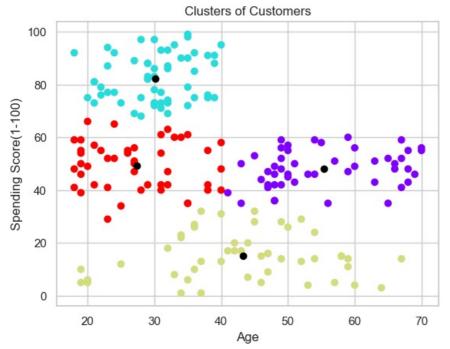
plt.figure(figsize=(12,6))

plt.grid()
plt.plot(range(1,11), wcss, linewidth=2, color="red", marker ="8")
plt.xlabel("K Value")
plt.ylabel("WCSS")

plt.show()
```



```
In [18]: kmeans = KMeans(n_clusters = 4)
label = kmeans.fit_predict(X1)
print(label)
```



 $\begin{bmatrix} 3 & 1 & 2 & 1 & 3 & 1 & 2 & 1 & 2 & 1 & 2 & 1 & 2 & 1 & 2 & 1 & 2 & 1 & 2 & 1 & 3 & 3 & 2 & 1 & 3 & 1 & 2$

```
In [21]: X2=df.loc[:, ["Annual Income (k$)", "Spending Score (1-100)"]].values
    from sklearn.cluster import KMeans

wcss = []

for k in range(1,11):
    kmeans = KMeans (n_clusters=k, init="k-means++")

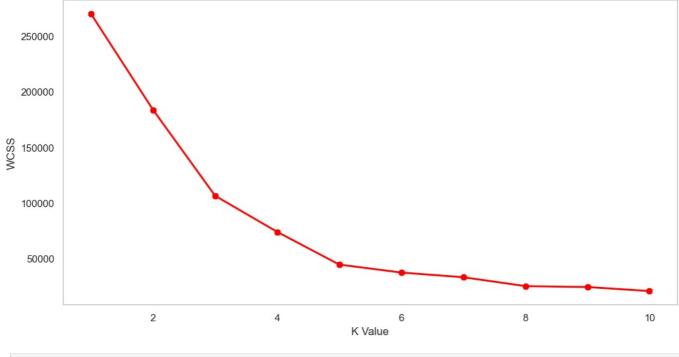
    kmeans.fit(X2)

    wcss.append(kmeans.inertia_)

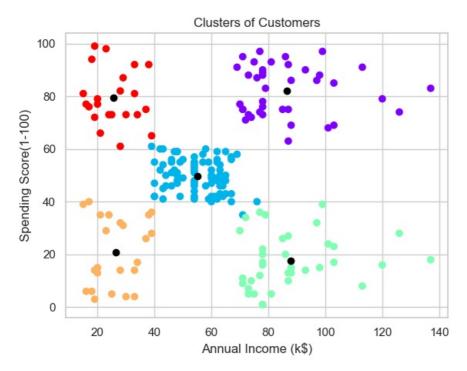
plt.figure(figsize=(12,6))

plt.grid()
plt.plot(range(1,11), wcss, linewidth=2, color="red", marker ="8")
plt.xlabel("K Value")
plt.ylabel("WCSS")

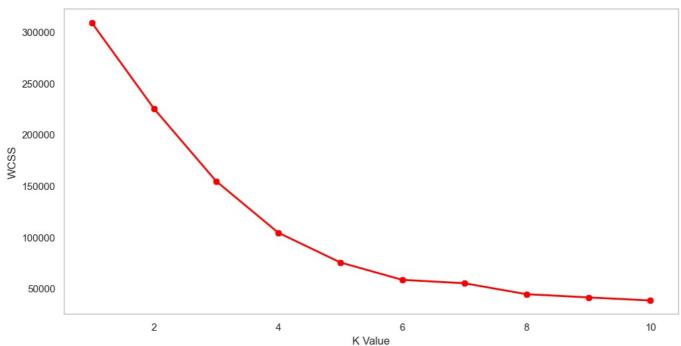
plt.show()
```



```
In [22]: kmeans = KMeans(n_clusters = 5)
                                                     label = kmeans.fit_predict(X2)
                                                     print(label)
                                                [ 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 \; 4 \; 3 
                                                   \begin{smallmatrix} 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 & 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 3
In [23]: print(kmeans.cluster_centers_)
                                               [[86.53846154 82.12820513]
                                                     [55.0875
                                                                                                                            49.7125
                                                     [87.75
                                                                                                                              17.58333333]
                                                     [26.30434783 20.91304348]
                                                     [25.72727273 79.36363636]]
In [24]: plt.scatter (X2[:,0], X1[:,1], c=kmeans.labels_, cmap='rainbow')
                                                     plt.scatter(kmeans.cluster_centers_[:,0],kmeans.cluster_centers_[:,1], color = 'black')
                                                     plt.title('Clusters of Customers')
                                                     plt.xlabel('Annual Income (k$)')
                                                     plt.ylabel('Spending Score(1-100)')
                                                     plt.show()
```



```
In [25]: X3=df.iloc[:,1:]
    wcss = []
    for k in range(1,11):
        kmeans = KMeans (n_clusters=k, init="k-means++")
        kmeans.fit(X3)
        wcss.append(kmeans.inertia_)
    plt.figure(figsize=(12,6))
    plt.grid()
    plt.plot(range (1,11), wcss, linewidth=2, color="red", marker ="8")
    plt.xlabel("K Value")
    plt.ylabel("WCSS")
    plt.show()
```

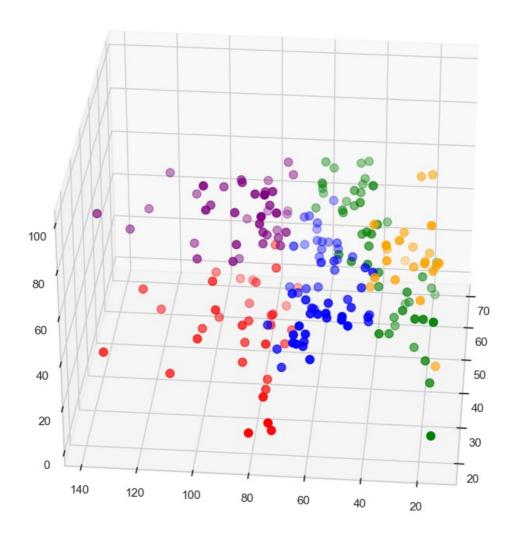


```
In [26]: kmeans = KMeans(n_clusters = 5)
label = kmeans.fit_predict(X3)
print(label)
```

```
\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 3
In [27]: print(kmeans.cluster_centers_)
                          [[45.2173913 26.30434783 20.91304348]
                             [32.69230769 86.53846154 82.12820513]
                             [40.66666667 87.75
                                                                                                            17.583333331
                             [24.96
                                                                      28.04
                                                                                                            77.
                             [43.72727273 55.48051948 49.32467532]]
In [28]: clusters = kmeans.fit predict(X3)
                             df["label"] = clusters
                             from mpl toolkits.mplot3d import Axes3D
                             fig = plt.figure(figsize=(20,10))
                             ax = fig.add subplot(111, projection='3d')
                             ax.scatter(df.Age[df.label==0], df["Annual Income (k$)"][df.label==0], df["Spending Score (1-100)"] [df.label==0]
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ax.scatter(df.Age[df.label== 1], df["Annual Income (k\$)"] [df.label ==1], df["Spending Score (1-100)"] [df.label ax.scatter(df.Age[df.label== 2], df["Annual Income (k\$)"] [df.label== 2], df["Spending Score (1-100)"] [df.label ax.scatter(df.Age [df.label ==3], df["Annual Income (k\$)"] [df.label== 3], df["Spending Score (1-100)"] [df.label ax.scatter(df.Age[df.label== 4], df["Annual Income (k\$)"] [df.label== 4], df["Spending Score (1-100)"] [df.label== 4]

 $[0\ 3\ 0\$



ax.view_init(30, 185)

In []:

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