

**Assignment -4**  
Python Programming

Assignment Date	13 October 2022
Student Name	Mr. Lavan R P
Student Roll Number	910619104045
Maximum Marks	2 Marks

## Customer Segmentation Analysis

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [1]:

### Importing Dataset

```
df = pd.read_csv('Mall_Customers.csv')

df.head()
```

In [6]:

In [8]:

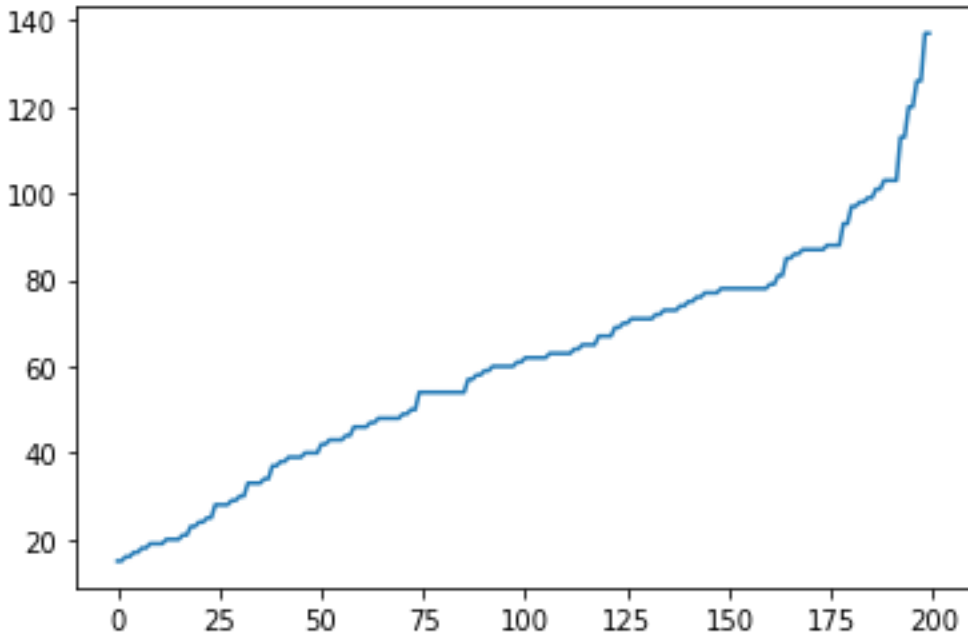
Out[8]:

	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

### 1. Univariate Analysis

```
plt.plot(df['Annual Income (k$)'])
plt.show()
```

In [11]:

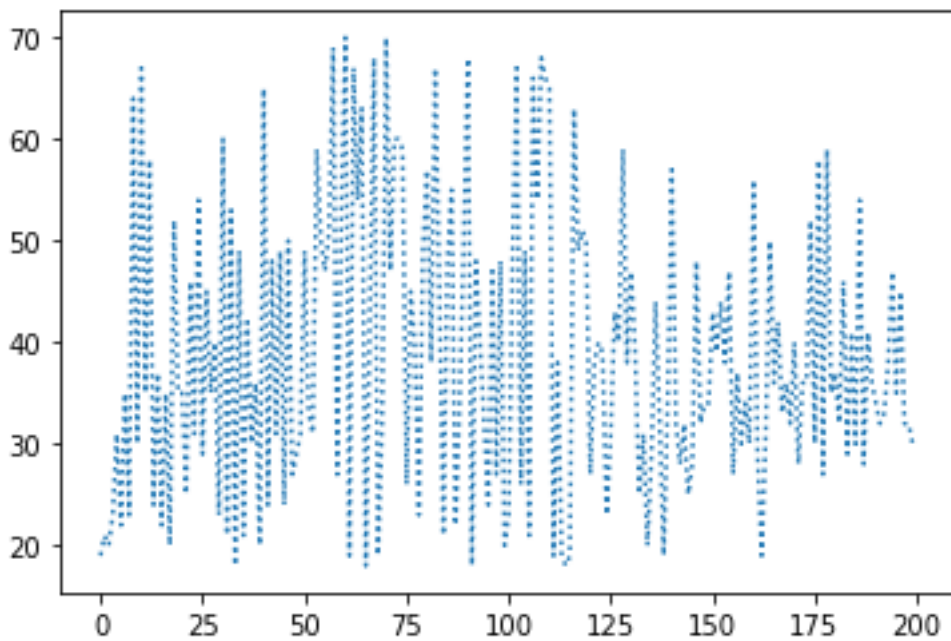


In [13]:

```
data=np.array(df['Age'])
plt.plot(data,linestyle='dotted')
```

Out[13]:

```
[<matplotlib.lines.Line2D at 0x173eeaf9ac0>]
```



In [14]:

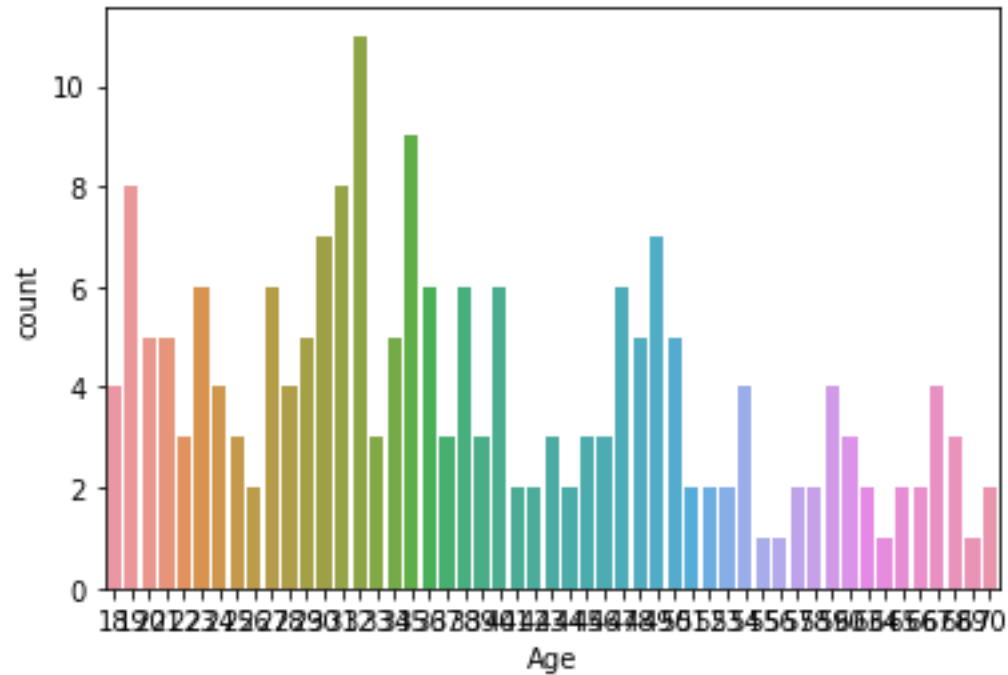
```
sns.countplot(df['Age'])
C:\Users\Ilyas\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
FutureWarning: Pass the following variable as a keyword arg: x. From version
0.12, the only valid positional argument will be `data`, and passing other
```

arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

```
<AxesSubplot:xlabel='Age', ylabel='count'>
```

Out[14]:

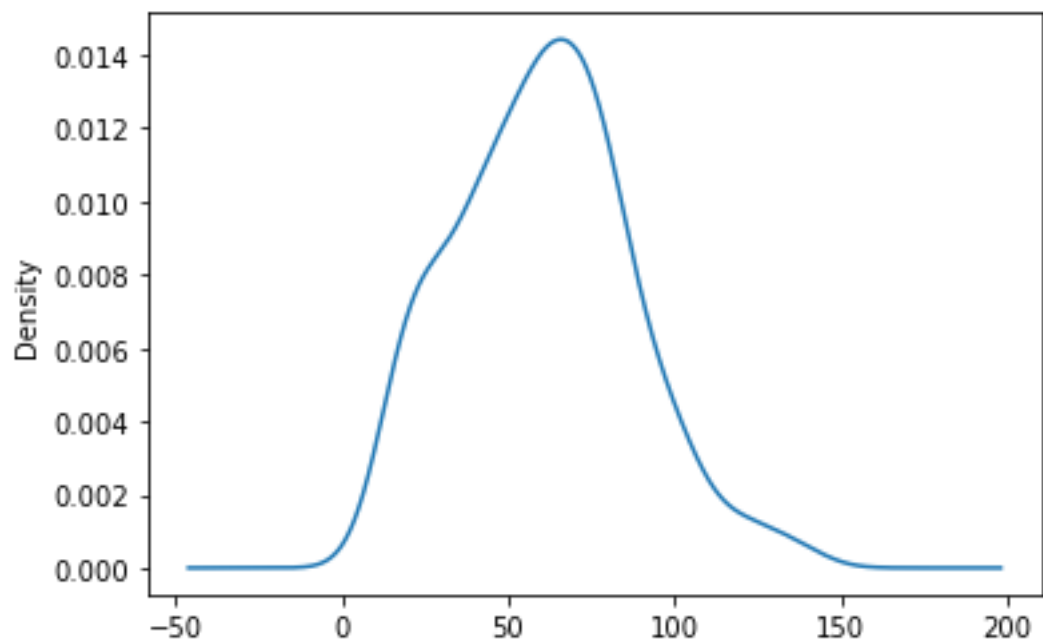


```
df['Annual Income (k$)'].plot(kind='density')
```

In [19]:

```
<AxesSubplot:ylabel='Density'>
```

Out[19]:



In [20]:

```
sns.countplot(df['Gender'])
```

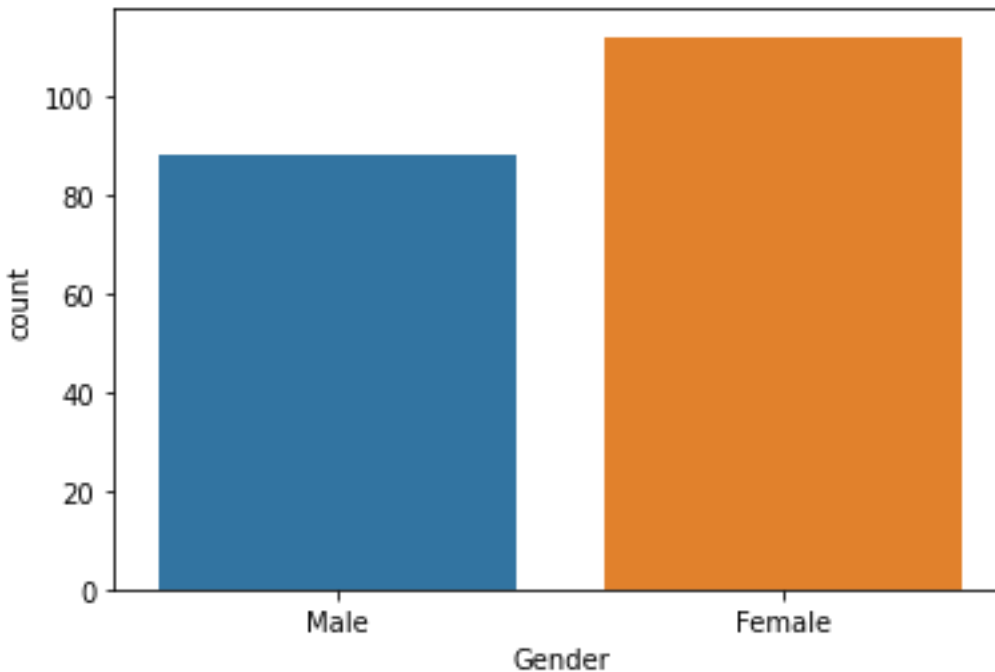
```
C:\Users\Ilyas\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
```

```
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(
```

Out[20]:

```
<AxesSubplot:xlabel='Gender', ylabel='count'>
```



In [21]:

```
sns.boxplot(df['Annual Income (k$)'])
```

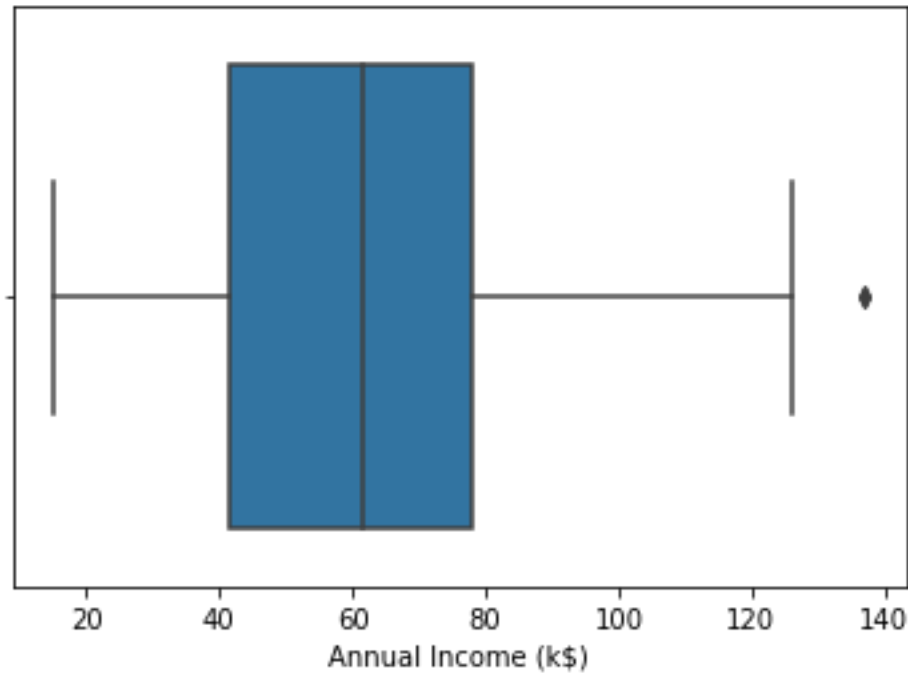
```
C:\Users\Ilyas\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
```

```
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(
```

Out[21]:

```
<AxesSubplot:xlabel='Annual Income (k$)'>
```

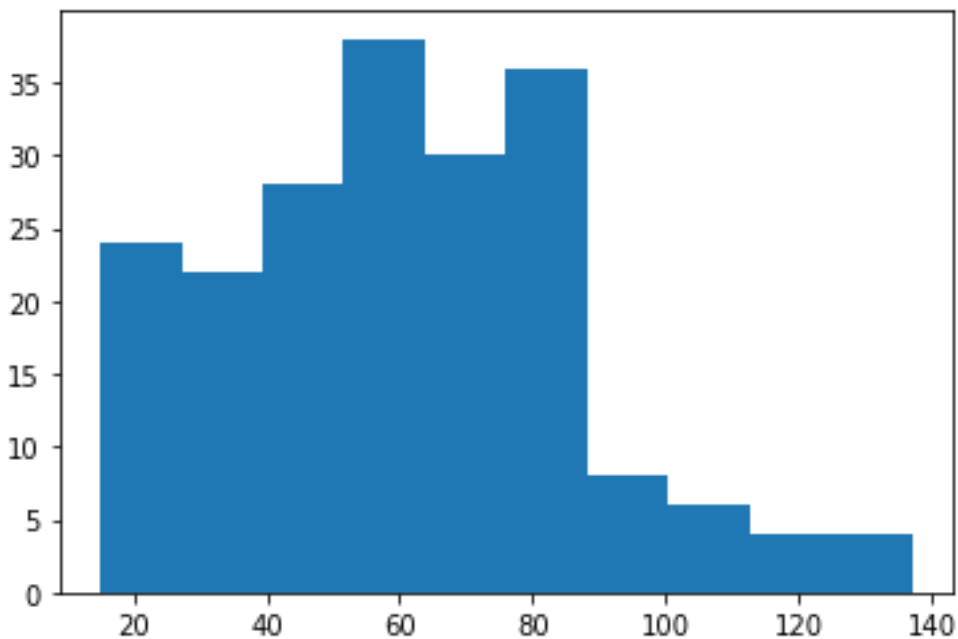


In [22]:

```
plt.hist(df['Annual Income (k$)'])
```

Out[22]:

```
(array([24., 22., 28., 38., 30., 36., 8., 6., 4., 4.]),
 array([ 15. , 27.2, 39.4, 51.6, 63.8, 76. , 88.2, 100.4, 112.6,
        124.8, 137. ]),
 <BarContainer object of 10 artists>)
```



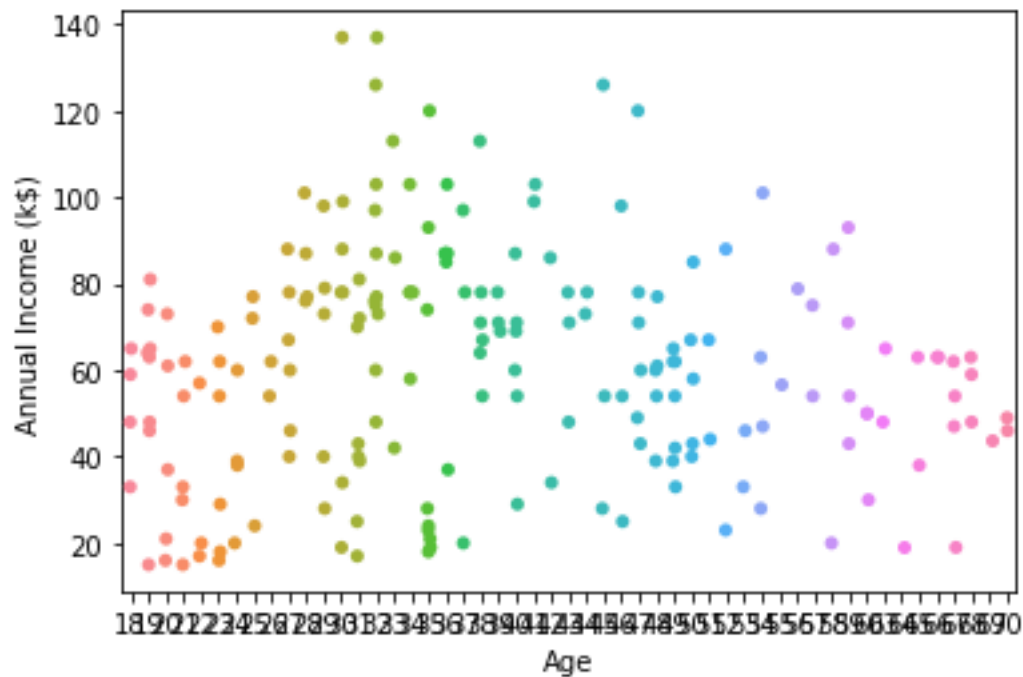
## 2. Bi-Variate Analysis

In [23]:

```
sns.stripplot(x=df['Age'],y=df['Annual Income (k$)'])
```

Out[23]:

```
<AxesSubplot:xlabel='Age', ylabel='Annual Income (k$) '>
```

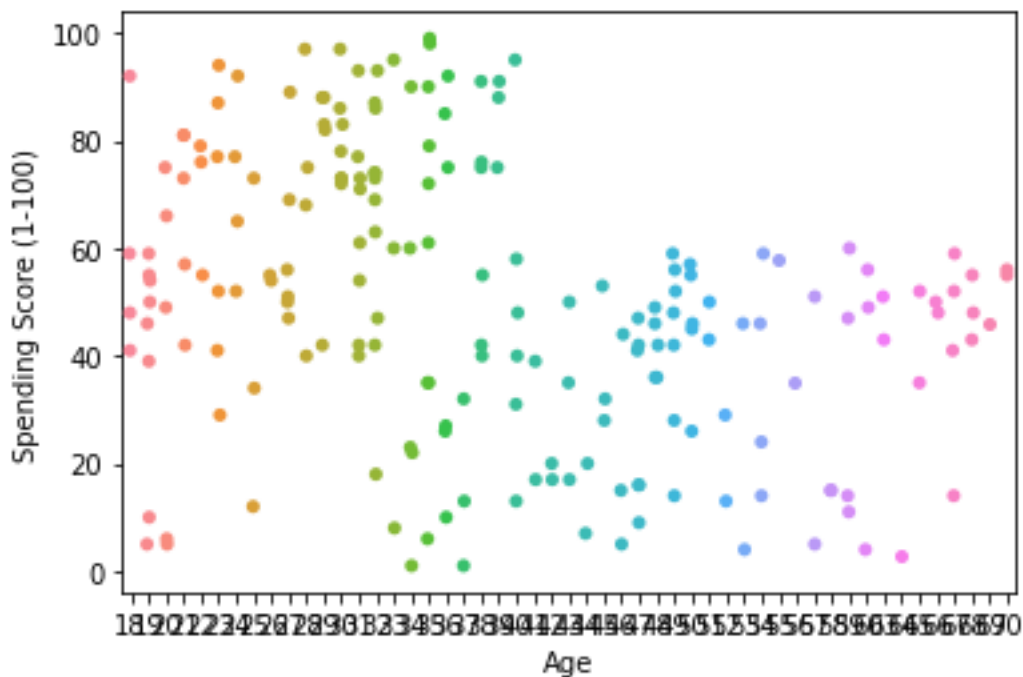


In [24]:

```
sns.stripplot(x=df['Age'],y=df['Spending Score (1-100)'])
```

Out[24]:

```
<AxesSubplot:xlabel='Age', ylabel='Spending Score (1-100) '>
```

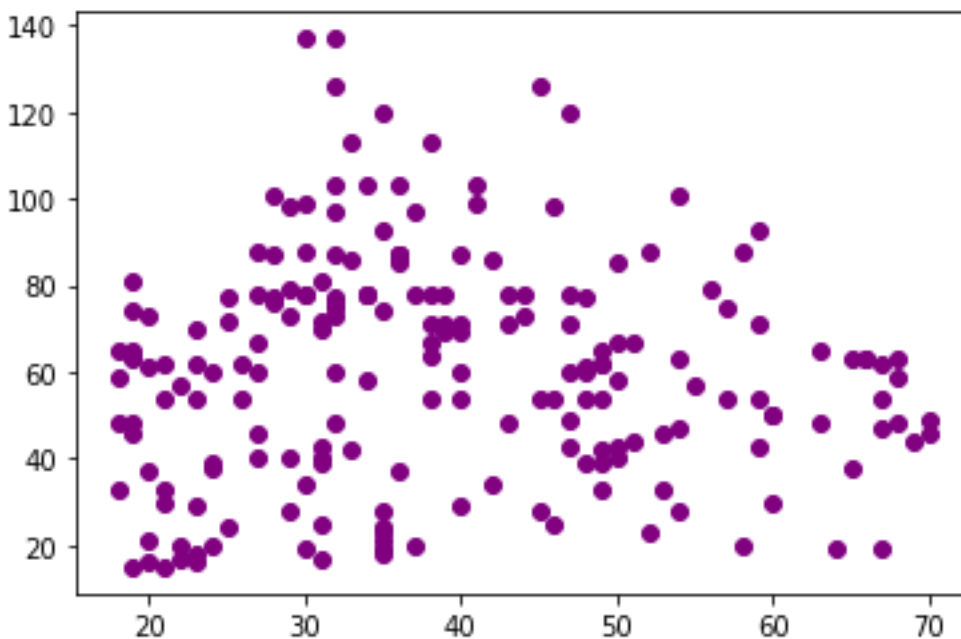


In [26]:

```
plt.scatter(df['Age'],df['Annual Income (k$)'],color='purple')
```

Out[26]:

```
<matplotlib.collections.PathCollection at 0x173f184f4c0>
```

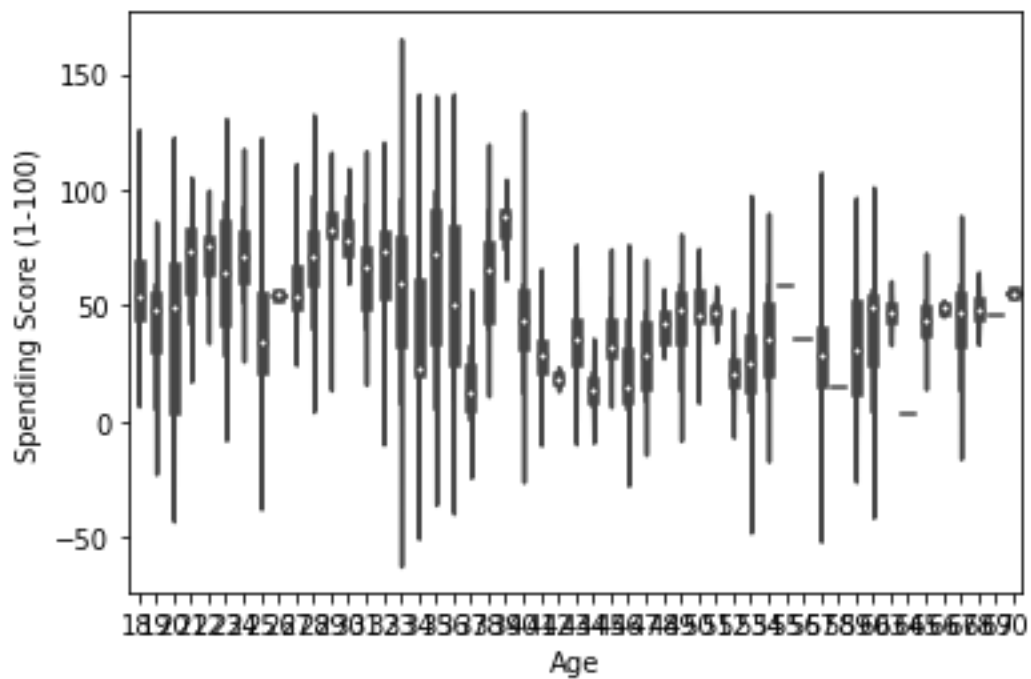


```
sns.violinplot(x='Age',y='Spending Score (1-100)',data=df)
```

In [27]:

Out[27]:

```
<AxesSubplot:xlabel='Age', ylabel='Spending Score (1-100) '>
```



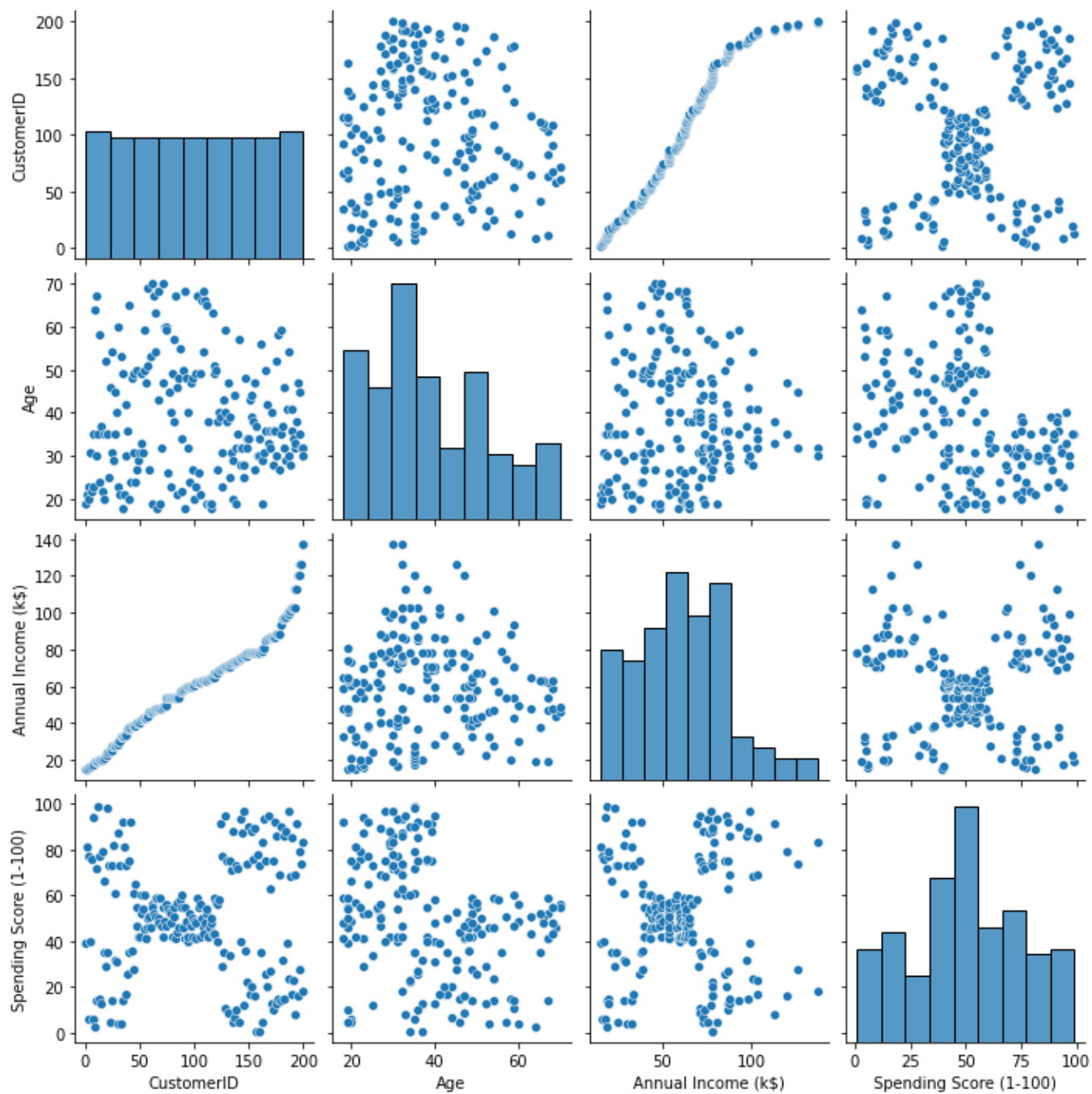
### 3. Multi-Variate Analysis

```
sns.pairplot(df)
```

In [28]:

```
<seaborn.axisgrid.PairGrid at 0x173f167d430>
```

Out[28]:





## 4. Discriptive Statistics

```
sns.heatmap(df.corr(),annot=True)
```

In [30]:

Out[30]:

<AxesSubplot:>



```
df.shape
```

In [31]:

```
(200, 5)
```

Out[31]:

```
df.isnull().sum()
```

In [32]:

Out[32]:

```
CustomerID      0
Gender          0
Age            0
Annual Income (k$)  0
Spending Score (1-100)  0
dtype: int64
```

```
df.info()
```

In [33]:

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   CustomerID                            200 non-null    int64
1   Gender                                200 non-null    object
2   Age                                    200 non-null    int64
3   Annual Income (k$)                    200 non-null    int64
4   Spending Score (1-100)                 200 non-null    int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB

```

In [34]:

```
df.describe()
```

Out[34]:

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

In [35]:

```

df.mean()
C:\Users\Ilyas\AppData\Local\Temp\ipykernel_10368\3698961737.py:1:
FutureWarning: Dropping of nuisance columns in DataFrame reductions (with
'numeric_only=None') is deprecated; in a future version this will raise
TypeError. Select only valid columns before calling the reduction.
df.mean()

```

Out[35]:

```

CustomerID          100.50
Age                  38.85
Annual Income (k$)   60.56
Spending Score (1-100)  50.20
dtype: float64

```

In [36]:

```

df.median()
C:\Users\Ilyas\AppData\Local\Temp\ipykernel_10368\530051474.py:1:
FutureWarning: Dropping of nuisance columns in DataFrame reductions (with
'numeric_only=None') is deprecated; in a future version this will raise
TypeError. Select only valid columns before calling the reduction.
df.median()

```

```
CustomerID          100.5
Age                 36.0
Annual Income (k$)  61.5
Spending Score (1-100) 50.0
dtype: float64
```

Out[36]:

```
df.mode()
```

In [37]:

Out[37]:

	CustomerID	Gender	Age	Annual Income (k\$)	Spending Score (1-100)
<b>0</b>	1	Female	32.0	54.0	42.0
<b>1</b>	2	NaN	NaN	78.0	NaN
<b>2</b>	3	NaN	NaN	NaN	NaN
<b>3</b>	4	NaN	NaN	NaN	NaN
<b>4</b>	5	NaN	NaN	NaN	NaN
...	...	...	...	...	...
<b>195</b>	196	NaN	NaN	NaN	NaN
<b>196</b>	197	NaN	NaN	NaN	NaN
<b>197</b>	198	NaN	NaN	NaN	NaN
<b>198</b>	199	NaN	NaN	NaN	NaN
<b>199</b>	200	NaN	NaN	NaN	NaN

200 rows × 5 columns

```
df['Gender'].value_counts()
```

In [38]:

```
Female    112
Male       88
Name: Gender, dtype: int64
```

Out[38]:

## 5. Missing Values

```
df.isna().sum()
```

In [39]:

```
CustomerID    0
Gender        0
Age           0
Annual Income (k$)  0
Spending Score (1-100)  0
dtype: int64
```

Out[39]:

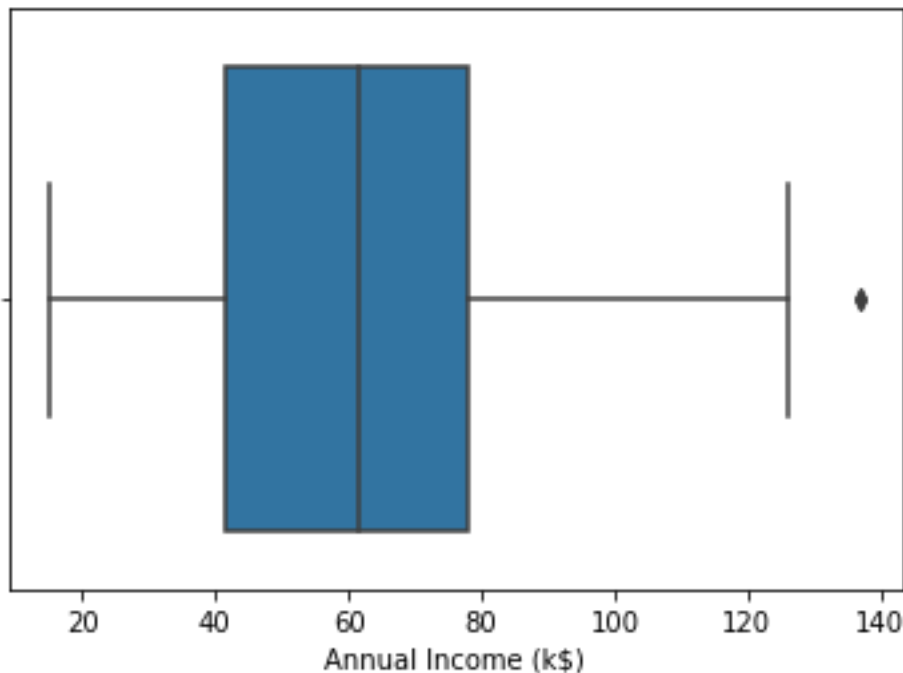
## 6. Handling outliers

In [40]:

```
sns.boxplot(df['Annual Income (k$)'])
C:\Users\Ilyas\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(
```

Out[40]:

```
<AxesSubplot:xlabel='Annual Income (k$) '>
```



In [41]:

```
Q1 = df['Annual Income (k$)'].quantile(0.25)
Q3 = df['Annual Income (k$)'].quantile(0.75)
IQR = Q3 - Q1

whisker_width = 1.5
lower_whisker = Q1 - (whisker_width*IQR)
upper_whisker = Q3 + (whisker_width*IQR)
df['Annual Income (k$)'] = np.where(df['Annual Income (k$)'] > upper_whisker,
upper_whisker, np.where(df['Annual Income (k$)'] <
lower_whisker, lower_whisker, df['Annual Income (k$)']))
```

In [43]:

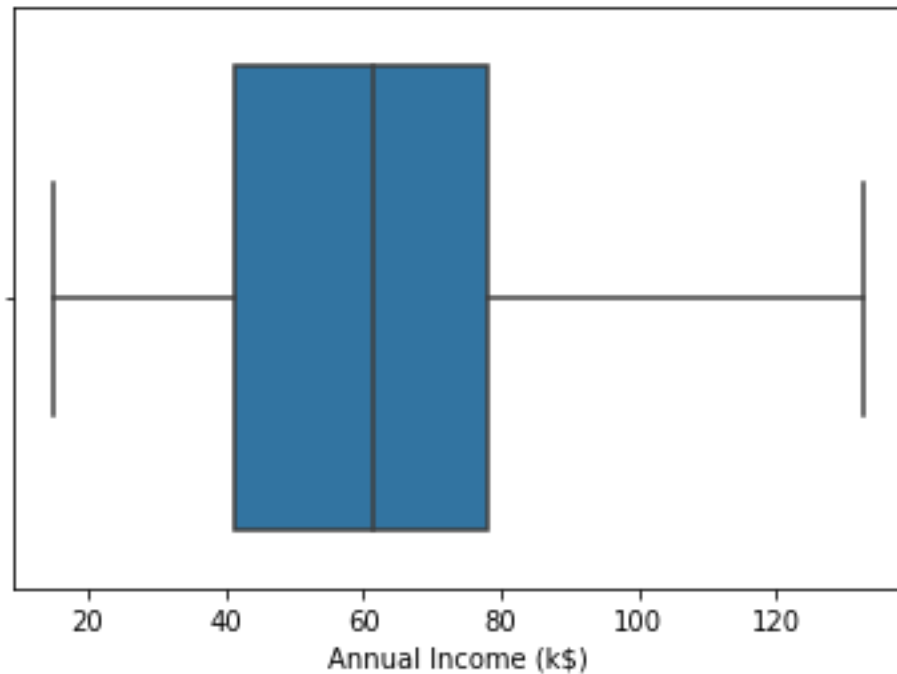
```
sns.boxplot(df['Annual Income (k$)'])
C:\Users\Ilyas\anaconda3\lib\site-packages\seaborn\_decorators.py:36:
FutureWarning: Pass the following variable as a keyword arg: x. From version
```

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(
```

Out[43]:

```
<AxesSubplot:xlabel='Annual Income (k$) '>
```



## 7. Encoding Categorical Values

```
numeric_data = df.select_dtypes(include = [np.number])  
categorical_data = df.select_dtypes(exclude = [np.number])
```

In [44]:

```
print("The number of numerical variables: ", numeric_data.shape[1])  
print("The number of categorical variables: ", categorical_data.shape[1])  
The number of numerical variables: 4  
The number of categorical variables: 1
```

In [45]:

```
print("The number of categorical variables: ", categorical_data.shape[1])
```

```
Categorical_variables = list(categorical_data.columns)  
Categorical_variables  
The number of categorical variables: 1
```

Out[45]:

```
['Gender']
```

In [47]:

```
df['Gender'].value_counts()
```

```
Female    112
Male       88
Name: Gender, dtype: int64
```

Out[47]:

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
label = le.fit_transform(df['Gender'])
df["Gender"] = label
```

In [50]:

```
df['Gender'].value_counts()
```

In [51]:

```
0    112
1     88
Name: Gender, dtype: int64
```

Out[51]:

## 8. Scaling the datas

```
X = df.drop("Age", axis=1)
Y = df['Age']
```

In [52]:

```
from sklearn.preprocessing import StandardScaler
object= StandardScaler()
scale = object.fit_transform(X)
print(scale)
[[-1.7234121  1.12815215 -1.74542941 -0.43480148]
 [-1.70609137  1.12815215 -1.74542941  1.19570407]
 [-1.68877065 -0.88640526 -1.70708307 -1.71591298]
 [-1.67144992 -0.88640526 -1.70708307  1.04041783]
 [-1.6541292  -0.88640526 -1.66873673 -0.39597992]
 [-1.63680847 -0.88640526 -1.66873673  1.00159627]
 [-1.61948775 -0.88640526 -1.6303904  -1.71591298]
 [-1.60216702 -0.88640526 -1.6303904  1.70038436]
 [-1.5848463  1.12815215 -1.59204406 -1.83237767]
 [-1.56752558 -0.88640526 -1.59204406  0.84631002]
 [-1.55020485  1.12815215 -1.59204406 -1.4053405 ]
 [-1.53288413 -0.88640526 -1.59204406  1.89449216]
 [-1.5155634  -0.88640526 -1.55369772 -1.36651894]
 [-1.49824268 -0.88640526 -1.55369772  1.04041783]
 [-1.48092195  1.12815215 -1.55369772 -1.44416206]
 [-1.46360123  1.12815215 -1.55369772  1.11806095]
 [-1.4462805  -0.88640526 -1.51535138 -0.59008772]
 [-1.42895978  1.12815215 -1.51535138  0.61338066]
 [-1.41163905  1.12815215 -1.43865871 -0.82301709]
 [-1.39431833 -0.88640526 -1.43865871  1.8556706 ]]
```

In [53]:

[-1.3769976 1.12815215 -1.40031237 -0.59008772]  
[-1.35967688 1.12815215 -1.40031237 0.88513158]  
[-1.34235616 -0.88640526 -1.36196603 -1.75473454]  
[-1.32503543 1.12815215 -1.36196603 0.88513158]  
[-1.30771471 -0.88640526 -1.24692702 -1.4053405 ]  
[-1.29039398 1.12815215 -1.24692702 1.23452563]  
[-1.27307326 -0.88640526 -1.24692702 -0.7065524 ]  
[-1.25575253 1.12815215 -1.24692702 0.41927286]  
[-1.23843181 -0.88640526 -1.20858069 -0.74537397]  
[-1.22111108 -0.88640526 -1.20858069 1.42863343]  
[-1.20379036 1.12815215 -1.17023435 -1.7935561 ]  
[-1.18646963 -0.88640526 -1.17023435 0.88513158]  
[-1.16914891 1.12815215 -1.05519534 -1.7935561 ]  
[-1.15182818 1.12815215 -1.05519534 1.62274124]  
[-1.13450746 -0.88640526 -1.05519534 -1.4053405 ]  
[-1.11718674 -0.88640526 -1.05519534 1.19570407]  
[-1.09986601 -0.88640526 -1.016849 -1.28887582]  
[-1.08254529 -0.88640526 -1.016849 0.88513158]  
[-1.06522456 -0.88640526 -0.90180999 -0.93948177]  
[-1.04790384 -0.88640526 -0.90180999 0.96277471]  
[-1.03058311 -0.88640526 -0.86346365 -0.59008772]  
[-1.01326239 1.12815215 -0.86346365 1.62274124]  
[-0.99594166 1.12815215 -0.82511731 -0.55126616]  
[-0.97862094 -0.88640526 -0.82511731 0.41927286]  
[-0.96130021 -0.88640526 -0.82511731 -0.86183865]  
[-0.94397949 -0.88640526 -0.82511731 0.5745591 ]  
[-0.92665877 -0.88640526 -0.78677098 0.18634349]  
[-0.90933804 -0.88640526 -0.78677098 -0.12422899]  
[-0.89201732 -0.88640526 -0.78677098 -0.3183368 ]  
[-0.87469659 -0.88640526 -0.78677098 -0.3183368 ]  
[-0.85737587 -0.88640526 -0.7100783 0.06987881]  
[-0.84005514 1.12815215 -0.7100783 0.38045129]  
[-0.82273442 -0.88640526 -0.67173196 0.14752193]  
[-0.80541369 1.12815215 -0.67173196 0.38045129]  
[-0.78809297 -0.88640526 -0.67173196 -0.20187212]  
[-0.77077224 1.12815215 -0.67173196 -0.35715836]  
[-0.75345152 -0.88640526 -0.63338563 -0.00776431]  
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[-0.71881007 -0.88640526 -0.55669295 0.03105725]  
[-0.70148935 1.12815215 -0.55669295 -0.16305055]  
[-0.68416862 1.12815215 -0.55669295 0.22516505]  
[-0.6668479 1.12815215 -0.55669295 0.18634349]  
[-0.64952717 -0.88640526 -0.51834661 0.06987881]  
[-0.63220645 -0.88640526 -0.51834661 0.34162973]  
[-0.61488572 1.12815215 -0.48000028 0.03105725]

[-0.597565 1.12815215 -0.48000028 0.34162973]  
[-0.58024427 -0.88640526 -0.48000028 -0.00776431]  
[-0.56292355 -0.88640526 -0.48000028 -0.08540743]  
[-0.54560282 1.12815215 -0.48000028 0.34162973]  
[-0.5282821 -0.88640526 -0.48000028 -0.12422899]  
[-0.51096138 1.12815215 -0.44165394 0.18634349]  
[-0.49364065 -0.88640526 -0.44165394 -0.3183368 ]  
[-0.47631993 -0.88640526 -0.4033076 -0.04658587]  
[-0.4589992 -0.88640526 -0.4033076 0.22516505]  
[-0.44167848 1.12815215 -0.24992225 -0.12422899]  
[-0.42435775 1.12815215 -0.24992225 0.14752193]  
[-0.40703703 -0.88640526 -0.24992225 0.10870037]  
[-0.3897163 1.12815215 -0.24992225 -0.08540743]  
[-0.37239558 -0.88640526 -0.24992225 0.06987881]  
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[-0.33775413 1.12815215 -0.24992225 0.03105725]  
[-0.3204334 1.12815215 -0.24992225 0.18634349]  
[-0.30311268 1.12815215 -0.24992225 -0.35715836]  
[-0.28579196 -0.88640526 -0.24992225 -0.24069368]  
[-0.26847123 -0.88640526 -0.24992225 0.26398661]  
[-0.25115051 1.12815215 -0.24992225 -0.16305055]  
[-0.23382978 -0.88640526 -0.13488324 0.30280817]  
[-0.21650906 -0.88640526 -0.13488324 0.18634349]  
[-0.19918833 -0.88640526 -0.0965369 0.38045129]  
[-0.18186761 -0.88640526 -0.0965369 -0.16305055]  
[-0.16454688 -0.88640526 -0.05819057 0.18634349]  
[-0.14722616 1.12815215 -0.05819057 -0.35715836]  
[-0.12990543 1.12815215 -0.01984423 -0.04658587]  
[-0.11258471 -0.88640526 -0.01984423 -0.39597992]  
[-0.09526399 -0.88640526 -0.01984423 -0.3183368 ]  
[-0.07794326 1.12815215 -0.01984423 0.06987881]  
[-0.06062254 -0.88640526 -0.01984423 -0.12422899]  
[-0.04330181 -0.88640526 -0.01984423 -0.00776431]  
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[ 0.00866036 -0.88640526 0.05684845 -0.35715836]  
[ 0.02598109 -0.88640526 0.05684845 -0.08540743]  
[ 0.04330181 1.12815215 0.05684845 0.34162973]  
[ 0.06062254 1.12815215 0.05684845 0.18634349]  
[ 0.07794326 1.12815215 0.05684845 0.22516505]  
[ 0.09526399 -0.88640526 0.05684845 -0.3183368 ]  
[ 0.11258471 -0.88640526 0.09519478 -0.00776431]  
[ 0.12990543 1.12815215 0.09519478 -0.16305055]  
[ 0.14722616 1.12815215 0.09519478 -0.27951524]  
[ 0.16454688 1.12815215 0.09519478 -0.08540743]



[ 0.18186761 1.12815215 0.09519478 0.06987881]  
[ 0.19918833 -0.88640526 0.09519478 0.14752193]  
[ 0.21650906 -0.88640526 0.13354112 -0.3183368 ]  
[ 0.23382978 1.12815215 0.13354112 -0.16305055]  
[ 0.25115051 -0.88640526 0.17188746 -0.08540743]  
[ 0.26847123 -0.88640526 0.17188746 -0.00776431]  
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[ 0.33775413 -0.88640526 0.24858013 0.26398661]  
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[ 0.3897163 -0.88640526 0.32527281 0.30280817]  
[ 0.40703703 1.12815215 0.32527281 1.58391968]  
[ 0.42435775 -0.88640526 0.36361914 -0.82301709]  
[ 0.44167848 -0.88640526 0.36361914 1.04041783]  
[ 0.4589992 1.12815215 0.40196548 -0.59008772]  
[ 0.47631993 1.12815215 0.40196548 1.73920592]  
[ 0.49364065 1.12815215 0.40196548 -1.52180518]  
[ 0.51096138 1.12815215 0.40196548 0.96277471]  
[ 0.5282821 1.12815215 0.40196548 -1.5994483 ]  
[ 0.54560282 1.12815215 0.40196548 0.96277471]  
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[ 0.58024427 -0.88640526 0.44031182 0.80748846]  
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[ 0.61488572 -0.88640526 0.47865816 1.46745499]  
[ 0.63220645 -0.88640526 0.47865816 -1.67709142]  
[ 0.64952717 1.12815215 0.47865816 0.88513158]  
[ 0.6668479 1.12815215 0.51700449 -1.56062674]  
[ 0.68416862 -0.88640526 0.51700449 0.84631002]  
[ 0.70148935 -0.88640526 0.55535083 -1.75473454]  
[ 0.71881007 1.12815215 0.55535083 1.6615628 ]  
[ 0.73613079 -0.88640526 0.59369717 -0.39597992]  
[ 0.75345152 -0.88640526 0.59369717 1.42863343]  
[ 0.77077224 1.12815215 0.6320435 -1.48298362]  
[ 0.78809297 1.12815215 0.6320435 1.81684904]  
[ 0.80541369 1.12815215 0.6320435 -0.55126616]  
[ 0.82273442 -0.88640526 0.6320435 0.92395314]  
[ 0.84005514 -0.88640526 0.67038984 -1.09476801]  
[ 0.85737587 1.12815215 0.67038984 1.54509812]  
[ 0.87469659 1.12815215 0.67038984 -1.28887582]  
[ 0.89201732 1.12815215 0.67038984 1.46745499]  
[ 0.90933804 -0.88640526 0.67038984 -1.17241113]  
[ 0.92665877 -0.88640526 0.67038984 1.00159627]  
[ 0.94397949 -0.88640526 0.67038984 -1.32769738]

```
[ 0.96130021 -0.88640526 0.67038984 1.50627656]
[ 0.97862094 1.12815215 0.67038984 -1.91002079]
[ 0.99594166 -0.88640526 0.67038984 1.07923939]
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[ 1.03058311 -0.88640526 0.67038984 0.88513158]
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[ 1.06522456 -0.88640526 0.70873618 1.27334719]
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[ 1.15182818 1.12815215 0.97716054 -1.17241113]
[ 1.16914891 -0.88640526 0.97716054 1.73920592]
[ 1.18646963 -0.88640526 1.01550688 -0.90066021]
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[ 1.22111108 1.12815215 1.01550688 -1.44416206]
[ 1.23843181 1.12815215 1.01550688 0.96277471]
[ 1.25575253 1.12815215 1.01550688 -1.56062674]
[ 1.27307326 1.12815215 1.01550688 1.62274124]
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[ 1.30771471 -0.88640526 1.05385321 1.38981187]
[ 1.32503543 1.12815215 1.05385321 -1.36651894]
[ 1.34235616 1.12815215 1.05385321 0.72984534]
[ 1.35967688 1.12815215 1.2455849 -1.4053405 ]
[ 1.3769976 1.12815215 1.2455849 1.54509812]
[ 1.39431833 -0.88640526 1.39897025 -0.7065524 ]
[ 1.41163905 -0.88640526 1.39897025 1.38981187]
[ 1.42895978 1.12815215 1.43731659 -1.36651894]
[ 1.4462805 -0.88640526 1.43731659 1.46745499]
[ 1.46360123 -0.88640526 1.47566292 -0.43480148]
[ 1.48092195 1.12815215 1.47566292 1.81684904]
[ 1.49824268 -0.88640526 1.5523556 -1.01712489]
[ 1.5155634 1.12815215 1.5523556 0.69102378]
[ 1.53288413 -0.88640526 1.62904827 -1.28887582]
[ 1.55020485 -0.88640526 1.62904827 1.35099031]
[ 1.56752558 -0.88640526 1.62904827 -1.05594645]
[ 1.5848463 -0.88640526 1.62904827 0.72984534]
[ 1.60216702 1.12815215 2.01251165 -1.63826986]
[ 1.61948775 -0.88640526 2.01251165 1.58391968]
[ 1.63680847 -0.88640526 2.28093601 -1.32769738]
[ 1.6541292 -0.88640526 2.28093601 1.11806095]
[ 1.67144992 -0.88640526 2.51101403 -0.86183865]
[ 1.68877065 1.12815215 2.51101403 0.92395314]
[ 1.70609137 1.12815215 2.76985181 -1.25005425]
[ 1.7234121 1.12815215 2.76985181 1.27334719]]
```

```
X_scaled = pd.DataFrame(scale, columns = X.columns)
X_scaled
```

In [54]:

Out[54]:

	CustomerID	Gender	Annual Income (k\$)	Spending Score (1-100)
0	-1.723412	1.128152	-1.745429	-0.434801
1	-1.706091	1.128152	-1.745429	1.195704
2	-1.688771	-0.886405	-1.707083	-1.715913
3	-1.671450	-0.886405	-1.707083	1.040418
4	-1.654129	-0.886405	-1.668737	-0.395980
...	...	...	...	...
195	1.654129	-0.886405	2.280936	1.118061
196	1.671450	-0.886405	2.511014	-0.861839
197	1.688771	1.128152	2.511014	0.923953
198	1.706091	1.128152	2.769852	-1.250054
199	1.723412	1.128152	2.769852	1.273347

200 rows × 4 columns

## Train test split

```
from sklearn.model_selection import train_test_split
```

In [55]:

### Split the dataset

```
X_train, X_test, Y_train, Y_test = train_test_split(X_scaled, Y, test_size
=0.20, random_state = 0)
X_train.shape
```

In [56]:

```
(160, 4)
```

Out[56]:

```
X_test.shape
```

In [57]:

```
(40, 4)
```

Out[57]:

```
Y_train.shape
```

In [58]:

```
(160,)
```

Out[58]:

```
Y_test.shape
```

In [59]:

```
(40,)
```

Out[59]:

## Clustering Algorithms

In [61]:

```
x = df.iloc[:, [3, 4]].values
```

In [62]:

```
from sklearn.cluster import KMeans
```

```
wcss_list= []
```

```
for i in range(1, 11):
```

```
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state= 42)
```

```
    kmeans.fit(x)
```

```
    wcss_list.append(kmeans.inertia_)
```

```
plt.plot(range(1, 11), wcss_list)
```

```
plt.title('The Elbow Method Graph')
```

```
plt.xlabel('Number of clusters(k)')
```

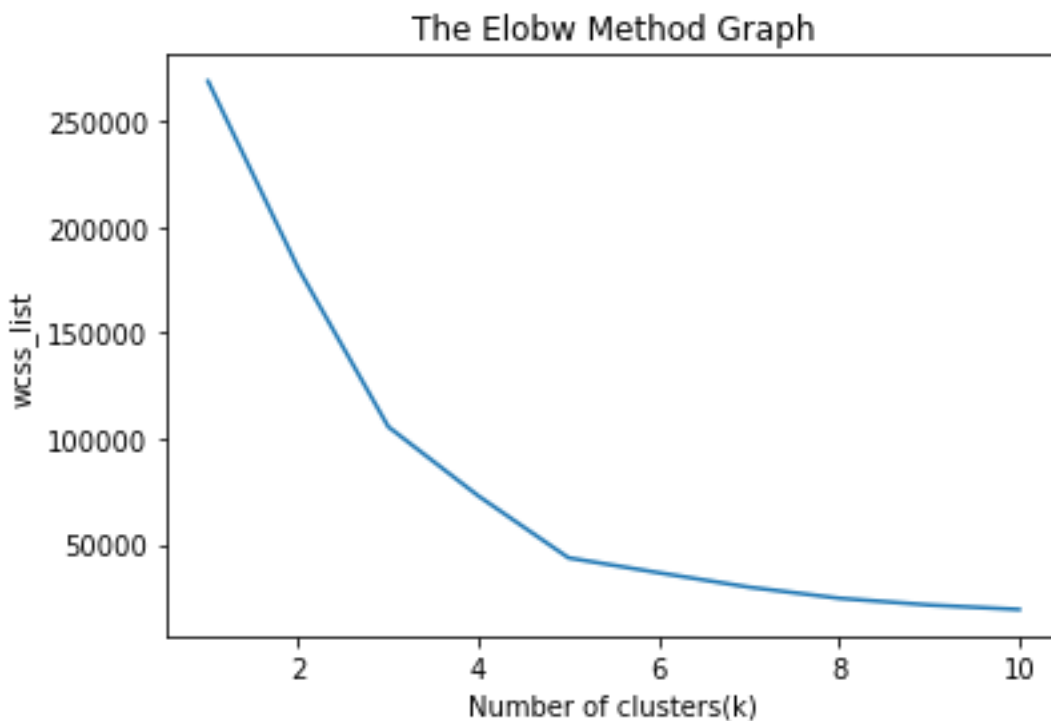
```
plt.ylabel('wcss_list')
```

```
plt.show()
```

```
C:\Users\Ilyas\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1036:
```

```
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 [63]:

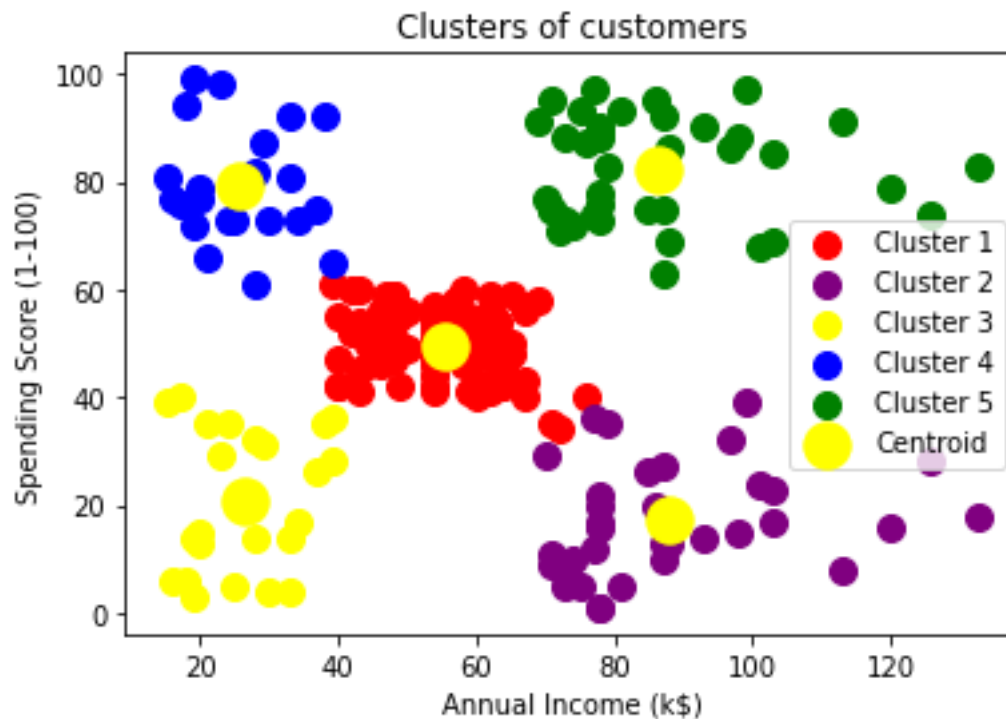
```
kmeans = KMeans(n_clusters=5, init = 'k-means++', random_state = 42)
```

```
y_predict = kmeans.fit_predict(x)
```

In [64]:

```
# Visualizing the clusters
```

```
plt.scatter(x[y_predict == 0, 0], x[y_predict == 0, 1], s = 100, c = 'red',  
label = 'Cluster 1') #for first cluster  
plt.scatter(x[y_predict == 1, 0], x[y_predict == 1, 1], s = 100, c =  
'purple', label = 'Cluster 2') #for second cluster  
plt.scatter(x[y_predict == 2, 0], x[y_predict == 2, 1], s = 100, c = 'yellow',  
label = 'Cluster 3') #for third cluster  
plt.scatter(x[y_predict == 3, 0], x[y_predict == 3, 1], s = 100, c = 'blue',  
label = 'Cluster 4') #for fourth cluster  
plt.scatter(x[y_predict == 4, 0], x[y_predict == 4, 1], s = 100, c = 'green',  
label = 'Cluster 5') #for fifth cluster  
plt.scatter(kmeans.cluster_centers[:, 0], kmeans.cluster_centers[:, 1], s =  
300, c = 'yellow', label = 'Centroid')  
plt.title('Clusters of customers')  
plt.xlabel('Annual Income (k$)')  
plt.ylabel('Spending Score (1-100)')  
plt.legend()  
plt.show()
```



In [ ]: