

Assignment -4

| | |
|---------------------|-----------------|
| Assignment Date | 18 October 2022 |
| Student Name | Kailash N |
| Student Roll Number | 811519104047 |
| Maximum Marks | 2 Marks |

```
[1]: import pandas as pd
      from matplotlib import pyplot as plt
      %matplotlib inline
      import seaborn as sns
      from sklearn.experimental import enable_iterative_imputer
      from sklearn.impute import IterativeImputer
      import pickle
      import numpy as np
      from sklearn.preprocessing import StandardScaler
```

```
[2]: df=pd.read_csv(r"C:\Users\Admin\Downloads\Mall_Customers.csv")
```

```
[3]: df.head()
```

```
[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 |

```
[4]: df.shape
```

```
[4]: (200, 5)
```

Checking the Null Values

```
[5]: df.isnull().sum()
```

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

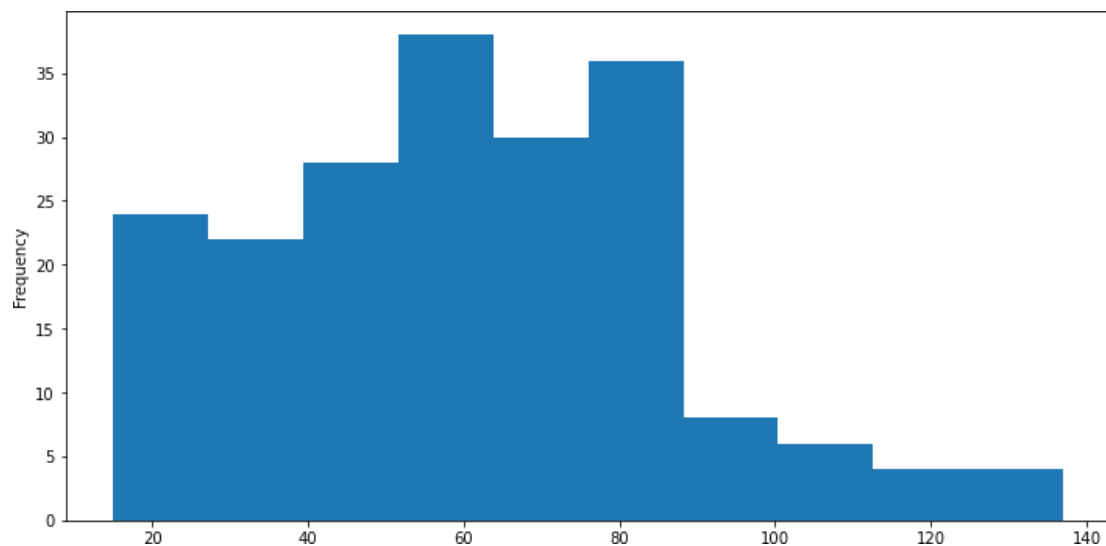
Renaming The Columns

```
[6]: df.columns=["CustomerID", "Gender", "Age", "Annual_Income", "Spending_Score"]
```

Univariant Plot

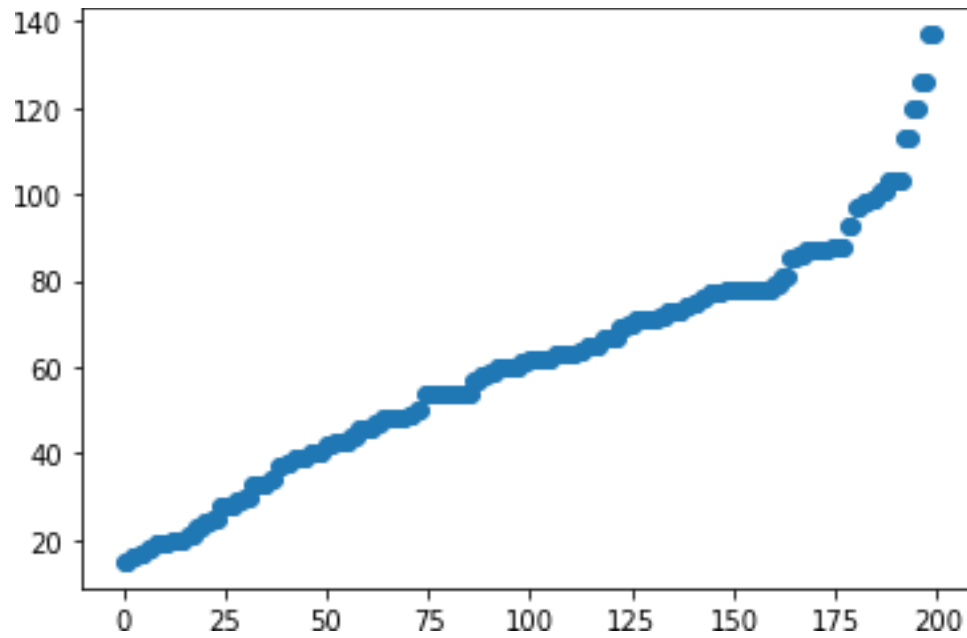
```
[7]: plt.figure()  
df.Annual_Income.plot(kind='hist', figsize=(12,6))
```

```
[7]: <AxesSubplot:ylabel='Frequency'>
```



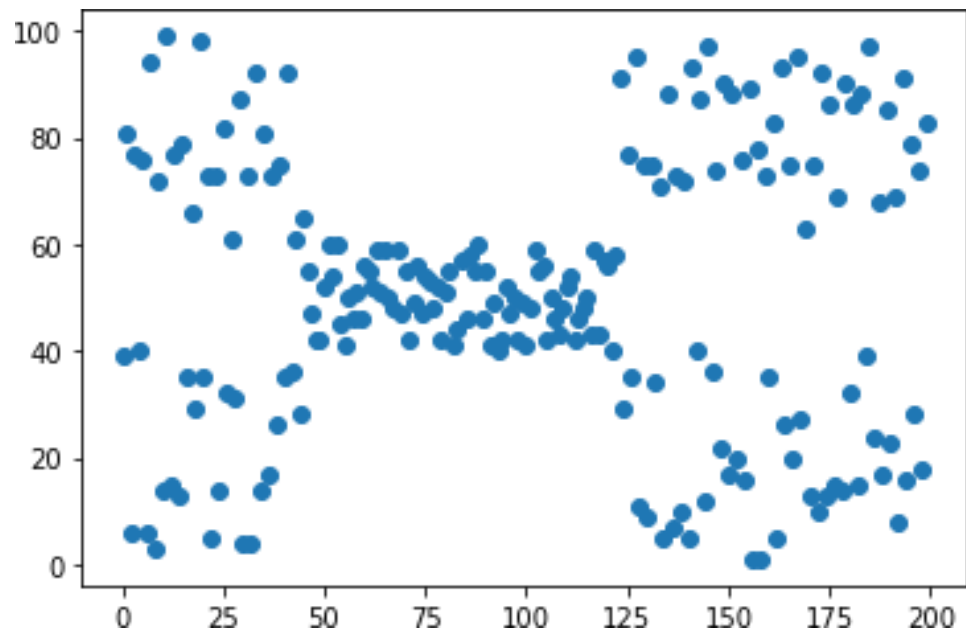
```
[8]: plt.scatter(df.index, df["Annual_Income"])
```

```
[8]: <matplotlib.collections.PathCollection at 0x15b1393a340>
```



```
[9]: plt.scatter(df.index,df['Spending_Score'])
```

```
[9]: <matplotlib.collections.PathCollection at 0x15b139bb7f0>
```



```
[10]: # z score computation
outliers=[]
def detect_outliers(data):
    threshold=3
    mean=np.mean(data)
    std=np.std(data)

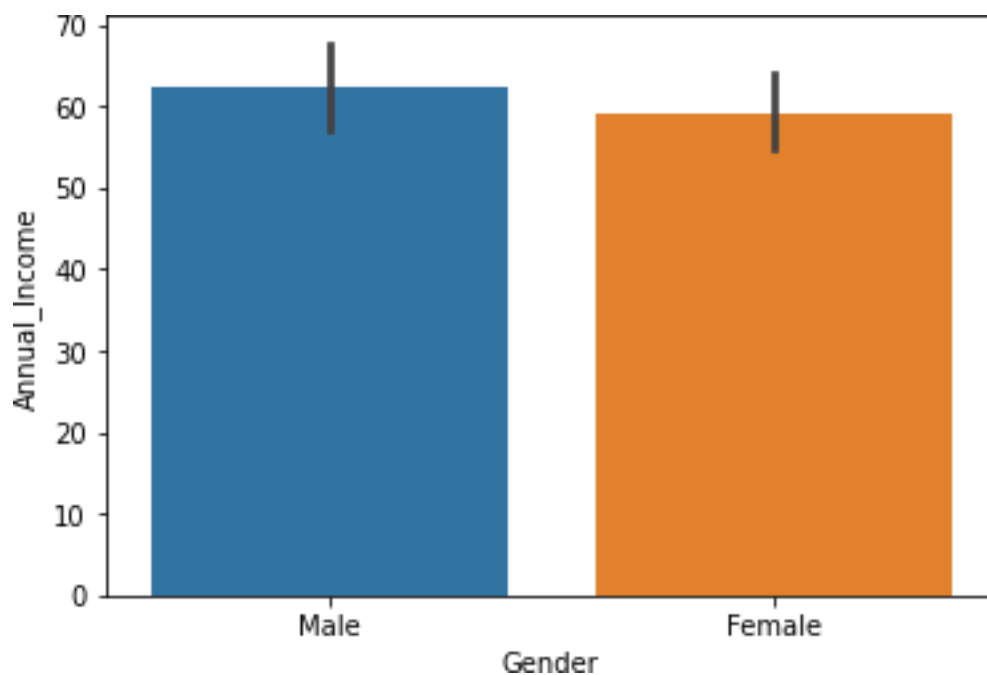
    for i in data:
        z_score=(i-mean)/std
        if np.abs(z_score)>threshold:
            outliers.append(i)

    return outliers
```

Bivariant Plot

```
[11]: sns.barplot(x='Gender',y='Annual_Income',data=df)
```

```
[11]: <AxesSubplot:xlabel='Gender', ylabel='Annual_Income'>
```



```
[12]: df
```

```
[12]:
```

| | CustomerID | Gender | Age | Annual_Income | Spending_Score |
|---|------------|--------|-----|---------------|----------------|
| 0 | 1 | Male | 19 | 15 | 39 |
| 1 | 2 | Male | 21 | 15 | 81 |

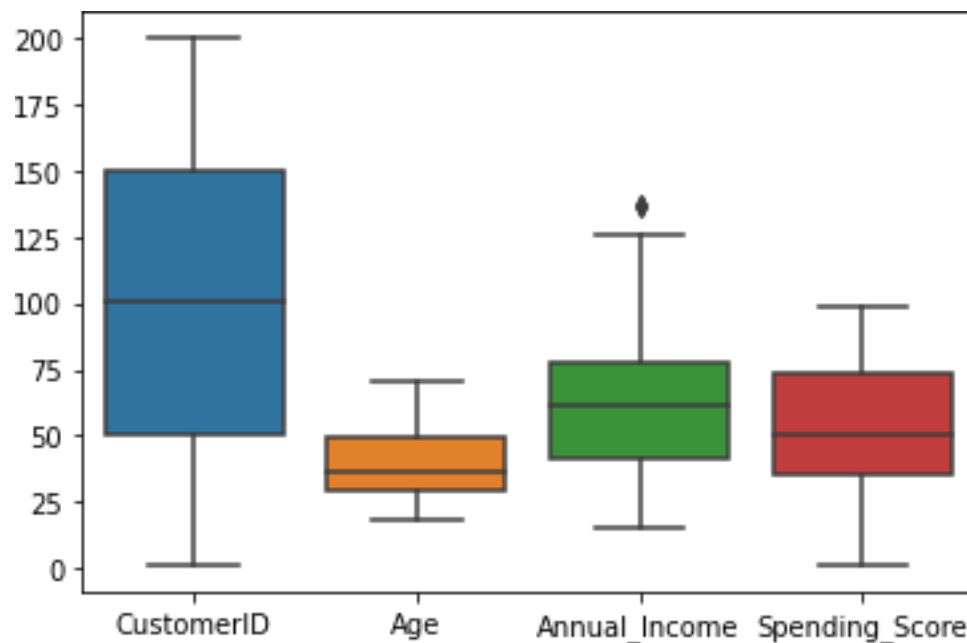
| | | | | | |
|-----|-----|--------|-----|-----|-----|
| 2 | 3 | Female | 20 | 16 | 6 |
| 3 | 4 | Female | 23 | 16 | 77 |
| 4 | 5 | Female | 31 | 17 | 40 |
| ... | ... | ... | ... | ... | ... |
| 195 | 196 | Female | 35 | 120 | 79 |
| 196 | 197 | Female | 45 | 126 | 28 |
| 197 | 198 | Male | 32 | 126 | 74 |
| 198 | 199 | Male | 32 | 137 | 18 |
| 199 | 200 | Male | 30 | 137 | 83 |

[200 rows x 5 columns]

Checking Outliers

```
[13]: sns.boxplot(data=df)
```

[13]: <AxesSubplot:>



Description About the Dataset

[14]:

```
df.describe()
```

```
[14]:
```

| | CustomerID | Age | Annual_Income | Spending_Score |
|-------|------------|------------|---------------|----------------|
| 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 |

[15]: df.info()

```
<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   200 non-null   int64
4   Spending_Score  200 non-null   int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB
Label Encoding for Gender
```

[16]: df["Gender"] = df["Gender"].replace(["Male", "Female"], [0, 1])

[17]: df.head()

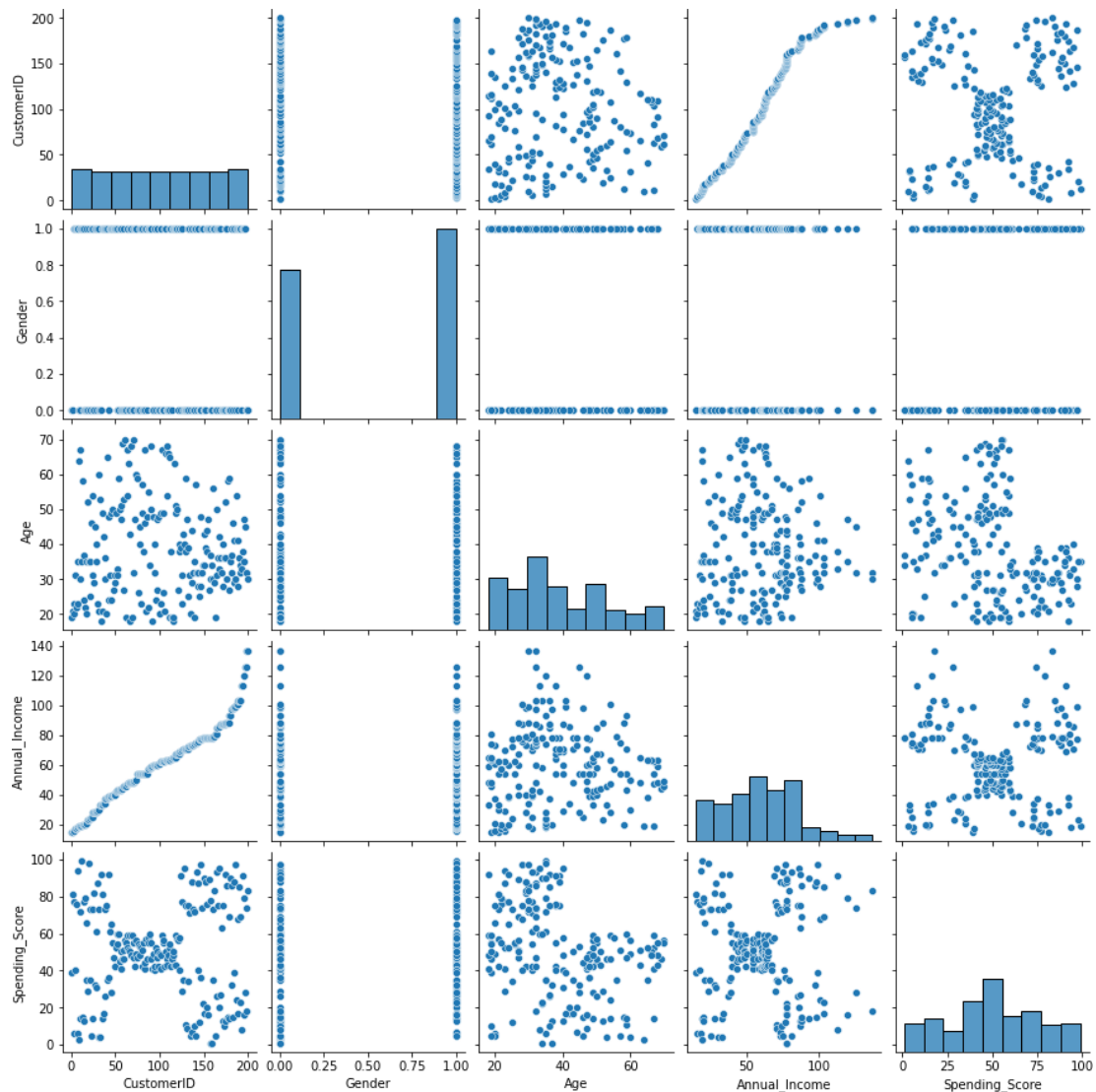
```
[17]:   CustomerID  Gender  Age  Annual_Income  Spending_Score
0           1       0   19           15           39
1           2       0   21           15           81
2           3       1   20           16            6
3           4       1   23           16           77
4           5       1   31           17           40
```

[18]: df.info()

```
<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   int64
2   Age             200 non-null   int64
3   Annual_Income   200 non-null   int64
4   Spending_Score  200 non-null   int64
dtypes: int64(5)
memory usage: 7.9 KB
Multivariant Plot
```

```
[19]: sns.pairplot(df)
```

```
[19]: <seaborn.axisgrid.PairGrid at 0x15b13bb63a0>
```



Scaling The Data

```
[20]: scaler = StandardScaler()  
scaler.fit(df)
```

```
[20]: StandardScaler()
```

```
[21]: print(scaler.transform(df))
```

```
[[-1.7234121 -1.12815215 -1.42456879 -1.73899919 -0.43480148]
```


[-1.70609137 -1.12815215 -1.28103541 -1.73899919 1.19570407]
 [-1.68877065 0.88640526 -1.3528021 -1.70082976 -1.71591298]
 [-1.67144992 0.88640526 -1.13750203 -1.70082976 1.04041783]
 [-1.6541292 0.88640526 -0.56336851 -1.66266033 -0.39597992]
 [-1.63680847 0.88640526 -1.20926872 -1.66266033 1.00159627]
 [-1.61948775 0.88640526 -0.27630176 -1.62449091 -1.71591298]
 [-1.60216702 0.88640526 -1.13750203 -1.62449091 1.70038436]
 [-1.5848463 -1.12815215 1.80493225 -1.58632148 -1.83237767]
 [-1.56752558 0.88640526 -0.6351352 -1.58632148 0.84631002]
 [-1.55020485 -1.12815215 2.02023231 -1.58632148 -1.4053405]
 [-1.53288413 0.88640526 -0.27630176 -1.58632148 1.89449216]
 [-1.5155634 0.88640526 1.37433211 -1.54815205 -1.36651894]
 [-1.49824268 0.88640526 -1.06573534 -1.54815205 1.04041783]
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 [-1.46360123 -1.12815215 -1.20926872 -1.54815205 1.11806095]
 [-1.4462805 0.88640526 -0.27630176 -1.50998262 -0.59008772]
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 [-1.29039398 -1.12815215 -0.70690189 -1.24279661 1.23452563]
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 [-1.20379036 -1.12815215 1.51786549 -1.16645776 -1.7935561]
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 [-1.15182818 -1.12815215 -1.49633548 -1.05194947 1.62274124]
 [-1.13450746 0.88640526 0.7284319 -1.05194947 -1.4053405]
 [-1.11718674 0.88640526 -1.28103541 -1.05194947 1.19570407]
 [-1.09986601 0.88640526 0.22606507 -1.01378004 -1.28887582]
 [-1.08254529 0.88640526 -0.6351352 -1.01378004 0.88513158]
 [-1.06522456 0.88640526 -0.20453507 -0.89927175 -0.93948177]
 [-1.04790384 0.88640526 -1.3528021 -0.89927175 0.96277471]
 [-1.03058311 0.88640526 1.87669894 -0.86110232 -0.59008772]
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 [1.09986601 0.88640526 -0.56336851 0.78018313 1.6615628]
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 [1.32503543 -1.12815215 1.37433211 1.04736914 -1.36651894]
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 [1.55020485 0.88640526 -0.20453507 1.61991057 1.35099031]
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 [1.60216702 -1.12815215 -0.41983513 2.00160487 -1.63826986]

```
[ 1.61948775  0.88640526 -0.06100169  2.00160487  1.58391968]
[ 1.63680847  0.88640526  0.58489852  2.26879087 -1.32769738]
[ 1.6541292   0.88640526 -0.27630176  2.26879087  1.11806095]
[ 1.67144992  0.88640526  0.44136514  2.49780745 -0.86183865]
[ 1.68877065 -1.12815215 -0.49160182  2.49780745  0.92395314]
[ 1.70609137 -1.12815215 -0.49160182  2.91767117 -1.25005425]
[ 1.7234121  -1.12815215 -0.6351352  2.91767117  1.27334719]]
```

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

```
[23]: X
```

```
[23]: array([[ 15,  39],
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[126, 28],
[126, 74],
[137, 18],
[137, 83]], dtype=int64)
```

DBSCAN Clustering Algorithm

```
[24]: from sklearn.cluster import DBSCAN
      dbscan=DBSCAN(eps=3,min_samples=4)
```

```
[25]: model=dbscan.fit(X)
      labels=model.labels_
```

```
[26]: from sklearn import metrics
```

```
[27]: sample_cores=np.zeros_like(labels,dtype=bool)
      sample_cores[dbscan.core_sample_indices_]=True
```

```
[28]: #Calculating the number of clusters
      n_clusters=len(set(labels))- (1 if -1 in labels else 0)
```

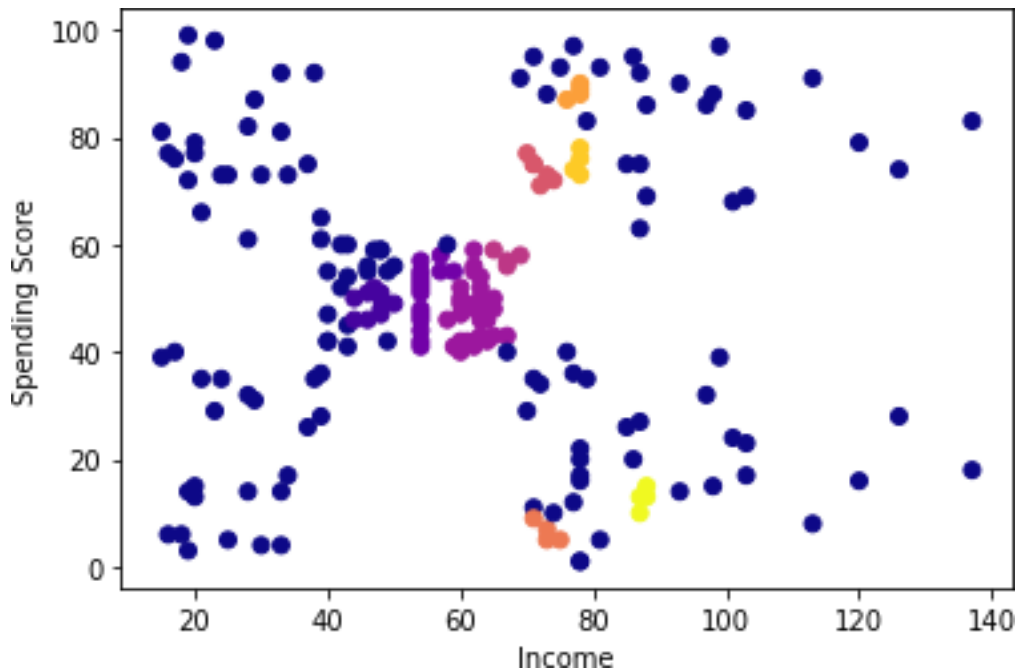
```
[29]: n_clusters
```

```
[29]: 9
```

```
[30]: print(metrics.silhouette_score(X,labels))
```

```
-0.1908319132560097
```

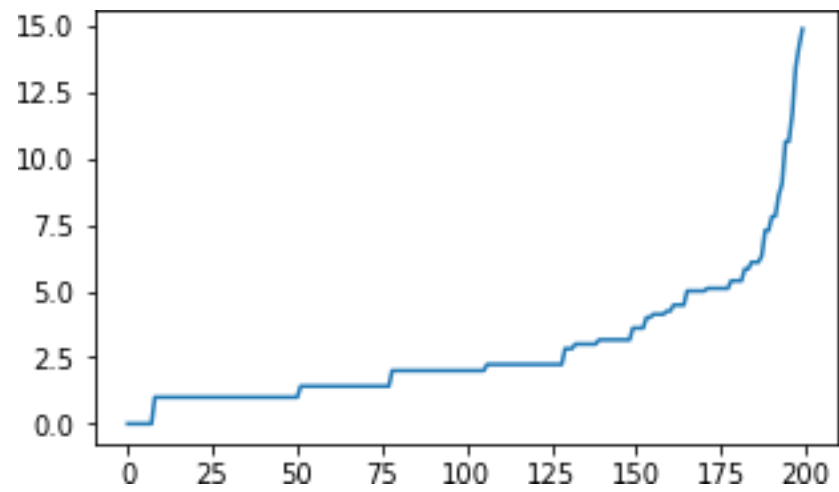
```
[31]: # Plot the clusters
plt.scatter(X[:, 0], X[:,1], c = labels, cmap= "plasma") # plotting the clusters
plt.xlabel("Income") # X-axis label
plt.ylabel("Spending Score") # Y-axis label
plt.show() # showing the plot
```



KNN Algorithm

```
[32]: from sklearn.neighbors import NearestNeighbors # importing the library
      neighb = NearestNeighbors(n_neighbors=2) # creating an object of the _
      ↪NearestNeighbors class
      nbrs=neighb.fit(X) # fitting the data to the object
      distances,indices=nbrs.kneighbors(X) # finding the nearest neighbours

[33]: # Sort and plot the distances results
      distances = np.sort(distances, axis = 0) # sorting the distances
      distances = distances[:, 1] # taking the second column of the sorted distances
      plt.rcParams["figure.figsize"] = (5,3) # setting the figure size
      plt.plot(distances) # plotting the distances
      plt.show() # showing the plot
```



[]:

[]: