Assignment Date	
Student Name	AMARA DHANUSH
Student Roll Number	111519104005
Maximum Marks	2 Marks

#Problem Statement: Customer Segmentation Analysis

###**Description:** You own the mall and want to understand the customers who can quickly converge [Target Customers] so that the insight can be given to the marketing team and plan the strategy accordingly.

Download and load Dataset

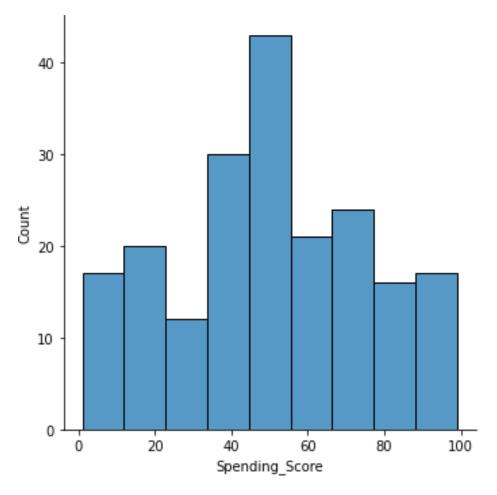
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib as rcParams
df=pd.read csv('Mall Customers.csv') #No Target Column - Unsupervised
Machine Learning
df.head()
  CustomerID Gender Age Annual Income (k$) Spending Score (1-100)
0
     1 Male 19
                              15
                                                                39
          2
              Male 21
                                         15
1
                                                                81
          3 Female 20
2
                                         16
                                                                 6
3
           4 Female 23
                                         16
                                                                77
           5 Female 31
                                                                40
df = df.rename(columns = {'Annual Income (k$)':
'Annual Income', 'Spending Score (1-100)': 'Spending Score')
df.head()
  CustomerID Gender Age Annual Income Spending Score
0
          1 Male 19
                                    15
                                                    39
              Male 21
                                    15
1
                                                   81
          3 Female 20
2
                                    16
                                                    6
3
          4 Female 23
                                    16
                                                    77
          5 Female 31
                                    17
                                                    40
df.shape
(200, 5)
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
```

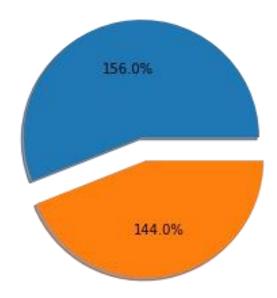
```
200 non-null
2
     Age
                                     int64
 3
     Annual Income 200 non-null
                                     int64
     Spending Score 200 non-null
                                     int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB
df.Gender.unique()
array(['Male', 'Female'], dtype=object)
df.Age.unique()
array([19, 21, 20, 23, 31, 22, 35, 64, 30, 67, 58, 24, 37, 52, 25, 46,
54,
       29, 45, 40, 60, 53, 18, 49, 42, 36, 65, 48, 50, 27, 33, 59, 47,
51,
       69, 70, 63, 43, 68, 32, 26, 57, 38, 55, 34, 66, 39, 44, 28, 56,
411)
df.Gender.value counts()
Female
          112
Male
           88
Name: Gender, dtype: int64
```

Visualizations

Univariate Analysis

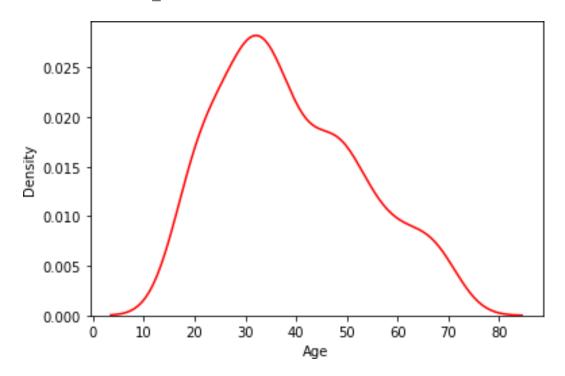
```
sns.displot(df.Spending_Score)
<seaborn.axisgrid.FacetGrid at 0x7f700626b950>
```





sns.kdeplot(df.Age,color="red")

<matplotlib.axes._subplots.AxesSubplot at 0x7f700549a450>



Bi-variate Analysis

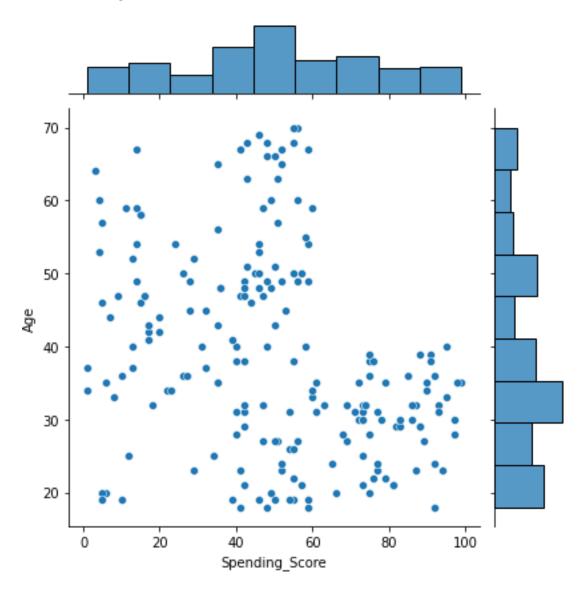
sns.jointplot(df.Spending_Score,df.Age)

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. 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.

FutureWarning

<seaborn.axisgrid.JointGrid at 0x7f7005459c50>

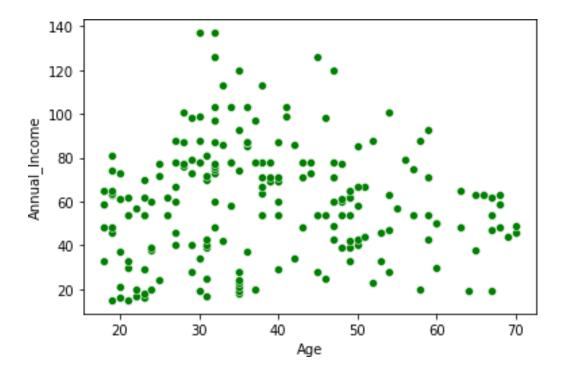


sns.scatterplot(df.Age,df.Annual Income,color="green")

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. 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.

FutureWarning

<matplotlib.axes. subplots.AxesSubplot at 0x7f7005268410>

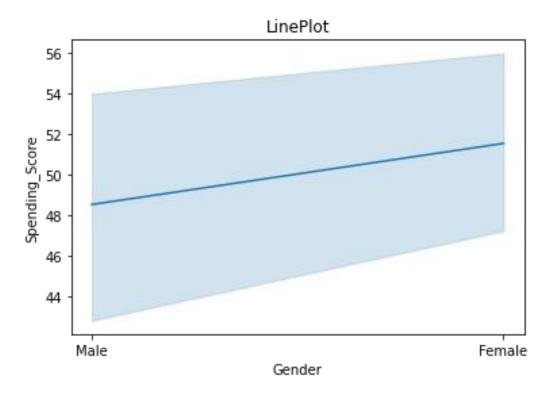


sns.lineplot(df.Gender,df.Spending_Score)
plt.xlabel('Gender')
plt.ylabel('Spending_Score')
plt.title('LinePlot')

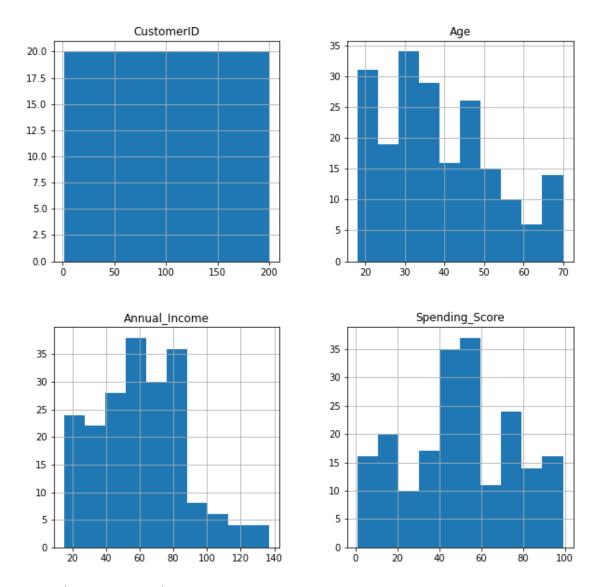
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

FutureWarning

Text(0.5, 1.0, 'LinePlot')

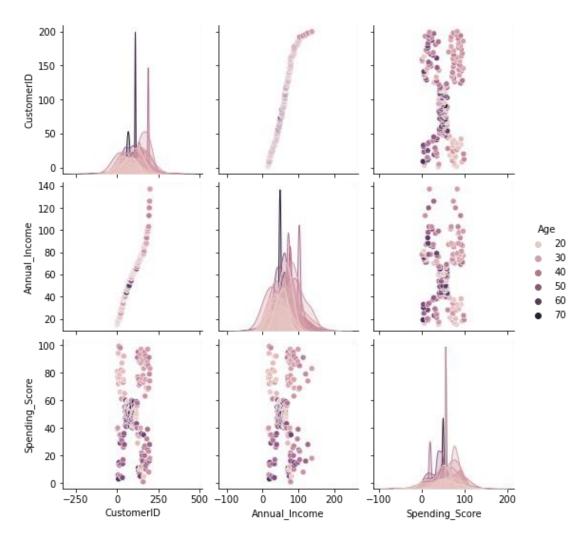


Multi-variate Analysis



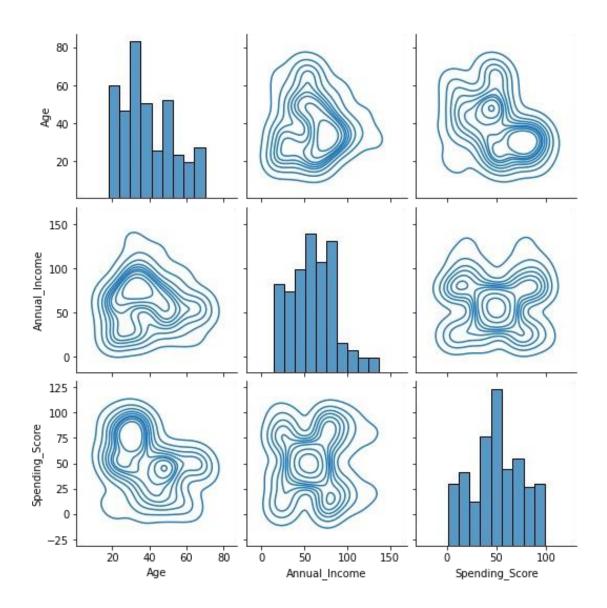
sns.pairplot(df,kind='scatter',hue='Age')

<seaborn.axisgrid.PairGrid at 0x7f700510cd90>



sns.pairplot(data=df[['Age','Annual_Income','Spending_Score']],kind='k
de',diag_kind='hist')

<seaborn.axisgrid.PairGrid at 0x7f7004bd3cd0>



Descriptive statistics

df.describe()

	CustomerID	Age	Annual Income	Spending Score
count	200.000000	200.000000	200.000000	$200.\overline{0}00000$
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

Handle missing data

df.isnull().any() #no missing data

CustomerID False
Gender False
Age False
Annual_Income False
Spending_Score False

dtype: bool

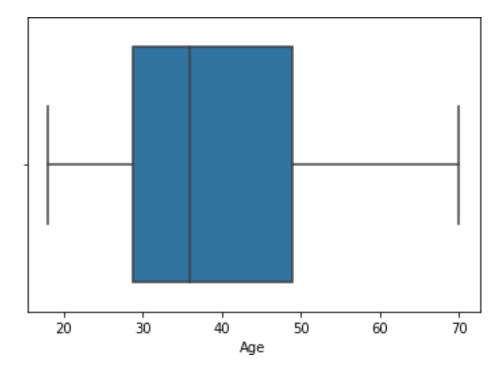
#Outliers Replacement

sns.boxplot(df.Age) #no outliers

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: 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.

FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7f7004604090>



Check for Categorical column and perform encoding

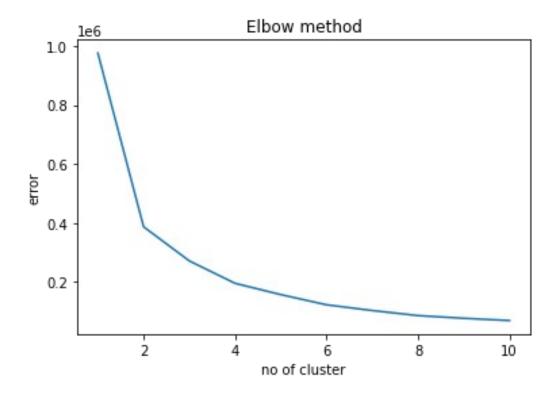
from sklearn.preprocessing import LabelEncoder

le = LabelEncoder()

```
df.Gender=le.fit transform(df.Gender)
df.head()
  CustomerID Gender Age Annual_Income Spending_Score
0
     1
             1
                   19
                                 15
         2
1
                 1 21
                                 1.5
                                               81
                                 16
2
         3
                 0 20
                                               6
3
                    23
                                               77
          4
                 0
                                 16
          5
                                 17
                                               40
                 0 31
```

Perform clustering algorithm

```
from sklearn import cluster
error =[]
for i in range (1,11):
    kmeans=cluster.KMeans(n clusters=i,init='k-means+
+', random state=0)
    kmeans.fit(df)
    error.append(kmeans.inertia)
error
[975512.0600000003,
 387065.71377137717,
271384.508782868,
195401.19855991466,
 157157.7579059829,
 122625.19813553878,
103233.01724386725,
 86053.67444777445,
76938.97565600359,
 69231.33607611558]
import matplotlib.pyplot as plt
plt.plot(range(1,11),error)
plt.title('Elbow method')
plt.xlabel('no of cluster')
plt.ylabel('error')
plt.show()
```



k_means_model.fit(df)

KMeans(n_clusters=3, random_state=0)

clustered data =k means model.predict(df)

Add the cluster data with the primary dataset

df['Clustered_data'] = pd.Series(clustered_data)
df.head()

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

Split the data into dependent and independent variables

```
y=df['Clustered data']
                        #y - target columns
У
0
       0
1
       0
2
       0
3
       0
4
      0
195
      2
196
      2
197
      2
198
199
      2
Name: Clustered data, Length: 200, dtype: int32
X=df.drop(columns=['Clustered data'],axis=1)
X.head()
                                            #X - predicting columns
   CustomerID Gender Age Annual Income
                                          Spending Score
0
       1 1
                       19
           2
1
                   1
                       21
                                      15
                                                      81
2
           3
                   0 20
                                      16
                                                       6
3
           4
                   0
                       23
                                      16
                                                      77
           5
                   0
                       31
                                      17
                                                      40
```

Scale the independent variables

from sklearn.preprocessing import scale

data=pd.DataFrame(scale(X),columns=X.columns)
data.head()

```
Annual Income
                                               Spending Score
  CustomerID
                Gender
                            Age
0
 -1.723412 1.128152 -1.424569
                                     -1.738999
                                                    -0.434801
   -1.706091 1.128152 -1.281035
                                     -1.738999
1
                                                    1.195704
   -1.688771 -0.886405 -1.352802
                                    -1.700830
                                                    -1.715913
 -1.671450 -0.886405 -1.137502
                                    -1.700830
                                                    1.040418
                                    -1.662660
   -1.654129 -0.886405 -0.563369
                                                    -0.395980
```

Split the data into training and testing

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test=train_test_split(data, y, test_size=0.3, ra
ndom_state=1)
X_train.shape, X_test.shape
((140, 5), (60, 5))
```

```
y_train.shape, y_test.shape
((140,), (60,))
```

Build the model

```
from sklearn.neighbors import KNeighborsClassifier
model = KNeighborsClassifier()

model.fit(X_train,y_train) # K - Nearest Neighbour model (KNN)
KNeighborsClassifier()
```

Train the model

Test the data

```
y test
58
         0
40
         0
34
         \cap
102
         1
         2
184
198
         2
95
         1
4
         0
29
         0
168
         2
         2
171
18
         \cap
11
         0
89
         1
```

```
110
       1
118
       1
159
       2
35
       0
       2
136
59
       0
51
       0
16
       0
44
       0
94
       1
31
       0
162
       2
       0
38
28
       0
193
       2
27
       0
47
       0
       2
165
194
      2
      2
177
      2
176
97
       1
       2
174
73
      1
       1
69
      2
172
108
      1
107
       1
189
       2
       0
14
       0
56
       0
19
114
       1
39
       0
       2
185
124
      1
98
       1
123
       1
119
       1
       0
53
33
       0
       2
179
      2
181
106
       1
       2
199
138
       2
Name: Clustered_data, dtype: int32
pred_test=model.predict(X_test)
pred_test
```

```
array([0, 1, 0, 1, 2, 2, 1, 0, 0, 2, 2, 0, 0, 1, 1, 1, 2, 0, 2, 1, 1,
       0, 1, 0, 2, 0, 0, 2, 0, 0, 2, 2, 2, 2, 1, 2, 1, 0, 2, 1, 1, 2,
0,
       0, 0, 1, 0, 2, 1, 1, 1, 1, 1, 0, 2, 2, 1, 2, 2], dtype=int32)
pred =
pd.DataFrame({'Actual value':y test,'Predicted value using KNN':pred t
pred.head()
     Actual value
                  Predicted value using KNN
58
40
                0
                                            1
34
                0
                                            0
102
                1
                                            1
                2
                                            2
184
```

Measure the performance using metrics

from sklearn.metrics import
accuracy_score,confusion matrix,classification report

#Accuracy Score

print('Training accuracy: ',accuracy_score(y_train,pred_train))
print('Testing accuracy: ',accuracy_score(y_test,pred_test))

#Confusion Matrix

pd.crosstab(y test,pred test)

col_0	0	1	2
Clustered_data			
0	19	4	0
1	1	16	0
2	0	0	2.0

#Classification Report

print(classification_report(y_test,pred_test))

support	f1-score	recall	precision	
23	0.88	0.83	0.95	0
17	0.86	0.94	0.80	1
20	1.00	1.00	1.00	2
60	0.92			accuracy
60	0.92	0.92	0.92	macro avg

weighted avg 0.92 0.92 0.92 60