```
import pandas as pd
from sklearn.cluster import KMeans
{\it from sklearn.preprocessing import MinMaxScaler}
import matplotlib.pyplot as plt
from google.colab import files
uploaded = files.upload()
    Choose Files income (1).csv
     • income (1).csv(text/csv) - 363 bytes, last modified: 7/27/2024 - 100% done
     Saving income (1).csv to income (1).csv
df = pd.read_csv("income (1).csv")
df.head()
\overline{2}
                                   Name Age Income($)
                  27
      0
           Rob
                          70000
      1 Michael
                  29
                          90000
         Mohan
                  29
                          61000
          Ismail
                  28
                          60000
                  42
                         150000
           Kory
 Next steps:
              Generate code with df
                                       View recommended plots
                                                                     New interactive sheet
plt.scatter(df.Age,df['Income($)'])
<matplotlib.collections.PathCollection at 0x7f6603122770>
      160000
      140000
      120000
      100000
       80000
       60000
        40000
                     27.5
                              30.0
                                       32.5
                                               35.0
                                                        37.5
                                                                 40.0
                                                                          42.5
\#k = 3
     KMeans(n_clusters=3)
km
₹
              KMeans
     KMeans(n_clusters=3)
```

y_predicted = km.fit_predict(df[["Age","Income(\$)"]])

df['cluster'] = y_predicted

df.head()



Next steps: Generate code with df View recommended plots New interactive sheet

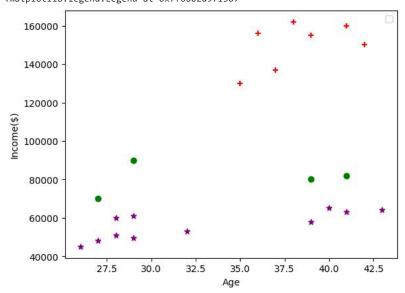
What happened here it, it applied the Kmeans algorithm and then it formed 3 clusters. It assigned them 3 clusters with labels 0,1,2

```
df_1 = df[df.cluster = = = 0]
df_2 = df[df.cluster = = = 1]
df_3 = df[df.cluster = = = 2]

plt.scatter(df_1['Age'], df_1['Income($)'], color = = 'red', marker = = "+")
plt.scatter(df_2['Age'], df_2['Income($)'], color = = 'purple', marker = = "*")
plt.scatter(df_3['Age'], df_3['Income($)'], color = = 'green')

plt.xlabel("Age")
plt.ylabel("Income($)")
plt.legend()
```

WARNING:matplotlib.legend:No artists with labels found to put in legend. Note that a contract watplotlib.legend.Legend at 0x7f6602d97130>



The plot does not look okay as the green ones and the purples ones look merged, it is happening because of scaling, the range of x axis is too narrow compared to the x axis, this is the reason why we use a MinMaxScaler()

```
₹
           Name
                 Age Income($) cluster
            Rob
                  27
                        0.213675
         Michael
                  29
                        0.384615
                                        2
      2
         Mohan
                  29
                        0.136752
      3
           Ismail
                  28
                        0.128205
                                        1
            Kory
                  42
                        0.897436
                                        0
              Generate code with df
                                       View recommended plots
                                                                      New interactive sheet
 Next steps:
scaler.fit(df[["Age"]])
df["Age"]=scaler.transform(df[["Age"]])
df.head()
\overline{\mathbf{T}}
                                                 \blacksquare
           Name
                      Age Income($) cluster
            Rob 0.058824
      0
                             0.213675
                                                  ılı.
      1 Michael
                 0.176471
                             0.384615
                                             2
         Mohan 0.176471
                             0.136752
                 0.117647
           Ismail
                             0.128205
            Kory 0.941176
                             0.897436
                                             0
 Next steps:
              Generate code with df
                                       View recommended plots
                                                                      New interactive sheet
km = KMeans(n_clusters=3)
y_predicted = km.fit_predict(df[["Age",'Income($)']])
df['cluster']=y_predicted
    /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change d
       super()._check_params_vs_input(X, default_n_init=10)
df_1 =df[df.cluster == 0]
df_2 =df[df.cluster == 1]
df_3 =df[df.cluster == 2]
plt.scatter(df_1['Age'], df_1['Income($)'],color = 'red',label = "cluster 1")
plt.scatter(df_2['Age'], df_2['Income($)'],color = 'purple',label = "cluster 2")
plt.scatter(df_3['Age'], df_3['Income($)'],color = 'green', label = "cluster 3")
plt.scatter(km.cluster_centers_[:,0],km.cluster_centers_[:,1],color ="black",marker = "*",label = "cendroid")
plt.xlabel("Age")
plt.ylabel("Income($)")
plt.legend()
<matplotlib.legend.Legend at 0x7f65fed5f4f0>
         1.0
                    cluster 1
                    cluster 2
                    cluster 3
                    cendroid
         0.8
         0.6
      Income($)
         0.4
         0.2
         0.0
               0.0
                           0.2
                                       0.4
                                                   0.6
                                                               0.8
                                                                           1.0
                                             Age
```

It is much better now and it has been solved.

In a real life problem it would be more complicated and we have to use the elbow method. We choose a number of k and we find the SSE

```
k_range = range(1,10)
SSE =[]
for k in k_range:
  km = KMeans(n_clusters=k)
  km.fit(df[["Age","Income($)"]])
  SSE.append(km.inertia_)
🚁 /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change 🕆
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change ⊣
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/ kmeans.py:1416: FutureWarning: The default value of `n init` will change +
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change <code>+</code>
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change ⊣
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change ⊣
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change +
       super()._check_params_vs_input(X, default_n_init=10)
     /usr/local/lib/python3.10/dist-packages/sklearn/cluster/_kmeans.py:1416: FutureWarning: The default value of `n_init` will change ⊣
       super()._check_params_vs_input(X, default_n_init=10)
    4
SSE

→ [5.434011511984241,
      2.091136388689264,
      0.4750783498520276,
      0.3491047094404182,
      0.26640301246863574,
      0.21066678487917875.
      0.17796706251972708,
      0.1326541982744777,
      0.101887877250499]
plt.plot(k_range,SSE)
plt.ylabel("SSE")
plt.xlabel("K")
```

→ Text(0.5, 0, 'K')

