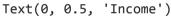
```
from sklearn.cluster import KMeans
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
from matplotlib import pyplot as plt
%matplotlib inline
```

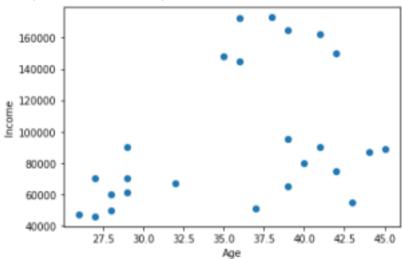
df = pd.read\_csv("/content/income.csv")
df.head()

#### Name Age Income

- **0** A 27 70000
- 1 B 29 90000
- 2 C 29 61000
- 3 D 28 60000
- 4 E 42 150000

```
plt.scatter(df.Age,df['Income'])
plt.xlabel('Age')
plt.ylabel('Income')
```





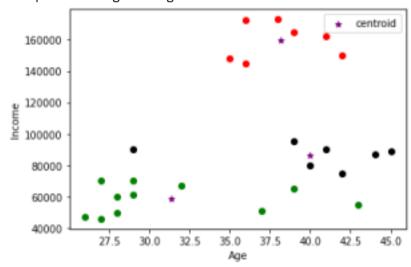
km = KMeans(n\_clusters=3)
y\_predicted = km.fit\_predict(df[['Age','Income']])
y\_predicted

df['cluster']=y\_predicted
df.head()

```
Name Age Income cluster
```

```
0 A 27 70000 0
      1 B 29 90000 2
      2 C 29 61000 0
      3 D 28 60000 0
      4 E 42 150000 1
km.cluster_centers_
     array([[3.13636364e+01, 5.83090909e+04],
      [3.81428571e+01, 1.59328571e+05],
      [4.00000000e+01, 8.65714286e+04]])
df1 = df[df.cluster==0]
df2 = df[df.cluster==1]
df3 = df[df.cluster==2]
plt.scatter(df1.Age,df1['Income'],color='green')
plt.scatter(df2.Age,df2['Income'],color='red')
plt.scatter(df3.Age,df3['Income'],color='black')
plt.scatter(km.cluster_centers_[:,0],km.cluster_centers_[:,1],color='purple',marker='*',la
plt.xlabel('Age')
plt.ylabel('Income')
plt.legend()
```

# <matplotlib.legend.Legend at 0x7f95a19f7d10>



## Preprocessing using min max scaler

```
scaler = MinMaxScaler()

scaler.fit(df[['Income']])

df['Income'] = scaler.transform(df[['Income']])

scaler.fit(df[['Age']])

df['Age'] = scaler.transform(df[['Age']])
```

 $https://colab.research.google.com/drive/1hQjR8EyMf1wbU_ljY1yLLb-wmYWvPlOf\#printMode=true~2/5~10/13/22,~2:37~PM~Clustering~With~K~Means.ipynb~-~Colaboratory$ 

#### Name Age Income cluster

```
0 A 0.052632 0.190252 0
```

1 B 0.157895 0.347484 2

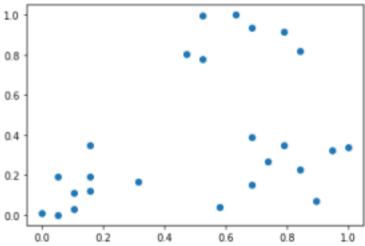
2 C 0.157895 0.119497 0

3 D 0.105263 0.111635 0

4 E 0.842105 0.819182 1

## plt.scatter(df.Age,df['Income'])

<matplotlib.collections.PathCollection at 0x7f95a1931e10>



km = KMeans(n\_clusters=3)
y\_predicted = km.fit\_predict(df[['Age','Income']])
y\_predicted

array([1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 2, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2], dtype=int32)

df['cluster']=y\_predicted
df.head()

#### Name Age Income cluster

**0** A 0.052632 0.190252 1

**1** B 0.157895 0.347484 1

2 C 0.157895 0.119497 1

3 D 0.105263 0.111635 1

4 E 0.842105 0.819182 0

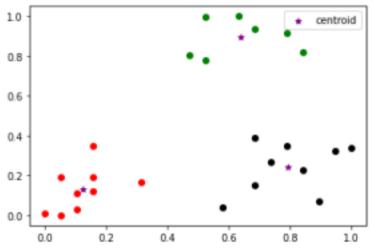
km.cluster\_centers\_

https://colab.research.google.com/drive/1hQjR8EyMf1wbU\_ljY1yLLb-wmYWvPlOf#printMode=true 3/5 10/13/22, 2:37 PM Clustering With K Means.ipynb - Colaboratory

```
array([[0.63909774, 0.89252022], [0.12280702, 0.12945493], [0.79532164, 0.24004193]])
```

```
df1 = df[df.cluster==0]
df2 = df[df.cluster==1]
df3 = df[df.cluster==2]
plt.scatter(df1.Age,df1['Income'],color='green')
plt.scatter(df2.Age,df2['Income'],color='red')
plt.scatter(df3.Age,df3['Income'],color='black')
plt.scatter(km.cluster_centers_[:,0],km.cluster_centers_[:,1],color='purple',marker='*',la
plt.legend()
```

## <matplotlib.legend.Legend at 0x7f95a18ade90>

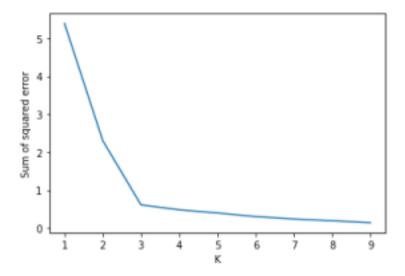


#### **Elbow Plot**

```
sse = []
k_rng = range(1,10)
for k in k_rng:
    km = KMeans(n_clusters=k)
    km.fit(df[['Age','Income']])
    sse.append(km.inertia_) # inertia will calculate SSE

plt.xlabel('K')
plt.ylabel('Sum of squared error')
plt.plot(k_rng,sse)
```

 $https://colab.research.google.com/drive/1hQjR8EyMf1wbU_ljY1yLLb-wmYWvPlOf\#printMode=true~4/5~10/13/22,~2:37~PM~Clustering~With~K~Means.ipynb~-~Colaboratory$ 



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