

```
In [1]: # import the library
import pandas as pd
import numpy as np
```

Read and understand the data

```
In [2]: # load the data
df = pd.read_csv('Mall_Customers.csv')
```

```
In [3]: # first five rows
df.head()
```

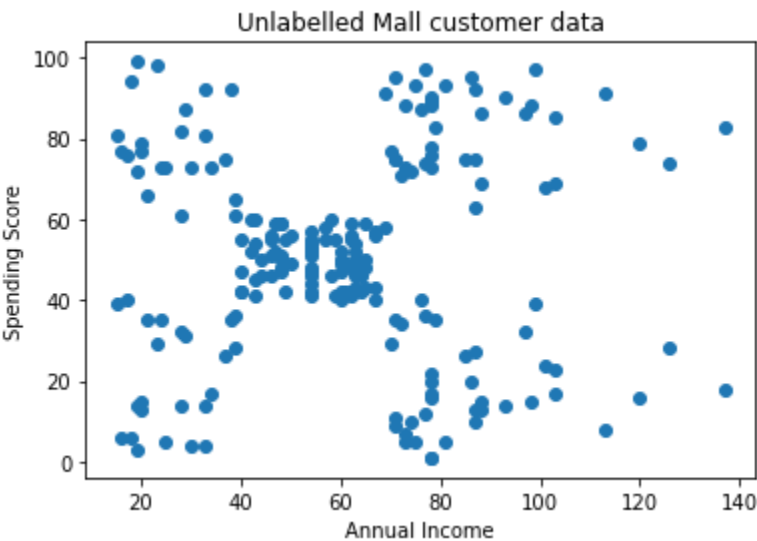
	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
2	3	Female	20	16	6
3	4	Female	23	16	77
4	5	Female	31	17	40

```
In [4]: # data types
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   Genre                 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
```

Visualise the data

```
In [7]: # import library
import matplotlib.pyplot as plt
plt.scatter(df['Annual Income (k$)'], df['Spending Score (1-100)'])
plt.xlabel('Annual Income')
plt.ylabel('Spending Score')
plt.title('Unlabelled Mall customer data')
plt.show()
```



Define the value of X

```
In [8]: # define X , unsupervised learning no labels , no Y
X = df[['Annual Income (k$)','Spending Score (1-100)']]
X.head()
```

	Annual Income (k\$)	Spending Score (1-100)
0	15	39
1	15	81
2	16	6
3	16	77
4	17	40

Build the model

```
In [9]: # import the library to build the model
from sklearn.cluster import KMeans
kmeans = KMeans(n_clusters=5 , random_state=42)
```

Train the Model

```
In [11]: kmeans.fit(X)
```

Out[11]: KMeans(n_clusters=5, random_state=42)

Predict

```
In [13]: pred = kmeans.predict(X)
```

```
In [14]: df['clusters'] = pred
```

```
In [15]: df.head()
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)	clusters
0	1	Male	19	15	39	2
1	2	Male	21	15	81	3
2	3	Female	20	16	6	2
3	4	Female	23	16	77	3
4	5	Female	31	17	40	2

```
In [16]: ## Show the value for cluster 2
df[df['clusters']==2].head(10)
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)	clusters
0	1	Male	19	15	39	2
2	3	Female	20	16	6	2
4	5	Female	31	17	40	2
6	7	Female	35	18	6	2
8	9	Male	64	19	3	2
10	11	Male	67	19	14	2
12	13	Female	58	20	15	2
14	15	Male	37	20	13	2
16	17	Female	35	21	35	2
18	19	Male	52	23	29	2

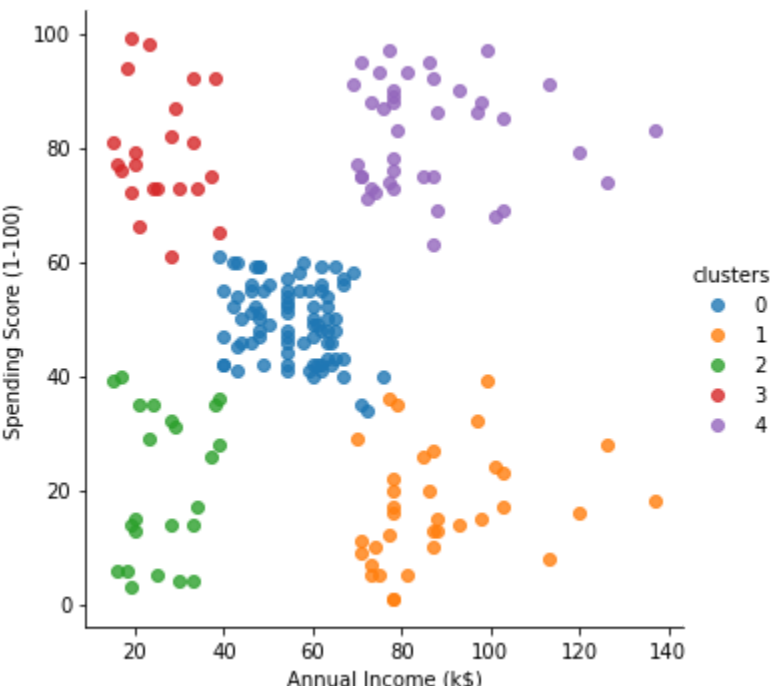
```
In [17]: ## Show the value for cluster 2 and only Annual Income (k$)
df[df['clusters']==2]['Annual Income (k$)'].head(10)
```

Out[17]: 0 15
2 16
4 17
6 18
8 19
10 19
12 20
14 20
16 21
18 23
Name: Annual Income (k\$), dtype: int64

```
In [19]: ## Show the value for cluster 2 and only Spending Score (1-100)
df[df['clusters']==2]['Spending Score (1-100)'].head(10)
```

Out[19]: 0 39
2 6
4 40
6 6
8 3
10 14
12 15
14 13
16 35
18 29
Name: Spending Score (1-100), dtype: int64

```
In [21]: # seaborn is a datavisualisation library
# visualising the cluster
import seaborn as sns
sns.lmplot(data=df , x='Annual Income (k$)' , y='Spending Score (1-100)' , hue='clusters',fit_reg=False)
plt.show()
```



```
In [22]: ## Coordinates of centroids
kmeans.cluster_centers_
```

Out[22]: array([[55.2962963 , 49.51851852],
 [88.2 , 17.11428571],
 [26.30434783, 20.91304348],
 [25.72727273, 79.36363636],
 [86.53846154, 82.12820513]])

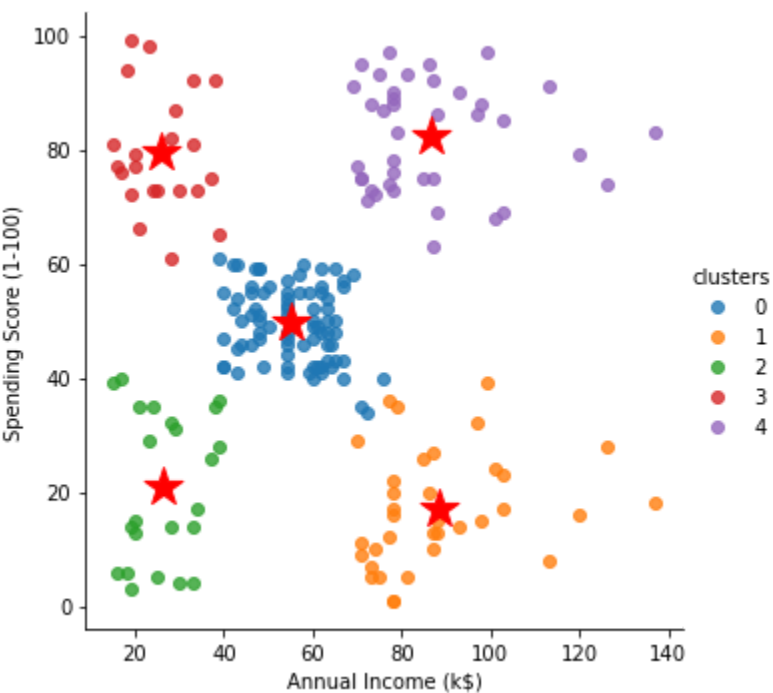
```
In [23]: # X_coordinates of centroid
kmeans.cluster_centers_[0,0]
```

Out[23]: array([55.2962963 , 88.2 , 26.30434783, 25.72727273, 86.53846154])

```
In [24]: # Y_coordinates of centroid
kmeans.cluster_centers_[0,1]
```

Out[24]: array([49.51851852, 17.11428571, 20.91304348, 79.36363636, 82.12820513])

```
In [30]: # visualise with centroids
sns.lmplot(data=df , x='Annual Income (k$)' , y='Spending Score (1-100)' , hue='clusters',fit_reg=False)
plt.scatter(kmeans.cluster_centers_[0,0],kmeans.cluster_centers_[0,1], c='red',marker='*',s=400 , label='centroid')
plt.legend()
plt.show()
```



In [] :