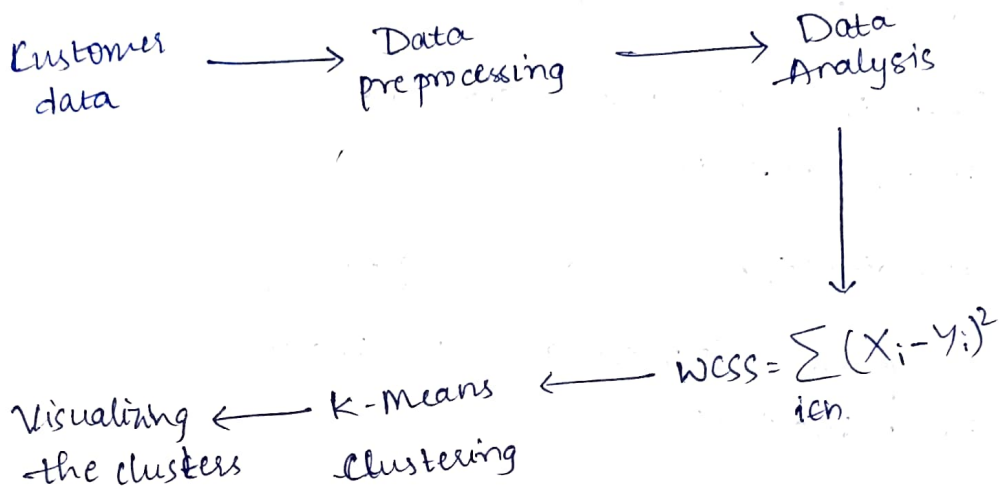


# Customer Segmentation using k-means Clustering with Python.

## work flow:-



## Code:-

\* Importing the Dependencies (required libraries & functions)

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans
```

numpy - A way to handle matrices and vectors in python  
pandas - Pandas is used for making dataframes i.e structured table.  
matplotlib & seaborn - These are the data visualization libraries  
sklearn.cluster - From sklearn library in cluster module we import kmeans algorithm.

## \* Data Collection & Analysis

```
#loading the data from csv
```

```
customer_data = pd.read_csv('/content/Mall-Customer.csv')
```

```
#first 5 rows in the dataframe
```

```
customer_data.head()
```

head() - It is used to give first 5 rows in dataframe.

```
#finding the number of rows and columns
```

```
customer_data.shape
```

Shape = gives the total no of rows & columns.

```
#getting some information about the dataset
```

```
customer_data.info()
```

```
#checking for missing values
```

```
customer_data.isnull().sum()
```

## \* Choosing the annual income column & spending score column.

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

```
print(X)
```

iloc[:, [3, 4]].values = Used to choose the 3rd & 4th columns from the dataframe.

":" is used to specify the we need 3rd & 4th "columns" other it will consider rows.

## \* Choosing the number of cluster

# finding wcss value for different number of clusters

```
wcss = []
```

```
for i in range(1, 11):
```

```
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
```

```
    kmeans.fit(X)
```

```
    wcss.append(kmeans.inertia_)
```

range(1, 11) = Here it checks for 10 clusters.

kmeans = variable created to load KMeans cluster-function.

n\_clusters = Represents the num of clusters we want.

kmeans.fit(X) = Here we are fitting our data to kmeans.

kmeans.inertia\_ = Gives us the wcss value and that value will be stored in the empty list.

WCSS (Within Cluster Sum Of Squares)

# plot an elbow graph

```
sns.set()
```

```
plt.plot(range(1, 11), wcss)
```

```
plt.title('The Elbow Point Graph')
```

```
plt.xlabel('Number of Clusters')
```

```
plt.ylabel('WCSS')
```

```
plt.show()
```

- By running this code we will get an elbow graph where we can have elbow points. We have to select a point so that after that point there should not be any sharp significant drop.
- By this we can say correct optimum number of clusters

## \* Training the k-means clustering model.

```
kmeans = KMeans(n_clusters=5, init='k-means++', random_state=0)
# returning a label for each datapoint based on their cluster
Y = kmeans.fit_predict(X)
print(Y)
```

## \* Visualizing all the clusters

# plotting all the clusters and their centroids

```
plt.figure(figsize=(8,8))
```

```
plt.scatter(X[Y==0,0], X[Y==0,1], s=50, c='green', label='Cluster1')
```

```
plt.scatter(X[Y==1,0], X[Y==1,1], s=50, c='red', label='Cluster2')
```

```
plt.scatter(X[Y==2,0], X[Y==2,1], s=50, c='yellow', label='Cluster3')
```

```
plt.scatter(X[Y==3,0], X[Y==3,1], s=50, c='blue', label='Cluster4')
```

```
plt.scatter(X[Y==4,0], X[Y==4,1], s=50, c='violet', label='Cluster5')
```

# plot the centroids

```
plt.scatter(kmeans.cluster_centers_[:,0], kmeans.cluster_centers_[:,1], s=100,
            c='cyan', label='Centroids')
```

```
plt.title('Customer Groups')
```

```
plt.xlabel('Annual Income')
```

```
plt.ylabel('Spending Score')
```

```
plt.show()
```

`plt.scatter(X[Y==0,0], X[Y==0,1])`

← this indicates the cluster 1

← this is the index value of annual income

← this is the index value of spending score

`s=50`

↓  
indicates the size of the datapoints in the cluster

`s=100`

↓  
indicates the size of the centroids in each cluster



## Conclusion:

- this the process to make clusters (groups among people) and make better recommendations for them.
- For example: Consider a movie ~~watching~~ streaming platform like netflix. Some people want to watch thriller and some other want ~~horror~~. ~~so~~ when a ~~new~~ person watches a horror movie, it automatically recommend some more horror movies.
- By this method we can give better recommendations for the people.