# **Major Project Synopsis**

On

### **Customer Segmentation using Machine Learning**

In partial fulfilment of requirements for the degree

**Of** 

### **BACHELOR OF TECHNOLOGY**

IN

#### **COMPUTER SCIENCE & ENGINEERING**

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<u>JUL-DEC – 2022</u>

SHRI VAISHNAV INSTITUTE OF INFORMATION TECHNOLOGY

### **INTRODUCTION**

In this Data Science Project, we will perform one of the most essential applications of machine learning — Customer Segmentation. In this project, we will implement customer segmentation . Whenever we need to find your best customer, customer segmentation is the ideal methodology. Also, in this data science project, we will see the descriptive analysis of our data and then implement several versions of the K-means algorithm. So, follow the complete data science customer segmentation project using machine learning in R and become a pro in Data Science.

### PROBLEM STATEMENT

Consider a mall a very famous mall and we are a very experienced data scientist and this mall wants all information about their customers and other expects. As a data scientist you can build a system that can cluster customer in different groups.

One group of customer are those who purchase more from that mall and other group are those who don't purchase too much from that mall. So having these group of customer these is easy to understand and better details to make and understand marketing strategies.



## WORK FLOW

First we want to make a small customer data to train or instruct our machine learning project model. So first is getting those customer data and then we have to process these data, we cannot feed these data directly in machine learning model so we need to select key features that particular dataset contain is all come in analysis process and after that we need to choose the current no.

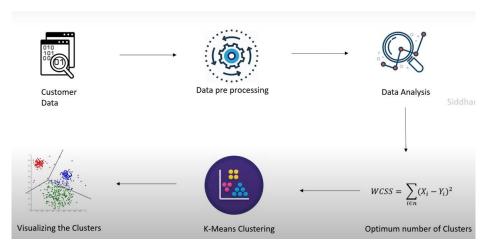
Of clusters and tell the machine learning project how many clusters are exists.

So we find the no. of clusters using method within cluster sum of square, so in this process we find the value of WCSS.

Once we have no. of cluster we can find the data in K-mean cluster Algorithm.

So once we f feed this algorithm to this model it can grew the data dependent on the similarities or similar expending pattern etc.

After that we can visualizing these cluster data by putting these data on the prediction made by the clustering models to get better insect of data.



### **K-Means Algorithm**

- K-Means Clustering is an Unsupervised Learning algorithm, which groups the unlabeled dataset into different clusters.
- Here K defines the number of pre-defined clusters that need to be created in the process, as if K=2, there will be two clusters, and for K=3, there will be three clusters, and so on.
- It allows us to cluster the data into different groups and a convenient way to discover the categories of groups in the unlabeled dataset on its own without the need for any training.
- It is a centroid-based algorithm, where each cluster is associated with a centroid. The main aim of this algorithm is to minimize the sum of distances between the data point and their corresponding clusters.

## **Step-1: Data pre-processing Step**

### **Importing the Dependencies**

In the above code, the numpy we have imported for the performing mathematics calculation, **matplotlib** is for plotting the graph, and **pandas** are for managing the dataset.

```
In [1]: import numpy as np
   import pandas as pd
   import matplotlib.pyplot as plt
   import seaborn as sns
   from sklearn.cluster import KMeans
```

#### Importing dataset.

Next, we will import the dataset that we need to use. So here, we are using the Mall\_Customer\_data.csv dataset. It can be imported using the below code:

```
In [7]: #read dataset
    print("Prachi Gothwal")
    print("Sakshi Patel")
    customer_data = pd.read_csv("C:\\Users\\Abhishek\\Desktop\\Mall_Customers.csv")

Prachi Gothwal
    Sakshi Patel
```

#### **Data Analysis**

```
In [8]: print("Prachi Gothwal")
         print("Sakshi Patel")
         customer_data.head()
         Prachi Gothwal
         Sakshi Patel
Out[8]:
             CustomerID Gender Age Annual Income (k$) Spending Score (1-100)
          0
                          Male
                                 19
                                                   15
                                                                        39
          1
                     2
                          Male
                                 21
                                                   15
                                                                        81
                      3 Female
                                  20
                                                   16
                                                                         6
          3
                     4 Female
                                 23
                                                   16
                                                                        77
                     5 Female
                                 31
                                                   17
                                                                        40
```

```
In [9]: # finding the number of rows and columns
    print("Prachi Gothwal")
    print("Sakshi Patel")
    customer_data.shape

    Prachi Gothwal
    Sakshi Patel

Out[9]: (200, 5)
```

```
[10]: print("Prachi Gothwal")
      print("Sakshi Patel")
      customer_data.info()
      Prachi Gothwal
      Sakshi Patel
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 200 entries, 0 to 199
      Data columns (total 5 columns):
         Column
                               Non-Null Count Dtype
                               -----
         CustomerID
                               200 non-null
                                            int64
         Gender
                               200 non-null
                                            object
         Age
                               200 non-null
                                            int64
         Annual Income (k$)
                              200 non-null
                                            int64
       4 Spending Score (1-100) 200 non-null
                                           int64
      dtypes: int64(4), object(1)
      memory usage: 7.9+ KB
1 [12]: # checking for missing values
         print("Prachi Gothwal")
         print("Sakshi Patel")
         customer data.isnull().sum()
         Prachi Gothwal
         Sakshi Patel
it[12]: CustomerID
                                         0
         Gender
                                         0
         Age
         Annual Income (k$)
         Spending Score (1-100)
         dtype: int64
```

### **Choosing the Annual Income Column & Spending Score column**

```
In [13]: print("Prachi Gothwal")
         print("Sakshi Patel")
         X = customer_data.iloc[:,[3,4]].values
         Prachi Gothwal
         Sakshi Patel
In [14]: print(X)
         [[ 15
               39]
          [ 15 81]
          [ 16
                6]
          [ 16 77]
          [ 17 40]
           17
               76]
           18
                6]
           18 94]
           19
                3]
           19
               72]
           19
               14]
           19
               99]
           20
               15]
           20
               77]
           20
               13]
           20
               79]
           21 35]
           21
                66]
           23
                29]
```

### Step-2: Finding the optimal number of clusters using the elbow method

In the second step, we will try to find the optimal number of clusters for our clustering problem.

- Choosing the number of clusters
- WCSS -> Within Clusters Sum of Squares

```
In [15]: # finding wcss value for different number of clusters
print("Prachi Gothwal")
print("Sakshi Patel")
wcss = []

for i in range(1,11):
    kmeans = KMeans(n_clusters=i, init='k-means++', random_state=42)
    kmeans.fit(X)

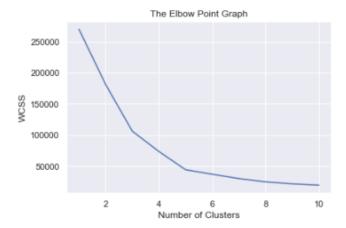
wcss.append(kmeans.inertia_)

Prachi Gothwal
Sakshi Patel

C:\Users\Abhishek\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1036: UserWarning: KMeans is known to have a memory le ak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=1.
    warnings.warn(
```

```
In [16]: # plot an elbow graph
    print("Prachi Gothwal")
    print("Sakshi Patel")
    sns.set()
    plt.plot(range(1,11), wcss)
    plt.title('The Elbow Point Graph')
    plt.xlabel('Number of Clusters')
    plt.ylabel('WCSS')
    plt.show()

Prachi Gothwal
Sakshi Patel
```



- From the above plot, we can see the elbow point is at 5. So the number of clusters here will be 5.
- Optimum Number of Clusters = 5

### Step- 3: Training the K-means algorithm on the training dataset

As we have got the number of clusters, so we can now train the model on the dataset.

To train the model, we will use the same two lines of code as we have used in the above section, but here instead of using i, we will use 5, as we know there are 5 clusters that need to be formed. The code is given below:

5 Clusters - 0, 1, 2, 3, 4

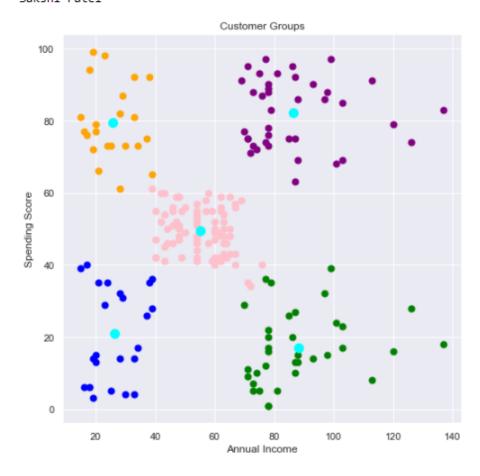
### **Step-4: Visualizing the Clusters**

The last step is to visualize the clusters. As we have 5 clusters for our model, so we will visualize each cluster one by one.

```
# plotting all the clusters and their Centroids
print("Prachi Gothwal")
print("Sakshi Patel")
plt.figure(figsize=(8,8))
plt.scatter(X[Y==0,0], X[Y==0,1], s=50, c='green', label='Cluster 1')
plt.scatter(X[Y==1,0], X[Y==1,1], s=50, c='pink', label='Cluster 2')
plt.scatter(X[Y==2,0], X[Y==2,1], s=50, c='purple', label='Cluster 3')
plt.scatter(X[Y==3,0], X[Y==3,1], s=50, c='orange', label='Cluster 4')
plt.scatter(X[Y==4,0], X[Y==4,1], s=50, c='blue', label='Cluster 5')

# 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()
```

#### Prachi Gothwal Sakshi Patel



## **CONCLUSION**

As we see we have multiple cluster here in different different color. All the cluster are portioned differently on the screen and the cluster represent the group of customer we are getting .As we focus on particular cluster it means a group of people we can see not too many have more annual income but they are regular customer of the mall except these others are not regular customers of mall.

So through these survey we can focus on that group of customer who are not buying to much from the mall, we can give them offers so they can be regular customers.

Through these survey we can improve the market value and the profit of mall and improve their status.