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### **EXPERIMENT-8**

#### **Problem Statement**

Apply the k-means algorithm and apply it to the selected data. Evaluate the process by measuring the sum of euclidean distance of each example from its class center. Test the performance of the algorithm as a function of the parameter k.

# Algorithm

K-Means Clustering Algorithm

K-Means Clustering is an unsupervised learning algorithm that is used to solve the clustering problems in machine learning or data science. In this topic, we will learn what is K-means clustering algorithm, how the algorithm works, along with the Python implementation of k-means clustering.

What is K-Means Algorithm?

K-Means Clustering is an <u>Unsupervised Learning algorithm</u>, 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 is an iterative algorithm that divides the unlabeled dataset into k different clusters in such a way that each dataset belongs only one group that has similar properties.

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.

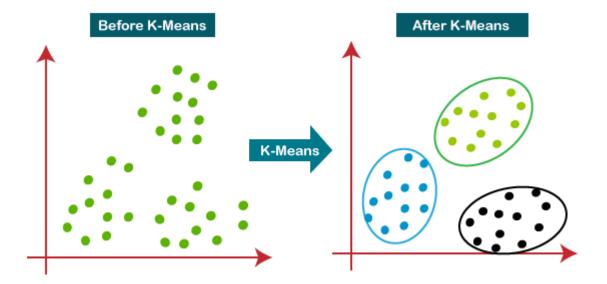
The algorithm takes the unlabeled dataset as input, divides the dataset into k-number of clusters, and repeats the process until it does not find the best clusters. The value of k should be predetermined in this algorithm.

The k-means <u>clustering</u> algorithm mainly performs two tasks:

- o Determines the best value for K center points or centroids by an iterative process.
- o Assigns each data point to its closest k-center. Those data points which are near to the particular k-center, create a cluster.

Hence each cluster has datapoints with some commonalities, and it is away from other clusters.

The below diagram explains the working of the K-means Clustering Algorithm:



How does the K-Means Algorithm Work?

The working of the K-Means algorithm is explained in the below steps:

**Step-1:** Select the number K to decide the number of clusters.

**Step-2:** Select random K points or centroids. (It can be other from the input dataset).

**Step-3:** Assign each data point to their closest centroid, which will form the predefined K clusters.

**Step-4:** Calculate the variance and place a new centroid of each cluster.

**Step-5:** Repeat the third steps, which means reassign each datapoint to the new closest centroid of each cluster.

**Step-6:** If any reassignment occurs, then go to step-4 else go to FINISH.

**Step-7**: The model is ready.

# **Program Screenshots**

```
In [1]: import pandas as pd
                                import numpy as np
                                import matplotlib.pyplot as plt
                                from sklearn.cluster import KMeans
In [2]: df1 = pd.read_csv("Country-data.csv")
    df2 = pd.read_csv("creditcard.csv")
 In [3]: df1.head()
Out[3]:
                                                                             country child_mort exports health imports income inflation life_expec total_fer
                                                                                                                                                                                                                                                                                                                                                  gdpp
                                    0
                                                                    Afghanistan
                                                                                                                            90.2
                                                                                                                                                                              7.58
                                                                                                                                                                                                                                   1610
                                                                                                                                                                                                                                                                                                    56.2
                                                                                                                                                                                                                                                                                                                                                        553
                                    1
                                                                                Albania
                                                                                                                            16.6
                                                                                                                                                       28.0
                                                                                                                                                                             6.55
                                                                                                                                                                                                         48.6
                                                                                                                                                                                                                                  9930
                                                                                                                                                                                                                                                                 4.49
                                                                                                                                                                                                                                                                                                  76.3
                                                                                                                                                                                                                                                                                                                                1.65
                                                                                                                                                                                                                                                                                                                                                    4090
                                    2
                                                                                                                            27.3
                                                                                                                                                       38.4
                                                                                                                                                                                                                                                              16.10
                                                                                                                                                                                                                                                                                                  76.5
                                                                                                                                                                                                                                                                                                                                2.89
                                                                                                                                                                                                                                                                                                                                                    4460
                                                                                 Algeria
                                                                                                                                                                              4.17
                                                                                                                                                                                                         31.4
                                                                                                                                                                                                                               12900
                                    3
                                                                                 Angola
                                                                                                                          119.0
                                                                                                                                                       62.3
                                                                                                                                                                             2.85
                                                                                                                                                                                                         42.9
                                                                                                                                                                                                                                  5900
                                                                                                                                                                                                                                                             22.40
                                                                                                                                                                                                                                                                                                  60.1
                                                                                                                                                                                                                                                                                                                                6.16
                                                                                                                                                                                                                                                                                                                                                    3530
                                    4 Antigua and Barbuda
                                                                                                                             10.3
                                                                                                                                                       45.5
                                                                                                                                                                             6.03
                                                                                                                                                                                                                               19100
                                                                                                                                                                                                                                                                                                   76.8
                                                                                                                                                                                                                                                                                                                                2.13 12200
 In [4]: df2.head()
Out[4]:
                        0 0.0 -1.359807 -0.072781 2.536347 1.378155 -0.338321 0.462388 0.239599 0.098698 0.363787 ... -0.018307 0.277838 -0.110474 0.066928 0.12853
                         1 0.0 1.191857 0.266151 0.166480 0.448154 0.060018 -0.082361 -0.078803 0.085102 -0.255425 ... -0.225775 -0.638672 0.101288 -0.339846 0.16717
                        2 1.0 -1.358354 -1.340163 1.773209 0.379780 -0.503198 1.800499 0.791461 0.247676 -1.514654 ... 0.247998 0.771679 0.909412 -0.689281 -0.32764
                                  1.0 \quad -0.966272 \quad -0.185226 \quad 1.792993 \quad -0.863291 \quad -0.010309 \quad 1.247203 \quad 0.237609 \quad 0.377436 \quad -1.387024 \quad \dots \quad -0.108300 \quad 0.005274 \quad -0.190321 \quad -1.175575 \quad 0.6473761 \quad -0.108300 \quad 0.005274 \quad -0.108300 \quad -0.1083000 \quad -0.108300 \quad -0.108300 \quad -0.108300 \quad -0.108300 \quad -0.1083000 
                                   5 rows × 31 columns
 In [5]: np.unique(df2["Class"].values)
Out[5]: array([0, 1], dtype=int64)
In [6]: df11 = df2[df2.columns[1:]]
    df21 = df2[df2.columns[:-1]]
In [7]: X1 = df11.values
X2 = df21.values
```

```
In [8]: X1.shape, X2.shape
Out[8]: ((284807, 30), (284807, 30))
          Country Data for k = 2
 In [9]: model = KMeans(n_clusters=2, n_init=4, verbose=1, max_iter=3)
In [10]: model.fit(X1)
          Initialization complete
          Iteration 0, inertia 15355477175.982346
          Iteration 1, inertia 11029741483.103212
          Iteration 2, inertia 10392052620.156
          Initialization complete
          Iteration 0, inertia 11114980702.64728
          Iteration 1, inertia 9264719549.548733
Iteration 2, inertia 9102721435.752607
          Initialization complete
          Iteration 0, inertia 12335933988.245417
          Iteration 1, inertia 10054582314.42063
          Iteration 2, inertia 9625872693.42902
          Initialization complete
          Iteration 0, inertia 12731282274.103073
          Iteration 1, inertia 10247849537.812023
          Iteration 2, inertia 9758780502.942104
Out[10]: KMeans(max iter=3, n clusters=2, n init=4, verbose=1)
```

```
In [11]: model.cluster_centers_
 Out[11]: array([[ 3.26057855e-02, 5.76572977e-02,
                                                            2.51826835e-02,
                     -1.17976378e-02, 3.44565898e-02, -1.85262634e-02,
                     -3.69604248e-02, 9.59436742e-03, 3.28556924e-03,
                      7.94181926e-03, -7.87280491e-04, 8.33871426e-04,
                     -5.53236011e-04, -3.21306631e-03, -6.21150418e-04,
                     -1.23735478e-03, -3.89169934e-04, -1.20998738e-03,
                      3.51151916e-03, -2.11103215e-02, -6.01010005e-03,
                      4.99521799e-03, 6.32069988e-03, -7.33291634e-04,
                      2.19754330e-03, -1.24054131e-04, -3.76137105e-04,
                     -9.19022240e-04, 7.06226133e+01, 1.71319122e-03],
                    [-3.19518565e+00, -5.65009452e+00, -2.46776293e+00,
                      1.15610289e+00, -3.37655418e+00, 1.81547079e+00,
                      3.62191609e+00, -9.40194650e-01, -3.21967514e-01,
                     -7.78254120e-01, 7.71491098e-02, -8.17147623e-02,
                      5.42140524e-02, 3.14862629e-01, 6.08692865e-02,
                      1.21253887e-01, 3.81364892e-02, 1.18572034e-01,
                     -3.44109348e-01, 2.06869410e+00, 5.88956381e-01,
                     -4.89503583e-01, -6.19393437e-01, 7.18585020e-02,
                     -2.15347022e-01, 1.21566150e-02, 3.68593445e-02,
                      9.00590685e-02, 1.82549762e+03, 3.12825860e-03]])
 In [12]: model.inertia
 Out[12]: 8993108022.120855
            Credit Card Data for k = 2
In [13]: model = KMeans(n_clusters=2, n_init=4, verbose=1, max_iter=3)
       model.fit(X2)
        Initialization complete
        Iteration 0, inertia 133197499804769.97
        Iteration 1, inertia 118387122772311.6
        Iteration 2, inertia 118320729057132.44
        Converged at iteration 2: center shift 4116.304743573539 within tolerance 7517.262262351321.
        Initialization complete
        Iteration 0, inertia 217397134618389.03
        Iteration 1, inertia 119969342982985.11
        Iteration 2, inertia 118332620375948.28
        Initialization complete
        Iteration 0, inertia 186568546424018.4
        Iteration 1, inertia 119451025740449.81
        Iteration 2, inertia 118329474501541.7
        Initialization complete
        Iteration 0, inertia 179793823891819.7
        Iteration 1, inertia 118406789111879.89
       Iteration 2, inertia 118323452868119.0
Out[13]: KMeans(max_iter=3, n_clusters=2, n_init=4, verbose=1)
```

```
In [14]: model.cluster centers
Out[14]: array([[ 1.41083313e+05,
                                   2.73767049e-01, -5.15444169e-02,
                  -7.06911008e-01, -1.55103617e-01, 2.79380502e-01,
                  -8.09820247e-02, 1.21060177e-01, -6.12691562e-02,
                  4.03453498e-02, 3.45519063e-02, -2.43862400e-01,
                   3.77698228e-02, -3.37779537e-02, -6.54987759e-02,
                  -2.12086853e-01, 1.37932481e-02, -5.10405524e-02,
                   8.24898455e-02, 2.46551972e-02, -4.59766665e-02,
                   4.32965774e-02, 1.27668887e-01, 3.32715680e-02,
                  -1.36283144e-02, -1.35737230e-01, -2.67064777e-02,
                  -2.18643134e-03, -3.32139465e-03, 8.80042894e+01],
                 [ 5.50524450e+04, -2.35260292e-01, 4.42944270e-02,
                   6.07480303e-01, 1.33287488e-01, -2.40084184e-01,
                   6.95914823e-02, -1.04032434e-01, 5.26513312e-02,
                  -3.46705668e-02, -2.96920012e-02, 2.09561887e-01,
                  -3.24573010e-02, 2.90269091e-02, 5.62860329e-02,
                   1.82255736e-01, -1.18531561e-02, 4.38614336e-02,
                  -7.08872202e-02, -2.11873157e-02, 3.95098096e-02,
                  -3.72066889e-02, -1.09711595e-01, -2.85917491e-02,
                   1.17114212e-02, 1.16645083e-01, 2.29500729e-02,
                   1.87889841e-03, 2.85422324e-03, 8.86463767e+01]])
In [15]: model.inertia
Out[15]: 118320095020223.72
         Country Data for k = 5
In [16]: model = KMeans(n clusters=5, n init=4, verbose=1, max iter=3)
         model.fit(X1)
         Initialization complete
         Iteration 0, inertia 2889028093.155498
         Iteration 1, inertia 2717269558.1921086
         Iteration 2, inertia 2704110391.975314
         Initialization complete
         Iteration 0, inertia 3295333163.6207714
         Iteration 1, inertia 2869522403.4372406
         Iteration 2, inertia 2832973997.1933174
         Initialization complete
         Iteration 0, inertia 3025270588.6172094
         Iteration 1, inertia 2909840001.0567784
         Iteration 2, inertia 2874583597.3048234
         Initialization complete
         Iteration 0, inertia 3040410093.5436487
         Iteration 1, inertia 2717410920.4940886
         Iteration 2, inertia 2693101368.7125025
Out[16]: KMeans(max_iter=3, n_clusters=5, n_init=4, verbose=1)
```

```
In [17]: model.cluster_centers_
Out[17]: array([[ 9.83651344e-02, 2.54103108e-01, 6.04003529e-02,
                       1.70041712e-02, 1.27166399e-01, -6.45736185e-02,
                     -8.38463653e-02, 2.53674301e-02, 1.18147742e-02, 2.45559224e-02, 5.11664845e-03, -5.22905530e-03,
                       6.41475476e-03, -4.40675480e-03, 1.05158842e-02,
                      2.29428429e-02, -4.59008062e-03, -1.39598504e-02, 1.08309624e-02, -5.92953649e-02, -1.97182136e-02,
                      1.03818799e-03, 6.74124616e-03, 2.97953032e-03, 1.75356983e-03, -1.64555584e-03, 2.97617850e-03,
                      -1.46491495e-03, 3.16655816e+01, 1.59640849e-03],
                    [-3.70843022e+00, -6.38774863e+00, -2.80870489e+00,
                       1.32446444e+00, -3.81013619e+00, 2.09938947e+00,
                       4.12031962e+00, -1.12419876e+00, -2.94638143e-01,
                      -8.68205805e-01, 4.89664347e-02, -9.68806826e-02,
                       9.78763645e-02, 3.40398805e-01, 9.45513707e-02,
                     1.77688719e-01, 4.07348875e-02, 1.55725960e-01, -4.07262315e-01, 2.37250672e+00, 7.02706113e-01,
                     -5.78397900e-01, -7.62235684e-01, 9.56489667e-02,
                      -2.85532479e-01, -1.01650919e-02, 5.41269191e-02,
                    9.12386135e-02, 2.07124164e+03, 2.23048327e-03],
[-3.10957233e-01, -1.11101288e+00, -1.20167167e-01,
                      2.16317855e-02, -5.01885500e-01, 2.48264810e-01,
                       1.54256143e-01, -7.00612325e-02, -4.37929135e-02,
                      -7.79515240e-02, -4.54305461e-02, 4.89125795e-02,
                      -4.97971230e-02, -1.13612576e-02, -6.95372340e-02,
                     -1.79293164e-01, 2.27674197e-02, 1.01438459e-01, -4.33594737e-02, 1.51677950e-01, 6.53152144e-02,
                      5.41138015e-02, 1.52359310e-02, -2.67435074e-02, 1.26740819e-02, 7.45773661e-03, -2.61530043e-02,
                       2.83325223e-03, 2.78443409e+02, 2.26011234e-03],
                    [-1.00340340e+01, -1.52117704e+01, -7.44986489e+00,
                       3.91149207e+00, -1.17452096e+01, 6.79241210e+00,
                      1.31767681e+01, -2.84844360e+00, -6.84045575e-01, -2.79192665e+00, 3.53040705e-01, -5.03014438e-01,
                       6.00109117e-01, 5.92747029e-01, 7.29773808e-01,
                     1.28400942e+00, -1.76304914e-01, 4.77845733e-01, -1.16044822e+00, 5.33346697e+00, 1.35002813e+00,
                      -1.60514740e+00, -1.86909508e+00, 2.38337193e-01,
                      -7.61303979e-01, -2.71437588e-01, 5.89841604e-01,
                       1.10187178e-01, 5.56343753e+03, 2.16840434e-19],
                    [-1.31802813e+00, -2.87535853e+00, -1.00018534e+00,
                       3.88334705e-01, -1.47011877e+00, 7.24064284e-01,
                       1.33647871e+00, -3.46292734e-01, -1.71545086e-01,
                     -3.33045022e-01, -2.07692763e-03, 3.52741020e-03, -4.73464181e-02, 1.37751503e-01, -1.14694799e-01,
                     -1.08085396e-01, 6.51827229e-02, 1.77667089e-02,
```

## Github link-

https://github.com/avnish9898/Ml-

Experiment/blob/main/Machine%20Learning%20Experiment%208.ipynb