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K-Means Clustering Algorithm for Analysis Cyber **Crime Data**

Nagendra Kumar Sahu

Research Scholar Kalinga University, New Raipur (C.G.)

Dr. Sanjay Kumar Supervisor **Associate Processor**

Kalinga University, New Raipur (C.G.)

ABSTRACT

Cyber Crime is technology based crime committed by technocrats. This paper deals with Variants of cyber crime held in Chhattisgarh between 2005to 2013. Under this, the Age wise Clustering of arrested people has been displayed on basis of cybercrime in Chhattisgarh. Data mining k-Means algorithm is used for clustering. In k-means clustering, we are given a set of n data points in ddimensional space Rd and an integer k and the problem is to determine a set of k points in Rd, called centers, so as to minimize the mean squared distance from each data point to its nearest center. Python Software has been used to implement the K-Mean Algorithm in cyber crime dataset.

Key Words: Cyber Crime, Types of Cyber Crime, k-Mean Algorithm, Python, Cyber Crime Dataset, Result Analysis.

1. Introduction

Cyber crime always involves some degree of infringement on the privacy of others or damage to computer-based property such as files, web pages or software. This paper is completely focused on cyber crime case register and number of person arrested in Chhattisgarh. The paper also includes Chhattisgarh cybercrime Statistics according age wise people arrested. According to the age of the arrested person based on cyber crime in Chhattisgarh from 2005 to 2013, the clustering has been made through the k- mean algorithm which is based on cyber crime dataset. We can do k-means algorithm using python.

2. Methodology

Cluster analysis or clustering is the process of partitioning aset of data objects into subsets. Each subset is a cluster, such that objects in a cluster resemble one another, yet dissimilar toobjects in other clusters. In this context, k -Means clustering methods may generate different clustering's on the Cyber Crime dataset. The partitioning is not performed by humans, but by the clustering algorithm K-mean clustering algorithms were used for formation of clusters on cyber crime database. The data was collected from the National crime record bureau (2005 to 2013) data set converted into iris dataset using in python. The data set contains the various instances and the 6 attributes. The attributes are year, Crime type (act according), People arrested, age, Crime type. The algorithm is used in following manner:

K-Means: Technique

The k-means algorithm defines the centroid of a cluster as the mean value of the points within the cluster. First, it randomly selects k of the objects in D, each of which initially represents a cluster mean or center. For each of the remaining objects, an object is assigned to the cluster to which it is the most similar, based on the Euclidean distance between the object and the cluster mean. The k-means algorithm then iteratively improves the within-cluster variation. For each cluster, it computes the new mean using the objects assigned to the cluster in the previous iteration clusters formed in the current round are the same as those formed in the previous round. The k-means procedure along with algorithm given below. :

Input = K:

The number of clusters= D: A dataset containing n objects Output = A set of K clusters Method:

- 2. Arbitrarily choose K-objects from D as the initial cluster centers
- 3. Repeat
- Re-assign each object to the cluster to which the object is the most similar, based on the 4. mean value of the objects in the cluster
- 5. Update the cluster means, i.e. calculate the mean value of the objects for each clusters
- Until no changer 6.

The time complexity of the k-means algorithm is O(nkt), where n is the total number of objects, k is the number of clusters, and t is the number of iterations. Normally, $k \ll n$ and $t \ll n$. Therefore, the method is relatively scalable and efficient in processing large data sets.

3. Technology & Dataset

K-Means Clustering is one of the popular clustering algorithms. The goal of this algorithm is to find groups (clusters) in the given data. We implement K-Means algorithm using Python packages: pandas, NumPy, scikit-learn, Seaborn and Matplotlib.

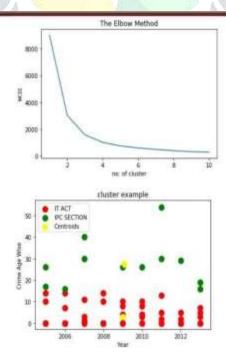
- 1. Pandas: Pandas is used to working with "relational" or "labeled" data both easy and intuitive. It aims to be the fundamental high-level building block for doing practical, real world data analysis in Python.
- 2. Numpy: NumPy is used for N-dimensional array object and sophisticated (broadcasting) functions.
- 3. Scikit-learn:Scikit-learn provideK-Mean algorithms via a consistent interface in Python.
- 4. 1Seaborn: Seaborn is use for data visualization and a high-level interface for drawing attractive and informative statistical graphics.
- 5. Matplotlib:Matplotlib is used for 2D plotting library which produces publication quality figures in a variety of hard copy formats and interactive environments across platforms.

The data was collected from the National crime record bureau (2005 to 2013) data set converted into iris dataset using in python. The data set contains the various instances and the 6 attributes. The attributes are year, Crime type (act according), People arrested, age, Crime type. Describe below:

```
print("***** CyberCrimeRecord *****")
In [11]:
          print(cty.head(20))
          print("\n")
             ***** CyberCrimeRecord *****
                        CrimeType
                                     People Arested
                                                          AGE Case Register
             0
                 2005
                             ITACT
                                                  14
                                                         1t30
                                                                         18.0
                 2005
                             ITACT
                                                  10
                                                       30to45
                                                                          NaN
             2
                 2005
                             ITACT
                                                       45to60
                                                                          NaN
             3
                 2005
                             ITACT
                                                   a
                                                                          NaN
                                                      above60
             4
                 2005
                       IPCSection
                                                  17
                                                         1t30
                                                                          28.0
             5
                 2005
                       IPCSection
                                                  26
                                                       30to45
                                                                          NaN
             6
                 2005
                       IPCSection
                                                   0
                                                       45to60
                                                                          NaN
            7
                 2005
                       IPCSection
                                                   0
                                                      above60
                                                                          NaN
             8
                 2006
                             ITACT
                                                         1t30
                                                                           0.0
             9
                 2006
                             ITACT
                                                   0
                                                       30to45
                                                                          NaN
             10
                 2006
                                                   0
                             ITACT
                                                       45to60
                                                                           NaN
             11
                 2006
                             ITACT
                                                   0
                                                      above60
                                                                           NaN
             12
                 2006
                       IPCSection
                                                  16
                                                         1t30
                                                                          30.0
                 2006
                       IPCSection
                                                  14
                                                       30to45
                                                                          NaN
            13
             14
                 2006
                       IPCSection
                                                       45to60
                                                                           NaN
             15
                 2006
                       IPCSection
                                                      above60
                                                                           NaN
             16
                 2007
                             ITACT
                                                   1
                                                         1+30
                                                                           5.0
                                                   3
             17
                 2007
                             ITACT
                                                       30to45
                                                                           NaN
             18
                 2007
                             ITACT
                                                       45to60
                                                                           NaN
                 2007
                             ITACT
                                                   0
            19
                                                      above60
                                                                           NaN
```

4. Result

In every model, the accuracy and the cost analysis plays an important role in the acceptance of model for the application. The result of the cyber crime data set is being displayed as a cluster form. The result is displayed o the basis of axis x-axis represent year and y-axis represents age of arrest people and the cluster are represent in form of dot yellow and green ,centroids cluster represent as yellow color.



Conclusion

This paper presents a k-Mean clustering using python. It is taking cyber crime dataset (Chhattisgarh) from (2005 to 2013) and classification of peoples arrested in that year by cluster. it also helpful for other prescribe dataset.

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