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Advances in Computer and Computational Sciences

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Role of Clustering in Crime Detection: Application of Fuzzy K-means

Nidhi Tomar and Amit Kumar Manjhvar

Abstract The rising rate of crime has devastated everything seriously. The reason of working on crime dataset is to make a better system, which can make people more aware about the increasing type of crime and crime rate in various fields. The proposed paper works on the detection of crime count and the factors that decide the increasing nature of crime in a better way. Utilization of fuzzy k-means has lead to a better technology that detects the crime rate in a better and effective way. The termination measure is an important factor to define the clusters that are formed over the years. They help in easy detection whether the crime is increasing over the years or not. The dataset from the Indian government's website is taken and been processed so that the results that are calculated can be as correct and near to reality as possible.

Keywords Data mining • Crime detection • Fuzzy • K-means • Clustering • Termination measure

1 Introduction

Crime is a very serious problem in the world. Malefaction is a crime against the society that is often prosecuted and realizable by the law [1]. Criminals commit crime at the place anywhere in type. Traditionally solving crime has been the privilege of the crime equity and law enforcement specialists. With the incrementing utilization of the computerized system to track crime, computer data analysis has commented availing the law enforcement [2]. Many challenges are increasing encountered by decision-makers in the law enforcement department in detecting, identifying the public crime or and tracing or tracking the social crime or

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actions according to their timeline is becoming a tedious task [3]. The Process of information divided into similar object groups known as clusters. Object consist that are similar to one another and exceptional to objects of other collection is called clustering in this implementing work we differentiate a cluster through a c-means clustering methods using the concept of fuzzy [4].

1.1 Fuzzy K-means Approach

It is the type of clustering algorithms, it is the procedure of partitioning the points of data into the clusters like k and S_l ($l = 1, 2, \dots, k$) and clusters S_l are related to representatives (cluster center) C_l . The relationship between a cluster and data points belong to a fuzzy [5]. That is, a membership is $u_{i,j} \in [0, 1]$ is used to show the degree is belong to data point X_i and cluster center C_j . Denote data points set as $S = \{X_i\}$. Fuzzy K-means method is based on minimizing following distortion:

Clustering methods Fuzzy k-means is used for division of points of records into the k clusters S_l ($l = 1, 2, \dots, k$) and clusters S_l are related with a representative (cluster center) C_l . The correlation between a data point and cluster representative is fuzzy [6]. That is a membership $u_{i,j} \in [0, 1]$ is used to show the degree which is belongings of data point X_i and cluster center C_j —Denote the set of data points as $S = \{X_i\}$. This algorithm of Fuzzy k-means is based on the minimizing the following distortion:

$$J = \sum_{j=1}^k \sum_{i=1}^N u_{i,j}^m d_{ij} \quad (1)$$

With respect to the memberships $u_{i,j}$ and cluster representatives C_j , where N is various data points; m is the fuzzifier parameter; k is numerous clusters; and d_{ij} is squared Euclidean distance between data points X_i and also representative of cluster C_j . It is noted that $u_{i,j}$ should satisfy the following constraint:

$$\sum_{j=1}^k u_{i,j} = 1, \text{ for } i = 1 \text{ to } N. \quad (2)$$

1.2 Working Steps of Fuzzy K-Means

- (1) Input is a set of initial cluster centers $SC_0 = \{C_j(0)\}$ and the value of is set $P = 1$.
- (2) Set of cluster centers SC_p , are given compute d_{ij} for $i = 1$ to N and $j = 1$ to K .
Update memberships value $u_{i,j}$ by the following equation:

$$u_{i,j} = \left((d_{ij})^{\frac{1}{m}-1} \sum_{l=1}^k \left(\frac{1}{d_{il}} \right)^{\frac{1}{m}-1} \right)^{-1} \quad (3)$$

If $d_{ij} > \eta$, set $u_{ij} = 1$, where η is a small positive number.

(3) Calculate center of all clusters applying Eq. (4). To find a new cluster set representatives SC_{p+1} .

$$C_j(P) = \frac{\sum_{i=1}^N u_{ij}^m X_i}{\sum_{i=1}^N u_{ij}^m} \quad (4)$$

(4) If $\| C_j(p) - C_j(p-1) \| > \epsilon$ and for $j = 1$ to K , stop, where $\epsilon > 0$ is a very small positive number. Otherwise, set $P + 1 \rightarrow P$ and go to step 2.

The computational complexity of FKM in the form of phase 2 and 3. Though, the computational complexity of phase 3 is much less than that of phase 2. For that cause, this complexity, in various distance calculations terms, of FKM is $O(Nkt)$, where t is the amount of iterations [7].

2 Literature Survey

Wang Shunye et al. [8] Stimulated with the random determination problem of initial centroid and similarity measures, the researcher presented a make novel k-means methods of clustering dissimilarity based. The algorithm which proposed gives enhanced accuracy rates and results. Pallavi Purohit and Ritesh Joshi et al. [9] proposed an enhanced approach designed for original K-means algorithm due to its certain limitations. The main reason of this method is poor performance in the initial centroid random selection. The proposed algorithm deals with this problem and improves the performance and cluster quality of original k-means methods. It first finds out the closest data point by calculating Euclidian distance between each data point and then these points are deleted from population. Juntao Wang and Xiao log [10] discussed about an improved version of k-means clustering algorithm to deal with the problem of outlier detection of existing k-means algorithm. The proposed algorithm usages noise information filter to the deal with this issue [11]. Density based outlier detection method is applied to the data to be clustered so as to remove the outliers [12]. The motive of this method is that the outliers may not be engaged in computation of initial cluster centers.

3 Proposed Work

The proposed work show that the number of clusters formed in the graph displays the crime count yearwise in all the cities. The proposed algorithm is applied fuzzy k-means of the crime data which leads to cluster formation in a better way. The crime detection being an important aspect needs to be taken seriously. The main motive of providing this proposed is to highlight the use of clustering technique on real-world based dataset which is based on crime. The database contains crime type, location, city, crime id and few more columns. The database seems like this: (Fig. 1).

A flowchart representation has been shown below here for the proposed work (Fig. 2).

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S
2	8023285	HIN7449	02-25-2008 11:00:00 PM	JAPAN	414	BATTERY	DOMESTIC BATTERY SIMPLE GAS STATION	False	True	7528	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
3	8023247	HIN7450	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE STREET	False	True	7529	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
4	8023248	HIN7451	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE STREET	False	True	7530	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
5	8023278	HIN7454	02-25-2008 11:00:00 PM	JAPAN	7433	BURGLARY	DOMESTIC BATTERY SIMPLE RESIDENCE-GANG VALUE	False	True	7531	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
6	8023238	HIN7455	02-25-2008 11:00:00 PM	JAPAN	5663	ASSAULT	DOMESTIC BATTERY SIMPLE RESIDENCE	False	True	7532	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
7	8023239	HIN7456	02-25-2008 11:00:00 PM	JAPAN	5663	ASSAULT	DOMESTIC BATTERY SIMPLE RESIDENCE	False	True	7533	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
8	8023430	HIN7458	02-25-2008 11:00:00 PM	JAPAN	5663	BATTERY	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7534	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
9	8023445	HIN7459	02-25-2008 11:00:00 PM	JAPAN	5663	BATTERY	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7535	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
10	8023446	HIN7460	02-25-2008 11:00:00 PM	JAPAN	120	CRIMINAL DAMAGE	DOMESTIC BATTERY SIMPLE APARTMENT	False	True	7536	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
11	8023238	HIN7461	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE STREET	False	True	7537	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
12	8023239	HIN7462	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE STREET	False	True	7538	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
13	8023239	HIN7463	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE STREET	False	True	7539	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
14	8023239	HIN7464	02-25-2008 11:00:00 PM	JAPAN	120	CRIMINAL DAMAGE	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7540	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
15	8023239	HIN7465	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7541	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
16	8023280	HIN7466	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7542	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
17	8023414	HIN7467	02-25-2008 11:00:00 PM	JAPAN	2094	ROBBERY	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7543	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
18	8023239	HIN7468	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7544	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
19	8023239	HIN7469	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7545	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
20	8023272	HIN7470	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7546	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
21	8023239	HIN7471	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7547	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
22	8023239	HIN7473	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7548	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
23	8023239	HIN7474	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7549	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
24	8023239	HIN7475	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7550	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
25	8023239	HIN7476	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7551	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
26	8023239	HIN7477	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7552	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
27	8023491	HIN7478	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7553	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
28	8023295	HIN7479	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7554	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
29	8023239	HIN7480	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7555	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
30	8023492	HIN7481	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7556	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
31	8023239	HIN7482	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7557	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
32	8023239	HIN7483	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7558	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
33	8023239	HIN7484	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7559	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
34	8023239	HIN7485	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7560	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
35	8023224	HIN7486	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7561	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
36	8023239	HIN7487	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7562	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
37	8023239	HIN7488	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7563	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
38	8023239	HIN7489	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7564	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
39	8023239	HIN7490	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7565	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
40	8023239	HIN7491	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7566	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
41	8023239	HIN7492	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7567	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
42	8023239	HIN7493	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7568	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
43	8023239	HIN7494	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7569	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
44	8023239	HIN7495	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7570	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
45	8023239	HIN7496	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7571	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
46	8023239	HIN7497	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7572	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
47	8023239	HIN7498	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7573	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
48	8023239	HIN7499	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7574	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
49	8023239	HIN7500	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7575	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
50	8023239	HIN7501	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7576	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
51	8023239	HIN7502	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7577	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
52	8023239	HIN7503	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7578	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
53	8023239	HIN7504	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7579	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00	02:00:00
54	8023239	HIN7505	02-25-2008 11:00:00 PM	JAPAN	414	WEAPONS VIOLATION	DOMESTIC BATTERY SIMPLE SIDEWALK	False	True	7580	2008	02-25-2008	02:00:00	02:00:00	02:00:00	02:00:00		

Proposed Algorithm Pseudocode:**Pseudo_Yf_FKMC1 (X, c, options, init_V)**

1. Define maximum iterations $max_it = 100$;
termination threshold = $1.0e+03^*$;
 $initial_velocity = 0$.
2. Define n : Number of feature vectors
 p : Length of each feature vector
3. if use_init_V,
 $V = init_V$;
else
 $V = Yf_FKMC1_InitV (c, p);$ % Initial cluster centers
end
4. for $i = 1:max_iter$,
 $[V, U, E(i)] = Yf_FKMC1_Step (X, V, c, m);$
Show iteration count & termination measure value. End.

Pseudo_Yf_FKMC1_InitV(c, p)

Generate initial cluster centers for FKM clustering using formula:

$V = rand(c, p)$

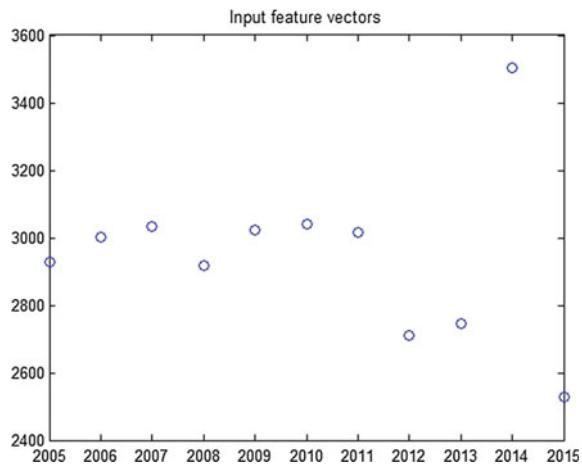
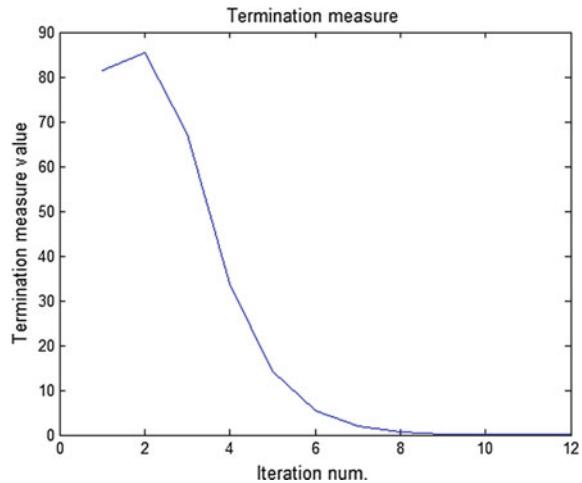
Pseudo_Yf_FKMC1_Step(X, V, c, m)

1. Initialize $n = size (X, 1)$
 $p = size (X, 2)$, where, X is the input data of the crime
2. Distance calculation using Euclidean distance formula
 $dist = Yf_EuDistArrayOfVectors1 (V, X)$
3. Now calculating the new membership degrees using a variable temp.
 $tmp = dist.^{(-2/(m-1))}$
 $U = tmp ./ (ones(c, 1) * sum(tmp))$
4. Check constraint by checking
 $tmp = ((sum (U) - ones (1, n)) > 0.0001)$
5. Update V , mf , and E
 $mf = U.^m$ % MF matrix after exponential modification
 $V = mf * X ./ ((ones(p, 1) * sum(mf'))')$ % new center
 $E = norm (V - V_old, 1)$
6. End.

4 Result Analysis

The results show various factors and results:

1. Cluster formation in the base: these are the crime counts on an overall basis. The x-axis shows the year and y-axis shows the total number of crimes (Fig. 3).
2. Termination measure: this can be defined as the value of the termination measure that goes along with the number of iterations. The value denotes the

Fig. 3 Crime count yearwise**Fig. 4** Termination measure

termination of the proposed algorithm. The termination value is set initially and the algorithm is terminated when its value reaches a very less point. The termination value taken in proposed algorithm is $1.0e + 03^*$ (Figs. 3 and 4).

3. Membership function: membership function is used in fuzzy as the degree of truth and evaluation of extension of the outcomes. The figure shows the membership values for both base and proposed respectively. From the figure it can be derived as the fact that membership value for proposed is better than previous values (Fig. 5).
4. Cluster formation: Number of clusters defined = 2. The clusters formed are shown in the Fig. 6.

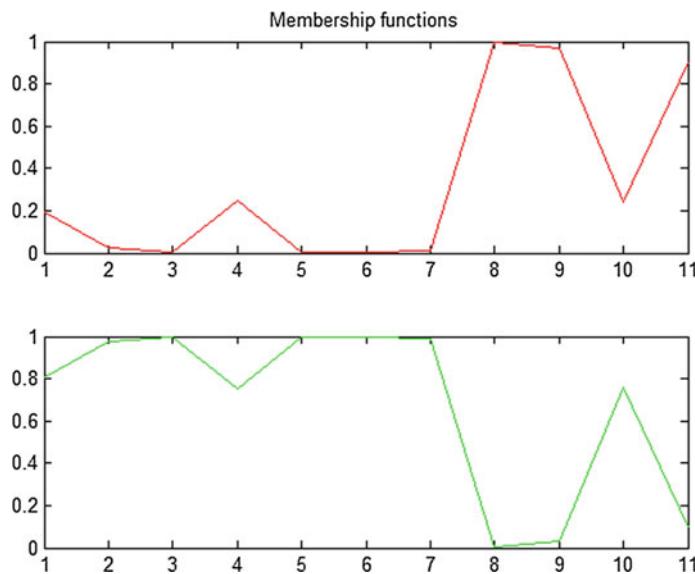


Fig. 5 Membership function

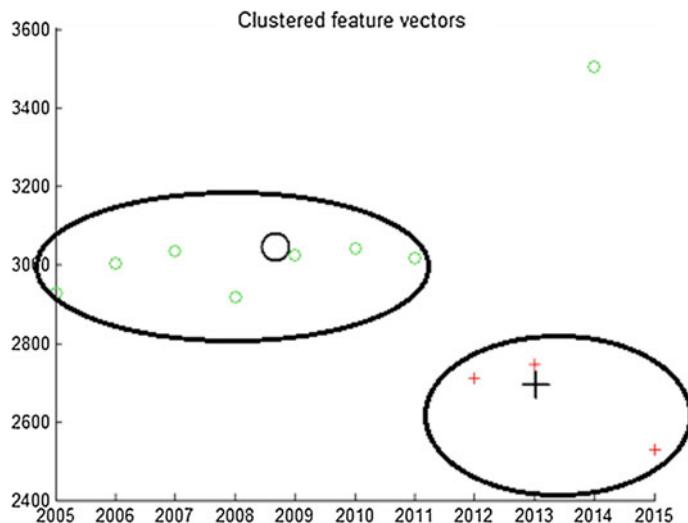


Fig. 6 Clustered feature vectors

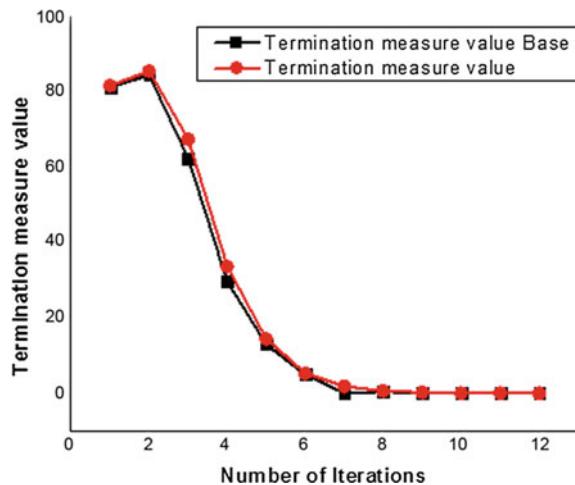
4.1 Results of FKMC

The results are generated in iterative manner which are shown below in tabular form:

Iteration	Termination measure value
1	81.444251
2	85.274352
3	67.243550
4	33.692983
5	14.442164
6	5.292763
7	1.842184
8	0.629867

4.2 Comparison Results

Fig. 7 Termination measure value graph



5 Conclusion

Crime detection plays a vital role in our lives for the reason that of the increasing rate of the crime happening in every area of the country. The crime is also of various types which has made it complex further. The government is not able to find a proper way out towards the removal of the crime and have better control over it. The proposed works above is a step ahead in this field. A further work needs also to be done so as to detect the category of crime, crime count, citywise distribution, increasing rate, etc. The various factors which affect crime rate are discussed.

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