### **Pattern Recognition**

### **Project Report**

# Clustering by fast search and find of density peaks

## **Implementation:**

- 1) Loading data set: We have taken 3 datasets and asked user to choose one
  - a) Crescent moon random dataset
  - b) Dataset given in IEEE paper
  - c) Flame dataset
- 2) Calculating the distance matrix ,we have used eucledian distance
- 3) Calculate cut off distance
  - Dc( cut off distance) depends up on Average number of neighbours is around 1 to 2 % of total number of points in the datasets
- 4) Calculating local density Rho
- 5) Calculating delta for each point delta is measured by computing the min distance between the point I and any other point with higher density
- 6) Based on rho and delta values we assign points to the cluster or as an outlier

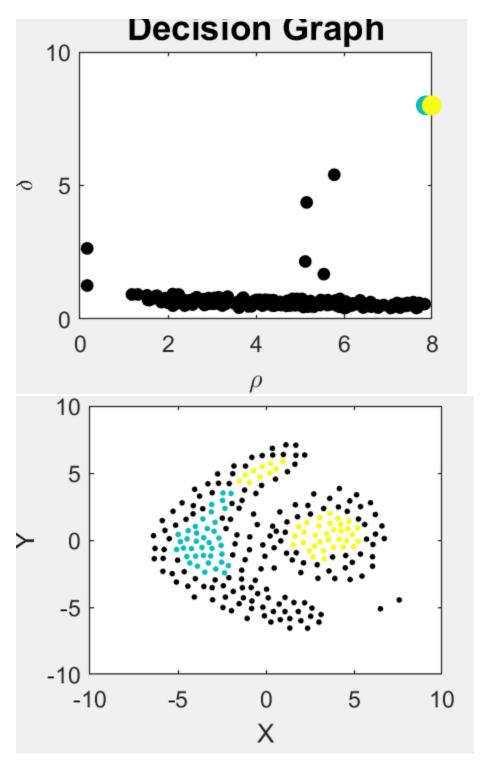
### **OUTPUTS**

**DataSet: Flame.txt** 

Enter 1 for flame dataset

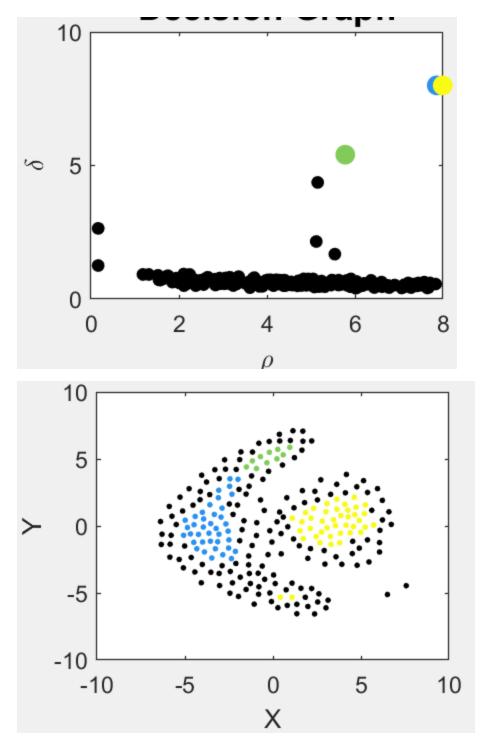
No of clusters: 2(pick from decision graph)

No of points:240



CLUSTER: 1 CENTER: 69 ELEMENTS: 102 CORE: 39 HALO: 63 CLUSTER: 2 CENTER: 230 ELEMENTS: 138 CORE: 43 HALO: 95

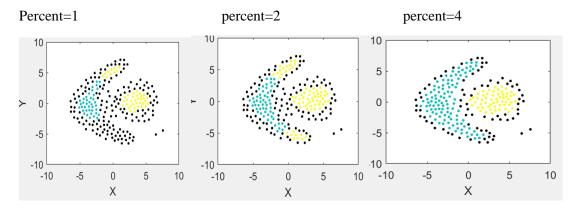
**Clusters: 3 (pick from decision graph)** 



CLUSTER: 1 CENTER: 69 ELEMENTS: 102 CORE: 39 HALO: 63 CLUSTER: 2 CENTER: 138 ELEMENTS: 32 CORE: 10 HALO: 22 CLUSTER: 3 CENTER: 230 ELEMENTS: 106 CORE: 41 HALO: 65

# Experiment1: changing percentage of average number of neighbours

no of clusters=2 we got the following result



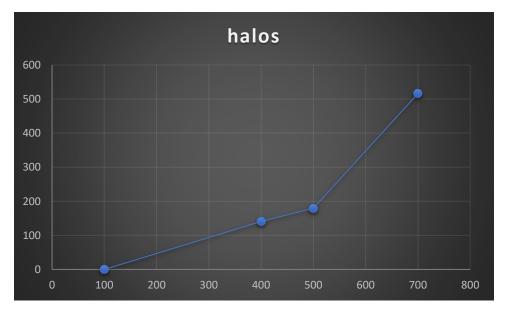
We observed for percentage of avg, number of neighbours is 4 we got better clustering.

## **Experiment2:**

Number of Halos(outliers) Vs number of Samples

X axis: number of samples

Y-axis: number of outliers



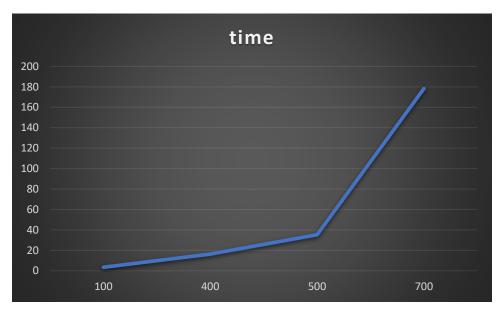
As number of samples increases halos increased.

### **Experiment3:**

### **ExecutionTime vs number of samples:**

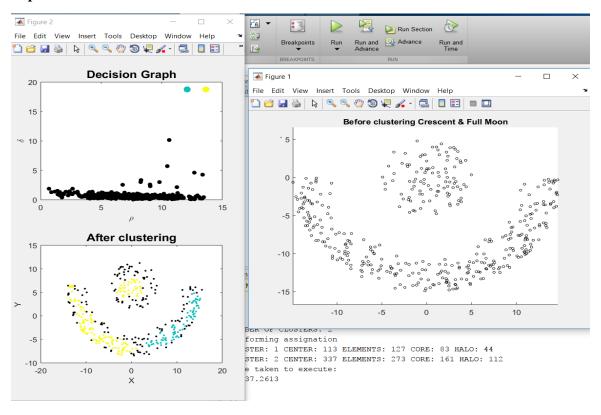
X-axis: number of sample

Y-axis: time in secs



As number of points increases execution time increases.

### **Experiment4:**



enter 1 for Random CrescentFullMoon data 2 for data set present in IEEE paper 3 for Flame data set :1

Generated file: DECISION GRAPH

column 1:Density

column 2:Delta

Select a rectangle region to pick cut off values of rho and delta for cluster centers

ans =

0.3000

NUMBER OF CLUSTERS: 2

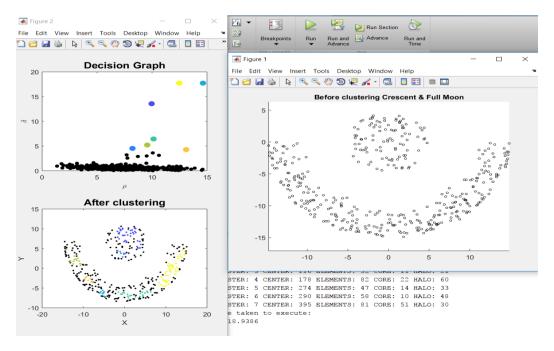
Performing assignation

CLUSTER: 1 CENTER: 113 ELEMENTS: 127 CORE: 83 HALO: 44

CLUSTER: 2 CENTER: 337 ELEMENTS: 273 CORE: 161 HALO: 112

time taken to execute:

37.2613



NUMBER OF CLUSTERS: 7

#### Performing assignation

CLUSTER: 1 CENTER: 2 ELEMENTS: 72 CORE: 41 HALO: 31

CLUSTER: 2 CENTER: 6 ELEMENTS: 28 CORE: 7 HALO: 21

CLUSTER: 3 CENTER: 116 ELEMENTS: 32 CORE: 11 HALO: 21

CLUSTER: 4 CENTER: 178 ELEMENTS: 82 CORE: 22 HALO: 60

CLUSTER: 5 CENTER: 274 ELEMENTS: 47 CORE: 14 HALO: 33

CLUSTER: 6 CENTER: 290 ELEMENTS: 58 CORE: 10 HALO: 48

CLUSTER: 7 CENTER: 395 ELEMENTS: 81 CORE: 51 HALO: 30

time taken to execute:

18.9386

#### **Advantages:**

1) We can select choose number of clusters dynamically

2) It is robust to number of instances

### **Limitations:**

- 1) We have taken 2 dimensions data only
- 2) Selecting rectangular region is for cut off rho and delta values
- 3) Varying cutoff distance produce consistent results if it is 1-2%

### **Drawbacks:**

- 1) The paper does not talk about noise
- 2) As number of samples increases performances reduces

#### **References:**

https://www.mathworks.com/matlabcentral/fileexchange/41459-6-functions-for-generatingartificial-datasets

http://cs.uef.fi/sipu/datasets/

https://www.mathworks.com/matlabcentral/fileexchange/53922-densityclust