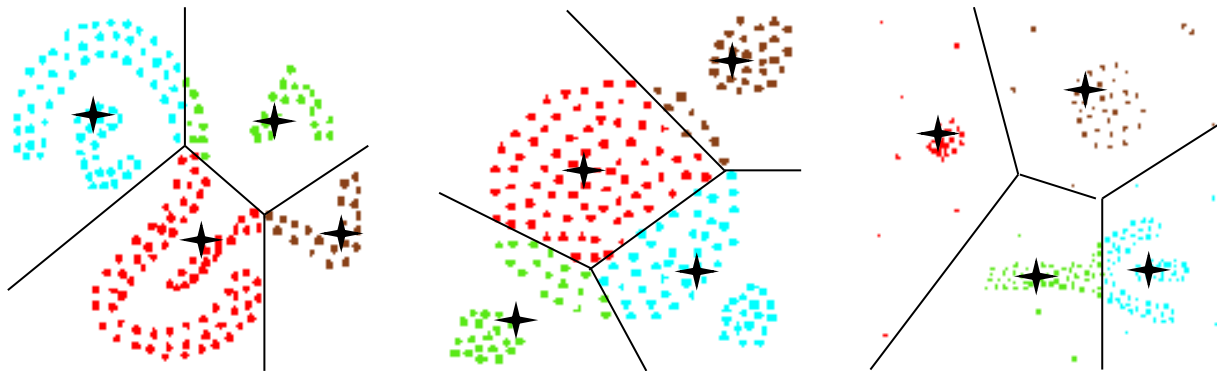
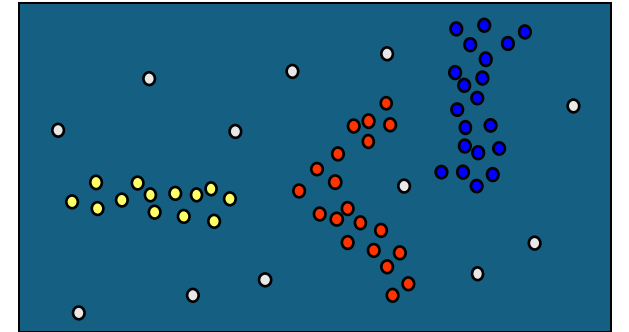


Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

- Most partitioning (clustering) methods cluster objects based on the distance between objects
- Such methods can find only spherical-shaped clusters and encounter difficulty at discovering of arbitrary shapes

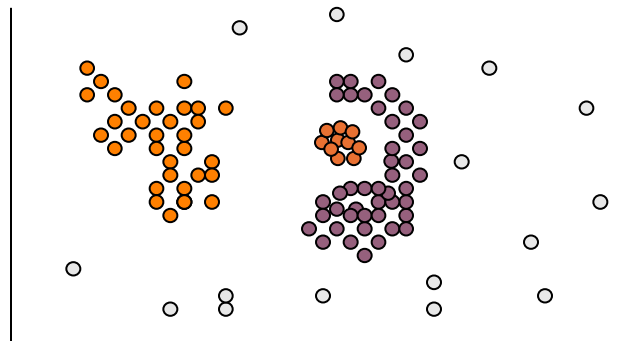


Results of a k -means algorithm for $k=4$

Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

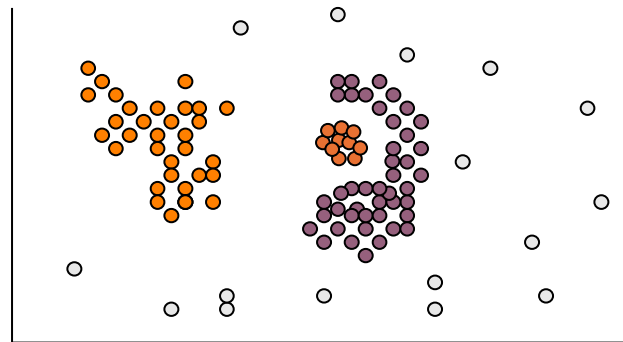
- Other methods have been developed based on the notion of density
- Their general idea is to continue growing the given cluster as long as the density (number of data points) in the neighbourhood exceeds some threshold



Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

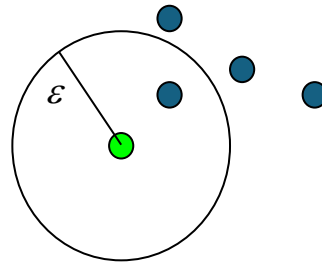
- That is, for each data point within a given cluster, the neighbourhood of a given radius has to contain at least a minimum number of points
- Such methods can be used to filter out noise (outliers), and discover clusters of arbitrary shape



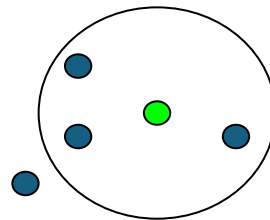
Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

- The neighbourhood within a radius ϵ of a given object (data point) is called the ϵ neighbourhood of the object.



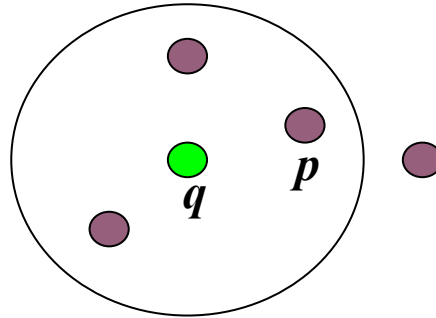
- If the ϵ neighbourhood of an object contains at least a minimum number, *MinPts*, of objects, then the object is called a **core** object.



Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

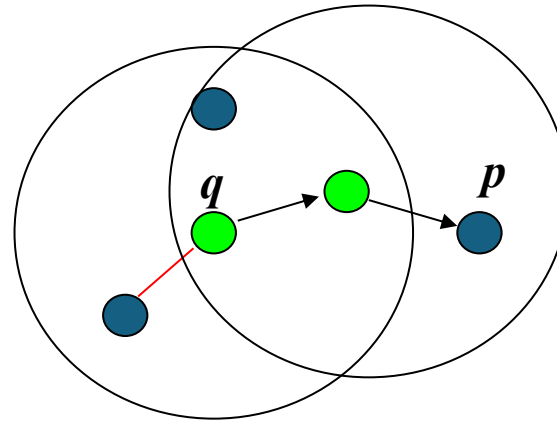
- We say that an object p , is **directly density-reachable** from a core object q if p is within the ϵ neighbourhood of q



Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

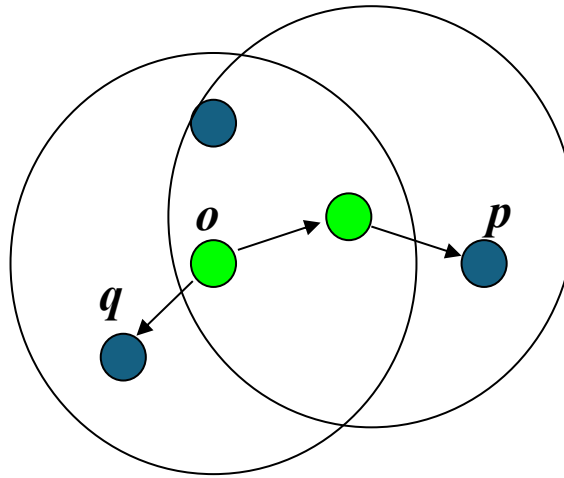
- An object p is **density-reachable** from object q , if there is a chain of objects p_1, \dots, p_n ; where $p_1 = q$ and $p_n = p$ such that p_{i+1} is directly density reachable from p_i



Density Based Clustering

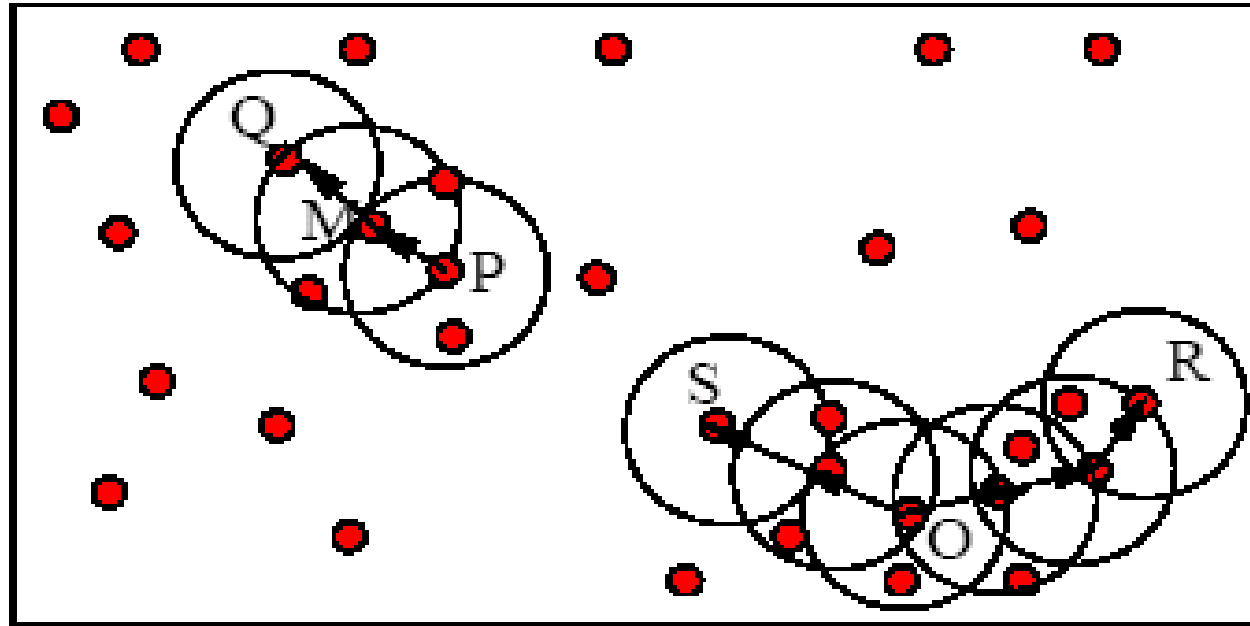
DBSCAN (Density based spatial clustering of applications with noise)

- An object p is **density-connected** to object q , if there is an object o such that both p and q are density reachable from o



Density Based Clustering

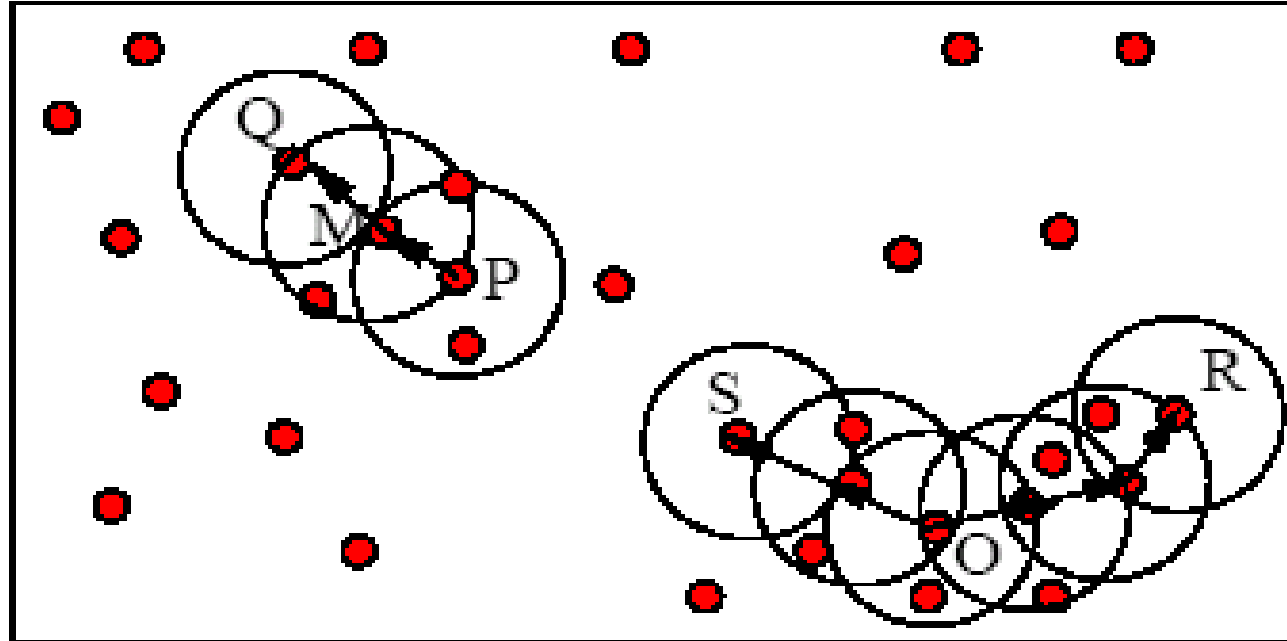
DBSCAN (Density based spatial clustering of applications with noise)



- M, P, O and R are **core** objects, if $MinPts = 3$
- P and Q are **directly density reachable** from M and M is directly density reachable from P

Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)



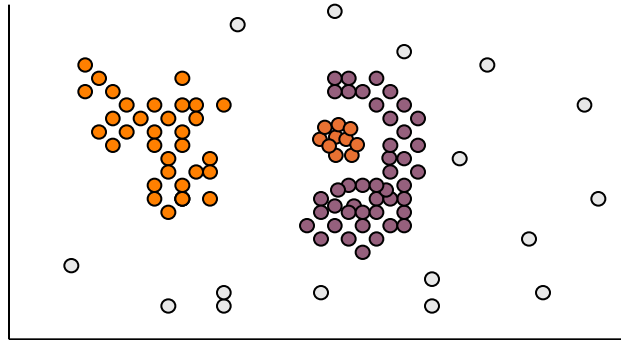
- Q is **density reachable** from P, but P is not density reachable from Q
- S, O, R are all **density connected**

Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

A **cluster** is a set of density-connected objects

Noise is the set of objects not contained in any cluster.



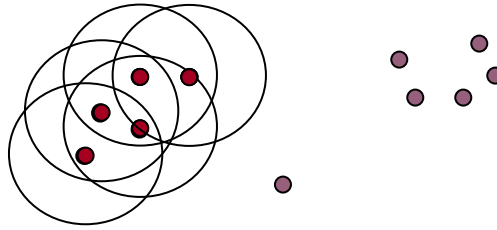
Density Based Clustering

DBSCAN (Density based spatial clustering of applications with noise)

- Arbitrary select a point p
- Retrieve all points density-reachable from p wrt Eps and $MinPts$.
- If p is a core point, a cluster is formed.
- If p is a border point, no points are density-reachable from p and DBSCAN visits the next point of the database.
- Continue the process until all of the points have been processed.

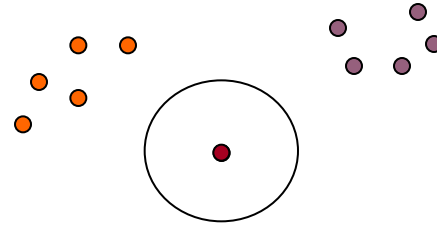
DBSCAN Algorithm: Example

- Parameter
 - $\varepsilon = 2$ cm
 - $MinPts = 3$



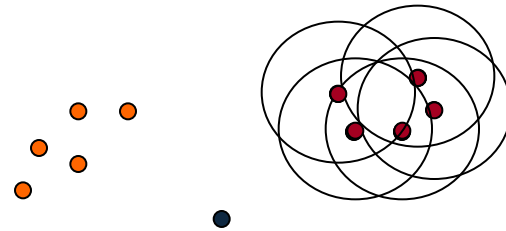
DBSCAN Algorithm: Example

- Parameter
 - $\varepsilon = 2$ cm
 - $MinPts = 3$



DBSCAN Algorithm: Example

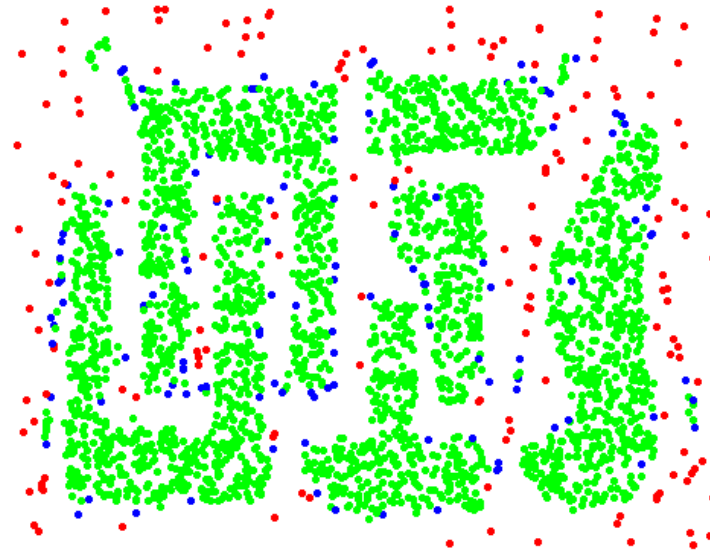
- Parameter
 - $\varepsilon = 2$ cm
 - $MinPts = 3$



Example



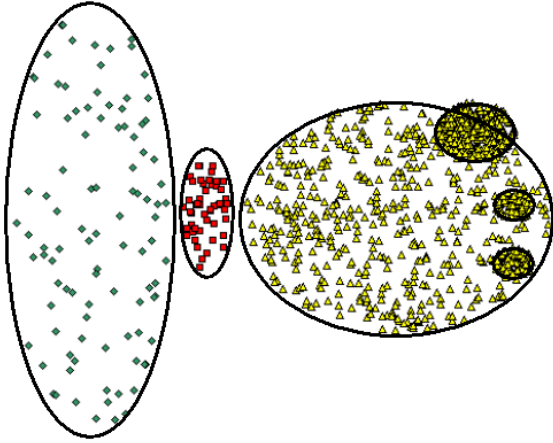
Original Points



Point types: **core**,
border and **outliers**

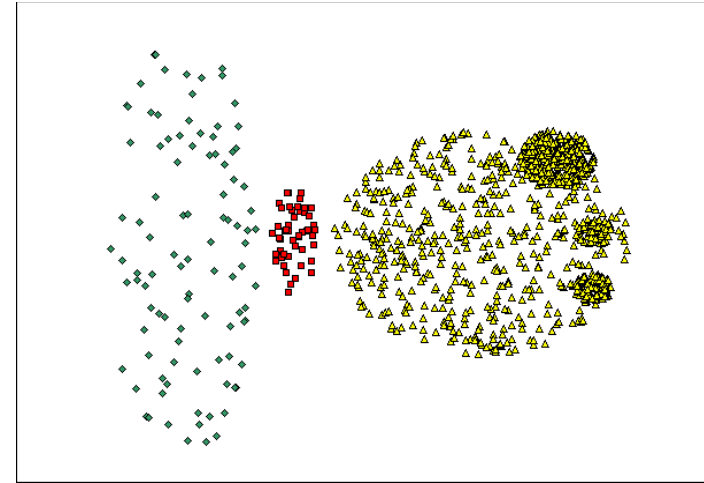
$\epsilon = 10$, MinPts = 4

When DBSCAN Does NOT Work Well

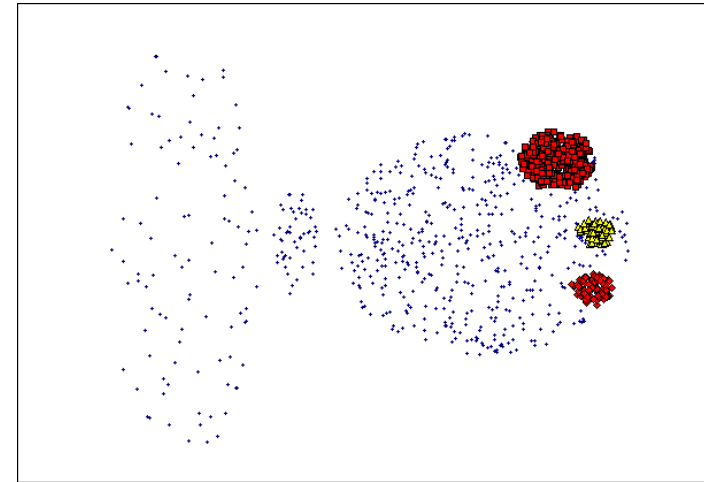


Original Points

- cannot handle Varying densities
- sensitive to parameters



(MinPts=4, Eps=9.92).



(MinPts=4, Eps=9.75)