

# Density Based Clustering

# Necessity

- Clusters with arbitrary shapes
- Clusters of different sizes
- Clusters with different densities
- Finding noise and outliers

# DBSCAN

- Find all the neighbor points within  $\epsilon$  and Identify the core points
- For each core point create a new cluster if it doesn't belong to any
- Find recursively all its density connected points and assign them to the same cluster as the core point
- A point  $a$  and  $b$  are said to be density connected if there exist a point  $c$  which has a sufficient number of points in its neighbors and both the points  $a$  and  $b$  are within the  $\epsilon$  distance. This is a chaining process. So, if  $b$  is neighbor of  $c$ ,  $c$  is neighbor of  $d$ ,  $d$  is neighbor of  $e$ , which in turn is neighbor of  $a$  implies that  $b$  is neighbor of  $a$ .
- Iterate through the remaining unvisited points in the dataset.
- Those points that do not belong to any cluster are noise

# Definitions

- ***r***: The radius of the neighborhoods around a data point
- ***minPts***: The minimum number of data points need to be a cluster
- ***Core Points***: A data point  $p$  is a core point if it has at least ***minPts*** neighbors
- ***Border Points***: A data point  $q$  is a border point if it is not a core point but a neighbor of core point
- ***Outlier***: A data point  $o$  is an outlier if it is neither a core point nor a border point

# Examples

Not Possible using K-Means

*Thank You*

# References

- <https://blog.dominodatalab.com/topology-and-density-based-clustering/>
- <https://www.kdnuggets.com/2020/02/understanding-density-based-clustering.html>
- <https://www.geeksforgeeks.org/dbscan-clustering-in-ml-density-based-clustering/>
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