



Unsupervised Learning

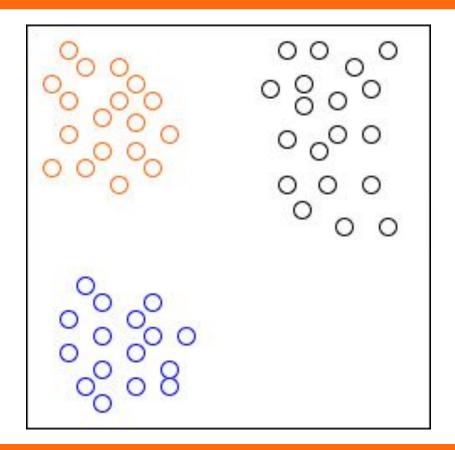
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Introduction To DBSCAN

- DBSCAN stands for Density-Based Spatial Clustering of Applications with Noise.
- 2. It also does not require the number of clusters to be told beforehand, unlike K-Means, where we have to specify the number of centroids.
- 3. K-Means (distance between points)
- 4. DBSCAN (distance between nearest points)
- 5. The key idea is that for each point of a cluster, the neighborhood of a given radius has to contain at least a minimum number of points.

Introduction To DBSCAN



Introduction To DBSCAN

- Partitioning methods K-means and hierarchical clustering work for finding spherical-shaped clusters or convex clusters.
- 2. But several data may contain irregularities, like
 - a. i) Clusters can be of convex shape or mixed manner it is difficult to separable out.
 - b. ii) Data may contain noise.
- 3. This algorithm does it by identifying different clusters in the dataset and connects the areas of high densities into clusters.

- DBSCAN requires only two parameters:
 - a. epsilon and minPoints.
 - Epsilon is the radius of the circle to be created around each data point to check the density.
 - ii. minPoints is the minimum number of data points required inside that circle for that data point to be classified as a Core point.

1. Epsilon:

- a. If the distance between two points is lower or equal to 'eps' then they are considered as neighbors.
- b. If it is chosen very large then the clusters will merge and majority of the data points will be in the same clusters.

minPoints:

- a. Minimum number of neighbors (data points) within eps radius.
- b. Larger the dataset, the larger value of MinPts must be chosen.
- c. The minimum value of MinPts must be chosen at least 3.

In this algorithm, we have 3 types of data points. :-

- a. Core Point: A point is a core point if it has more than MinPts points within eps.
- b. Border Point: A point which has fewer than MinPts within eps but it is in the neighborhood of a core point.
- c. Noise or outlier: A point which is not a core point or border point.

