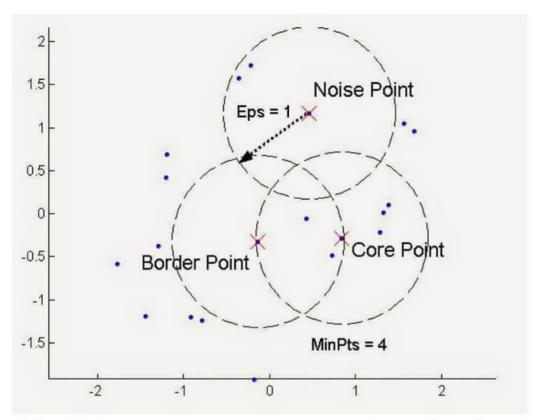
- It is a Density based clustering (Density-based spatial clustering of applications with noise)
- In density based clustering we partition points into dense regions separated by not-so-dense regions.
- DBScan works well even if your data contains nonconvex clusters or noise.
- In DBSCAN, there are no centroids, and clusters are formed by linking nearby points to one another.
- Important things in DBScan
- Parameters
- Eps
- Minpoints
- Types of data points
- Core point
- Border point
- Noise or Outlier

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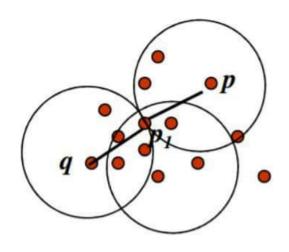
- eps: It defines the neighborhood around a data point i.e. if the distance between two points is lower or equal to 'eps' then they are considered as neighbors.
- If eps is small, then large no of points will be left out as noise
- If eps is large, then most of thepoints will be in same cluster
- We can find the best ps value using kdistance graph.

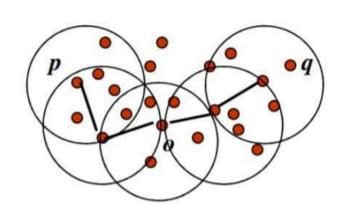
- MinPoints: Minimum number of data points within eps radius.
- Larger the dataset, the larger value of MinPts must be chosen.
- As a general rule, the minimum MinPts can be derived from the number of dimensions D in the dataset as, MinPts >= D+1.
- Or MinPts = 2*D



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

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Density edge

Density connected

- Density connected points
- Density edge: We place an edge between two core points q and p if they are within distance Eps.
- Density-connected: A point p is densityconnected to a point q if there is a path of edges from p to q.

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- steps
- Label points as core, border and noise
- Eliminate noise points
- For every core point p that has not been assigned to a cluster
- Create a new cluster with the point p and all the points that are density-connected to p.
- Assign border points to the cluster of the closest core point

- Pros
- Resistant to Noise
- Can handle clusters of different shapes and sizes
- Cons
- Varying densities
- High-dimensional data
- sensitive to parameters