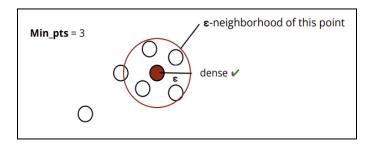
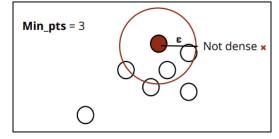
I. Density Based Clustering

A clustering method that groups together data points that are closely packed (high-density regions) and separates them from areas of low density (sparse regions or noise).

It's different from K-means because it doesn't assume a specific cluster shape or number of clusters — it finds clusters based on the density of points in space.

<u>How should we define density:</u> density is defined by counting how many points fall within a fixed radius (ϵ) around a point. If there are at least min_pts points in that area, **it's considered a dense region.**





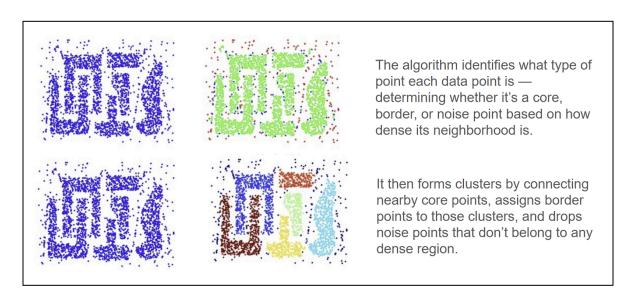
A. This is a dense region

B. This is not a dense region

But the datapoint in visual B was part of a dense section in visual A this shows that a point can appear "not dense" on its own (fewer than min_pts points within its ϵ -radius) but still belong to a dense cluster if it's near other dense (core) points.

So, we distinguish between three types of datapoints:

- Core point: has at least min_pts points within its ε-radius (the center of a dense region).
- Border point: doesn't meet the density requirement itself, but lies within the ε-radius of a core point.
- Noise point: is too far from any dense region not a core or border point, so it's treated as an outlier.



II. DBScan Algorithm

In DBScan, the parameters ϵ (the neighborhood radius) and min_pts (the minimum number of points needed to form a dense region) are given as inputs.

The algorithm steps are:

First take a datapoint:

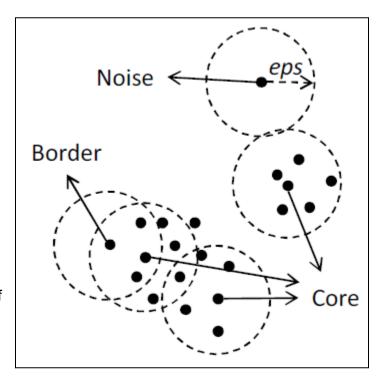
- Find the datapoint's ε-neighborhood (identify points within the chosen radius)
- 2. Check if it's a core point (those with at least min pts neighbors inside their ε-radius)

Then if it's a core point:

- 3. Iterate through neighborhood points to find any other core points within the neighborhood assign them to our cluster.
- 4. Then go to the next datapoint in the dataset (and restart steps)

But if it's not a core point:

- 5. Check if it's a border point, label it as such if it is
- 6. Otherwise, it's a noise point, so label it as such
- 7. Then go to the next datapoint in the dataset (and restart steps)



Lastly once we iterate through all points and perform the above steps on them:

8. Assign border points to their nearest cluster.

In short, DBSCAN builds clusters by chaining together dense areas and ignores sparse or isolated points as noise.