

# DBSCAN Algorithm

DBSCAN (Density-Based Spatial Clustering of Applications with Noise) is a density-based clustering algorithm that groups together points that are closely packed together while marking points that lie alone in low-density regions as outliers.

## Steps of the DBSCAN Algorithm

### 1. Initialization:

- Choose the parameters  $\epsilon$  (epsilon) and  $\text{min\_samples}$ .
  - $\epsilon$ : The maximum distance between two samples for them to be considered as in the same neighborhood.
  - $\text{min\_samples}$ : The minimum number of samples in a neighborhood for a point to be considered a core point.

### 2. Classify Points:

- For each point in the dataset, check if it is a core point:
  - A core point has at least  $\text{min\_samples}$  points (including itself) within  $\epsilon$  distance.

### 3. Expand Clusters:

- Start with an unvisited point and perform the following steps:
  - If the point is a core point, create a new cluster.
  - Add all points within  $\epsilon$  distance of the core point to the cluster.
  - Recursively visit each point in the cluster's neighborhood and add their neighbors to the cluster if they are also core points.

### 4. Handle Noise:

- Points that are not core points and not reachable from any core point are considered noise or outliers.

## Example

Consider a small dataset with the following points and parameters:  $\epsilon = 2$  and  $\text{min\_samples} = 3$ .

### Sample Dataset

Points: (1, 2), (2, 2), (2, 3), (8, 7), (8, 8), (25, 80)

## Step-by-Step Execution

### 1. Initialize Parameters:

- $\epsilon = 2$
- $\text{min\_samples} = 3$

## 2. Classify Points:

- Calculate the number of points within eps distance for each point.

## 3. Expand Clusters:

- For each unvisited point, check if it is a core point and expand the cluster.

## Detailed Execution

### 1. Point (1, 2):

- Neighbors within eps: (1, 2), (2, 2), (2, 3)
- Core point: Yes (3 neighbors including itself)
- Create a new cluster:  $C1 = \{(1, 2), (2, 2), (2, 3)\}$

### 2. Point (2, 2):

- Already visited and part of cluster C1.

### 3. Point (2, 3):

- Already visited and part of cluster C1.

### 4. Point (8, 7):

- Neighbors within eps: (8, 7), (8, 8)
- Core point: No (2 neighbors, less than min\_samples)
- Not a core point, check neighbors.
  - Point (8, 8) is a core point with neighbors (8, 7), (8, 8).
  - Expand cluster:  $C2 = \{(8, 7), (8, 8)\}$

### 5. Point (8, 8):

- Already visited and part of cluster C2.

### 6. Point (25, 80):

- Neighbors within eps: (25, 80)
- Core point: No (1 neighbor, less than min\_samples)
- Not a core point, mark as noise.

## Resulting Clusters and Noise

- Clusters:
  - C1:  $\{(1, 2), (2, 2), (2, 3)\}$
  - C2:  $\{(8, 7), (8, 8)\}$
- Noise:  $\{(25, 80)\}$