



Session 4

DBSCAN





Blue AI Team





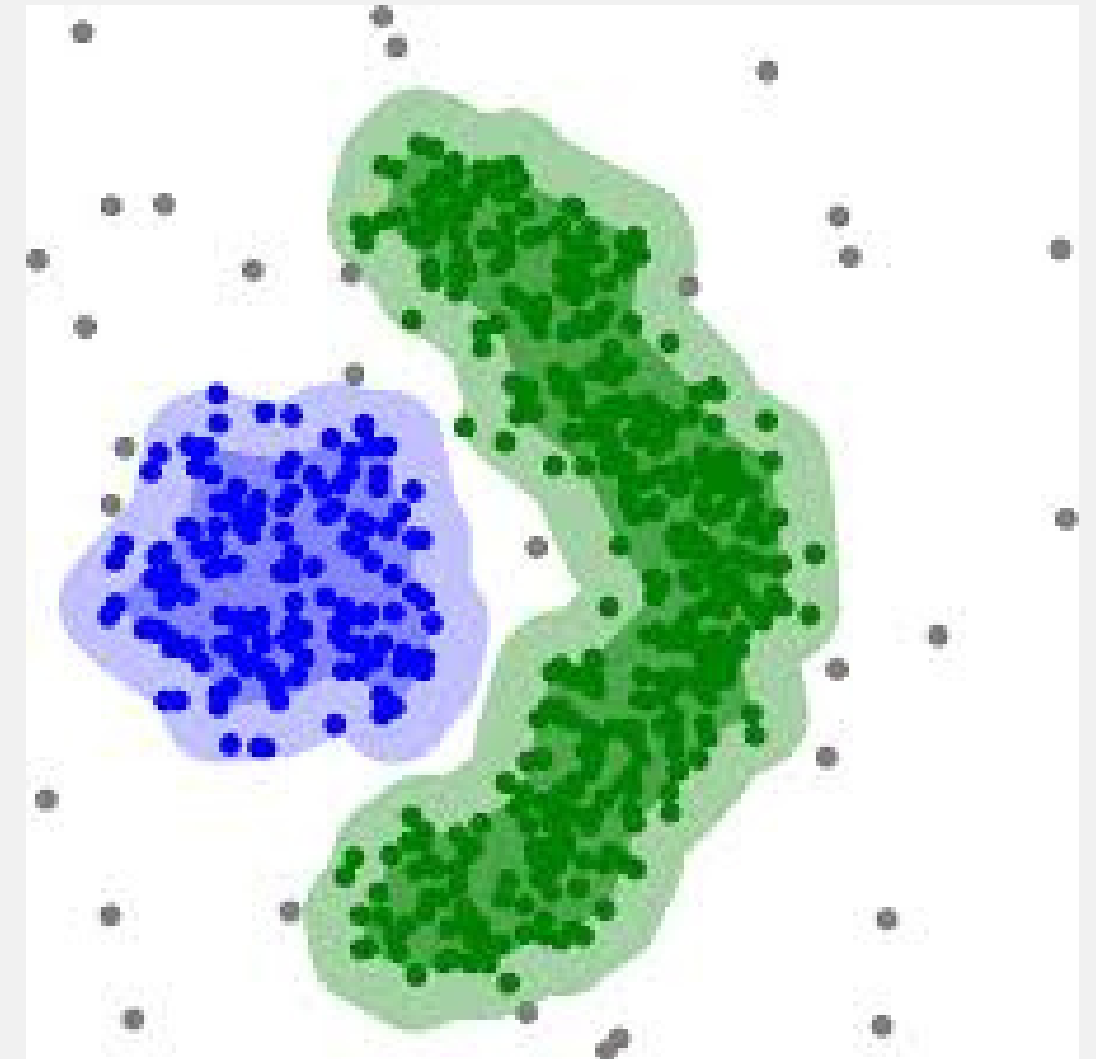
Collecting Samples

1. Simple Random
2. Systematic
3. Stratified
4. Cluster

- **Cluster Sampling** is a data collection technique.
 - **Machine learning clustering** is a pattern recognition technique.
- 
- 
- 
- 

What is DBSCAN?

- A density-based clustering algorithm.
- Clusters data points that are closely packed together
- Marks outliers as noise.
- automatically finds clusters based on density.



How Does DBSCAN Work?

DBSCAN works by categorizing data points into three types:

1 Core Points

- ✓ A core point has at least MinPts neighbors within a specified radius ϵ (epsilon).
- ✓ It forms the center of a cluster and helps expand it.

How Does DBSCAN Work?

1 Border Points

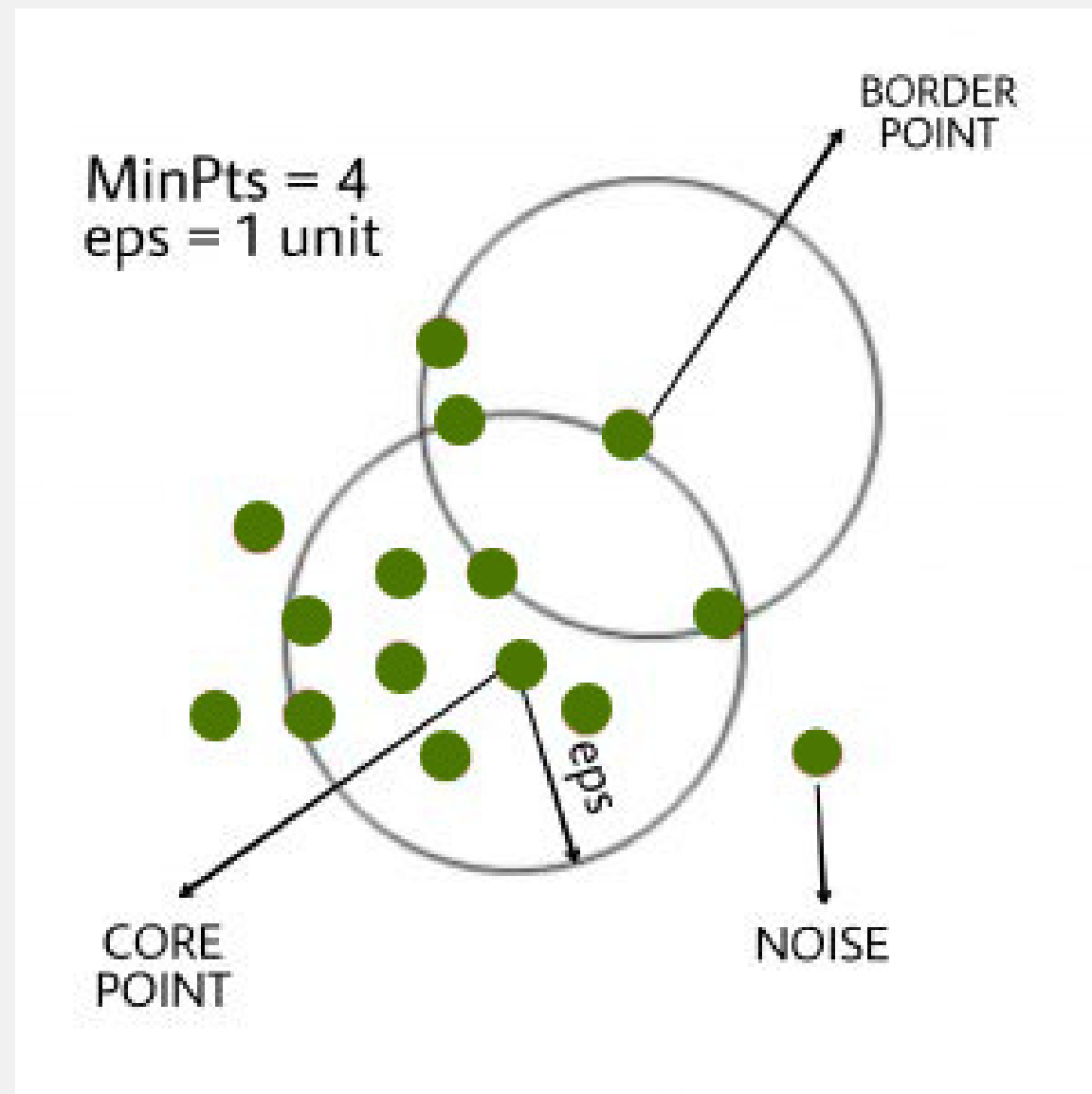
- ✓ A border point is close to a core point but does not have enough neighbors to be a core itself.
- ✓ It belongs to a cluster but does not expand it.

How Does DBSCAN Work?

1 Noise (Outlier) Points

- ✓ A noise point is isolated with too few neighbors within ϵ .
- ✓ It does not belong to any cluster.



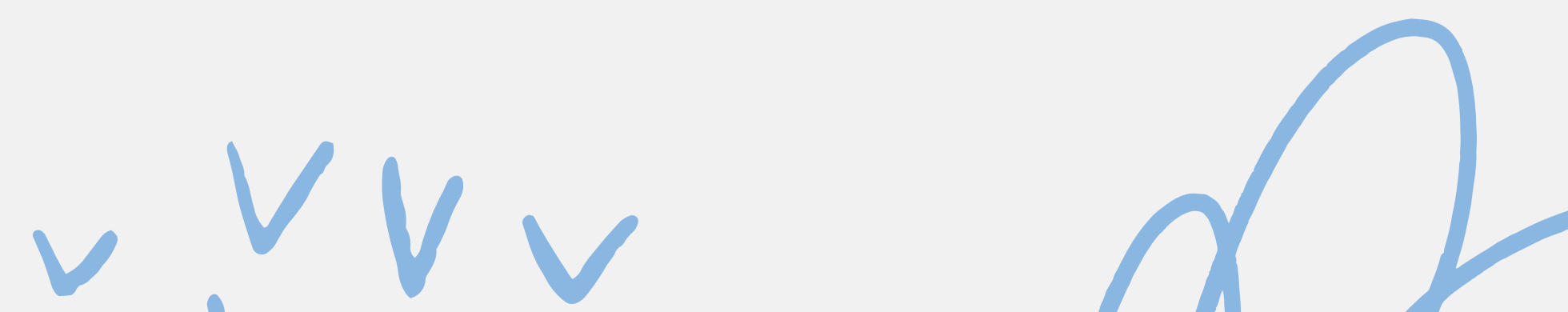
How Does DBSCAN Work?



1. Pick a random unvisited point and check how many neighbors are within ϵ .
2. If MinPts are found, mark it as a core point and start forming a cluster.
3. Expand the cluster by adding all reachable border points.
4. If a point doesn't meet the density threshold, it's labeled noise.
5. Repeat until all points are visited and categorized.



Key Parameters in DBSCAN

- 1) **eps**: This defines the radius of the neighborhood around a data point.
 - 2) **MinPts**: This is the minimum number of points required within the **eps** radius to form a dense region.
- 
- 
- 

The slide features several hand-drawn blue scribbles in the corners. Top-left: a star-like shape. Top-center: three overlapping loops. Top-right: a vertical oval. Bottom-left: a wavy line. Bottom-center: a scribble resembling a ball. Bottom-right: four small 'v' marks and a large loop.

How to set your parameters?

Choosing Epsilon (ϵ)

If the distance between two points is less than or equal to **ϵ** , they are considered neighbors. Choosing the right **ϵ** is crucial:

If **ϵ** is too small, most points will be classified as noise.



Choosing Epsilon (eps)

If **eps** is too large, clusters may merge, and the algorithm may fail to distinguish between them.



Choosing Epsilon (eps)

Keep it balanced :)



Choosing MinPts in DBSCAN

MinPts (Minimum Points) determines how many neighboring points are needed to form a core point and start a cluster.

How to Set MinPts?

A general rule of thumb:

- ◆ $\text{MinPts} \geq D + 1$, where D is the number of dimensions in the dataset.
- ◆ In most cases, a minimum value of $\text{MinPts} = 3$ is recommended.

Choosing MinPts in DBSCAN

Why is MinPts Important?

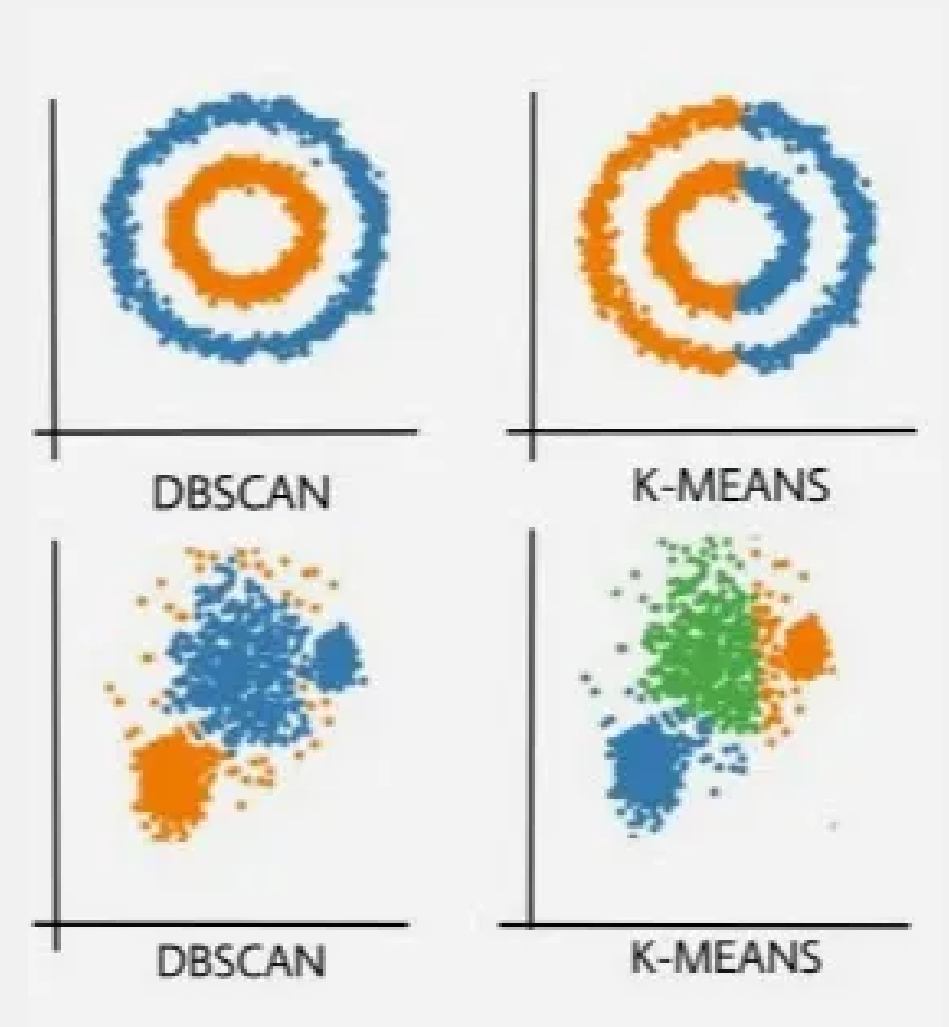
- ✓ **Too low MinPts:** Can form small, meaningless clusters and misclassify noise.
- ✓ **Too high MinPts:** Can miss smaller clusters and label important points as noise.

When to Use DBSCAN?

DBSCAN is the best choice when dealing with **complex datasets** that traditional clustering methods struggle with.

Use DBSCAN When:

- ✓ Clusters are irregularly shaped: Unlike K-Means, which assumes circular clusters, DBSCAN detects arbitrary shapes.



When to Use DBSCAN?

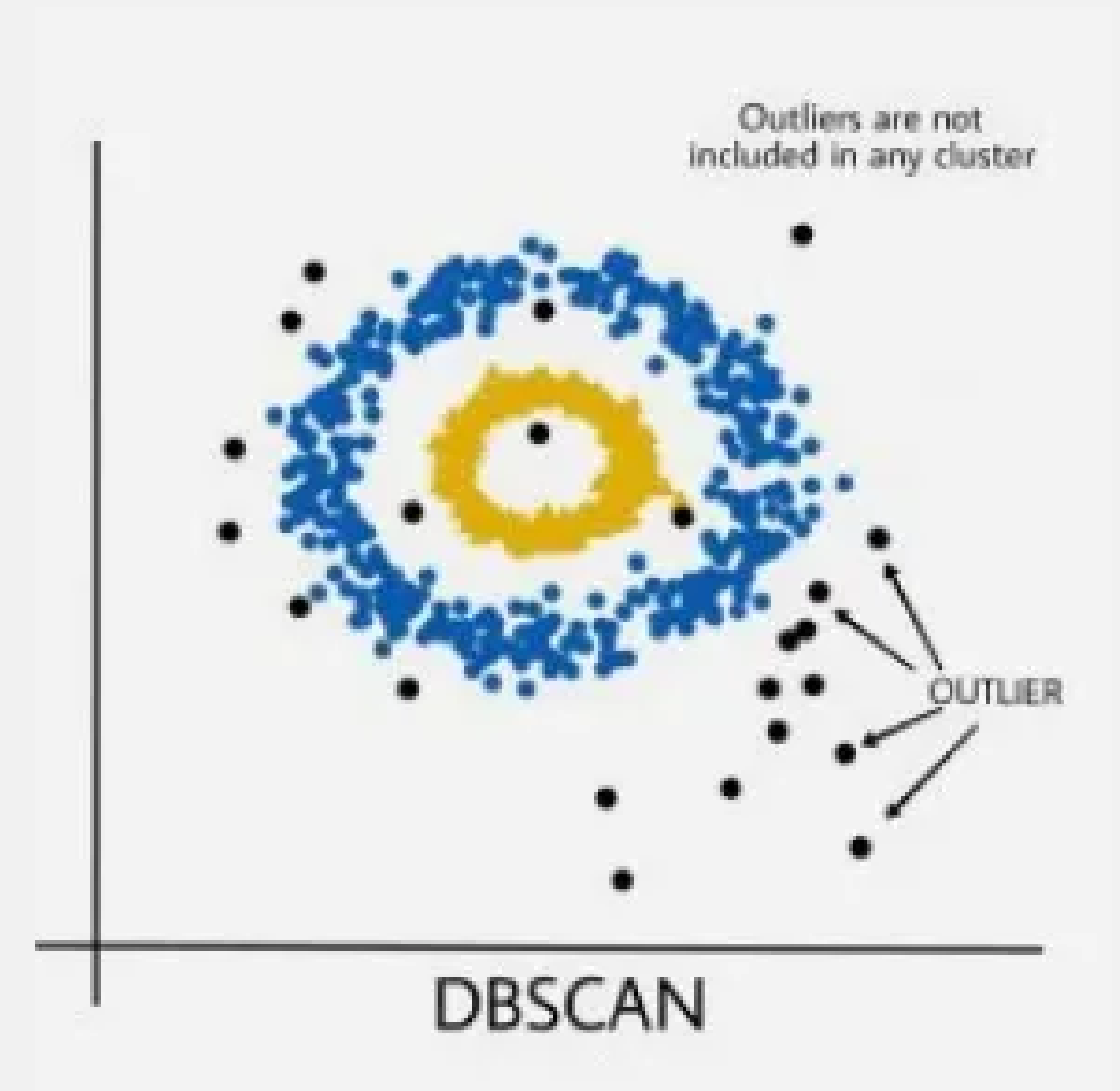
Use DBSCAN When:

- ✓ **Clusters have different densities:** Unlike Hierarchical Clustering, DBSCAN can adapt to varying densities without merging everything into one big cluster.
- ✓ **You don't know the number of clusters:** K-Means requires predefining K, but DBSCAN discovers clusters naturally.

When to Use DBSCAN?

Use DBSCAN When:

- ✓ **Clusters have different densities:** Unlike Hierarchical Clustering, DBSCAN can adapt to varying densities without merging everything into one big cluster.



When to Use DBSCAN?

✗ Avoid DBSCAN When:

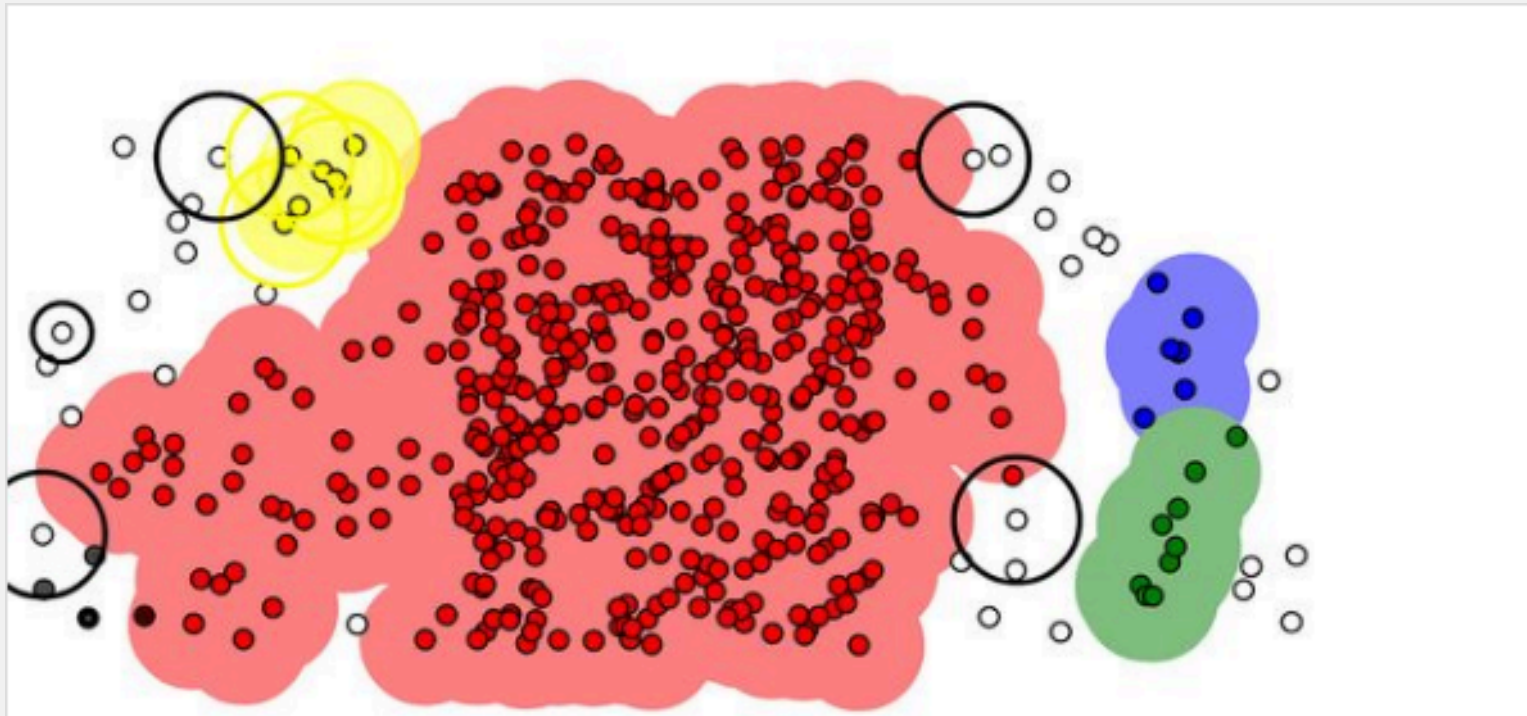
- ✗ Clusters have similar densities and well-separated → K-Means is better.
- ✗ **Dataset is too large:** DBSCAN has higher computational complexity than K-Means.
- ✗ **You have high-dimensional data:** DBSCAN struggles because distances become meaningless.
 - ♦ Solution? Use **PCA** to reduce dimensions before applying DBSCAN.

Code for DBSCAN




[https://colab.research.google.com/
drive/1wrEFL9mVXu_IRdD3zrD58LqBKzAb7HB7?
usp=sharing](https://colab.research.google.com/drive/1wrEFL9mVXu_IRdD3zrD58LqBKzAb7HB7?usp=sharing)


Resources



Clustering Like a Pro: A Beginner's Guide to DBSCAN


Data clustering is a fundamental task in machine learning and data analysis. One powerful technique that has gained prominence is...

 Medium / Dec 26, 2023



DBSCAN Clustering in ML | Density based clustering

DBSCAN is a density-based clustering algorithm that effectively identifies arbitrary-shaped clusters and handles noise, distinguishing it from K-Means and hierarchical clustering, which assume compact, spherical...

 GeeksforGeeks / Jan 29



Thank you very much!

Blue AI Team

Youssef Moustafa

Nagham Wael