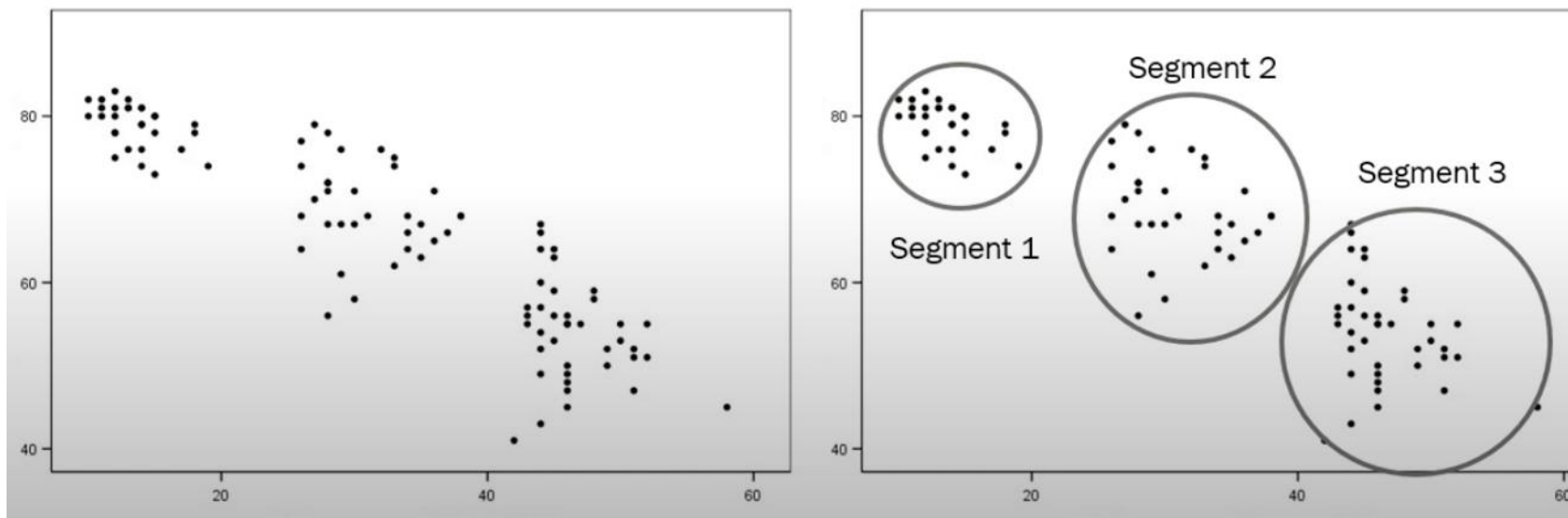


The background of the slide is an abstract pattern consisting of a grid of lines in red, green, and blue, creating a 3D effect of receding planes.

CLUSTERING ALGORITHMS

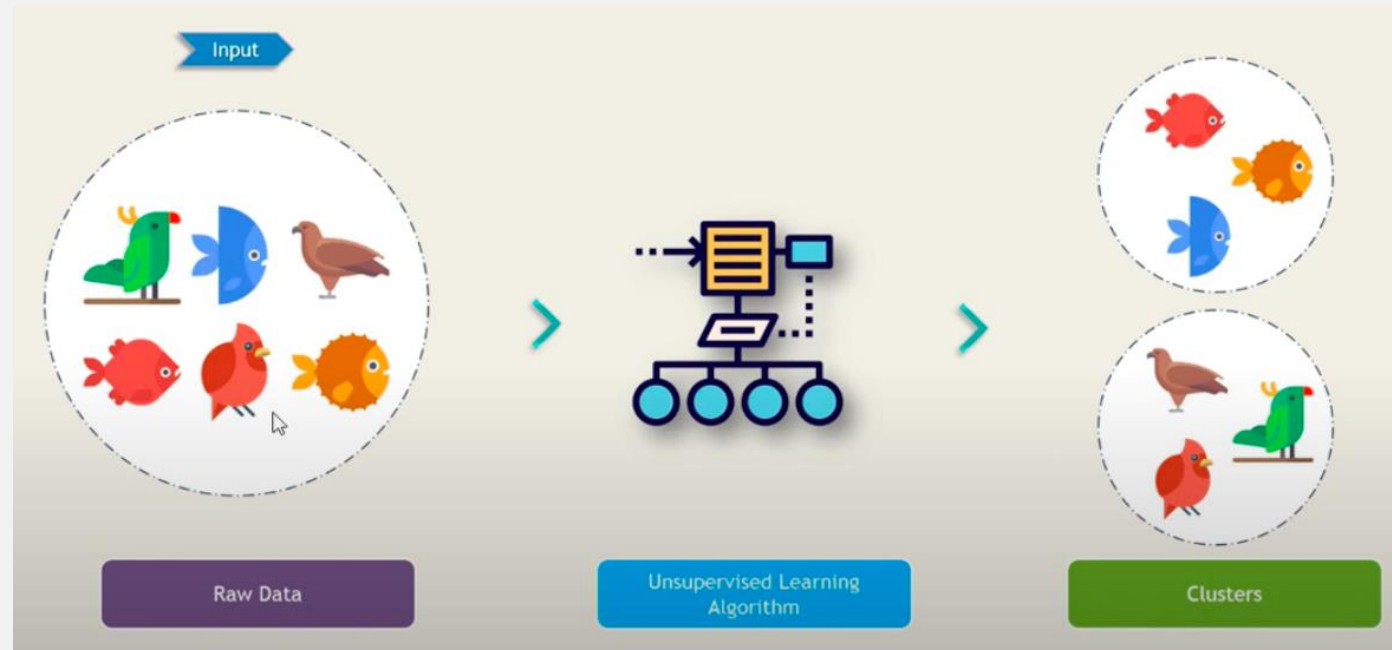
Students: Bruno R. dos Santos and Clément Perucca

WHAT IS CLUSTERING ?



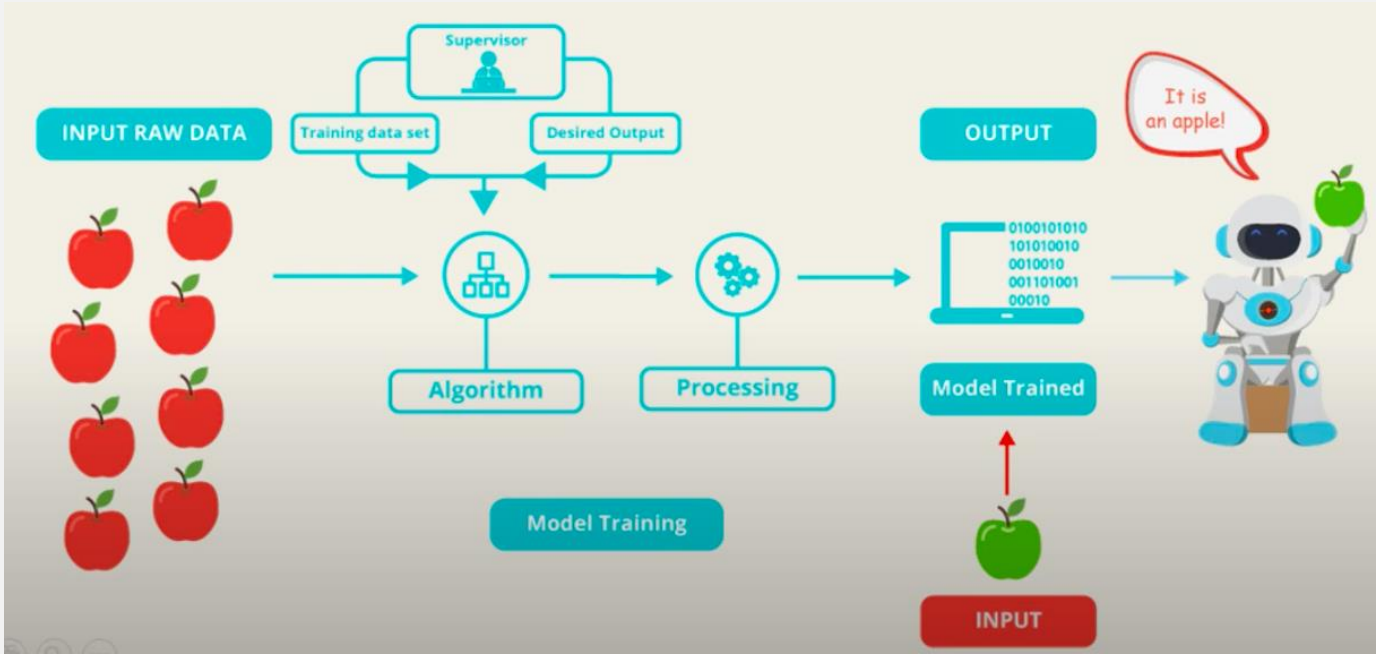
CLUSTERING OR CLASSIFICATION ?

CLUSTERING



CLUSTERING OR CLASSIFICATION ?

CLASSIFICATION



HOW IT'S WORKS ?

1. Specify the number of clusters K
2. Initialise the centroids
3. Continue iterating
 - Calculate the sum of the square of the distance between the data points and all centroids.
 - Assign each data point to the next closest cluster (centroid).
 - Calculate the centroids of the clusters by taking the average of all data points that belong to each cluster.

K MEANS - CODE

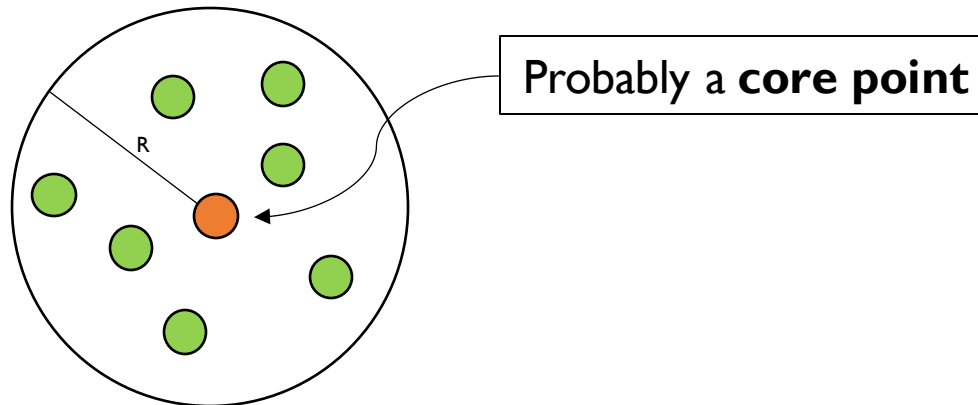
[OTPA0011/kMeans \(Class\).py at main · bruniculos08/OTPA0011 \(github.com\)](https://github.com/bruniculos08/OTPA0011/blob/main/OTPA0011/kMeans%20(Class).py)

DB SCAN

- **Density-based spatial clustering of applications with noise;**
- **Parameters:** a ratio **R** and a minimum group size **k**;
- To implement the algorithm, first, it is good to have in mind the concepts of **core points**, **border points** and **noise points**.

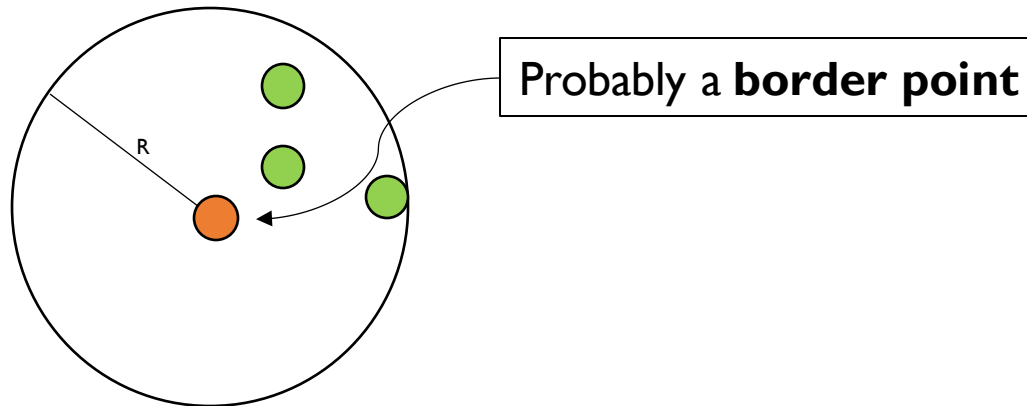
DB SCAN – CORE POINT

- A **core point** is a point that has at least **k** neighbors in a distance **R**;
- A **core point** always belong to a group of points (basically it is not a **noise point**);



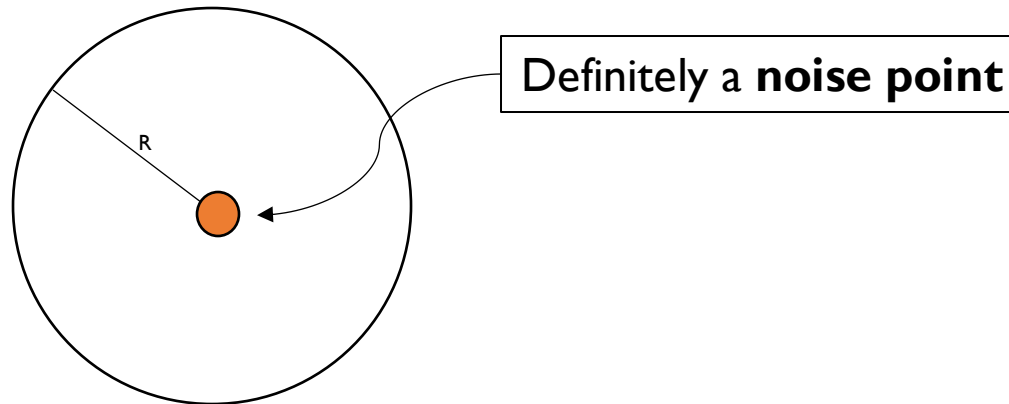
DB SCAN – BORDER POINT

- A **border point** is a point that has fewer than **k** neighbors in a distance **R**, but this point has at least one **core point** as neighbor;
- A **border points** belongs to the cluster of his **core point** neighbors;



DB SCAN – NOISE POINT

- A **noise point** is a point that has fewer than **k** neighbors in a distance **R**, and none of these points is **core point**;



DB SCAN – PSEUDOCODE

DB-Scan(data, R, k):

1. Create a list of clusters
2. Chose any point β from data that has not been selected yet
3. Create a cluster, set β as selected and add to this cluster the selected point β and the return of **DB-ScanBuild**(data, β , R, k)
4. Add the result cluster to the list of clusters
5. If there still exists unselected points in the data list, go back to step 2

DB SCAN – PSEUDOCODE

DB-ScanBuild(data, β , R, k):

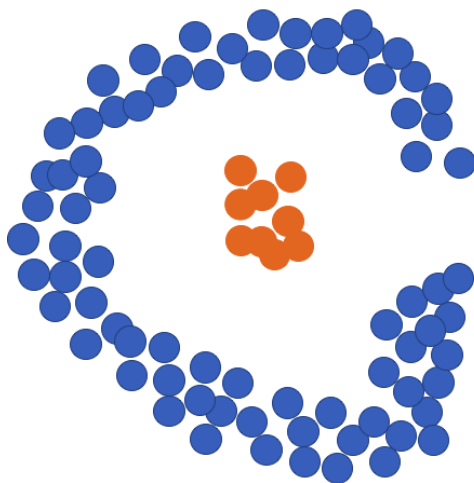
1. Create an empty list for the neighbors of β
2. For each point μ in data, if μ is in a distance R from β then put this point in the neighbors list
3. If the length of the list is less than k, return an empty list, else go to step 4
4. Create a new cluster and add to this cluster the result of **DBScanBuild**(data, μ , R, k) for each point μ in the list
5. Return the new cluster concatenated with the neighbors list

DB SCAN – CODE

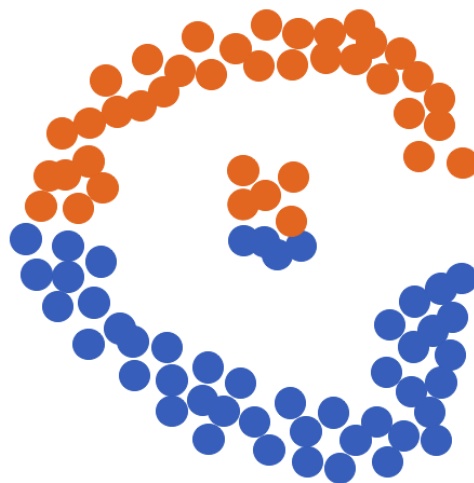
[OTPA0011/DB-Scan \(Class\).py at main · bruniculos08/OTPA0011 \(github.com\)](https://github.com/bruniculos08/OTPA0011/tree/main/OTPA0011/DB-Scan%20(Class).py)

COMPARING

DBSCAN



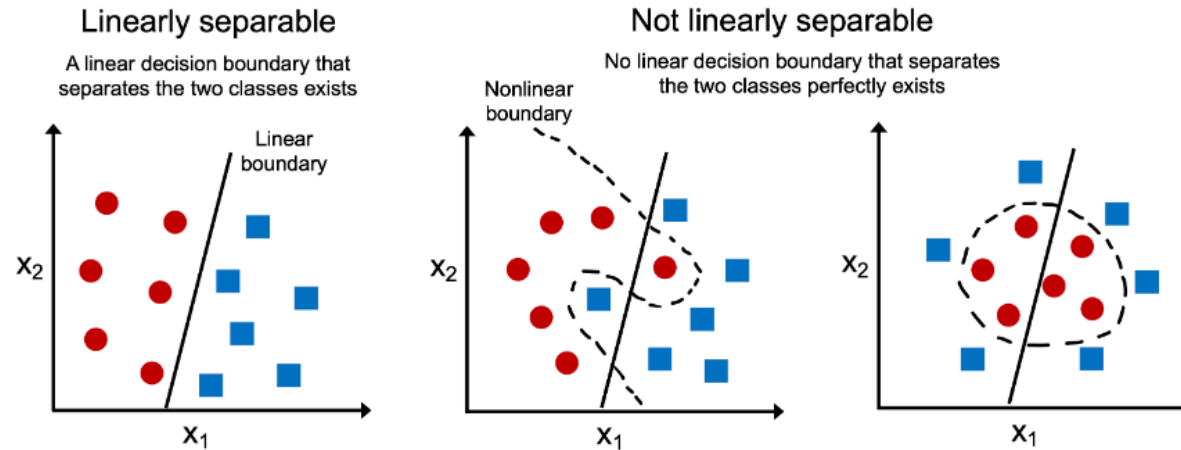
K-Means



Source: Yufeng, 2022.

COMPARING

- K-Means is clustering algorithm used for linear datasets and DB-Scan is used for non-linear datasets.



Source: Kumar, 2022.

REFERENCES

- MUND, Kumar Shritam; How does DBSCAN clustering algorithm works? 26 de julho de 2019. Disponível em: <<https://shritam.medium.com/how-dbscan-algorithm-works-2b5bef80fb3>>. Acesso em 3 de dezembro de 2022.
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- YUFENG; Understanding DB-Scan and Implementation with Python. 22 de janeiro de 2022. Disponível em: <<https://towardsdatascience.com/understanding-dbscan-and-implementation-with-python-5de75a786f9f/>>. Acesso em 3 de dezembro de 2022.