

APPLIED MACHINE LEARNING

Clustering

Part I - Principle

Clustering Principle

Clustering aims at grouping data that resemble one another into *clusters*.

Applications

- Grouping documents (e.g. webpages) into similar topics
 - *Speed up search of documents based on matches for similar queries*
- Group customers (e.g. on amazon) into types / profiles for marketing
 - Offer related products that may interest the customer
- Find groups of genes with similar phylogenetic characteristics
 - Relate some diseases and determine risks of developing a disease

Clustering Principle

Clustering can be used as:

- Feature extraction method: for identifying underlying structure in data and salient features, best visualized through cluster prototype.
- Compression method: for organizing the data and summarizing it through cluster prototypes.

A cluster prototype can be:

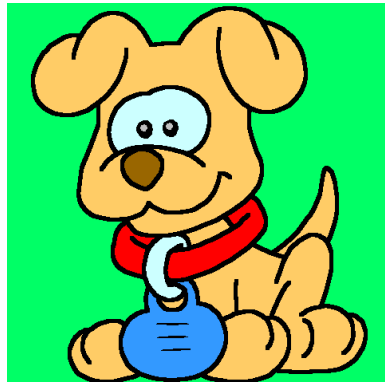
- A typical datapoint, best representative of the cluster
- The average (centroid) of the datapoints in the cluster

Clustering Principle

Groups of points are said to belong to the same **cluster** if they are similar enough.

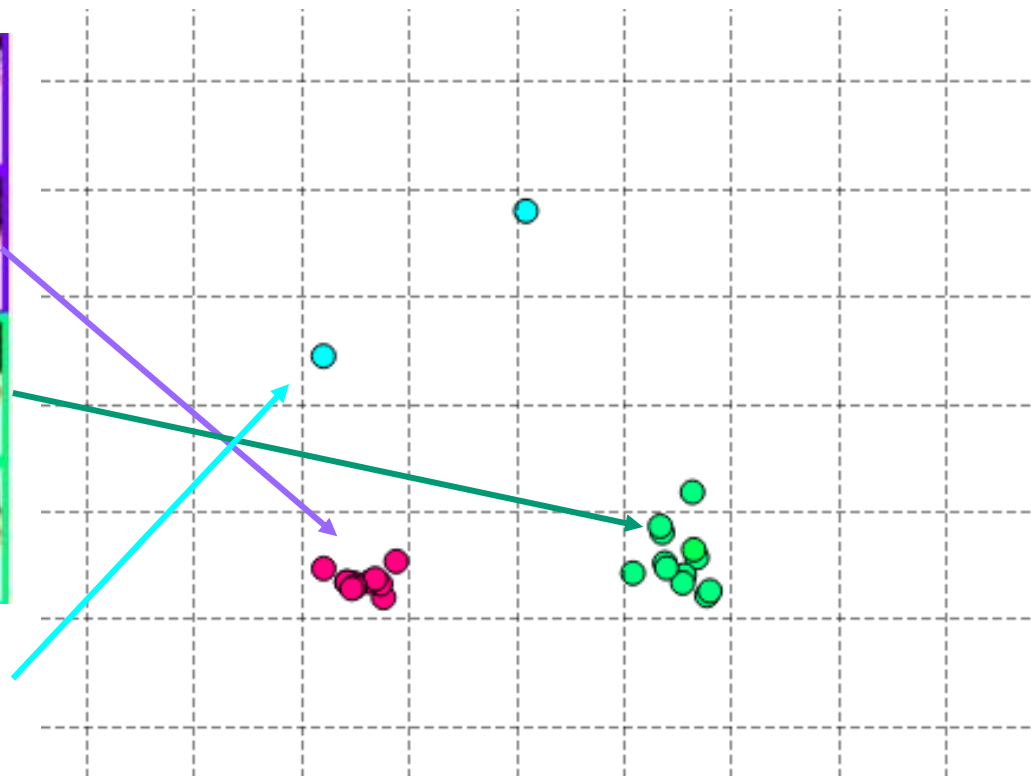
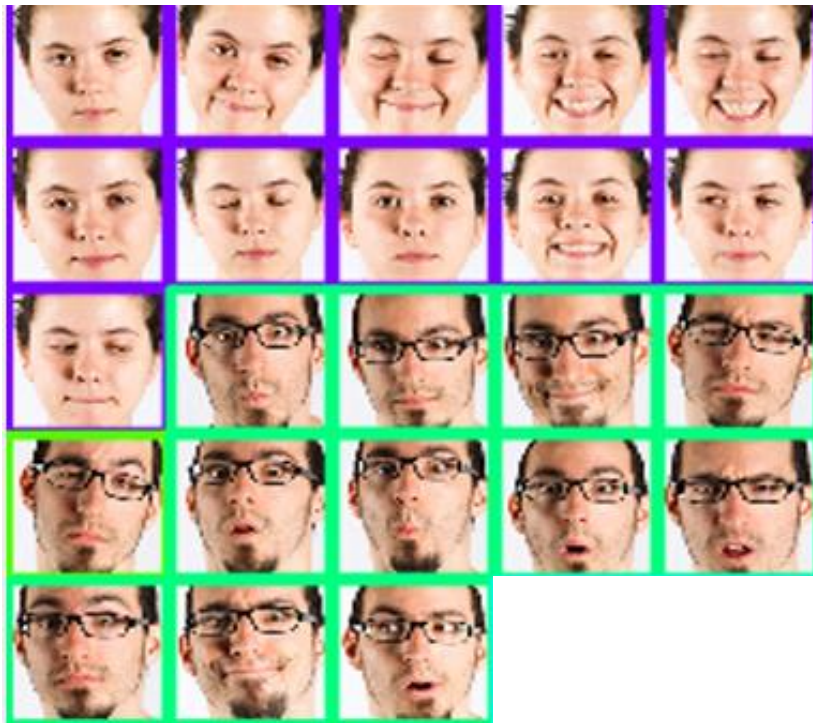
→ measure of similarity.

Which clusters?



Which two subgroups of pictures are similar and why?

Dataset with outliers



What is good clustering?

A good clustering method produces high quality clusters when:

Outliers can be determined

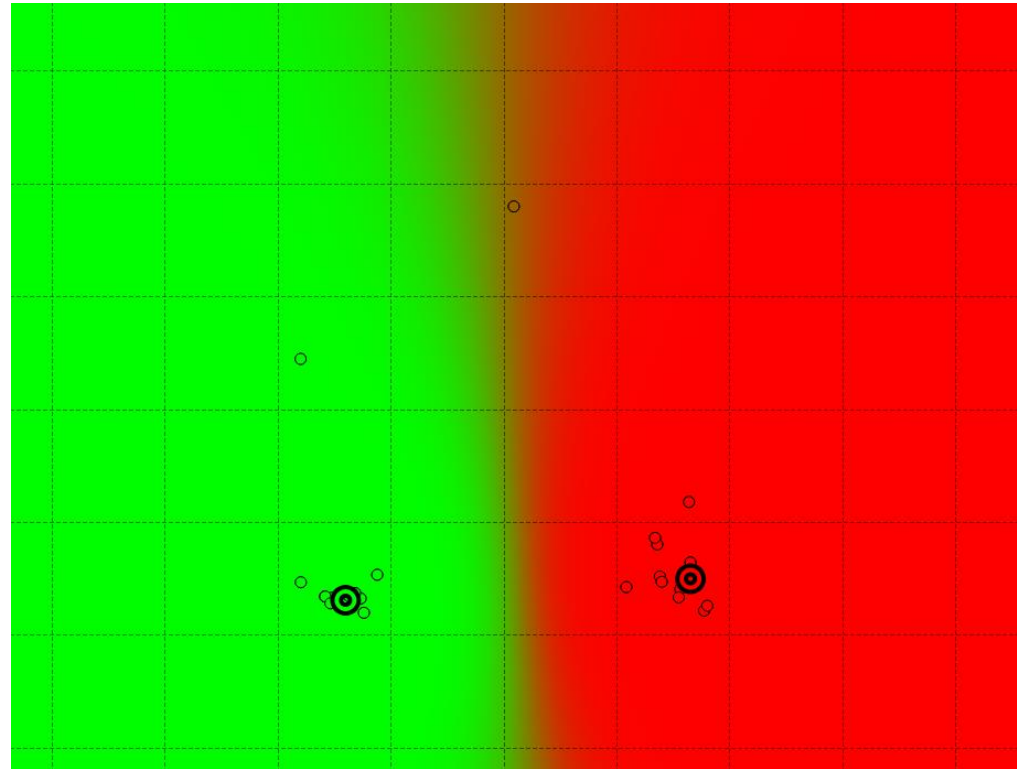
The intra-cluster similarity is high.

The inter-cluster similarity is low.

The quality measure of a cluster depends on the similarity measure used!

The role of prior information

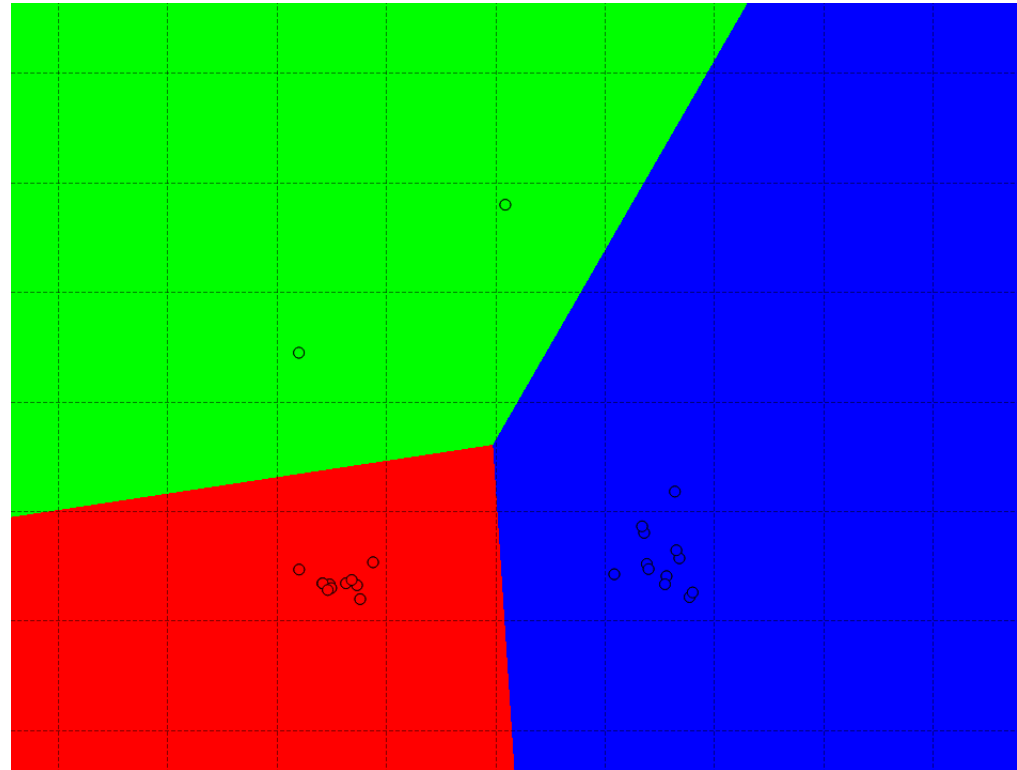
Clustering approaches often require to have a good *guess* as to how many clusters exist.



*If we assume only 2 clusters,
the outliers cannot be excluded.*

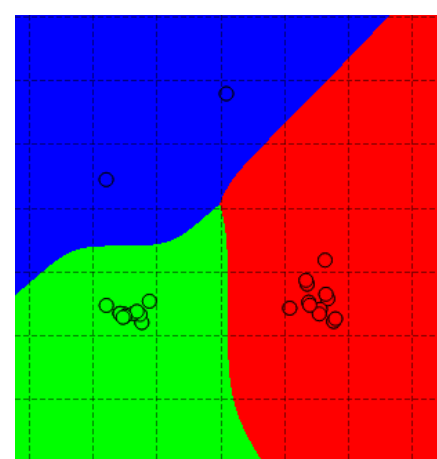
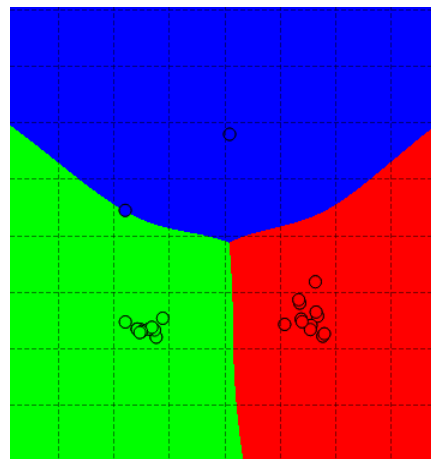
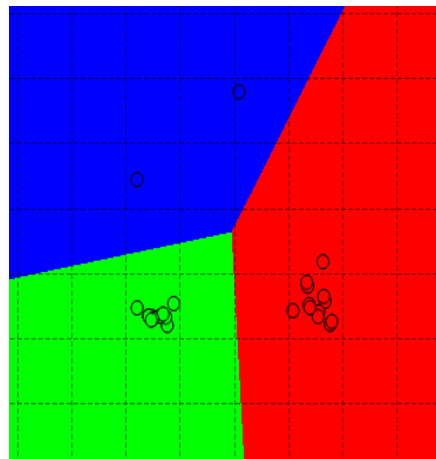
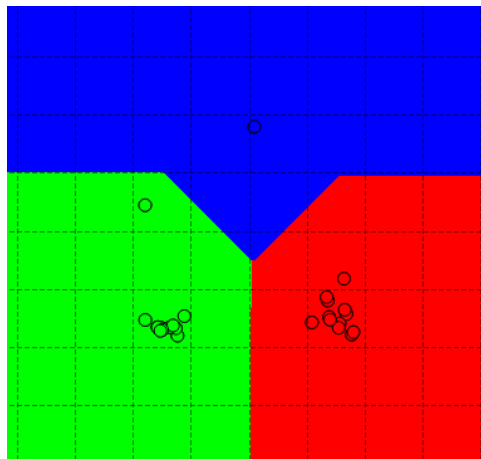
The role of prior information

Clustering approaches often require to have a good *guess* as to how many clusters exist.



If we assume 3 clusters, the outliers are grouped in one cluster.

The role of the measure of similarity



L1-norm

$$d(x^1, x^2) = |x^1 - x^2|$$

$$= \sum_{i=1}^N |x_i^1 - x_i^2|$$

L2-norm

$$d(x^1, x^2) = \|x^1 - x^2\|$$

$$= \sqrt{\sum_{i=1}^N (x_i^1 - x_i^2)^2}$$

$p = 3$

$p = 5$

Lp-norm

$$d(x^1, x^2) = \|x^1 - x^2\|^p$$

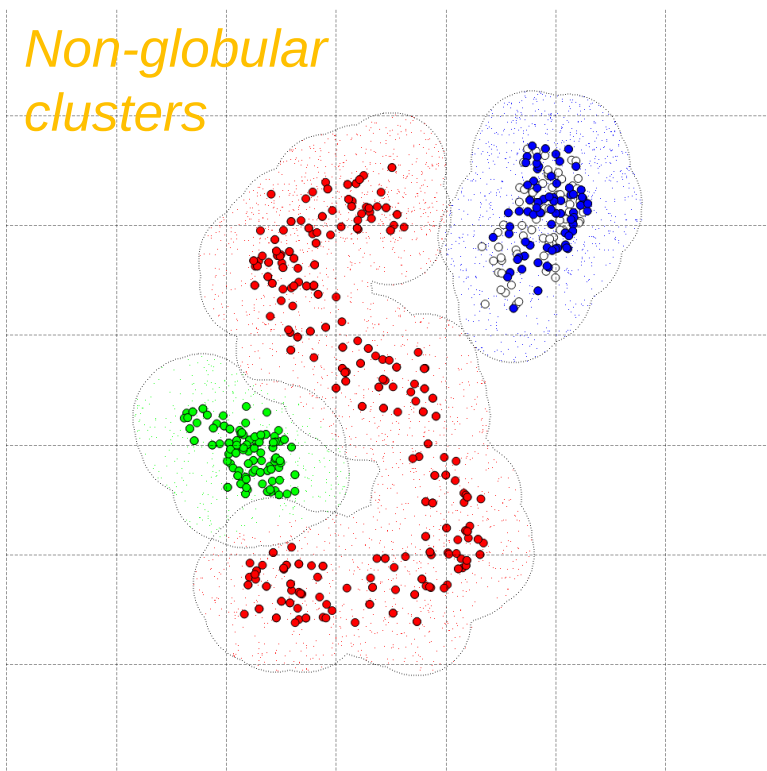
$$= \sqrt[p]{\sum_{i=1}^N |x_i^1 - x_i^2|^p}$$

$$x^1, x^2 \in \mathbb{R}^N$$

The shape of the clusters

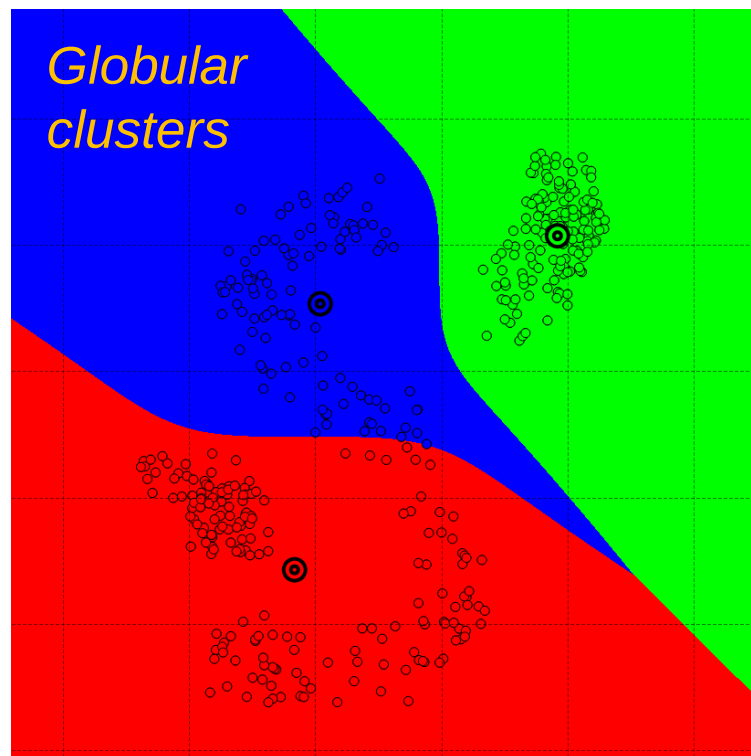
Clustering techniques differ in the complexity of the clusters generated.

*Non-globular
clusters*



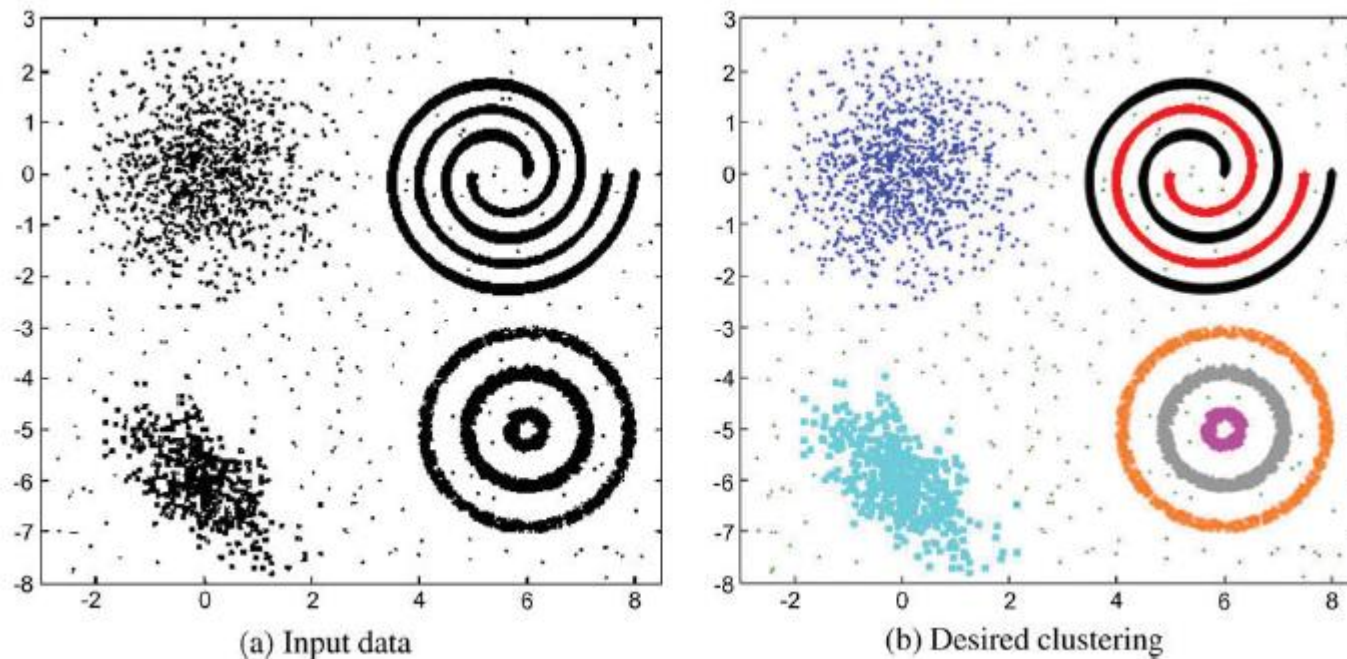
DBSCAN

*Globular
clusters*



K-means

Clustering: Challenges



Clustering may fail when metric changes depending on the region of space