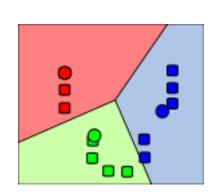


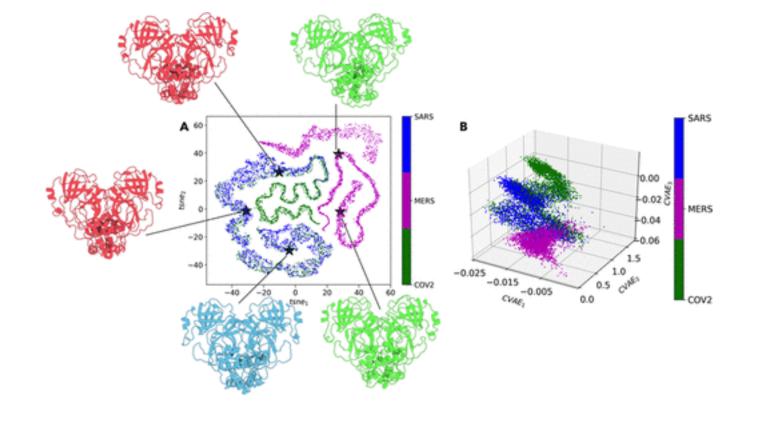
# From biomolecular data to information



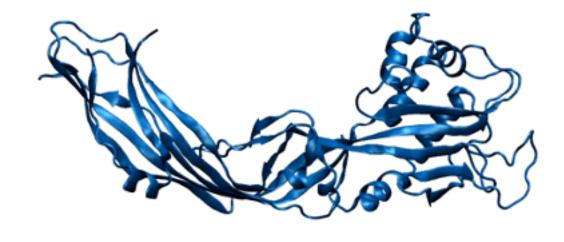
Antonia Mey

Matteo Degiacomi













matteo.t.degiacomi@dur.ac.uk



@MatteoDegiacomi



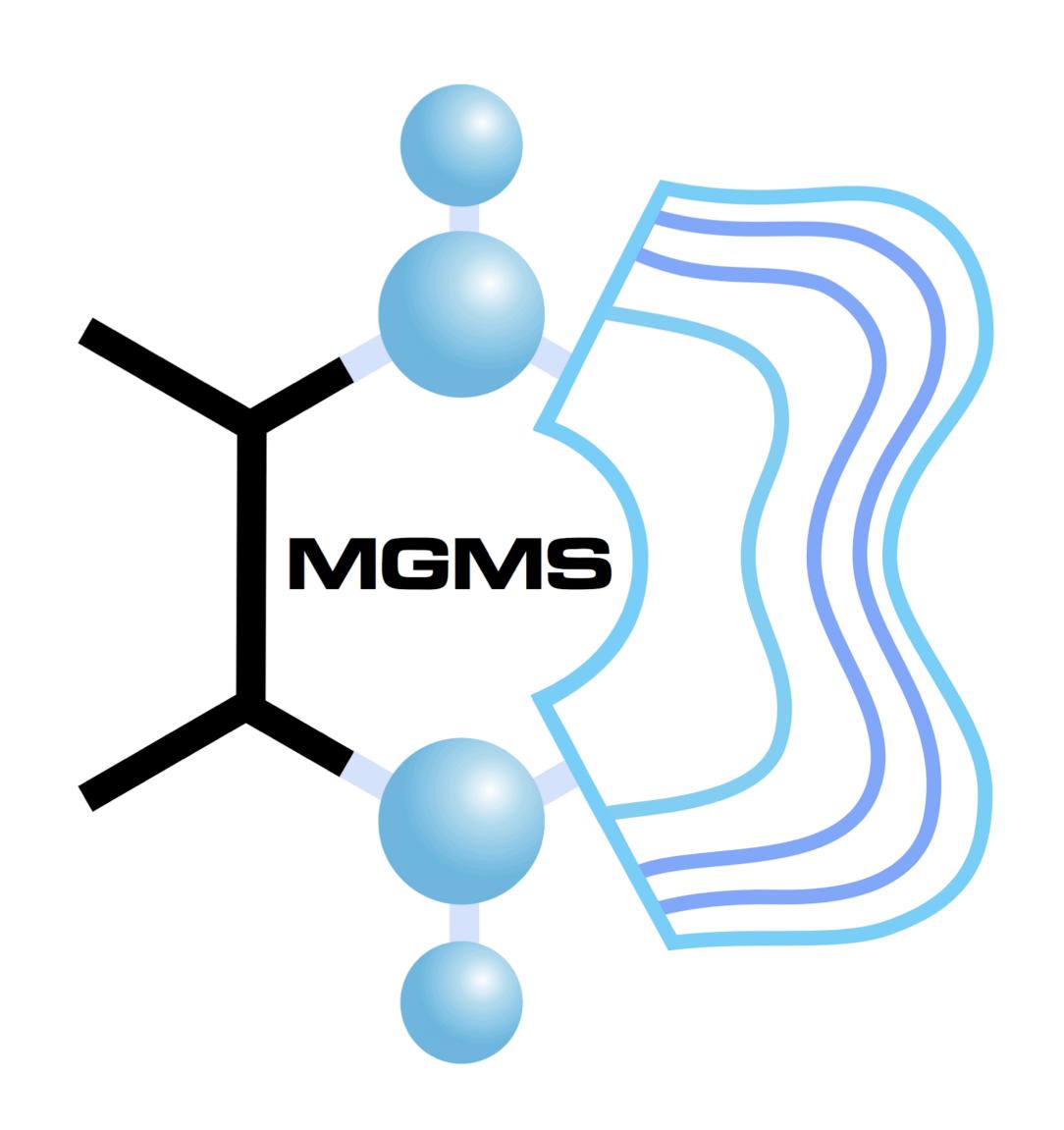








# Thanks to our sponsors



### Schedule

## Thursday — Antonia Mey

Friday — N	Matteo	Degiacomi
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13:30-14:45	ML Clustering
14:45-15:00	break
15:00-16:20	ML Dimensionality Reduction
16:20-16:30	Closing remarks
18:00-onwards	Informal social event

09:30-10:45	ML Classification
10:45-11:00	break
11:00-12:30	ML Regressions and Neural Networks
12:30-13:30	Lunch
13:30-onwards	Bring your own problem

## What is machine learning?

## Artificial intelligence

Design an intelligent agent that perceives its environment and makes decisions to maximise chances of achieving its goal.

## Machine learning

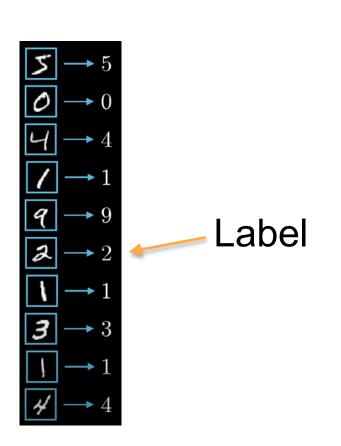
Gives computers the ability to learn without specifically being programmed (Arthur Samuel 1959)

Supervised learning

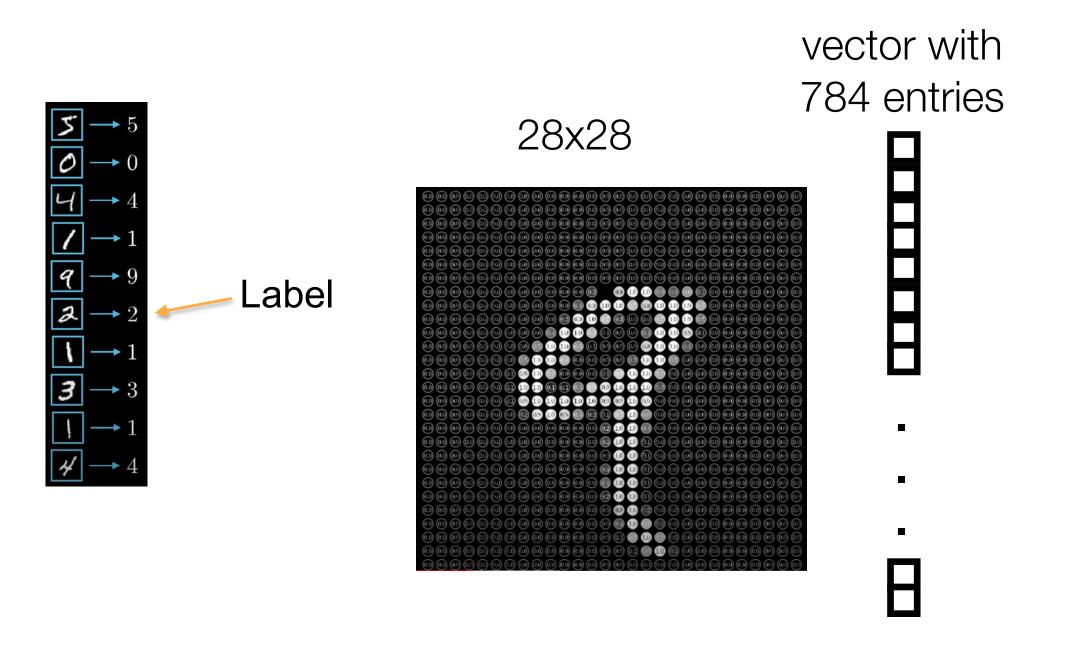
**Unsupervised learning** 

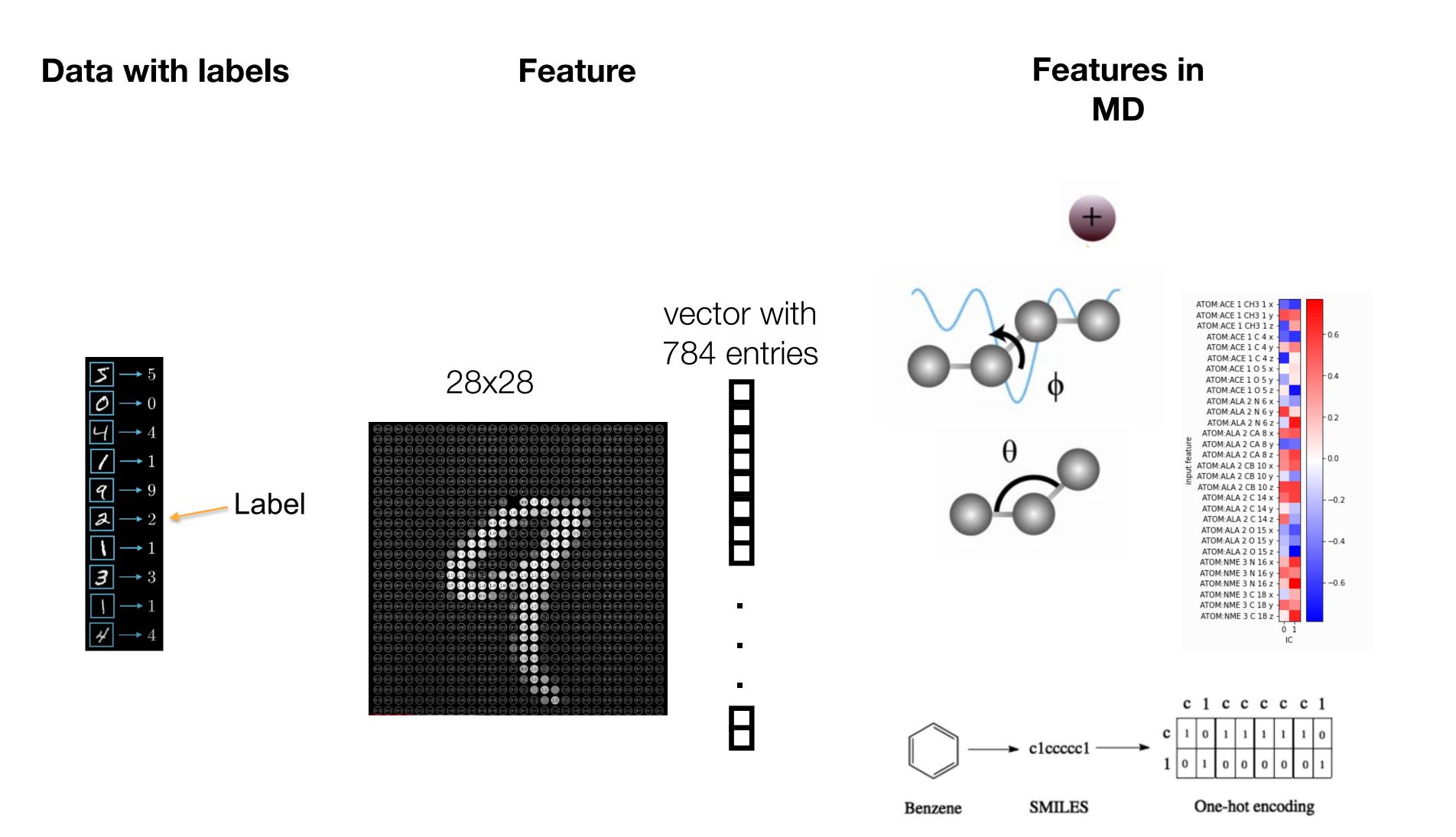
reinforcement learning

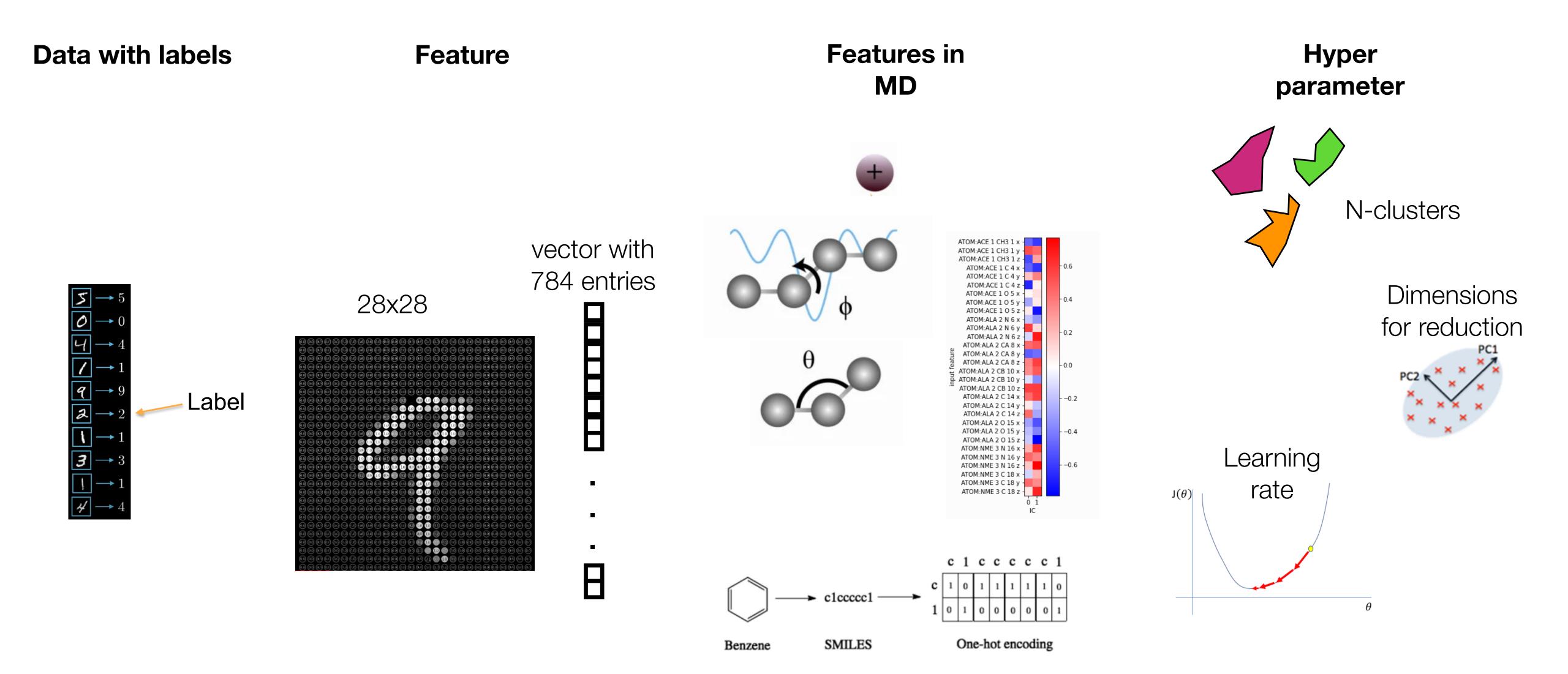
#### Data with labels



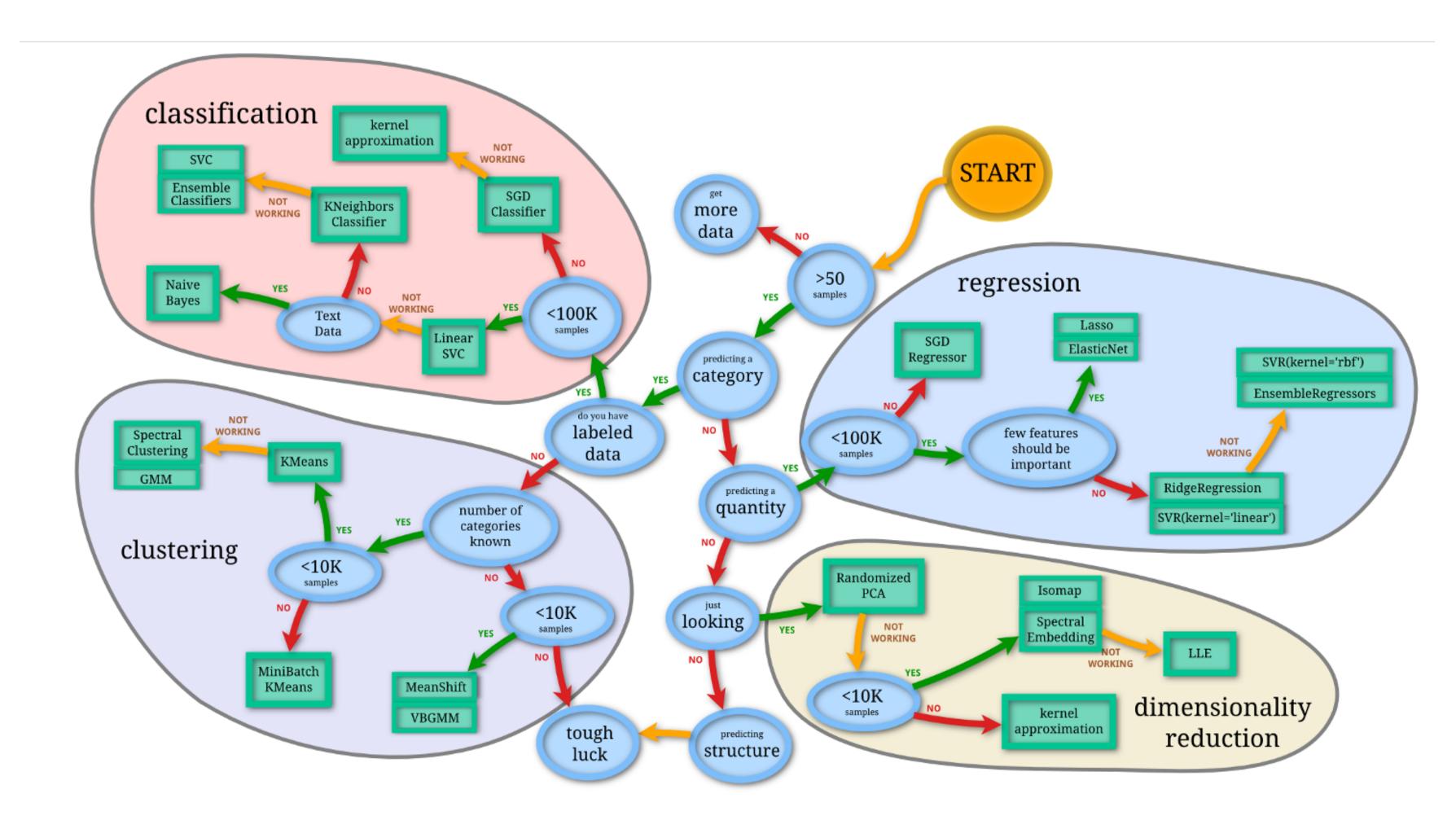
Data with labels Feature





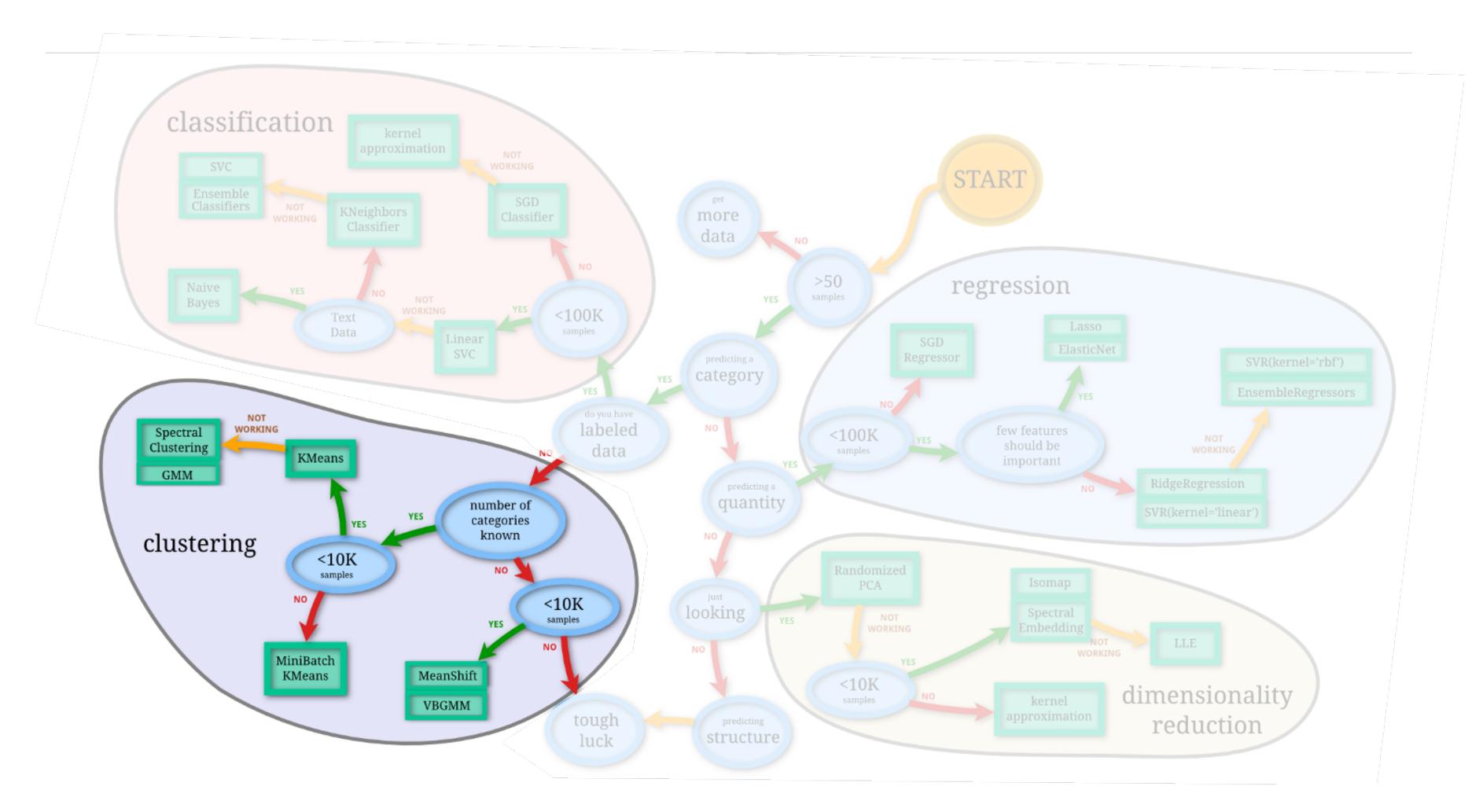


## The Data Mining World

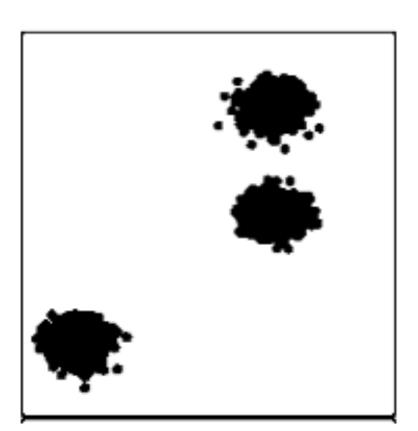


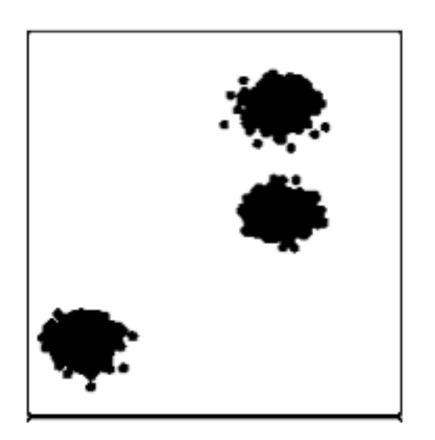
From scikit-learn.org

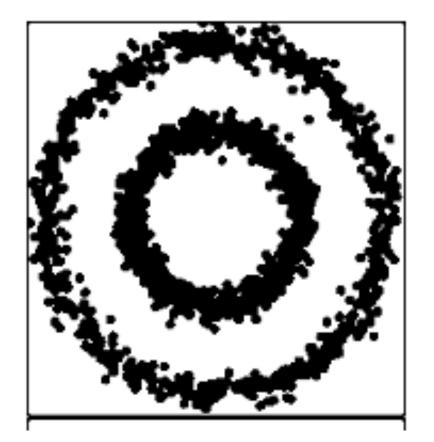
## The Data Mining World — Clustering

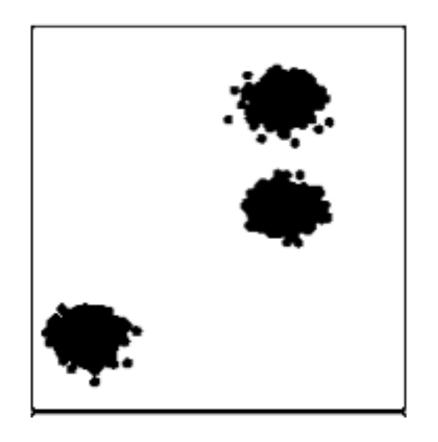


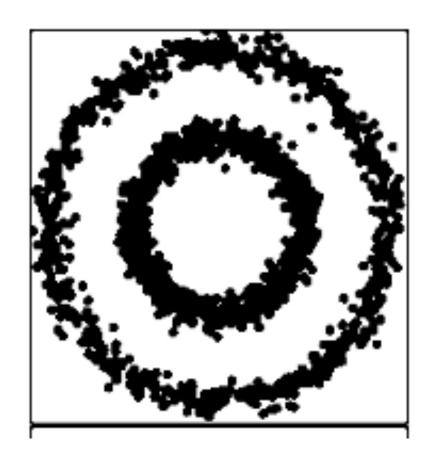
From scikit-learn.org



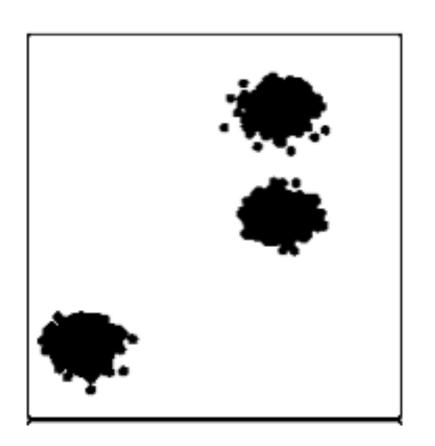


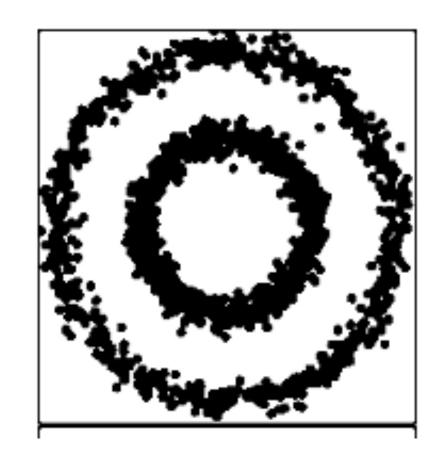


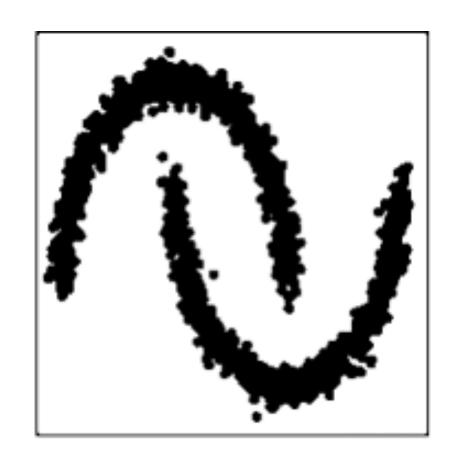


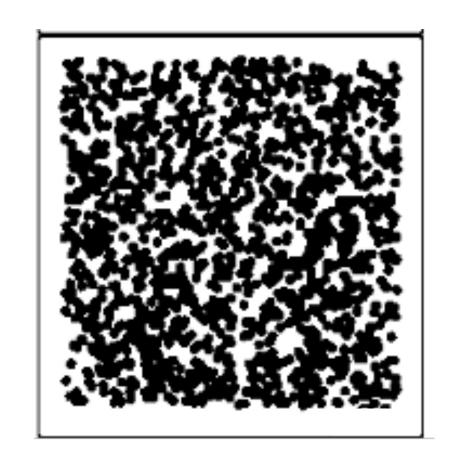


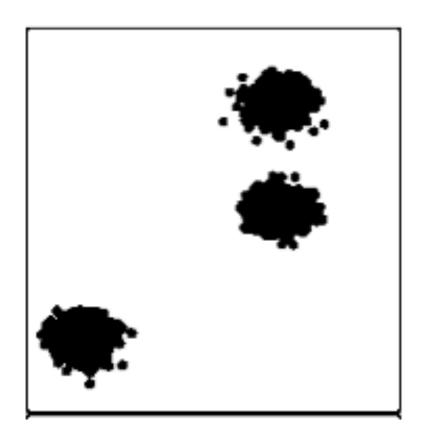


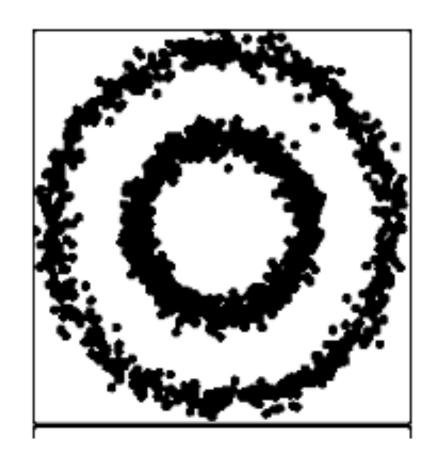


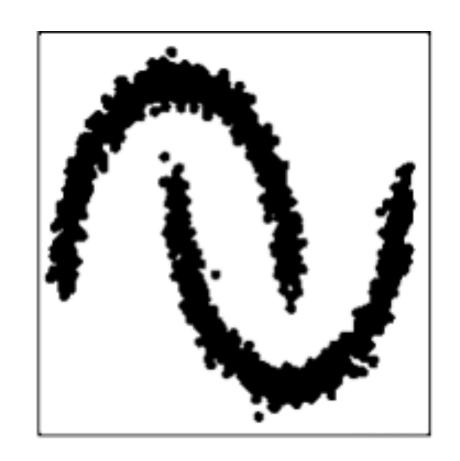




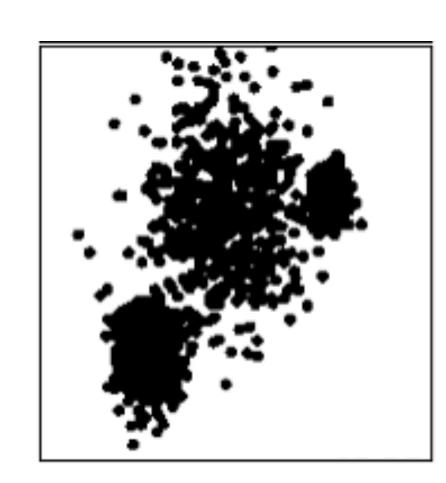




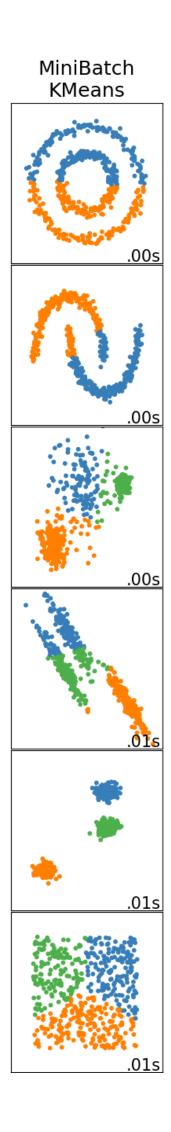




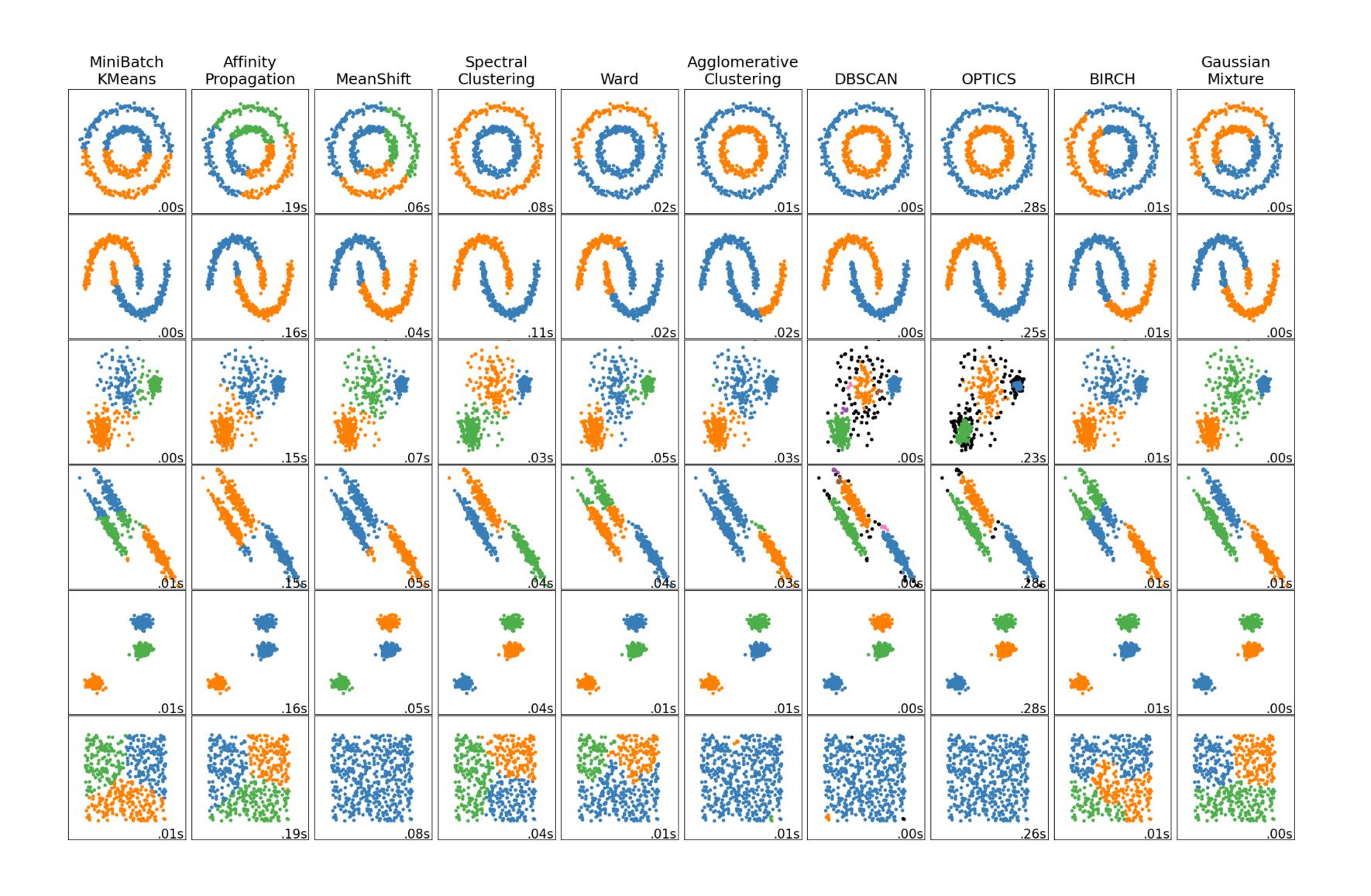




# There are many different clustering algorithms

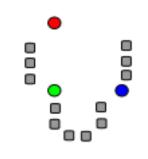


## There are many different clustering algorithms

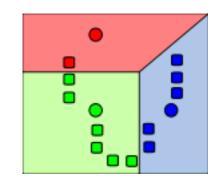


# K-means, DBSCAN and spectral clustering

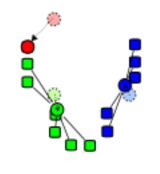
#### K-means



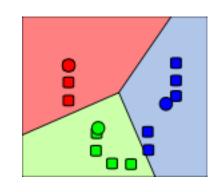
Initial guess



K-clusters are generated with the nearest mean



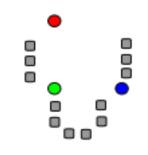
Centroid of the k-lcusters becomes the new mean



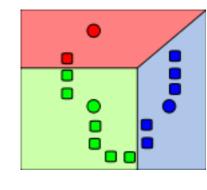
Iterate until convergence

# K-means, DBSCAN and spectral clustering

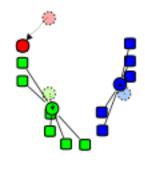
#### K-means



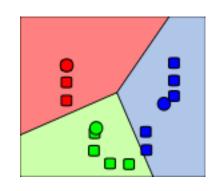
Initial guess



K-clusters are generated with the nearest mean

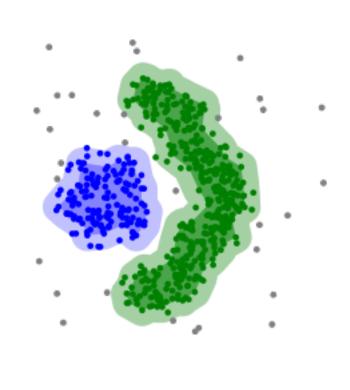


Centroid of the k-lcusters becomes the new mean



Iterate until convergence

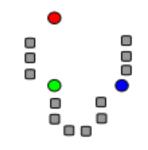
#### **DBSCAN**



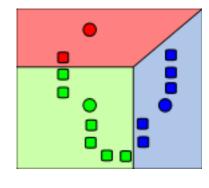
- 1. Find the points in the ε (eps) neighbourhood of every point, and identify the core points with more than minPts neighbours.
- 2. Find the connected components of core points on the neighbour graph, ignoring all non-core points.
- 3. Assign each non-core point to a nearby cluster if the cluster is an  $\epsilon$  (eps) neighbour, otherwise assign it to noise.

# K-means, DBSCAN and spectral clustering

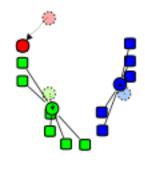
#### K-means



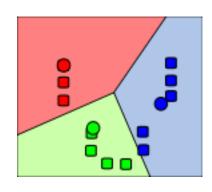
Initial guess



K-clusters are generated with the nearest mean

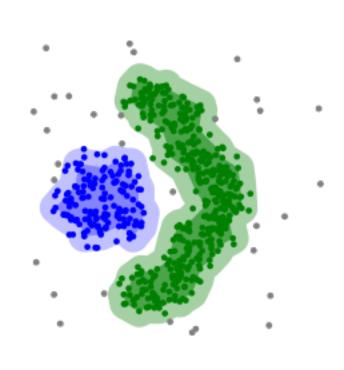


Centroid of the k-lcusters becomes the new mean



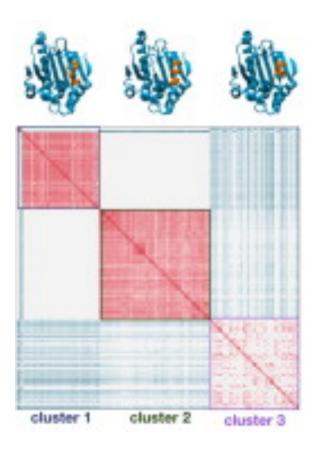
Iterate until convergence

#### **DBSCAN**



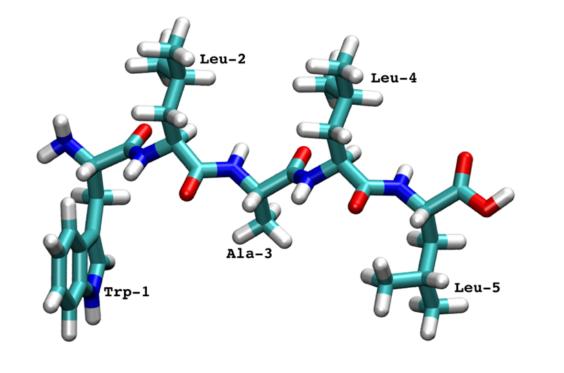
- 1. Find the points in the  $\epsilon$  (eps) neighbourhood of every point, and identify the core points with more than minPts neighbours.
- 2. Find the connected components of core points on the neighbour graph, ignoring all non-core points.
- 3. Assign each non-core point to a nearby cluster if the cluster is an  $\epsilon$  (eps) neighbour, otherwise assign it to noise.

#### **Spectral clustering**



In spectral clustering clusters are found by doing an eigenvalue decomposition of the Laplacian

# K-means example



Clustering is one of the first steps in building a Markov State Model

