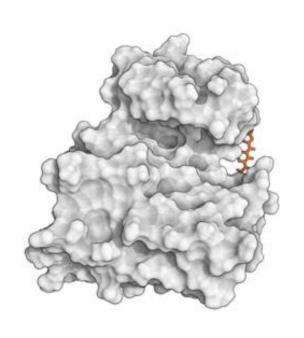
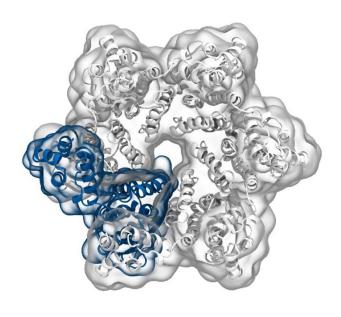
## Simulation of Biomolecules



### Clustering



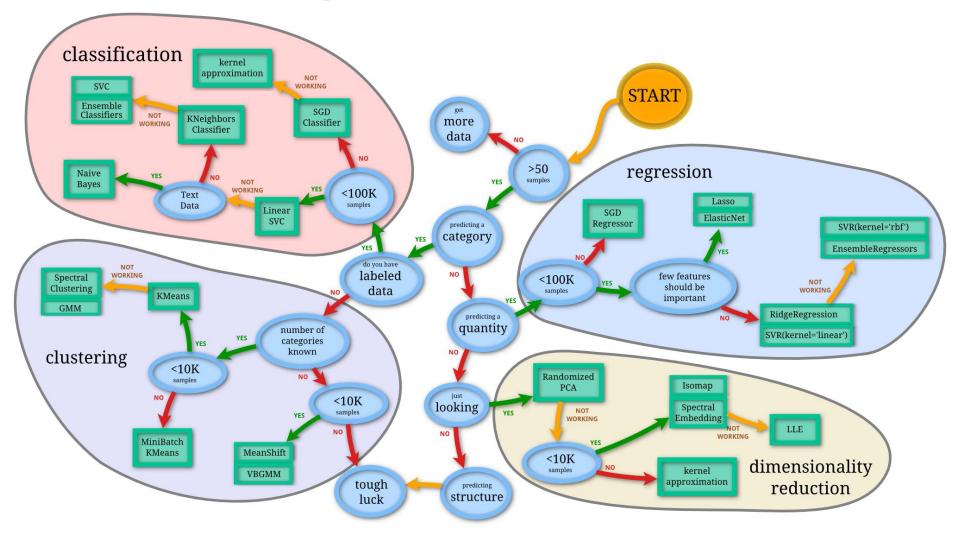
Dr Matteo Degiacomi University of Edinburgh

matteo.degiacomi@ed.ac.uk

Dr Antonia Mey
University of Edinburgh

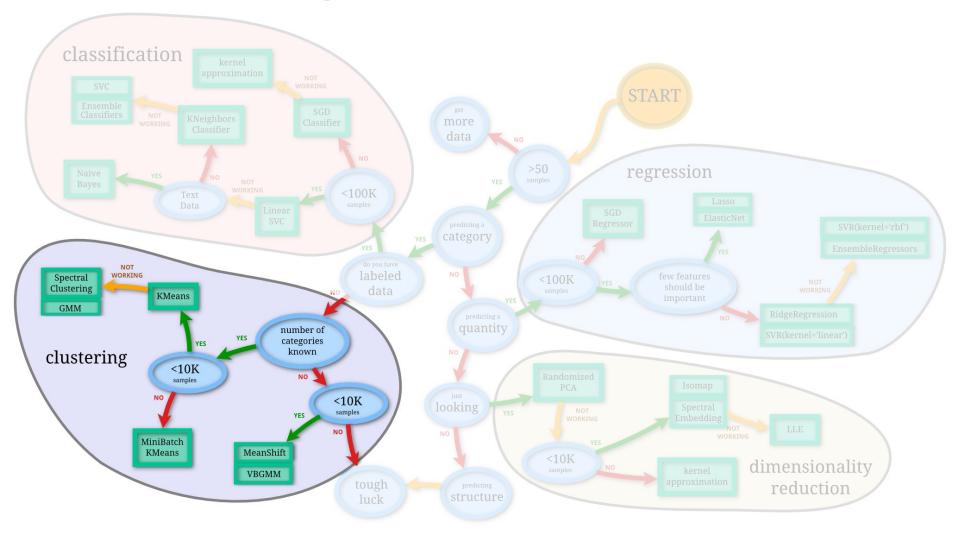
antonia.mey@ed.ac.uk

### The Data Mining world



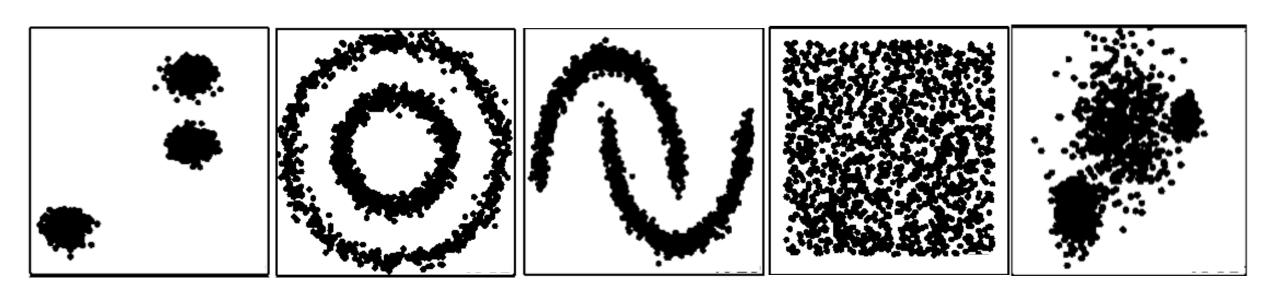
From scikit-learn.org 2

### The Data Mining world



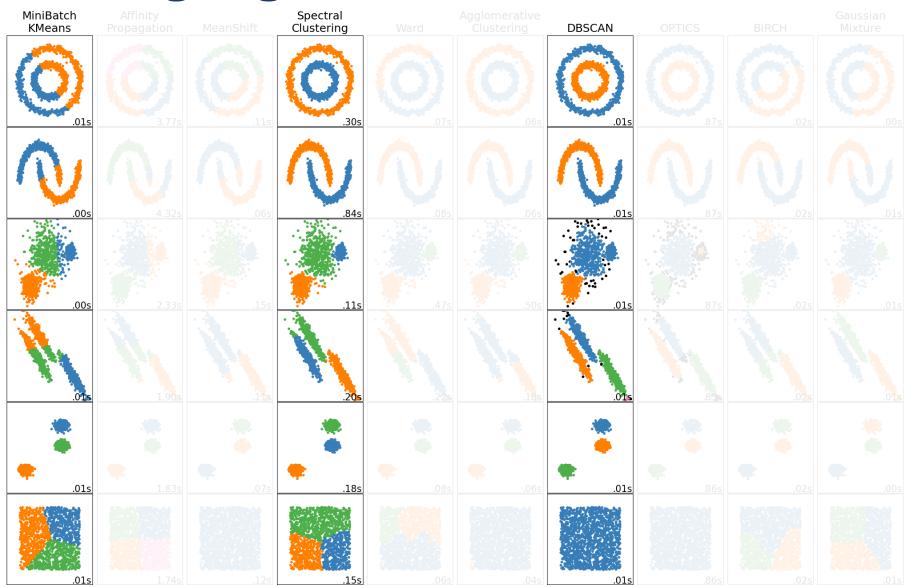
From scikit-learn.org

### Clustering (i.e., unsupervised learning)



Known number of clusters? Flat geometry? Even cluster size? Outliers? Centroids needed?

## **Clustering algorithms**



### How does k-means work?

**Input**: K, set of points  $x_1...x_n$  (can be in N-dimensional)

Place centroids, c<sub>1</sub>..., c<sub>n</sub> at random locations

Repeat until convergence:

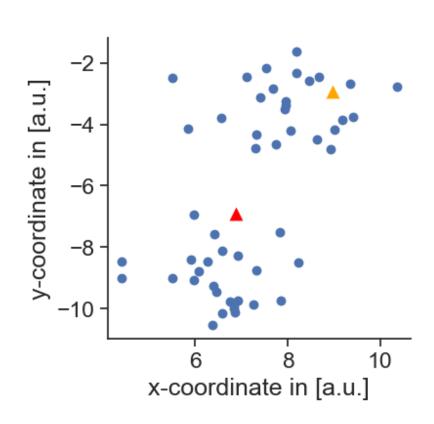
For each point x<sub>i</sub>:

Find nearest centroid  $c_{j} = \arg \min_{j} D(x_{i}, c_{j})$ Assign the point  $x_{i}$  to cluster j

For each cluster j = 1...K:

Compute the centroid mean for all points in one cluster and update the centroid

$$c_j(a) = \frac{1}{n_j} \sum_{x_i \to c_j} x_i(a)$$



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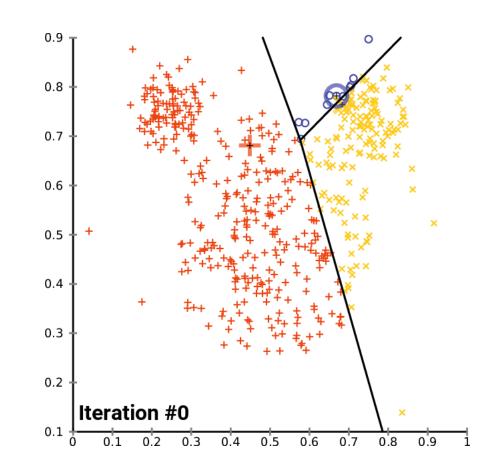
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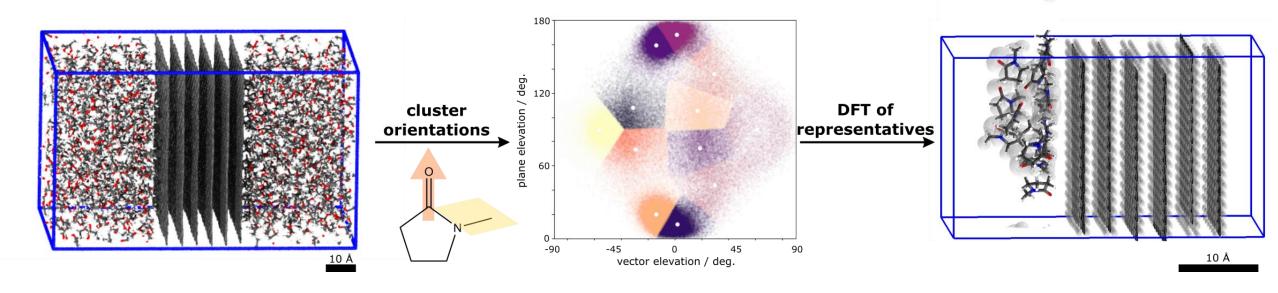
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## [Example] k-means vs solvent-graphite interactions

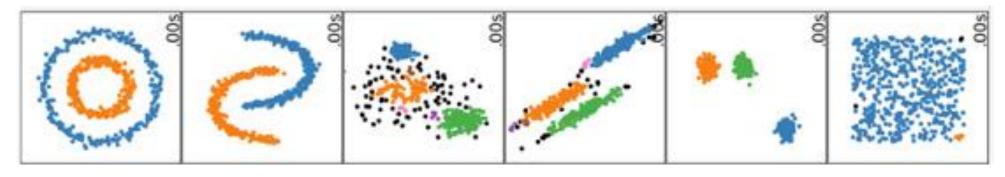
- k-means useful when number of clusters can be estimated, and cluster are approximately circular. Useful if cluster centres are required.
- Example: Molecular Dynamics simulation of graphite immersed in solvents.
   Centroids as representatives of >100k individual solvent-graphene interactions



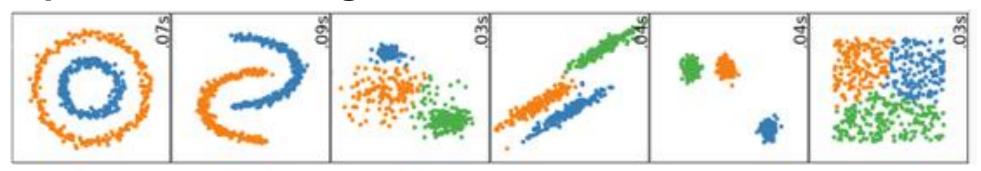
### Density-based and spectral clustering

Useful when number of clusters is unknown, or clusters are not circular

#### **DBSCAN**



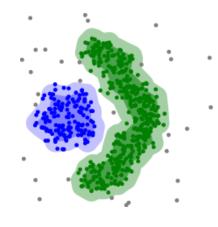
#### **Spectral Clustering**



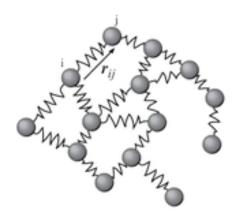
### Density-based and spectral clustering

Useful when number of clusters is unknown, or clusters are not circular

**DBSCAN** 



## Spectral Clustering



- Find points in the  $\varepsilon$  neighbourhood of every point, identify core points with more than n neighbours.
- Find the connected components of core points on the neighbour graph, ignoring all non-core points.
- Assign each non-core point to a nearby cluster if the cluster is an ε neighbour, otherwise assign to noise otherwise.

- Calculate the Laplacian
- Calculate the first k eigenvectors
- Consider the matrix formed by the first k-eigenvectors
- Cluster the graph nodes based on these features (e.g., k-means)

# [Example] DBSCAN for noise detection in EM maps

- Load electron density map and chose a threshold t
- Place pseudoatoms where intensity > t
- Cluster beads and delete small clusters (<1% of total beads</li>

