

Class: Machine Learning

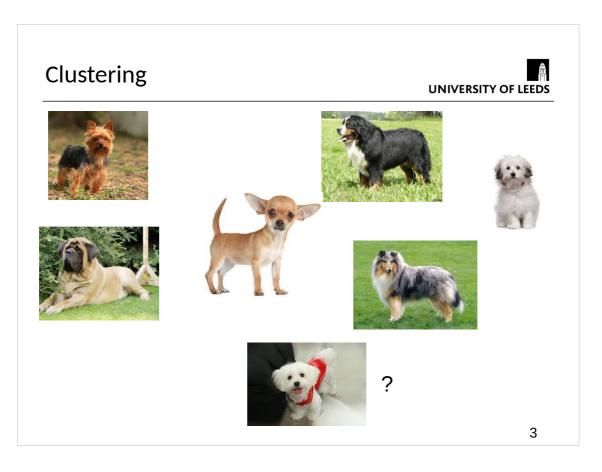
Unsupervised learning: k-means

Instructor: Matteo Leonetti

Learning outcomes



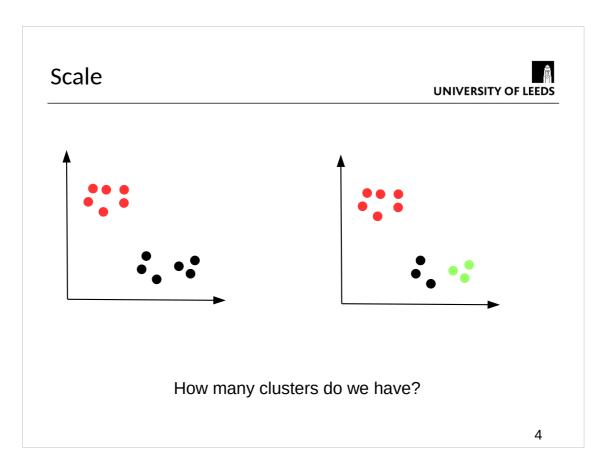
• Apply the k-means algorithm to a dataset



We go now back to the first lecture, when we discussed *clustering*.

Clustering is the problem of grouping elements into *clusters*, because of their proximity in some sense.

This is an *unsupervised* problem, so there are no labels, and therefore no correct or incorrect clustering. It is up to the design to decide whether a certain method and parameters give an acceptable result.

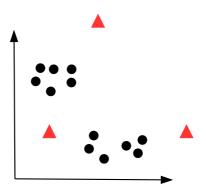


Depending on the granularity with which you look at a problem, the number of cluster may change.

In this example, considering either two or three clusters seems equally reasonable.

K-means





- 1. Choose the number of clusters (in the example: k=3)
- 2. Place k centroids randomly (the triangles)

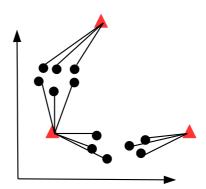
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We look now at the simplest and most popular clustering algorithm: k-means.

The algorithm begins by choosing the number of clusters we are going to find k, and placing k cluster centres (or centroids) randomly in the feature space.

K-means



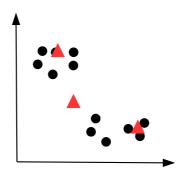


- 3. Identify the closest centroid to each point
- 4. Compute the new centroids for the clusters

6

K-means

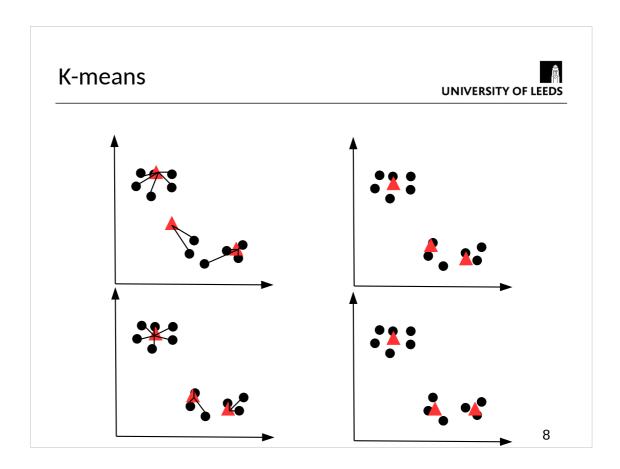




5. Repeat until the centroids do not move

Cluster new points with the closest cluster centre (centroid)

7



K-means - characteristics



Very easy to implement



You have to choose the number of clusters

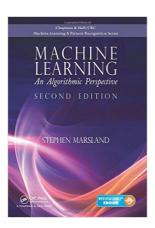


Subject to local minima (clusters depend on initial positions of the centroids)



∼Yet, a popular first thing to try!

9



Chapter 14 (intro) Chapter 14.1