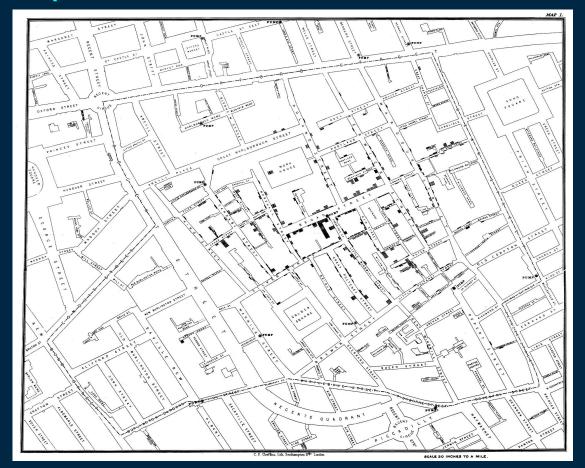
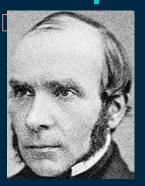


Final thoughts

Once upon a time ...

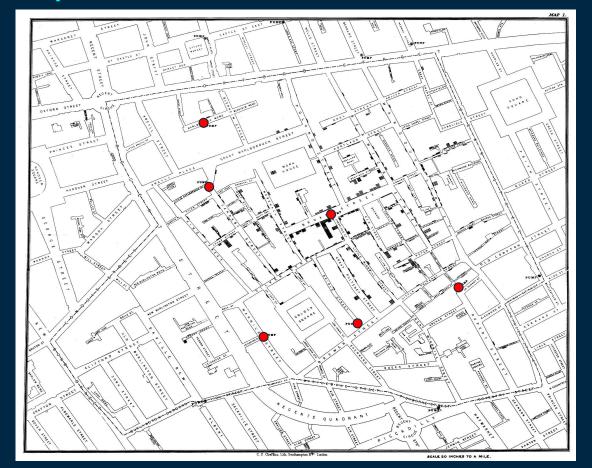


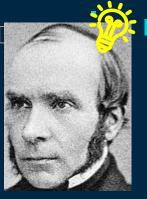


John SNOW

British physician (1813 – 1858)

Once upon a time ...



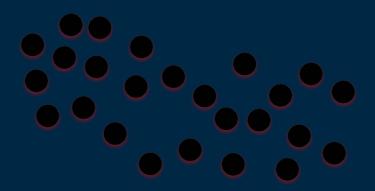


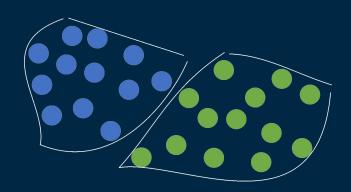
John SNOW

British physician (1813 – 1858)

Clustering

- The organization of unlabeled data into similarity groups called clusters.
- A cluster is a collection of data items which are "similar" between them, and "dissimilar" to data items in other clusters





Proximity measure

Criterion function

Algorithm



1. Proximity measure

• Similarity s(x,y): Large if x and y are similar.

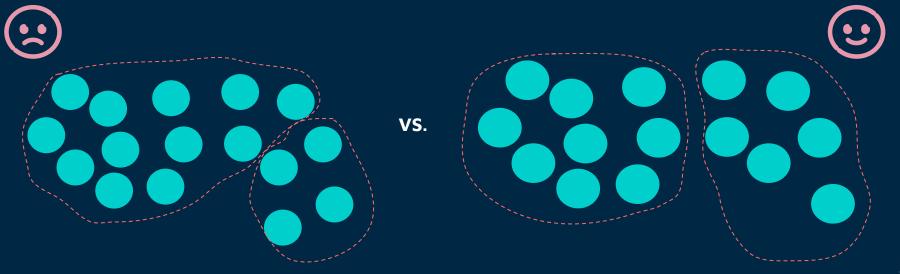
• Dissimilarity (distance) d(x,y): small if x and y are similar.

Small s, Large d

Large s, Small d

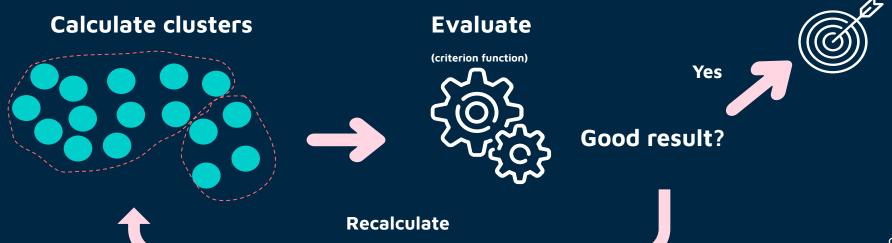
2. Criterion function

A formula to evaluate the quality of the clustering



3. Algorithm to compute the clustering

Iterate over data to optimize the criterion function



Clustering techniques Clustering Hierarchical Bayesian **Devise Decision Agglomerative Partitional** Nonparametric based Centroid Model Graph **Spectral** based Theoretic

Clustering techniques Clustering Hierarchical Bayesian **Devise Decision Agglomerative Partitional** Nonparametric based K-means Centroid **Spectral** Model Graph

Theoretic

based

Partitioning the given data into k clusters :

- Each cluster has a cluster center, called centroid.
- Centroid is the mean point of the cluster.
- **K** is the **clusters number** specified by the user.

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How can we calculate the centroid of a cluster?

Sum of Squared Error (SSE)

$$SSE = \sum_{j=1}^{k} \sum_{\mathbf{x} \in C_j} d(\mathbf{x}, \mathbf{m}_j)^2$$

- 1. **Cj** is the cluster number j.
- 2. **m**j the **centroid** of cluster Cj.
- 3. d(x,y) is the **distance** between x and y.

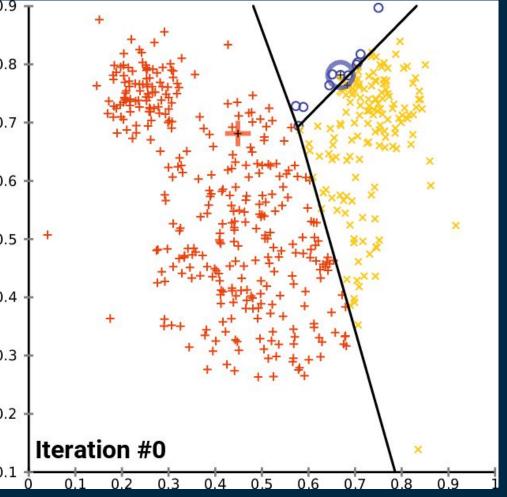
Partitioning the given data into k clusters :

- Choose k (random) data points as the initial centroids.
- Assign each data point to the closest centroid.
- Re-compute the centroids using the current cluster memberships.
- 4. Repeat 2 & 3 untill **SSE** is **minimized** or **stabillized**.

A cluster is a:

Centroid : Ci

Sum of Distances d(Ci,X)

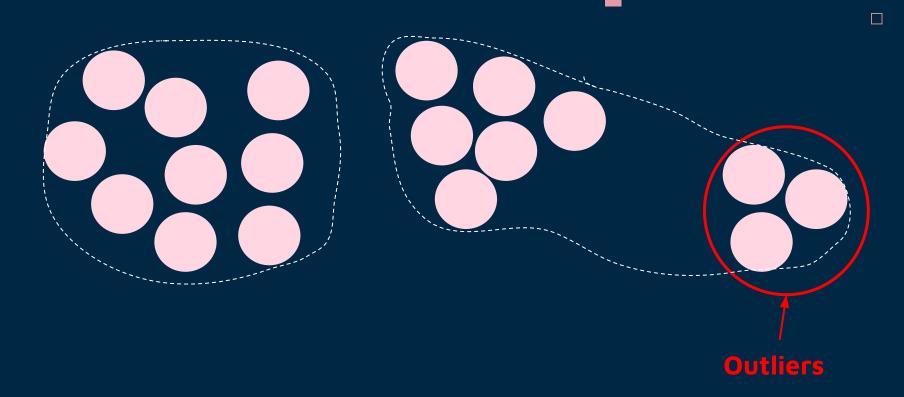


Why K-means?

• **Simple**: easy to understand and to implement

- **Efficient**: time complexity: O(tkn) for the Euclidean distance
 - on: number of points
 - o k : number of clusters
 - o t : number of iterations

Yes! ... BUT



Code Demo

Check my github Repo:

https://github.com/Hamzandj/Open-Week-3.0

In a Nutshell

- **Clustering**: unsupervised machine learning.
- The clustering output can serve to train a supervised model.
- A good understanding of the data is mandatory.
- Results interpretation is a key to have a good model that helps in decision making.
- Clustering can be used to **solve complex problems**









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