

Clustering

**MACHINE LEARNING**



# Topics

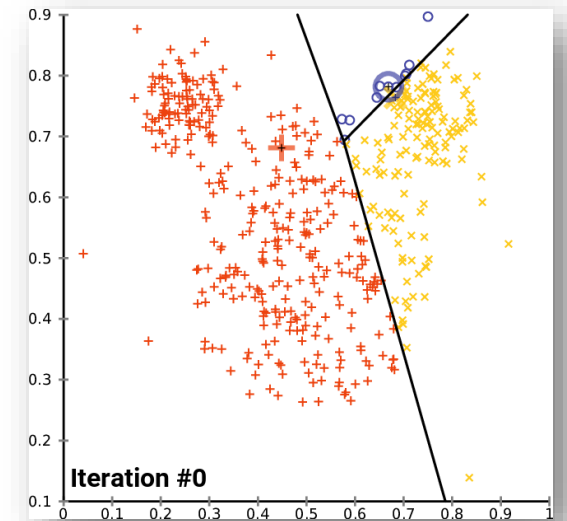
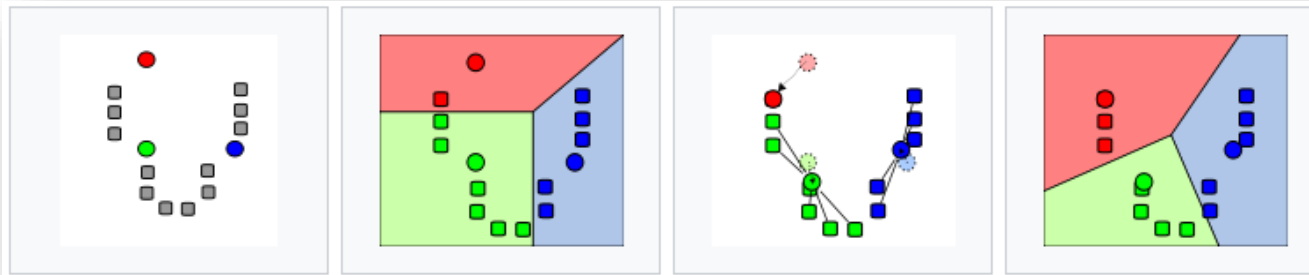
- **Introduction:** GUI and basic calculations
- **Coding 1:** Scripts, style, and variable classes
- **Coding 2:** Control statements and loops
- **Visualization 1:** Basics, subplots, get and set
- **Coding 3:** Functions
- **Visualization 2:** Descriptive plots
- **Coding 4:** Basic input and output
- **Visualization 3:** Distribution and 3D plots
- **Coding 5:** Input and output specials – last lecture before holidays
- **Machine Learning 1:** Introduction and dimension reduction
- **Machine Learning 2: Clustering**
- **Machine Learning 3:** Classification
- **Coding 6:** Efficiency and debugging basics
- **Coding 7:** Advanced functions and debugging

# Cluster Analysis

- Group a **set of data points** (each with a number of features) in a way that objects in the **same group** (cluster) are **more similar to each other than those in other groups**
- Common tool for exploratory data analysis and statistics
- Distance in n-D is generally the used measure
  - Distance in n-D can be defined in various ways, not just euclidean
  - Another measure can be density
- Invented in 1932 (Driver & Kroeber)
  - Famously used for personality analysis (Cattell, 1943)

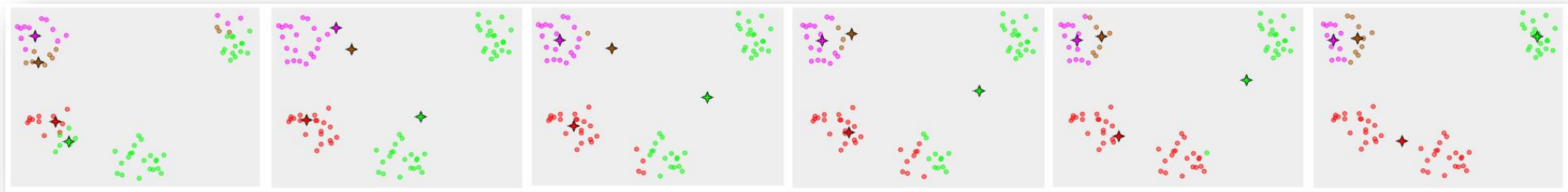
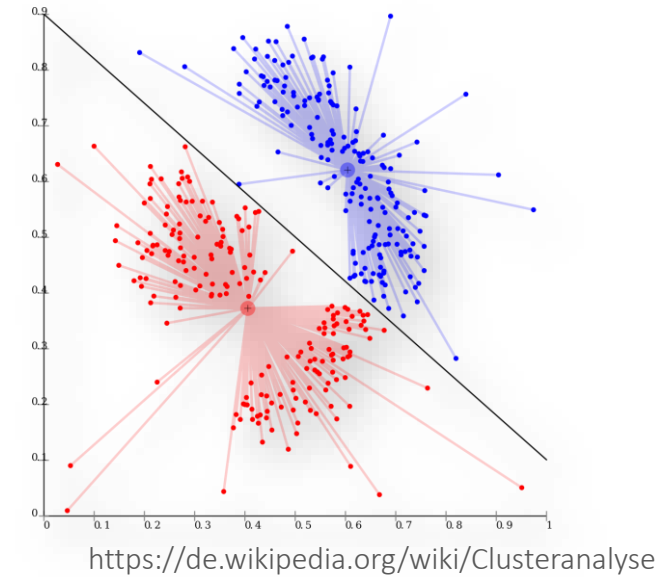
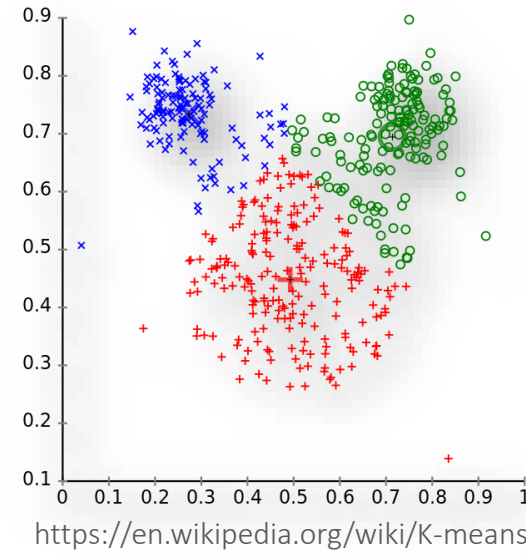
# K-Means

- A number of clusters (k) are represented by a centroid (artificial data point)
- The centroids move around and „collect“ the closest data points
- Distance within the cluster (spread) is minimized while distance to other clusters is maximized
- Random initialization, iterative approach
  - Not converging to the same solution all the time!



# K-Means Limitations

- K needs to be specified!
- Works well only with spherical distributions
- Assumes same-sized clusters
- Can run into local minima



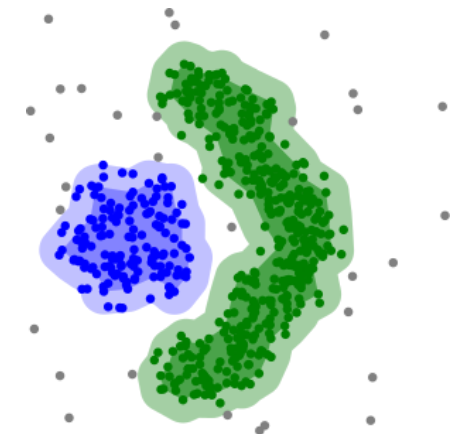
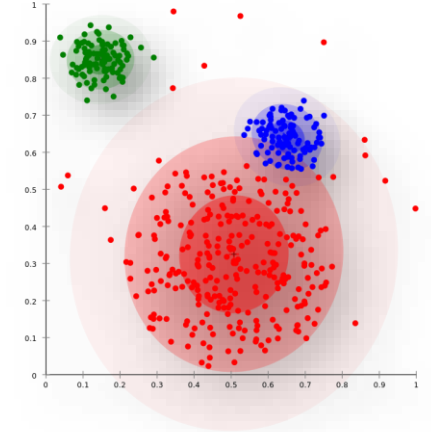
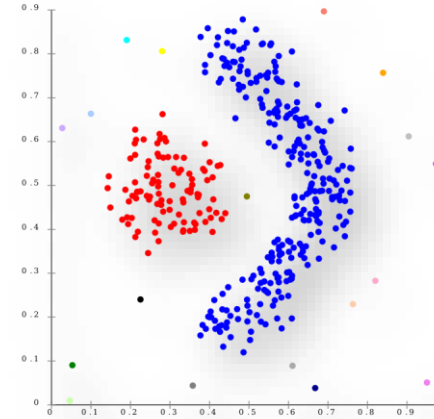
[https://en.wikipedia.org/wiki/K-means\\_clustering](https://en.wikipedia.org/wiki/K-means_clustering)

# Other Algorithms

- Distribution-based
  - E.g. gaussian mixture models
- Connectivity-based
  - Clusters based on linkage of data points
  - Nice for swiss roll
- Density-based
  - Arbitrarily shaped clusters of dense regions
  - Sparse regions are „noise“/border-points
  - Have problems with overlapping boundaries



-[https://en.wikipedia.org/wiki/Swiss\\_roll](https://en.wikipedia.org/wiki/Swiss_roll)



<https://de.wikipedia.org/wiki/Clusteranalyse>