

# **Computational Intelligence**

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Prof. Dr. Günter Rudolph

Lehrstuhl für Algorithm Engineering (LS 11)

Fakultät für Informatik

TU Dortmund

• Fuzzy Clustering





## idea:

- select, say, 2000 men at random and measure their "body lengths"
- arrange these 2000 men into five disjoint groups

## such that

- deviations from mean of group as small as possible
- differences between group means as large as possible





arm's length, collar size, chest girth, ... **numerical example**: 1000 points uniformly sampled in [0,1] x [0,1]  $\rightarrow$  form 5 cluster



given data points  $x_1, x_2, ..., x_N \in \mathbb{R}^n$ 

<u>objective:</u>

group data points into cluster such that

- points within cluster are as homogeneous as possible
- points across clusters are as heterogeneous as possible

 $\Rightarrow$  crisp clustering is just a partitioning of data set {  $x_1, x_2, ..., x_N$  }, i.e.,

$$\bigcup_{k=1}^{K} C_k = \{ \mathbf{x}_1, \mathbf{x}_2, ..., \mathbf{x}_N \} \text{ and } \forall j \neq k : C_j \cap C_k = \emptyset$$

where  $C_k$  is Cluster k and K denotes the number of clusters.

Constraint:  $\forall k = 1, \dots, K : |C_k| \ge 1$  hence  $1 \le K \le N$ 

**Complexity:** How many choices to assign N objects into K clusters?

more precisely:

- $\rightarrow$  objects are distinguishable / labelled
- $\rightarrow$  clusters are nondistinguishable / unlabelled and nonempty

$$\Rightarrow \text{Stirling number of 2nd kind} \quad S(\mathsf{N},K) = \frac{1}{K!} \sum_{i=1}^{K} (-1)^{k-i} \binom{k}{i} \cdot i^{\mathsf{N}} \quad \sim \frac{K^{\mathsf{N}}}{K!}$$

${\sf N}/K$	1	2	3	4	5	
10	1	511	9,330	34,105	42,525	
11	1	1,023	28,501	145,750	246,730	$S(100,5) = 6.6 \times 10^{67}$ $S(1000,5) = 7.8 \times 10^{696}$ $S(2000,5) = 7.2 \times 10^{1395}$
12	1	2,047	86,526	611, 501	1,379,400	
13	1	4,095	261,625	2,532,530	7,508,501	
14	1	8,191	788,970	10,391,745	40,075,035	$S(2000, 5) = 7.3 \times 10^{-15}$
15	1	16,383	2,375,101	42,355,950	210,766,920	

 $\Rightarrow$  enumeration hopeless!  $\Rightarrow$  iterative improvement procedure required!

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define objective function idea:

that measures compactness of clusters and quality of partition

- $\rightarrow$  elements in cluster C<sub>i</sub> should be as homogeneous as possible!
- $\rightarrow$  sum of squared distances to unknown center y should be as small as possible

$$\rightarrow$$
 find y with  $\sum_{i \in C_j} d(x_i, y)^2 \rightarrow \min!$ 

typically,  $d(x_i, y) = ||x_i - y|| = \sqrt{(x_i - y)'(x_i - y)}$ 

$$\frac{d}{dy} \sum_{i \in C_j} (x_i - y)'(x_i - y) = -2 \sum_{i \in C_j} (x_i - y) \stackrel{!}{=} 0$$

$$\Rightarrow \sum_{i \in C_j} x_i \stackrel{!}{=} \sum_{i \in C_j} y = |C_j| \cdot y \qquad \Rightarrow y = \frac{1}{|C_j|} \sum_{i \in C_j} x_i =: \bar{x}_j$$

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# Hard / Crisp Clustering

$$\rightarrow$$
 find partition  $C = (C_1, \dots, C_K)$  with  $D(C) = \sum_{j=1}^K \sum_{i \in C_j} d(x_i, \bar{x}_j)^2 \rightarrow \min!$ 

### Definition

A partition  $C^*$  is optimal if

$$D(C^*) = \min\{ D(C) \, : \, C \in P(\mathsf{N}, K) \,\}$$

where P(N, K) denotes all partitions of N elements in K clusters.

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## Theorem

$$\min_{C \in P(\mathsf{N},K)} D(C) = \max_{C \in P(\mathsf{N},K)} \sum_{j=1}^{N} |C_j| \cdot \|\bar{x}_j - \bar{x}\|$$
  
where  $\bar{x}$  is the mean of all  $x$ 

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$$\begin{aligned} \forall k = 1, \dots, K: \text{ set } C_k &= \emptyset \\ \forall x \in \{x_1, \dots, x_N\}: \text{ assign } x \text{ to some cluster } C_k \\ \text{set } t = 0 \text{ and } D^{(t)} = \infty \end{aligned}$$

repeat

$$\begin{split} t &= t+1 \\ \forall k = 1, \dots, K: \ \bar{x}_k = \frac{1}{|C_k|} \sum_{x \in C_k} x \\ \forall i = 1, \dots, N: \ d_{ik} = d(x_i, \bar{x}_k) \quad \text{ distance to center of cluster } k \\ \text{ let } k^* \text{ be such that } d_{ik^*} &= \min\{d_{ik} : k = 1, \dots, K\} \\ \text{ assign } x_i \text{ to } C_{k^*} \\ D^{(t)} &= \sum_{k=1}^K \sum_{x \in C_k} d(x, \bar{x}_k) \\ \text{ until } D^{(t-1)} - D^{(t)} < \varepsilon \end{split}$$

## objective for crisp clustering:

find partition 
$$C = (C_1, \ldots, C_K)$$
 with  $D(C) = \sum_{j=1}^K \sum_{i \in C_j} d(x_i, \bar{x}_j)^2 \to \min!$ 

 $\rightarrow$  rewrite objective:

find partition 
$$C = (C_1, \dots, C_K)$$
 with  $D(C) = \sum_{j=1}^K \sum_{i=1}^N u_{ij} \cdot d(x_i, \bar{x}_j)^2 \to \min!$   
expresses membership  $\longrightarrow u_{ij} = \begin{cases} 1 & \text{if } x_i \in C_j \\ 0 & \text{otherwise} \end{cases}$ 

## objective for fuzzy clustering:

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find partition 
$$C = (C_1, \ldots, C_K)$$
 with  $D(C) = \sum_{j=1}^K \sum_{i=1}^N u_{ij}^m \cdot d(x_i, \bar{x}_j)^2 \to \min!$   
 $u_{ij} \in [0, 1] \subset \mathbb{R}, m > 1$ 

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find partition 
$$C = (C_1, \dots, C_K)$$
 with  $D(C) = \sum_{j=1}^K \sum_{i=1}^N u_{ij}^m \cdot d(x_i, \bar{x}_j)^2 \to \min!$ 

#### where

 $u_{ij} \in [0,1] \subset \mathbb{R}$  denotes membership of  $x_i$  to cluster  $C_j$ 

m > 1 denotes a fixed *fuzzifier* (controls / affects membership function)

subject to

$$\sum_{j=1}^{K} u_{ij} = 1 \qquad \forall i = 1, \dots, N$$
$$0 < \sum_{i=1}^{N} u_{ij} < N \qquad \forall j = 1, \dots, K$$

each  $x_i$  distributes membership completely over clusters  $C_1, \ldots, C_K$  $\rightarrow$  normalization at least one element belongs to some extent to a certain cluster, but not all elements to a single cluster

## two questions:

- (a) how to define and calculate centers  $\bar{x}_j$ ?
- (b) how to obtain optimal memberships  $u_{ij}$ ?

ad a) let 
$$d(x_i, \bar{x}_j) = ||x_i - \bar{x}_j||_2$$

$$\frac{d}{d\bar{x}_j} \sum_{i=1}^N u_{ij}^m \cdot (x_i - \bar{x}_j)'(x_i - \bar{x}_j) = -2 \sum_{i=1}^N u_{ij}^m \cdot (x_i - \bar{x}_j) \stackrel{!}{=} 0$$

$$\Leftrightarrow \sum_{i=1}^N u_{ij}^m x_i \stackrel{!}{=} \sum_{i=1}^N u_{ij}^m \bar{x}_j \quad \Leftrightarrow \quad \left[ \bar{x}_j = \frac{\sum_{i=1}^N u_{ij}^m x_i}{\sum_{i=1}^N u_{ij}^m} \right] \quad \rightarrow \text{weighted mean!}$$

ad b) let  $d_{ij} := d(x_i, \bar{x}_j) = ||x_i - \bar{x}_j||_2$ 

apply Lagrange multiplier method:

$$\frac{\partial}{\partial u_{ij}} \sum_{j=1}^{K} \sum_{i=1}^{N} u_{ij}^{m} \cdot d_{ij}^{2} - \sum_{i=1}^{N} \lambda_{i} \left( \sum_{j=1}^{K} u_{ij} - 1 \right) = m u_{ij}^{m-1} \cdot d_{ij}^{2} - \lambda_{i} \stackrel{!}{=} 0$$
without constraints  $\rightarrow u_{ij}^{*} = 0$ 

$$u_{ij}^{*} = \left( \frac{\lambda_{i}}{m \cdot d_{ij}^{2}} \right)^{\frac{1}{m-1}}$$

$$\sum_{j=1}^{K} \sqrt{u_{ij}} = \sum_{j=1}^{K} \left( \frac{\lambda_{i}}{m \cdot d_{ij}^{2}} \right)^{\frac{1}{m-1}} = \sum_{j=1}^{K} \frac{\lambda_{i}^{\frac{1}{q}}}{(m \cdot d_{ij}^{2})^{\frac{1}{q}}} = \lambda_{i}^{\frac{1}{q}} \sum_{j=1}^{K} \frac{1}{(m \cdot d_{ij}^{2})^{\frac{1}{q}}} \stackrel{!}{=} 1$$

$$\sum_{set q = m-1}^{K} \sum_{i=1}^{K} \frac{1}{(m \cdot d_{ij}^{2})^{\frac{1}{q}}} = \lambda_{i}^{*} \sum_{j=1}^{K} \frac{1}{(m \cdot d_{ij}^{2})^{\frac{1}{q}}} \stackrel{!}{=} 1$$

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after insertion:

$$u_{ij}^{*} = \left(\frac{1}{m \cdot d_{ij}^{2}} \left[\frac{1}{\sum_{k=1}^{K} \left(\frac{1}{m \cdot d_{ik}^{2}}\right)^{\frac{1}{m-1}}}\right]^{m-1}\right)^{\frac{1}{m-1}}$$

 $= \left[\sum_{k=1}^{K} \left(\frac{d_{ij}}{d_{ik}}\right)^{\frac{2}{m-1}}\right]^{-1}$ 

choose  $K \in \mathbb{N}$  and m > 1choose  $u_{ij}$  at random (obeying constraints) repeat

 $\begin{aligned} \forall j = 1, \dots, K: \text{ calculate centers } \bar{x}_j \\ \forall i = 1, \dots, N: \\ \text{let } J_i = \{j : x_i = \bar{x}_j\} \\ \text{if } J_i = \emptyset \text{ determine memberships } u_{ij} \\ \text{else} \\ \text{choose } u_{ij} \text{ such that } \sum_{i \in J_i} u_{ij} = 1 \end{aligned}$ 

and  $u_{ij} = 0$  for  $j \notin J_i$ until  $D(C^{(t)}) - D(C^{(t+1)}) < \varepsilon$  or  $t = t_{max}$ 

## problems:

- <u>choice of K</u> calculate quality measure for each #cluster; then choose best
- <u>choice of m</u> try some values;
   typical: m=2;
   use interval → fuzzy type-2



 $u_{ii} = 1 / |J_i|$  for  $j \in J_i$  appears plausible

but: different values algorithmically better

 $\rightarrow$  cluster centers more likely to separate again ( $\rightarrow$  tiny randomization?)

## • Partition Coefficient

$$\mathsf{PC}(C_1, \dots, C_K) = \frac{1}{N} \sum_{i=1}^N \sum_{j=1}^K u_{ij}^2$$

( "larger is better" )

# • Partition Entropy