# Exercise Sheet 8

Additional Exercise Application of Clustering and Classification Algorithms

This exercise will be discussed on June 14th/20th. The deadline for submissions is June 13th, 23:59:00 CEST.

• As presented during the lecture on May 25th: sign up at

### http://learning-challenge.de/

with your URZ login name as Nick- or Teamname and your real name as real name.

- Choose the group IDA2016.
- You can now select the *Classification* and the *Clustering* rounds. Parameterize the algorithms available in each round to obtain the best possible result for each data set.
- *Classification* tasks will be ranked by the achieved accuracy; *Clustering* tasks by the minimum of Homogenity and Completeness.
- The five best students per round (across all data sets) (right hand side, scored with *stars*), will receive bonus credit for the exercise.

**Exercise 29** Fuzzy Clustering

Consider the one-dimensional data set

1, 3, 4, 5, 8, 10, 11, 12.

We want to process this data set with fuzzy c-means clustering with c = 2 (two clusters) and a fuzzifier of w = 2. Assume that the cluster centers are initialized to 1 and 5. Execute one step of alternating optimization as it is used for fuzzy clustering, i.e.:

- a) Compute the membership degrees of the data points for the initial cluster centers!
- b) Compute new cluster centers from the membership degrees computed in this way!

#### **Exercise 30** Expectation Maximization

Consider again the one-dimensional data set used in exercise 29, which we want to process in this exercise with the expectation maximization algorithm to estimate the parameters of a mixture of two normal/Gaussian distributions. Let the prior probabilities of the two clusters be fixed to  $\theta_i = \frac{1}{2}$  and the variances to  $\sigma_i^2 = 1$ , i = 1, 2. (That is, only the expected values of the normal distributions — the cluster centers — are to be adapted.) Use the same values for the initial cluster centers as in exercise 29, that is, 1 and 5. Computeone expectation step and one maximization step, i.e.:

- a) Compute the posterior probabilities of the data points for the initial cluster centers!
- b) Estimate new cluster centers from the data point weights computed in this way!

#### **Exercise 31** Fuzzy Clustering

Consider the objective function of fuzzy clustering with a fuzzifier w = 1, that is,

$$J(\mathbf{X}, \mathbf{B}, \mathbf{U}) = \sum_{i=1}^{c} \sum_{j=1}^{n} u_{ij} d^{2}(\beta_{i}, \vec{x}_{j}),$$

which is to be minimized under the constraint

$$\forall j \in \{1, \dots, n\}: \quad \sum_{i=1}^{c} u_{ij} = 1.$$

Show that one obtains a hard/crisp assignment of the data points even if the membership degrees  $u_{ij}$  may come from the interval [0, 1]. That is, show that for the minimum of the objective function J it is  $\forall i \in \{1, \ldots, c\} : \forall j \in \{1, \ldots, n\} : u_{ij} \in \{0, 1\}$ . (Hint: You may find it easier to consider the special case c = 2 (two clusters) and to examine the term for a single data point  $\vec{x}_j$ . Then generalize the result.)

#### **Exercise 32** Agglomerative Clustering

Let the following one-dimensional data set be given:

Process this data set with hierarchical agglomerative clustering using

- a) the centroid method,
- b) the single linkage methode,
- c) the complete linkage methode!

Draw a dendrogram for each case!

## Additional Exercise Fuzzy Clustering

Consider the objective function

$$J(\mathbf{X}, \mathbf{B}, \mathbf{U}) = \sum_{i=1}^{c} \sum_{j=1}^{n} u_{ij} d^{2}(\beta_{i}, \vec{x}_{j}),$$

which is to be minimized under the constraint

$$\forall j \in \{1, \dots, n\}: \quad \sum_{i=1}^{c} \sqrt{u_{ij}} = 1.$$

Derive the update formulae for the membership degrees and the cluster centers using the Euclidean distance. How does the result differ from standard fuzzy clustering with a fuzzifier w = 2? (In particular, consider the cluster centers.)