# K-means Clustering

In this problem set we will implement and apply the standard (batch) K-means algorithm, the online version, and the "soft" clustering procedures. The file cluster.dat contains a data set of p=500 (2-dimensional) observations generated from four different Gaussians with four different means.

### 10.1 K-means Clustering (3 points)

Write a program that implements the *standard* version of K-means clustering and partitions the given data set into K clusters. Repeat the clustering procedure for different initializations of the prototypes and K = 2, 3, 4, 5, 6, 7, 8. Include the following steps:

#### Initialization -

- ullet Set the initial prototypes  ${\bf w}_q$  randomly around the data set mean
- Set the maximum number of iterations  $t_{max}$ , e.g. 5

#### Optimization -

Implement the k-means update (see lecture notes). Each iteration should contain the following two steps

- assign all datapoints to their closest prototype
- re-compute the new positions of the prototypes for this assignment

### Visualization -

- (a) Visualize data points and prototypes for each iteration in a sequence of scatter plots.
- (b) Plot the error function E against the iteration number t

$$E_{\{m_q^{(\alpha)}\},\{\mathbf{w}_q\}} = \frac{1}{2p} \sum_{q=1}^K \sum_{\alpha=1}^p m_q^{(\alpha)} \|\mathbf{x}^{(\alpha)} - \mathbf{w}_q\|^2$$

(c) Create a plot (Voronoi-Tesselation) to show how the resulting solution assigns different regions of input space (e.g. new data points  $\mathbf{x} \in \mathbb{R}^2$ ) to the different clusters.

## 10.2 Online K-means Clustering (3 points)

Write a program that implements the *online* version of K-means clustering (see lecture notes) and partitions the given data set into K=4 clusters. Include the following steps:

### Initialization -

- ullet Set the initial prototypes  ${\bf w}_q$  randomly around the data set mean
- Select an initial learning step  $\eta_0$
- Set the maximum number of iterations  $t_{max}$ , e.g. equal to the data set size p.

#### Optimization -

• Choose a suitable  $\tau < 1$  and implement online K-means clustering using the following "annealing" schedule for  $\eta$ :

$$\eta_t = \eta_0 \quad \text{ for } t = 0,..., rac{t_{max}}{4} \quad \text{ and } \quad \eta_t = au \eta_{t-1} \quad \text{ for } t = rac{t_{max}}{4} + 1,...,t_{max}$$

### Visualization -

- (a) Visualize data points and the prototypes for each iteration in a sequence of scatter plots, but only show the first, the final, and four intermediate iterations. In the final plot additionally show for each cluster the sequence of centroid positions  $\mathbf{w}_q$  by connecting them with straight lines.
- (b) Plot the error function E (as above) against the iteration number t

# 10.3 Soft K-means Clustering (4 points)

"Soft" clustering is a mean-field approximation of pairwise clustering with squared Euclidean distances. Implement the *soft* K-means algorithm with squared Euclidean distances (cf. lecture notes) and apply it to the same data as before. Proceed as follows:

- (a) Set K=8 initial prototypes  $\mathbf{w}_q$  randomly around the data set mean and choose a convergence tolerance  $\gamma$ .
- (b) For fixed  $\beta$  (no annealing), let the optimization procedure run until convergence, that is  $\|\mathbf{w}_q^{new} \mathbf{w}_q^{old}\| < \gamma \ \forall q$ . Repeat this for different  $\beta \in [0.2, 20]$  e.g. in steps of  $\Delta\beta = 0.2$ . Use the same initial prototypes for all runs.
- (c) Visualize the data set, initial and final prototypes for each (fixed)  $\beta$  in one scatter plot. Therein show how "soft" each data point  $\mathbf{x}^{(\alpha)}$  is assigned to a cluster, e.g. by scaling the brightness of the respective plot symbol with the largest assignment probability, i.e.,  $\max_{q} \langle m_q^{(\alpha)} \rangle$ .
- (d) Plot in two separate subplots the first and second coordinate of the final prototypes  $\mathbf{w}_q$  against the  $\beta$  (i.e. K lines per subplot) and interpret the result.
- (e) In additional simulations, run the optimization for K=2,4,6,8 using an annealing schedule: increase  $\beta$  after each iteration. E.g.  $\beta_0=0.2,\ \tau=1.1,\ \beta_{t+1}=\tau\beta_t$ .
- (f) Show the data set, initial and final prototypes of the "annealed" clustering solutions for K=2,4,6,8 in a scatter plot. How "soft" are data points assigned now?

Total points: 10