

## Machine Learning II

Exercise Sheet 4

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**Exercise 1.** Consider a data set  $\mathbf{X} = (\mathbf{x}_1, \dots, \mathbf{x}_N)^T$  in which the observations  $\{\mathbf{x}_n\}$  are assumed to be drawn independently from a multivariate Gaussian distribution. Write down the log likelihood function  $\ln p(\mathbf{X}|\mu, \Sigma)$  for this distribution. Estimate the mean  $\mu$  by maximum likelihood. Interpret the result.

**2 points**

**Exercise 2.** Mixture of Gaussians: the conditional distribution of  $\mathbf{x}$  given a particular value for the latent variable  $\mathbf{z}$  is assumed to be a Gaussian, i.e.  $p(\mathbf{x}|z_k = 1) = \mathcal{N}(\mathbf{x}|\mu_{\mathbf{k}}, \Sigma_{\mathbf{k}})$ . The probability  $p(z_k = 1) = \pi_k$  is called mixing coefficient. Derive an expression for the conditional distribution of  $\mathbf{z}$  given  $\mathbf{x}$ , i.e.  $p(z_k = 1|\mathbf{x})$  in terms of  $\pi_k$  and  $\mathcal{N}(\mathbf{x}|\mu_{\mathbf{k}}, \Sigma_{\mathbf{k}})$ . Use the derived mathematical expression to interpret  $\gamma(z_k) \equiv p(z_k = 1|\mathbf{x})$  (with words and by drawing a cartoon).

**2 points**

**Exercise 3.** a) Write down the log of the likelihood function of the Gaussian mixture model  $\ln p(\mathbf{X}|\pi, \mu, \Sigma)$ . Explain why it is difficult to find the parameters for which this function has a maximum.

**1 point**

b) Write down the conditions for  $\mu_k$  that must be satisfied at a maximum of the log likelihood function. Derive the expression

$$\mu_k = \frac{1}{N_k} \sum_{n=1}^N \gamma(z_{nk}) \mathbf{x}_n$$

with

$$N_k = \sum_{n=1}^N \gamma(z_{nk})$$

end interpret the result.

**2 points**

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c) Now maximize the log likelihood function with respect to the mixing coefficients  $\pi_k$ . To do so maximize

$$\ln p(\mathbf{X}|\pi, \mu, \Sigma) + \lambda \left( \sum_{k=1}^K \pi_k - 1 \right)$$

where  $\lambda$  is a Lagrange multiplier implementing the constraint that the  $\pi_k$  must sum to 1. Show that

$$\pi_k = \frac{N_k}{N}$$

and interpret the result.

**2 points**

d) Assuming  $\Sigma$  is fixed, why do these expression not constitute a closed-form solution for the parameters of the mixture model?

**1 point**