

# Machine Learning – WS'12

## Exercise-7

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### *Linear Discriminant Analysis*

#### **Problem-1:**

Scientists of Iowa state have acquired the samples of water from state's reservoirs. Some water samples contain a particular bacterium (class 1) while other do not contain (class 2). The samples have two observed variables  $x_1$  (pH) and  $x_2$  (Nitrogen content). The number of instances in each class, the average of the variable vectors and the covariance matrices for the two types of water samples are given as follows:

$$\begin{aligned} \hat{n}_1 &= 13, & n_2 &= 10 \\ \hat{\mu}_1 &= \begin{pmatrix} 7.8 \\ 45 \end{pmatrix}, & \hat{\mu}_2 &= \begin{pmatrix} 5.9 \\ 20.8 \end{pmatrix} \\ \hat{\Sigma}_1 &= \begin{pmatrix} 0.5 & 4.5 \\ 4.5 & 147.2 \end{pmatrix}, & \hat{\Sigma}_2 &= \begin{pmatrix} 0.1 & 0.2 \\ 0.2 & 24.2 \end{pmatrix} \end{aligned}$$

- Determine discriminant function for the two classes.
- Assign the observation  $x = (6 \ 52.5)^T$  to one of the classes.
- Are these LDA or QDA ?

#### **Problem 2:**

- Quadratic discriminant analysis generates non-linear decision boundaries. Why it is sometimes necessary to allow non-linear decision boundaries? What are the disadvantages to use more complex decision boundaries?
- Why QDA produces non-linear decision boundaries? But LDA does not? As one could transform the data to use LDA rather than complex QDA (see LDA-QDA discussion in lecture), the distinction between two classes  $Y=\{0,1\}$  is sufficient? Can the disadvantages of complex QDA be eliminated?

#### **Problem-3:**

Suppose we have the following training sample:

	$t = 1$			$t = 2$		
$x_1$	2	4	3	5	3	4
$x_2$	12	10	8	7	9	5

We assume  $x_1$  and  $x_2$  follow a bivariate normal distribution within each group, where the covariance matrix is assumed to be the same in both groups.

- a) Estimate the group means, covariance matrix (unbiased), and group prior probabilities  $\pi_k = p(Y = k)$  from this training sample.
- b) Estimate the linear discriminant functions  $a_1(x_1, x_2)$  and  $a_2(x_1, x_2)$  for class 1 and 2 respectively.
- c) Give one linear classification rule for this problem and construct a *confusion matrix* by applying to the training sample. What is in-sample error rate?
- d) Draw the border lines between the areas that belong to class 1 and 2, respectively, in a scatter plot of the data. Can you find a straight line that has low in-sample error rate?