## Assignment no. 2

**Exercise 2 (10 points)** Consider a one-dimensional classification problem with  $X = \mathbb{R}$  and  $Y = \{-1, 1\}$ . The marginal distribution of labels is given as follows:

$$p(y = -1) = \frac{3}{4}$$
  $p(y = +1) = \frac{1}{4}$ 

The conditional distribution p(x | y) is given in the following way:

$$p(x \mid y = -1) = \frac{1}{2\sqrt{\pi}}e^{-\frac{(x+1)^2}{4}}$$
Gaussian with  $\mu = -1$  and  $\sigma^2 = 2$ 

$$p(x \mid y = +1) = \frac{\sqrt{2}}{\sqrt{\pi}}e^{-2(x-3)^2}$$
Gaussian with  $\mu = 3$  and  $\sigma^2 = \frac{1}{4}$ 

Visualize the marginal distribution p(x) and the conditional distributions  $p(y = -1 \mid x)$  and  $p(y = +1 \mid x)$ . Guess from the visualization of  $p(y = -1 \mid x)$  and  $p(y = +1 \mid x)$  what the Bayes-optimal classifier is like (**hint:** visualize the two conditional distributions in one plot).

Exercise 3 (20 points) Consider the following one-dimensional regression task: inputs x are uniformly distributed in [-1,3] and targets y are given as

$$y = f(x) = 0.6x^4 + 2x^3 - 8x^2$$

plus independent normally distributed noise with  $\mu = 0$  and  $\sigma^2 = 0.09$ . What are  $E(y \mid x_0)$  and the unavoidable error  $Var(y \mid x_0)$  in this setting?

Perform polynomial regression to illustrate the bias-variance decomposition. To this end, perform the following steps for each degree n = 1, ..., 7:

- 1. Create 200 training sets with l = 20 samples each.
- 2. For each of the training sets, train a polynomial model with degree n and compute the predicted value for  $x_0 = 1.8$ .
- 3. Estimate the squared bias and the variance from the 200 predicted values and compute an overall estimate for the expected prediction error for  $x_0 = 1.8$ .

After having followed these steps, visualize your results appropriately. Discuss how the results illustrate the bias-variance decomposition.

**Submission:** electronically via Moodle:

Please take the submission instructions into account! Deadline: Monday, November 20, 2017, 1:00pm.