

# Problem Set 5 – Loss Functions

DS 542 – DL4DS

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The following problems assume a data set with one input  $x$  and one output  $y$ . The input  $x$  is a binary variable only taking on values zero and one, and  $Pr(y|x = i)$  is normally distributed with mean  $\mu_i$  and standard deviation  $\sigma_i$ . In the training data,  $Pr(x = 1) = p$ . Both the  $\mu_i$ 's and  $\sigma_i$ 's are unknown.

**Problem 1** If we model this distribution as  $y = \phi_0 + \phi_1 x$  and choose the  $L_2$  loss function, what is the loss function  $L[]$  as parameterized by  $\phi_0$ ,  $\phi_1$ , and  $p$ ?

## Problem 2

What are the gradients of the loss loss function  $L[]$  with respect to  $\phi_0$  and  $\phi_1$ ? Assume the data set is large enough that the analytical solution is accurate (i.e. ignore statistical variations not explicitly modeled).

## Problem 3

If you train this model with gradient descent to the global minimum loss, what will the final values of parameters  $\phi_0$  and  $\phi_1$  be? Write your answer in terms of the  $\mu_i$ 's and  $\sigma_i$ 's and  $p$ . Again, assume the data set is large enough that the analytical solution is accurate.

## Problem 4

Are there any values of  $\phi_0$  and  $\phi_1$  where gradient descent could get stuck? That is, are there any parameter choices where the gradients become zero, but they are not at the global minimum?