# STAT 542: Homework 2

Due: Feb. 24 midnight on Canvas

Please make sure that your solutions are readable and the file size is reasonable. Typing the answers is highly encouraged.

#### Problem 1.

[2pts] Complete Exercise 4.2 in ESL to show the equivalence of LDA and linear regression in a certain setting. Our Notes LDA.pdf on Canvas already provided most of the ingredients of the calculations. Address the following question to complete the exercise:

• Let  $\hat{\delta}_1(x)$  and  $\hat{\delta}_2(x)$  be the discriminant functions in LDA (note that the condition in part (a) of Ex 4.2 is  $\hat{\delta}_2(x) > \hat{\delta}_1(x)$ ). Let  $\hat{f}(x)$  be the linear regression function in part (e). Show that when  $N_1 = N_2$  we have

$$\hat{f}(x) = \lambda(\hat{\delta}_2(x) - \hat{\delta}_1(x)) \tag{1}$$

for some scalar  $\lambda > 0$  depending on the training data (but not on the new input x). What is  $\lambda$ ?

### Problem 2.

[2pts] Run the iris dataset example for 1da documentation:

https://www.rdocumentation.org/packages/MASS/versions/7.3-58.2/topics/lda After training the model, write codes to use the test set to compute the confusion matrix and then the values of sensitivity and specificity (see s6\_logistic.pdf for definitions). Please include a screenshot to show your codes and results after running the code.

### Problem 3.

[3pts] Repeat Problem 2 using logistic regression instead of LDA. Compare the errors (sensitivity and specificity) with the LDA method. Also compare the performances with and without the intercept.

## Problem 4.

[1 bonus point] In our slides page "Advantages of LDA over logistic regression", it was mentioned that logistic regression may be unstable in certain settings. Consider the following example: In binary logistic regression, suppose that p > n, X has full row rank (rank=n), and there is no intercept. Show that the likelihood function is not maximized by any finite  $\beta$ ; in fact we can let  $\beta$  diverge while the likelihood tends to 1.

Hint: pick a  $\beta$  so that there is no training error in classification, and then set  $\beta \leftarrow t\beta$  with the scalar  $t \to \infty$ .