

# ST451 Bayesian Machine Learning

## Week 4

### Exercises

1. Consider the case of Linear Discriminant Analysis with a scalar  $x$  with data  $(y_i, x_i)_{i=1}^n$ , where  $y_i$ 's are binary random variables and  $x_i$ 's continuous. Assume that  $x_i \sim N(\mu_0, \sigma^2)$  in category  $c_0$  and that  $x_i \sim N(\mu_1, \sigma^2)$  in category  $c_1$  and that they are independent. Further assume that each  $y_i$  is a Bernoulli random variable with probability of success  $p(y \in c_1|x)$  and that the  $y_i$ 's are independent. Finally, the prior probability  $\pi(y \in c_1) = \pi$ . Write down the likelihood function and provide the maximum likelihood estimators for  $\pi, \mu_0, \mu_1$  and  $\sigma^2$ .
2. Let  $y = (y_1, \dots, y_n)$  be a r.s. from a  $\text{Poisson}(\lambda)$  and assign the an improper prior to  $\lambda$  such that  $\pi(\lambda) \propto \lambda^{-1/2}$ . Find the Laplace approximation to the posterior based on the mode and the Hessian matrix of  $\pi^*(\lambda|y) = f(y|\lambda)\pi(\lambda)$ .
3. In the dataset used in the computer class compute the maximum likelihood estimates without using the relevant *sklearn* function but with the *numpy* library only.
4. The file 'CreditCardFraud.csv' contains a subsample from the following Kaggle competition <https://www.kaggle.com/mlg-ulb/creditcardfraud>. A smaller sample has been taken with more balanced data, as the issue of unbalanced data is not the subject of this week's material. The features 'Amount' and 'Time' were also removed, leaving the 28 features labeled V1-V28. Fit logistic regression models on the data using both the MLE and Bayesian approaches and compare their predictive performance.
5. The file 'Default.csv' contains the data from the first motivating example in the lecture slides. Fit logistic regression models on the data (both the MLE and Bayesian) as well as the Linear Discriminant Analysis model and compare their predictive performances.