## **ASSIGNMENT 5** — Advanced Techniques I

For all questions below, provide all programming code and plots in the report. Unless stated otherwise, assume  $\alpha=0.05$ 

## **Maximum Likelihood**

- 1. You have collected the following data on a group of individuals: [8.453532, 10.025041, 11.495339, 9.367600, 8.333229, 9.788753, 10.883344, 10.543059, 9.869095, 10.799819]. (10 marks).
  - a. Work through the math of calculating the Normal distribution MLE estimate for practice. No need to show this work.
  - b. What is the MLE assuming a Normal distribution. (2 mark)
  - c. Use a brute force method to find the maximum likelihood estimates. (4 mark).
    - . vary  $\mu$  from 5 to 15 and  $\sigma$  from 0.5 to 1.5, both in increments of 0.1.
  - d. Use an optimizer to find the maximum likelihood estimates. Start your initial guesses as  $\mu = 8$  and  $\sigma^2 = 0.5$  (4 mark).
    - . hint: https://stackoverflow.com/questions/40620277/r-how-can-the-nlm-function-be-used-for-optimization-with-multiple-variables

## **Bayesian Statistics**

- 2. Point Probability (5 marks):
  - a. As in the lecture example, let's say you are getting COVID tests and want to know the probability that you have caught it. Initial conditions: p(+covid) = 0.5, p(+test|+covid) = 0.6, p(+test|-covid) = 0.35. You observe 3 positive tests in row. List the probability that you have covid for each test (2 mark).
  - b. Change your initial prior to 0.01. List the probability that you have covid for each test. What happens and why (0.5 mark)?
  - c. Change your initial prior to 0.0. List the probability that you have covid for each test. What happens and why (0.5 mark)?
  - d. Once again, lets assume p(+covid) = 0.5, p(+test|+covid) = 0.6, p(+test|-covid) = 0.35. This time, estimate the probability p(+covid|-test) given you observe 3 negative tests in a row. Tip, p(-test|+covid) is the complement of p(+test|+covid), and p(-test|-covid) is the complement of p(+test|-covid) (2 mark).

- 3. Continuous Probability (5 marks):
  - a. The conjugate prior of a Normal distribution is a Normal distribution. This can be summarized as follows:

Prior =  $\mathcal{N}(\mu_0, \sigma_0^2)$ ; Likelihood =  $\mathcal{N}(\mu_1, \sigma_1^2)$ ; Posterior =  $\mathcal{N}(\mu_2, \sigma_2^2)$ 

$$\mu_2 = \frac{1}{\frac{1}{\sigma_0^2} + \frac{n}{\sigma_1^2}} \cdot \left(\frac{\mu_0}{\sigma_0^2} + \frac{\sum_{i=1}^n x_i}{\sigma_1^2}\right)$$

$$\sigma_2^2 = \left(\frac{1}{\sigma_0^2} + \frac{n}{\sigma_1^2}\right)^{-1}$$

These formula allow us to get an estimate  $\mu_2$  (given  $\sigma_0$ ,  $\sigma_1$ , and  $x_i$  are known). Use the same data from Question 1 above. Assume  $\mu_0 = 11.0$ ,  $\sigma_0 = 1.0$ , and  $\sigma_1 = 1.0$ . Calculate the posterior after observing all ten individuals (2 points).

- b. Plot the prior and posterior. Make sure you consider  $\sigma_2$  vs.  $\sigma_2^2$  when plotting (1 marks).
- c. Why do our estimates of  $\mu$  differ between Question 1 and 3 (1 mark)?
- d. What does the posterior represent and how does this differ MLE (1 mark)?

