## QUEEN MARY, UNIVERSITY OF LONDON

## MTH6102: Bayesian Statistical Methods

## Exercise sheet 10

## 2023-2024

This exercise sheet 10 is assessed and counts for 4% of the module total. The deadline for submission is Monday the 11th December at 11am.

Submit the R code used as an R script file (with extension .R). But you need to write the answers in a separate file. This can be a Word document, pdf or a clearly legible image of hand-written work. So you need to submit two files.

- 1. **50 marks.** Let the observed data be y = (6, 4, 9, 2, 0, 3), a random sample from the Poisson distribution with mean  $\lambda$ , where  $\lambda > 0$  is unknown. Suppose that we assume a Gamma(1,1) prior distribution for  $\lambda$ . The posterior density,  $p(\lambda \mid y)$ , for  $\lambda$  is Gamma(1+S,1+n), where  $S = \sum_{i=1}^{6} y_i$  and n = 6. Suppose that you want to construct a symmetric Metropolis-Hastings on the log-scale to generate a sample from this posterior distribution by using a normal proposal distribution with standard deviation b = 0.2.
  - (a) Write down the steps in this symmetric Metropolis-Hastings (on the log-scale) to simulate realisations from the posterior density  $p(\lambda \mid y)$ .
  - (b) Implement the algorithm in R and plot the observations as a function of the iterations. Use M=5000 for the number of iterations.
  - (c) To assess the accuracy compare the empirical distribution of the sample with the exact posterior density, Gamma(1 + S, 1 + n).
  - (d) Rerun the algorithm in R using a smaller b = 0.01 and a larger b = 20. What are the effects on the behaviour of the algorithm of making b smaller? What are the effects of making it larger?
  - (e) Add code to count how many times the proposed value for  $\lambda$  was accepted. Rerun the algorithm using values of b = 0.01, b = 0.2 and b = 20, and each time calculate the proportion of steps that were accepted. Then plot this acceptance probability against b. Examine how the acceptance probability for this algorithm depends b.
- 2. **50 marks.** Let  $y_1, \ldots, y_n$  be a sample from a Poisson distribution with mean  $\lambda$ , where  $\lambda$  is given a Gamma $(\alpha, \beta)$  prior distribution.
  - (a) It is observed that  $y_1 = y_2 = \cdots = y_n = 0$ , and we take  $\alpha = 1, \beta = 1$ .
    - i. What is the posterior distribution for  $\lambda$ ?
    - ii. What is the posterior mean?
    - iii. What is the posterior median and an equal tail 95% credible interval for  $\lambda$  (without using R)?

(b) Show that if a new data-point x is generated from the same Poisson distribution, the posterior predictive probability that x=0 is

$$p(x = 0 \mid y) = \frac{n+1}{n+2}.$$

- (c) Now suppose that we have general  $y_1, \ldots, y_n$ ,  $\alpha$  and  $\beta$ ; and that again x is a new data-point from the same Poisson distribution.
  - i. Find the mean and variance of x.
  - ii. Derive the full posterior predictive distribution for x.
- (d) Use R to check as many of these results as you can.