

### MA676: Bayesian Methods – Additional Exercises 3

1. Find the Jeffrey's prior for  $\theta$  where the observations are drawn from Poisson( $\theta$ ).
2. Suppose that  $X_1, \dots, X_n$  are independent observations from  $N(\theta, \sigma^2)$  where  $\sigma^2$  is a known constant. Further assume that  $\theta$  follows  $N(\mu, \tau^2)$  for known values of  $\mu$  and  $\tau^2$ . Obtain the posterior predictive distribution of  $X_{n+1}$  given  $x_1, \dots, x_n$ .
3. A random variable  $X$  has a gamma distribution  $\text{gamma}(m, \beta)$  with pdf

$$f(x) = \frac{\beta^m}{\Gamma(m)} x^{m-1} \exp(-\beta x) \quad x > 0.$$

Show that if  $Y = 1/X$  then  $Y$  has p.d.f.

$$f(y) = \frac{\beta^m}{\Gamma(m)} \frac{1}{y^{m+1}} \exp(-\beta/y) \quad y > 0.$$

This is the inverse gamma distribution.

A particular measuring device has normally distributed error with mean zero and unknown variance  $\sigma^2$ . In an experiment to estimate  $\sigma^2$ ,  $n$  independent evaluations of this error are obtained.

If the prior distribution for  $\sigma^2$  is inverse gamma with parameters  $\beta$  and  $m$ , show that the posterior distribution is also inverse gamma, with parameters  $\beta^*$  and  $m^*$ , and derive expressions for  $\beta^*$  and  $m^*$ . Show that the predictive distribution for the error,  $Z$ , of a further observation made by this device has p.d.f.

$$f(z) \propto \left(1 + \frac{z^2}{2\beta^*}\right)^{-m^* - \frac{1}{2}} \quad z \in \mathbb{R}.$$

4. Assume  $Y_1, Y_2, \dots, Y_n$  are independent observations which have the distribution  $Y_i \sim N(\beta x_i, \sigma^2)$ ,  $i = 1, 2, \dots, n$ , where the  $x_i$ s and  $\sigma^2$  are known constants, and  $\beta$  is an unknown parameter, which has a normal prior distribution with mean  $\beta_0$  and variance  $\sigma_0^2$ , where  $\beta_0$  and  $\sigma_0^2$  are known constants.
  - (i) Derive the posterior distribution of  $\beta$ .
  - (ii) Show that the mean of the posterior distribution is a weighted average of the prior mean  $\beta_0$ , and the maximum likelihood estimator of  $\beta$ .
  - (iii) Find the limit of the posterior distribution as  $\sigma_0^2 \rightarrow \infty$ , and discuss the result.
  - (iv) How would you predict a future observation from the population  $N(\beta x_{n+1}, \sigma^2)$ , where  $x_{n+1}$  is known?

5. Let  $Y_1, Y_2, \dots, Y_n$  be a sequence of independent, identically distributed random variables with probability density function

$$f(y|\lambda) = \begin{cases} \lambda e^{-\lambda y}, & y > 0 \\ 0, & \text{otherwise} \end{cases}$$

where  $\lambda$  is an unknown, positive parameter with a gamma( $m, \beta$ ) prior distribution (see above).

- (i) Show that the posterior distribution of  $\lambda$  given  $Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n$  is gamma( $n + m, \beta + t$ ) where  $t = \sum_{i=1}^n y_i$ .
- (ii) Show that the (predictive) density of  $Y_{n+1}$  given the  $n$  observations  $Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n$  is

$$\pi(y_{n+1}|y_1, \dots, y_n) = \frac{(n+m)(\beta+t)^{n+m}}{(y_{n+1} + \beta + t)^{n+m+1}}.$$

- (iii) Find the joint (predictive) density of  $Y_{n+1}$  and  $Y_{n+2}$  given  $Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n$ .