

MA676: Bayesian Methods – Exercise Sheet 1

1. Suppose that X_1, \dots, X_n is a random sample from the distribution with pdf

$$f(x|\theta) = \begin{cases} \theta x^{\theta-1} & \text{if } 0 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Suppose also that the value of the parameter θ is unknown ($\theta > 0$) and that the prior distribution of θ is a gamma distribution with parameters α and β ($\alpha > 0$ and $\beta > 0$). Determine the posterior distribution of θ and hence obtain the Bayes estimator of θ under a squared error loss function.

2. A Bayes estimator is required for θ under the loss function

$$L(a, \theta) = e^{c(a-\theta)} - c(a - \theta) - 1,$$

where c is a positive constant. As the constant c varies, the loss function varies from very asymmetric to almost symmetric. This is called the LINEX (LINear-EXponential) loss.

By minimising the expected loss, show that the Bayes estimator is

$$\hat{\theta} = \frac{-1}{c} \log E \left(e^{-c\theta} | \mathbf{x} \right),$$

where the expectation is under the posterior distribution $\pi(\theta|\mathbf{x})$.

Suppose that X_1, \dots, X_n is a random sample from the normal distribution $N(\theta, \sigma^2)$ where σ^2 is known. A priori θ follows the normal distribution with mean μ and variance τ^2 where both μ and τ^2 are known. Find the Bayes estimators of θ under the above loss function and under the squared error loss. Compare the two estimators. **Hint:** Recall the definition and the expression for the moment generating function of a normal distribution.

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 σ^2 . In an experiment to estimate σ^2 , n independent evaluations of this error are obtained.

If the prior distribution for σ^2 is inverse gamma with parameters β and m , show that the posterior distribution is also inverse gamma, with parameters β^* and m^* , and derive expressions for β^* 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 σ^2 are known constants, and β is an unknown parameter, which has a normal prior distribution with mean β_0 and variance σ_0^2 , where β_0 and σ_0^2 are known constants.
- (i) Derive the posterior distribution of β .
 - (ii) Show that the mean of the posterior distribution is a weighted average of the prior mean β_0 , and the maximum likelihood estimator of β .
 - (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 λ is an unknown, positive parameter with a gamma(m, β) prior distribution (see above).

- (i) Show that the posterior distribution of λ 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 $Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n$.