

Statistical Modeling 2

Exercise 2

February 3, 2017

1 A simple Gaussian location model

A

The joint prior over the mean parameter θ and precision parameter ω is:

$$p(\theta, \omega) \approx \omega^{(d+1)/2-1} \exp \left\{ -\omega \frac{\kappa(\theta - \mu)^2}{2} \right\} \exp \left\{ -\omega \frac{\eta}{2} \right\}$$

To get the marginal prior, we integrate out the parameter ω :

$$\begin{aligned} p(\theta) &\propto \int_0^\infty \omega^{(d+1)/2-1} \exp \left\{ -\omega \frac{\kappa(\theta - \mu)^2 + \eta}{2} \right\} \\ &\propto \left(\frac{\kappa(\theta - \mu)^2 + \eta}{2} \right)^{-(d+1)/2} \\ &= \left(\frac{\eta}{2} + \frac{\kappa(\theta - \mu)^2}{2} \right)^{-(d+1)/2} \\ &= \left(1 + \frac{\kappa(\theta - \mu)^2}{\eta} \right)^{-(d+1)/2} \left(\frac{\eta}{2} \right)^{-(d+1)/2} \\ &\propto \left(1 + \frac{\kappa(\theta - \mu)^2}{\eta} \right)^{-(d+1)/2} \\ &= \left(1 + \frac{1}{d} \frac{\kappa(\theta - \mu)^2}{\eta} \right)^{-(d+1)/2} \end{aligned}$$

Let $\nu = d, m = \mu$ and $s = \sqrt{\eta/\kappa}$, we have a Student t distribution with ν degrees of freedom and scale s :

$$p(\theta) \propto \left(1 + \frac{1}{\nu} \frac{(\theta - m)^2}{s^2} \right)^{-(\nu+1)/2}$$

B

The sampling model is:

$$(y_i \mid \theta, \omega) \sim N(\theta, 1/\omega)$$

where y_1, \dots, y_n are the datapoints, θ is the mean and ω is the precision. We have that the likelihood for all the datapoints can be written as:

$$\begin{aligned} p(\mathbf{y} \mid \theta, \omega) &\propto \prod_{i=1}^n \omega^{1/2} \exp \left\{ -\frac{1}{2} \omega (y_i - \theta)^2 \right\} \\ &= \omega^{n/2} \exp \left\{ -\frac{1}{2} \omega \sum_{i=1}^n (y_i - \theta)^2 \right\} \\ &= \omega^{n/2} \exp \left\{ -\frac{1}{2} \omega \left(\sum_{i=1}^n y_i^2 + \sum_{i=1}^n \theta^2 - 2 \sum_{i=1}^n y_i \theta \right) \right\} \\ &= \omega^{n/2} \exp \left\{ -\frac{1}{2} \omega \left(\sum_{i=1}^n y_i^2 + n\theta^2 - 2n\bar{y}\theta + n\bar{y}^2 - n\bar{y}^2 \right) \right\} \end{aligned}$$

where $\bar{y} = (\sum_{i=1}^n y_i) / n$. Let $S_y = \sum_{i=1}^n (y_i - \bar{y})^2$, we have:

$$\begin{aligned} S_y &= \sum_{i=1}^n y_i^2 + n\bar{y}^2 - 2 \sum_{i=1}^n y_i \bar{y} \\ &= \sum_{i=1}^n y_i^2 - n\bar{y}^2 \end{aligned}$$

Therefore, the likelihood is:

$$\begin{aligned} p(\mathbf{y} \mid \theta, \omega) &= \omega^{n/2} \exp \left\{ -\frac{1}{2} \omega [S_y + n(\theta^2 - 2\bar{y}\theta + \bar{y}^2)] \right\} \\ &= \omega^{n/2} \exp \left\{ -\frac{1}{2} \omega [S_y + n(\bar{y} - \theta)^2] \right\} \end{aligned}$$

The posterior is proportional to the product of the likelihood and the prior:

$$p(\theta, \omega \mid \mathbf{y}) \propto \omega^{(d+1)/2-1} \exp \left\{ -\omega \frac{\kappa(\theta - \mu)^2}{2} \right\} \exp \left\{ -\omega \frac{\eta}{2} \right\}$$