Consolidation Activities

2. Exercise 2

Let $x=(x_1,\ldots,x_n)$ be a random sample from a $N(\theta,\sigma^2)$ distribution with σ^2 known.

a. Show that the likelihood is proportional to

$$f(x| heta) \propto \expigg(-rac{n(ar x- heta)^2+(n-1)S^2}{2\sigma^2}igg).$$

where \bar{x} is the sample mean and S^2 is the sample variance

$$S^2 = rac{1}{n-1} \sum_{i=1}^n (x_i - ar{x})^2.$$

Hence the likelihood simplifies to

$$f(x| heta) \propto \exp\left(-rac{(heta-ar{x})^2}{2rac{\sigma^2}{n}}
ight)$$

▼ Solution

The joint density of the sample x is

$$egin{align} f(x| heta,\sigma^2) &= f(x_1,\ldots,x_n| heta,\sigma^2) = \prod_{i=1}^n (2\pi\sigma^2)^{-1/2} \expigg(-rac{(x_i- heta)^2}{2\sigma^2}igg) \ &= (2\pi\sigma^2)^{-n/2} \expigg(-rac{\sum_{i=1}^n (x_i- heta)^2}{2\sigma^2}igg) \end{split}$$

Hence, it suffices to show that

$$\sum_{i=1}^n (x_i - heta)^2 = n(ar{x} - heta)^2 + \sum_{i=1}^n (x_i - ar{x})^2.$$

Note that

$$\sum_{i=1}^n (x_i - heta)^2 = \sum_{i=1}^n (x_i^2 - 2 heta x_i + heta^2) = \sum_{i=1}^n x_i^2 - 2 heta nar{x} + n heta^2 \ \sum_{i=1}^n (x_i - ar{x})^2 = \sum_{i=1}^n (x_i^2 - 2ar{x}x_i + ar{x}^2) = \sum_{i=1}^n x_i^2 - 2nar{x}^2 + nar{x}^2 = \sum_{i=1}^n x_i^2 - nar{x}^2$$

Subtracting the second of these equations from the first yields

$$\sum_{i=1}^n (x_i - heta)^2 - \sum_{i=1}^n (x_i - ar{x})^2 = nar{x}^2 - 2 heta nar{x} + n heta^2 = n(ar{x} - heta)^2$$

Since σ^2 is known, we are interested in $f(x|\theta)$ which is proportional to

$$f(x|\theta) = (2\pi\sigma^2)^{-n/2} \exp\left(-\frac{n(\bar{x}-\theta)^2 + (n-1)S^2}{2\sigma^2}\right)$$

$$\propto \exp\left(-\frac{n(\bar{x}-\theta)^2}{2\sigma^2}\right) \exp\left(-\frac{(n-1)S^2}{2\sigma^2}\right)$$

$$\propto \exp\left(-\frac{(\theta-\bar{x})^2}{2\frac{\sigma^2}{n}}\right)$$

b. Set the prior for θ to be $N(\mu, \tau^2)$ and derive its posterior distribution. (You can use the above result)

▼ Solution

The prior for θ is set to be $N(\mu, \tau^2)$. Hence we can write:

$$\pi(\theta) \propto \exp\left(-\frac{(\theta-\mu)^2}{2\tau^2}\right).$$

Using the result of part (a), the posterior is then proportional to

$$\begin{split} \pi(\theta|x) &\propto f(x|\theta)\pi(\theta) \propto \exp\left(-\frac{(\theta-\bar{x})^2}{2\frac{\sigma^2}{n}}\right) \exp\left(-\frac{(\theta-\mu)^2}{2\tau^2}\right) \\ &= \exp\left(-\frac{\theta^2 + 2\theta\bar{x}}{2\frac{\sigma^2}{n}} - \frac{\theta^2 - 2\theta\mu}{2\tau^2}\right) \exp\left(-\frac{\bar{x}^2}{2\frac{\sigma^2}{n}}\right) \exp\left(-\frac{\mu^2}{2\tau^2}\right) \\ &\propto \exp\left(-\frac{\theta^2 + 2\theta\bar{x}}{2\frac{\sigma^2}{n}} - \frac{\theta^2 - 2\theta\mu}{2\tau^2}\right) \\ &= \exp\left(-\frac{\tau^2\theta^2 - 2\theta\bar{x}\tau^2 + \frac{\sigma^2}{n}\theta^2 - 2\theta\mu\frac{\sigma^2}{n}}{2\frac{\sigma^2}{n}\tau^2}\right) \\ &= \exp\left(-\frac{(\frac{\sigma^2}{n} + \tau^2)\theta^2 - 2\theta(\bar{x}\tau^2 + \mu\frac{\sigma^2}{n})}{2\frac{\sigma^2}{n}\tau^2}\right) \\ &= \exp\left(-\frac{\theta^2 - 2\theta\frac{\bar{x}\tau^2 + \mu\frac{\sigma^2}{n}}{(\frac{\sigma^2}{n} + \tau^2)}}{2\frac{\sigma^2}{(\frac{\sigma^2}{n} + \tau^2)}}\right) \stackrel{\mathcal{D}}{=} \operatorname{N}\left(\frac{\frac{\sigma^2}{n}\mu + \tau^2\bar{x}}{\tau^2 + \frac{\sigma^2}{n}}, \frac{\tau^2\frac{\sigma^2}{n}}{\tau^2 + \frac{\sigma^2}{n}}\right) \end{split}$$