

## MATH 557 - MID-TERM 2008 - SOLUTIONS

1. (a) Note first that by standard expansion into a quartic polynomial

$$\left(\frac{x-\theta}{\sigma}\right)^4 = w_0(\theta, \sigma) + \sum_{j=1}^k w_j(\theta, \sigma)x^j = w_0(\theta, \sigma) + \sum_{j=1}^k w_j(\theta, \sigma)t_j(x)$$

say, where  $w_j(\theta, \sigma)$  are constant functions of  $\theta$  and  $\sigma$ . Thus

$$f_{X|\theta, \sigma}(x|\theta, \sigma) = h(x)c(\theta, \sigma) \exp\left\{\sum_{j=1}^k w_j(\theta, \sigma)t_j(x)\right\}$$

where

$$h(x) = 1 \quad c(\theta, \sigma) = \exp\{w_0(\theta, \sigma) - \kappa(\theta, \sigma)\} \quad t_j(x) = x^j, \quad j = 1, \dots, 4.$$

and hence the distribution is an Exponential Family distribution.

6 Marks

- (b) By inspection, and using the Neyman factorization theorem in this Exponential family setting, we have

$$\underline{T}(\underline{X}) = (T_1(\underline{X}), T_2(\underline{X}), T_3(\underline{X}), T_4(\underline{X}))^T \quad T_j(\underline{X}) = \sum_{i=1}^n t_j(X_i) = \sum_{i=1}^n X_i^j \quad j = 1, \dots, 4$$

is a sufficient statistic.

6 Marks

2. (a) We have

$$\begin{aligned} f_{\underline{X}|\theta}(\underline{x}|\theta) &= \left(\frac{1}{2\pi\theta^2}\right)^{n/2} \exp\left\{-\frac{1}{2\theta^2} \sum_{i=1}^n (x_i - \theta)^2\right\} \\ &= \left(\frac{1}{2\pi\theta^2}\right)^{n/2} \exp\left\{-\frac{1}{2\theta^2} \sum_{i=1}^n x_i^2 + \frac{1}{\theta} \sum_{i=1}^n x_i - \frac{n}{2}\right\} \\ &= \left(\frac{1}{2\pi\theta^2}\right)^{n/2} e^{-n/2} \exp\left\{-\frac{1}{2\theta^2} T_2(\underline{x}) + \frac{1}{\theta} T_1(\underline{x})\right\} \end{aligned}$$

so that  $\underline{T}(\underline{X}) = (T_1(\underline{X}), T_2(\underline{X}))^T$  where

$$T_1(\underline{X}) = \sum_{i=1}^n X_i \quad T_2(\underline{X}) = \sum_{i=1}^n X_i^2$$

is a sufficient statistic. Note that  $T_1(\underline{X})$  and  $T_2(\underline{X})$  are linearly independent, and that for two vectors  $\underline{x}, \underline{y}$

$$\frac{f_{\underline{X}|\theta}(\underline{x}|\theta)}{f_{\underline{X}|\theta}(\underline{y}|\theta)} = \exp\left\{-\frac{1}{2\theta^2}(T_2(\underline{x}) - T_2(\underline{y})) + \frac{1}{\theta}(T_1(\underline{x}) - T_1(\underline{y}))\right\}$$

does not depend on  $\theta$  iff  $T_1(\underline{x}) = T_1(\underline{y})$  and  $T_2(\underline{x}) = T_2(\underline{y})$ . Thus  $\underline{T}$  is minimal sufficient.

6 Marks

(b) For each  $i$ , we have from the formula sheet properties of Normal distributions that

$$E_{f_{X_i|\theta}}[X_i^2] = 2\theta^2$$

so that

$$E_{f_{T_2|\theta}}[T_2] = 2n\theta^2$$

Also from mgf results.

$$T_1(\underline{X}) = \sum_{i=1}^n X_i \sim \text{Normal}(n\theta, n\theta^2)$$

Thus if  $S(\underline{X}) = \{T_1(\underline{X})\}^2$ .

$$E_{f_{S|\theta}}[S] = n\theta^2 + (n\theta)^2 = n(n+1)\theta^2.$$

Therefore the function

$$g(t_1, t_2) = \frac{t_1^2}{n(n+1)} - \frac{t_2}{2n}$$

is such that  $E_{f_{T_1, T_2|\theta}}[g(T_1, T_2)] = 0$ , and hence  $\underline{T}(\underline{X})$  is not complete.

6 Marks

3. (a) In the Poisson model, the likelihood is

$$L(\lambda|\underline{x}) \propto \lambda^{\sum_{i=1}^n x_i} e^{-n\lambda}$$

so therefore the conjugate prior is  $\text{Gamma}(\alpha, \beta)$

$$\pi_\lambda(\lambda) \propto \lambda^{\alpha-1} e^{-\beta\lambda}$$

yielding posterior

$$\pi_{\lambda|\underline{x}}(\lambda|\underline{x}) \propto \lambda^{\alpha + \sum_{i=1}^n x_i - 1} e^{-(\beta+n)\lambda} \equiv \text{Gamma}\left(\alpha + \sum_{i=1}^n x_i, \beta + n\right).$$

From lectures, the estimate under squared-error loss is the posterior mean, that is, from the formula sheet

$$\hat{\lambda}_B(\underline{x}) = \frac{\alpha + \sum_{i=1}^n x_i}{\beta + n} = \frac{n}{\beta + n} \bar{x}_n + \frac{\beta}{\beta + n} \left(\frac{\alpha}{\beta}\right) = w_n \bar{x}_n + (1 - w_n)m$$

10 Marks

(b) If  $T(\underline{X}) = a\bar{X}_n + b$ , then

$$\begin{aligned} R_T(\lambda) &= \int_{\mathcal{X}} (a\bar{X}_n + b - \lambda)^2 f_{\underline{X}|\lambda}(\underline{x}|\lambda) d\underline{x} = \int_{\mathcal{X}} ((a\bar{X}_n - a\lambda) + (b + (a-1)\lambda))^2 f_{\underline{X}|\lambda}(\underline{x}|\lambda) d\underline{x} \\ &\geq \int_{\mathcal{X}} (a\bar{X}_n - a\lambda)^2 f_{\underline{X}|\lambda}(\underline{x}|\lambda) d\underline{x} \\ &> \int_{\mathcal{X}} (\bar{X}_n - \lambda)^2 f_{\underline{X}|\lambda}(\underline{x}|\lambda) d\underline{x} = R_{T_0}(\lambda) \end{aligned}$$

as  $a > 1$ , where

$$T_0(\underline{X}) = \bar{X}_n.$$

This follows by expanding the integrand in the second integral of line 1, noting that the integral of the cross term is zero, and that the integral of the  $(b + (a-1)\lambda)^2$  term is non-negative. Hence  $T$  is inadmissible.

6 Marks