

## Statistics 852 Fall 2011 Midterm Exam – Solutions

1. Since  $M_X(t) = \mathbb{E}(e^{tX})$  exists for  $t \in (-h, h)$  for some  $h > 0$  we know that

$$M'_X(0) = \mathbb{E}(X) \quad \text{and} \quad M''_X(0) = \mathbb{E}(X^2).$$

Moreover, it is always the case that  $M_X(0) = 1$ . From the chain rule, it follows that

$$\frac{d}{dt} \log M_X(t) = \frac{M'_X(t)}{M_X(t)}$$

and

$$\frac{d^2}{dt^2} \log M_X(t) = \frac{d}{dt} \left[ \frac{M'_X(t)}{M_X(t)} \right] = \frac{M''_X(t)M_X(t) - M'_X(t)M'_X(t)}{M_X(t)^2}.$$

Therefore,

$$\left. \frac{d}{dt} \log M_X(t) \right|_{t=0} = \frac{M'_X(0)}{M_X(0)} = M'_X(0) = \mathbb{E}(X)$$

and

$$\begin{aligned} \left. \frac{d^2}{dt^2} \log M_X(t) \right|_{t=0} &= \frac{M''_X(0)M_X(0) - M'_X(0)M'_X(0)}{M_X(0)^2} = M''_X(0) - [M'_X(0)]^2 = \mathbb{E}(X^2) - [\mathbb{E}(X)]^2 \\ &= \text{Var}(X) \end{aligned}$$

as required.

2. (a) If  $\alpha$  is known and  $\beta$  is unknown, then

$$f(x|\beta) = \beta\alpha^\beta x^{-\beta-1} I(x > \alpha) = x^{-1} I(x > \alpha) \cdot \beta\alpha^\beta \cdot e^{-\beta \log x}.$$

Hence, if we take

$$h(x) = x^{-1} I(x > \alpha), \quad c(\beta) = \beta\alpha^\beta, \quad w(\beta) = \beta, \quad t(x) = -\log x,$$

then we see that

$$f(x|\beta) = h(x)c(\beta)e^{w(\beta)t(x)}$$

proving that  $\{f(x|\beta) : \beta > 0\}$  does, in fact, form an exponential family.

2. (b) If  $\alpha$  is unknown and  $\beta$  is known, then

$$f(x|\alpha) = \beta\alpha^\beta x^{-\beta-1} I(x > \alpha) = \beta x^{-\beta-1} \cdot \alpha^\beta \cdot I(x > \alpha).$$

Since the support of the density, namely  $\{x : x > \alpha\}$ , depends on the parameter  $\alpha$ , we conclude that  $\{f(x|\alpha) : \alpha > 0\}$  does not form an exponential family.

**2. (c)** In order to show that  $\mathbb{E}(X^2)$  does not exist for  $0 < \beta \leq 2$ , we will consider the integral

$$\int_{-\infty}^{\infty} x^2 f(x|\alpha, \beta) dx = \int_{\alpha}^{\infty} x^2 \cdot \beta \alpha^{\beta} x^{-\beta-1} dx = \beta \alpha^{\beta} \int_{\alpha}^{\infty} x^{1-\beta} dx. \quad (*)$$

In order to evaluate the integral, there are three separate cases to treat. First, suppose that  $\beta = 1$  so that

$$(*) = \alpha \int_{\alpha}^{\infty} dx = \alpha \lim_{N \rightarrow \infty} \int_{\alpha}^N dx = \alpha \lim_{N \rightarrow \infty} (N - \alpha) = \infty.$$

Now suppose that  $\beta = 2$  so that

$$(*) = 2\alpha^2 \int_{\alpha}^{\infty} x^{-1} dx = 2\alpha^2 \lim_{N \rightarrow \infty} \int_{\alpha}^N x^{-1} dx = 2\alpha^2 \lim_{N \rightarrow \infty} (\log N - \log \alpha) = \infty.$$

Finally, suppose that  $\beta \in (0, 1) \cup (1, 2)$  so that

$$\begin{aligned} (*) &= \beta \alpha^{\beta} \int_{\alpha}^{\infty} x^{1-\beta} dx = \beta \alpha^{\beta} \lim_{N \rightarrow \infty} \int_{\alpha}^N x^{1-\beta} dx = \beta \alpha^{\beta} \lim_{N \rightarrow \infty} \left. \frac{x^{2-\beta}}{2-\beta} \right|_{x=\alpha}^{x=N} \\ &= \frac{\beta \alpha^{\beta}}{2-\beta} \lim_{N \rightarrow \infty} (N^{2-\beta} - \alpha^{2-\beta}) \\ &= \infty \end{aligned}$$

since  $\beta \in (0, 1) \cup (1, 2)$  implies  $2 - \beta > 0$ .

**3.** In order to show that  $T(X_1, X_2) = X_1 + X_2$  is not sufficient for  $\theta$ , we will show that for some choice of  $(x_1, x_2)$  and  $t$ , the ratio

$$\frac{P_{\theta}\{(X_1, X_2) = (x_1, x_2)\}}{P_{\theta}\{T(X_1, X_2) = t\}}$$

does depend on  $\theta$ . Consider  $t = 2$  so that

$$\begin{aligned} P_{\theta}\{T(X_1, X_2) = 2\} &= P_{\theta}\{X_1 + X_2 = 2\} \\ &= P_{\theta}\{(X_1, X_2) = (1, 1)\} + P_{\theta}\{(X_1, X_2) = (0, 2)\} + P_{\theta}\{(X_1, X_2) = (2, 0)\} \\ &= \theta e^{-\theta} \cdot \theta e^{-\theta} + e^{-\theta} \cdot (1 - e^{-\theta} - \theta e^{-\theta}) + (1 - e^{-\theta} - \theta e^{-\theta}) \cdot e^{-\theta} \\ &= 2e^{-\theta} + e^{-2\theta}(\theta^2 - 2\theta - 2). \end{aligned}$$

Consider  $(x_1, x_2) = (1, 1)$  so that

$$P_{\theta}\{(X_1, X_2) = (1, 1)\} = \theta e^{-\theta} \cdot \theta e^{-\theta} = \theta^2 e^{-2\theta}.$$

Therefore,

$$\frac{P_{\theta}\{(X_1, X_2) = (1, 1)\}}{P_{\theta}\{T(X_1, X_2) = 2\}} = \frac{\theta^2 e^{-2\theta}}{2e^{-\theta} + e^{-2\theta}(\theta^2 - 2\theta - 2)} = \frac{\theta^2}{2e^{\theta} + \theta^2 - 2\theta - 2}.$$

Since this ratio obviously depends on  $\theta$ , we conclude that  $T(X_1, X_2) = X_1 + X_2$  is not sufficient for  $\theta$ .