

Solutions 1
out of 20

MAT5190

M. ALVO

3.2 a) let $\theta = \#$ of defectives. Let $f(\theta, k) = P(\text{accepting the lot} | \theta)$

$$\text{Then } f(\theta, k) = \binom{N-\theta}{k} \binom{\theta}{0} / \binom{N}{k}.$$

(3) Note $\frac{f(\theta+1, k)}{f(\theta, k)} = \frac{N-\theta-k}{N-\theta} < 1$; hence $f \downarrow$ in θ for fixed k and \downarrow in k .

If $\theta \geq 6$, $f(\theta, k) \leq f(6, k)$ for $\theta \geq 6$.
We find k such that $f(6, k) \leq 0.1$ by trial and error
the smallest k is $k=32$

(3) b) Write $g(\theta, k) = \left[\binom{100-\theta}{k} + \binom{100-\theta}{k-1} \binom{\theta}{1} \right] / \left[\binom{100}{k} \right]$

As before we can show $g \downarrow \theta$

The smallest value of k for which $g(\theta, k) < 0.1$ is $k=51$.

3.21 The mgf is $E e^{tx} = \int_{-\infty}^{\infty} \frac{1}{\pi} \frac{e^{tx}}{1+x^2} dx$

(3) On $(0, \infty)$, $e^{tx} > x$ and hence $\int_0^{\infty} \frac{e^{tx}}{1+x^2} dx > \int_0^{\infty} \frac{x}{1+x^2} dx = \infty$

Hence the mgf does not exist as expected since the moments of the Cauchy do not exist.

3.32 $f(x|\eta) = h(x) c(\eta) \exp\left(\sum_{i=1}^k \eta_i t_i(x)\right)$. Since f is a density,

$$1 = \int c(\eta) h(x) \exp\left(\sum_{i=1}^k \eta_i t_i(x)\right) dx$$

Differentiate both sides with respect to η_j (this can be done for the exponential family) twice.

(3) ① $0 = \frac{\partial}{\partial \eta_j} \int c(\eta) h(x) \exp\left(\sum \eta_i t_i(x)\right) dx$

$$= \int [h(x)] \left[\frac{\partial c(\eta)}{\partial \eta_j} + t_j(x) \right] \exp\left(\sum \eta_i t_i(x)\right) dx$$

$$\Rightarrow 0 = \left[\int h(x) c(\eta) \exp\left(\sum \eta_i t_i(x)\right) dx \right] \frac{1}{c(\eta)} \frac{\partial c(\eta)}{\partial \eta_j}$$

$$\Rightarrow E t_j(x) = - \frac{\partial}{\partial \eta_j} \log c(\eta)$$

② $0 = \int [h(x)] \left[\frac{\partial^2 c(\eta)}{\partial \eta_j^2} \right] \exp\left(\sum \eta_i t_i(x)\right) dx$

$$+ \int [h(x)] \left[\frac{\partial c(\eta)}{\partial \eta_j} + t_j(x) \right] t_j(x) \exp\left(\sum \eta_i t_i(x)\right) dx$$

Noting that $\frac{\partial^2 \ln c(\eta)}{\partial \eta_j^2} = - \frac{1}{c^2} \frac{\partial c(\eta)}{\partial \eta_j} + \frac{1}{c} \frac{\partial^2 c(\eta)}{\partial \eta_j^2}$ and

$$\text{Var } t_j(x) = E t_j^2(x) - [E t_j(x)]^2 \text{ yields the result}$$

b) Let $\sigma_1 > \sigma_2$, $X_1 \sim f(x/\sigma_1)$, $X_2 \sim f(x/\sigma_2)$. Let $F(x)$ be the cdf corresponding to $f(x)$, and $Z \sim f(x)$. Then

$$F(x/\sigma_1) = P(X_1 \leq x) = P(\sigma_1 Z \leq x) = F(x/\sigma_1) \leq F(x/\sigma_2) = P(Z \leq x/\sigma_2)$$

$$(3) \quad = P(X_2 \leq x) = F(x/\sigma_2)$$

To get strict inequality for some x , let $(a, b]$ be an interval such that $a > 0$, $b/a = \sigma_1/\sigma_2$ and $P(a < Z \leq b) = F(b) - F(a) > 0$. Let

$$x = a\sigma_1. \text{ Then } F(x/\sigma_1) = F\left(\frac{x}{\sigma_1}\right) = F\left(\frac{a\sigma_1}{\sigma_1}\right) = F(a) < F(b) \\ = F\left(\frac{a\sigma_1}{\sigma_2}\right) = F\left(\frac{x}{\sigma_2}\right) = F(x/\sigma_2)$$

5.3 $Y_i \sim \text{Bernoulli}$ with $p_i = P(X_i \geq \mu) = 1 - F(\mu)$. Since the

Y_i are iid,

$$(2) \quad \sum_1^n Y_i \sim \text{binomial}(n, p) \text{ where } p = 1 - F(\mu)$$

5.6 a) For $Z = X - Y$ set $W = X$. Then $J = \begin{vmatrix} \frac{\partial Z}{\partial X} & \frac{\partial Z}{\partial Y} \\ \frac{\partial W}{\partial X} & \frac{\partial W}{\partial Y} \end{vmatrix} = 1$

(3) and $f_{Z,W}(z,w) = f_X(w) \cdot f_Y(w-z) \cdot 1$

$$\therefore f_Z(z) = \int_{-\infty}^{\infty} f_X(w) f_Y(w-z) dw$$

b) For $z = XY$, set $w = X$. Similarly we find

$$f_{z,w}(z,w) = f_X(w) f_Y(z/w) \left| \frac{1}{w} \right| \text{ and}$$

$$f_z(z) = \int_{-\infty}^{\infty} f_{z,w}(z,w) dw$$

c) For $z = X/Y$, set $w = X$. Similarly we find

$$f_{z,w}(z,w) = f_X(w) f_Y(w/z) \left| \frac{w}{z^2} \right|$$

$$f_z(z) = \int_{-\infty}^{\infty} f_X(w) f_Y(w/z) \left| \frac{w}{z^2} \right| dw$$