## Math 362, Problem set 9

## Due 4/20/10

1. (7.2.2) Prove that the sum of the observations of a random sample of size n from a Poisson distribution of having parameter  $\theta$ ,  $0 < \theta < \infty$ , is a sufficient statistic for  $\theta$ .

Answer:

$$f_{X_1,\dots,X_n}(x_1,\dots,x_n;\theta) = \left(e^{-\theta n}\theta^{\sum x_i}\right) \left(\frac{1}{\prod(x_i)!}\right)$$

and so  $\sum X_i$  is sufficient by the factorization theorem.

2. (7.2.6) Let  $X_1, \ldots, X_n$  be a random sample of size n from a beta distribution with parameters  $\alpha = \theta$  and  $\beta = 2$ . Show that the product  $X_1 X_2 \cdots X_n$  is a sufficient statistic for  $\theta$ .

Answer:

$$f_{X_1,\dots,X_n}(x_1,\dots,x_n;\theta) = \left(\left(\frac{\Gamma(\theta+2)}{\Gamma(\theta)\Gamma(2)}\right)^n \left(\prod(x_i)\right)^{\theta-1}\right) \prod(1-x_i)$$

and by the factorization theorem  $\prod X_i$  is a sufficient statistic for  $\theta$ .

3. (7.2.8) What is the sufficient statistic for  $\theta$  if the sample arises from a beta distribution in which  $\alpha = \beta = \theta > 0$ .

Answer: Here,

$$f_{X_1,\dots,X_n}(x_1,\dots,x_n;\theta) = \left(\left(\frac{\Gamma(2\theta)}{\Gamma(\theta)^2}\right)^n \left(\prod(x_i)(1-x_i)\right)^{\theta-1}\right) \cdot 1$$

and so by the factorization theorem  $\prod X_i(1 - X_i)$  is a sufficient statistic for  $\theta$ .

4. (7.3.3) If  $X_1, X_2$  is a random sample of size 2 from a distribution having pdf  $f(x; \theta) = (1/\theta)e^{-x/\theta}, 0 < x < infty, 0 < \theta < \infty$ , zero elsewhere, find the joint pdf of the sufficient statistic  $Y_1 = X_1 + X_2$  and  $Y_2 = X_2$ . Show that

 $Y_2$  is an unbiased estimator of  $\theta$  with variance  $\theta^2$ . Find  $\mathbb{E}[Y_2|y_1] = \varphi(y_1)$ and the variance of  $\varphi(Y_1)$ .

We have  $X_1 = Y_1 - Y_2$  and  $X_2 = Y_2$ , so |J| = 1 and we have

$$f_{Y_1, Y_2}(y_1, y_2; \theta) = (1/\theta)^2 e^{-y_1/\theta} \quad 0 < y_2 < y_1 < \infty$$

We have that

$$\mathbb{E}[Y_2] = \theta$$
  $\operatorname{Var}(Y_2) = \theta^2$ 

since  $Y_2 \sim \Gamma(1, \theta)$ .

We have

$$f_{Y_1}(y_1) = \int_{y_2=0}^{y_1} (1/\theta)^2 e^{-y_1/\theta} dy_2 = (1/\theta)^2 y_1 e^{-y_1/\theta} \quad 0 < y_1 < \infty$$

$$f_{Y_2|Y_1} = \frac{1}{y_1} \quad 0 < y_2 < y_1.$$

(That is conditioned on  $Y_1, Y_2$  is uniform.)

Therefore

$$\mathbb{E}[Y_2|y_1] = \frac{y_1}{2} = \varphi(y_1).$$

and

$$\operatorname{Var}(\varphi(Y_1)) = \operatorname{Var}((X_1 + X_2)/2) = \theta^2/2.$$

- 5. (7.3.4) Let  $f(x, y) = (2/\theta^2)e^{-(x+y)/\theta}$ ,  $0 < x < y < \infty$ , zero elsewhere, be the joint pdf of the random variables X and Y.
  - Show that the mean and variance of Y are respectively  $3\theta/2$  and  $5\theta^2/4$ .
  - Show that  $\mathbb{E}[Y|x] = x + \theta$ . In accordance with the theory we built up last semester, the expected value of  $X + \theta$  is that of Y, namely,  $3\theta/2$ , and the variance of  $X + \theta$  is less than that of Y. Show that the variance of  $X + \theta$  is in fact  $\theta^2/4$ .

Answer: We have

$$f_Y(y) = \frac{2}{\theta} (e^{-y/\theta} - e^{-2y/\theta}) \quad 0 < y < \infty.$$

$$\mathbb{E}[Y] = \int_0^\infty \frac{2}{\theta} y(e^{-y/\theta} - e^{-2y/\theta})$$

We use the fact that this is twice the expectation of a  $\Gamma(1,\theta)$  minus the expectation of a  $\Gamma(1,\theta/2)$  to get that this is

$$\mathbb{E}[Y] = 2\theta - \theta/2 = \frac{3\theta}{2}.$$

Also

$$\mathbb{E}[Y^2] = \int_0^\infty \frac{2}{\theta} y^2 (e^{-y/\theta} - e^{-2y/\theta})$$

This is twice the expectation of the square of a  $\Gamma(1,\theta)$  R.V. (which is  $2\theta^2$ ) minus the expectation of the square of a  $\Gamma(1,\theta/2)$  (which is  $\theta^2/2$ . In total we have

$$\mathbb{E}[Y^2] - \mathbb{E}[Y]^2 = (3.5)\theta^2 - 9/4\theta^2 = 5\theta^2/4.$$

$$\mathbb{E}[Y^2] = \int_{-\infty}^{\infty} (2/\theta^2) e^{-(x+y)/\theta} = 2e^{-2x/\theta} \text{ for } 0 < x$$

We have  $f_X(x) = \int_{y=x}^{\infty} (2/\theta^2) e^{-(x+y)/\theta} = \frac{2}{\theta} e^{-2x/\theta}$  for  $0 < x < \infty$ .

$$\mathbb{E}[Y|x] = \int_x^\infty y e^{(x-y)/theta} = x + \theta.$$

Since  $X \sim \Gamma(1, \theta/2)$ , we have that  $\operatorname{Var}(X + \theta) = \operatorname{Var}(X) = \theta^2/4$  as desired.

- 6. (7.4.6) Let a random sample of size n be taken from a distribution of the discrete type with pmf  $f(x; \theta), x = 1, 2, ..., \theta$ , zero elsewhere, where  $\theta$  is an unknown positive integer.
  - Show that the largest observation, say Y, of the sample is a complete sufficient statistic for  $\theta$ .
  - Prove that

$$[Y^{n+1} - (Y-1)^{n+1}]/[Y^n - (Y-1)^n]$$

is the unique MVUE for  $\theta$ .

Answer:

We have that  $f(x;\theta) = \frac{1}{\theta} \mathbb{1}_{\{1,2,\dots,\theta\}}(x)$ , and hence

$$f(x_1, \dots, x_n; \theta) = \frac{1}{\theta^n} \prod \mathbb{1}_{\{1, 2, \dots, \theta\}}(x_i) = \frac{1}{\theta^n} \mathbb{1}_{\{1, 2, \dots, \theta\}}(\max\{x_i\}).$$

Thus  $Y_n = \max\{X_i\}$  is a sufficient statistic by the factorization theorem. For (b), note that  $f_{Y_n}(y) = \frac{y^n}{\theta^n} - \frac{(y-1)^n}{\theta^n}$ , so

$$\mathbb{E}\left[\frac{Y^{n+1} - (Y-1)^{n+1}}{Y^n - (Y-1)^n}\right] = \sum_{y=1}^{\theta} \frac{y^{n+1} - (y-1)^{n+1}}{y^n - (y-1)^n} \left(\frac{y^n}{\theta^n} - \frac{(y-1)^n}{\theta^n}\right)$$
$$= \frac{1}{\theta^n} \sum_{y=1}^{\theta} y^{n+1} - (y-1)^{n+1}$$
$$= \frac{1}{\theta^n} \theta^{n+1} = \theta.$$

Since this is an unbiased estimator in terms of the sufficient statistic Y, it is an mvue.