

Turn in solutions for any 3 of the 4 problems. If you attempt all 4 problems, each will count 1/4 instead of 1/3. You may use one formula sheet, but exchanging formula sheets is not allowed.

1. Let  $X_1$  and  $X_2$  be independent EXP(1) random variables (pdf  $e^{-x}$ ,  $x > 0$ ). Find the joint pdf of (a)  $Y_1 = X_1/X_2$  and (b)  $Y_2 = X_1 + X_2$ .

Hint for inverse transformation: Solve eq. (a) for  $X_1$  in terms of  $Y_1$  and  $X_2$  and call this eq. (c); solve eq. (b) for  $X_2$  in terms of  $Y_2$  and  $X_1$  and call this eq. (d). Substitute eq. (d) into eq. (c) and solve for  $X_1$ .

Sol. (c)  $X_1 = Y_1 X_2$ . (d)  $X_2 = Y_2 - X_1$ . Hence  $X_1 = Y_1(Y_2 - X_1) = Y_1 Y_2 - Y_1 X_1$  or  $(1 + Y_1)X_1 = Y_1 Y_2$ , hence

$$X_1 = \frac{Y_1 Y_2}{1 + Y_1}, \quad X_2 = \frac{Y_2}{1 + Y_1}.$$

Jacobian is

$$J = \det \begin{pmatrix} y_2/(1+y_1)^2 & y_1/(1+y_1) \\ -y_2/(1+y_1)^2 & 1/(1+y_1) \end{pmatrix} = \frac{y_2 + y_1 y_2}{(1+y_1)^3} = \frac{y_2}{(1+y_1)^2},$$

Conclude that

$$f_{Y_1, Y_2}(y_1, y_2) = f_{X_1} \left( \frac{y_1 y_2}{1 + y_1} \right) f_{X_2} \left( \frac{y_2}{1 + y_1} \right) \frac{y_2}{(1 + y_1)^2} = y_2 e^{-y_2} / (1 + y_1)^2,$$

where  $y_1 > 0$  and  $y_2 > 0$ .

2. Let  $X_1, X_2, X_3$  be a random sample of size 3 from the UNIF(0, 1) distribution.

(a) Find the joint pdf of the order statistics  $Y_1, Y_2, Y_3$ . Part of the problem is to identify the  $y_1, y_2, y_3$  values for which the formula is valid.

(b) Find the (marginal) joint pdf of  $Y_1$  and  $Y_3$ . Part of the problem is to identify the  $y_1, y_3$  values for which the formula is valid.

(c) Find the mean of the sample range  $R = Y_3 - Y_1$ . (It is not necessary to find the pdf of  $R$ .)

Sol. (a) By a theorem,  $f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) = 6$ ,  $0 < y_1 < y_2 < y_3 < 1$ .

(b)  $f_{Y_1, Y_3}(y_1, y_3) = \int_{y_1}^{y_3} f_{Y_1, Y_2, Y_3}(y_1, y_2, y_3) dy_2 = 6(y_3 - y_1)$ ,  $0 < y_1 < y_3 < 1$ .

(c)  $E[Y_3 - Y_1] = \int \int (y_3 - y_1) 6(y_3 - y_1) dy_1 dy_3 = \int_0^1 \int_{y_1}^1 6(y_3 - y_1)^2 dy_3 dy_1 = \int_0^1 2(1 - y_1)^3 dy_1 = \frac{1}{2}$ .

Alternatively,  $E[Y_3 - Y_1] = E[Y_3] - E[Y_1] = \int_0^1 y_3 3y_3^2 dy_3 = \int_0^1 y_1 3(1 - y_1)^2 dy_1 = \frac{3}{4} - \frac{1}{4} = \frac{1}{2}$ .

3. Let  $X_1, X_2, \dots, X_n$  be independent LOGN( $\mu, \sigma^2$ ) (i.e.,  $\ln(X_1), \ln(X_2), \dots, \ln(X_n)$  are independent N( $\mu, \sigma^2$ )). Use a theorem to show that the sample

median is asymptotically normal and find its asymptotic mean and asymptotic variance. For simplicity, assume that  $n$  is odd. In case the LOGN density is needed, it has the form  $f(x) = (2\pi\sigma^2)^{-1/2}x^{-1}e^{-(\ln x - \mu)^2/(2\sigma^2)}$ ,  $x > 0$ .

Sol.  $X_1 = e^{Y_1}, \dots, X_n = e^{Y_n}$  with  $Y_1, \dots, Y_n$  i.i.d.  $N(\mu, \sigma^2)$ . What is  $x_{1/2}$  for  $X_1$ ?  $\frac{1}{2} = P(X_1 > x_{1/2}) = P(Y_1 > \ln x_{1/2})$  implies  $\ln x_{1/2} = \mu$  or  $x_{1/2} = e^\mu$ . Conclude by a theorem that the sample median  $X_{(n+1)/2:n}$  is asymptotically normal with asymptotic mean  $x_{1/2} = e^\mu$  and asymptotic variance  $c^2/n$ , where

$$c^2 = \frac{\frac{1}{2}(1 - \frac{1}{2})}{f_{X_1}(x_{1/2})^2} = \frac{1/4}{[(\sqrt{2\pi\sigma^2} e^\mu)^{-1}]^2} = (\pi/2)\sigma^2 e^{2\mu}.$$

Alternative solution:  $Y_{(n+1)/2:n}$  is asymptotically normal with asymptotic mean  $\mu$  and asymptotic variance  $c^2/n$ , where

$$c^2 = \frac{\frac{1}{2}(1 - \frac{1}{2})}{f_{Y_1}(\mu)^2} = \frac{1/4}{[(\sqrt{2\pi\sigma^2})^{-1}]^2} = (\pi/2)\sigma^2.$$

Now  $X_{(n+1)/2:n} = e^{Y_{(n+1)/2:n}}$ , which is therefore asymptotically normal with asymptotic mean  $e^\mu$  and asymptotic variance  $e^{2\mu}c^2/n$  with  $c^2$  as above.

4. Let  $V_n$  be  $\chi^2$  with  $n$  degrees of freedom.

(a) Show that  $(1/n)V_n$  is asymptotically normal, and determine the asymptotic mean and asymptotic variance. Hint: If  $Z$  is  $N(0, 1)$ , then  $\text{Var}(Z^2) = 2$ .

(b) Explain why  $e^{(1/n)V_n}$  is asymptotically normal, and find its asymptotic mean and asymptotic variance.

Sol. (a) By the central limit theorem,  $\bar{X}_n$  is asymptotically normal with asymptotic mean  $\mu$  and asymptotic variance  $\sigma^2/n$ . Now  $(1/n)V_n$  is of the form of  $\bar{X}_n$  with  $\mu = E[Z^2] = 1$  and  $\sigma^2 = \text{Var}(Z^2) = 2$ , so  $(1/n)V_n$  is asymptotically normal with asymptotic mean 1 and asymptotic variance  $2/n$ .

(b) We apply a theorem with  $g(x) = e^x$ , hence  $g'(x) = e^x$  also. We get that  $e^{(1/n)V_n}$  is asymptotically normal with asymptotic mean  $g(1) = e$  and asymptotic variance  $g'(1)^2(2/n) = 2e^2/n$ .