

1. Show that if  $\beta_x(t) = f_x(t|\mathbf{x} > t)$ ,  $\beta_y(t) = f_y(t|\mathbf{y} > t)$  and  $\beta_x(t) = k\beta_y(t)$  then  $1 - F_x(x) = [1 - F_y(x)]^k$
2. Express the density  $f_x(y)$  of the RV  $\mathbf{y} = g(\mathbf{x})$  in terms of  $f_x(x)$  if (a)  $g(x) = |x|$ ; (b)  $g(x) = e^{-x}U(x)$ .
3. The RVs  $\mathbf{x}$  and  $\mathbf{y}$  are independent with exponential densities  
 $f_x(x) = \alpha e^{-\alpha x}U(x)$ ,  $f_y(y) = \beta e^{-\beta y}U(y)$   
 Find the densities of the following RVs:  $2\mathbf{x} + \mathbf{y}$ ;  $\frac{\mathbf{x}}{\mathbf{y}}$
4. The RVs  $\mathbf{x}$  and  $\mathbf{y}$  are  $N(0; \sigma)$  and independent. Show that, if  $\mathbf{z} = |\mathbf{x} - \mathbf{y}|$ , then

$$E\{\mathbf{z}\} = 2\sigma/\sqrt{\pi}, E\{\mathbf{z}^2\} = 2\sigma^2.$$

5. Use the moment generating function, show that the linear transformation of a Gaussian random vector is also Gaussian.
6. Let  $\{x_k(n)\}_{k=1}^4$  be four IID random variables with exponential distribution with  $\alpha = 1$ .

$$y_k(n) = \sum_{l=1}^k x_l(n), 1 \leq k \leq 4$$

- (a) Determine and plot the pdf of  $y_2(n)$
  - (b) Determine and plot the pdf of  $y_3(n)$
  - (c) Determine and plot the pdf of  $y_4(n)$
  - (d) Compare the pdf of  $y_4(n)$  with that of the Gaussian density.
7. The mean and covariance of a Gaussian random vector  $\mathbf{x}$  are given by, respectively,

$$\mu_x = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

and

$$\mathbf{\Gamma}_x = \begin{bmatrix} 1 & \frac{1}{2} \\ \frac{1}{2} & 1 \end{bmatrix}$$

Plot the  $1\sigma$ ,  $2\sigma$ , and  $3\sigma$  concentration ellipses representing the contours of the density function in the  $(x_1, x_2)$  plane. *Hints:* The radius of an ellipse with major axis  $a$  (along  $x_1$ ) and minor axis  $b < a$  (along  $x_2$ ) is given by

$$r^2 = \frac{a^2 b^2}{a^2 \sin^2 \theta + b^2 \cos^2 \theta}$$

where  $0 \leq \theta \leq 2\pi$ . Compute the  $1\sigma$  ellipse specified by  $a = \sqrt{\lambda_1}$  and  $b = \sqrt{\lambda_2}$  and then rotate and translate each point  $\mathbf{x}^{(i)} = [x_1^{(i)} x_2^{(i)}]$  using the transformation  $\mathbf{w}^{(i)} = \mathbf{Q}_x \mathbf{x}^{(i)} + \mu_x$ .