

**Question 6.1**

A scalar random variable  $X$  is sent through a multiplicative coding channel which generates the output  $Z = XY$ : The independent positive scalar random variables  $X$  and  $Y$  have the densities  $\{f_X(x) = x^{-2}, 1 \leq x < \infty\}$  and  $\{f_Y(y) = 2y^{-3}, 1 \leq y < \infty\}$  respectively.

- (a) Find the densities of the random variables:  $X' = \log X$ ,  $1 \leq X < \infty$ , and  $Y' = \log Y$ ,  $1 \leq Y < \infty$ ; on  $R_+ = [0, 1)$ .
- (b) Find the characteristic functions of the densities of  $X'$ ,  $Y'$  and  $Z' = \log Z$ .
- (c) Give the probability density of the coded version  $Z$  of the signal  $X$ .

**Q 6.1 Solution:**

- (a) Given  $X' = \log X$  and  $Y' = \log Y$ , find  $f_{X'}(x')$  and  $f_{Y'}(y')$  respectively.

By definition, the densities for  $X'$  and  $Y'$  are defined as follows:

$$f_{X'}(x') = f_X(x(x')) \left| \frac{\partial x(x')}{\partial x'} \right|$$

$$f_{Y'}(y') = f_Y(y(y')) \left| \frac{\partial y(y')}{\partial y'} \right|$$

It follows that

$$x'(x) = \log x \Rightarrow x(x') = e^{x'} \Rightarrow \frac{\partial x(x')}{\partial x'} = e^{x'}$$

$$y'(y) = \log y \Rightarrow y(y') = e^{y'} \Rightarrow \frac{\partial y(y')}{\partial y'} = e^{y'}$$

and

$$\begin{aligned}f_X(x(x')) &= (e^{x'})^{-2} = e^{-2x'}, \quad 0 \leq x' < \infty \\f_Y(y(y')) &= 2(e^{y'})^{-3} = 2e^{-3y'}, \quad 0 \leq y' < \infty\end{aligned}$$

Then the final densities will be as follows:

$$\begin{aligned}f_{X'}(x') &= e^{-2x'} e^{x'} = e^{-x'}, \quad 0 \leq x' < \infty \\f_{Y'}(y') &= 2e^{-3y'} e^{y'} = 2e^{-2y'}, \quad 0 \leq y' < \infty\end{aligned}$$

(b) Find the characteristic functions of the densities of  $X'$ ,  $Y'$  and  $Z' = \log Z$ .

Following the definition of the characteristic function

$$\begin{aligned}\Phi_{X'}(\omega) &= \int_{-\infty}^{\infty} e^{j\omega x'} f_{X'}(x') dx' \\ \Phi_{Y'}(\omega) &= \int_{-\infty}^{\infty} e^{j\omega y'} f_{Y'}(y') dy'\end{aligned}$$

Since  $x'$  and  $y'$  are defined in certain ranges, and substituting for  $f_{X'}(x')$  and  $f_{Y'}(y')$  respectively, the characteristic functions become as follows:

$$\begin{aligned}\Phi_{X'}(\omega) &= \int_0^{\infty} e^{j\omega x'} e^{-x'} dx' = \frac{1}{(1 - j\omega)} \\ \Phi_{Y'}(\omega) &= \int_0^{\infty} e^{j\omega y'} 2e^{-2y'} dy' = \frac{2}{(2 - j\omega)}\end{aligned}$$

For  $Z' = \log Z$ , its expression can be simplified as follows:

$$\begin{aligned}Z' &= \log Z \\ &= \log XY = \log X + \log Y = X' + Y'\end{aligned}$$

which is simply the summation of two independent random variables. The characteristic function for  $Z'$ , by definition, will be as follows:

$$\begin{aligned}
 \Phi_{Z'}(\omega) &= \Phi_{(X'+Y')}(\omega) \\
 &= \Phi_{X'}(\omega)\Phi_{Y'}(\omega) \\
 &= \frac{1}{(1-j\omega)} \frac{2}{(2-j\omega)} \\
 &= \frac{2}{(1-j\omega)(2-j\omega)}, \quad \omega \in R
 \end{aligned}$$

(c) Give the probability density of the coded version  $Z$  of the signal  $X$ .

Using the property of characteristic functions that they have unique inverses, we can restore the density  $f_{Z'}(z')$  from  $\Phi_{Z'}(\omega)$  then use the transformation  $Z = e^{Z'}$  to restore the original density  $f_Z(z)$ . First, let us expand the above solution for  $\Phi_{Z'}(\omega)$  and using partial fraction to see it clearly.

$$\begin{aligned}
 \Phi_{Z'}(\omega) &= \frac{2}{(1-j\omega)(2-j\omega)} \\
 &= \frac{a}{(1-j\omega)} + \frac{b}{(2-j\omega)} \\
 &= \frac{2}{(1-j\omega)} + \frac{-2}{(2-j\omega)} \\
 \Phi_{(X'+Y')}(\omega) &= 2\Phi_{X'}(\omega) - \Phi_{Y'}(\omega)
 \end{aligned}$$

By the inverse Fourier transform, we know that  $\Phi_{(X'+Y')}(\omega)$  must be the characteristic function of the density  $f_{X'+Y'}(x' + y')$  which is actually the density  $f_{Z'}(z')$ . Therefore,

$$f_{Z'}(z') = 2e^{-z'} - 2e^{-2z'}, \quad z' \geq 0$$

Using the transformation rule

$$f_Z(z) = f_{Z'}(z'(z)) \left| \frac{\partial z'(z)}{\partial z} \right|,$$
$$z'(z) = \log z \Rightarrow \frac{\partial z'(z)}{\partial z} = 1/z$$
$$f_{Z'}(z'(z)) = 2e^{-\log z} - 2e^{-2\log z}$$

This makes the final density

$$f_Z(z) = \left( \frac{2}{z} - \frac{2}{z^2} \right) \frac{1}{z}, \quad z \geq 1$$

You can verify the answers by integrating  $f_Z(z)$  from 1 to  $\infty$  and  $f_{Z'}(z')$  from 0 to  $\infty$ .

### Question 6.2

Find (a) the mean value  $\mu$ , and (b) the variance  $\sigma^2$  of an RV  $X$  with the *Laplace density*

$$f_X(x) = \frac{1}{2b} e^{-2|x-m|/2b},$$

where  $b$  and  $m$  are real constants,  $b > 0$  and  $-\infty < m < \infty$ .

Find the corresponding characteristic function  $\Phi_X(\omega)$  and verify the values found above for  $\mu$ ,  $\sigma^2$  by use of the Moment Theorem.

**Q 6.2 Solution:**

We can find the mean and variance of  $X$  from a table or by integrating

$$\begin{aligned} EX &= \int_{-\infty}^{\infty} (x - m + m) f_X(x) dx \\ &= \int_{-\infty}^{\infty} (x - m) \frac{1}{2b} e^{-2|x-m|/2b} dx + \int_{-\infty}^{\infty} m f_X(x) dx \\ &= \int_{-\infty}^{\infty} y \frac{1}{2b} e^{-2|y|/2b} dy + m \\ &= 0 + m \quad (\text{by symmetry of the integrand}). \end{aligned}$$

Integration by parts can be used to obtain  $EX^2$ . In the end, we find:

$$\mu = m,$$

$$\sigma^2 = 2b^2.$$

The characteristic function is given by:

$$\Phi_X(\omega) = \int_{-\infty}^{\infty} e^{j\omega x} f_X(x) dx = \int_{-\infty}^{\infty} e^{j\omega x} \frac{1}{2b} e^{-2|x-m|/2b} dx.$$

Using the hint, we set the parameter  $m=0$ , making  $X$  a zero-mean R.V. The characteristic

function in this case is given by:

$$\begin{aligned}
 \Phi_X(\omega) &= \int_{-\infty}^{\infty} e^{j\omega x} \frac{1}{2b} e^{-2|x|/2b} dx \\
 &= \frac{1}{2b} \int_{-\infty}^0 e^{x(\frac{1}{b}+j\omega)} dx + \frac{1}{2b} \int_0^{\infty} e^{x(-\frac{1}{b}+j\omega)} dx \\
 &= \frac{1}{2b} \left( \frac{1}{\frac{1}{b}+j\omega} e^{x(\frac{1}{b}+j\omega)} \Big|_{-\infty}^0 - \frac{1}{-\frac{1}{b}+j\omega} e^{x(-\frac{1}{b}+j\omega)} \Big|_0^{\infty} \right) \\
 &= \frac{1}{2b} \left( \frac{1}{\frac{1}{b}+j\omega} + \frac{1}{\frac{1}{b}-j\omega} \right) \\
 &= \frac{1}{1+b^2\omega^2}
 \end{aligned}$$

Now for a R.V.  $X$  of mean  $m$ , and using the formula on page 8 of the Lecture to account for the shift, the characteristic function becomes:

$$\Phi_X(\omega) = \frac{e^{jm\omega}}{1+b^2\omega^2}.$$

By the moment theorem,

$$\begin{aligned}
 EX &= (1/j) \frac{d}{d\omega} \frac{e^{jm\omega}}{1+b^2\omega^2} \Big|_{\omega=0} \\
 &= (1/j) \left( \frac{-2b^2\omega}{(1+b^2\omega^2)^2} e^{jm\omega} + \frac{jme^{jm\omega}}{1+b^2\omega^2} \right) \Big|_{\omega=0} \\
 &= m.
 \end{aligned}$$

By taking one more derivative, we find

$$\begin{aligned}
 EX^2 &= -\frac{d}{d\omega} \left( \frac{-2b^2\omega}{(1+b^2\omega^2)^2} e^{jm\omega} + \frac{jm e^{jm\omega}}{1+b^2\omega^2} \right) \Big|_{\omega=0} \\
 &= -\left( \frac{-(2b^2)(1+b^2\omega^2)^2 + 2b^2\omega(1+b^2\omega^2)(2b^2\omega)}{(1+b^2\omega^2)^4} e^{jm\omega} + \frac{-2b^2\omega}{(1+b^2\omega^2)^2} (jm) e^{jm\omega} \right. \\
 &\quad \left. + \frac{-2b^2\omega}{(1+b^2\omega^2)^2} (jm) e^{jm\omega} + \frac{1}{(1+b^2\omega^2)^2} (jm)^2 e^{jm\omega} \right) \Big|_{\omega=0} \\
 &= 2b^2 + m^2.
 \end{aligned}$$

Hence, the variance is  $2b^2$ .

### Question 6.3

- (a) The exponential random variable  $Z$  has the density  $f_Z(\cdot) = \{5\mu e^{-5\mu z}, z \in R_+, \mu > 0\}$ .

Find the characteristic function  $\Phi_Z(\omega), \omega \in R$ .

- (b) Let the random variable  $W$  be defined by  $W = 3(Z_1) - 3(Z_2)$ , where  $Z_1$  and  $Z_2$  are independent identically distributed exponential random variables with parameter  $\lambda$ .

Find  $\Phi_W(\omega), \omega \in R$ .

- (c) Using the one-to-one relation of characteristic functions and densities and part (b), find the density of  $W$ . (Hint: check Question 2.)

### Q 6.3 Solution:

- (a) We can obtain the characteristic function from a table or by integration. Let's integrate.

$$\Phi_Z(w) = \int_0^{\infty} e^{jtx} 5\mu e^{-5\mu x} dx = \frac{5\mu}{5\mu - jw}.$$

(b) First, observe that the characteristic function of  $-3Z_2$  is

$$\Phi_{-3Z_2}(w) = \Phi_{Z_2}(-3w),$$

by the scaling property seen earlier. By independence, we have

$$\begin{aligned} \Phi_W(w) &= \Phi_{3Z_1}(w)\Phi_{-3Z_2}(w) \\ &= \Phi_{Z_1}(3w)\Phi_{Z_2}(-3w) \\ &= \frac{\lambda}{\lambda - j3w} \frac{\lambda}{\lambda + j3w} \\ &= \frac{\lambda^2}{\lambda^2 + 9w^2} = \Phi_X(w). \end{aligned}$$

(c) Since  $\Phi_W(t) = \Phi_X(t)$ , by the inversion formula,  $X$  and  $W$  have the same distribution if we let  $\lambda = \frac{3}{b}$  and  $m = 0$ . Hence, for all  $x$

$$f_W(x) = \frac{\lambda e^{-\frac{\lambda|x|}{3}}}{6}$$

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#### Question 6.4

$X$  is a scalar random variable distributed  $N(\mu, \sigma^2) = N(1, 4)$ ,  $\sigma > 0$ . Let  $Z = g(X) = \frac{X^2}{2}$ .

Find the density  $f_Z(z)$  of  $Z$  for  $z > 0$ .

#### Q 6.4 Solution:

$X$  is a random variable with Gaussian distribution  $N(1, 4)$ , so

$$f_X(x) = \frac{1}{2\sqrt{2\pi}} e^{-\frac{(x-1)^2}{8}}.$$

Since  $g(x) = \frac{x^2}{2}$  then  $g^{-1}(x) = \pm\sqrt{2x}$ . we know

$$f_Z(z) = \sum_{g^{-1}(z) \in A_i} f_X(g^{-1}(z)) \cdot \left| \frac{dg^{-1}(z)}{dz} \right|.$$



Therefore

$$f_Z(z) = \frac{1}{2\sqrt{2\pi}} e^{-\frac{(\sqrt{2z}-1)^2}{8}} \times \frac{1}{\sqrt{2z}} + \frac{1}{2\sqrt{2\pi}} e^{-\frac{(-\sqrt{2z}-1)^2}{8}} \times \frac{-1}{\sqrt{2z}}$$

which equals to

$$f_Z(z) = \frac{1}{4\sqrt{\pi z}} \left[ e^{-\frac{(\sqrt{2z}-1)^2}{8}} - e^{-\frac{(-\sqrt{2z}-1)^2}{8}} \right],$$

for  $z \geq 0$  and 0 otherwise.

### Question 6.5

Many people believe that the daily change of price of a company's stock on the stock market is a random variable with mean 0 and variance  $\sigma^2$ . That is, if  $Y_n$  represents the price of the stock on the  $n$ th day, then

$$Y_n = Y_{n-1} + X_n \quad n \geq 1$$

where  $X_1, X_2, \dots$  are independent and identically distributed random variables with mean 0 and variance  $\sigma^2$ . Suppose that the stock's price today is 100. If  $\sigma^2 = 1$ , what can you say about the probability that the stock's price will exceed 105 after 10 days?

### Q 6.5 Solution:

We have  $Y_{10} = Y_0 + \sum_{i=1}^{10} X_i$ ,  $E(Y_{10}) = \mu = 100$  and  $E((Y_{10} - \mu)^2) = \sigma^2 = 10$ . We wish to apply Chebyshev's inequality

$$P(|Y_{10} - \mu| \geq k\sigma) \leq \frac{1}{k^2}$$

Noting the symmetry of  $Y_{10}$ , we can half Chebyshev's inequality, drop the absolute value and hence obtain a bound for  $P(Y_{10} > 105)$  (with  $k = \frac{5}{\sqrt{10}}$ )

$$P(Y_{10} - 100 \geq 5) \leq \frac{1}{2k^2} = \frac{1}{5}$$