ECE541: Stochastic Signals and Systems

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Chapter 3: Functions of Random Variables

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1 Functions of Random Variables of the Type Y = g(X)

Example 1. Let X be a uniform RV on (0,1), that is, X:U(0,1), and let Y=2X+3. What is the pdf of Y?

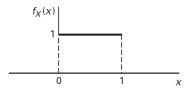
$$F_Y(y) = Pr(Y \le y)$$

$$= Pr(2X + 3 \le y)$$

$$= Pr\left(X \le \frac{y - 3}{2}\right)$$

$$= F_X\left(\frac{y - 3}{2}\right).$$

$$f_Y(y) = \frac{dF_Y(y)}{dy} = \frac{1}{2}f_X\left(\frac{y - 3}{2}\right).$$



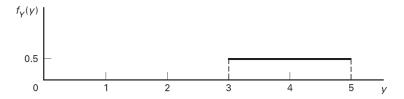


Figure 1: PDF of X and Y.

Generalization: Let Y = aX + b, where $a \ (a \neq 0)$ and b are certain constants and X is continuous RV with pdf $f_X(x)$. Then the pdf of Y is given by:

$$f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right).$$

Example 2. Let X be a RV with continuous CDF $F_X(x)$ and let $Y = X^2$. What is the pdf of Y?

$$F_Y(y) = Pr(Y \le y)$$

= $Pr(X^2 \le y)$.

For $y \geq 0$,

$$\begin{split} F_Y(y) &= Pr(-\sqrt{y} \le X \le \sqrt{y}) \\ &= F_X(+\sqrt{y}) - F_X(-\sqrt{y}) + Pr(X = -\sqrt{y}). \\ f_Y(y) &= \frac{dF_Y(y)}{dy} \\ &= \frac{dF_X(+\sqrt{y})}{d(\sqrt{y})} \frac{d(\sqrt{y})}{dy} - \frac{dF_X(-\sqrt{y})}{d(-\sqrt{y})} \frac{d(-\sqrt{y})}{dy} \\ &= f_X(\sqrt{y}) \times \frac{1}{2\sqrt{y}} + f_X(-\sqrt{y}) \times \frac{1}{2\sqrt{y}} \\ &= \frac{1}{2\sqrt{y}} (f_X(\sqrt{y}) + f_X(-\sqrt{y})). \end{split}$$

Suppose that $X \sim N(0,1)$, then the pdf of Y in this case is given by:

$$f_Y(y) = \begin{cases} \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}} & \text{if } y \ge 0, \\ 0 & \text{if } y < 0. \end{cases}$$

Theorem 1. Given a continuous RV X with pdf $f_X(x)$, and a differentiable function g(X). The pdf of Y = g(X) is given by,

$$f_Y(y) = \sum_{i=1}^n \frac{f_X(x_i)}{|g'(x_i)|},$$

where the x_i 's, i = 1, ..., n, are the roots of y = g(x) and $g'(x_i) \neq 0$.

Example 3. Let X be a random variable uniformly distributed over $(-\pi, +\pi)$ and let $Y = \sin X$. What is the pdf of Y?

In this example, $g(x) = \sin x$ and $g'(x) = \cos x$. The equation g(x) = y has two roots for |y| < 1, which are given by $x_1 = \sin^{-1} y$ and $x_2 = \pi - \sin^{-1} y$. By applying Theorem 1,

$$f_Y(y) = \frac{f_X(x_1)}{|g'(x_1)|} + \frac{f_X(x_2)}{|g'(x_2)|}$$

$$= \frac{1}{2\pi} \frac{1}{|\cos(\sin^{-1}y)|} + \frac{1}{2\pi} \frac{1}{|\cos(\pi - \sin^{-1}y)|}$$

$$= \frac{1}{\pi} \frac{1}{|\cos(\sin^{-1}y)|}.$$

To evaluate $\cos(\sin^{-1} y)$ we make use of figure 2.

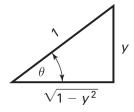


Figure 2: Evaluating $\cos(\sin^{-1} y)$.

As shown in figure 2, $\theta = \sin^{-1} y$ and $\cos \theta = \sqrt{1 - y^2} = \cos(\sin^{-1} y)$. Hence,

$$f_Y(y) = \begin{cases} \frac{1}{\pi} \frac{1}{\sqrt{1 - y^2}} & \text{if } |y| < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Example 4. A student at a train station awaits the arrival of either a red or a green train. At this station, red and green trains arrive independently with a rate $\lambda_r = 0.1$ train/min for red trains and a rate of $\lambda_g = 0.5$ trains/min for green trains. Let T_R be the time the student waits until a red train arrives, and T_G be the time the students waits until a green train arrives. Given $T_G \sim \exp(\lambda_g)$ and $T_R \sim \exp(\lambda_r)$.

1. What is the probability that the green train arrives first?

$$Pr(T_G < T_R) = \int_0^{+\infty} Pr(T_R > t | T_G = t) f_{T_G}(t) dt$$

$$= \int_0^{+\infty} (1 - Pr(T_R \le t | T_G = t)) f_{T_G}(t) dt$$

$$= \int_0^{+\infty} (1 - F_{T_R}(t)) f_{T_G}(t) dt$$

$$= \int_0^{+\infty} e^{-\lambda_r t} \lambda_g e^{-\lambda_g t} dt$$

$$= \frac{\lambda_g}{\lambda_g + \lambda_r}$$

$$= \frac{5}{6}.$$

2. Let T be the time the student waits until a red or a green train arrives. What is the pdf of T? Intuitively, T can be expressed as

$$T = min\{T_R, T_G\}.$$

Therefore for $t \geq 0$,

$$Pr(T \le t) = Pr(min\{T_R, T_G\} \le t)$$

$$= 1 - Pr(min\{T_R, T_G\} > t)$$

$$= 1 - Pr(T_R > t, T_G > t)$$

$$= 1 - Pr(T_R > t)Pr(T_G > t)$$

$$= 1 - (1 - Pr(T_R \le t))(1 - Pr(T_G \le t))$$

$$= 1 - (1 - F_{T_R}(t))(1 - F_{T_G}(t))$$

$$= 1 - e^{-(\lambda_r + \lambda_g)t}.$$

Therefore, $T \sim exp(\lambda_r + \lambda_q) = exp(0.6)$.

2 Functions of Random Variables of the Type Z = g(X, Y)

Theorem 2. Given two independent random variables X and Y with pdfs $f_X(x)$ and $f_Y(y)$ respectively, the pdf of Z = X + Y is given by,

$$f_Z(z) = (f_X * f_Y)(z) = \int_{-\infty}^{+\infty} f_X(y) f_Y(z - y) dy.$$

Example 5. Let X and Y be independent random variables such that $X \sim exp(1)$ and $Y \sim Uniform(-1,1)$, and let Z = X + Y. What is the pdf of Z?

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(y) f_Y(z - y) dy.$$

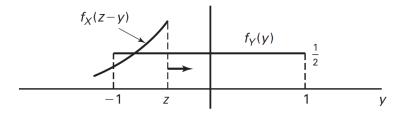


Figure 3: Relative positions of $f_X(z-y)$ and $f_Y(y)$.

1. If
$$z \leq -1$$
,

$$f_Z(z) = 0.$$

2. If
$$-1 \le z \le 1$$
,

$$f_Z(z) = \frac{1}{2} \int_{-1}^{z} e^{-(z-y)} dy = \frac{1}{2} (1 - e^{-1-z}).$$

3. If
$$z \ge 1$$
,

$$f_Z(z) = \frac{1}{2} \int_{-1}^{1} e^{-(z-y)} dy = \frac{1}{2} \left(e^{1-z} - e^{-1-z} \right).$$

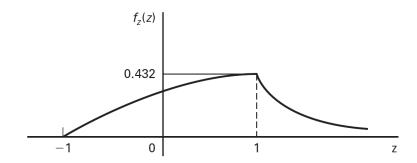


Figure 4: The pdf $f_Z(z)$.

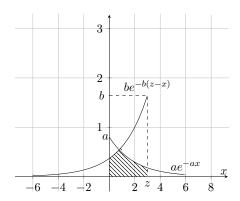
Example 6. Let X and Y be independent random variables such that $X \sim exp(a)$ and $Y \sim exp(b)$, and let Z = X + Y. What is the pdf of Z?

$$f_X(x) = \begin{cases} ae^{-ax} & \text{if } x \ge 0\\ 0 & \text{if } x < 0 \end{cases}$$

$$f_Y(y) = \begin{cases} be^{-by} & \text{if } y \ge 0\\ 0 & \text{if } y < 0 \end{cases}$$

$$f_Y(y) = \begin{cases} be^{-by} & \text{if } y \ge 0\\ 0 & \text{if } y < 0 \end{cases}$$

$$f_Z(z) = \int_{-\infty}^{+\infty} f_X(y) f_Y(z-y) dy.$$



1. If
$$z \leq 0$$
,

$$f_Z(z) = 0.$$

2. If
$$z \ge 0$$
,

$$f_Z(z) = \int_0^z abe^{-ay}e^{-b(z-y)}dy$$
$$= abe^{-bz}\int_0^z e^{(b-a)y}dy.$$

Therefore, for $z \geq 0$,

$$f_Z(z) = \begin{cases} \frac{ab}{a-b} \left(e^{-bz} - e^{-az} \right) & \text{if } a \neq b, \\ abz e^{-bz} & \text{if } a = b. \end{cases}$$

Example 7. Let X and Y be two iid (independent and identically distributed) random variables such that $X, Y \sim N(0, 1)$.

1. What is the pdf of $Z = X^2 + Y^2$?

Method 1:

$$Z = T + W$$
$$T = X^{2}$$
$$W = Y^{2}$$

Since T and W are independent, $f_Z = f_T * f_W$. Method 2:

$$F_{Z}(z) = Pr(Z \le z)$$

$$= Pr(X^{2} + Y^{2} \le z)$$

$$= \iint_{x^{2} + y^{2} \le z} f_{X,Y}(x, y) dx dy$$

$$= \iint_{x^{2} + y^{2} \le z} f_{X}(x) f_{Y}(y) dx dy$$

$$= \frac{1}{2\pi} \iint_{x^{2} + y^{2} \le z} e^{-\frac{x^{2}}{2}} e^{-\frac{y^{2}}{2}} dx dy$$

$$= \frac{1}{2\pi} \iint_{x^{2} + y^{2} \le z} e^{-\frac{x^{2} + y^{2}}{2}} dx dy.$$

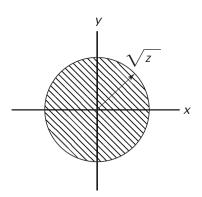


Figure 5: The region of the event $\{X^2 + Y^2 \le z\}$ for $z \ge 0$.

We evaluate this integral by transforming to polar coordinates,

$$x = r \cos \theta.$$
$$y = r \sin \theta.$$
$$x^{2} + y^{2} = r^{2}.$$

Therefore,

$$F_Z(z) = \int_0^{\sqrt{z}} \int_0^{2\pi} \frac{e^{-\frac{r^2}{2}}}{|J|} d\theta dr.$$

Where |J| is determinant of the Jacobian of the transformation and is given by,

$$J = \begin{vmatrix} \frac{\partial r}{\partial x} & \frac{\partial r}{\partial y} \\ \\ \frac{\partial \theta}{\partial x} & \frac{\partial \theta}{\partial y} \end{vmatrix}$$

|J| is also equal to the inverse of the Jacobian of the inverse transformation

$$|J|^{-1} = \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r \cos^2 \theta + r \sin^2 \theta = r.$$

Therefore, for $z \geq 0$,

$$F_Z(z) = \frac{1}{2\pi} \int_0^{\sqrt{z}} \int_0^{2\pi} r e^{-\frac{r^2}{2}} d\theta dr$$

$$= \frac{1}{2\pi} \int_0^{\sqrt{z}} r e^{-\frac{r^2}{2}} dr \int_0^{2\pi} d\theta$$

$$= 1 - e^{-\frac{z}{2}}.$$

$$f_Z(z) = \frac{1}{2} e^{-\frac{z}{2}}.$$

Therefore, $Z \sim exp(0.5)$.

2. What is the pdf of $Z' = \sqrt{X^2 + Y^2}$?

$$F_{Z'}(z') = Pr(Z' \le z')$$

$$= Pr(Z'^2 \le z'^2)$$

$$= Pr(Z \le z'^2)$$

$$= F_Z(z'^2)$$

$$= 1 - e^{-\frac{z'^2}{2}}.$$

$$f_{Z'}(z') = z'e^{-\frac{z'^2}{2}}. \quad (Rayleigh \ distribution)$$

Example 8. Let X and Y be two iid random variables such that $X, Y \sim N(0, 1)$, and let Z = Y/X. What is the pdf of Z?

By conditioning on X (fixing) and applying the general linear transformation we get,

$$f_{Z|X=x}(z|X=x) = \frac{|x|}{\sqrt{2\pi}}e^{-\frac{x^2z^2}{2}}.$$

Therefore, by applying the total law of probability,

$$f_Z(z) = \int_{-\infty}^{+\infty} f_{Z|X=x}(z|X=x) f_X(x) dx$$

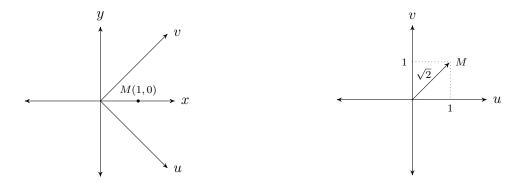
$$= \int_{-\infty}^{+\infty} \frac{|x|}{\sqrt{2\pi}} e^{-\frac{x^2 z^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{2\pi} \int_{-\infty}^{+\infty} |x| e^{-\frac{x^2 (1+z^2)}{2}} dx$$

$$= \frac{1}{\pi} \frac{1}{1+z^2}.$$

3 Functions of Random Variables of the Type U = g(X,Y) and V = h(X,Y)

Example 9. Let X and Y be two iid random variables such that $X, Y \sim N(0, 1)$. Let U = X + Y and V = X - Y. What is joint pdf of U and V? Consider the point M shown in the figures below.



This figures illustrate a case of a one-to-one mapping, because the linear system of equations

$$\begin{cases} X + Y &= 1 \\ X - Y &= 1 \end{cases}$$

is invertible, i.e. $\begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2 \neq 0$. In fact,

$$f_{U,V}(1,1) = \frac{f_{X,Y}(1,0)}{|J|}.$$

Where,

$$J = \begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{vmatrix} = \begin{vmatrix} 1 & 1 \\ 1 & -1 \end{vmatrix} = -2.$$

$$\Rightarrow f_{U,V}(1,1) = \frac{f_{X,Y}(1,0)}{2}.$$

Generalizing,

$$f_{U,V}(u,v) = \frac{f_{X,Y}(x,y)}{|J|},$$

such that,

$$x = \frac{u+v}{2},$$
$$y = \frac{u-v}{2}.$$

Therefore,

$$f_{U,V}(u,v) = \frac{1}{2} f_{X,Y} \left(\frac{u+v}{2}, \frac{u-v}{2} \right)$$
$$= \frac{1}{4\pi} \exp \left[-\frac{1}{8} \left[(u+v)^2 - (u-v)^2 \right] \right].$$

Theorem 3. Given two continuous RVs X and Y with pdfs $f_X(x)$ and $f_Y(y)$ respectively, and two differentiable functions $g_1(x)$ and $g_2(x)$. The joint pdf of $U = g_1(X,Y)$ and $V = g_2(X,Y)$ is given by,

$$f_{U,V}(u,v) = \sum_{i=1}^{n} \frac{f_{X,Y}(x_i, y_i)}{|J(x_i, y_i)|},$$

where the pairs (x_i, y_i) , i = 1, ..., n, are the solutions of the system of equations given by,

$$\begin{cases} g_1(x,y) = u \\ g_2(x,y) = v \end{cases}.$$