

Random-Process

1.1 Introduction

Random Processes:

- Any physical quantity that varies with time is a signal.
- Examples of signals are an electrocardiogram (ECG), and an electroencephalogram (EEG) signals.
- There are two types signals which are
 1. Continuous time signals
 2. Discrete time signals
- Continuous time signals have the time variable t takes values from $-\infty$ to ∞ or in an interval between t_1 to t_2 .
- Continuous time signals are indicated as $x(t)$, $y(t)$, $z(t)$ and so on.
- A discrete-time signal is a set of measurements taken sequentially in time (e.g., at every millisecond).
- Each measurement point is usually called a sample, and a discrete-time signal is indicated by $x(n)$, $y(n)$, $z(n)$, where the index n is an integer that points to the order of the measurements in the sequence.
- A random process is a time-varying function that contains the outcome of a random experiment for each time instant, $X(t)$.
- A random process is a time varying function, e.g., a signal.
- A random process consists of infinite number of random variables.
- Random Process are of two types
 1. Continuous random process
 2. Discrete random process
- Real random process also called stochastic process
- The collection of all possible sample functions of $X(t)$ is called an ensemble of $X(t)$.

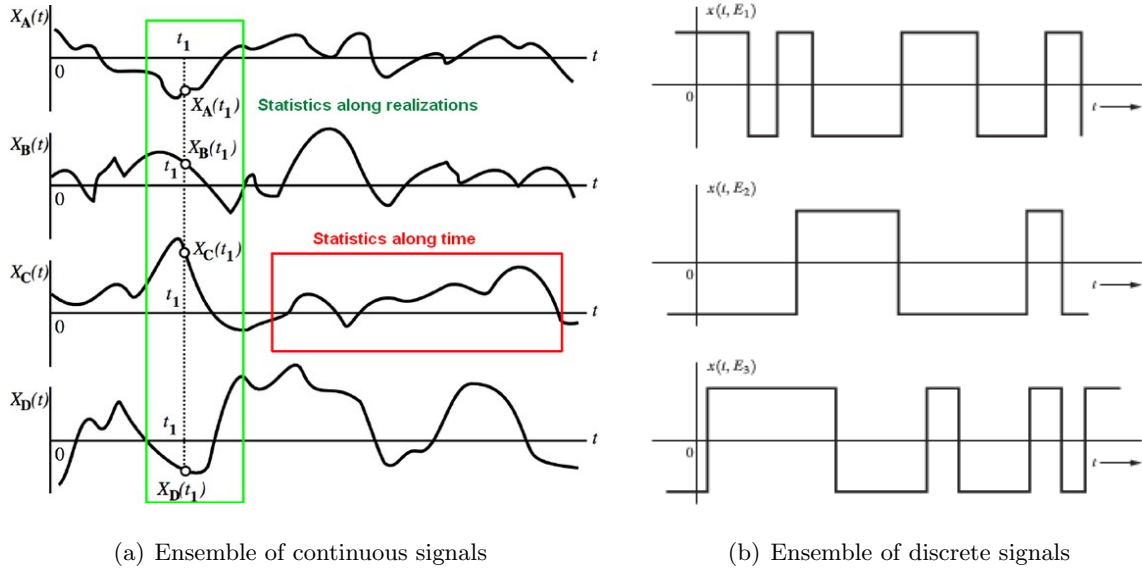


Figure 1.1: An Ensemble of Signals

1.2 Probability Distribution and Density Functions

The cdf of a random process is defined as

$$F_X(x, t) = P\{x(t) \leq x\}$$

The pdf of a random process is defined as

$$f_X(x, t) = \frac{dF_X(x, t)}{dx}$$

The bivariate cdf of a random process is defined as

$$F_{X(t_1)X(t_2)}(x_1, x_2) = P\{X(t_1) \leq x_1, X(t_2) \leq x_2\}$$

The bivariate pdf of a random process is defined as

$$f_{X(t_1)X(t_2)}(x_1, x_2) = \frac{\partial^2}{\partial x_1 \partial x_2} F_{X(t_1)X(t_2)}$$

The bivariate cdf of a nth order multivariate random process is defined as

$$F_{X(t_1)X(t_2)\dots X(t_n)}(x_1, x_2, \dots, x_n) = P\{X(t_1) \leq x_1, X(t_2) \leq x_2, \dots, X(t_n) \leq x_n\}$$

Similarly nth order multivariate density functions of order n exist

$$f_{X(t_1)X(t_2)\dots X(t_n)}(x_1, x_2, \dots, x_n) = \frac{\partial^n}{\partial x_1 \partial x_2 \dots \partial x_n} F_{X(t_1)X(t_2)\dots X(t_n)}$$

The mean of a random process is defined as

$$E[X(t)] = \int_{-\infty}^{\infty} x f_X(x, t) dx$$

The mean square μ_X of a random process is defined as

$$\overline{X^2(t)} = E[X^2(t)] = \int_{-\infty}^{\infty} x^2 f_X(x, t) dx$$

The variance σ_X^2 of a random process is defined as

$$\begin{aligned}
 \sigma_X^2 &= E[(X - \mu_X)^2] \\
 &= E[X^2 - 2\mu_X X + \mu_X^2] \\
 &= E[X^2] - 2\mu_X E[X] + \mu_X^2 \\
 &= E[X^2] - 2\mu_X^2 + \mu_X^2 \\
 &= E[X^2] - \mu_X^2
 \end{aligned}$$

1.3 Stationary

A random process $\{X(t)\}$ is stationary if the pdfs and its statistical properties are invariant with changes in time. For example, for a stationary process, $X(t)$ and $X(t + \Delta)$ have the same probability distributions.

$$F_{X(t)}(x) = F_{X(t+\Delta)}(x) \quad \text{for all } t, t + \Delta$$

Stationary of order one: Consider a pdf of random process $X(t)$ is

$$f_{X(t+t_s)}(x) = f_{X(t)}(x) = f_X(x)$$

The pdf is independent of any time shift hence its mean of the random process $X(t)$ is also a constant

$$\mu_{X(t)}(x) = \mu_X$$

The variance of the random process $X(t)$ is also a constant

$$\sigma_{X(t)}^2 = \sigma_X^2$$

Stationary of order two: Consider a pdf of random process $X(t)$ is

$$f_{X(t_1+t_s)X(t_2+t_s)}(x_1, x_2) = f_{X(t_1)X(t_2)}(x_1, x_2)$$

Let

$$\begin{aligned}
 t_1 + t_s &= t \\
 t_s &= t - t_1 \\
 t_2 + t_s &= t_2 + t - t_1 \\
 t_2 + t_s &= t + (t_2 - t_1) \\
 &= t + \tau
 \end{aligned}$$

The mean of the random process $X(t)$ is also a constant

$$\begin{aligned}
 E[X(t)] &= \mu_X \\
 E[X(t_1)X(t_2)] &= E[X(t)X(t + \tau)]
 \end{aligned}$$

The correlation of a wide sense stationer is independent of time, it depends upon τ . When the time difference is 0 i.e., $\tau = t_2 - t_1 = 0 \Rightarrow t_1 = t_2 = t$

$$E[X(t)X(t)] = E[X^2(t)] = \sigma_X^2 + \mu_X^2$$

The above relation is a autocorrelation $R_X(\tau)$ of a random process $X(t)$

$$R_X(\tau) = E[X^2(t)] = \sigma_X^2 + \mu_X^2$$

If a random process $X(t)$ has first and second order stationary, then it called Wide-Sense Stationary Random Processes

Wide-Sense Stationary Random Processes A continuous-time random process $X(t)$ is wide-sense stationary (WSS) if it follows the following properties

1. The mean is independent of time t

$$E[X(t)] = \mu_{X(t)} = \mu_X = \text{constant}$$

The variance of the random process $X(t)$ is also a constant

$$\sigma_{X(t)}^2 = \sigma_X^2$$

2. The autocorrelation function only depends on time difference

$$R_X(\tau) = E[X(t)X(t)] = E[X^2(t)] = \sigma_X^2 + \mu_X^2$$

A discrete-time random process $\{X(n), n \in Z\}$ is weak-sense stationary or wide-sense stationary (WSS) if

1. $\mu_X(n) = \mu_X$ for all $n \in Z$
2. $R_X(n_1, n_2) = R_X(n_1 - n_2)$ for all $n_1, n_2 \in Z$

1.4 Correlation Functions:

The correlation functions between two random variables is a measure of the similarity between the variables.

There are two types of correlation functions

1. Auto-Correlation Functions
2. Cross-Correlation Functions

1.4.1 Auto-Correlation Functions:

The correlation of a signal $X(t)$ with itself is called as autocorrelation. This is denoted as

$$R_X(\tau) = E[X(t)X(t + \tau)]$$

where t and τ are arbitrary. When $\tau = 0$ the autocorrelation function is the average of the random process squared. It is also called average power.

$$R_X(0) = E[X(t)X(t)] = E[X^2(t)] = \sigma_X^2 + \mu_X^2$$

For $-\tau$

$$R_X(-\tau) = E[X(t)X(t - \tau)]$$

Let $t' = t - \tau$

$$\begin{aligned} R_X(-\tau) &= E[X(t' + \tau)X(t')] \\ &= E[X(t')X(t' + \tau)] \end{aligned}$$

$$R_X(-\tau) = R_X(\tau)$$

The autocorrelation function is an even function of τ .

$$E[X_i X_j] = \begin{cases} E[X_i^2] = E[X^2] = \mu_X^2 + \sigma_X^2 & j = i \\ E[X_i X_i] = \mu_X^2 & j \neq i \end{cases}$$

The bounds on auto-correlation function is

$$\begin{aligned} E[\{X(t) \pm X(t + \tau)\}^2] &\geq 0 \\ E[\{X^2(t) \pm 2X(t)X(t + \tau) + X^2(t + \tau)\}^2] &\geq 0 \end{aligned}$$

By performing expectations

$$R_X(0) \pm 2R_X(\tau) + R_X(0) \geq 0$$

$$|R_X(\tau)| \leq R_X(0)$$

If $X(t)$ is a random process with non zero mean, then it is defined as **auto-covariance function** $C_X(\tau)$

$$\begin{aligned} C_X(\tau)R_X(0) &= E[\{X(t) - \mu_X\}\{X(t + \tau) - \mu_X\}] \\ &= R_X(\tau) - \mu_X^2 \end{aligned}$$

The power spectral density of is defined as

$$\int_{-\infty}^{\infty} R_X(\tau) e^{-j\omega\tau} d\tau \geq 0 \text{ for all } \omega$$

Properties of Auto Correlation Functions

- $R_X(\tau)$ is bounded as

$$|R_X(\tau)| \leq R_X(0)$$

- Auto Correlation function is even symmetry

$$R_X(-\tau) = R_X(\tau)$$

- The mean value of the random process is obtained using autocorrelation using the following relation

$$E[X^2(t)] = \overline{X^2(t)} = \sigma_X^2 + \mu_X^2 = R_X(0)$$

1.4.2 Cross-Correlation Functions:

Consider a two random process $X(t)$ and $Y(t)$ are wide sense stationary. When we consider both the random process it is called as jointly wide sense stationary. Then their cross-correlation function is defined as

$$\begin{aligned} R_{XY}(\tau) &= E[X(t)Y(t+\tau)] \\ R_{YX}(\tau) &= E[Y(t)X(t+\tau)] \end{aligned}$$

where t and τ are arbitrary. The order of the subscript is

$$R_{XY}(-\tau) = R_{YX}(\tau)$$

If $Y(t)$ is periodic with period T , then

$$R_{XY}(\tau + T) = R_{XY}(\tau)$$

If $X(t)$ is periodic with T , then

$$R_{YX}(\tau + T) = R_{YX}(\tau)$$

The bounds on cross-correlation function is

$$\begin{aligned} E[\{X(t) \pm kY(t+\tau)\}^2] &\geq 0 \\ E[\{X^2(t) \pm 2kX(t)Y(t+\tau) + k^2Y^2(t+\tau)\}^2] &\geq 0 \end{aligned}$$

By performing expectations

$$R_X(0) \pm 2kR_{XY}(\tau) + k^2R_Y(0) \geq 0$$

If $k = 1$

$$R_X(0) \pm 2R_{XY}(\tau) + R_Y(0) \geq 0$$

Then it becomes

$$|R_{XY}(\tau)| \leq \frac{1}{2}[R_X(0) + R_Y(0)]$$

If k is a positive and real constant then

$$k^2R_Y(0) + 2kR_{XY}(\tau) + R_X(0) \geq 0$$

The quadratic will be never negative if it does not have real roots. If its discriminant is

$$4R_{XY}^2(\tau) - 4R_X(0)R_Y(0) \leq 0$$

$$|R_{XY}(\tau)| \leq \sqrt{R_X(0)R_Y(0)}$$

For example for a quadratic equation

$$ax^2 + bx + c^2 = 0$$

then its discriminant is $b^2 - 4ac$

The geometric mean is

$$\sqrt{R_X(0)R_Y(0)} \leq \frac{1}{2}[R_X(0) + R_Y(0)]$$

A cross correlation function is

$$\begin{aligned} C_{XY}(\tau) &= E[(X(t) - \mu_X)(Y(t + \tau) - \mu_Y)] \\ &= R_{XY}(\tau) - \mu_X\mu_Y \end{aligned}$$

$$\begin{aligned} C_{YX}(\tau) &= E[(Y(t) - \mu_Y)(X(t + \tau) - \mu_X)] \\ &= R_{YX}(\tau) - \mu_Y\mu_X \end{aligned}$$

Addition and Subtraction

$$\begin{aligned} \sin(A + B) &= \sin A \cos B + \cos A \sin B \\ \sin(A - B) &= \sin A \cos B - \cos A \sin B \\ \cos(A + B) &= \cos A \cos B - \sin A \sin B \\ \cos(A - B) &= \cos A \cos B + \sin A \sin B \\ \tan(A + B) &= \frac{\tan A + \tan B}{1 - \tan A \tan B} \\ \tan(A - B) &= \frac{\tan A - \tan B}{1 + \tan A \tan B} \end{aligned}$$

Product Identities

$$\begin{aligned} \sin A \cos B &= \frac{1}{2}(\sin(A + B) + \sin(A - B)) \\ \cos A \sin B &= \frac{1}{2}(\sin(A + B) - \sin(A - B)) \\ \cos A \cos B &= \frac{1}{2}(\cos(A + B) + \cos(A - B)) \\ \sin A \sin B &= \frac{1}{2}(\cos(A - B) - \cos(A + B)) \end{aligned}$$

Example 4.2 The random process described by

$$Y(t) = A \cos(\omega_c t + \Theta)$$

where A and $\omega_c t$ are constants and Θ is a random variable distributed uniformly between $\pm\pi$. Find the pdf of random variable Θ , mean and autocorrelation function of Y

Solution:

The pdf of random variable Θ

$$\begin{aligned} f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (-\pi)} \\ &= \frac{1}{2\pi} \\ f_{\Theta}(\theta) &= \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

The mean of random variable Y

$$\begin{aligned} \mu_Y &= \int_{-\pi}^{\pi} A \cos(\omega_c t + \theta) \frac{1}{2\pi} d\theta \\ &= \frac{A}{2\pi} \int_{-\pi}^{\pi} \cos \omega_c t \cos \theta - \sin \omega_c t \sin \theta d\theta \\ &= \frac{A}{2\pi} [\cos \omega_c t \sin \theta + \sin \omega_c t \cos \theta]_{-\pi}^{\pi} \\ &= \frac{A}{2\pi} [0 + \sin \omega_c t (-1 - (-1))] \\ &= 0 \end{aligned}$$

Autocorrelation function $R_Y(\tau)$

$$\begin{aligned} E[Y(t)Y(t+\tau)] &= \int_{-\pi}^{\pi} \left(\frac{1}{2\pi} A \cos(\omega_c t + \theta) \right) \left(\frac{1}{2\pi} A \cos(\omega_c(t+\tau) + \theta) \right) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \cos(\omega_c t + \theta) \cos(\omega_c(t+\tau) + \theta) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\theta) + \cos \omega_c \tau] d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\theta) + \cos \omega_c \tau] d\theta \\ x &= 2\omega_c t + \omega_c \tau + 2\theta \\ \frac{dx}{d\theta} &= 0 + 0 + 2 \\ d\theta &= \frac{dx}{2} \end{aligned}$$

Limits, when $\theta = \pi$

$$x = 2\omega_c t + \omega_c \tau + 2\pi$$

when $\theta = -\pi$

$$x = 2\omega_c t + \omega_c \tau - 2\pi$$

$$\begin{aligned} \int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta &= \int_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} (\cos x) \frac{dx}{2} \\ &= \frac{1}{2} [\sin x]_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} \\ &= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\ &= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\ &= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau) - \sin(2\omega_c t + \omega_c \tau)] \\ &= 0 \end{aligned}$$

$$\begin{aligned}\int_{-\pi}^{\pi} [\cos \omega_c \tau] d\theta &= \cos \omega_c \tau [\theta]_{-\pi}^{\pi} \\ &= 2\pi \cos \omega_c \tau\end{aligned}$$

$$\begin{aligned}E[Y(t)Y(t+\tau)] &= \frac{A^2}{2\pi} 2\pi \cos \omega_c \tau \\ &= \frac{A^2}{2} \cos \omega_c \tau\end{aligned}$$

The ensemble variance is constant

$$\sigma_Y^2 = R_Y(0) = \frac{A^2}{2}$$

From the above discussions it is observed that

1. $E[Y(t)] = 0 = \text{Constant}$ and
2. $E[Y(t)Y(t+\tau)] = R_Y(\tau)$, it is independent of absolute time hence

$Y(t)$ is a wide sense stationary.

Example 4.6 Consider two random process $X(t)$ and $Y(t)$ which are independent, jointly wide sense stationary random process described by

$$\begin{aligned}X(t) &= A \cos(\omega_1 t + \Theta_1) \\ Y(t) &= B \cos(\omega_2 t + \Theta_2)\end{aligned}$$

where A, B and ω_1, ω_2 are constants and Θ_1 and Θ_2 are random variable distributed uniformly between $\pm\pi$. Let $X(t)$ and $Y(t)$ are related by

$$W(t)X(t) = X(t)Y(t)$$

Find the autocorrelation function of W

Solution:

The autocorrelation function of $X(t)$ and $Y(t)$ are

$$\begin{aligned}R_X(\tau) &= \frac{A^2}{2} \cos \omega_1 \tau \\ R_Y(\tau) &= \frac{B^2}{2} \cos \omega_2 \tau\end{aligned}$$

$$\begin{aligned}R_W(\tau) &= \frac{A^2}{2} \cos \omega_1 \tau \frac{B^2}{2} \cos \omega_2 \tau \\ &= \frac{A^2 B^2}{4} \cos \omega_1 \tau \cos \omega_2 \tau \\ &= \frac{A^2 B^2}{8} [\cos(\omega_1 + \omega_2)\tau + (\omega_1 - \omega_2)\tau]\end{aligned}$$

6. The random process described by

$$X(t) = A \sin(\omega_c t + \Theta)$$

where A and ω_c are constants and Θ is a random variable uniformly distributed between $\pm\pi$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned} f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (-\pi)} \\ &= \frac{1}{2\pi} \end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= \int_{-\pi}^{\pi} A \sin(\omega_c t + \theta) \frac{1}{2\pi} d\theta \\ &= \frac{A}{2\pi} \int_{-\pi}^{\pi} \sin \omega_c t \cos \theta - \cos \omega_c t \sin \theta d\theta \\ &= \frac{A}{2\pi} [\sin \omega_c t \sin \theta + \cos \omega_c t \cos \theta]_{-\pi}^{\pi} \\ &= \frac{A}{2\pi} [0 + \sin \omega_c t (-1 - (-1))] \\ &= 0 \end{aligned}$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned} E[X(t)X(t+\tau)] &= \int_{-\pi}^{\pi} \frac{1}{2\pi} (A \sin(\omega_c t + \theta)) (A \sin(\omega_c(t+\tau) + \theta)) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \sin(\omega_c t + \theta) \sin(\omega_c(t+\tau) + \theta) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \frac{1}{2} [\cos \omega_c \tau - \cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta \\ &= \frac{A^2}{4\pi} \int_{-\pi}^{\pi} [\cos \omega_c \tau - \cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta \end{aligned}$$

$$\begin{aligned} x &= 2\omega_c t + \omega_c \tau + 2\theta \\ \frac{dx}{d\theta} &= 0 + 0 + 2 \\ d\theta &= \frac{dx}{2} \end{aligned}$$

Limits, when $\theta = \pi$

$$x = 2\omega_c t + \omega_c \tau + 2\pi$$

when $\theta = -\pi$

$$x = 2\omega_c t + \omega_c \tau - 2\pi$$

$$\begin{aligned}
\int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta &= \int_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} (\cos x) \frac{dx}{2} \\
&= \frac{1}{2} [\sin x]_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau) - \sin(2\omega_c t + \omega_c \tau)] \\
&= 0
\end{aligned}$$

$$\begin{aligned}
\int_{-\pi}^{\pi} [\cos \omega_c \tau] d\theta &= \cos \omega_c \tau [\theta]_{-\pi}^{\pi} \\
&= 2\pi \cos \omega_c \tau
\end{aligned}$$

$$\begin{aligned}
E[X(t)X(t + \tau)] &= \frac{A^2}{4\pi} 2\pi \cos \omega_c \tau \\
R_X(\tau) &= \frac{A^2}{2} \cos \omega_c \tau
\end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = 0 = \text{Constant}$ and
2. $E[X(t)X(t + \tau)] = R_X(\tau)$, it is independent of absolute time hence

$X(t)$ is a wide sense stationary.

7. The random process described by

$$X(t) = A \cos(\omega_c t + \phi + \Theta)$$

where A , ω_c and ϕ are constants and Θ is a random variable uniformly distributed between $\pm\pi$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned}
f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (-\pi)} \\
&= \frac{1}{2\pi}
\end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\mu_X = \int_{-\pi}^{\pi} \cos(\omega_c t + \phi + \Theta) \frac{1}{2\pi} d\theta$$

$$\begin{aligned} x &= \omega_c t + \phi + \theta \\ \frac{dx}{d\theta} &= 0 + 0 + d\theta \\ d\theta &= dx \end{aligned}$$

Limits, when $\theta = \pi$

$$x = \omega_c t + \phi + \pi$$

when $\theta = -\pi$

$$x = \omega_c t + \phi - \pi$$

$$\begin{aligned} \mu_X &= \int_{-\pi}^{\pi} \cos(\omega_c t + \phi + \Theta) \frac{1}{2\pi} d\theta \\ &= \frac{A}{2\pi} \int_{\omega_c t + \phi - \pi}^{\omega_c t + \phi + \pi} \cos x dx \\ &= \frac{A}{2\pi} [\sin x]_{\omega_c t + \phi - \pi}^{\omega_c t + \phi + \pi} \\ &= \frac{A}{2\pi} [\sin(\omega_c t + \phi + \pi) - \sin(\omega_c t + \phi - \pi)] \\ &= \frac{A}{2\pi} [-\sin(\omega_c t + \phi) + \sin(\omega_c t + \phi)] \\ &= 0 \end{aligned}$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned} E[X(t)X(t+\tau)] &= \int_{-\pi}^{\pi} \frac{1}{2\pi} (A \cos(\omega_c t + \phi + \theta)) (A \cos(\omega_c(t+\tau) + \phi + \theta)) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \cos(\omega_c t + \phi + \theta) \cos(\omega_c(t+\tau) + \phi + \theta) d\theta \\ &= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \frac{1}{2} [\cos \omega_c \tau - \cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta \\ &= \frac{A^2}{4\pi} \int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\phi + 2\theta) + \cos \omega_c \tau] d\theta \end{aligned}$$

$$\begin{aligned} x &= 2\omega_c t + \omega_c \tau + 2\theta \\ \frac{dx}{d\theta} &= 0 + 0 + 2 \\ d\theta &= \frac{dx}{2} \end{aligned}$$

Limits, when $\theta = \pi$

$$x = 2\omega_c t + \omega_c \tau + 2\pi$$

when $\theta = -\pi$

$$x = 2\omega_c t + \omega_c \tau - 2\pi$$

$$\begin{aligned}
\int_{-\pi}^{\pi} [\cos(2\omega_c t + \omega_c \tau + 2\theta)] d\theta &= \int_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} (\cos x) \frac{dx}{2} \\
&= \frac{1}{2} [\sin x]_{2\omega_c t + \omega_c \tau - 2\pi}^{2\omega_c t + \omega_c \tau + 2\pi} \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau + 2\pi) - \sin(2\omega_c t + \omega_c \tau - 2\pi)] \\
&= \frac{1}{2} [\sin(2\omega_c t + \omega_c \tau) - \sin(2\omega_c t + \omega_c \tau)] \\
&= 0
\end{aligned}$$

$$\begin{aligned}
\int_{-\pi}^{\pi} [\cos \omega_c \tau] d\theta &= \cos \omega_c \tau [\theta]_{-\pi}^{\pi} \\
&= 2\pi \cos \omega_c \tau
\end{aligned}$$

$$E[X(t)X(t + \tau)] = R_X(\tau) = \frac{A^2}{4\pi} 2\pi \cos \omega_c \tau = \frac{A^2}{2} \cos \omega_c \tau$$

From the above discussions it is observed that

1. $E[X(t)] = 0 = \text{Constant}$ and
2. $E[X(t)X(t + \tau)] = R_X(\tau)$, it is independent of absolute time hence

$X(t)$ is a wide sense stationary.

8. The random process described by

$$X(t) = A \cos(\omega_c t + \Theta) + B$$

where A , B and ω_c are constants and Θ is a random variable uniformly distributed between $\pm\pi$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned}
f_{\Theta}(\theta) &= \frac{1}{b - a} = \frac{1}{\pi - (-\pi)} \\
&= \frac{1}{2\pi}
\end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned}
\mu_X &= E[X(t)] = E[A \cos(\omega_c t + \Theta) + B] \\
&= \int_{-\pi}^{\pi} A \cos(\omega_c t + \theta) \frac{1}{2\pi} d\theta + B \int_{-\pi}^{\pi} \frac{1}{2\pi} d\theta
\end{aligned}$$

$$\begin{aligned}
x &= \omega_c t + \theta \\
dx &= 0 + d\theta \\
d\theta &= dx
\end{aligned}$$

Limits, when $\theta = \pi$

$$x = \omega_c t + \pi$$

when $\theta = -\pi$

$$x = \omega_c t - \pi$$

$$\begin{aligned}
\mu_X &= \int_{-\pi}^{\pi} A \cos(\omega_c t + \theta) \frac{1}{2\pi} d\theta \\
&= \frac{A}{2\pi} \int_{\omega_c t - \pi}^{\omega_c t + \pi} \cos x dx \\
&= \frac{A}{2\pi} [\sin x]_{\omega_c t - \pi}^{\omega_c t + \pi} \\
&= \frac{A}{2\pi} [\sin(\omega_c t + \pi) - \sin(\omega_c t - \pi)] \\
&= \frac{A}{2\pi} [-\sin(\omega_c t) + \sin(\omega_c t)] \\
&= 0
\end{aligned}$$

$$\begin{aligned}
B \int_{-\pi}^{\pi} \frac{1}{2\pi} d\theta &= \frac{B}{2\pi} [\theta]_{-\pi}^{\pi} \\
&= \frac{B}{2\pi} [2\pi] \\
&= B
\end{aligned}$$

Autocorrelation function $R_X(\tau)$

$$\cos A \cos B = \frac{1}{2} (\cos(A+B) + \cos(A-B))$$

$$\begin{aligned}
E[X(t)X(t+\tau)] &= E[(A \cos(\omega_c t + \phi + B)) (A \cos(\omega_c(t+\tau) + B))] \\
&= E[(A^2 \cos(\omega_c t + \phi)) (\cos(\omega_c(t+\tau) + \phi)) + AB (\cos(\omega_c(t+\tau) + \phi)) + AB (\cos(\omega_c t + \phi)) + B^2] \\
&= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \frac{1}{2} [\cos(2\omega_c t + \omega_c \tau + 2\theta) + \cos(\omega_c \tau)] d\theta + 0 + 0 + B^2 \\
&= \frac{A^2}{2} \cos(\omega_c \tau) + B^2
\end{aligned}$$

$$\begin{aligned}
E[X(t)X(t+\tau)] &= \frac{A^2}{2} \cos(\omega_c \tau) + B^2 \\
R_X(\tau) &= \frac{A^2}{2} \cos(\omega_c \tau) + B^2
\end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = B = \text{Constant}$ and
2. $E[X(t)X(t+\tau)] = R_X(\tau)$, it is independent of absolute time hence

$X(t)$ is a wide sense stationary.

10. The random process described by

$$X(t) = A^2 \cos^2(\omega_c t + \Theta)$$

where A and ω_c are constants and Θ is a random variable uniformly distributed between $\pm\pi$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned} f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (-\pi)} \\ &= \frac{1}{2\pi} \end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[A^2 \cos^2(\omega_c t + \Theta)] \\ &= \frac{A^2}{2} E[1 + \cos(2\omega_c t + 2\Theta)] \\ &= \frac{A^2}{2} + \frac{A^2}{2} \int_{-\pi}^{\pi} \cos(2\omega_c t + 2\theta) d\theta \end{aligned}$$

$$\begin{aligned} x &= 2\omega_c t + 2\theta \\ dx &= 0 + 2d\theta \\ d\theta &= \frac{dx}{2} \end{aligned}$$

Limits, when $\theta = \pi$

$$x = 2\omega_c t + 2\pi$$

when $\theta = -\pi$

$$x = 2\omega_c t - 2\pi$$

$$\begin{aligned} &= \int_{-\pi}^{\pi} \cos(2\omega_c t + 2\theta) \frac{1}{2\pi} d\theta \\ &= \frac{A^2}{4\pi} \int_{2\omega_c t - 2\pi}^{2\omega_c t + 2\pi} \cos x dx \\ &= \frac{A^2}{4\pi} [\sin x]_{2\omega_c t - 2\pi}^{2\omega_c t + 2\pi} \\ &= \frac{A^2}{4\pi} [\sin(2\omega_c t + 2\pi) - \sin(2\omega_c t - 2\pi)] \\ &= \frac{A^2}{4\pi} [\sin(2\omega_c t) - \sin(2\omega_c t)] = 0 \\ \mu_X &= \frac{A^2}{2} \end{aligned}$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned}
E[X(t)X(t+\tau)] &= E[(A^2 \cos^2(\omega_c t + \Theta)) (A \cos(\omega_c(t+\tau) + \Theta) + B) A^2 \cos^2(\omega_c t + \Theta)] \\
&= \frac{A^2}{4} E[\{1 + \cos(2\omega_c t + 2\Theta)\} \{1 + \cos(2\omega_c(t+\tau) + 2\Theta)\}] \\
&= \frac{A^2}{4} \left\{ 1 + \int_{-\pi}^{\pi} \cos(2\omega_c t + 2\theta) \frac{1}{2\pi} d\theta + \int_{-\pi}^{\pi} \cos(2\omega_c(t+\tau) + 2\theta) \frac{1}{2\pi} d\theta \right. \\
&\quad \left. + \int_{-\pi}^{\pi} \cos(2\omega_c t + 2\theta) \cos(2\omega_c(t+\tau) + 2\theta) \frac{1}{2\pi} d\theta \right\} \\
&= \frac{A^2}{4} \left\{ 1 + 0 + 0 + \frac{1}{4\pi} \int_{-\pi}^{\pi} [\cos(4\omega_c t + 2\omega_c \tau + 4\theta) + \cos(2\omega_c \tau)] d\theta \right\} \\
&= \frac{A^2}{4} \left\{ 1 + \frac{1}{2} \cos(2\omega_c \tau) \right\}
\end{aligned}$$

$$\begin{aligned}
E[X(t)X(t+\tau)] &= \frac{A^2}{4} \left\{ 1 + \frac{1}{2} \cos(2\omega_c \tau) \right\} \\
R_X(\tau) &= \frac{A^2}{4} \left\{ 1 + \frac{1}{2} \cos(2\omega_c \tau) \right\}
\end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = \frac{A^2}{2} = \text{Constant}$ and
2. $E[X(t)X(t+\tau)] = R_X(\tau)$, it is independent of absolute time hence

$X(t)$ is a wide sense stationary.

12. The random process described by

$$X(t) = A \cos(\omega_c t + \Theta) + B \cos(\omega_s t)$$

where A, B, ω_c , and ω_s $\omega_c \neq \omega_s$ are constants and Θ is a random variable uniformly distributed between $\pm\pi$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned}
f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (-\pi)} \\
&= \frac{1}{2\pi}
\end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned}
\mu_X &= E[X(t)] = E[A \cos(\omega_c t + \Theta) + B \cos(\omega_s t)] \\
&= A \int_{-\pi}^{\pi} \cos(\omega_c t + 2\theta) \frac{1}{2\pi} d\theta + B \cos(\omega_s t) \\
&= B \cos(\omega_s t)
\end{aligned}$$

From the above discussions it is observed that $X(t)$ is not a wide sense stationary, because its mean is not a constant.

13. The random process described by

$$X(t) = A \cos(\omega_c t + \Theta)$$

where A and ω_c , are constants and Θ is a random variable uniformly distributed between $\pm \frac{\pi}{2}$. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned} f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi/2 - (-\pi/2)} \\ &= \frac{1}{\pi} \end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{\pi} & -\pi/2 < \theta < \pi/2 \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[A \cos(\omega_c t + \Theta) \frac{1}{\pi}] \\ &= A \int_{-\pi/2}^{\pi/2} \cos(\omega_c t + \theta) \frac{1}{\pi} d\theta \end{aligned}$$

$$\begin{aligned} x &= \omega_c t + \theta \\ d\theta &= dx \end{aligned}$$

Limits, when $\theta = \pi/2$

$$x = \omega_c t + \pi/2$$

when $\theta = -\pi/2$

$$x = \omega_c t - \pi/2$$

$$\begin{aligned} \mu_X &= A \int_{-\pi/2}^{\pi/2} \cos(\omega_c t + \theta) \frac{1}{\pi} d\theta \\ &= \frac{A}{\pi} \int_{\omega_c t - \pi/2}^{\omega_c t + \pi/2} \cos(x) dx \\ &= \frac{A}{\pi} [\sin(\omega_c t + \pi/2) - \sin(\omega_c t - \pi/2)] \\ &= \frac{2A}{\pi} \cos(\omega_c t) \end{aligned}$$

From the above discussions it is observed that $X(t)$ is not a wide sense stationary, because its mean is not a constant.

14. The random process described by

$$X(t) = A \cos(\omega_c t + \Theta)$$

where A and ω_c , are constants and Θ is a random variable uniformly distributed between 0 and π . Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable Θ

$$\begin{aligned} f_{\Theta}(\theta) &= \frac{1}{b-a} = \frac{1}{\pi - (0)} \\ &= \frac{1}{\pi} \end{aligned}$$

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{\pi} & 0 < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[A \cos(\omega_c t + \Theta)] \frac{1}{\pi} \\ &= A \int_0^{\pi} \cos(\omega_c t + \theta) \frac{1}{\pi} d\theta \end{aligned}$$

$$\begin{aligned} x &= \omega_c t + \theta \\ d\theta &= dx \end{aligned}$$

Limits, when $\theta = 0$

$$x = \omega_c t$$

when $\theta = \pi$

$$x = \omega_c t + \pi$$

$$\begin{aligned} \mu_X &= A \int_0^{\pi} \cos(\omega_c t + \theta) \frac{1}{\pi} d\theta \\ &= \frac{A}{\pi} \int_{\omega_c t}^{\omega_c t + \pi} \sin(x) dx \\ &= \frac{A}{\pi} [\sin(\omega_c t + \pi) - \sin(\omega_c t)] \\ &= \frac{-2A}{\pi} \sin(\omega_c t) \end{aligned}$$

From the above discussions it is observed that $X(t)$ is not a wide sense stationary, because its mean is not a constant.

15. A random process described by

$$X(t) = V$$

where V is a random variable uniformly distributed between 0 and 4. That means that each realization of $X(t) = V$ is constant v , that the constant varies from one realization to the next, and that the variation is described as uniformly distributed between 0 and 4. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable X

$$\begin{aligned} f_V(v) &= \frac{1}{b-a} = \frac{1}{4-(0)} \\ &= \frac{1}{4} \\ f_X(x) &= \begin{cases} \frac{1}{4} & 0 < \theta < 4 \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[V] = \frac{1}{4} \int_0^4 v dv = \frac{1}{4} \left[\frac{v^2}{2} \right]_0^4 = \frac{1}{8} [4^2 - 0] \\ &= 2 \end{aligned}$$

$$\begin{aligned} E[X(t)X(t+\tau)] &= E[V^2] = \frac{1}{4} \int_0^4 v^2 dv = \frac{1}{4} \left[\frac{v^3}{3} \right]_0^4 \\ &= \frac{1}{12} [4^3 - 0] \\ &= \frac{16}{3} \end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = 2 = \text{Constant}$ and

2. $E[X(t)X(t+\tau)] = R_X(\tau) = \frac{16}{3}$, it is independent of absolute time hence

$X(t)$ is a wide-sense stationary.

16. A random process described by

$$X(t) = V$$

where V is a random variable uniformly distributed between 0 and 2. That means that each realization of $X(t) = V$ is constant v , that the constant varies from one realization to the next, and that the variation is described as uniformly distributed between 0 and 2. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable X

$$\begin{aligned} f_V(v) &= \frac{1}{b-a} = \frac{1}{2-(0)} \\ &= \frac{1}{2} \end{aligned}$$

$$f_X(x) = \begin{cases} \frac{1}{2} & 0 < \theta < 2 \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[V] = \frac{1}{2} \int_0^2 v dv = \frac{1}{2} \left[\frac{v^2}{2} \right]_0^2 = \frac{1}{4} [2^2 - 0] \\ &= 1 \end{aligned}$$

$$\begin{aligned} E[X(t)X(t+\tau)] &= E[V^2] = \frac{1}{2} \int_0^2 v^2 dv = \frac{1}{2} \left[\frac{v^3}{3} \right]_0^2 \\ &= \frac{1}{6} [2^3 - 0] \\ &= \frac{4}{3} \end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = 1 = \text{Constant}$ and

2. $E[X(t)X(t+\tau)] = R_X(\tau) = \frac{4}{3}$, it is independent of absolute time hence

$X(t)$ is a wide-sense stationary.

17. A random process described by

$$X(t) = V$$

where V is a random variable uniformly distributed between 0 and 3. That means that each realization of $X(t) = V$ is constant v , that the constant varies from one realization to the next, and that the variation is described as uniformly distributed between 0 and 3. Is $X(t)$ wide-sense stationary? If not, then why not? If so, then what are the mean and autocorrelation function for the random process? [?]

Solution:

The pdf of random variable X

$$\begin{aligned} f_V(v) &= \frac{1}{b-a} = \frac{1}{3-(0)} \\ &= \frac{1}{3} \end{aligned}$$

$$f_X(x) = \begin{cases} \frac{1}{3} & 0 < \theta < 3 \\ 0 & \text{otherwise} \end{cases}$$

The mean of random variable X

$$\begin{aligned} \mu_X &= E[X(t)] = E[V] = \frac{1}{3} \int_0^3 v dv = \frac{1}{3} \left[\frac{v^2}{2} \right]_0^3 = \frac{1}{6} [3^2 - 0] \\ &= \frac{3}{2} \end{aligned}$$

$$\begin{aligned}
E[X(t)X(t+\tau)] &= E[V^2] = \frac{1}{3} \int_0^3 v^2 dv = \frac{1}{3} \left[\frac{v^3}{3} \right]_0^3 \\
&= \frac{1}{9} [3^3 - 0] \\
&= 3
\end{aligned}$$

From the above discussions it is observed that

1. $E[X(t)] = \frac{3}{2}$ = Constant and

2. $E[X(t)X(t+\tau)] = R_X(\tau) = 3$, it is independent of absolute time hence

$X(t)$ is a wide-sense stationary.

1.4.3 Addition of Random Processes:

Consider the addition of two random independent, jointly wide-sense stationary random process $X(t)$ and $Y(t)$ is

$$W(t) = X(t) + Y(t)$$

The autocorrelation function of the sum $W(t)$ is

$$\begin{aligned} R_W(\tau) &= E[\{X(t) + Y(t)\}\{X(t + \tau) + Y(t + \tau)\}] \\ &= E[X(t)X(t + \tau) + X(t)Y(t + \tau) + Y(t)X(t + \tau) + Y(t)Y(t + \tau)] \\ &= R_X(\tau) + R_{XY}(\tau) + R_{YX}(\tau) + R_Y(\tau) \end{aligned}$$

If $X(t)$ and $Y(t)$ are independent, then the cross-correlation function is

$$R_{XY}(\tau) = R_{YX}(\tau) = \mu_X \mu_Y$$

If $X(t)$ or $Y(t)$ has a zero mean then

$$R_W(\tau) = R_X(\tau) + R_Y(\tau)$$

18. The autocorrelation function for random process $Z(t)$ is

$$R_Z(\tau) = \begin{cases} 50(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

A random process $X(t)$ is the sum

$$X(t) = Z(t) + U$$

where U is a random variable with a mean $\mu_U = 4$, a variance $\sigma_U^2 = 25$, and is independent of $Z(t)$. Find the autocorrelation function of $X(t)$ [?]

Solution:

$$E[Z(t)] = 0$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned} R_X(\tau) &= E[\{Z(t) + U\}\{Z(t + \tau) + U\}] \\ &= E[Z(t)Z(t + \tau) + UZ(t + \tau) + UZ(t) + U^2] \\ &= R_Z(\tau) + 0 + 0 + E[U^2] \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + 25 + 4^2 \\ &= R_Z(\tau) + 41 \end{aligned}$$

19. The autocorrelation function for random process $Z(t)$ is

$$R_Z(\tau) = \begin{cases} 40(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

A random process $X(t)$ is the sum

$$X(t) = Z(t) + U$$

where U is a random variable with a mean $\mu_U = 5$, a variance $\sigma_U^2 = 15$, and is independent of $Z(t)$. Find the autocorrelation function of $X(t)$ [?]

Solution:

$$E[Z(t)] = 0$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned} R_X(\tau) &= E[\{Z(t) + U\}\{Z(t + \tau) + U\}] \\ &= E[Z(t)Z(t + \tau) + UZ(t + \tau) + UZ(t) + U^2] \\ &= R_Z(\tau) + 0 + 0 + E[U^2] \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + 15 + 5^2 \\ &= R_Z(\tau) + 40 \end{aligned}$$

20. The autocorrelation function for random process $Z(t)$ is

$$R_Z(\tau) = \begin{cases} 60(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

A random process $X(t)$ is the sum

$$X(t) = Z(t) + U$$

where U is a random variable with a mean $\mu_U = 4$, a variance $\sigma_U^2 = 20$, and is independent of $Z(t)$. Find the autocorrelation function of $X(t)$ [?]

Solution:

$$E[Z(t)] = 0$$

Autocorrelation function $R_X(\tau)$

$$\begin{aligned} R_X(\tau) &= E[\{Z(t) + U\}\{Z(t + \tau) + U\}] \\ &= E[Z(t)Z(t + \tau) + UZ(t + \tau) + UZ(t) + U^2] \\ &= R_Z(\tau) + 0 + 0 + E[U^2] \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + \sigma_U^2 + \mu_U^2 \\ &= R_Z(\tau) + 20 + 4^2 \\ &= R_Z(\tau) + 36 \end{aligned}$$

21. The random process $X(t)$ has the autocorrelation function

$$R_X(\tau) = \begin{cases} 10(1 - |\tau|/\tau_N) & -\tau_N \leq \tau \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is independent of $X(t)$ and has the autocorrelation function

$$R_Y(\tau) = \begin{cases} 15(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

$$X(t) = Z(t) + U$$

where $T \gg \tau_N$. The random process $Z(t) = X(t) + Y(t)$. For $Z(t)$, find the autocorrelation function, its total power, its dc power, and its ac power. Is $Z(t)$ wide-sense stationary? [?]

Solution:

$$\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 15, \mu_Y = 0, \text{ Autocorrelation function } R_Z(\tau)$$

$$R_Z(\tau) = R_X(\tau) + R_Y(\tau)$$

It is mentioned that $T \gg \tau_N$, $R_Z(\tau)$ is summation of the autocorrelation function of $R_X(\tau)$ and $R_Y(\tau)$, $R_X(\tau)$ is within the limits of $R_Y(\tau)$, hence we have to consider the limits for $R_Z(\tau)$ from $0 \leq |\tau| \leq \tau_N$ and $\tau_N \leq |\tau| \leq T$
From $0 \leq |\tau| \leq \tau_N$ the $R_Z(\tau)$ is

$$R_Z(\tau) = 15(1 - |\tau|/T) + 10(1 - |\tau|/\tau_N)$$

From $\tau_N \leq |\tau| \leq T$ the $R_Z(\tau)$ is

$$R_Z(\tau) = 15(1 - |\tau|/T)$$

$$R_Z(\tau) = \begin{cases} 15(1 - |\tau|/T) + 10(1 - |\tau|/\tau_N) & 0 \leq |\tau| \leq \tau_N \\ 15(1 - |\tau|/T) & \tau_N \leq |\tau| \leq T \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is

$$\begin{aligned} &= \sigma_X^2 + \sigma_Y^2 = 10 + 15 \\ &= 25 \end{aligned}$$

22. The random process $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. Given that $W(t) = X(t) + Y(t)$ and

$$R_X(\tau) = 10 \exp\left(-\frac{|\tau|}{3}\right)$$

$$R_Y(\tau) = \begin{cases} 10 \left(\frac{3-|\tau|}{3}\right) & -3 \leq \tau \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

For $W(t)$, find its autocorrelation function, its total power, its dc power, and its ac power. Is $W(t)$ wide-sense stationary? [?]

Solution:

It is mentioned that $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. $\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 10, \mu_Y = 0$, **Autocorrelation function** $R_Z(\tau)$

$$R_W(\tau) = R_X(\tau) + R_Y(\tau)$$

$R_Z(\tau)$ is summation of the autocorrelation function of $R_X(\tau)$ and $R_Y(\tau)$, $R_X(\tau)$ is within the limits of $R_Y(\tau)$, hence we to consider the limits for $R_Z(\tau)$ from $0 \leq |\tau| \leq 3$ and other than this limits From $0 \leq |\tau| \leq 3$ the $R_Z(\tau)$ is

$$R_Z(\tau) = 10 \exp\left(-\frac{|\tau|}{3}\right) + 10 \left(\frac{3-|\tau|}{3}\right)$$

Otherwise

$$R_Z(\tau) = 10 \left(\frac{3-|\tau|}{3}\right)$$

$$R_Z(\tau) = \begin{cases} 10 \exp\left(-\frac{|\tau|}{3}\right) + 10 \left(\frac{3-|\tau|}{3}\right) & 0 \leq |\tau| \leq 3 \\ 10 \left(\frac{3-|\tau|}{3}\right) & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is

$$\begin{aligned} &= \sigma_X^2 + \sigma_Y^2 = 10 + 10 \\ &= 20 \end{aligned}$$

23. The random process $X(t)$ has the autocorrelation function

$$R_X(\tau) = \begin{cases} 10 \exp(1 - |\tau|/\tau_N) & -\tau_N \leq \tau \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is independent of $X(t)$ and has the autocorrelation function

$$R_Y(\tau) = 13 \left(\frac{\sin(\omega_B \tau)}{\omega_B \tau} \right)$$

where $(2\pi)/\omega_B \gg \tau_N$. The random process $Z(t) = X(t) + Y(t)$. For $Z(t)$, find the autocorrelation function, its total power, its dc power, and its ac power. Is $Z(t)$ wide-sense stationary? [?]

Solution:

It is mentioned that $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. $\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 13, \mu_Y = 0$, **Autocorrelation function** $R_Z(\tau)$

$$R_Z(\tau) = R_X(\tau) + R_Y(\tau)$$

$R_Z(\tau)$ is summation of the autocorrelation function of $R_X(\tau)$ and $R_Y(\tau)$, $R_X(\tau)$ is within the limits of $R_Y(\tau)$, hence we have to consider the limits for $R_Z(\tau)$ from $0 \leq |\tau| \leq \tau_N$ and other than this limits From $0 \leq |\tau| \leq \tau_N$ the $R_Z(\tau)$ is

$$R_Z(\tau) = 10 \exp(1 - |\tau|/\tau_N) + 13 \left(\frac{\sin(\omega_B \tau)}{\omega_B \tau} \right)$$

Otherwise

$$R_Z(\tau) = 13 \left(\frac{\sin(\omega_B \tau)}{\omega_B \tau} \right)$$

$$R_Z(\tau) = \begin{cases} 10(1 - |\tau|/\tau_N) + 13 \left(\frac{\sin(\omega_B \tau)}{\omega_B \tau} \right) & 0 \leq |\tau| \leq \tau_N \\ 13 \left(\frac{\sin(\omega_B \tau)}{\omega_B \tau} \right) & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is

$$\begin{aligned} &= \sigma_X^2 + \sigma_Y^2 = 10 + 13 \\ &= 23 \end{aligned}$$

24. A random process $X(t)$ has the autocorrelation function

$$R_X(\tau) = 15\exp(-2|\tau|) \quad -\infty < \tau < \infty$$

The random process $Y(t)$ is $Y(t) = X(t) - 3$

- a. What is the autocorrelation function of $Y(t)$?
- b. What are the total power, the dc power, and the ac power of $Y(t)$?
- c. What is the cross-correlation $R_{XY}(\tau)$? [?]

Solution:

It is given that $R_X(\tau)$ has $\sigma_X^2 = 15$ and $\mu_X = 0$.

It is given that $Y(t) = X(t) - 3$ $\sigma_Y^2 = 15$ and $\mu_Y = -3$.

- a. The autocorrelation function of $Y(t)$ is

$$\begin{aligned} R_Y(\tau) &= E[Y(t)Y(t+\tau)] = E[\{X(t) - 3\}\{X(t+\tau) - 3\}] \\ &= E[X(t)X(t+\tau) - 3X(t+\tau) - 3X(t) + 9] \\ &= R_X(\tau) + 9 \end{aligned}$$

- b. The total power, the dc power, and the ac power of $Y(t)$ are

The total power

$$E[Y^2(t)] = \sigma_Y^2 + \mu_Y^2 = 15 + 9 = 24$$

The DC power is

$$\mu_Y^2 = 9$$

The AC power is

$$\sigma_Y^2 = 15$$

- c. The cross-correlation $R_{XY}(\tau)$ is

$$\begin{aligned} R_{XY}(\tau) &= E[X(t)Y(t+\tau)] = E[\{X(t)\}\{X(t+\tau) - 3\}] \\ &= E[X(t)X(t+\tau) - 3X(t)] \\ &= R_X(\tau) \end{aligned}$$

25. A random process $X(t)$ has the autocorrelation function

$$R_X(\tau) = 10\cos(\omega_c\tau) \quad -\infty < \tau < \infty$$

The random process $Y(t)$ is $Y(t) = X(t) - 4$

- a. What is the autocorrelation function of $Y(t)$?
- b. What are the total power, the dc power, and the ac power of $Y(t)$?
- c. What is the cross-correlation $R_{XY}(\tau)$? [?]

Solution:

It is given that $R_X(\tau)$ has $\sigma_X^2 = 10$ and $\mu_X = 0$.

It is given that $Y(t) = X(t) - 4$ $\sigma_Y^2 = 10$ and $\mu_X = -4$.

a. The autocorrelation function of $Y(t)$ is

$$\begin{aligned} R_Y(\tau) &= E[Y(t)Y(t+\tau)] = E[\{X(t) - 4\}\{X(t+\tau) - 4\}] \\ &= E[X(t)X(t+\tau) - 4X(t+\tau) - 4X(t) + 16] \\ &= R_{XY}(\tau) + 16 \end{aligned}$$

b. The total power, the dc power, and the ac power of $Y(t)$ are

The total power

$$E[Y^2(t)] = \sigma_Y^2 + \mu_Y^2 = 10 + 16 = 26$$

The DC power is

$$\mu_Y^2 = 16$$

The AC power is

$$\sigma_Y^2 = 10$$

c. The cross-correlation $R_{XY}(\tau)$ is

$$\begin{aligned} R_{XY}(\tau) &= E[X(t)Y(t+\tau)] = E[\{X(t)\}\{X(t+\tau) - 4\}] \\ &= E[X(t)X(t+\tau) - 4X(t)] \\ &= R_X(\tau) \end{aligned}$$

26. A random process $X(t)$ has the autocorrelation function

$$R_X(\tau) = \begin{cases} 13(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is $Y(t) = X(t) - 2$

a. What is the autocorrelation function of $Y(t)$?

b. What are the total power, the dc power, and the ac power of $Y(t)$?

c. What is the cross-correlation $R_{XY}(\tau)$? [?]

Solution:

It is given that $R_X(\tau)$ has $\sigma_X^2 = 13$ and $\mu_X = 0$.

It is given that $Y(t) = X(t) - 2$ $\sigma_Y^2 = 13$ and $\mu_X = -2$.

a. The autocorrelation function of $Y(t)$ is

$$\begin{aligned} R_Y(\tau) &= E[Y(t)Y(t+\tau)] = E[\{X(t) - 2\}\{X(t+\tau) - 2\}] \\ &= E[X(t)X(t+\tau) - 2X(t+\tau) - 2X(t) + 4] \\ &= R_{XY}(\tau) + 4 \end{aligned}$$

b. The total power, the dc power, and the ac power of $Y(t)$ are

The total power

$$E[Y^2(t)] = \sigma_Y^2 + \mu_Y^2 = 13 + 4 = 17$$

The DC power is

$$\mu_Y^2 = 4$$

The AC power is

$$\sigma_Y^2 = 13$$

c. The cross-correlation $R_{XY}(\tau)$ is

$$\begin{aligned} R_{XY}(\tau) &= E[X(t)Y(t+\tau)] = E[\{X(t)\}\{X(t+\tau) - 2\}] \\ &= E[X(t)X(t+\tau) - 2X(t)] \\ &= R_X(\tau) \end{aligned}$$

27. $X(t)$ and $Y(t)$ are zero mean jointly wide-sense stationary random process. The random process $Z(t)$ is

$$Z(t) = 3X(t) + Y(t)$$

Find the correlations $R_Z(\tau), R_{ZX}(\tau), R_{XZ}(\tau), R_{ZY}(\tau)$ and $R_{YZ}(\tau)$ [?]

Solution:

$$\begin{aligned} R_Z(\tau) &= E[Z(t)Z(t+\tau)] = E[\{3X(t) + Y(t)\}\{3X(t+\tau) + Y(t+\tau)\}] \\ &= E[\{9X(t)X(t+\tau)\} + \{3X(t)Y(t+\tau)\} + \{3X(t+\tau)Y(t)\} + \{Y(t)Y(t+\tau)\}] \\ &= 9R_X(\tau) + 3R_{XY}(\tau) + 3R_{YX}(\tau) + R_Y(\tau) \end{aligned}$$

$$\begin{aligned} R_{ZX}(\tau) &= E[Z(t)X(t+\tau)] = E[\{3X(t) + Y(t)\}X(t+\tau)] \\ &= E[\{3X(t)X(t+\tau)\} + \{Y(t)X(t+\tau)\}] \\ &= 3R_X(\tau) + R_{YX}(\tau) \end{aligned}$$

$$\begin{aligned} R_{XZ}(\tau) &= E[X(t)Z(t+\tau)] = E[X(t)\{3X(t+\tau) + Y(t+\tau)\}] \\ &= E[\{3X(t)X(t+\tau)\} + \{X(t)Y(t+\tau)\}] \\ &= 3R_X(\tau) + R_{XY}(\tau) \end{aligned}$$

$$\begin{aligned} R_{ZY}(\tau) &= E[Z(t)Y(t+\tau)] = E[\{3X(t) + Y(t)\}Y(t+\tau)] \\ &= E[\{3X(t)Y(t+\tau)\} + \{Y(t)Y(t+\tau)\}] \\ &= 3R_{XY}(\tau) + R_Y(\tau) \end{aligned}$$

$$\begin{aligned} R_{YZ}(\tau) &= E[Y(t)Z(t+\tau)] = E[Y(t)\{3X(t+\tau) + Y(t+\tau)\}] \\ &= E[\{3Y(t)X(t+\tau)\} + \{Y(t)Y(t+\tau)\}] \\ &= 3R_{YX}(\tau) + R_Y(\tau) \end{aligned}$$

28. $X(t)$ and $Y(t)$ are zero mean jointly wide-sense stationary random process. The random process $Z(t)$ is

$$Z(t) = 3X(t) + 2Y(t)$$

Find the correlations $R_Z(\tau)$, $R_{ZX}(\tau)$, $R_{XZ}(\tau)$, $R_{ZY}(\tau)$ and $R_{YZ}(\tau)$ [?]

Solution:

$$\begin{aligned} R_Z(\tau) &= E[Z(t)Z(t+\tau)] = E[\{3X(t) + 2Y(t)\}\{3X(t+\tau) + 2Y(t+\tau)\}] \\ &= E[\{9X(t)X(t+\tau)\} + \{6X(t)Y(t+\tau)\} + \{6X(t+\tau)Y(t)\} + 4\{Y(t)Y(t+\tau)\}] \\ &= 9R_X(\tau) + 6R_{XY}(\tau) + 6R_{YX}(\tau) + 4R_Y(\tau) \end{aligned}$$

$$\begin{aligned} R_{ZX}(\tau) &= E[Z(t)X(t+\tau)] = E[\{3X(t) + 2Y(t)\}X(t+\tau)] \\ &= E[\{3X(t)X(t+\tau)\} + 2\{Y(t)X(t+\tau)\}] \\ &= 3R_X(\tau) + 2R_{YX}(\tau) \end{aligned}$$

$$\begin{aligned} R_{XZ}(\tau) &= E[X(t)Z(t+\tau)] = E[X(t)\{3X(t+\tau) + 2Y(t+\tau)\}] \\ &= E[\{3X(t)X(t+\tau)\} + 2\{X(t)Y(t+\tau)\}] \\ &= 3R_X(\tau) + 2R_{XY}(\tau) \end{aligned}$$

$$\begin{aligned} R_{ZY}(\tau) &= E[Z(t)Y(t+\tau)] = E[\{3X(t) + 2Y(t)\}Y(t+\tau)] \\ &= E[\{3X(t)Y(t+\tau)\} + 2\{Y(t)Y(t+\tau)\}] \\ &= 3R_{XY}(\tau) + 2R_Y(\tau) \end{aligned}$$

$$\begin{aligned} R_{YZ}(\tau) &= E[Y(t)Z(t+\tau)] = E[Y(t)\{3X(t+\tau) + 2Y(t+\tau)\}] \\ &= E[\{6Y(t)X(t+\tau)\} + 2\{Y(t)Y(t+\tau)\}] \\ &= 6R_{YX}(\tau) + 2R_Y(\tau) \end{aligned}$$

29. $X(t)$ and $Y(t)$ are zero mean jointly wide-sense stationary random process. The random process $Z(t)$ is

$$Z(t) = X(t) + 2Y(t)$$

Find the correlations $R_Z(\tau)$, $R_{ZX}(\tau)$, $R_{XZ}(\tau)$, $R_{ZY}(\tau)$ and $R_{YZ}(\tau)$ [?]

Solution:

$$\begin{aligned} R_Z(\tau) &= E[Z(t)Z(t+\tau)] = E[\{X(t) + 2Y(t)\}\{X(t+\tau) + 2Y(t+\tau)\}] \\ &= E[\{X(t)X(t+\tau)\} + \{2X(t)Y(t+\tau)\} + \{2X(t+\tau)Y(t)\} + 4\{Y(t)Y(t+\tau)\}] \\ &= R_X(\tau) + 2R_{XY}(\tau) + 2R_{YX}(\tau) + 4R_Y(\tau) \end{aligned}$$

$$\begin{aligned} R_{ZX}(\tau) &= E[Z(t)X(t+\tau)] = E[\{X(t) + 2Y(t)\}X(t+\tau)] \\ &= E[\{X(t)X(t+\tau)\} + 2\{Y(t)X(t+\tau)\}] \\ &= R_X(\tau) + 2R_{YX}(\tau) \end{aligned}$$

$$\begin{aligned}
R_{XZ}(\tau) &= E[X(t)Z(t+\tau)] = E[X(t)\{X(t+\tau) + 2Y(t+\tau)\}] \\
&= E[\{X(t)X(t+\tau)\} + 2\{X(t)Y(t+\tau)\}] \\
&= R_X(\tau) + 2R_{XY}(\tau)
\end{aligned}$$

$$\begin{aligned}
R_{ZY}(\tau) &= E[Z(t)Y(t+\tau)] = E[\{X(t) + 2Y(t)\}Y(t+\tau)] \\
&= E[\{X(t)Y(t+\tau)\} + 2\{Y(t)Y(t+\tau)\}] \\
&= R_{XY}(\tau) + 2R_Y(\tau)
\end{aligned}$$

$$\begin{aligned}
R_{YZ}(\tau) &= E[Y(t)Z(t+\tau)] = E[Y(t)\{X(t+\tau) + 2Y(t+\tau)\}] \\
&= E[\{Y(t)X(t+\tau)\} + 2\{Y(t)Y(t+\tau)\}] \\
&= R_{YX}(\tau) + 2R_Y(\tau)
\end{aligned}$$

1.4.4 Multiplication of Random Process:

Consider the multiplication of two random independent, jointly wide-sense stationary random process $X(t)$ and $Y(t)$ is

$$W(t) = X(t)Y(t)$$

The mean of $W(t)$ is

$$E[W(t)] = \mu_W = E[X(t)]E[Y(t)] = \mu_X\mu_Y$$

The variance of $W(t)$ is

$$\begin{aligned}
\sigma_W^2 &= E[X(t)]E[Y(t) - \mu_X\mu_Y]^2 \\
&= (\sigma_X^2 + \mu_X^2)(\sigma_Y^2 + \mu_Y^2)\mu_X^2\mu_Y^2
\end{aligned}$$

The autocorrelation function of the sum $W(t)$ is

$$\begin{aligned}
R_W(\tau) &= E[\{X(t)Y(t)\}\{X(t+\tau) + Y(t+\tau)\}] \\
&= E[X(t)X(t+\tau)]E[Y(t)Y(t+\tau)] \\
&= R_X(\tau)R_Y(\tau)
\end{aligned}$$

30. The random process $X(t)$ is noise with the autocorrelation function

$$R_X(\tau) = \begin{cases} 10(1 - |\tau|/\tau_N) & -\tau_N \leq \tau \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is independent of $X(t)$ and has the autocorrelation function

$$R_Y(\tau) = \begin{cases} 15(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

where $T \gg \tau_N$. The random process $Z(t) = X(t) \times Y(t)$. For $Z(t)$, find the autocorrelation function, its total power, its dc power, and its ac power. Is $Z(t)$ wide-sense stationary? [?]

Solution:

$$\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 15, \mu_Y = 0, \text{ Autocorrelation function } R_Z(\tau)$$

$$R_Z(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned}
R_Z(\tau) &= 15(1 - |\tau|/T) \times 10(1 - |\tau|/\tau_N) \\
&= 150(1 - |\tau|/T)(1 - |\tau|/\tau_N)
\end{aligned}$$

$$R_Z(\tau) = \begin{cases} 150(1 - |\tau|/T)(1 - |\tau|/\tau_N) & |\tau| \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is 150

31. The random process $X(t)$ is noise with the autocorrelation function

$$R_X(\tau) = \begin{cases} 10(1 - |\tau|/\tau_N) & -\tau_N \leq \tau \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is independent of $X(t)$ and has the autocorrelation function

$$R_Y(\tau) = 18\cos(\omega_c\tau) \quad -\infty \leq \tau \leq \infty$$

where $2\pi/\omega_c \gg \tau_N$. The random process $Z(t) = X(t) \times Y(t)$. For $Z(t)$, find the autocorrelation function, its total power, its dc power, and its ac power. Is $Z(t)$ wide-sense stationary? [?]

Solution:

$$\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 18, \mu_Y = 0, \text{ Autocorrelation function } R_Z(\tau)$$

$$R_Z(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned}
R_Z(\tau) &= 15(1 - |\tau|/T) \times 18\cos(\omega_c\tau) \\
&= 180(1 - |\tau|/T)(\cos(\omega_c\tau))
\end{aligned}$$

$$R_Z(\tau) = \begin{cases} 180(\cos(\omega_c\tau))(1 - |\tau|/T) & |\tau| \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is 180

32. The random process $X(t)$ is noise with the autocorrelation function

$$R_X(\tau) = \begin{cases} 10(1 - |\tau|/\tau_N) & -\tau_N \leq \tau \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

The random process $Y(t)$ is independent of $X(t)$ and has the autocorrelation function

$$R_Y(\tau) = 13 \frac{\sin(\omega_B \tau)}{(\omega_B \tau)} \quad -\infty \leq \tau \leq \infty$$

where $2\pi/\omega_B \gg \tau_N$. The random process $Z(t) = X(t) \times Y(t)$. For $Z(t)$, find the autocorrelation function, its total power, its dc power, and its ac power. Is $Z(t)$ wide-sense stationary? [?]

Solution:

$\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 13, \mu_Y = 0$, Autocorrelation function $R_Z(\tau)$

$$R_Z(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned} R_Z(\tau) &= 10(1 - |\tau|/T) \times 13 \frac{\sin(\omega_B \tau)}{(\omega_B \tau)} \\ &= 130(1 - |\tau|/T) \frac{\sin(\omega_B \tau)}{(\omega_B \tau)} \end{aligned}$$

$$R_Z(\tau) = \begin{cases} 130 \frac{\sin(\omega_B \tau)}{(\omega_B \tau)} (1 - |\tau|/T) & |\tau| \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_Z(\tau) = 0$
2. $R_Z(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_Z(\tau)$ is 130

33. The random process $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. Given that $W(t) = X(t) \times Y(t)$ and

$$R_X(\tau) = 10 \exp\left(-\frac{|\tau|}{3}\right) \quad -\infty \leq \tau \leq \infty$$

$$R_Y(\tau) = \begin{cases} 11 \left(\frac{3-|\tau|}{3}\right) & -3 \leq \tau \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

For $W(t)$ find its autocorrelation function, its total power, its dc power, and its ac power. Is $W(t)$ wide-sense stationary? [?]

Solution:

$\sigma_X^2 = 10, \mu_X = 0, \sigma_Y^2 = 11, \mu_Y = 0$, Autocorrelation function $R_W(\tau)$

$$R_W(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned} R_W(\tau) &= 10 \exp\left(-\frac{|\tau|}{3}\right) \times 11 \left(\frac{3-|\tau|}{3}\right) \\ &= 110 \exp\left(-\frac{|\tau|}{3}\right) \left(\frac{3-|\tau|}{3}\right) \end{aligned}$$

$$R_W(\tau) = \begin{cases} 110 \exp\left(-\frac{|\tau|}{3}\right) \left(\frac{3-|\tau|}{3}\right) & |\tau| \leq \tau_N \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_W(\tau) = 0$
2. $R_W(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_W(\tau)$ is 110

34. The random process $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. Given that $W(t) = X(t) \times Y(t)$ and

$$R_X(\tau) = 12 \exp\left(-\frac{|\tau|}{4}\right) \quad -\infty \leq \tau \leq \infty$$

$$R_Y(\tau) = \begin{cases} 10 \left(\frac{4-|\tau|}{4}\right) & -4 \leq \tau \leq 4 \\ 0 & \text{otherwise} \end{cases}$$

For $W(t)$ find its autocorrelation function, its total power, its dc power, and its ac power. Is $W(t)$ wide-sense stationary? [?]

Solution:

$$\sigma_X^2 = 12, \mu_X = 0, \sigma_Y^2 = 10, \mu_Y = 0, \text{ Autocorrelation function } R_W(\tau)$$

$$R_W(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned} R_W(\tau) &= 12 \exp\left(-\frac{|\tau|}{4}\right) \times 10 \left(\frac{4-|\tau|}{4}\right) \\ &= 120 \exp\left(-\frac{|\tau|}{4}\right) \left(\frac{4-|\tau|}{4}\right) \end{aligned}$$

$$R_W(\tau) = \begin{cases} 120 \exp\left(-\frac{|\tau|}{4}\right) \left(\frac{4-|\tau|}{4}\right) & |\tau| \leq 4 \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_W(\tau) = 0$
2. $R_W(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_W(\tau)$ is 120

35. The random process $X(t)$ and $Y(t)$ are jointly wide-sense stationary and they are independent. Given that $W(t) = X(t) \times Y(t)$ and

$$R_X(\tau) = 11 \exp\left(-\frac{|\tau|}{5}\right) \quad -\infty \leq \tau \leq \infty$$

$$R_Y(\tau) = \begin{cases} 12 \left(\frac{5-|\tau|}{5} \right) & -5 \leq \tau \leq 5 \\ 0 & \text{otherwise} \end{cases}$$

For $W(t)$ find its autocorrelation function, its total power, its dc power, and its ac power. Is $W(t)$ wide-sense stationary? [?]

Solution:

$$\sigma_X^2 = 11, \mu_X = 0, \sigma_Y^2 = 12, \mu_Y = 0, \text{ Autocorrelation function } R_W(\tau)$$

$$R_W(\tau) = R_X(\tau) \times R_Y(\tau)$$

$$\begin{aligned} R_W(\tau) &= 11 \exp\left(-\frac{|\tau|}{5}\right) \times 12 \left(\frac{5-|\tau|}{5}\right) \\ &= 132 \exp\left(-\frac{|\tau|}{5}\right) \left(\frac{5-|\tau|}{5}\right) \end{aligned}$$

$$R_W(\tau) = \begin{cases} 132 \exp\left(-\frac{|\tau|}{5}\right) \left(\frac{5-|\tau|}{5}\right) & |\tau| \leq 5 \\ 0 & \text{otherwise} \end{cases}$$

From the above discussions it is observed that

1. The mean of $R_W(\tau) = 0$
2. $R_W(\tau)$ is independent of absolute time, hence it is wide-sense stationary.

Total power or AC power of $R_W(\tau)$ is 132

1.5 Ergodic Random Processes

Any random process is wide-sense stationarity if it satisfies the conditions like mean and autocorrelation function are independent of time. To calculate mean and autocorrelation function of a random process, it requires an ensemble of sample functions (data records). It is difficult to collect the data in real time situations. In many real-life applications Its convenient to calculate the averages from a single data record. This is possible in certain random processes called ergodic processes.

A random process is ergodic if time averages with a sample function equal to ensemble. The avaragieng of random process is defined as

$$\langle x(t) \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t) dt$$

The autocorrelation function for random process $x(t)$ is defined as

$$\langle x(t)x(t+\tau) \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t)x(t+\tau) dt$$

36. Assume that the data in the following table are obtained from a windowed sample function obtained from an ergodic random process. Use(4.73) to estimate the autocorrelation for $\tau = 0, 2$ and 4 ms, where $\Delta t = 2$ ms

$x(t)$	1.5	2.1	1.0	2.2	-1.6	-2.0	-2.5	2.5	1.6	-1.8
k	0	1	2	3	4	5	6	7	8	9

Solution:

Autocorrelation function for discrete sequences is

$$r_X(i) = \frac{1}{n} \sum_{k=0}^{k=N-1-i} x(k)x(k+i)$$

It is given that $\tau = 0, 2$ and 4 ms, where $\Delta t = 2$ ms, when $\tau = 0$ it is an autocorrelation when $\tau = 2$ and $\Delta t = 2$ ms $i = 2/2 = 1$, $\tau = 4$ and $\Delta t = 2$ ms $i = 4/2 = 2$

$$\begin{aligned}
 r_X(i) &= \frac{1}{n} \sum_{k=0}^{k=N-1-i} x(k)x(k+i) \\
 r_X(0) &= \frac{1}{n} \sum_{k=0}^{k=N-1} x(k)x(k) = \frac{1}{n} [x(k)^2] \\
 &= \frac{1}{10} [(1.5)^2 + (2.1)^2 + (1.0)^2 + (2.2)^2 + (-1.6)^2 + (-2.0)^2 + (-2.5)^2 + (2.5)^2 + (1.6)^2 + (-1.8)^2] \\
 &= 3.736 \\
 r_X(2 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{k=N-2} x(k)x(k+1) \\
 &= \frac{1}{10} [(1.5)(2.1) + (2.1)(1.0) + (1.0)(2.2) + (2.2)(-1.6) + (-1.6)(-2.0) + (-2.0)(-2.5) \\
 &\quad + (-2.5)(2.5) + (2.5)(1.6) + (1.6)(-1.8)] \\
 &= \frac{1}{10} [3.15 + 2.1 + 2.2 - 3.52 + 3.2 + 5 - 6.25 + 4 - 2.88] \\
 &= -0.700
 \end{aligned}$$

$$\begin{aligned}
r_X(4 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{N-3} x(k)x(k+2) \\
&= \frac{1}{10} [(1.5)(1.0) + (2.1)(2.2) + (1.0)(-1.6) + (2.2)(-2.0) + (-1.6)(-2.5) + (-2.0)(2.5) \\
&\quad + (-2.5)(1.6) + (2.5)(-1.8)] \\
&= -0.938
\end{aligned}$$

37. Assume that the data in the following table are obtained from a windowed sample function obtained from an ergodic random process. Use(4.73) to estimate the autocorrelation for $\tau = 0, 3$ and 6 ms , where $\Delta t = 3 \text{ ms}$

x(t)	1.0	2.2	1.5	-3.0	-0.5	1.7	-3.5	-1.5	1.6	-1.3
k	0	1	2	3	4	5	6	7	8	9

Solution:

Autocorrelation function for discrete sequences is

$$r_X(i) = \frac{1}{n} \sum_{k=0}^{N-1-i} x(k)x(k+i)$$

It is given that $\tau = 0, 3$ and 6 ms , where $\Delta t = 3 \text{ ms}$, when $\tau = 0$ it is an autocorrelation when $\tau = 3$ and $\Delta t = 3 \text{ ms}$ $i = 3/3 = 1$, $\tau = 6$ and $\Delta t = 3 \text{ ms}$ $i = 6/3 = 2$

$$\begin{aligned}
r_X(i) &= \frac{1}{n} \sum_{k=0}^{N-1-i} x(k)x(k+i) \\
r_X(0) &= \frac{1}{n} \sum_{k=0}^{N-1} x(k)x(k) = \frac{1}{n} [x(k)^2] \\
&= \frac{1}{10} [(1.0)^2 + (2.2)^2 + (1.5)^2 + (-3.0)^2 + (-0.5)^2 + (1.7)^2 + (-3.5)^2 + (-1.5)^2 + (1.6)^2 + (-1.3)^2] \\
&= 3.898
\end{aligned}$$

$$\begin{aligned}
r_X(3 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{N-2} x(k)x(k+1) \\
&= \frac{1}{10} [(1.0)(2.2) + (2.2)(1.5) + (1.5)(-3.0) + (-3.0)(-0.5) + (-0.5)(1.7) + (1.7)(-3.5) \\
&\quad + (-3.5)(-1.5) + (-1.5)(1.6) + (1.6)(-1.3)] \\
&= -0.353
\end{aligned}$$

$$\begin{aligned}
r_X(6 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{N-3} x(k)x(k+2) \\
&= \frac{1}{10} [(1.0)(1.5) + (2.2)(-3.0) + (1.5)(-0.5) + (-3.0)(1.7) + (-0.5)(-3.5) + (1.7)(-1.5) \\
&\quad + (-3.5)(1.6) + (-1.5)(-1.3)] \\
&= -1.540
\end{aligned}$$

38. Assume that the data in the following table are obtained from a windowed sample function obtained from an ergodic random process. Use(4.73) to estimate the autocorrelation for $\tau = 0, 7$ and 14 ms, where $\Delta t = 7$ ms

x(t)	1.5	0.4	0.8	0.3	-0.4	-1.7	2.0	-2.0	0.8	-0.2
k	0	1	2	3	4	5	6	7	8	9

Solution:

Autocorrelation function for discrete sequences is

$$r_X(i) = \frac{1}{n} \sum_{k=0}^{k=N-1-i} x(k)x(k+i)$$

It is given that $\tau = 0, 2$ and 4 ms, where $\Delta t = 2$ ms, when $\tau = 0$ it is an autocorrelation when $\tau = 7$ and $\Delta t = 7$ ms $i = 7/7 = 1$, $\tau = 14$ and $\Delta t = 7$ ms $i = 14/7 = 2$

$$\begin{aligned} r_X(i) &= \frac{1}{n} \sum_{k=0}^{k=N-1-i} x(k)x(k+i) \\ r_X(0) &= \frac{1}{n} \sum_{k=0}^{k=N-1} x(k)x(k) = \frac{1}{n} [x(k)^2] \\ &= \frac{1}{10} [(1.5)^2 + (0.4)^2 + (0.8)^2 + (0.3)^2 + (-0.4)^2 + (-1.7)^2 + (2.0)^2 + (-2.0)^2 + (0.8)^2 + (-0.2)^2] \\ &= 1.487 \end{aligned}$$

$$\begin{aligned} r_X(7 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{k=N-1} x(k)x(k+1) \\ &= \frac{1}{10} [(1.5)(0.4) + (0.4)(0.8) + (0.8)(0.3) + (0.3)(-0.4) + (-0.4)(-1.7) + (-1.7)(2.0) \\ &\quad + (2.0)(-2.0) + (-2.0)(0.8) + (0.8)(-0.2)] \\ &= -0.744 \\ r_X(14 \text{ ms}) &= \frac{1}{n} \sum_{k=0}^{k=N-2} x(k)x(k+2) \\ &= \frac{1}{10} [(1.5)(0.8) + (0.4)(0.3) + (0.8)(-0.4) + (0.3)(-1.7) + (-0.4)(2.0) + (-1.7)(-2.0) \\ &\quad + (2.0)(0.8) + (-2.0)(-0.2)] \\ &= 0.509 \end{aligned}$$

1.6 Power Spectral Densities

The power spectral density of a wide-sense stationary random process is the Fourier transform of the autocorrelation function. The distribution of power over range frequencies is described by power power density spectrum.

Fourier transform pair is

$$\begin{aligned} X_T(j\omega) &= \int_{-\infty}^{\infty} x_T(t) e^{-j\omega t} dt \\ x_T(t) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} X_T(j\omega) e^{j\omega t} d\omega \end{aligned}$$

The power of signal is its square. Averaging the power of the signal $x(t)$ within a window of width T is

$$\begin{aligned} P_T &= \frac{1}{T} \int_{-T/2}^{T/2} x^2(t) dt \\ &= \frac{1}{T} \int_{-\infty}^{\infty} x_T^2(t) dt \\ &= \frac{1}{T} \int_{-\infty}^{\infty} x_T(t) \frac{1}{2\pi} \int_{-\infty}^{\infty} x_T(t) e^{j\omega t} d\omega dt \\ &= \frac{1}{2\pi T} \int_{-\infty}^{\infty} x_T(j\omega) \frac{1}{2\pi} \int_{-\infty}^{\infty} x_T(t) e^{j\omega t} dt d\omega \\ &= \frac{1}{2\pi T} \int_{-\infty}^{\infty} x_T(j\omega) x_T(-j\omega) d\omega \end{aligned}$$

$$x_T(-j\omega) = X_T^*(j\omega)$$

$$\begin{aligned} P_T &= \frac{1}{T} \int_{-T/2}^{T/2} X^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X_T(j\omega)|^2 d\omega \\ \int_{-T/2}^{T/2} X^2(t) dt &= \frac{1}{2\pi} \int_{-\infty}^{\infty} |X_T(j\omega)|^2 d\omega \end{aligned}$$

Expectation of the above equation is

$$\begin{aligned} E \left[\int_{-T/2}^{T/2} X^2(t) dt \right] &= \frac{1}{2\pi} E \left[\int_{-\infty}^{\infty} |X_T(j\omega)|^2 d\omega \right] \\ \int_{-T/2}^{T/2} E[X^2(t)] dt &= \frac{1}{2\pi} \int_{-\infty}^{\infty} E[|X_T(j\omega)|^2] d\omega \\ TE[X^2(t)] &= \frac{1}{2\pi} \int_{-\infty}^{\infty} E[|X_T(j\omega)|^2] d\omega \\ TR_X(0) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} E[|X_T(j\omega)|^2] d\omega \\ R_X(0) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \lim_{T \rightarrow \infty} E[|X_T(j\omega)|^2] d\omega \end{aligned}$$

The above equation can be split into two parts as

$$\begin{aligned} R_X(0) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\ S_X(\omega) &= \lim_{T \rightarrow \infty} E[|X_T(j\omega)|^2] \end{aligned}$$

39. A PSD is as shown in Figure 1.2 where the constants are $a = 55$, $b = 5$, $\omega_o = 1000$ and $\omega_1 = 100$. Calculate values for $E[X^2(t)]$, the σ_X^2 and $|\mu_X|$.

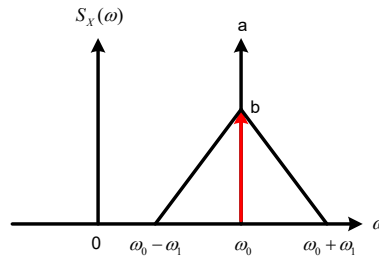


Figure 1.2

Solution:

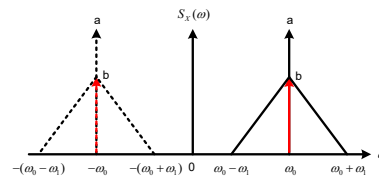


Figure 1.3

$$\begin{aligned}\omega_o - \omega_1 &= 1000 - 100 = 900 \\ \omega_o + \omega_1 &= 1000 + 100 = 1100 \\ (\omega_o + \omega_1) - (\omega_o - \omega_1) &= 1100 - 900 = 200\end{aligned}$$

The given spectrum is in the triangular form, its base is 200 and its height is 5 and it also has $a = 55$. The modified two sided spectrum is redrawn and is as shown in Figure 1.3. The x axis is ω radians/sec. The Autocorrelation function is

$$\begin{aligned}R_X(0) &= E[X^2(t)] = \text{Area under PSD} = \text{Area of impulse} + \text{Area of Triangle} \\ &= 2 \times \frac{1}{2\pi} [\text{Area of impulse} + \text{Area of Triangle}] \\ &= \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\ &= \frac{1}{2\pi} \left(55 + \frac{1}{2}(5)(200) \right) \\ &= 176.6620 \\ |\mu_X| &= 0\end{aligned}$$

40. A PSD is as shown in Figure 1.2 where the constants are $a = 450$, $b = 6$, $\omega_o = 10,000$ and $\omega_1 = 1000$. Calculate values for $E[X^2(t)]$, the σ_X^2 and $|\mu_X|$.

Solution:

The given spectrum has the following details constant $a = 450$

$$\begin{aligned}\omega_o - \omega_1 &= 10,000 - 1000 = 9900 \\ \omega_o + \omega_1 &= 10000 + 1000 = 11000 \\ (\omega_o + \omega_1) - (\omega_o - \omega_1) &= 11000 - 9000 = 2000\end{aligned}$$

The signal magnitude is $b = 6$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= E[X^2(t)] = \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} \left(450 + \frac{1}{2}(6)(2000) \right) \\
 &= 2053.0988 \\
 |\mu_X| &= 0
 \end{aligned}$$

41. A PSD is as shown in Figure 1.2 where the constants are $a = 72$, $b = 4$, $\omega_o = 1000$ and $\omega_1 = 50$. Calculate values for $E[X^2(t)]$, the σ_X^2 and $|\mu_X|$.

Solution:

The given spectrum has the following details constant $a = 72$

$$\begin{aligned}
 \omega_o - \omega_1 &= 1000 - 50 = 950 \\
 \omega_o + \omega_1 &= 1000 + 50 = 1050 \\
 (\omega_o + \omega_1) - (\omega_o - \omega_1) &= 1050 - 950 = 100
 \end{aligned}$$

The signal magnitude is $b = 4$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= E[X^2(t)] = \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} \left(72 + \frac{1}{2}(4)(100) \right) \\
 &= 86.5803 \\
 |\mu_X| &= 0
 \end{aligned}$$

42. A PSD is as shown in Figure 1.4 where the constants are $a = 3$, $b = 5$, $\omega_1 = 8$ and $\omega_2 = 12$. Calculate values for $R_X(0)$, the variance and μ_X^2 .

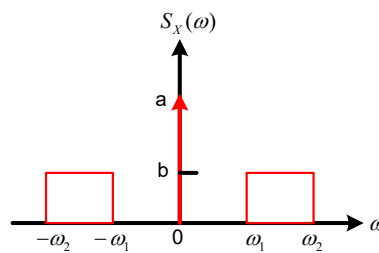


Figure 1.4

Solution:

$$\begin{aligned}
 \omega_1 &= 8 \\
 \omega_2 &= 12 \\
 \omega_2 - \omega_1 &= 12 - 8 = 4
 \end{aligned}$$

The given spectrum is in the rectangular form, its base is 4 and its height is 5 and it also has $a = 3$ at $\omega = 0$. The x axis is ω radians/sec. The Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= E[X^2(t)] = \text{Area under PSD} = \text{Area of impulse} + \text{Area of Rectangle} \\
 &= \frac{1}{2\pi} [\text{Area of impulse} + \text{Area of Rectangle}] \\
 &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (3 + 2[(5)(12 - 8)]) \\
 &= 6.8437 \\
 \mu_X^2 &= \frac{3}{2\pi} = 0.4775 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 6.8437 - 0.4775 \\
 &= 6.3662
 \end{aligned}$$

43. A PSD is as shown in Figure 1.4 where the constants are $a = 5$, $b = 3$, $\omega_1 = 7$ and $\omega_2 = 13$. Calculate values for $R_X(0)$, the variance and μ_X^2 .

Solution:

The given spectrum has the following details constant $a = 5$

$$\begin{aligned}
 \omega_1 &= 7 \\
 \omega_2 &= 13 \\
 \omega_2 - \omega_1 &= 13 - 7 = 6
 \end{aligned}$$

The signal magnitude is $b = 3$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (5 + 2(3)(13 - 7)) \\
 &= 6.5254 \\
 \mu_X^2 &= \frac{3}{2\pi} = 0.7958 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 6.5254 - 0.7958 \\
 &= 5.7296
 \end{aligned}$$

44. A PSD is as shown in Figure 1.4 where the constants are $a = 4$, $b = 3$, $\omega_1 = 9$ and $\omega_2 = 14$. Calculate values for $R_X(0)$, the variance and μ_X^2 .

Solution:

The given spectrum has the following details constant $a = 4$

$$\begin{aligned}
 \omega_1 &= 9 \\
 \omega_2 &= 14 \\
 \omega_2 - \omega_1 &= 14 - 9 = 5
 \end{aligned}$$

The signal magnitude is $b = 3$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (4 + 2(3)(14 - 9)) \\
 &= 5.4113 \\
 \mu_X^2 &= \frac{3}{2\pi} = 0.6366 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 5.4113 - 0.6366 \\
 &= 4.7746
 \end{aligned}$$

45. A PSD is as shown in Figure 1.5 where the constants are $a = 5$, $\omega_o = 100$ and $w = 8$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

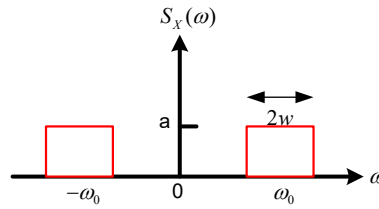


Figure 1.5

Solution:

The given spectrum has the following details

$$\begin{aligned}
 w &= 8 \\
 2w &= 16
 \end{aligned}$$

The signal magnitude is $a = 5$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{2}{2\pi} (5(16)) \\
 &= 25.4648 \\
 \mu_X &= 0
 \end{aligned}$$

46. A PSD is as shown in Figure 1.5 where the constants are $a = 3$, $\omega_o = 150$ and $w = 7$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

Solution:

The given spectrum has the following details

$$\begin{aligned}
 w &= 7 \\
 2w &= 14
 \end{aligned}$$

The signal magnitude is $a = 3$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{2}{2\pi} (3(14)) \\
 &= 13.3690 \\
 \mu_X &= 0
 \end{aligned}$$

47. A PSD is as shown in Figure 1.6 where the constants are $a = 4$, $\omega_o = 125$ and $w = 6$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

Solution:

The given spectrum has the following details

$$\begin{aligned}
 w &= 6 \\
 2w &= 12
 \end{aligned}$$

The signal magnitude is $a = 3$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{2}{2\pi} (4(12)) \\
 &= 15.2789 \\
 \mu_X &= 0
 \end{aligned}$$

48. A PSD is as shown in Figure 1.6 where the constants are $a = 300$, $b = 10$, $\omega_M = 100$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

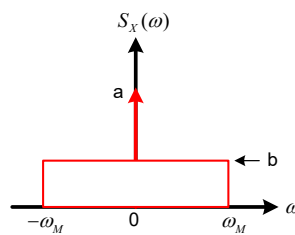


Figure 1.6

Solution:

The given spectrum has the following details

$$-\omega_M + \omega_M = 100 + 100 = 200$$

The signal magnitude is $a = 300$, $b = 10$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (300 + (10)(200)) \\
 \sigma_X^2 &= \frac{2000}{2\pi} = 318.3099 \\
 \mu_X^2 &= \frac{300}{2\pi} = 47.7465 \\
 |\mu_X| &= 6.9099 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 318.3099 - 47.7465 = 366.0564
 \end{aligned}$$

49. A PSD is as shown in Figure 1.6 where the constants are $a = 200$, $b = 20$, $\omega_M = 80$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

Solution:

The given spectrum has the following details

$$-\omega_M + \omega_M = 80 + 80 = 160$$

The signal magnitude is $a = 200$, $b = 20$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (200 + (20)(160)) \\
 &= 541.1268 \\
 \mu_X^2 &= \frac{300}{2\pi} = 47.7465 \\
 |\mu_X| &= 6.9099 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 541.1268 - 31.8310 = 509.2958
 \end{aligned}$$

50. A PSD is as shown in Figure 1.6 where the constants are $a = 300$, $b = 15$, $\omega_M = 75$. Calculate values for $R_X(0)$, the variance and $|\mu_X|$.

Solution:

The given spectrum has the following details

$$-\omega_M + \omega_M = 75 + 75 = 150$$

The signal magnitude is $a = 300$, $b = 15$

Autocorrelation function is

$$\begin{aligned}
 R_X(0) &= \mu_X^2 + \sigma_X^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) d\omega \\
 &= \frac{1}{2\pi} (300 + (15)(150)) \\
 &= 405.8451 \\
 \mu_X^2 &= \frac{300}{2\pi} = 47.7465 \\
 |\mu_X| &= 6.9099 \\
 \sigma_X^2 &= R_X(0) - \mu_X^2 = 405.8451 - 47.7465 = 358.0986
 \end{aligned}$$

	$R_X(\tau)$	$S_X(\omega)$
1	$\cos\omega_c\tau$	$\pi[\delta(\omega - \omega_c) + \delta(\omega + \omega_c)]$
2	$\delta(\tau)$	1
3	$\exp(-a \tau) \ a > 0$	$\frac{2a}{\omega^2 + a^2}$
4	$\begin{cases} T(1 - \tau /T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$	$T^2 \left(\frac{\sin(\omega T/2)}{\omega T/2} \right)^2$
5	$(\omega_B/\pi) \frac{\sin(\omega_B\tau)}{\omega_B\tau}$	$\begin{cases} 1 & -\omega_B \leq \omega \leq \omega_B \\ 0 & \text{otherwise} \end{cases}$

1.7 Wiener-Khinchin Relations

Weiner-Khinchin Relation states that the PSD and the autocorrelation function for wide-sense stationary random process are Fourier transform pair:

$$S_X(\omega) = \int_{-\infty}^{\infty} R_X(\tau) e^{-j\omega\tau} d\tau$$

$$R_X(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) e^{j\omega\tau} d\omega$$

Let a rectangular function of time τ be

$$\begin{cases} T(1 - |\tau|/T) & -T \leq \tau \leq T \\ 0 & \text{otherwise} \end{cases}$$

51. Use the property of an integrated unit-impulse function. (E.3; see Appendix E) to verify the property item 1 in table 4.1.

Solution:

$$S_X(\omega) = \pi[\delta(\omega - \omega_c) + \delta(\omega + \omega_c)]$$

Show that

$$R_X(\tau) = \cos\omega_c\tau$$

It is given that

$$\int_{-\infty}^{\infty} g(x) \delta(x - x_0) dx = g(x_0)$$

Based on the above relation

$$\begin{aligned} \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) e^{j\omega\tau} d\omega &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \pi[\delta(\omega - \omega_c) + \delta(\omega + \omega_c)] e^{j\omega\tau} d\omega \\ &= \frac{1}{2} [e^{j\omega_c\tau} + e^{-j\omega_c\tau}] \\ &= \cos\omega_c\tau \end{aligned}$$

52. Use the property of an integrated unit-impulse function. (E.3; see Appendix E) to verify the property item 2 in table 4.1.

Solution:

$$S_X(\omega) = 1$$

Show that

$$R_X(\tau) = \delta(\tau)$$

It is given that

$$\int_{-\infty}^{\infty} g(x)\delta(x - x_0)dx = g(x_0)$$

Based on the above relation

$$\begin{aligned} \int_{-\infty}^{\infty} R_X(\tau)e^{-j\omega\tau}d\tau &= \int_{-\infty}^{\infty} \delta(\tau)e^{-j\omega\tau}d\tau = e^0 \\ &= 1 \end{aligned}$$

53. Use the property of an integrated unit-impulse function. (E.3; see Appendix E) to verify the property item 3 in table 4.1.

$$R_X(\tau) = \exp(-a|\tau|) \quad a > 0$$

Show that

$$S_X(\omega) = \frac{2a}{\omega^2 + a^2}$$

Solution: It is given that

$$\int_{-\infty}^{\infty} g(x)\delta(x - x_0)dx = g(x_0)$$

Based on the above relation

$$\begin{aligned} S_X(\omega) &= \int_{-\infty}^0 e^{-a\tau}e^{-j\omega\tau}d\tau + \int_{-\infty}^{\infty} e^{a\tau}e^{-j\omega\tau}d\tau \\ &= \int_{-\infty}^0 e^{(a-j\omega)\tau}d\tau + \int_0^{\infty} e^{-(a+j\omega)\tau}d\tau \\ &= \left[\frac{e^{(a-j\omega)\tau}}{(a-j\omega)} \right]_{-\infty}^0 + \left[\frac{e^{-(a+j\omega)\tau}}{(a+j\omega)} \right]_0^{\infty} \\ &= \frac{1}{(a-j\omega)} - \frac{1}{(a+j\omega)} = \frac{2a}{(a^2 + \omega^2)} \end{aligned}$$

Multiplication of Two Random Process

Consider a two random process $X(t)$ and $Y(t)$ have the autocorrelation functions $R_X(\tau)$ and $R_Y(\tau)$, multiplication of two random process is

$$W(t) = X(t)Y(t)$$

The multiplication of autocorrelation functions is

$$R_W(\tau) = R_X(\tau)R_Y(\tau)$$

$$\begin{aligned} S_W(\omega) &= \int_{-\infty}^{\infty} R_X(\tau)R_Y(\tau)d\tau \\ &= \int_{-\infty}^{\infty} R_X(\tau) \left(\frac{1}{2\pi} \int_{-\infty}^{\infty} S_Y(u)e^{ju\tau}du \right) e^{-j\omega\tau}d\tau \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_Y(u) \int_{-\infty}^{\infty} R_X(\tau)e^{-j\omega\tau}d\tau \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_Y(u)S_X(\omega - u)du = \frac{1}{2\pi} S_X(\omega) * S_Y(\omega) \end{aligned}$$

56. The random process $X(t)$ and $Y(t)$ have the autocorrelation functions

$$R_X(\tau) = e^{(-10|\tau|)} \quad \text{and} \quad R_Y(\tau) = 5\cos(600\tau)$$

If $Z(t) = X(t)Y(t)$, and if $X(t)$ and $Y(t)$ are independent, what is the PSD for $Z(t)$?

Solution:

$$R_X(\tau) = e^{(-10|\tau|)}$$

$$S_X(\omega) = \frac{20}{\omega^2 + 100}$$

$$R_Y(\tau) = 5\cos(600\tau)$$

$$S_Y(\omega) = 5\pi[\delta(\omega - 600) + \delta(\omega + 600)]$$

$$\begin{aligned} S_Z(\omega) &= \frac{1}{2\pi} S_X(\omega)S_Y(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 5\pi[\delta(\omega - u - 600) + \delta(\omega - u + 600)] \frac{20}{u^2 + 100} du \\ &= 50 \left(\frac{1}{(\omega - 600)^2 + 100} + \frac{1}{(\omega + 600)^2 + 100} \right) \end{aligned}$$

57. The random process $X(t)$ and $Y(t)$ have the autocorrelation functions

$$R_X(\tau) = e^{(-10|\tau|)} \quad \text{and} \quad R_Y(\tau) = 6\cos(400\tau)$$

If $Z(t) = X(t)Y(t)$, and if $X(t)$ and $Y(t)$ are independent, what is the PSD for $Z(t)$?

Solution:

$$R_X(\tau) = e^{(-10|\tau|)}$$

$$S_X(\omega) = \frac{20}{\omega^2 + 100}$$

$$R_Y(\tau) = 6\cos(400\tau)$$

$$S_Y(\omega) = 6\pi[\delta(\omega - 400) + \delta(\omega + 400)]$$

$$\begin{aligned} S_Z(\omega) &= \frac{1}{2\pi} S_X(\omega) S_Y(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 6\pi[\delta(\omega - u - 400) + \delta(\omega - u + 400)] \frac{20}{u^2 + 100} du \\ &= 60 \left(\frac{1}{(\omega - 400)^2 + 100} + \frac{1}{(\omega + 400)^2 + 100} \right) \end{aligned}$$

58. The random process $X(t)$ and $Y(t)$ have the autocorrelation functions

$$R_X(\tau) = e^{(-10|\tau|)} \quad \text{and} \quad R_Y(\tau) = 7\cos(500\tau)$$

If $Z(t) = X(t)Y(t)$, and if $X(t)$ and $Y(t)$ are independent, what is the PSD for $Z(t)$?

Solution:

$$R_X(\tau) = e^{(-10|\tau|)}$$

$$S_X(\omega) = \frac{20}{\omega^2 + 100}$$

$$R_Y(\tau) = 7\cos(500\tau)$$

$$S_Y(\omega) = 7\pi[\delta(\omega - 500) + \delta(\omega + 500)]$$

$$\begin{aligned} S_Z(\omega) &= \frac{1}{2\pi} S_X(\omega) S_Y(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 7\pi[\delta(\omega - u - 500) + \delta(\omega - u + 500)] \frac{20}{u^2 + 100} du \\ &= 70 \left(\frac{1}{(\omega - 500)^2 + 100} + \frac{1}{(\omega + 500)^2 + 100} \right) \end{aligned}$$

59. A bandlimited wide-sense stationary random process $X(t)$ has the PSD

$$S_X(\omega) = \begin{cases} 7\cos(\pi\omega/2\omega_M) & -\omega_M \leq \omega \leq \omega_M \\ 0 & \text{otherwise} \end{cases}$$

A carrier random process $C(t)$ which is independent of $X(t)$, is

$$C(t) = \sqrt{60}\cos(\omega_C t + \Theta)$$

where $\omega_C \gg \omega_M$, and where Θ is a random variable uniformly distributed between $\pm\pi$. If $Y(t) = X(t)C(t)$, what is the PSD for $Y(t)$?

Solution:

$$\begin{aligned} C(t) &= \sqrt{60}\cos(\omega_C t + \Theta) \\ R_C(\tau) &= 30\cos(\omega_C \tau) \\ S_C(\omega) &= 30\pi[\delta(\omega - \omega_C) + \delta(\omega + \omega_C)] \\ S_X(\omega) &= 7\cos(\pi\omega/2\omega_M) \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= \frac{1}{2\pi} S_X(\omega) \star S_C(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 7\cos(\pi u/2\omega_M) 30\pi[\delta(\omega - u - \omega_C) + \delta(\omega - u + \omega_C)] du \\ &= 105\cos(\pi(\omega - \omega_C)/2\omega_M) + 105\cos(\pi(\omega + \omega_C)/2\omega_M) \end{aligned}$$

60. A bandlimited wide-sense stationary random process $X(t)$ has the PSD

$$S_X(\omega) = \begin{cases} 5\cos(\pi\omega/2\omega_M) & -\omega_M \leq \omega \leq \omega_M \\ 0 & \text{otherwise} \end{cases}$$

A carrier random process $C(t)$ which is independent of $X(t)$, is

$$C(t) = \sqrt{200}\cos(\omega_C t + \Theta)$$

where $\omega_C \gg \omega_M$, and where Θ is a random variable uniformly distributed between $\pm\pi$. If $Y(t) = X(t)C(t)$, what is the PSD for $Y(t)$?

Solution:

$$\begin{aligned} C(t) &= \sqrt{200}\cos(\omega_C t + \Theta) \\ R_C(\tau) &= 100\cos(\omega_C \tau) \\ S_C(\omega) &= 100\pi[\delta(\omega - \omega_C) + \delta(\omega + \omega_C)] \\ S_X(\omega) &= 5\cos(\pi\omega/2\omega_M) \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= \frac{1}{2\pi} S_X(\omega) \star S_C(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 5\cos(\pi u/2\omega_M) 100\pi[\delta(\omega - u - \omega_C) + \delta(\omega - u + \omega_C)] du \\ &= 250\cos(\pi(\omega - \omega_C)/2\omega_M) + 250\cos(\pi(\omega + \omega_C)/2\omega_M) \end{aligned}$$

61. A bandlimited wide-sense stationary random process $X(t)$ has the PSD

$$S_X(\omega) = \begin{cases} 6\cos(\pi\omega/2\omega_M) & -\omega_M \leq \omega \leq \omega_M \\ 0 & \text{otherwise} \end{cases}$$

A carrier random process $C(t)$ which is independent of $X(t)$, is

$$C(t) = \sqrt{34}\cos(\omega_C t + \Theta)$$

where $\omega_C \gg \omega_M$, and where Θ is a random variable uniformly distributed between $\pm\pi$. If $Y(t) = X(t)C(t)$, what is the PSD for $Y(t)$?

Solution:

$$\begin{aligned} C(t) &= \sqrt{34}\cos(\omega_C t + \Theta) \\ R_C(\tau) &= 17\cos(\omega_C \tau) \\ S_C(\omega) &= 17\pi[\delta(\omega - \omega_C) + \delta(\omega + \omega_C)] \\ S_X(\omega) &= 6\cos(\pi\omega/2\omega_M) \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= \frac{1}{2\pi} S_X(\omega) \star S_C(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 6\cos(\pi u/2\omega_M) 17\pi[\delta(\omega - u - \omega_C) + \delta(\omega - u + \omega_C)] du \\ &= 51\cos(\pi(\omega - \omega_C)/2\omega_M) + 51\cos(\pi(\omega + \omega_C)/2\omega_M) \end{aligned}$$

62. The wide sense stationary random process $X(t)$ has a PSD that is a constant $5 \times 10^{-6} V^2$ when $|f| < 1 kHz$ and is 0 otherwise. The random process $Y(t) = 10\cos(\omega_o t + \Theta)$ V where $f_o = 100 kHz$ and Θ is uniformly distributed between $\pm\pi$. $X(t)$ and $Y(t)$ are independent $Z(t) = X(t)Y(t)$. What is the PSD for $Z(t)$?

Solution:

$$\begin{aligned} Y(t) &= 10\cos(\omega_o t + \Theta) \\ R_Y(\tau) &= 50\cos(\omega_o \tau) \\ S_Y(\omega) &= 50\pi[\delta(\omega - \omega_o) + \delta(\omega + \omega_o)] \end{aligned}$$

$$\begin{aligned} S_Z(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) \star S_Y(\omega) \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) S_Y(\omega - u) du \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} 5 \times 10^{-6} \times 50\pi[\delta(\omega - u - \omega_o) + \delta(\omega - u + \omega_o)] du \\ &= 125 \times 10^{-6} V^4 \quad |\pm \omega_o - \omega| < 2\pi \times 10^3 \end{aligned}$$

63. The wide sense stationary random process $X(t)$ has a PSD that is a constant $7 \times 10^{-6} V^2$ when $|f| < 0.9 kHz$ and is 0 otherwise. The random process $Y(t) = 8\cos(\omega_o t + \Theta)$ V where $f_o = 100 kHz$ and Θ is uniformly distributed between $\pm\pi$. $X(t)$ and $Y(t)$ are independent $Z(t) = X(t)Y(t)$. What is the PSD for $Z(t)$?

Solution:

$$\begin{aligned} Y(t) &= 8\cos(\omega_o t + \Theta) \\ R_Y(\tau) &= 32\cos(\omega_o \tau) \\ S_Y(\omega) &= 32\pi[\delta(\omega - \omega_o) + \delta(\omega + \omega_o)] \end{aligned}$$

$$\begin{aligned}
S_Z(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) \star S_Y(\omega) \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) S_Y(\omega - u) du \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} 7 \times 10^{-6} \times 32\pi [\delta(\omega - u - \omega_o) + \delta(\omega - u + \omega_o)] du \\
&= 112 \times 10^{-6} V^4 \quad |\pm \omega_o - \omega| < 2\pi \times 10^3
\end{aligned}$$

64. The wide sense stationary random process $X(t)$ has a PSD that is a constant $6 \times 10^{-6} V^2$ when $|f| < 1.1 kHz$ and is 0 otherwise. The random process $Y(t) = 9 \cos(\omega_o t + \Theta)$ V where $f_o = 100 kHz$ and Θ is uniformly distributed between $\pm\pi$. $X(t)$ and $Y(t)$ are independent $Z(t) = X(t)Y(t)$. What is the PSD for $Z(t)$?

Solution:

$$\begin{aligned}
Y(t) &= 9 \cos(\omega_o t + \Theta) \\
R_Y(\tau) &= 40.5 \cos(\omega_o \tau) \\
S_Y(\omega) &= 40.5\pi [\delta(\omega - \omega_o) + \delta(\omega + \omega_o)]
\end{aligned}$$

$$\begin{aligned}
S_Z(\omega) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) \star S_Y(\omega) \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(u) S_Y(\omega - u) du \\
&= \frac{1}{2\pi} \int_{-\infty}^{\infty} 6 \times 10^{-6} \times 40.5\pi [\delta(\omega - u - \omega_o) + \delta(\omega - u + \omega_o)] du \\
&= 121.5 \times 10^{-6} V^4 \quad |\pm \omega_o - \omega| < 2\pi \times 10^3
\end{aligned}$$

1.8 Linear Systems:

Consider a signal $x(t)$ is passed through a filter with $h(t)$, then the output of the filter $y(t)$ is expressed as

$$y(t) = \int_0^\infty h(u)x(t-u)du$$

where $h(t)$ is the impulse response of the system. Also above mentioned equation represents the convolution operation between input signal $x(t)$ with $h(t)$. The laplace transform the above system is

$$Y(s) = H(s)X(s)$$

where $X(s)$ and $Y(s)$ are the Laplace transforms of the input signal $x(t)$ and output $t(t)$ signals.

The above relation is applied for the random process $X(t)$ and which expressed as

$$y(t) = \int_0^\infty h(u)X(t-u)du$$

1.8.1 The mean of $Y(t)$:

The mean of the output random process is

$$\mu_Y = E[Y(t)] = \int_0^\infty h(u)E[X(t-u)]du$$

$$\mu_Y = E[Y(t)] = \mu_X \int_0^\infty h(u)du$$

The Laplace transform of $h(t)$

$$H(s) = \int_0^\infty h(t)e^{-st}dt$$

$$\mu_Y = E[Y(t)] = \mu_X H(0)$$

1.8.2 Cross-Correlating $Y(t)$ and $X(t)$:

Consider a two random process $X(t)$ and $Y(t)$ are wide sense stationary. When we consider both the random process it is called as jointly wide sense stationary. Then their cross-correlation function is defined as

$$R_{XY}(\tau) = E[X(t)Y(t+\tau)] = \int_0^\infty h(u)E[X(t)X(t+\tau-u)]du$$

$$R_{XY}(\tau) = \int_0^\infty h(u)R_X(\tau-u)du$$

$$S_{XY}(j\omega) = \int_0^\infty h(u) \int_{-\infty}^\infty R_X(\tau-u)e^{-j\omega\tau}d\tau du$$

Let $v = \tau - u$

$$\begin{aligned} S_{XY}(j\omega) &= \int_0^\infty h(u)e^{-j\omega u} \int_{-\infty}^\infty R_X(v)e^{-j\omega v}dv \\ &= H(j\omega)S_X(\omega) \end{aligned}$$

1.8.3 Autocorrelation of $Y(t)$:

$$\begin{aligned}
R_Y(\tau) &= E[Y(t)Y(t+\tau)] \\
&= \int_0^\infty h(v) \int_0^\infty h(u) E[X(t-v)X(t+\tau-u)] du dv \\
&= \int_0^\infty h(v) \int_0^\infty h(u) R_X(\tau+v-u) du dv
\end{aligned}$$

$$S_Y(\omega) = \int_0^\infty h(v) \int_0^\infty h(u) \int_{-\infty}^\infty R_X(\tau+v-u) e^{-j\omega\tau} d\tau du dv$$

Let $w = \tau + v - u$

$$\begin{aligned}
S_Y(\omega) &= \int_0^\infty h(v) e^{j\omega v} dv \int_0^\infty h(u) e^{-j\omega u} \int_{-\infty}^\infty R_X(w) e^{-j\omega w} dw \\
&= H(-j\omega) H(j\omega) S_X(\omega) \\
&= |H(j\omega)|^2 S_X(\omega)
\end{aligned}$$

65. Suppose in a given application

$$R_{XY}(\tau) = \begin{cases} aKe^{-a\tau} & \tau \geq 0 \\ 0 & \tau \leq 0 \end{cases}$$

$$S_{XY}(j\omega) = \frac{aK}{j\omega + a}$$

What are $R_{YX}(\tau)$ and $S_{YX}(j\omega)$ in this case? Interpret your results. [?]

Solution:

We Know that

$$R_{YX}(\tau) = R_{XY}(-\tau)$$

$$R_{YX}(\tau) = \begin{cases} aKe^{a\tau} & \tau \geq 0 \\ 0 & \tau \leq 0 \end{cases}$$

And also

$$S_{YX}(j\omega) = S_{YX}^*(j\omega) = \frac{aK}{-j\omega + a}$$

66. Suppose that the PSD input to a linear system is $S_X(\omega) = K$. The cross-correlation of the input $X(t)$ with the output $Y(t)$ of the linear system is found to be

$$R_{XY}(\tau) = K \begin{cases} e^{-\tau} + 3e^{-2\tau} & \tau \geq 0 \\ 0 & \tau \leq 0 \end{cases}$$

What is the power filter function $|H(j\omega)|^2$? . [?]

Solution:

If $S_X(\omega) = K$ then

$$R_X(\tau) = K\delta(\tau)$$

$$R_{XY}(\tau) = \int_0^\infty h(u)R_X(\tau - u)du$$

$$\begin{aligned} R_{XY}(\tau) &= K \int_0^\infty h(u)\delta(\tau - u)du \\ &= Kh(\tau) \end{aligned}$$

Similarly

$$h(t) = K \begin{cases} e^{-t} + 3e^{-2t} & t \geq 0 \\ 0 & t \leq 0 \end{cases}$$

$$H(s) = \frac{1}{s+1} + \frac{3}{s+2} = \frac{(s+2) + 3(s+1)}{(s+1)(s+2)} = \frac{4s+5}{s^2+3s+2}$$

$$H(j\omega) = \frac{j4\omega + 5}{-\omega^2 + 3j\omega + 2}$$

$$|H(j\omega)|^2 = \frac{16\omega^2 + 25}{\omega^4 + 5\omega^2 + 4}$$

67. Suppose that the PSD input to a linear system is $S_X(\omega) = K$. The cross-correlation of the input $X(t)$ with the output $Y(t)$ of the linear system is found to be

$$R_{XY}(\tau) = K \begin{cases} 3e^{-\tau} + e^{-2\tau} & \tau \geq 0 \\ 0 & \tau \leq 0 \end{cases}$$

What is the power filter function $|H(j\omega)|^2$? . [?]

Solution:

If $S_X(\omega) = K$ then

$$R_X(\tau) = K\delta(\tau)$$

$$R_{XY}(\tau) = \int_0^\infty h(u)R_X(\tau - u)du$$

$$\begin{aligned} R_{XY}(\tau) &= K \int_0^\infty h(u)\delta(\tau - u)du \\ &= Kh(\tau) \end{aligned}$$

Similarly

$$h(t) = K \begin{cases} 3e^{-t} + e^{-2t} & t \geq 0 \\ 0 & t \leq 0 \end{cases}$$

$$H(s) = \frac{3}{s+1} + \frac{1}{s+2} = \frac{3(s+2) + (s+1)}{(s+1)(s+2)} = \frac{4s+7}{s^2+3s+2}$$

$$H(j\omega) = \frac{j4\omega + 7}{-\omega^2 + 3j\omega + 2}$$

$$|H(j\omega)|^2 = \frac{16\omega^2 + 49}{\omega^4 + 5\omega^2 + 4}$$

68. Suppose that the PSD input to a linear system is $S_X(\omega) = K$. The cross-correlation of the input $X(t)$ with the output $Y(t)$ of the linear system is found to be

$$R_{XY}(\tau) = K \begin{cases} 2e^{-2\tau} + 3e^{-\tau} & \tau \geq 0 \\ 0 & \tau \leq 0 \end{cases}$$

What is the power filter function $|H(j\omega)|^2$? . [?]

Solution:

If $S_X(\omega) = K$ then

$$R_X(\tau) = K\delta(\tau)$$

$$R_{XY}(\tau) = \int_0^\infty h(u)R_X(\tau - u)du$$

$$\begin{aligned} R_{XY}(\tau) &= K \int_0^\infty h(u)\delta(\tau - u)du \\ &= Kh(\tau) \end{aligned}$$

Similarly

$$h(t) = K \begin{cases} 2e^{-t} + 3e^{-t} & t \geq 0 \\ 0 & t \leq 0 \end{cases}$$

$$H(s) = \frac{2}{s+2} + \frac{3}{s+1} = \frac{2(s+1) + 3(s+2)}{(s+1)(s+2)} = \frac{5s+8}{s^2+3s+2}$$

$$H(j\omega) = \frac{j5\omega + 8}{-\omega^2 + 3j\omega + 2}$$

$$|H(j\omega)|^2 = \frac{25\omega^2 + 64}{\omega^4 + 5\omega^2 + 4}$$

69. The PSD of the random process $X(t)$ and the transfer function of a network are

$$S_X(\omega) = \frac{1}{\omega^2 + (100)^2} \quad \text{and} \quad H(s) = \frac{s}{(s+10)(s+9000)}$$

$Y(s) = H(s)X(s)$. Find μ_Y , $S_{XY}(j\omega)$ and $S_Y(\omega)$. [?]

Solution:

$S_X(\omega)$ doesn't have any dc component (unit impulse function) when $\omega = 0$, hence

$$\mu_X = 0$$

Also the relation is

$$\mu_Y = \mu_X H(0) = 0$$

$$\begin{aligned} S_{XY}(j\omega) &= H(j\omega)S_X(\omega) \\ &= \frac{j\omega}{(\omega^2 + (100)^2)(j\omega + 10)(j\omega + 9000)} \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= |H(j\omega)|^2 S_X(\omega) \\ &= \frac{\omega^2}{(\omega^2 + (100)^2)(\omega^2 + (10)^2)(\omega^2 + (9000)^2)} \end{aligned}$$

70. The PSD of the random process $X(t)$ and the transfer function of a network are

$$S_X(\omega) = \frac{1}{\omega^2 + (80)^2} \quad \text{and} \quad H(s) = \frac{s}{(s+9)(s+10000)}$$

$Y(s) = H(s)X(s)$. Find μ_Y , $S_{XY}(j\omega)$ and $S_Y(\omega)$. [?]

Solution:

$S_X(\omega)$ doesn't have any dc component (unit impulse function) when $\omega = 0$, hence

$$\mu_X = 0$$

Also the relation is

$$\mu_Y = \mu_X H(0) = 0$$

$$\begin{aligned} S_{XY}(j\omega) &= H(j\omega)S_X(\omega) \\ &= \frac{j\omega}{(\omega^2 + (80)^2)(j\omega + 9)(j\omega + 10000)} \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= |H(j\omega)|^2 S_X(\omega) \\ &= \frac{\omega^2}{(\omega^2 + (80)^2)(\omega^2 + (9)^2)(\omega^2 + (10000)^2)} \end{aligned}$$

71. The PSD of the random process $X(t)$ and the transfer function of a network are

$$S_X(\omega) = \frac{1}{\omega^2 + (70)^2} \quad \text{and} \quad H(s) = \frac{s}{(s+10)(s+11000)}$$

$Y(s) = H(s)X(s)$. Find μ_Y , $S_{XY}(j\omega)$ and $S_Y(\omega)$. [?]

Solution:

$S_X(\omega)$ doesn't have any dc component (unit impulse function) when $\omega = 0$, hence

$$\mu_X = 0$$

Also the relation is

$$\mu_Y = \mu_X H(0) = 0$$

$$\begin{aligned} S_{XY}(j\omega) &= H(j\omega)S_X(\omega) \\ &= \frac{j\omega}{(\omega^2 + (70)^2)(j\omega + 10)(j\omega + 11000)} \end{aligned}$$

$$\begin{aligned} S_Y(\omega) &= |H(j\omega)|^2 S_X(\omega) \\ &= \frac{\omega^2}{(\omega^2 + (70)^2)(\omega^2 + (10)^2)(\omega^2 + (11000)^2)} \end{aligned}$$