

ECE-GY 6303, PROBABILITY & STOCHASTIC PROCESSES

Solution to Homework # 8

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Problem 1

a.) Find the correlation function of

$$X(t) = \sum_{k=1}^n a_k \cos(\omega_k t + \phi_k),$$

where a_k and ω_k are constants and ϕ_k are independent random variables that are uniformly distributed in $(0, 2\pi)$.

b.) Show that $R(t_1, t_2) = \min(t_1, t_2)$ is an auto-correlation function.

Solution:

$$\begin{aligned} \text{a.)} \quad R(t_2, t_1) &= E[X(t_1)X^*(t_2)] \\ &= E \left[\left(\sum_{k=1}^n a_k \cos(\omega_k t_1 + \phi_k) \right) \left(\sum_{i=1}^n a_i^* \cos(\omega_i t_2 + \phi_i) \right) \right] \\ &= \sum_{k=1}^n \sum_{i=1}^n a_k a_i^* E [\cos(\omega_k t_1 + \phi_k) \cos(\omega_i t_2 + \phi_i)] \\ &= \frac{1}{2} \sum_{k=1}^n \sum_{i=1}^n a_k a_i^* E [\cos(\omega_k t_1 - \omega_i t_2 + \phi_k - \phi_i) + \cos(\omega_k t_1 + \phi_k + \omega_i t_2 + \phi_i)] \\ &= \frac{1}{2} \sum_{k=1}^n a_k^2 E [\cos(\omega_k(t_2 - t_1))], \end{aligned}$$

as $E [\cos(\omega_k t_1 - \omega_i t_2 + \phi_k - \phi_i)] = E [\cos(\omega_k t_1 + \phi_k + \omega_i t_2 + \phi_i)] = 0$ for $i \neq k$.

b.) Let $t_0 = 0$. Define the matrix

$$\begin{aligned}
 T &= \begin{bmatrix} t_1 & t_1 & \cdots & t_1 \\ t_1 & t_2 & \cdots & t_2 \\ \vdots & \vdots & \ddots & \vdots \\ t_1 & t_2 & \cdots & t_n \end{bmatrix} = \begin{bmatrix} t_1 & t_1 & \cdots & t_1 \\ t_1 & t_1 & \cdots & t_1 \\ \vdots & \vdots & \ddots & \vdots \\ t_1 & t_2 & \cdots & t_1 \end{bmatrix} + \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & t_2 - t_1 & \cdots & t_2 - t_1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & t_2 - t_1 & \cdots & t_n - t_1 \end{bmatrix} \\
 &= t_1 \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & \cdots & 1 \end{bmatrix} + (t_2 - t_1) \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & \cdots & 1 \end{bmatrix} + \cdots + (t_n - t_{n-1}) \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & \cdots & 1 \end{bmatrix} \\
 &= \sum_{k=1}^n \mu_k a_k a_k^T,
 \end{aligned}$$

where $\mu_k = t_k - t_{k-1}$. As $\det(T) > 0$, $R(t_1, t_2) = \min(t_1, t_2)$ is an auto-correlation function.

Problem 2

$X(t) = N(0, t)$ is a Poisson process with parameter λt , i.e., k arrivals in $(0, t)$ is governed by

$$P[X(t) = k] = e^{-\lambda t} \frac{(\lambda t)^k}{k!}.$$

- a.) Show that “the duration of the first arrival” is an exponential random variable;
- b.) Show that “the duration of the n -th arrival” is a Gamma distributed random variable. Find its parameters.

Solution:

- a.) Let τ_1 denote the time to the first arrival. Then,

$$P[\tau_1 > t] = P[N(0, t) = 0] = e^{-\lambda t}.$$

Thus,

$$F_{\tau_1}(t) = 1 - P[\tau_1 > t] = 1 - e^{-\lambda t}, \quad f_{\tau_1}(t) = \frac{dF_{\tau_1}(t)}{dt} = \lambda e^{-\lambda t}, \quad t \geq 0.$$

So τ_1 is exponentially distributed.

- b.) Similarly with a.), let τ_n denote the time to the n -th arrival.

$$P[\tau_n > t] = P[N(0, t) \leq n - 1] = \sum_{k=0}^{n-1} P[X(t) = k].$$

Hence,

$$F_{\tau_n}(t) = 1 - P[\tau_n > t] = 1 - \sum_{k=0}^{n-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!},$$

and

$$\begin{aligned} f_{\tau_n}(t) &= \frac{dF_{\tau_n}(t)}{dt} = \sum_{k=0}^{n-1} \left(-k e^{-\lambda t} \frac{\lambda^k t^{k-1}}{k!} + \right) + \lambda \sum_{k=0}^{n-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!} \\ &= -\lambda \sum_{k=0}^{n-2} e^{-\lambda t} \frac{(\lambda t)^k}{k!} + \lambda \sum_{k=0}^{n-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!} \\ &= \lambda^n e^{-\lambda t} \frac{t^{n-1}}{(n-1)!}. \end{aligned}$$

The parameters are $\alpha = n$ and $\beta = \lambda$.

Problem 3

Let

$$X(t) = e^{j(\omega_0 t + W(t))},$$

where $W(t)$ is a Wiener process and ω_0 is a constant. Calculate the mean and autocorrelation function of $X(t)$.

Solution:

$$\begin{aligned} R_{XX}(t_1, t_2) &= E[X(t_1)X^*(t_2)] \\ &= E[e^{j(\omega_0 t_1 + W(t_1))} e^{-j(\omega_0 t_2 + W(t_2))}] \\ &= e^{j\omega_0(t_1 - t_2)} E[e^{j(W(t_1) - W(t_2))}]. \end{aligned}$$

Let

$$z = W(t_1) - W(t_2) \sim \mathcal{N}(0, \sigma_Z^2), \quad \sigma_Z^2 = E[(W(t_1) - W(t_2))^2] = \begin{cases} t_2 - t_1 & t_2 \geq t_1, \\ t_1 - t_2 & \text{otherwise.} \end{cases}$$

Hence,

$$R_{XX}(t_1, t_2) = e^{j\omega_0(t_1 - t_2)} e^{j \cdot j \frac{\sigma_Z^2}{2}} = e^{j\omega_0(t_1 - t_2)} e^{-\frac{|t_2 - t_1|}{2}}.$$

Problem 4

Show that

- a.) all strict sense stationary processes are wide sense stationary;
- b.) wide sense stationarity implies strict sense stationarity for Gaussian process.

Solution:

- a.) $X(t)$ is S.S.S. means that with t_1, t_2, \dots, t_n and $X(t_i) = X_i$, $X(t_i + \Delta t) = \tilde{X}_i$,

$$f_X(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n) = f_{\tilde{X}}(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n), \quad \forall \Delta t.$$

Thus $f_X(x, t) = f_X(x)$, and

$$f_{X_1 X_2}(x_1, x_2; t_1, t_2) = f_{X_1 X_2}(x_1, x_2; \tau), \quad \tau = t_2 - t_1.$$

Hence,

$$E[X(t)] = \int x f_X(x) dx = \text{Constant},$$

$$R_{X_1 X_2}(t_1, t_2) = \int \int x_1 x_2 f_{X_1 X_2}(x_1, x_2; \tau) dx_1 dx_2 = \rho(\tau), \quad \tau = t_2 - t_1.$$

So $X(t)$ is also W.S.S.

- b.) For a Gaussian process $X(t)$, its mean and auto-correlation function define the process. Let $X_i = X(t_i)$, $i = 1, 2, \dots, n$. The joint distribution is given by

$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{(2\pi)^{n/2} |\mathbf{R}|} e^{-\mathbf{x}^T \mathbf{R}^{-1} \mathbf{x}},$$

where

$$\mathbf{X} = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_n \end{bmatrix}, \quad \mathbf{R} = \begin{bmatrix} R_{11} & R_{12} & \cdots & R_{1n} \\ R_{21} & R_{22} & \cdots & R_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ R_{n1} & R_{n2} & \cdots & R_{nn} \end{bmatrix}, \quad R_{ij} = R_{XX}(t_i, t_j).$$

If $X(t)$ is W.S.S., then $R_{XX}(t_i, t_j) = R_{XX}(t_j - t_i)$ and $E[X(t)] = \text{constant}$. Hence, the Gaussian process $X(t)$ is also S.S.S.

Problem 5

Let $X(t)$ be a zero-mean real stochastic process with

$$E[(X(t) - X(s))^2] = |t - s|, \quad \text{and} \quad X(0) = 0.$$

a.) Show that

$$R_{XX}(t, s) = E[X(t)X(s)] = \frac{1}{2}(|t| + |s| - |t - s|) = \min(t, s).$$

b.) Define

$$Z_n(t) = n \left[X \left(t + \frac{1}{n} \right) - X(t) \right], \quad n = 1, 2, \dots$$

Given n , find the auto-correlation function of $Z_n(t)$ and plot it. Is $Z_n(t)$ W.S.S.?

c.) What happens to $Z_n(t)$ when $n \rightarrow \infty$? (Does it appear to have the characteristics of any known stochastic process?)

Solution:

a.)

$$\begin{aligned} R_{XX}(t, s) &= E[X(t)X(s)] \\ &= \frac{1}{2} (E[(X(t))^2] + E[(X(s))^2] - E[(X(t) - X(s))^2]) \\ &= \frac{1}{2} (E[(X(t) - X(0))^2] + E[(X(s) - X(0))^2] - E[(X(t) - X(s))^2]) \\ &= \frac{1}{2} (|t| + |s| - |t - s|) \\ &= \begin{cases} t & s \geq t, \\ s & \text{otherwise} \end{cases} \\ &= \min(t, s). \end{aligned}$$

b.)

$$\begin{aligned} R_{ZZ}(t, s) &= E[Z(t)Z(s)] \\ &= n^2 E \left[\left(X \left(t + \frac{1}{n} \right) - X(t) \right) \left(X \left(s + \frac{1}{n} \right) - X(s) \right) \right] \\ &= n^2 \left(R_{XX} \left(t + \frac{1}{n}, s + \frac{1}{n} \right) - R_{XX} \left(t + \frac{1}{n}, s \right) - R_{XX} \left(t, s + \frac{1}{n} \right) + R_{XX}(t, s) \right) \\ &= \frac{n^2}{2} (|t - s + 1/n| + |t - s - 1/n| - 2|t - s|) \\ &= R_{ZZ}(t - s). \end{aligned}$$

Hence, $Z(t)$ is W.S.S.

c.) As $n \rightarrow \infty$,

$$R_{ZZ}(t - s) = \delta(t - s).$$

Thus,

$$\lim_{n \rightarrow \infty} Z_n(t) = W(t), \quad \text{White Noise.}$$