

Solutions to Chapter 10 Home Exercises

Problem 10.8

$$S_{X,X}(f) = FT[R_{X,X}(\tau)] = FT[1] = \delta(f).$$

That is, all power in the process is at d.c.

Problem 10.10

$$\begin{aligned} R_{X,X}(t, t + \tau) &= E[b^2 \cos(\omega t + \Theta) \cos(\omega(t + \tau) + \Theta)] \\ &= \frac{b^2}{2} \cos(\omega\tau) + \frac{b^2}{2} E[\cos(\omega(2t + \tau) + 2\Theta)] \\ R_{X,X}(\tau) &= \langle R_{X,X}(t, t + \tau) \rangle = \frac{b^2}{2} \cos(\omega\tau) \\ S_{X,X}(f') &= \frac{b^2}{4} \delta(f' - f) + \frac{b^2}{4} \delta(f' + f) \end{aligned}$$

This PSD is independent of the distribution of Θ . This is expected because the process has all its power at frequency, f , regardless of the phase Θ .

Problem 10.13

(a)

$$R_{Z,Z}[k] = R_{X,X}[k] + R_{Y,Y}[k] = \left(\frac{1}{2}\right)^{|k|} + \left(\frac{1}{3}\right)^{|k|}$$

(see Exercise 8.18 for details)

(b) For a function of the form $R[k] = p^{|k|}$, the Fourier Transform is (t_o is the time between samples of the discrete time process)

$$\begin{aligned} S(f) &= \sum_k R[k] e^{-j2\pi k f t_o} \\ &= 1 + \sum_{k=1}^{\infty} p^k \{e^{-j2\pi k f t_o} + e^{j2\pi k f t_o}\} \\ &= 1 + \frac{pe^{-j2\pi k f t_o}}{1 - pe^{-j2\pi k f t_o}} + \frac{pe^{j2\pi k f t_o}}{1 - pe^{j2\pi k f t_o}} \\ &= \frac{1 - p^2}{1 + p^2 - 2p \cos(2\pi f t_o)} \end{aligned}$$

Therefore,

$$\begin{aligned} S_{X,X}(f) &= \frac{3/4}{5/4 - \cos(2\pi f t_o)} \\ S_{Y,Y}(f) &= \frac{8/9}{10/9 - (2/3) \cos(2\pi f t_o)} \\ S_{Z,Z}(f) &= S_{X,X}(f) + S_{Y,Y}(f). \end{aligned}$$

Problem 10.21 a

$$X[n] = \frac{1}{2}X[n-1] + E[n]. \quad (1)$$

Taking expectations of both sides of (1) results in

$$\mu[n] = \frac{1}{2}\mu[n-1], \quad n = 1, 2, 3, \dots$$

Hence $\mu[n] = (1/2)^n \mu[0]$. Noting that $X(0) = 0$, then $\mu[0] = 0 \Rightarrow \mu[n] = 0$. Multiply both sides of (1) by $X[k]$ and then take expected values to produce

$$E[X[k]X[n]] = \frac{1}{2}E[X[k]X[n-1]] + E[X[k]E[n]].$$

Assuming $k < n$, $X[k]$ and $E[n]$ are independent. Thus, $E[X[k]E[n]] = 0$ and therefore

$$\begin{aligned} R_{X,X}[k, n] &= \frac{1}{2}R_{X,X}[k, n-1]. \\ \Rightarrow R_{X,X}[k, n] &= \left(\frac{1}{2}\right)^{n-k} R_{X,X}[k, k], \quad n = k, k+1, k+2, \dots \end{aligned}$$

Following a similar procedure, it can be shown that if $k > n$

$$R_{X,X}[k, n] = \left(\frac{1}{2}\right)^{k-n} R_{X,X}[k, k].$$

Hence in general

$$R_{X,X}[k, n] = \left(\frac{1}{2}\right)^{|n-k|} R_{X,X}[m, m], \text{ where } m = \min(n, k).$$

Note that $R_{X,X}[m, m]$ can be found as follows:

$$\begin{aligned} R_{X,X}[m, m] &= E[X^2[m]] = E\left[\left(\frac{1}{2}X[m-1] + E[m]\right)^2\right] \\ &= \frac{1}{4}R_{X,X}[m-1, m-1] + E[X[m-1]E[m]] + E[E^2[m]]. \end{aligned}$$

Since $X[m-1]$ and $E[m]$ are uncorrelated, we have the following recursion

$$\begin{aligned} R_{X,X}[m, m] &= \frac{1}{4}R_{X,X}[m-1, m-1] + \sigma_E^2 \\ \Rightarrow R_{X,X}[m, m] &= \left(\frac{1}{4}\right)^m R_{X,X}[0, 0] + \sigma_E^2 \sum_{i=0}^{m-1} \left(\frac{1}{4}\right)^i. \end{aligned}$$

Note that since $X(0) = 0$, $R_{X,X}(0,0) = 0$. Therefore

$$\begin{aligned} R_{X,X}[m,m] &= \sigma_E^2 \frac{1 - (1/4)^m}{1 - 1/4} = \frac{4\sigma_E^2}{3}(1 - (1/4)^m) \\ \Rightarrow R_{X,X}[k,n] &= \frac{4\sigma_E^2}{3}(1 - (1/4)^m) \left(\frac{1}{2}\right)^{|n-k|}. \end{aligned}$$

Since $m = \min(n, k)$ is not a function of $n - k$, the process is not WSS.

Problem 10.24

(a)

$$\begin{aligned} E[\epsilon^2] &= E[(Y[n+1] - \sum_{k=1}^p a_k Y[n-k+1])^2] \\ &= E[Y^2[n+1]] - 2 \sum_{k=1}^p a_k E[Y[n+1]Y[n+1-k]] \\ &\quad + \sum_{k=1}^p \sum_{m=1}^p a_k a_m E[Y[n+1-k]Y[n+1-m]] \\ &= R_{Y,Y}[0] - 2 \sum_{k=1}^p a_k R_{Y,Y}[k] + \sum_{k=1}^p \sum_{m=1}^p a_k a_m R_{Y,Y}[m-k] \end{aligned}$$

To simplify the notation, introduce the following vectors and matrices:

$$\begin{aligned} \mathbf{r} &= [R_{Y,Y}[1] \quad R_{Y,Y}[2] \quad \dots \quad R_{Y,Y}[p]]^T, \\ \mathbf{a} &= [a_1 \quad a_2 \quad \dots \quad a_p]^T, \\ \mathbf{R} &= p \times p \text{ matrix whose } (k,m)\text{th element is } R_{Y,Y}[m-k]. \end{aligned}$$

Then the mean squared error is

$$E[\epsilon^2] = R_{Y,Y}[0] - 2\mathbf{r}^T \mathbf{a} + \mathbf{a}^T \mathbf{R} \mathbf{a}.$$

(b)

$$\begin{aligned} \nabla_{\mathbf{a}} &= -2\mathbf{r} + 2\mathbf{R}\mathbf{a} = 0 \\ \Rightarrow \mathbf{a} &= \mathbf{R}^{-1}\mathbf{r} \end{aligned}$$