

---

## 10.2.5 Solved Problems

### Problem 1

Consider a WSS random process  $X(t)$  with

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

Find the PSD of  $X(t)$ , and  $E[X(t)^2]$ .

**Solution**

First, we have

$$E[X(t)^2] = R_X(0) = 1.$$

We can write triangular function,  $R_X(\tau) = \Lambda(\tau)$ , as

$$R_X(\tau) = \Pi(\tau) * \Pi(\tau),$$

where

$$\Pi(\tau) = \begin{cases} 1 & -\frac{1}{2} \leq \tau \leq \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$

Thus, we conclude

$$\begin{aligned} S_X(f) &= \mathcal{F}\{R_X(\tau)\} \\ &= \mathcal{F}\{\Pi(\tau) * \Pi(\tau)\} \\ &= \mathcal{F}\{\Pi(\tau)\} \cdot \mathcal{F}\{\Pi(\tau)\} \\ &= [\text{sinc}(f)]^2. \end{aligned}$$

---

### Problem 2

Let  $X(t)$  be a random process with mean function  $\mu_X(t)$  and autocorrelation function  $R_X(s, t)$  ( $X(t)$  is not necessarily a WSS process). Let  $Y(t)$  be given by

$$Y(t) = h(t) * X(t),$$

where  $h(t)$  is the impulse response of the system. Show that

a.  $\mu_Y(t) = \mu_X(t) * h(t)$ .

b.  $R_{XY}(t_1, t_2) = h(t_2) * R_X(t_1, t_2) = \int_{-\infty}^{\infty} h(\alpha) R_X(t_1, t_2 - \alpha) d\alpha$ .

### Solution

a. We have

$$\begin{aligned} \mu_Y(t) &= E[Y(t)] = E \left[ \int_{-\infty}^{\infty} h(\alpha) X(t - \alpha) d\alpha \right] \\ &= \int_{-\infty}^{\infty} h(\alpha) E[X(t - \alpha)] d\alpha \\ &= \int_{-\infty}^{\infty} h(\alpha) \mu_X(t - \alpha) d\alpha \\ &= \mu_X(t) * h(t). \end{aligned}$$

b. We have

$$\begin{aligned} R_{XY}(t_1, t_2) &= E[X(t_1)Y(t_2)] = E \left[ X(t_1) \int_{-\infty}^{\infty} h(\alpha) X(t_2 - \alpha) d\alpha \right] \\ &= E \left[ \int_{-\infty}^{\infty} h(\alpha) X(t_1) X(t_2 - \alpha) d\alpha \right] \\ &= \int_{-\infty}^{\infty} h(\alpha) E[X(t_1) X(t_2 - \alpha)] d\alpha \\ &= \int_{-\infty}^{\infty} h(\alpha) R_X(t_1, t_2 - \alpha) d\alpha. \end{aligned}$$

### Problem 3

Prove the third part of [Theorem 10.2](#): Let  $X(t)$  be a WSS random process and  $Y(t)$  be given by

$$Y(t) = h(t) * X(t),$$

where  $h(t)$  is the impulse response of the system. Show that

$$R_Y(s, t) = R_Y(s - t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\beta) R_X(s - t - \alpha + \beta) d\alpha d\beta.$$

Also, show that we can rewrite the above integral as  $R_Y(\tau) = h(\tau) * h(-\tau) * R_X(\tau)$ .

**Solution**

$$\begin{aligned}
R_Y(s, t) &= E[X(s)Y(t)] \\
&= E \left[ \int_{-\infty}^{\infty} h(\alpha) X(s - \alpha) d\alpha \int_{-\infty}^{\infty} h(\beta) X(t - \beta) d\beta \right] \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\beta) E[X(s - \alpha) X(t - \beta)] d\alpha d\beta \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\beta) R_X(s - t - \alpha + \beta) d\alpha d\beta.
\end{aligned}$$

We now compute  $h(\tau) * h(-\tau) * R_X(\tau)$ . First, let  $g(\tau) = h(\tau) * h(-\tau)$ . Note that

$$\begin{aligned}
g(\tau) &= h(\tau) * h(-\tau) \\
&= \int_{-\infty}^{\infty} h(\alpha) h(\alpha - \tau) d\alpha.
\end{aligned}$$

Thus, we have

$$\begin{aligned}
g(\tau) * R_X(\tau) &= \int_{-\infty}^{\infty} g(\theta) R_X(\theta - \tau) d\theta \\
&= \int_{-\infty}^{\infty} \left[ \int_{-\infty}^{\infty} h(\alpha) h(\alpha - \theta) d\alpha \right] R_X(\theta - \tau) d\theta \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\alpha - \theta) R_X(\theta - \tau) d\alpha d\theta \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\beta) R_X(\alpha - \beta - \tau) d\alpha d\beta \\
&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha) h(\beta) R_X(\tau - \alpha + \beta) d\alpha d\beta \quad (\text{since } R_X(-\tau) = R_X(\tau)).
\end{aligned}$$


---

**Problem 4**

Let  $X(t)$  be a WSS random process. Assuming that  $S_X(f)$  is continuous at  $f_1$ , show that  $S_X(f_1) \geq 0$ .

**Solution**

Let  $f_1 \in \mathbb{R}$ . Suppose that  $X(t)$  goes through an LTI system with the following transfer function

$$H(f) = \begin{cases} 1 & f_1 < |f| < f_1 + \Delta \\ 0 & \text{otherwise} \end{cases}$$

where  $\Delta$  is chosen to be very small. The PSD of  $Y(t)$  is given by

$$S_Y(f) = S_X(f)|H(f)|^2 = \begin{cases} S_X(f) & f_1 < |f| < f_1 + \Delta \\ 0 & \text{otherwise} \end{cases}$$

Thus, the power in  $Y(t)$  is

$$\begin{aligned} E[Y(t)^2] &= \int_{-\infty}^{\infty} S_Y(f) df \\ &= 2 \int_{f_1}^{f_1+\Delta} S_X(f) df \\ &\approx 2\Delta S_X(f_1). \end{aligned}$$

Since  $E[Y(t)^2] \geq 0$ , we conclude that  $S_X(f_1) \geq 0$ .

### Problem 5

Let  $X(t)$  be a white Gaussian noise with  $S_X(f) = \frac{N_0}{2}$ . Assume that  $X(t)$  is input to an LTI system with

$$h(t) = e^{-t}u(t).$$

Let  $Y(t)$  be the output.

- Find  $S_Y(f)$ .
- Find  $R_Y(\tau)$ .
- Find  $E[Y(t)^2]$ .

### Solution

First, note that

$$\begin{aligned} H(f) &= \mathcal{F}\{h(t)\} \\ &= \frac{1}{1 + j2\pi f}. \end{aligned}$$

- To find  $S_Y(f)$ , we can write

$$\begin{aligned}
 S_Y(f) &= S_X(f)|H(f)|^2 \\
 &= \frac{N_0/2}{1 + (2\pi f)^2}.
 \end{aligned}$$

b. To find  $R_Y(\tau)$ , we can write

$$\begin{aligned}
 R_Y(\tau) &= \mathcal{F}^{-1}\{S_Y(f)\} \\
 &= \frac{N_0}{4}e^{-|\tau|}.
 \end{aligned}$$

c. We have

$$\begin{aligned}
 E[Y(t)^2] &= R_Y(0) \\
 &= \frac{N_0}{4}.
 \end{aligned}$$


---