## 7.3.4 Autocorrelation Functions for Random Pulse Trains

As another example of calculating autocorrelation functions, consider a random process with sample functions that can be expressed as

$$X(t) = \sum_{k=-\infty}^{\infty} a_k p(t - kT - \Delta)$$
 (7.49)

where ...  $a_{-1}, a_0, a_1, ..., a_k$  ... is a doubly-infinite sequence of random variables with

$$E[a_k a_{k+m}] = R_m \tag{7.50}$$

The function p(t) is a deterministic pulse-type waveform where T is the separation between pulses;  $\Delta$  is a random variable that is independent of the value of  $a_k$  and uniformly distributed in the interval (-T/2, T/2). The autocorrelation function of this waveform is

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

$$= E\left\{\sum_{k=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} a_k a_{k+m} p(t-kT-\Delta) p\left[t+\tau-(k+m)T-\Delta\right]\right\}$$
(7.51)

Taking the expectation inside the double sum and making use of the independence of the sequence  $\{a_k a_{k+m}\}$  and the delay variable  $\Delta$ , we obtain

$$R_{X}(\tau) = \sum_{k=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} E\left[a_{k} a_{k+m}\right] E\left\{p(t-kT-\Delta) p\left[t+\tau-(k+m)T-\Delta\right]\right\}$$

$$= \sum_{m=-\infty}^{\infty} R_{m} \sum_{k=-\infty}^{\infty} \int_{-T/2}^{T/2} p(t-kT-\Delta) p\left[t+\tau-(k+m)T-\Delta\right] \frac{d\Delta}{T}$$
 (7.52)

The change of variables  $u = t - kT - \Delta$  inside the integral results in

$$R_{X}(\tau) = \sum_{m=-\infty}^{\infty} R_{m} \sum_{k=-\infty}^{\infty} \int_{t-(k+1/2)T}^{t-(k-1/2)T} p(u) p(u+\tau-mT) \frac{du}{T}$$

$$= \sum_{m=-\infty}^{\infty} R_{m} \left[ \frac{1}{T} \int_{-\infty}^{\infty} p(u+\tau-mT) p(u) du \right]$$
(7.53)

Finally we have

$$R_X(\tau) = \sum_{m=-\infty}^{\infty} R_m \, r(\tau - mT) \tag{7.54}$$

where

$$r(\tau) \triangleq \frac{1}{T} \int_{-\infty}^{\infty} p(t+\tau) p(t) dt$$
 (7.55)

is the pulse-correlation function. We consider the following example as an illustration.

## **EXAMPLE 7.6**

In this example we consider a situation where the sequence  $\{a_k\}$  has memory built into it by the relationship

$$a_k = g_0 A_k + g_1 A_{k-1} (7.56)$$

where  $g_0$  and  $g_1$  are constants and the  $A_k$ 's are random variables such that  $A_k = \pm A$  where the sign is determined by a random coin toss independently from pulse to pulse for all k (note that if  $g_1 = 0$ , there is no memory). It can be shown that

$$E[a_k a_{k+m}] = \begin{cases} \left(g_0^2 + g_1^2\right) A^2, & m = 0\\ g_0 g_1 A^2, & m = \pm 1\\ 0, & \text{otherwise} \end{cases}$$
 (7.57)

The assumed pulse shape is  $p(t) = \Pi\left(\frac{t}{T}\right)$  so that the pulse-correlation function is

$$r(\tau) = \frac{1}{T} \int_{-\infty}^{\infty} \Pi\left(\frac{t+\tau}{T}\right) \Pi\left(\frac{t}{T}\right) dt$$
$$= \frac{1}{T} \int_{-T/2}^{T/2} \Pi\left(\frac{t+\tau}{T}\right) dt = \Lambda\left(\frac{\tau}{T}\right)$$
(7.58)

Thus, the autocorrelation function (7.58) becomes

$$R_X\left(\tau\right) = A^2 \left\{ \left[ g_0^2 + g_1^2 \right] \Lambda\left(\frac{\tau}{T}\right) + g_0 g_1 \left[ \Lambda\left(\frac{\tau + T}{T}\right) + \Lambda\left(\frac{\tau - T}{T}\right) \right] \right\} \tag{7.59}$$

From: PRINCIPLES OF COMMUNICATIONS Systems, Modulation, and Noise, 7<sup>th</sup> ed. R. Ziemer, and W. Tranter, Wiley, 2014.