

# **DIGITAL SIGNAL PROCESSING**

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### 11.3.1 Definition of the Periodogram

As an estimate of the power density spectrum let us consider the Fourier transform of the biased autocorrelation estimate  $c_{xx}(m)$ . That is,

$$I_N(\omega) = \sum_{m=-(N-1)}^{N-1} c_{xx}(m)e^{-j\omega m} \quad (11.24)$$

Since the Fourier transform of the real finite-length sequence  $x(n)$ ,  $0 \leq n \leq N-1$ , is

$$X(e^{j\omega}) = \sum_{n=0}^{N-1} x(n)e^{-j\omega n}$$

it can be shown that (see Problem 1 of this chapter).

$$I_N(\omega) = \frac{1}{N} |X(e^{j\omega})|^2 \quad (11.25)$$

The spectrum estimate  $I_N(\omega)$  is often called the *periodogram*.

As before, it is of interest to determine the bias and variance of the periodogram as an estimate of the power spectrum. The expected value of  $I_N(\omega)$  is

$$E[I_N(\omega)] = \sum_{m=-(N-1)}^{N-1} E[c_{xx}(m)]e^{-j\omega m} \quad (11.26)$$

Since we have shown that for a zero mean process

$$E[c_{xx}(m)] = \frac{N - |m|}{N} \phi_{xx}(m), \quad |m| < N$$

then

$$E[I_N(\omega)] = \sum_{m=-(N-1)}^{N-1} \left( \frac{N - |m|}{N} \right) \phi_{xx}(m)e^{-j\omega m} \quad (11.27)$$

Thus because of the finite limits of summation and the factor  $(N - |m|)/N$ ,  $E[I_N(\omega)]$  is not equal to the Fourier transform of  $\phi_{xx}(m)$ , and therefore the periodogram is a biased estimate of the power spectrum,  $P_{xx}(\omega)$ .

Alternatively, consider the Fourier transform of the estimate  $c'_{xx}(m)$ ; i.e.,

$$P_N(\omega) = \sum_{m=-(N-1)}^{N-1} c'_{xx}(m)e^{-j\omega m} \quad (11.28)$$

The expected value of  $P_N(\omega)$  is

$$\begin{aligned} E[P_N(\omega)] &= \sum_{m=-(N-1)}^{N-1} E[c'_{xx}(m)]e^{-j\omega m} \\ &= \sum_{m=-(N-1)}^{N-1} \phi_{xx}(m)e^{-j\omega m} \end{aligned} \quad (11.29)$$

Again, because of the finite limits of summation,  $P_{xx}(\omega)$ , even though  $c_{xx}(m)$  is an unbiased estimator of  $P_{xx}(\omega)$ , is biased.

We can interpret Eqs. (11.27) and (11.29) as the Fourier transforms of windowed autocorrelation sequences. In the case of  $w_B(m)$ , the window is the triangular window

$$w_B(m) = \begin{cases} \frac{N - |m|}{N}, & |m| < N \\ 0, & \text{otherwise} \end{cases}$$

In Chapter 5 we called this the Bartlett window. The rectangular window is rectangular; i.e.,

$$w_R(n) = \begin{cases} 1, & |m| < N \\ 0, & \text{otherwise} \end{cases}$$

Using the concepts introduced in Chapter 5 and (11.29) can be interpreted in the frequency domain as

$$E[I_N(\omega)] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P_{xx}(\theta) W_B(e^{j\omega}) d\theta$$

and

$$E[P_N(\omega)] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P_{xx}(\theta) W_R(e^{j\omega}) d\theta$$

where

$$W_B(e^{j\omega}) = \frac{1}{N} \left( \frac{\sin [\omega N/2]}{\sin [\omega/2]} \right)^2$$

and

$$W_R(e^{j\omega}) = \frac{\sin [\omega(2N - 1)/2]}{\sin [\omega/2]}$$

are the Fourier transforms of the Bartlett and rectangular windows, respectively.

### 11.3.2 Variance of the Periodogram

To obtain an expression for the variance of the periodogram, it is convenient to first assume that the sequence  $x(n)$ , is a realization of a real, white, zero-mean process with constant power spectral density  $G$ . The periodogram  $I_N(\omega)$  can be expressed as

$$I_N(\omega) = \frac{1}{N} |X(e^{j\omega})|^2$$

#### 544 Power Spectrum Estimation

To evaluate the covariance of  $I_N(\omega)$  at two frequencies  $\omega_1$  and  $\omega_2$  we first consider

$$E[I_N(\omega_1)I_N(\omega_2)] = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} E[x(k)x(l)x(m)x(n)] e^{j[\omega_1(k-l) + \omega_2(m-n)]} \quad (11.36)$$

To obtain a useful result, we must simplify Eq. (11.36). In general, it is not possible to obtain a very simple result even when  $x(n)$  is white, because  $E[x(n)x(n+m)] = \sigma_x^2 \delta(m)$  does not guarantee a simple expression for  $E[x(k)x(l)x(m)x(n)]$  for all combinations of  $k, l, m,$  and  $n$ . However, in the case of a white Gaussian process, it can be shown [7] that

$$\begin{aligned} E[x(k)x(l)x(m)x(n)] &= E[x(k)x(l)]E[x(m)x(n)] \\ &\quad + E[x(k)x(m)]E[x(l)x(n)] \\ &\quad + E[x(k)x(n)]E[x(l)x(m)] \end{aligned}$$

Therefore,

$$E[x(k)x(l)x(m)x(n)] = \begin{cases} \sigma_x^4, & k = l \text{ and } m = n \\ & \text{or } k = m \text{ and } l = n \\ & \text{or } k = n \text{ and } l = m \\ 0, & \text{otherwise} \end{cases} \quad (11.37)$$

For other than Gaussian joint density functions, the result will not necessarily be so simple. However, our objective is to give a result that will lend insight into the problems of spectrum estimation rather than to give a general formula with wide validity which would be difficult to interpret. Thus, if we substitute Eq. (11.37) into Eq. (11.36), we obtain

$$E[I_N(\omega_1)I_N(\omega_2)] = \frac{\sigma_x^4}{N^2} \left\{ N^2 + \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} e^{j(m-n)(\omega_1+\omega_2)} + \sum_{m=0}^{N-1} \sum_{n=0}^{N-1} e^{j(n-m)(\omega_1-\omega_2)} \right\}$$

or

$$\begin{aligned} E[I_N(\omega_1)I_N(\omega_2)] &= \sigma_x^4 \left\{ 1 + \left( \frac{\sin [(\omega_1 + \omega_2)N/2]}{N \sin [(\omega_1 + \omega_2)/2]} \right)^2 \right. \\ &\quad \left. + \left( \frac{\sin [(\omega_1 - \omega_2)N/2]}{N \sin [(\omega_1 - \omega_2)/2]} \right)^2 \right\} \quad (11.38) \end{aligned}$$

(If the signal is non-Gaussian, Eq. (11.38) contains additional terms which are proportional to  $1/N$  [4, 8].) The covariance of the periodogram is

$$\text{cov} [I_N(\omega_1), I_N(\omega_2)] = E[I_N(\omega_1)I_N(\omega_2)] - E[I_N(\omega_1)]E[I_N(\omega_2)] \quad (11.39)$$

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