04: Discrete-time random signal characterization and transmission

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Sampling

Autocorrelation and Autocovariance

Power Spectrum

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EE21: Random Signal Processing

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Given an analog process X(t), we form the digital process $\mathbf{Y}(t)$

$$\mathbf{X}[n] = \mathbf{X}(nT)$$

where T is a given constant. From this it follows that

$$\mu[n] = \mu_a(nT)$$
 $R[n_1, n_2] = R_a(n_1T, n_2T)$ (1)

where $\mu_a(t)$ is the mean and $R_a(t_1,t_2)$ the autocorrelation of $\mathbf{X}(t)$.

mean $\mu=\mu_a$ and autocorrelation

$$R[m] = R_a(mT)$$

If $\mathbf{X}(t)$ is a stationary process, then $\mathbf{X}[n]$ is also stationary with

From this it follows that the power spectrum of X[n] equals

$$\mathbf{S}_X(e^{j\omega}) = \sum_{\alpha=0}^{\infty} R_a(mT)e^{-jm\omega} = \frac{1}{T} \sum_{\alpha=0}^{\infty} S_a(\frac{\omega + 2\pi n}{T})$$
 (2)

where $S_a(\omega)$ is the power spectrum of X(t). The above is a consequence of Poisson's sum formula.

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$$\mathbf{X}(t) = \sum_{i=1}^{M} \sigma_i e^{j\omega_i t} \qquad S_a(\omega) = 2\pi \sum_{i=1}^{M} \sigma_i^2 \delta(\omega - \omega_i)$$

where $\sigma_i^2=E\{\mathbf{c}_i^2\}$. We shall determine the power spectrum $\mathbf{S}(e^{j\omega})$ of the process $\mathbf{X}[n]=\mathbf{X}(nT)$. Since $\delta(\omega/T)=T\delta(\omega)$, it follows from (2) that

$$\mathbf{S}(e^{j\omega}) = 2\pi \sum_{n=-\infty}^{\infty} \sum_{i=1}^{M} \sigma_i^2 \delta(\omega - T\omega_i + 2\pi n)$$

In the interval $(-\pi,\pi)$, this consists of M lines:

$$\mathbf{S}(e^{j\omega}) = 2\pi \sum_{i=1}^{M} \sigma_i^2 \delta(\omega - \beta_i) \qquad |\omega| < \pi$$

where $\beta_i = T\omega_i - 2\pi n_i$ and such that $|\beta_i| < \pi$.

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The autocorrelation and autocovariance of X[n] are given by

$$R[n_1, n_2] = E\{X[n_1]X^*[n_2]\}$$
(3)

$$C[n_1, n_2] = R[n_1, n_2] - \mu[n_1]\mu^*[n_2]$$
 (4)

respectively where $\mu[n] = E\{X[n]\}$ is the mean of X[n].

A process X[n] is SSS if its statistical properties are invariant to a shift of the origin.

A process X[n] is WSS if $\mu[n] = \mu = {\rm constant}$ and

$$R[n+m,n] = E(X[n+m]X^*[n]) = R[m]$$
 (5)

- **1** A process X[n] is strictly white noise if the random variables $X[n_i]$ are independent.
- 2 It is white noise if the random variables $X[n_i]$ are uncorrelated.
- 3 The autocorrelation of a white-noise process with zero mean is thus given by

$$R[n_1, n_2] = q[n_1]\delta[n_1 - n_2]$$
 where $\delta[n] = \begin{cases} 1 & n = 0 \\ 0 & n \neq 0 \end{cases}$ (6)

and $q[n] = E\{X^2[n]\}.$

4 If X[n] is also stationary, then $R[m] = q\delta[m]$. Thus a WSS white noise is a sequence of i.i.d. random variables with variance q.

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Given a WSS process X[n], we form the z transform S(z) of its autocorrelation R[m]:

$$S(z) = \sum_{m = -\infty}^{\infty} R[m] z^{-m} \tag{7}$$

The power spectrum of X[n] is the function

$$S(\omega) = S(e^{j\omega}) = \sum_{m = -\infty}^{\infty} R[m]e^{-jm\omega} \ge 0$$
 (8)

Thus $S(e^{j\omega})$ is the discrete Fourier transform (DFT) of R[m]. The function $S(e^{j\omega})$ is periodic with period 2π and Fourier series coefficients R[m]. Hence

$$R[m] = \frac{1}{2\pi} \int_{-\pi}^{\pi} S(e^{j\omega}) e^{jm\omega} d\omega$$
 (9)

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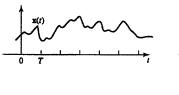
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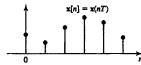


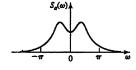
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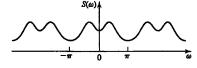
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It suffices, therefore, to specify $S(e^{j\omega})$ for $|w|<\pi$ only (see Fig.).









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If X[n] is a real process, then R[-m] = R[m] and (8) yields

$$S(e^{j\omega}) = R[0] + 2\sum_{m=1}^{\infty} R[m]\cos m\omega$$
 (10)

This shows that the power spectrum of a real process is a function of $\cos \omega$ because $\cos m\omega$ is a function of $\cos \omega$.

The nonnegativity condition can be expressed in terms of certain Hermitian Toeplitz matrices. Let

$$r_k \stackrel{\Delta}{=} R[k] \tag{11}$$

and define

$$T_{n} = \begin{pmatrix} r_{0} & r_{1} & r_{2} & \dots & r_{n} \\ r_{1}^{*} & r_{0} & r_{1} & r_{2} & \dots & r_{n-1} \\ r_{2}^{*} & & r_{0} & & \dots & r_{n-2} \\ \vdots & & & & & & \\ \vdots & & & & & & \\ r_{n}^{*} & r_{n-1}^{*} & \dots & r_{1}^{*} & r_{0} \end{pmatrix}$$

$$(12)$$

In that case

$$S(\omega) \ge 0 \Leftrightarrow T_n \ge 0 \qquad n = 0 \to \infty$$
 (13)

i.e., the nonnegative nature of the spectrum is equivalent to the nonnegativity of every Hemitian Toeplitz matrix $T_n,\,n=0\to\infty$

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To prove this result due to Schur, first assume that $S(\omega) \geq 0$ in (8). Then letting

$$\boldsymbol{a} = [a_0, a_1, a_2, ..., a_n]^T \tag{14}$$

we have

$$a^{H}T_{n}a = \sum_{i=0}^{n} \sum_{m=0}^{n} a_{i}^{*}a_{m}r_{i-m}$$

$$= \sum_{i=0}^{n} \sum_{m=0}^{n} a_{i}^{*}a_{m}\frac{1}{2\pi} \int_{-\pi}^{\pi} S(\omega)e^{j(i-m)\omega}d\omega \qquad (15)$$

$$= \frac{1}{2\pi} \int_{-\pi}^{\pi} S(\omega) \left| \sum_{m=0}^{n} a_{m}e^{-jm\omega} \right|^{2} d\omega \ge 0$$

Since a is arbitrary, this gives

$$S(\omega) \ge 0 \Rightarrow T_n \ge 0 \qquad n = 0 \to \infty$$
 (16)

Conversely, assume that every T_n , $n=0\to\infty$ are nonnegative definite matrices.

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Proof.

Further, for any ρ , $0<\rho<1$, and ω_0 , $0<\omega_0<2\pi$, define the vector ${\bf a}$ in (14) with

$$a_m = \sqrt{1 - \rho^2} \rho^m e^{jm\omega_0}$$

Then T_n nonnegative implies that

$$0 \le \boldsymbol{a}^{H} \boldsymbol{T}_{n} \boldsymbol{a} = \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 - \rho^{2}) \left| \sum_{m=0}^{n} \rho^{m} e^{jm(\omega - \omega_{0})} \right|^{2} S(\omega) d\omega$$

and letting $n \to \infty$, the above intergrad tends to

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{(1 - \rho^2)}{1 - 2\rho \cos(\omega - \omega_0) + \rho^2} S(\omega) d\omega \ge 0 \tag{17}$$

The left-hand side of (17) represents the Poisson integral, and its interior ray limit as $p\to 1-0$ equals $S(\omega)$ for almost all ω_0 . Thus

$$\mathbf{T}_n \geq 0 \quad n=0 \to \infty \Rightarrow S(\omega) \geq 0 \quad \text{almost everywhere (a.e.)}$$
 (18)

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$$\frac{1}{2\pi} \int_{-\pi}^{\pi} \ln S(\omega) d\omega > -\infty \tag{19}$$

every T_k , $k = 0 \to \infty$, must be positive definite.

This follows from (15). In fact, if some T_k is singular, then there exists a nontrivial vector \mathbf{a} such $T_k \mathbf{a} = 0$ and, from (15),

$$\mathbf{a}^{H} \mathbf{T}_{k} \mathbf{a} = \frac{1}{2\pi} \int_{-\pi}^{\pi} S(\omega) \left| \sum_{m=0}^{k} a_{m} e^{jm\omega} \right|^{2} d\omega = 0$$

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Proof.

Since $S(\omega) \ge 0$, a.e., this expression gives

$$S(\omega) \left| \sum_{m=0}^{k} a_m e^{jm\omega} \right|^2 = 0$$
 a.e.

and $\sum_{m=0}^{k} a_m e^{-jm\omega} \neq 0$, a.e., implies

$$S(\omega) = 0$$
 a.e.

and

$$\int_{-\pi}^{\pi} \ln S(\omega) d\omega = -\infty$$

contradicting (19). Hence subject to (19), every

$$\mathbf{T}_k > 0 \qquad k = 0 \to \infty \tag{20}$$

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Example

If $R[m] = a^{|m|}$, then

$$S(z) = \sum_{m=-\infty}^{-1} a^{-m} z^{-m} + \sum_{m=0}^{\infty} a^m z^{-m} = \frac{az}{1-az} + \frac{z}{z-a}$$
$$= \frac{a^{-1} - a}{(a^{-1} + a) - (z^{-1} + z)}$$

Hence
$$S(\omega) = S(e^{j\omega}) = \frac{a^{-1} - a}{a^{-1} + a - 2\cos\omega}$$

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Example

Proceeding as in the analog case, we can show that the process

$$X[n] = \sum_{i} c_i e^{j\omega_i n}$$

is WSS iff the coefficients c_i , are uncorrelated with zero mean. In this case,

$$R[m] = \sum_{i} \sigma_{i}^{2} e^{j\beta_{i}|m|} \qquad S(\omega) = 2\pi \sum_{i} \sigma_{i}^{2} \delta(\omega - \beta_{i}) \qquad |\omega| < \pi$$
 (21)

where $\sigma_i^2 = E\{c_i^2\}$, $\omega_i = 2\pi k_i + \beta_i$, and $|\beta_i| < \pi$.