Stochastic Signal Processing

Lesson 9
Memoryless and Linear Systems II

Weize Sun

Memoryless and Linear Systems – Outline

- The average power Time domain
- Differentiators (微分器)
- Power spectrum of the output of a linear system
 - The average power Frequency domain
 - Differentiators
- Hilbert transforms

The average power – Time domain

We now give the average power $E\{|\mathbf{y}(t)|^2\}$ of the output of a system if its input is white noise.

• If the input to a linear system h(t) is (complex) stationary white noise with autocorrelation $R_X(\tau) = q\delta(\tau)$, then:

$$E\{|\mathbf{y}(t)|^{2}\} = R_{Y}(0) = R_{X}(0) * h(0) * h^{*}(0) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{X}(0 + \alpha - \beta)h(\alpha)h^{*}(\beta)d\alpha d\beta$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} q\delta(\alpha - \beta)h(\alpha)h^{*}(\beta)d\alpha d\beta \quad \text{(for the } \delta(\alpha - \beta), \text{ now-zero only when } \alpha = \beta = t)$$

$$= \int_{-\infty}^{\infty} qh(t)h^{*}(t)dt = qE \qquad \text{where } E = \int_{-\infty}^{\infty} |h(t)|^{2}dt \text{ is the energy of } h(t).$$

Linear Systems – output autocorrelation

• In exam, we consider only the Steady state analysis (稳态分析) of stochastic processes, in this case

If input is stationary, output is stationary, for Steady state analysis (稳态分析) only

Properties when WSS ($\tau = t_1 - t_2$): (complex valued case)

$$R_{XY}(t_1, t_2) = R_X(t_1, t_2) * h^*(t_2) \qquad \Rightarrow \qquad R_{XY}(\tau) = R_X(\tau) * h^*(-\tau)$$

$$R_{YX}(t_1, t_2) = R_X(t_1, t_2) * h(t_1) \qquad \Rightarrow \qquad R_{YX}(\tau) = R_X(\tau) * h(\tau)$$

$$R_Y(t_1, t_2) = R_X(t_1, t_2) * h(t_1) * h^*(t_2) \qquad \Rightarrow \qquad R_Y(\tau) = R_X(\tau) * h(\tau) * h^*(-\tau)$$

$$R_Y(t_1, t_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_X(t_1 - \alpha, t_2 - \beta) h(\alpha) h^*(\beta) d\alpha d\beta$$

$$R_Y(\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_X(\tau + \alpha - \beta) h(\alpha) h^*(\beta) d\alpha d\beta$$

Differentiators

Recall

Linear Systems – output autocorrelation

• In exam, we consider only the Steady state analysis (稳态分析) of stochastic processes, in this case

If input is stationary, output is stationary, for Steady state analysis (稳态分析) only

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$$R_Y(t_1, t_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_X(t_1 - \alpha, t_2 - \beta) h(\alpha) h^*(\beta) d\alpha d\beta$$

$$R_Y(\tau) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_X(\tau + \alpha - \beta) h(\alpha) h^*(\beta) d\alpha d\beta$$

Now we discuss a widely used system called Differentiators

Differentiators

A differentiator is a linear system whose output is the derivative of the input

$$L[\mathbf{x}(t)] = \frac{\mathrm{d}\mathbf{x}(t)}{\mathrm{d}t} = \mathbf{x}'(t)$$

We can, therefore, use the preceding results to find the mean of $\mathbf{x}'(t)$:

$$m_{X'}(t) = L[m_X(t)] = \frac{\mathrm{d}m_X(t)}{\mathrm{d}t} = m_X'(t)$$

Which is, the find the mean of $\mathbf{x}'(t)$, we can first take the mean of $\mathbf{x}(t)$ and thus get $m_X(t)$, and then take the derivation $m_X(t)$

• If $\mathbf{x}(t)$ is WSS, then $m_X(t)$ is constant; hence $\frac{\mathrm{d}m_X(t)}{\mathrm{d}t} = 0$, which is:

$$E\{\mathbf{x}'(t)\}=0$$

Differentiators

• A differentiator is a linear system whose output is the derivative of the input $L[\mathbf{x}(t)] = \frac{d\mathbf{x}(t)}{dt} = \mathbf{x}'(t)$

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We can, therefore, use the preceding results to find the autocorrelation of
$$\mathbf{x}'(t)$$
:
$$R_{XX'}(t_1, t_2) = L_2[R_{XX}(t_1, t_2)] = \frac{\partial R_{XX}(t_1, t_2)}{\partial t_2}$$

$$R_{X'X'}(t_1, t_2) = L_1[R_{XX'}(t_1, t_2)] = \frac{\partial R_{XX'}(t_1, t_2)}{\partial t_1}$$

$$R_{X'X'}(t_1, t_2) = L_1[R_{XX'}(t_1, t_2)] = \frac{\partial R_{XX'}(t_1, t_2)}{\partial t_1}$$

• Furthermore, for WSS process, $R_X(t_1, t_2) = R_X(\tau)$, $\tau = t_1 - t_2$, then

$$\frac{\partial R_X(t_1 - t_2)}{\partial t_2} = -\frac{dR_X(\tau)}{d\tau}$$

$$\frac{\partial^2 R_X(t_1 - t_2)}{\partial t_1 \partial t_2} = -\frac{d^2 R_X(\tau)}{d\tau^2}$$

$$R_{XX'}(\tau) = -R'_{XX}(\tau)$$

$$R_{XX'}(\tau) = -R''_{XX}(\tau)$$

• For the linear system and a stationary input $\mathbf{x}(t)$:

$$\mathbf{y}(t) = \mathbf{x}(t) * h(t) = \int_{-\infty}^{\infty} \mathbf{x}(t - \alpha)h(\alpha)d\alpha$$

• the autocorrelation/cross-correlation $R_Y(\tau)/R_{XY}(\tau)$ and power spectrum/cross power spectrum $S_Y(\omega)/S_{XY}(\omega)$ of the response are Fourier Transform pairs:

$$R_{XY}(\tau) = R_X(\tau) * h^*(-\tau) \quad \longleftrightarrow \quad S_{XY}(\omega) = S_X(\omega)H^*(\omega)$$

• Note that for real process, $R_{XY}(\tau) = R_X(\tau) * h(-\tau) \iff S_{XY}(\omega) = S_X(\omega)H^*(\omega)$, the conjugate in $H^*(\omega)$ refers to the Fourier Transform ' $h(-\tau)$ ', no matter it is in conjugate (complex valued) or not (real valued)

$$R_Y(\tau) = R_{XY}(\tau) * h(\tau) \qquad \longleftrightarrow \qquad S_Y(\omega) = S_{XY}(\omega)H(\omega)$$

Where $H(\omega)$ is the Fourier Transform of h(t), also referred to as Transfer function.

Convolution in the time domain is multiplication in the frequency domain, and vice versus

(时域上的卷积等价于频域上的乘,反之亦然)

$$R_{XY}(\tau) = R_X(\tau) * h^*(-\tau) \qquad \longleftrightarrow \qquad S_{XY}(\omega) = S_X(\omega)H^*(\omega)$$

$$R_Y(\tau) = R_{XY}(\tau) * h(\tau) \qquad \longleftrightarrow \qquad S_Y(\omega) = S_{XY}(\omega)H(\omega)$$

Combining the above two equations we get:

$$R_{Y}(\tau) = R_{X}(\tau) * h(\tau) * h^{*}(-\tau) = R_{X}(\tau) * \rho(\tau)$$

$$\Leftrightarrow S_{Y}(\omega) = S_{X}(\omega)H(\omega)H^{*}(\omega) = S_{X}(\omega)|H(\omega)|^{2}$$

where

$$\rho(\tau) = h(\tau) * h^*(-\tau) = \int_{-\infty}^{\infty} h(t+\tau)h^*(t)dt \leftrightarrow |H(\omega)|^2$$

In particular, if $\mathbf{x}(t)$ is white noise with average power q, then

$$R_X(\tau) = q\delta(\tau) \iff S_X(\omega) = q$$

 $R_Y(\tau) = q\rho(\tau) \iff S_Y(\omega) = q|H(\omega)|^2$

Some widely used Fourier Transform pairs

Circuit System	$H(\omega)$	h(t)
R C	$\frac{1}{1+j\omega RC}$	$\frac{1}{RC}e^{-t/RC}U(t)$
$\begin{array}{c c} C \\ \hline R \\ \hline \end{array}$	$\frac{j\omega RC}{1+j\omega RC}$	$\delta(t) - \frac{1}{RC} e^{-t/RC} U(t)$
R	$\frac{R}{R+j\omega L}$	$\frac{R}{L}e^{-Rt/L}U(t)$
R L	$\frac{j\omega L}{R+j\omega L}$	$\delta(t) - \frac{R}{L}e^{-Rt/L}U(t)$

The average power – Frequency domain

As shown in page 3 of this ppt, the average power $E\{|y(t)|^2\}$ of the output of a system is:

$$E\{|\mathbf{y}(t)|^2\} = R_Y(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_X(\omega) |H(\omega)|^2 d\omega \ge 0$$

- This equation describes the filtering properties of a system when the input is a random process. It shows, for example, if $H(\omega) = 0$ for $|\omega| > \omega_0$ and $S_{xx}(\omega) = 0$ for $|\omega| < \omega_0$, then $E\{|\mathbf{v}(t)|^2\} = 0$.
- If $\mathbf{x}(t)$ is white noise with average power q (it means $R_X(\tau) = q\delta(\tau)$), and note that the Fourier Transform of $R_X(\tau) = q\delta(\tau)$ is $S_X(\omega) = q$, we get

$$E\{|\mathbf{y}(t)|^2\} = R_Y(0) = q \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 d\omega$$

• And $E = \int_{-\infty}^{\infty} |h(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 d\omega$ is the energy of h(t). (时域和频域求积分都可以 得到系统或信号的平均功率

The average power - Time domain

We now give the average power $E\{|\mathbf{y}(t)|^2\}$ of the output of a system if its input is white noise.

• If the input to a linear system h(t) is (complex) stationary white noise with autocorrelation $R_X(\tau) = q\delta(\tau)$, then:

$$E\{y^{2}(t)\} = R_{Y}(0) = R_{X}(0) * h(0) * h^{*}(0) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{X}(0 + \alpha - \beta)h(\alpha)h^{*}(\beta)d\alpha d\beta$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} q\delta(\alpha - \beta)h(\alpha)h^{*}(\beta)d\alpha d\beta \text{ (for the } \delta(\alpha - \beta), now-zero only when } \alpha = \beta = t)$$

$$= \int_{-\infty}^{\infty} qh(t)h^{*}(t)dt = qE$$
Linear Systems – output autocorrelation

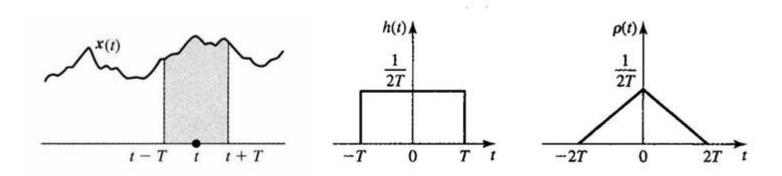
• In exam, we consider only the Steady state analysis (稳态分析) of stochastic processes, in this case If input is stationary, output is stationary, for Steady state analysis (稳态分析) only Properties when WSS ($\tau = t_1 - t_2$): (for complex process)

where $E = \int_{-\infty}^{\infty} |h(t)|^2 dt$ is the energy of h(t).

• Quick example – Moving average: The integral $\mathbf{y}(t) = \frac{1}{2T} \int_{t-T}^{t+T} \mathbf{x}(\alpha) d\alpha$ is the average of the process $\mathbf{x}(t)$ in the interval (t-T,t+T). Clearly, $\mathbf{y}(t)$ is the output of a system with input $\mathbf{x}(t)$ and impulse response a rectangular pulse as in the below Fig. The corresponding $\rho(\tau)$ is a triangle. In this case,

$$H(\omega) = \frac{1}{2T} \int_{-T}^{T} 1 \times e^{-j\omega\tau} d\tau = \frac{\sin T\omega}{T\omega} \qquad S_Y(\omega) = S_X(\omega) \frac{\sin^2 T\omega}{T^2 \omega^2}$$

- Thus $H(\omega)$ takes significant values only in the interval [-T, T] centered at the origin. Hence the moving average suppresses the high-frequency components of the input
- It is a simple low-pass filter.

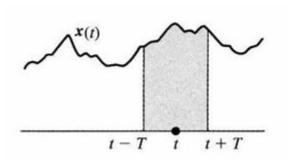


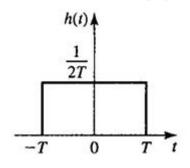
• Since $\rho(\tau)=h(\tau)*h^*(-\tau)$ is a triangle, it follows from $R_Y(\tau)=R_X(\tau)*h(\tau)*h^*(-\tau)=R_X(\tau)*\rho(\tau)$ that

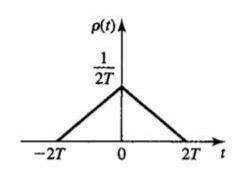
$$R_Y(\tau) = \frac{1}{2T} \int_{-2T}^{2T} \left(1 - \frac{|\alpha|}{2T} \right) R_X(\tau - \alpha) d\alpha$$

• Note that we can define the time average of the process $\mathbf{x}(t)$ as:

$$\boldsymbol{m}_T = \frac{1}{2T} \int_{-T}^T \mathbf{x}(t) dt$$







• Example 1: A stationary process x(t) with autocorrelation $R_X(\tau) = \sigma_X^2 e^{-\beta|\tau|}$ is applied to a linear system with

$$h(t) = \begin{cases} \alpha e^{-\alpha t}, & t \ge 0 \\ 0, & t < 0 \end{cases}$$

Please calculate the autocorrelation of the output, or says, calculate $R_Y(\tau)$ where $\tau=t_1-t_2$.

(This is the example 2 from last lesson, this time, we use spectrum domain kind of method to solve this problem)

Circuit System	$H(\omega)$	h(t)
R	$\frac{1}{1+j\omega RC}$	$\frac{1}{RC}e^{-t/RC}U(t)$
$\begin{array}{c c} C \\ \hline \\ R \end{array}$	$\frac{j\omega RC}{1+j\omega RC}$	$\delta(t) - \frac{1}{RC} e^{-t/RC} U(t)$
L R	$\frac{R}{R+j\omega L}$	$\frac{R}{L}e^{-Rt/L}U(t)$
R	$\frac{j\omega L}{R+j\omega L}$	$\delta(t) - \frac{R}{L}e^{-Rt/L}U(t)$

• Example 2: For the two linear system $h_Y(t)$ and $h_Z(t)$ whose Transfer functions are $H_Y(\omega) = \frac{\alpha}{2\alpha + j\omega}$ and $H_Z(\omega) = \frac{\alpha + j\omega}{2\alpha + j\omega}$, and the power spectrum of the stationary input X(t) is $S_X(\omega) = \frac{4\lambda}{4\lambda^2 + \omega^2}$, calculate $S_Y(\omega)$, $S_Z(\omega)$ and $S_{ZY}(\omega)$

Solution:

• Example 2: For the two linear system $h_Y(t)$ and $h_Z(t)$ whose Transfer functions are $H_Y(\omega) = \frac{\alpha}{2\alpha + j\omega}$ and $H_Z(\omega) = \frac{\alpha + j\omega}{2\alpha + j\omega}$, and the power spectrum of the stationary input X(t) is $S_X(\omega) = \frac{4\lambda}{4\lambda^2 + \omega^2}$, calculate $S_Y(\omega)$, $S_Z(\omega)$ and $S_{ZY}(\omega)$

Quick quesion:

Then
$$S_{YZ}(\omega) = ?$$

Powe spectrum of the output of a linear system - Differentiators

• The derivative $\mathbf{x}'(t)$ of a process $\mathbf{x}(t)$ can be considered as the output of a linear system with input $\mathbf{x}(t)$ and system function $j\omega$. Which is:

$$R_{XX'}(\tau) = -\frac{dR_X(\tau)}{d\tau} \qquad \qquad S_{XX'}(\omega) = -j\omega S_X(\omega)$$

$$R_{X'X}(\tau) = \frac{dR_X(\tau)}{d\tau} \qquad \qquad S_{X'X}(\omega) = j\omega S_X(\omega)$$

$$R_{X'X'}(\tau) = -\frac{d^2R_{XX}(\tau)}{d\tau^2} \qquad \qquad S_{X'X'}(\omega) = \omega^2 S_X(\omega)$$

• The *n*-th derivative $y(t) = \mathbf{x}^{(n)}(t)$ of $\mathbf{x}(t)$ is the output of a system with input x(t) and system function $(j\omega)^n$, thus:

$$S_Y(\omega) = |j\omega|^{2n} S_X(\omega); \quad R_Y(\tau) = (-1)^n R_X^{(2n)}(\tau)$$

Powe spectrum of the output of a linear system - Differentiators

Quick example – The differential equation

$$\mathbf{y}'(t) + c\mathbf{y}(t) = \mathbf{x}(t)$$

specifies a linear system with input $\mathbf{x}(t)$, output $\mathbf{y}(t)$, and system function $1/(j\omega + c)$.

This is because:
$$\mathbf{y}'(t) + c\mathbf{y}(t) = \mathbf{x}(t) \rightarrow j\omega Y(\omega) + cY(\omega) = X(\omega)$$
, then $H(\omega) = Y(\omega)/X(\omega) = 1/(j\omega + c)$

We assume that $\mathbf{x}(t)$ is white noise with $R_X(\tau) = q\delta(\tau)$. Then we obtain

$$S_Y(\omega) = S_X(\omega) \frac{1}{\omega^2 + c^2} = \frac{q}{\omega^2 + c^2}; \qquad R_Y(\tau) = \frac{q}{2c} e^{-c|\tau|}$$

- Note that $E\{y^2(t)\} = R_{yy}(0) = q/2c$.
- Similarly, if $\mathbf{y}''(t) + b\mathbf{y}'(t) + c\mathbf{y}(t) = \mathbf{x}(t)$, and $S_X(\omega) = q$, then we have: $H(\omega) = \frac{1}{-\omega^2 + jb\omega + c} \qquad S_Y(\omega) = \frac{q}{(c \omega^2)^2 + b^2\omega^2}$

$$H(\omega) = \frac{1}{-\omega^2 + jb\omega + c}$$
 $S_Y(\omega) = \frac{1}{(c - \omega)^2 + jb\omega + c}$

$$R_{XX'}(\tau) = -\frac{dR_X(\tau)}{d\tau} \qquad \qquad S_{XX'}(\omega) = -j\omega S_X(\omega)$$

$$R_{X'X}(\tau) = \frac{dR_X(\tau)}{d\tau} \qquad \qquad S_{X'X}(\omega) = j\omega S_X(\omega)$$

$$R_{X'X'}(\tau) = -\frac{d^2R_{XX}(\tau)}{d\tau^2} \qquad \qquad S_{X'X'}(\omega) = \omega^2 S_X(\omega)$$

Powe spectrum of the output of a linear system - Differentiators

• Example 3: given a linear time-invariant system with input x(t) and output y(t) satisfying

$$\frac{d^2y(t)}{dt^2} - 4\frac{dy(t)}{dt} + 2y(t) = 3\frac{dx(t)}{dt} + x(t)$$

Find $H(\omega)$ and $|H(\omega)|^2$

Solution:

Hilbert transforms

A system with system function

$$H(\omega) = -j \operatorname{sgn} \omega = \begin{cases} -j & \omega > 0 \\ j & \omega < 0' \end{cases} \text{ where } \operatorname{sgn} \omega = \begin{cases} 1 & \omega > 0 \\ -1 & \omega < 0 \end{cases}$$

is called a **quadrature filter** (正交滤波器). The corresponding impulse response is $1/\pi t$.

- $H(\omega)$ is all-pass with -90° phase shift; hence its response to $\cos \omega t$ is $\cos(\omega t 90^{\circ}) = \sin \omega t$ and its response to $\sin \omega t$ is $\sin(\omega t 90^{\circ}) = -\cos \omega t$.
- The response of a quadrature filter to a real process $\mathbf{x}(t)$ is denoted by $\hat{\mathbf{x}}(t)$ and it is called the Hilbert transform of $\mathbf{x}(t)$. Thus

$$\hat{\mathbf{x}}(t) = \mathbf{x}(t) * \frac{1}{\pi t} = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\mathbf{x}(\alpha)}{t - \alpha} d\alpha$$

where * is the convolution operator

Hilbert transforms

• The response of a quadrature filter to a real process $\mathbf{x}(t)$ is denoted by $\hat{\mathbf{x}}(t)$ and it is called the Hilbert transform of $\mathbf{x}(t)$. Thus

$$\hat{\mathbf{x}}(t) = \mathbf{x}(t) * \frac{1}{\pi t} = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\mathbf{x}(\alpha)}{t - \alpha} d\alpha$$

$$S_{X\hat{X}}(\omega) = jS_{XX}(\omega) \operatorname{sgn}\omega = -S_{\hat{X}X}(\omega)$$

$$S_{\hat{X}\hat{X}}(\omega) = S_{XX}(\omega)$$

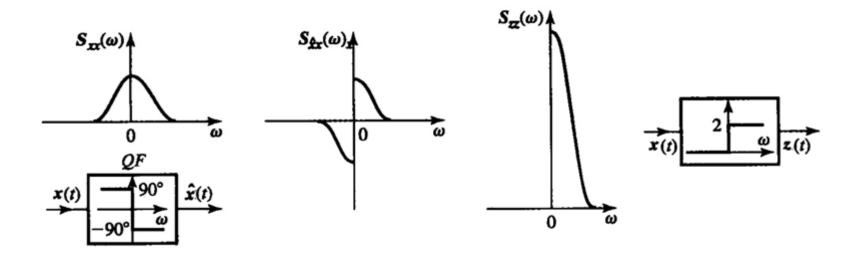
Hilbert transforms

• The complex process $\mathbf{z}(t) = \mathbf{x}(t) + j\hat{\mathbf{x}}(t)$ is calked the analytic signal associated with $\mathbf{x}(t)$. Clearly, $\mathbf{z}(t)$ is the response of the system

$$H(\omega) = 1 + j(-j\operatorname{sgn}\omega) = 2U(\omega)$$

• with input $\mathbf{x}(t)$. we have:

$$S_{ZZ}(\omega) = 4S_{XX}(\omega)U(\omega) = 2S_{XX}(\omega) + 2jS_{\hat{X}X}(\omega)$$
$$R_{ZZ}(\tau) = 2R_{XX}(\tau) + 2jR_{\hat{X}X}(\tau)$$



More examples

1: find the mean and autocorrelation of Y(t) if it is an Poisson impact process (泊松冲击过程) inputted to the Differentiator.

Hint: for Poisson impact process X(t):

$$E[X(t)] = \lambda t$$
 $R_X(t_1, t_2) = \lambda^2 t_1 t_2 + \lambda \min(t_1, t_2)$

2: given a process $X(t) = Acos(\omega_0 t + \varphi)$ where A and ω_0 are constants, φ uniformly distributed in $(0,2\pi)$. Input the X(t) to an differentiator and get X'(t), calculate the mean, autocorrelation and power spectrum of X'(t).

Reading

• This week:

• Text book: 7.3 (part of 'power spectrum together with systems')

• Red book: 3.2.2, 3.2.3

Next week:

Text book: 7.4

• Red book: 3.3 (限带过程) 注: 限带过程的讲解将以此书为主导

Experiment

• Go on with Experiment 2