

# Stochastic Process

## Ch9

40 Midterm 第8週 - 2024/02/1

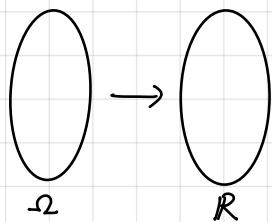
40 Final 第16週 - 2024/2/16

20 homework / Participation

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random variable  $X : \Omega \rightarrow \mathbb{R}$

$r.v X$  is a mapping from sample space  $\Omega$  to real number  $\mathbb{R}$



discrete : Bernoulli, Binomial, Geometric, Poisson ex:  $P(X=k) = \frac{\lambda^k e^{-\lambda}}{k!} \quad \forall k \in [0, 1, 2, \dots, k]$

RVs continuous : uniform  $[a, b]$ , Gaussian  $X \sim N(\mu, \sigma^2)$

other

$$P(X \leq t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

NOT probability, 積分後才是  $\rightarrow P(X \in [a, b]) = \int_a^b f_X(x) dx$

probability density function :  $f_X(x)$

Random variable  $X$



Random vector  $X = (X_1, X_2, \dots, X_n)$

(Ex)  $X$  is a  $n$ -dimensional Gaussian random vector  $X \sim N(\mu_X, \Lambda_X)$

$$\cdot \mu_X = E[X] \in \mathbb{R}^{n \times 1}$$

$$\cdot \Lambda_X = E[(X - \mu_X)(X - \mu_X)^T] \in \mathbb{R}^{n \times n} \text{ Auto-covariance matrix of } X$$

$$\cdot f_X(\vec{x}) = \frac{1}{\sqrt{(2\pi)^n \det(\Lambda_X)}} \cdot e^{-\frac{1}{2} \underbrace{(x - \mu_X)^T \Lambda_X^{-1} (x - \mu_X)}_{(x - \mu_X)^T (x - \mu_X) = \|x - \mu_X\|^2}} \geq 0$$

$$\vec{x} = (X_1, X_2, \dots, X_n)^T$$

$$(x - \mu_X)^T (x - \mu_X) = \|x - \mu_X\|^2$$

Theorem 1.

$$\Lambda_X \triangleq E[(X - \mu_X)(X - \mu_X)^T]$$

is a symmetric and semi-definite matrix

$$\cdot \Lambda_X^{-1} = \Lambda_X$$

$$\cdot \vec{u}^T \Lambda_X \vec{u} \geq 0 \quad \forall \vec{u} \in \mathbb{R}^{n \times 1}$$

1.  $\Lambda_X^{-1}$  exist  $\leftarrow \det(\Lambda_X) > 0$

2.  $\det(\Lambda_X) \geq 0$

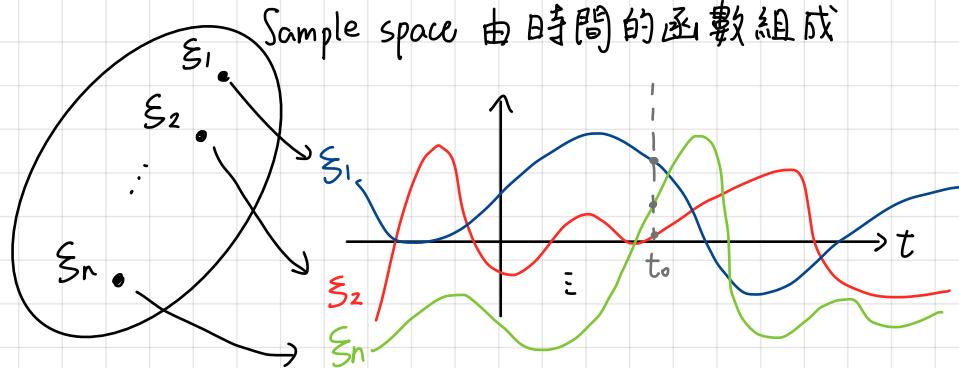
# Chapter 9

## 9-1 Definition

Stochastic Process  $X(t, \xi)$

is a two-variable function

$$X: \mathbb{R} \times \Omega \rightarrow \Omega$$



such that

(1) for each fixed  $t_0 \in \mathbb{R}$   $X(t_0)$  is a random variable

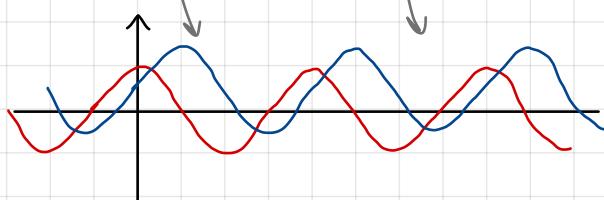
(2) for each fixed  $\xi \in \Omega$ ,  $\{X(t), t \in \mathbb{R}\}$  is a deterministic real-value function

Ex 9-4.

$$X(t) = r \cos(\omega t + \varphi), \forall t \in \mathbb{R}$$

$$X(t, \xi) \stackrel{\Delta}{=} r(\xi) \cos(\omega t + \varphi(\xi)) \quad \forall t \in \mathbb{R} \quad \forall \xi \in \Omega$$

- $r$  and  $\varphi$  are indep continuous rVs
- $\varphi \sim U(-\pi, \pi)$
- $r \perp \varphi$  (statistically independent)



$$\begin{aligned} r(\xi_1) &= 1.5 \\ \varphi(\xi_1) &= \frac{\pi}{2} \\ r(\xi_2) &= 1 \\ \varphi(\xi_2) &= 0 \end{aligned}$$

$$\begin{aligned} \textcircled{1} \quad E[X(t)] &= E[r \cos(\omega t + \varphi)] \\ &= E[r] \cdot E[\cos(\omega t + \varphi)] = 0 \\ &= 0 \quad \text{# 常數} \end{aligned} \quad \begin{aligned} \textcircled{1} \quad E[\cos(\omega t + \varphi)] &= \int_{-\pi}^{\pi} \cos(\omega t + \theta) \cdot \frac{1}{2\pi} d\theta \\ &= \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos(\omega t + \theta) d\theta \\ &= \frac{1}{2\pi} \cdot \sin(\omega t + \theta) \Big|_{-\pi}^{\pi} \\ &= \frac{1}{2\pi} \cdot [\sin(\omega t + \pi) - \sin(\omega t - \pi)] \\ &= 0 \end{aligned}$$

$$\textcircled{2} \quad R_{XX}(t_1, t_2) = E[r \cos(\omega t_1 + \varphi) \cdot r \cos(\omega t_2 + \varphi)]$$

$$= E[r^2] \cdot E[\cos(\omega t_1 + \varphi) \cdot \cos(\omega t_2 + \varphi)]$$

$$= E[r^2] \cdot \frac{1}{2} E[\cos(\omega t_1 + \omega t_2 + 2\varphi) + \cos(\omega t_1 - \omega t_2)] \stackrel{=0}{=} 0$$

$$= \frac{1}{2} E[r^2] \cdot \cos(\omega(t_1 - t_2)) \quad \text{#}$$

$$\tau = t_1 - t_2$$

$$\rightarrow E[\cos(\omega t_1 + \omega t_2 + 2\varphi)]$$

$$\begin{aligned} &= \frac{1}{2\pi} \int_{-\pi}^{\pi} \cos(\omega t_1 + \omega t_2 + 2\theta) d\theta \\ &= 0 \end{aligned}$$

### Ex 9-3

$X(t)$  is a random process

$$S \triangleq \int_a^b X(t) dt$$

$$X(t) \rightarrow \boxed{\int_a^b (\cdot) dt} \rightarrow S$$

$$S(\xi) \triangleq \int_a^b X(t, \xi) dt, \forall t \in \Omega$$

(1)  $S : \Omega \rightarrow \mathbb{R}$  is a random variable

$$E[X(t)] = \int_{-\infty}^{\infty} u f_{X(t)}(u) du$$

(2) •  $\eta_x(t) \triangleq E[X(t)], \forall t \in \mathbb{R}$

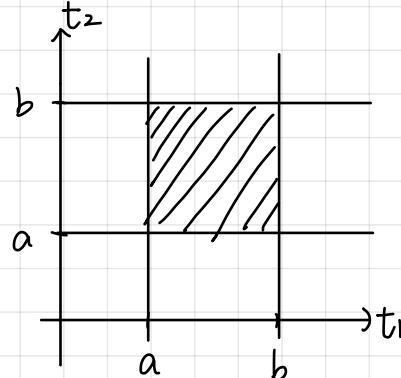
在不同樣本點之間做平均

- $R_{xx}(t_1, t_2) \triangleq E[X(t_1)X(t_2)], \forall t_1, t_2 \in \mathbb{R}$

- $C_{xx} \triangleq R_{xx}(t_1, t_2) - \eta_x(t_1)\eta_x(t_2)$

Obtain  $E[S]$  and  $E[S^2]$  in terms of statistical property of  $X(t)$

$$\textcircled{1} E[S] = E\left[\int_a^b X(t) dt\right] = \int_a^b \underline{\eta_x(t)} dt$$



$$\textcircled{2} E[S^2] = E\left[\int_a^b X(t_1) dt_1 \cdot \int_a^b X(t_2) dt_2\right]$$

$$= E\left[\int_a^b \int_a^b X(t_1)X(t_2) dt_1 dt_2\right]$$

$$= \int_a^b \int_a^b E[X(t_1)X(t_2)] dt_1 dt_2$$

$$= \int_a^b \int_a^b \underline{R_{xx}(t_1, t_2)} dt_1 dt_2$$

2024 9/2

Let  $X(t)$  and  $y(t)$  be two stochastic process (SP)

$Z(t) \triangleq X(t) + jy(t)$  is complex SP

$$(1) \eta_z(t) \triangleq E[Z(t)] = E[X(t)] + jE[y(t)], \forall t \in \mathbb{R}$$

$$(2) R_{zz}(t_1, t_2) \triangleq E[Z(t_1) Z^*(t_2)], \forall t_1, t_2 \in \mathbb{R}$$

Lemma 1: Let  $Z(t)$  be complex SP

$$(1) R_{zz}(t, t) \geq 0, \forall t \in \mathbb{R}$$

positive-definite

$$(2) \sum_{i=1}^n \sum_{k=1}^n a_i a_k^* R_{zz}(t_i, t_k) \geq 0, \forall n \in \mathbb{N}, a = (a_1 \dots a_n) \in \mathbb{C}^n \quad (t_1, t_2, \dots, t_n) \in \mathbb{R}^n$$

proof:

$$\textcircled{1} R_{zz}(t, t) = E[Z(t) Z^*(t)] = E[|Z(t)|^2] \geq 0 \quad \text{※}$$

energy/power of  $Z(t)$

\textcircled{2}

(i) Consider a fixed  $n$ ,  $a = (a_1, a_2, \dots, a_n) \in \mathbb{C}^n$  and  
 $t \triangleq (t_1, t_2, \dots, t_n) \in \mathbb{R}^n$ , Define  $U = \sum_{i=1}^n a_i Z(t_i)$

(ii) Then,  $U$  is a complex random variable. Thus  $E[|U|^2] \geq 0$

$$\text{(iii)} E[|U|^2] = E[U U^*]$$

$$\begin{aligned} &= E\left[\sum_{i=1}^n a_i Z(t_i) \cdot \sum_{k=1}^n a_k^* Z^*(t_k)\right] \\ &= E\left[\sum_{i=1}^n \sum_{k=1}^n a_i a_k^* Z(t_i) Z^*(t_k)\right] \\ &= \sum_{i=1}^n \sum_{k=1}^n a_i a_k^* R_{zz}(t_i, t_k) \geq 0 \quad \text{※} \end{aligned}$$

A matrix  $A \in \mathbb{C}^{n \times n}$  is positive-definite if

$$V^H A V \geq 0, \forall V \in \mathbb{C}^n$$

Ex 9-7

$a$  is RV,  $w \in \mathbb{R}$

$$X(t) \triangleq a \cdot e^{j\omega t}, \forall t \in \mathbb{R}$$

depend on  $t$  is general.

but become constant iff  $E[a] = 0$

$$(1) \eta_x(t) = E[a] e^{j\omega t} \quad \text{※}$$

$$(2) R_{xx}(t_1, t_2) = E[a e^{j\omega t_1} a^* e^{-j\omega t_2}] = E[|a|^2] e^{j\omega(t_1 - t_2)} \quad \text{※}$$

Definition:

A complex RP  $x(t)$  is said to be WSS, if

$$\textcircled{1} \quad \eta_x(t) = \eta_x \in C, \forall t \in \mathbb{R}$$

$$\textcircled{2} \quad R_{xx}(t_1+s, t_2+s) = R_{xx}(t_1, t_2) = R_{xx}(t_1 - t_2), \forall t_1, t_2 \in \mathbb{R}$$

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If  $x(t)$  is WSS.

$$\begin{aligned} \bullet \quad R_{xx}(\tau) &\triangleq E[x(t+\tau)x^*(t)] \\ &= E[x(t+\frac{\tau}{2})x^*(t-\frac{\tau}{2})] \end{aligned}$$

$$\begin{aligned} \bullet \quad R_{xx}(0) &= E[x(t)x^*(t)] \\ &= \underbrace{E[|x(t)|^2]}_{\text{energy}} \geq 0 \end{aligned}$$

## Ex 9-12

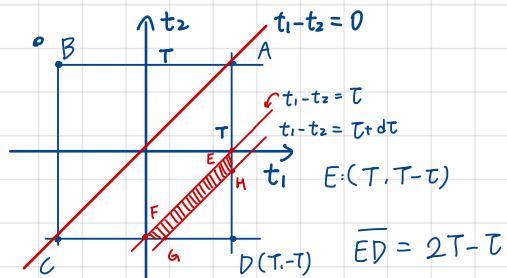
$X(t)$  is WSS

$$S \triangleq \int_{-T}^T X(t) dt \iff S(\omega) \triangleq \int_{-T}^T X(t, \omega) dt \quad \forall \omega \in \Omega$$

$$\text{Claim: } \sigma_S^2 = \int_{-T}^T \int_{-T}^T C_{xx}(t_1 - t_2) dt_1 dt_2 \\ = \int_{-2T}^{2T} (2T - |\tau|) C_{xx}(\tau) d\tau$$

proof

$$\begin{aligned} \bullet E[S] &= \int_{-T}^T E[x(t)] dt & \bullet \sigma_S^2 &= E[(S - E[S])^2] \\ &= \int_{-T}^T \eta_x(t) dt & &= E\left[\left(\int_{-T}^T X(t) dt - \int_{-T}^T \eta_x dt\right)^2\right] \\ &= \int_{-T}^T \eta_x dt & &= E\left[\int_{-T}^T X(t_1) - \eta_x dt_1 \cdot \int_{-T}^T X(t_2) - \eta_x dt_2\right] \\ &= 2T \cdot \eta_x & &= E\left[\int_{-T}^T \int_{-T}^T (X(t_1) - \eta_x)(X(t_2) - \eta_x) dt_1 dt_2\right] \\ & & &= \int_{-T}^T \int_{-T}^T E[(X(t_1) - \eta_x)(X(t_2) - \eta_x)] dt_1 dt_2 \\ & & &= \int_{-T}^T \int_{-T}^T C_{xx}(t_1, t_2) dt_1 dt_2 \end{aligned}$$



Consider the trapezoid  $\square EFGH$

the area  $\square EFGH$  is

$$= \frac{1}{2} (\overline{EF} + \overline{GH}) \cdot \frac{d\tau}{\sqrt{2}}$$

$$\cong \frac{1}{2} \cdot 2 \overline{EF} \cdot \frac{d\tau}{\sqrt{2}}$$

$$= (2T - \tau) \sqrt{2} \cdot \frac{d\tau}{\sqrt{2}}$$

$$\therefore \int_{-T}^T \int_{-T}^T C_{xx}(t_1 - t_2) dt_1 dt_2$$

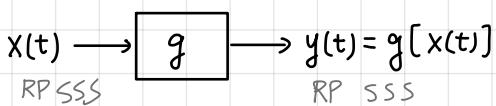
$$= \int_{-2T}^{2T} (2T - |\tau|) C_{xx}(\tau) d\tau *$$

$$\iint_{\square EFGH} C_{xx}(t_1 - t_2) dt_1 dt_2$$

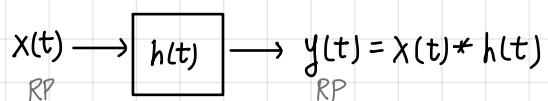
$$\approx (\text{Area } \square EFGH) C_{xx}(\tau)$$

## 9-2 system with stochastic input

(1) Memoryless system :  $g: \mathbb{R} \rightarrow \mathbb{R}$



(2) LTI



### Theorem 1.

Consider a memoryless system characterized by  $g$

- If the input  $X(t)$  is SSS, the output process  $y(t) = g[X(t)]$  is also SSS
- If in addition, the system  $g: y_1 = g(x_1), y_2 = g(x_2) \dots y_n = g(x_n)$  has a unique solution then,  $f_Y(y_1, y_2, \dots, y_n; t_1, t_2, \dots, t_n) = f_X(x_1, \dots, x_n; t_1, \dots, t_n) / |g'(x_1) g'(x_2) \dots g'(x_n)|$

### Theorem 0

Let  $X = (X_1, X_2, \dots, X_n)^T$  be a random vector

$g_1, g_2, \dots, g_n: \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $Y_k \triangleq g_k(X), \forall 1 \leq k \leq n$

Given  $(y_1, y_2, \dots, y_n) \in \mathbb{R}^n$ , if there exist a unique  $x \in \mathbb{R}^n$  such that

$g_k(x) = Y_k$  has a unique solution

$$f_Y(y_1, y_2, \dots, y_n) = \frac{f_X(x_1, x_2, \dots, x_n)}{|J(x_1, x_2, \dots, x_n)|}; J(x_1, \dots, x_n) = \begin{vmatrix} \frac{\partial g_1}{\partial x_1} & \dots & \frac{\partial g_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_n}{\partial x_1} & \dots & \frac{\partial g_n}{\partial x_n} \end{vmatrix}$$

- The  $n$ -th order CDF of  $X(t)$  at  $(t_1, t_2, \dots, t_n)$  ;

$$F_X(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n) \triangleq P(X(t_1) \leq x_1, X(t_2) \leq x_2, \dots, X(t_n) \leq x_n)$$

$$\underbrace{\quad}_{F_{X(t_1), X(t_2), \dots, X(t_n)}(x_1, x_2, \dots, x_n)}$$

- The  $n$ -th order PDF of  $X(t)$  at  $(t_1, \dots, t_n)$  ;

$$f_X(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n) \triangleq \frac{\partial^n F_X(x_1, \dots, x_n; t_1, \dots, t_n)}{\partial x_1 \partial x_2 \dots \partial x_n}$$

$$\underbrace{\quad}_{f_{X(t_1), X(t_2)}(x_1, x_2)}$$

Definite :

A RP  $X(t)$  is said to be "strict-sense stationary" (SSS)

if for each  $n \in \mathbb{N}$   $(t_1, t_2, \dots, t_n) \in \mathbb{R}^n$

$$f_X(x_1, x_2, \dots, x_n; t_1 + s, t_2 + s, \dots, t_n + s)$$

$$= f_X(x_1, x_2, \dots, x_n; t_1, t_2, \dots, t_n)$$

## Proof of Theorem 1

① Consider a fixed  $n \in N$ ,  $(t_1, t_2, \dots, t_n) \in \mathbb{R}^n$  and  $(y_1, y_2, \dots, y_n) \in \mathbb{R}^n$

② If there exist a  $k \in \{1, 2, \dots, n\}$  such that

$g_k(x) = y_k$  has no solution,  $f_Y(y_1, \dots, y_n) = 0$

③ Consider the non-trivial case in which  $g_k(x) = y_k$

has unique solution for each  $k$

Define  $g_k(x) = g(x_k)$ ,  $\forall x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ ,  $g_k : \mathbb{R}^n \rightarrow \mathbb{R}$

Then, if  $i \neq j$ ,  $\frac{\partial g_i}{\partial x_j} = 0$

On the other hand, if  $i = j$ ,  $\frac{\partial g_i}{\partial x_i} = g'(x_i)$

$$\textcircled{4} \quad J(x_1, x_2, \dots, x_n) = \begin{bmatrix} g'(x_1) & & & \\ & g'(x_2) & & \\ & & \ddots & \\ & & & g'(x_n) \end{bmatrix} \rightarrow \det(J(x_1, x_2, \dots, x_n)) = \prod_{i=1}^n g'(x_i)$$

⑤ Consider  $c \in \mathbb{R}$

$$f_Y(y_1, \dots, y_n; t_1 + c, \dots, t_n + c)$$

$$= \frac{f_X(x_1, \dots, x_n; t_1 + c, \dots, t_n + c)}{\prod_{k=1}^n g'(x_k)} = \frac{f_X(x_1, \dots, x_n; t_1, \dots, t_n)}{\prod_{k=1}^n g'(x_k)}$$

$$= f_Y(y_1, \dots, y_n; t_1, \dots, t_n)$$

⑥ Hence,  $y(t)$  is SSS \*

Ex.

$$y(t) = x^2(t) \quad (g(x) \triangleq x^2)$$

$$g' = 2x$$

$$g'(\pm\sqrt{y}) = \pm 2\sqrt{y}$$

$$f_Y(y; t) = [f_X(\sqrt{y}; t) + f_X(-\sqrt{y}; t)] \frac{1}{2\sqrt{y}}$$

$$= \frac{f_X(\sqrt{y}; t)}{|g'(\sqrt{y})|} + \frac{f_X(-\sqrt{y}; t)}{|g'(-\sqrt{y})|}$$

## LTI System

$$x(t) \xrightarrow[\text{RP}]{h(t)} y(t) = x(t) * h(t) = \int x(t-\tau)h(\tau) d\tau$$

- $E[y(t)] = \eta_x(t) * h(t)$
- $R_{XY}(t_1, t_2) = E[X(t_1) \cdot Y(t_2)], \forall t_1, t_2 \in \mathbb{R}$

Theorem 9-2

$$(a) R_{XY}(t_1, t_2) = \int_{-\infty}^{\infty} R_{XX}(t_1, t_2 - \alpha) h(\alpha) d\alpha$$

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

$$\curvearrowleft L_1 \{ R_{XY}(t_1, t_2) \}$$

$$\curvearrowright L_2 \{ R_{XX}(t_1, t_2) \}$$

## Differentiator

$$x(t) \xrightarrow{L} y(t) = x'(t) \leftrightarrow y(t; \omega) = \frac{\partial X(t, \omega)}{\partial t}$$

$$\bullet \quad \eta_x(t) = L \{ \eta_x(t) \} = \eta'_x(t)$$

$$\bullet \quad \text{is LTI system : } \frac{d}{dt}[x(t-t_0)] = x'(t-t_0) = y(t-t_0)$$

$$\bullet \quad H(\omega) = j\omega$$

$$\curvearrowleft R_{XX}(t_1 - t_2, 0) \triangleq R_{XX}(t_1, t_2)$$

$$(1) R_{XX'}(t_1, t_2) = L_2 [R_{XX}(t_1, t_2)] = \frac{\partial R_{XX}(t_1, t_2)}{\partial t_2}$$

$$(2) R_{XX''}(t_1, t_2) = L_1 [R_{XX'}(t_1, t_2)] = \frac{\partial R_{XX}(t_1, t_2)}{\partial t_1 \partial t_2}$$

若  $x(t)$  is WSS,  $y(t) \triangleq x'(t)$

$$\bullet \quad \eta_Y(t) = 0 \quad R_{XX'}(t_1, t_2) = R_{XX'}(t_1 - t_2, 0) = R_{XX'}(t_1, t_2) = \frac{\partial R_{XX}(t_1 - t_2)}{\partial t_2}$$

$$\bullet \quad R_{XX'}(\tau) = -R_{XX}'(\tau)$$

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

$$\bullet \quad R_{XX''}(\tau) = -R_{XX}'''(\tau)$$

則  $y(t) = x'(t)$  is also WSS

## 9-3 The Power Spectrum

In 9-3, all RPs are WSS

$$\begin{cases} S_x(\omega) \triangleq \int_{-\infty}^{\infty} R_{xx}(\tau) e^{-j\omega\tau} d\tau = \mathcal{F}\{R_{xx}(\tau)\} \\ R_{xx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_x(\omega) e^{j\omega\tau} d\omega = \mathcal{F}^{-1}\{S_x(\omega)\} \end{cases}$$

Table 9-1. 傳試會合

$R_{xx}(\tau)$	$S_x(\omega)$
$\delta(\tau)$	1
1	$2\pi \delta(\omega)$
$e^{j\beta\tau}$	$2\pi \delta(\omega - \beta)$
$e^{-\alpha \tau }$	$\frac{2\alpha}{\alpha^2 + \omega^2}$
$e^{-\alpha\tau^2}$	$\sqrt{\frac{\pi}{\alpha}} e^{-\frac{\omega^2}{4\alpha}}$
$\begin{cases} 1 &  \tau  \leq T/2 \\ 0 &  \tau  > T/2 \end{cases}$	$\frac{4\sin^2(\omega T/2)}{\pi \omega^2}$

Lemma 1.

If  $R_{xx}(\tau) = e^{-\alpha|\tau|}$ , where  $\alpha > 0$

$$S_x(\omega) = \frac{2\alpha}{\alpha^2 + \omega^2}$$

proof.

$$\begin{aligned} ① S_x(\omega) &= \int_{-\infty}^{\infty} R_{xx}(\tau) e^{-j\omega\tau} d\tau \\ &= \int_{-\infty}^0 e^{\alpha\tau} e^{-j\omega\tau} d\tau + \int_0^{\infty} e^{-\alpha\tau} e^{-j\omega\tau} d\tau \\ &= \frac{e^{(\alpha-j\omega)\tau}}{\alpha-j\omega} \Big|_{-\infty}^0 + \frac{e^{-(\alpha+j\omega)\tau}}{-(\alpha+j\omega)} \Big|_0^{\infty} \\ &= \frac{1}{\alpha-j\omega} + \frac{1}{\alpha+j\omega} = \frac{2\alpha}{\alpha^2 + \omega^2} \end{aligned}$$

Theorem 9-4.

$$R_{xy}(\tau) = R_{xx}(\tau) * h^*(-\tau) \longleftrightarrow S_{xy}(\omega) = S_{xx}(\omega) H^*(\omega)$$

$$R_{yy}(\tau) = R_{xy}(\tau) * h(\tau) \longleftrightarrow S_{yy}(\omega) = S_{xy}(\omega) H(\omega) = S_{xx}(\omega) |H(\omega)|^2$$

proof.  $R_{xy}(\tau) = R_{xx}(\tau) * h^*(-\tau) \longleftrightarrow S_{xy}(\omega) = S_{xx}(\omega) H^*(\omega)$

- $R_{xy}(\tau) = E[x(t+\tau) y^*(t)]$

$$= E[x(t+\tau) \int_{-\infty}^{\infty} x^*(t-\alpha) h^*(\alpha) d\alpha]$$

$$= \int_{-\infty}^{\infty} E[x(t+\tau) x^*(t-\alpha)] h^*(\alpha) d\alpha$$

$$= \int_{-\infty}^{\infty} R_{xx}(\tau+\alpha) h^*(\alpha) d\alpha \quad \text{--- ①}$$

$$\text{let } g(\tau) = h^*(-\tau)$$

$$R_{xx}(\tau) * h^*(-\tau) = R_{xx}(\tau) * g(\tau)$$

$$= \int_{-\infty}^{\infty} R_{xx}(\tau-\beta) g(\beta) d\beta$$

$$= \int_{-\infty}^{\infty} R_{xx}(\tau+\alpha) g(-\alpha) d\alpha$$

$$= \int_{-\infty}^{\infty} R_{xx}(\tau+\alpha) h^*(\alpha) d\alpha \quad \text{--- ②}$$

比較 ① & ②.  $R_{xy}(\tau) = R_{xx}(\tau) * h^*(-\tau)$

- $S_{xy}(\omega) \triangleq \mathcal{F}\{R_{xy}(\tau)\} = \int_{-\infty}^{\infty} R_{xy}(\tau) e^{-j\omega\tau} d\tau$

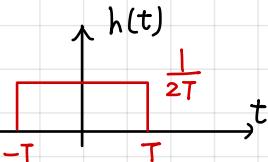
$$\rightarrow \int_{-\infty}^{\infty} h^*(-\tau) e^{-j\omega\tau} d\tau = \left[ \int_{-\infty}^{\infty} h(\alpha) e^{-j\omega\alpha} d\alpha \right]^* = H^*(\omega)$$

故  $S_{xy}(\omega) = S_{xx}(\omega) \cdot H^*(\omega)$

Ex 9-25

- $y(t) = \frac{1}{2T} \int_{t-T}^{t+T} x(\alpha) d\alpha$

- Define  $h(t)$



<sol>

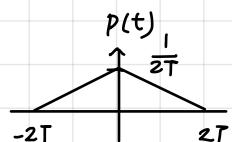
$$H(\omega) = \int_{-T}^T \frac{1}{2T} e^{-j\omega t} dt$$

$$= \frac{1}{2T} \left[ \int_{-T}^T \cos(\omega t) dt - j \int_{-T}^T \sin(\omega t) dt \right] \text{ odd}$$

$$= \frac{\sin(\omega T)}{\omega T}$$

- $S_{yy}(\omega) = S_{xx}(\omega) |H(\omega)|^2 = S_{xx}(\omega) \frac{\sin^2(\omega T)}{\omega^2 T^2}$

- $R_{yy}(\tau) = R_{xx}(\tau) \mathcal{F}^{-1} \left\{ \frac{\sin^2(\omega T)}{\omega^2 T^2} \right\} = R_{xx}(\tau) * P(t)$



- Then  $y(t) = x(t) * h(t)$

Question:  $R_{yy}(\tau) = ?$

$S_{yy}(\omega) = ?$

If  $y(t)$  is WSS RP

$$0 \leq E[|y(t)|^2] = R_{yy}(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{yy}(\omega) d\omega$$

### Ex 9-26

- $X(t) \rightarrow h(t) \rightarrow Y(t) = X'(t)$
- $Y(\omega) = j\omega X(\omega)$
- $H(\omega) = j\omega$

$$(1) S_{yy}(\omega) = S_{xx}(\omega) |H(\omega)|^2$$

$$= S_{xx}(\omega) \cdot \omega^2$$

$$(2) R_{yy}(\tau) = \mathcal{F}^{-1}\{S_{yy}(\omega)\} \quad \omega^2 = -(j\omega)^2$$

$$= \mathcal{F}^{-1}\left\{-\underbrace{(j\omega)^2 S_{xx}(\omega)}\right\}$$

$$= -\frac{d^2}{d\tau^2} R_{xx}(\tau) \cancel{\times}$$

$$(3) R_{yy}(0) =$$

### Ex 9-27

$$\bullet y'(t) + c y(t) = X(t)$$

$$\bullet R_{xx}(\tau) = g \delta(\tau)$$

$$\bullet S_{xx}(\omega) = g$$

$$R_{yy}(\tau) = ? \quad S_{yy}(\omega) = ?$$

$$\rightarrow (j\omega + c) Y(\omega) = X(\omega)$$

$$\rightarrow H(\omega) = \frac{Y(\omega)}{X(\omega)} = \frac{1}{j\omega + c}$$

$$(1) S_{yy}(\omega) = S_{xx}(\omega) |H(\omega)|^2 = \frac{g}{\omega^2 + c^2} \cancel{\#}$$

$$(2) R_{yy}(\tau) = \mathcal{F}^{-1}\left\{\frac{g}{2c} \cdot \frac{2c}{\omega^2 + c^2}\right\}$$

$$= \frac{g}{2c} \cdot e^{-c|\tau|} \cancel{\#}$$

## 9-4 Discrete Time Process

$X[n]$ : discrete-time SP

$X(t)$ : continuous-time SP

- $\eta_x[n] \triangleq E\{X[n]\}$
  - $R_{xx}[n_1, n_2] \triangleq E\{X[n_1] \cdot X^*[n_2]\} \quad \forall n_1, n_2 \in \mathbb{Z}$
  - $C_{xx}[n_1, n_2] \triangleq R_{xx}[n_1, n_2] - \eta_x[n_1] \eta_x^*[n_2]$
- 

$$X[n] \rightarrow h[n] \rightarrow y[n] = \sum_{k=-\infty}^{\infty} X[n-k] h[k]$$

- $H(z) = \sum_{n=-\infty}^{\infty} h[n] \cdot z^{-n}$

Theorem 1.

$$R_{xy}[n_1, n_2] = \sum R_{xx}[n_1, n_2-k] h^*[k]$$

$$R_{yy}[n_1, n_2] = \sum R_{xy}[n_1-r, n_2] h[r]$$

$$S_x(z) \triangleq \sum R_x[m] z^{-m}$$

- $S_x(\omega) \triangleq S_x(e^{j\omega}) = \sum R_x[m] e^{-jm\omega}$
  - $R_x[m] = \frac{1}{2\pi} \int_{-\pi}^{\pi} S_x(e^{j\omega}) e^{jm\omega} d\omega$
- $S_x(e^{j\omega})$  is a periodic function with period  $2\pi$
- 

Consider  $X[n]$ ,  $r_k \triangleq R_x[k]$

$$T_n \triangleq \begin{bmatrix} r_0 & r_1 & r_2 & & r_n \\ r_1^* & r_0 & r_1 & & \\ r_2^* & r_1^* & r_0 & & \\ \vdots & & & & \\ & & & & r_0 \end{bmatrix} \in \mathbb{C}^{(n+1)(n+1)} \text{ is autocorrelation matrix}$$

- $T_n = T_n^H$
- $T_n$  is a Toeplitz matrix
- $T_n$  is positive semi-definite,  $u^H T_n u \geq 0$

## Theorem 2

Let  $Y = (Y_1, Y_2, \dots, Y_n)^T$  be a  $n$ -dimensional complex random vector  
Define  $R_Y \triangleq E[YY^H]$ , is a positive semi-definite matrix

proof

1. Consider  $u \in C^{n \times 1}$ ,  
$$u^H R_Y u = E[u^H Y Y^H u]$$
$$= E[|u^H Y|^2] \geq 0$$

### Ex 9-31

$R_x[m] = \alpha^{|m|}$ ;  $\forall m, |\alpha| \leq 1$ , Find  $S_x(z)$  and  $S_x(\omega)$

$$\begin{aligned} \textcircled{1} \quad S_x(z) &= \sum_{-\infty}^{\infty} R_x[m] z^{-m} \\ &= \sum_{m=-\infty}^{-1} \alpha^{-m} z^{-m} + \sum_{m=0}^{\infty} \alpha^m z^{-m} \\ &= \sum_{n=1}^{\infty} \alpha^n z^n + \sum_{n=0}^{\infty} \alpha^n z^{-n} \\ &= \frac{\alpha z}{1-\alpha z} + \frac{1}{1-\alpha z^{-1}} \end{aligned}$$

ROC:  $|\alpha z| < 1$  and  $|\frac{1}{\alpha z}| < 1$

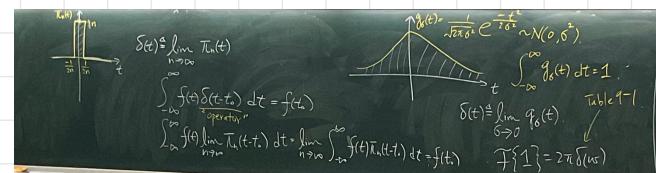
$$\Rightarrow |\alpha| < |z| < |\frac{1}{\alpha}|$$

Ex 9-32  $C_i$  is a random variable  $w_i \in \mathbb{R}$

$$X[n] = \sum C_i e^{j w_i n}, \forall n \in \mathbb{Z}$$

$$E[C_i] = 0, C_i \perp C_j$$

$$R_x[m], S_x(\omega) = ?$$



$$\textcircled{1} \quad R_x[n+m, n]$$

$$\begin{aligned} &= E\left[\sum_i C_i e^{j w_i (n+m)} \cdot \sum_j C_j^* e^{-j w_j n}\right] \\ &= E\left[\sum_i \sum_j C_i C_j^* e^{j[(w_i - w_j)n + w_i m]}\right] \\ &= \sum_i \sum_j E[C_i C_j^*] e^{j[(w_i - w_j)n + w_i m]} \\ &= \sum_i E[|C_i|^2] e^{j w_i m} \\ &= \sum_i \sigma_i^2 e^{j \beta_i m} = R_x[m] \end{aligned}$$

$$\textcircled{2} \quad S_x(f)$$

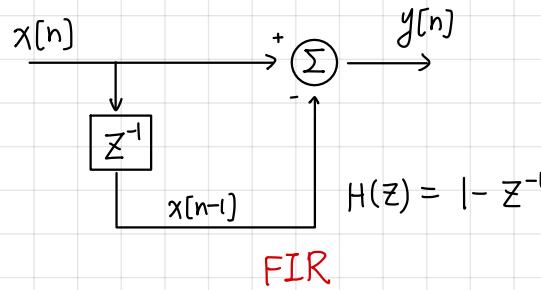
$$\begin{aligned} &= \sum_m R_x[m] e^{-j \omega m} \\ &= \sum_m \sum_i \sigma_i^2 e^{j \beta_i m} e^{-j \omega m} \\ &= \sum_i \sigma_i^2 \sum_m e^{-j(\omega - \beta_i)m} \\ &= \sum_i \sigma_i^2 \cdot 2\pi \delta(\omega - \beta_i) \end{aligned}$$

Ex 9-33

$X[n]$  is WSS,  $E[X[n]] = 0$

$$y[n] = x[n] - x[n-1]$$

$$R_{yy}[m], S_{yy}(\omega) = ?$$

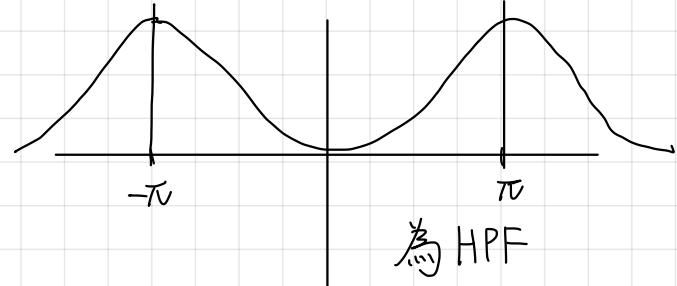


<Sol>

$$\begin{aligned} S_{yy}(z) &= S_{xx}(z) H(z) H(\frac{1}{z}) \\ &= S_{xx}(z) (1-z^{-1})(1-z) \\ &= S_{xx}(z) (2-z-z^{-1}) \end{aligned}$$

$$R_{yy}[m] = 2R_{xx}[m] - R_{xx}[m+1] - R_{xx}[m-1]$$

$$\begin{aligned} S_{yy}(\omega) &= S_{yy}(e^{j\omega}) \\ &= S_{xx}(e^{j\omega}) [2 - e^{-j\omega} - e^{j\omega}] \\ &= S_{xx}(e^{j\omega}) (2 - 2 \cos \omega) \end{aligned}$$

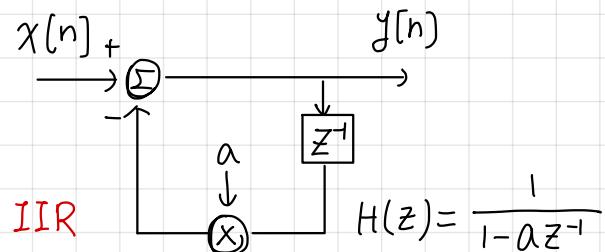


Ex 9-34

$X[n]$  is a WSS RP with zero mean:  $\alpha \in (-1, 1)$

$$y[n] - \alpha y[n-1] = x[n]$$

$$S_{xx}(z) = q$$



<Sol>

$$\begin{aligned} S_{yy}(z) &= S_{xx}(z) \cdot \frac{1}{1-\alpha z^{-1}} \cdot \frac{1}{1-\alpha z} \\ &= \frac{q}{1+\alpha^2 - \alpha(z+z^{-1})} \\ &= \frac{q\alpha^{-1}}{(\alpha+\alpha^{-1}) - (z+z^{-1})} \\ &= \frac{\alpha^2 - \alpha}{(\alpha+\alpha^{-1}) - (z+z^{-1})} \cdot \frac{q\alpha^{-1}}{\alpha^2 - \alpha} \end{aligned}$$

$$R_{yy}[m] = \frac{q\alpha^{-1}}{\alpha^2 - \alpha} \alpha^{|m|}$$

$$\begin{aligned} S_{yy}(\omega) &= S_{yy}(e^{j\omega}) \\ &= \frac{q}{1 + \alpha^2 - 2\alpha \cos \omega} \# \end{aligned}$$

