

# Chap 4 Spectral Analysis of Stochastic Processes

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# Chapter 4: Spectral Analysis



- References:

1. Stochastic Processes and Their Applications,  
Frank E. Beichelt, O211. W138  
Chap 8, Spectral analysis of stationary  
processes
2. Probability, Random Variables and Random  
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Chap 7, Spectral characteristic of random  
processes

# Chapter 4: Spectral Analysis



Content:

## 4.1 Spectral Density Functions

## 4.2 Spectral Analysis of Linear Systems

## 4.3 Spectrum of Amplitude-modulated Signals

## 4.4 Narrow-band Gaussian Processes

# 4.1 Spectral Density Functions



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4.1.1 Autospectral Density Functions

4.1.2 Wiener-Khintchine Theorem

4.1.3 Crossspectral Density Functions

4.1.4 S.D.F. of Derived Random Processes

# Average of the power



- If  $X(t)$  is ergodic, then

$$R_{XX}(0) = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T x^2(t) dt$$

- $R(0)$  represents the time average of the power of a random process  $X(t)$ .
- **Def. 1** average of the power

$$\bar{P}_X = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T x^2(t) dt$$

- **Parseval theorem**

$$\int_{-\infty}^{\infty} x^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} X^2(\omega) d\omega$$

- The representation of the average energy in terms of time and frequency of a signal.
- Thus,

$$\bar{P}_X = \lim_{T \rightarrow \infty} \frac{1}{4\pi T} \int_{-T}^T X^2(\omega) d\omega$$

# 4.1.1 Autospectral Density Functions



## Def.2 Spectral Density Function (S.D.F.)

- The S.D.F. of a random process  $X(t)$  is

$$S_{XX}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} |X_T(\omega)|^2 \quad S_{XX}(f) = \lim_{T \rightarrow \infty} \frac{1}{2T} |X_T(f)|^2$$

- (autospectral density function) (**Power Spectrum**)
- From Def.1 and Def. 2,

$$\begin{aligned} \overline{P}_X &= \lim_{T \rightarrow \infty} \frac{1}{4\pi T} \int_{-T}^T X^2(\omega) d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) d\omega = \frac{1}{\pi} \int_0^{\infty} S_{XX}(\omega) d\omega \\ &= \int_{-\infty}^{\infty} S_{XX}(f) df = 2 \int_0^{\infty} S_{XX}(f) df \end{aligned}$$

# Definitions associated with the S.D.F.



1. The  $k$ th moment of the S.D.F.

$$m_k = \int_0^{\infty} \omega^k S_{XX}(\omega) d\omega$$

2. Modal frequency  $\omega_m$

the frequency where the S.D.F. peaks.

3. Mean frequency

$$\bar{\omega}_X = \frac{\int_0^{\infty} \omega S_{XX}(\omega) d\omega}{\int_0^{\infty} S_{XX}(\omega) d\omega} = \frac{m_1}{m_0}$$

the mean frequency is not equal to the average value of the various frequencies involved in a random process; it is the mathematical mean of the S.D.F.



4. The  $k$ th moment about the mean of the S.D.F.

$$\mu_k = \int_0^\infty (\omega - \bar{\omega})^k S_{XX}(\omega) d\omega$$

Thus,

$$\mu_0 = m_0 = \int_0^\infty S_{XX}(\omega) d\omega = \pi \bar{P}_X$$

$$\mu_1 = 0$$

$$\mu_2 = m_2 - \frac{m_1^2}{m_0}$$





## 5. The bandwidth parameter of the S.D.F.

$$\varepsilon = \sqrt{1 - \frac{m_2^2}{m_0 m_4}}$$

$\varepsilon = 0$ , the random process has a narrow-band spectrum;

$\varepsilon = 1$ , the random process has a wide-band spectrum;



6. The spectral width parameter of the S.D.F.

$$\nu = \sqrt{\frac{\mu_2}{\mu_0} \frac{1}{\bar{\omega}}}$$

7. The spectral peakedness parameter

$$Q_P = \frac{\int_0^{\infty} \omega S_{XX}^2(\omega) d\omega}{\left( \int_0^{\infty} S_{XX}(\omega) d\omega \right)^2}$$

The value increases with increasing sharpness of the S.D.F.

# 4.1.1 Autospectral Density Functions



- Properties of the S.D.F.

- i) Nonnegative  $S_{XX}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} |X_T(\omega)|^2 \geq 0$

- ii) Real

- iii) Even, if  $X(t)$  is real,  $S_{XX}(\omega) = S_{XX}(-\omega)$

- iv) If  $\int_{-\infty}^{\infty} |R_{XX}(\tau)| d\tau < \infty$ , then  $S_{XX}(\omega)$  is a continuous function of  $\omega$

# 4.1 Spectral Density Functions



## Content:

4.1.1 Autospectral Density Functions

4.1.2 Wiener-Khintchine Theorem

4.1.3 Crossspectral Density Functions

4.1.4 S.D.F. of Derived Random  
Processes

# 4.1.2 Wiener-Khintchine Theorem



- The Fourier transform of the correlation function of an ergodic process:

$$\int_{-\infty}^{\infty} R_{XX}(\tau) e^{-i\omega\tau} d\tau = \lim_{T \rightarrow \infty} \frac{1}{2T} |X_T(\omega)|^2$$

- And the definition of S.D.F.,

$$S_{XX}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} |X_T(\omega)|^2$$

- Obtain,

$$S_{XX}(\omega) = \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-i\omega\tau} d\tau$$

$$R_{XX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) e^{i\omega\tau} d\omega$$

# 4.1.2 Wiener-Khintchine Theorem



## Theorem: Wiener-Khintchine Theorem

For a **weakly stationary random process**  $X(t)$ , its correlation function and the power spectrum are Fourier transform pair.

$$S_{XX}(\omega) = \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-i\omega\tau} d\tau$$

$$R_{XX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) e^{i\omega\tau} d\omega$$

## Standard Fourier Transform

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt$$

$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F(\omega) e^{i\omega t} d\omega$$

# 4.1.2 Wiener-Khintchine Theorem



- Both correlation function and power spectrum are **real** and **even** functions.

$$S_{xx}(\omega) = \int_{-\infty}^{\infty} R_{xx}(\tau) e^{-i\omega\tau} d\tau$$

$$= 2 \int_0^{\infty} R_{xx}(\tau) \cos \omega\tau d\tau$$

$$R_{xx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) e^{i\omega\tau} d\omega$$

$$= \frac{1}{\pi} \int_0^{\infty} S_{xx}(\omega) \cos \omega\tau d\omega$$

# 4.1.2 Wiener-Khintchine Theorem



- The average power and correlation function

$$\frac{1}{\pi} \int_0^{\infty} S_{XX}(\omega) d\omega = \bar{P}_X = R_{XX}(0) = E[X^2(t)]$$

- If  $E[X(t)] = 0$ ,

$$\text{Var}[X(t)] = \frac{1}{\pi} \int_0^{\infty} S_{XX}(\omega) d\omega = \bar{P}_X = R_{XX}(0)$$



# 4.1.2 Wiener-Khintchine Theorem



## *Example:* White Noise

The process with autocorrelation function,

$$R(\tau) = A\delta(\tau)$$

and power spectrum,

$$S(\omega) = A = \text{constant}$$

The **continuous white noise** is a **real, stationary, continuous-time** stochastic process with **constant spectral density**.

the “most random” stochastic process



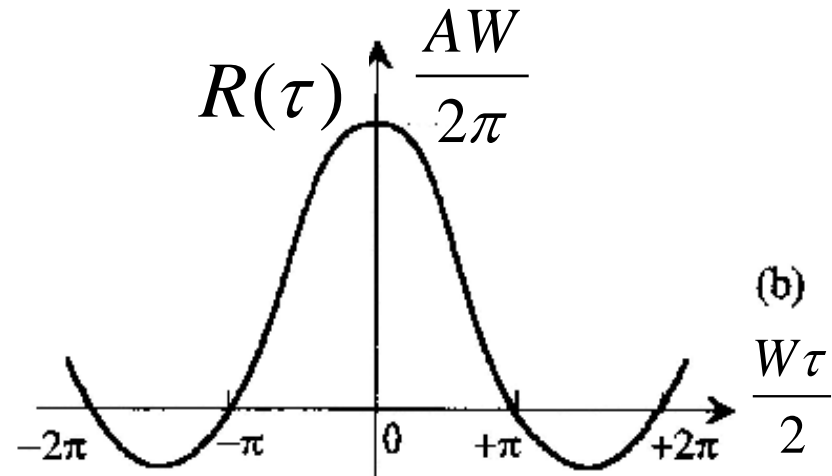
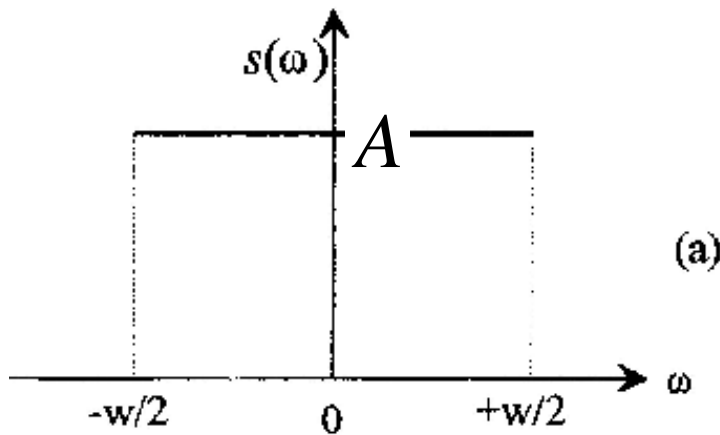
# 4.1.2 Wiener-Khintchine Theorem

*Example 2.* Band-limited White Noise

$$S(\omega) = A, \text{ where } -\frac{W}{2} < \omega < \frac{W}{2}$$

Thus,

$$R_{XX}(\tau) = \frac{1}{2\pi} \int_{-W/2}^{W/2} A e^{i\omega\tau} d\omega = \frac{A}{\pi\tau} \sin \frac{1}{2} W\tau$$



# Review 2.2.1 Stationary Processes



- weakly stationary

*Example 1.* Random phase processes

$X(t) = A \cos(\omega t + \varepsilon)$ ,  $t > 0$ , whereas  $A$  and  $\omega$  are constants and  $\varepsilon$  is random variable uniformly distributed between  $-\pi$  and  $\pi$ .

$$E[x(t)] = 0$$

$$\begin{aligned} C_{XX}(t_1, t_2) &= \frac{A^2}{2} \cos \omega_0 (t_2 - t_1) \\ &= \frac{A^2}{2} \cos \omega_0 \tau = C_{XX}(\tau) \end{aligned}$$

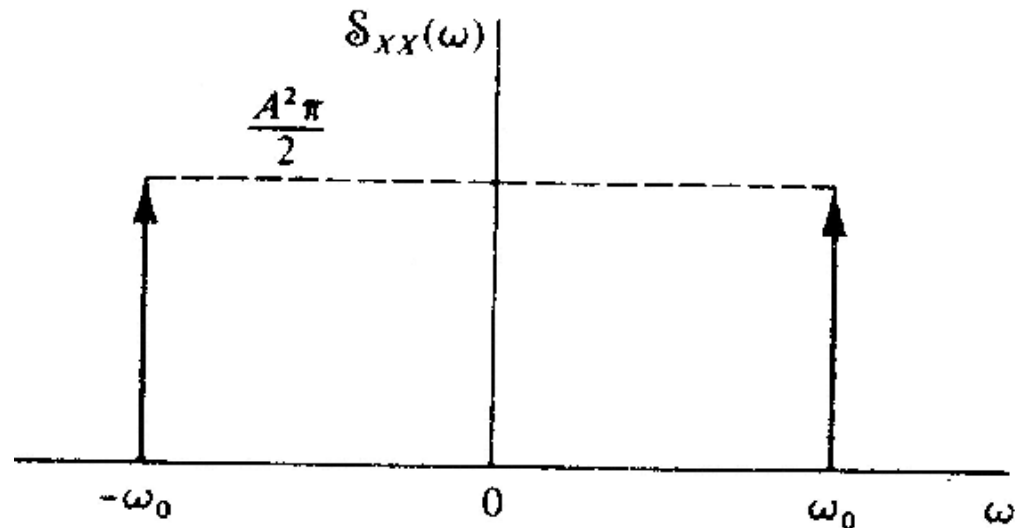
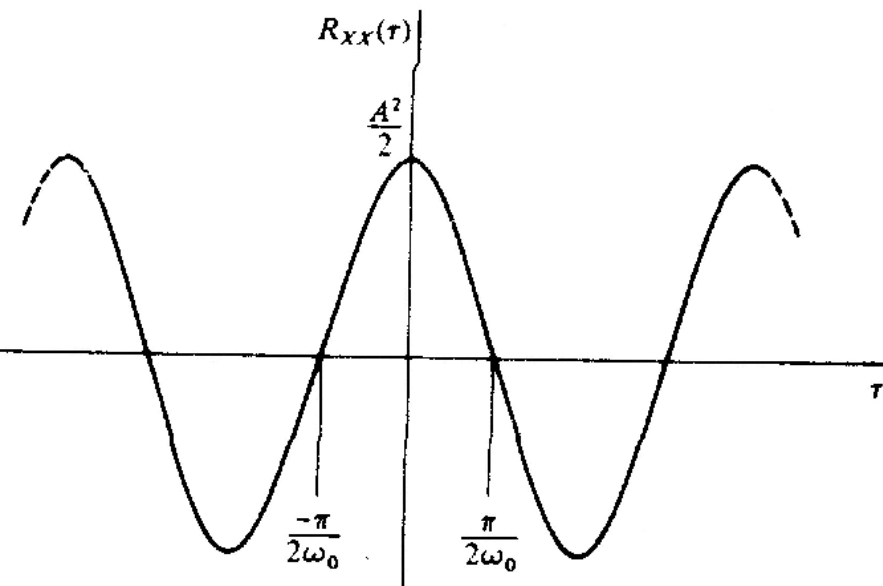
# 4.1.2 Wiener-Khintchine Theorem



## Example 3. Random Phase Processes

$$R_{XX}(\tau) = \frac{A^2}{2} \cos \omega_0 \tau$$

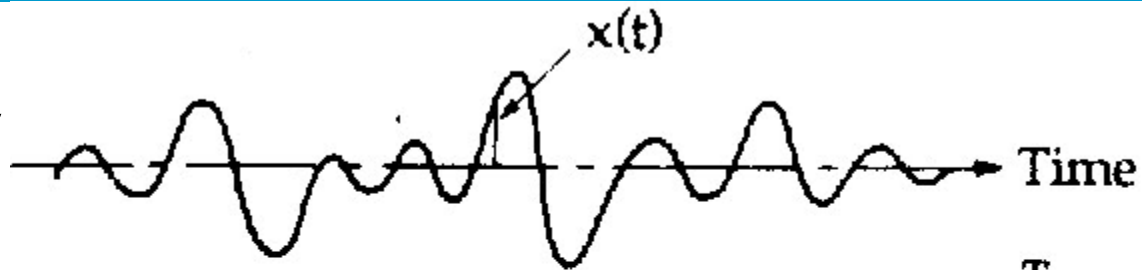
Obtain:  $S(\omega) = \frac{A^2 \pi}{2} [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$



# 4.1.2 Wiener-Khintchine Theorem



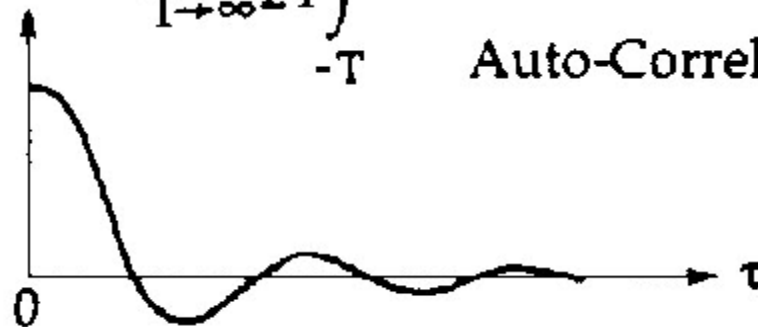
- Summary



$$\text{Ave. Energy } P_x = \frac{1}{2T} \int_{-T}^T \{x(t)\}^2 dt$$

$$R_{xx}(\tau) = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T x(t) x(t + \tau) dt$$

Auto-Correlation Function

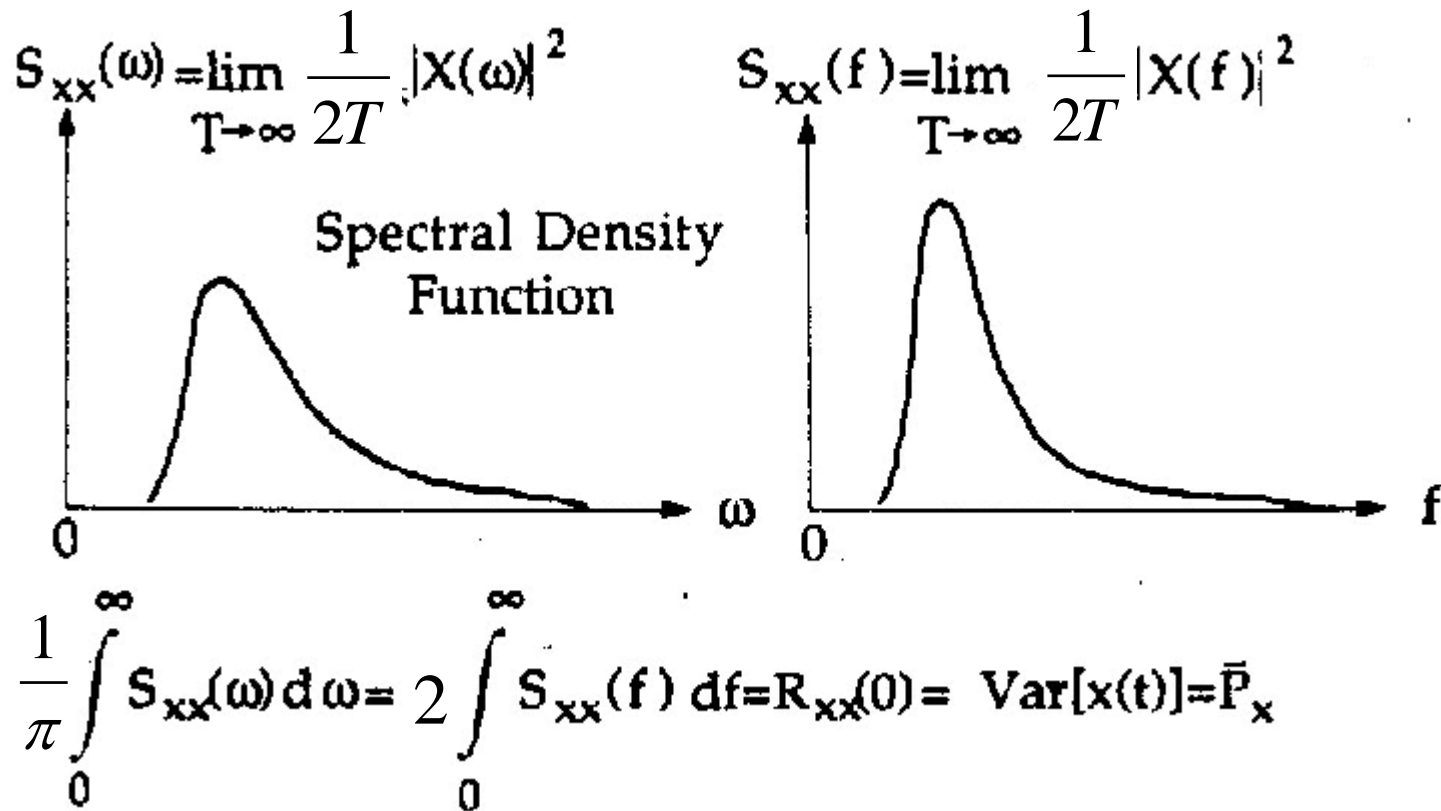


(Fourier Transform)

# 4.1.2 Wiener-Khintchine Theorem



- Summary



# 4.1.2 Wiener-Khintchine Theorem



- Summary

$R_X(\tau)$	$S_X(\omega)$
$\begin{cases} (T -  \tau )/T, &  \tau  < T, \\ 0, & \text{otherwise} \end{cases}$	$T \{\sin(\omega T/2)/(\omega T/2)\}^2 = T \operatorname{sinc}^2(fT)$
$1$	$2\pi\delta(\omega)$
$\delta(\tau)$	$1$
$\exp(-\alpha \tau )$	$2\alpha/(\alpha^2 + \omega^2)$
$\cos(\omega_0\tau)$	$\pi\delta(\omega - \omega_0) + \pi\delta(\omega + \omega_0)$
$\exp(-\alpha \tau ) \cos(\omega_0\tau)$	$\alpha/[\alpha^2 + (\omega - \omega_0)^2] + \{\alpha/[\alpha^2 + (\omega + \omega_0)^2]\}$
$2W \operatorname{sinc}(2W\tau) = \sin(2\pi W\tau)/\pi\tau$	$\begin{cases} 1, &  \omega  \leq 2\pi W \\ 0, & \text{otherwise} \end{cases}$
$\operatorname{sinc}(u) = \sin(\pi u)/\pi u, -\infty < u < \infty$	

# 4.1.2 Wiener-Khintchine Theorem



*e.g.* Suppose the continuous-time random process  $X(t)$  has autocorrelation function

$$R_X(\tau) = 1 + e^{-\alpha|\tau|}, \quad -\infty < \tau < \infty, \quad \alpha > 0$$

Find the spectral density function for  $X(t)$ .

$$S_X(\omega) = 2\pi\delta(\omega) + \frac{2\alpha}{\alpha^2 + \omega^2}$$



# 4.1 Spectral Density Functions



## Content:

4.1.1 Autospectral Density Functions

4.1.2 Wiener-Khintchine Theorem

4.1.3 Crossspectral Density  
Functions

4.1.4 S.D.F. of Derived Random  
Processes

## 4.1.3 Cross-Spectral Density Functions



### 1. Definition of Cross-Spectral Density Function (CS.D.F)

Cross-Spectral Density Function of two random processes  $X(t)$  and  $Y(t)$  is

$$S_{XY}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} X^*(\omega) Y(\omega)$$

$$S_{YX}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} X(\omega) Y^*(\omega)$$

## 4.1.3 Cross-Spectral Density Functions



### Theorem: Wiener-Khintchine Theorem

For two **jointly stationary random processes**  $X(t)$  and  $Y(t)$ , their cross-correlation function and cross-power density function are Fourier transform pair.

$$S_{XY}(\omega) = \int_{-\infty}^{\infty} R_{XY}(\tau) e^{-i\omega\tau} d\tau \quad S_{YX}(\omega) = \int_{-\infty}^{\infty} R_{YX}(\tau) e^{-i\omega\tau} d\tau$$

$$R_{XY}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XY}(\omega) e^{i\omega\tau} d\omega \quad R_{YX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{YX}(\omega) e^{i\omega\tau} d\omega$$

Standard Fourier Transform

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt$$
$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F(\omega) e^{i\omega t} d\omega$$

## 4.1.3 Cross-Spectral Density Functions



### 2. Real part and imaginary part

$$\begin{aligned} S_{XY}(\omega) &= \int_{-\infty}^{\infty} R_{XY}(\tau) e^{-i\omega\tau} d\tau \\ &= \int_{-\infty}^{\infty} R_{XY}(\tau) \cos \omega\tau d\tau - i \int_{-\infty}^{\infty} R_{XY}(\tau) \sin \omega\tau d\tau \\ &= C_{XY}(\omega) + iQ_{XY}(\omega) \end{aligned}$$

$$\begin{aligned} C_{XY}(\omega) &= \int_{-\infty}^{\infty} R_{XY}(\tau) \cos \omega\tau d\tau \\ &= \int_{-\infty}^0 R_{XY}(\tau) \cos \omega\tau d\tau + \int_0^{\infty} R_{XY}(\tau) \cos \omega\tau d\tau \\ &= \int_0^{\infty} R_{XY}(-\tau) \cos \omega\tau d\tau + \int_0^{\infty} R_{XY}(\tau) \cos \omega\tau d\tau \\ &= \int_0^{\infty} [R_{YX}(\tau) + R_{XY}(\tau)] \cos \omega\tau d\tau \end{aligned}$$

## 4.1.3 Cross-Spectral Density Functions



Real part and imaginary part

$$\begin{aligned} Q_{YX}(\omega) &= -\int_{-\infty}^{\infty} R_{XY}(\tau) \sin \omega \tau d\tau \\ &= \int_0^{\infty} [-R_{XY}(\tau) + R_{YX}(\tau)] \sin \omega \tau d\tau \end{aligned}$$

Amplitude Spectrum:

$$|S_{XY}(\omega)| = \sqrt{C_{XY}^2(\omega) + Q_{XY}^2(\omega)}$$

Phase Spectrum:

$$\varepsilon_{XY}(\omega) = \tan^{-1} \frac{Q_{XY}(\omega)}{C_{XY}(\omega)}$$

## 4.1.3 Cross-Spectral Density Functions



### 3. Properties of Cross-Spectral Density Function

- i)  $C_{XY}(\omega)$  and  $C_{YX}(\omega)$  are even functions
- ii)  $Q_{XY}(\omega)$  and  $Q_{YX}(\omega)$  are odd functions
- iii)  $S_{XY}(\omega) = S_{YX}(-\omega) = S_{YX}^*(\omega)$
- iv) If  $X(t)$  and  $Y(t)$  are jointly stationary random processes and uncorrelated,

$$S_{XY}(\omega) = S_{YX}(\omega) = 2\pi m_X m_Y \delta(\omega)$$

## 4.1.3 Cross-Spectral Density Functions



### 4. Coherency function

$$\gamma_{XY}(\omega) = \frac{|S_{XY}(\omega)|^2}{S_{XX}(\omega)S_{YY}(\omega)} = \frac{C_{XY}^2(\omega) + Q_{XY}^2(\omega)}{S_{XX}(\omega)S_{YY}(\omega)}$$

### Mutual correlation coefficients

$$\rho_{XY}[t_1, t_2] = \frac{C_{XY}(t_1, t_2)}{\sqrt{\text{Var}[X(t_1)]\text{Var}[Y(t_2)]}}$$

$$0 \leq \gamma_{XY}(\omega) \leq 1$$

$\gamma_{XY}(\omega) \equiv 1$  if and only if  $X(t)$  and  $Y(t)$  are exactly linearly related.

## 4.1.3 Cross-Spectral Density Functions



### *Example 1.*

Given: a cross-spectral density function

$$S_{XY}(\omega) = \begin{cases} a + i \frac{b\omega}{\omega_0} & -\omega_0 < \omega < \omega_0 \\ 0 & \text{otherwise} \end{cases}$$

where  $\omega_0 > 0$ ,  $a$  and  $b$  are real constants.

Obtain: the cross-correlation function.



# 4.1 Spectral Density Functions



## Content:

4.1.1 Autospectral Density Functions

4.1.2 Wiener-Khintchine Theorem

4.1.3 Crossspectral Density Functions

4.1.4 S.D.F. of Derived Random Processes

## 4.1.4 S.D.F. of Derived Random Processes



1. Autocorrelation function of derived random processes

$$R_{\dot{x}\dot{x}}(t_1, t_2) = \frac{\partial}{\partial t_2} R_{xx}(t_1, t_2) = -\frac{d}{d\tau} R_{xx}(\tau)$$

$$R_{\dot{x}\dot{x}}(\tau) = \frac{\partial^2}{\partial t_1 \partial t_2} R_{xx}(t_1, t_2) = -\frac{d^2}{d\tau^2} R_{xx}(\tau)$$

$$R_{\ddot{x}\ddot{x}}(\tau) = \frac{d^4}{d\tau^4} R_{xx}(\tau)$$

## 4.1.4 S.D.F. of Derived Random Processes



2. Spectral density function of derived random processes

$$\mathcal{F}\{x(t)\} = X(\omega)$$

$$\mathcal{F}\{\dot{x}(t)\} = \dot{X}(\omega) = i\omega X(\omega)$$

$$\mathcal{F}\{\ddot{x}(t)\} = \ddot{X}(\omega) = -\omega^2 X(\omega)$$

$$S_{\dot{x}\dot{x}}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{2T} |\dot{X}(\omega)|^2$$

$$= \omega^2 \lim_{T \rightarrow \infty} \frac{1}{2T} |X(\omega)|^2 = \omega^2 S_{xx}(\omega)$$

## 4.1.4 S.D.F. of Derived Random Processes



2. Spectral density function of derived random processes

$$\begin{aligned} S_{\ddot{x}\ddot{x}}(\omega) &= \lim_{T \rightarrow \infty} \frac{1}{2T} |\ddot{X}(\omega)|^2 \\ &= \omega^4 \lim_{T \rightarrow \infty} \frac{1}{2T} |X(\omega)|^2 = \omega^4 S_{xx}(\omega) \end{aligned}$$

# *Homework*

4.1

4.2

4.7