

Introduction to Higher-Order Spectra Analysis

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2.2.2005

Outline

- Motivation
- Moments and Cummulants (basic definitions, properties,..)
- Moments and Cummulants of Stationary Processes
- Cummulant Spectra (power spectrum,..)
- Linear Phase Shifts
- Regression Coefficient (linear system identification)
- Nonminimum Phase LTI Systems
- Coherency Function
- Nonlinear System Identification

Motivation

- Suppress Gaussian noise processes of unknown spectral characteristics in detection and parameter estimation problems
- Reconstruction of magnitude and phase information of signals and systems
- Detect and characterize nonlinearities in time series

Moments and Cummulants

$$\begin{aligned}
 Mom[x_1^{k_1}, x_2^{k_2}, \dots, x_n^{k_n}] &= E \{ x_1^{k_1} x_2^{k_2} \dots x_n^{k_n} \} = \\
 &= (-j)^r \frac{\partial^r \phi(\omega_1, \omega_2, \dots, \omega_n)}{\partial \omega_1^{k_1} \partial \omega_2^{k_2} \dots \partial \omega_n^{k_n}} \Big|_{\omega_1 = \omega_2 = \dots = \omega_n = 0}
 \end{aligned}$$

joint charact. Function (Fourier transform of the joint pdf!)

$$\begin{aligned}
 \phi(\omega_1, \dots, \omega_n) &= E \{ \exp(j(\omega_1 x_1 + \dots + \omega_n x_n)) \} = \\
 &= \int \dots \int \exp(j(\omega_1 x_1 + \dots + \omega_n x_n)) p(x_1, \dots, x_n) dx_1 \dots dx_n
 \end{aligned}$$

joint pdf

Moments and Cummulants

$$\begin{aligned} \text{Cum}[x_1^{k_1}, x_2^{k_2}, \dots, x_n^{k_n}] &= \\ &= (-j)^r \frac{\partial^r \psi(\omega_1, \omega_2, \dots, \omega_n)}{\partial \omega_1^{k_1} \partial \omega_2^{k_2} \dots \partial \omega_n^{k_n}} \Big|_{\omega_1 = \omega_2 = \dots = \omega_n = 0} \end{aligned}$$

Natural log. of the joint charact. Function

$$\psi(\omega_1, \dots, \omega_n) = \ln[\phi(\omega_1, \dots, \omega_n)]$$

Moments and Cummulants

$$m_1 = Mom[x] = E\{x\}$$

$$m_2 = Mom[x, x] = E\{x^2\}$$

.

.

$$c_1 = Cum[x] = -j \frac{d(\ln \phi(\omega))}{d\omega} \Big|_{\omega=0} = -j \frac{1}{\phi(\omega)} \frac{d\phi(\omega)}{d\omega} \Big|_{\omega=0} = m_1$$

$$c_2 = Cum[x, x] = (-j)^2 \frac{d^2 \psi(\omega)}{d\omega^2} \Big|_{\omega=0} =$$

$$= - \frac{d}{d\omega} \left(\frac{1}{\phi(\omega)} \frac{d\phi(\omega)}{d\omega} \right) \Big|_{\omega=0} = m_2 - m_1^2$$

Moments and Cummulants

Leonov-Shiryaev formula

$$\text{Cum} [x_1, \dots, x_n] = \sum (-1)^{p-1} (p-1)! \cdot E \left\{ \prod_{i \in s_1} x_i \right\} E \left\{ \prod_{i \in s_2} x_i \right\} \cdots E \left\{ \prod_{i \in s_p} x_i \right\}$$

$$(s_1, s_2, \dots, s_p), \quad p = 1, 2, \dots, n$$

Moments and Cummulants

$$\text{Cum}[x_1, x_2, x_3] = ?$$

$$p = 1 \quad s_1 = \{1, 2, 3\}$$

$$p = 2 \quad s_1 = \{1\}, s_2 = \{2, 3\}$$

$$s_1 = \{2\}, s_2 = \{1, 3\}$$

$$s_1 = \{3\}, s_2 = \{1, 2\}$$

$$p = 3 \quad s_1 = \{1\}, s_2 = \{2\}, s_3 = \{3\}$$

Moments and Cummulants

$$\begin{aligned} \text{Cum}[x_1, x_2, x_3] = & E\{x_1 x_2 x_3\} - E\{x_1\} E\{x_2 x_3\} - E\{x_2\} E\{x_1 x_3\} \\ & - E\{x_3\} E\{x_1 x_2\} + 2E\{x_1\} E\{x_2\} E\{x_3\} \end{aligned}$$

Calculation of Cum. of order r requires moments up to order r !

Properties of Moments and Cummulants (not complete!)

$$1.) \quad \text{Cum}[x_1, \dots, x_n] = 0, \quad \text{Mom}[x_1, \dots, x_n] \neq 0$$

$$\{x_1, \dots, x_\lambda\}, \{x_{\lambda+1}, \dots, x_n\} \rightarrow \psi(\omega_1, \dots, \omega_n) =$$

stat. independent

$$\underbrace{\psi_1(\omega_1, \dots, \omega_\lambda)} + \underbrace{\psi_2(\omega_{\lambda+1}, \dots, \omega_n)} = 0$$

$$\frac{\partial}{\partial \omega_1 \dots \partial \omega_n} = 0$$

Properties of Moments and Cummulants (not complete!)

2.)

$$\begin{aligned} \text{Cum}[x_1 + y_1, \dots, x_n + y_n] &= \\ &= \text{Cum}[x_1, \dots, x_n] + \text{Cum}[y_1, \dots, y_n] \\ \text{Mom}[x_1 + y_1, \dots, x_n + y_n] &\neq \\ &= \text{Mom}[x_1, \dots, x_n] + \text{Mom}[y_1, \dots, y_n] \end{aligned}$$

$$\{x_1, \dots, x_n\}, \{y_1, \dots, y_n\}$$

stat. independent

Proof is similar as in 1.)

Properties of Moments and Cummulants

(not complete!)

3.) $Cum[x_1, \dots, x_n] = 0, n > 2$ $\{x_1, \dots, x_n\}$ jointly Gaussian

$$\phi(\omega) = \exp\left(j\omega m_1 - \frac{1}{2}\omega^2 \sigma^2\right)$$

$$\psi(\omega) = \underbrace{j\omega m_1 - \frac{1}{2}\omega^2 \sigma^2}$$

$$\frac{d^n}{d\omega^n} = 0, n > 2$$

m_3, m_4, \dots

Intoduces no additional
Information !

Properties of Moments and Cummulants (not complete!)

Example: Noise supression

$$z_i = y_i + x_i, \quad i = 1, 2, 3$$

Gaussian noise
↓

$\{y_1, y_2, y_3\}$ joint prob. density function is not Gaussian !

$\{x_1, x_2, x_3\}$ joint prob. density function is Gaussian and Independent of $\{y_1, y_2, y_3\}$

with 2.) and 3.) \longrightarrow $Cum[z_1, z_2, z_3] = Cum[y_1, y_2, y_3]$

Moments and Cummulants of Stationary Processes

$$\{X(k)\}, k = 0, \pm 1, \pm 2, \dots$$



Real stationary random process

$$\underbrace{Mom[X(k), X(k + \tau_1), \dots, X(k + \tau_{n-1})]}_{m_n^x(\tau_1, \dots, \tau_{n-1})} = E\{X(k) X(k + \tau_1) \cdots X(k + \tau_{n-1})\}$$

$$m_n^x(\tau_1, \dots, \tau_{n-1})$$

$$c_n^x(\tau_1, \dots, \tau_{n-1}) = Cum[X(k), X(k + \tau_1), \dots, X(k + \tau_{n-1})]$$

Moments and Cummulants of Stationary Processes

1st-order cummulant

$$c_1^x = E \{ X(k) \} = m_1^x$$

mean value

2nd-order cummulant

$$c_2^x(\tau) = \underbrace{m_2^x(\tau)}_{\text{autocorrelation sequence}} - (m_1^x)^2$$

covariance sequence

$$c_2^x(0) = \gamma_2^x, m_1^x = 0 \quad \text{variance}$$

$$c_3^x(0,0) = \gamma_3^x, m_1^x = 0 \quad \text{skewness}$$

Cummulant Spectra

$$C_n^x(\omega_1, \dots, \omega_{n-1}) = \sum_{\tau_1=-\infty}^{\infty} \cdots \sum_{\tau_{n-1}=-\infty}^{\infty} c_n^x(\tau_1, \dots, \tau_{n-1}) \cdot \exp\{-j(\omega_1\tau_1 + \cdots + \omega_{n-1}\tau_{n-1})\}$$

complex periodic function

Power Spectrum: $n=2$

$$C_2^x(\omega) = \sum_{\tau} c_2^x(\tau) \exp(-j\omega\tau) \quad c_2^x(\tau) = \frac{1}{2\pi} \int_{-\pi}^{\pi} C_2^x(\omega) \exp(j\omega\tau) d\omega$$

if $\tau = 0 \rightarrow$

$$c_2^x(0) = \frac{1}{2\pi} \int_{-\pi}^{\pi} C_2^x(\omega) d\omega$$

power power spectrum

Linear Phase Shifts

$$Y(k) = X(k - D)$$

$$\begin{aligned} c_n^y(\tau_1, \dots, \tau_{n-1}) &= \text{Cum}[Y(k), Y(k + \tau_1), \dots, Y(k + \tau_{n-1})] \\ &= \text{Cum}[X(k - D), X(k - D + \tau_1), \dots, X(k - D + \tau_{n-1})] \\ &= c_n^x(\tau_1, \dots, \tau_{n-1}) \end{aligned}$$

Cummulant spectra suppress linear phase shifts!



Cross-cummulants

Linear Phase Shifts

$$\begin{aligned}C_{xyx\dots x}(\tau_1, \dots, \tau_{n-1}) &= \text{Cum}[X(k), Y(k + \tau_1), X(k + \tau_2), \dots, X(k + \tau_{n-1})] \\ &= \text{Cum}[X(k), X(k - D + \tau_1), X(k + \tau_2), \dots, X(k + \tau_{n-1})] \\ &= c_n^x(\tau_1 - D, \tau_2, \dots, \tau_{n-1})\end{aligned}$$



$$C_{xyx\dots x}(\omega_1, \dots, \omega_{n-1}) = C_n^x(\omega_1, \dots, \omega_{n-1}) \exp(-j\omega_1 D)$$

Cross-cummulant spectra preserve linear phase shifts!



e.g. for time delay estimation

Regression Coefficients

$$R(\omega) = \frac{C_{xy}(\omega)}{C_2^x(\omega)}$$

Example: linear system identification

$$Y(k) = \sum_i h(i) X(k-i), \quad E\{X(k)\} = 0$$

$$c_{xy}(\tau) = \text{Cum}[X(k), Y(k+\tau)] = \sum_i h(i) \text{Cum}[X(k), X(k+\tau-i)]$$

$$= \sum_i h(i) c_2^x(\tau-i) \longrightarrow \text{frequency-domain}$$

Regression Coefficients

$$C_{xy}(\omega) = \sum_{\tau} c_{xy}(\tau) \exp(-j\omega\tau) = \sum_{\tau} \sum_i h(i) \underbrace{c_2^x(\tau - i)}_{\text{variable change}} \exp(-j\omega\tau)$$

$$= \underbrace{\sum_v c_2^x(v) \exp(-j\omega v)}_{C_2^x(\omega)} \underbrace{\sum_i h(i) \exp(-j\omega i)}_{H(\omega)}$$



$$R(\omega) = H(\omega)$$

Nonminimum Phase LTI Systems

- Cumulant Spectra for $n > 2$ preserve phase information

$$Y(k) = \sum_i h(i) X(k-i), \quad E\{X(k)\} = 0$$

$$C_n^y(\omega_1, \dots, \omega_{n-1}) = H(\omega_1) \cdots H(\omega_{n-1}) H^*(\omega_1 + \cdots + \omega_{n-1}) \underbrace{C_n^x(\omega_1, \dots, \omega_{n-1})}_{\gamma_n^x \text{ if } \{X(k)\} \text{ is white}}$$

Example: Bispectrum $n=3$

$$C_3^y(\omega_1, \omega_2) = H(\omega_1) H(\omega_2) H^*(\omega_1 + \omega_2) \gamma_3^x$$

$$\Psi_3^y(\omega_1, \omega_2) = \phi_H(\omega_1) + \phi_H(\omega_2) - \phi_H(\omega_1 + \omega_2)$$

↑
phase of the LTI system

Nonminimum Phase LTI Systems

6th-order MA system

minimum phase system !

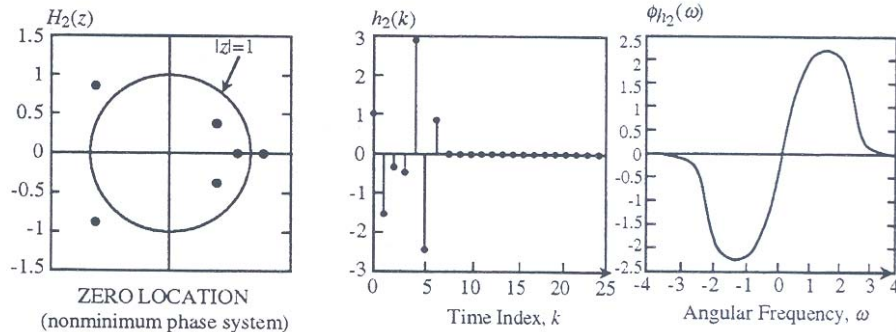
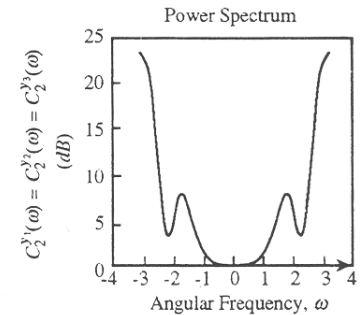
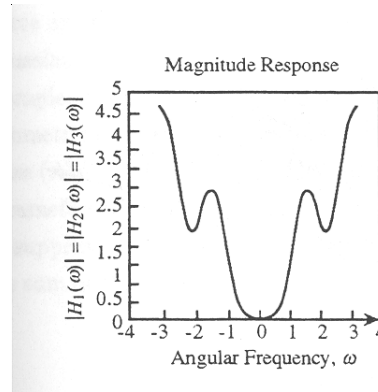
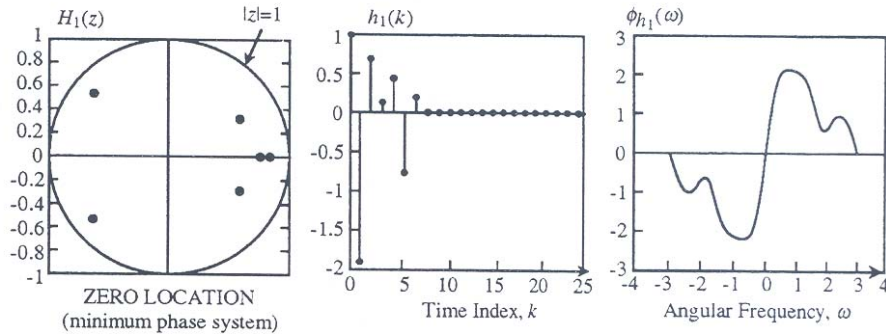
$$H_1(z) = \frac{(1 - 0.85z^{-1})(1 - 1.2z^{-1} + 0.45z^{-2})(1 - 0.869z^{-1})}{(1 + 1.1z^{-1} + 0.617z^{-2})}$$

$$H_2(z) = \frac{(1 - 0.85z^{-1})(1 - 1.2z^{-1} + 0.45z^{-2})(1 - 0.869z)}{(1 + 1.1z + 0.617z^2)}$$

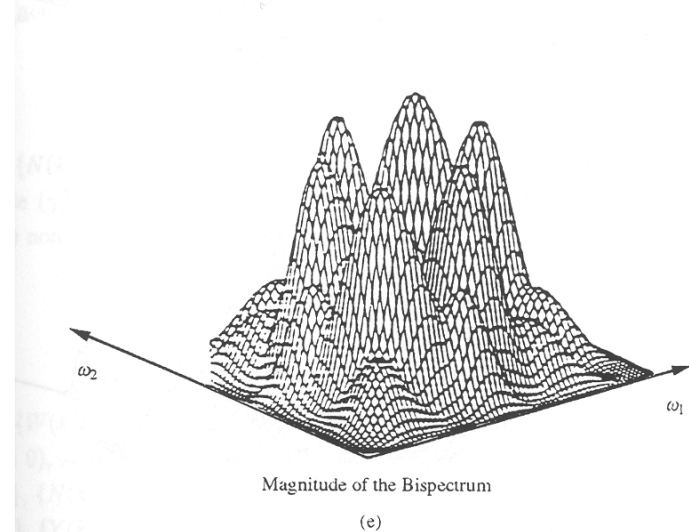
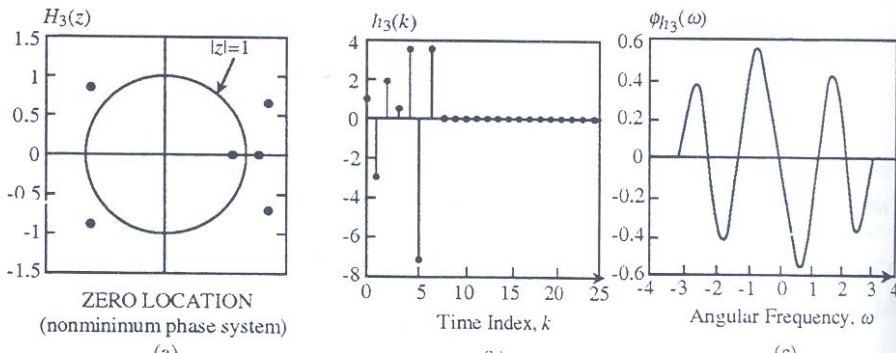
$$H_3(z) = \frac{(1 - 0.85z^{-1})(1 - 1.2z + 0.45z)(1 - 0.869z)}{(1 + 1.1z + 0.617z^2)}$$

same magnitude response !

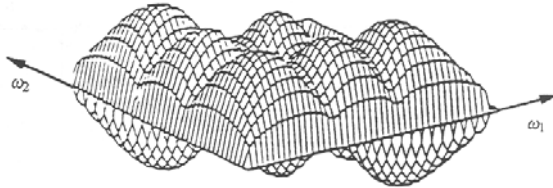
Nonminimum Phase LTI Systems



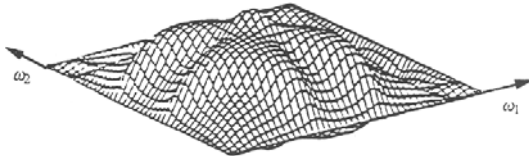
$$|C_3^{y_1,2,3}(\omega_1, \omega_2)| = |H(\omega_1)| |H(\omega_2)| |H(\omega_1 + \omega_2)|$$



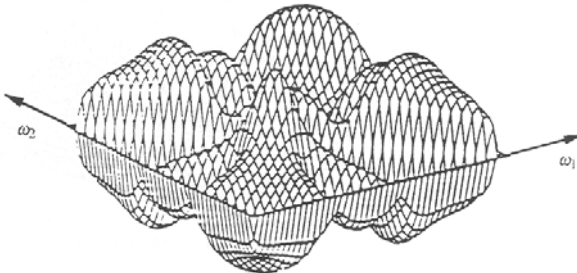
Nonminimum Phase LTI Systems

 $\Psi_3^{y_1}(\omega_1, \omega_2)$


$$\Psi_3^{y_1}(\omega_1, \omega_2) = \phi_{H_1}(\omega_1) + \phi_{H_1}(\omega_2) - \phi_{H_1}(\omega_1 + \omega_2)$$

 $\Psi_3^{y_2}(\omega_1, \omega_2)$


$$\Psi_3^{y_2}(\omega_1, \omega_2) = \phi_{H_2}(\omega_1) + \phi_{H_2}(\omega_2) - \phi_{H_2}(\omega_1 + \omega_2)$$

 $\Psi_3^{y_3}(\omega_1, \omega_2)$


$$\Psi_3^{y_3}(\omega_1, \omega_2) = \phi_{H_3}(\omega_1) + \phi_{H_3}(\omega_2) - \phi_{H_3}(\omega_1 + \omega_2)$$

(f)

Coherency Function

$$P_n^y(\omega_1, \dots, \omega_{n-1}) = \frac{C_n^y(\omega_1, \dots, \omega_{n-1})}{\left[C_2^y(\omega_1) \cdots C_2^y(\omega_{n-1}) C_2^y(\omega_1 + \dots + \omega_{n-1}) \right]^{\frac{1}{2}}}$$

Example: distinguish between linear non-Gaussian
and nonlinear processes

$$C_n^y(\omega_1, \dots, \omega_{n-1}) = H(\omega_1) \cdots H(\omega_{n-1}) H^*(\omega_1 + \dots + \omega_{n-1}) \underbrace{C_n^x(\omega_1, \dots, \omega_{n-1})}_{\gamma_n^x \text{ if } \{X(k)\} \text{ is white}}$$

$$C_2^y(\omega) = \gamma_2^x |H(\omega)|^2$$

coherency index

$$\left| P_n^y(\omega_1, \dots, \omega_n) \right| = \frac{\gamma_n^x}{(\gamma_2^x)^{\frac{n}{2}}}$$

Coherency Function

$$|P_n^y(\omega_1, \dots, \omega_n)| = \begin{cases} 0, & \text{if } Y(k) \text{ is Gaussian for } n > 2 \\ \frac{\gamma_n^x}{(\gamma_2^x)^{\frac{n}{2}}}, & \text{if } Y(k) \text{ is linear non-Gaussian} \\ f(\omega_1, \dots, \omega_n), & \text{if } Y(k) \text{ is nonlinear} \end{cases}$$

Nonlinear System Identification

2nd-order homogeneous Volterra system

$$Y(k) = \sum_{i_1} \sum_{i_2} h_2(i_1, i_2) X(k - i_1) X(k - i_2)$$

Gaussian



$$\begin{aligned} c_{yxx}(\tau_1, \tau_2) &= \text{Cum}[Y(k), X(k + \tau_1), X(k + \tau_2)] = \\ &= \sum_{i_1} \sum_{i_2} h_2(i_1, i_2) \underbrace{\text{Cum}[X(k - i_1) X(k - i_2), X(k + \tau_1), X(k + \tau_2)]}_{\text{Leonov-Shiryaev}} \end{aligned}$$

Wiener



Leonov-Shiryaev

$$\begin{aligned} &= \sum_{i_1} \sum_{i_2} h_2(i_1, i_2) \left[E\{X(k - i_1) X(k - i_2) X(k + \tau_1) X(k + \tau_2)\} - \right. \\ &\quad \left. E\{X(k - i_1) X(k - i_2)\} E\{X(k + \tau_1) X(k + \tau_2)\} \right] \end{aligned}$$

Nonlinear System Identification

2nd-order homogeneous Volterra system

$$= \sum_{i_1} \sum_{i_2} h_2(i_1, i_2) \left[c_2^x(i_1 + \tau_1) c_2^x(i_2 + \tau_2) + c_2^x(i_1 + \tau_2) c_2^x(i_2 + \tau_1) \right]$$

$$C_{yxx}(\omega_1, \omega_2) = \sum_{\tau_1} \sum_{\tau_2} c_{yxx}(\tau_1, \tau_2) \exp(-j(\omega_1 \tau_1 + \omega_2 \tau_2))$$

$$C_{yxx}(\omega_1, \omega_2) = 2C_2^x(\omega_1) C_2^x(\omega_2) H_2(-\omega_1, -\omega_2)$$

Nonlinear System Identification

2nd-order Volterra system

$$Y(k) = \underbrace{\sum_i h_1(i) X(k-i)}_{Y_1(k)} + \underbrace{\sum_{i_1} \sum_{i_2} h_2(i_1, i_2) X(k-i_1) X(k-i_2)}_{Y_2(k)}$$

1.)

$$\begin{aligned} c_{yx}(\tau) &= \text{Cum}[Y(k), X(k+\tau)] = \text{Cum}[Y_1(k) + Y_2(k), X(k+\tau)] = \\ &= \sum_i h_1(i) \text{Cum}[X(k-i), X(k+\tau)] \\ &+ \sum_{i_1} \sum_{i_2} h_2(i_1, i_2) \text{Cum}[X(k-i_1) X(k-i_2), X(k+\tau)] \end{aligned}$$

Wiener

Nonlinear System Identification

2nd-order Volterra system

$$c_{yx}(\tau) = \sum_i h_1(i) c_2^x(i + \tau) \xrightarrow{FT} C_{yx}(\omega) = C_2^x(\omega) H_1(-\omega)$$

2.)

$$c_{yxx}(\tau_1, \tau_2) = \text{Cum}[Y(k), X(k + \tau_1), X(k + \tau_2)] \longrightarrow$$

$$\sum_{i_1} \sum_{i_2} h_2(i_1, i_2) [c_2^x(i_1 + \tau_1) c_2^x(i_2 + \tau_2) + c_2^x(i_1 + \tau_2) c_2^x(i_2 + \tau_1)]$$

$$C_{yxx}(\omega_1, \omega_2) = 2C_2^x(\omega_1) C_2^x(\omega_2) H_2(-\omega_1, -\omega_2)$$