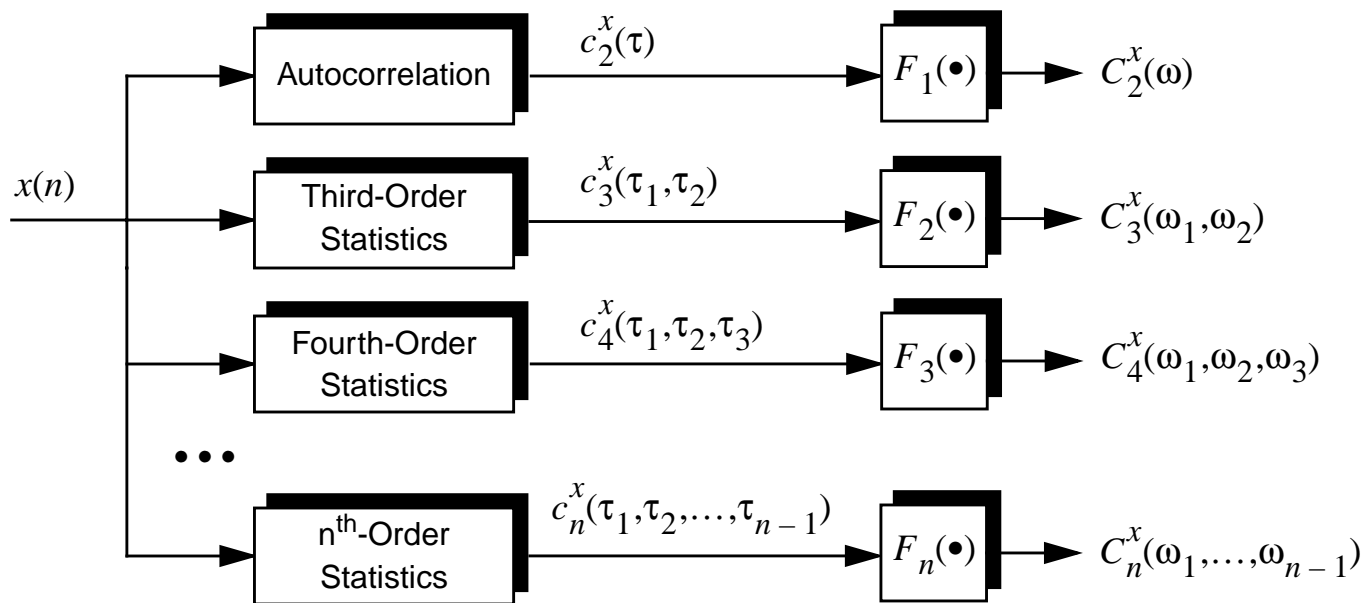


## Higher-Order Spectra

- ❑ Fact: the information contained in the power spectrum of a signal is essentially that which is contained in the autocorrelation sequence
- ❑ This is sufficient for a signal with Gaussian statistics (why?), but...
- ❑ Premise: there is useful information in higher-order moments of the signal (typically third and fourth-order moments or cumulants)

### Historical Perspective:

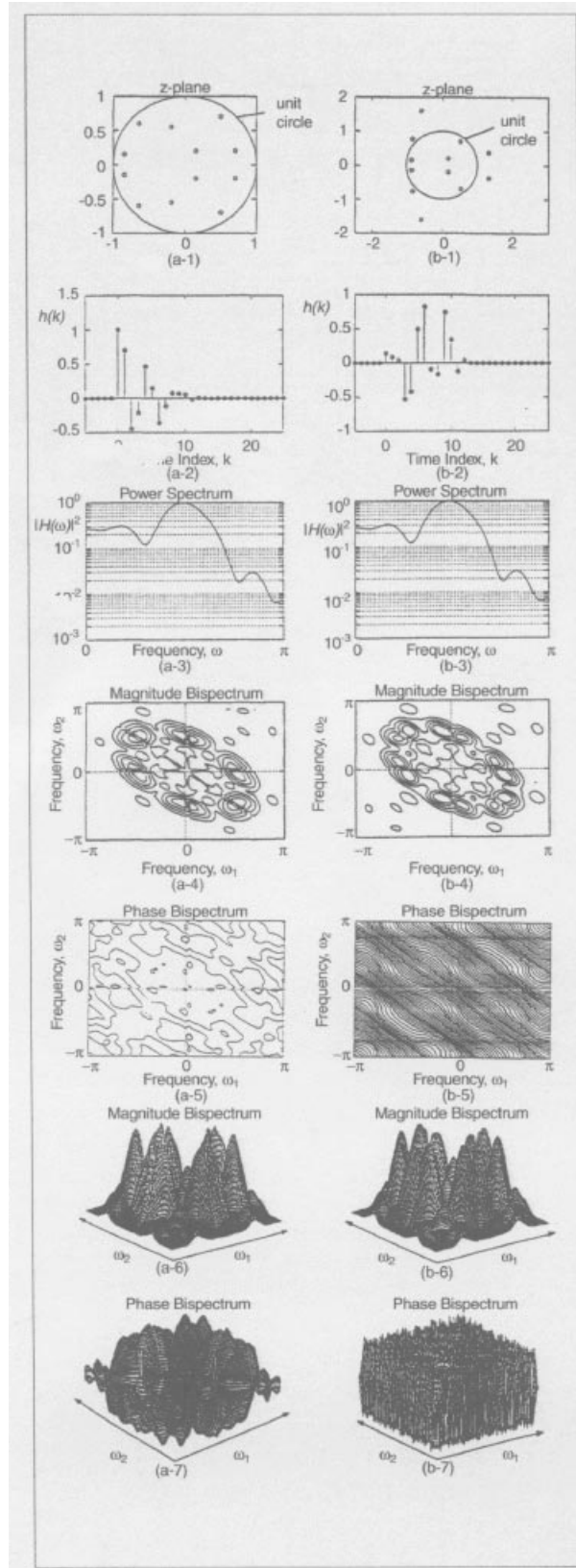
- ❑ The power spectrum is a member of the class of higher order spectra, and can be referred to as a second-order spectra:



- ❑ Such transformations can suppress additive colored Gaussian noise of unknown power spectra
- ❑ Identify and/or reconstruct non-minimum phase signals
- ❑ Characterize deviations from Gaussian statistics
- ❑ Identify nonlinear properties in signals



# An Example: Non-Minimum Phase Systems



## Mathematical Definitions

The  $n^{\text{th}}$ -order moment is defined as:

$$m_n^x(\tau_1, \tau_2, \dots, \tau_{n-1}) \equiv E[x(n)x(n + \tau_1)x(n + \tau_2) \dots x(n + \tau_{n-1})]$$

The  $n^{\text{th}}$ -order cumulant is defined as:

$$c_n^x(\tau_1, \tau_2, \dots, \tau_{n-1}) \equiv m_n^x(\tau_1, \tau_2, \dots, \tau_{n-1}) - m_n^G(\tau_1, \tau_2, \dots, \tau_{n-1})$$

where  $m_n^G(\tau_1, \tau_2, \dots, \tau_{n-1})$  is the  $n^{\text{th}}$ -order moment of an equivalent Gaussian signal with the same mean and autocorrelation as  $x(n)$ .

We can expand these functions as follows:

First-order moment:

$$c_1^x = m_1^x = E[x(n)] \quad (\text{mean value})$$

Second-order moment:

$$\begin{aligned} c_2^x(\tau_1) &= m_2^x(\tau_1) - (m_1^x)^2 \\ &= m_2^x(-\tau_1) - (m_1^x)^2 \quad (\text{covariance}) \\ &= c_2^x(\tau_1) \end{aligned}$$

Third-order moment:

$$c_3^x(\tau_1, \tau_2) = m_3^x(\tau_1, \tau_2) - m_1^x[m_2^x(\tau_1) + m_2^x(\tau_2) + m_2^x(\tau_1 - \tau_2)] + 2(m_1^x)^2$$

If the mean of the signal is zero, then the second and third-order cumulants are equivalent to the moments, and the fourth-order cumulant is given by:

$$\begin{aligned} c_4^x(\tau_1, \tau_2, \tau_3) &= m_4^x(\tau_1, \tau_2, \tau_3) - m_2^x(\tau_1)m_2^x(\tau_3 - \tau_2) - \\ &\quad m_2^x(\tau_2)m_2^x(\tau_3 - \tau_1) - m_3^x(\tau_3)m_2^x(\tau_2 - \tau_1) \end{aligned}$$

## Mathematical Properties of Deterministic Signals

- Spectrum

Energy spectrum:  $M_2^x(\omega) = X(\omega)X^*(\omega)$

Bispectrum:  $M_3^x(\omega_1, \omega_2) = X(\omega_1)X^*(\omega_2)X^*(\omega_1 + \omega_2)$

Trispectrum:  $M_4^x(\omega_1, \omega_2, \omega_3) = X(\omega_1)X(\omega_2)X(\omega_3)X^*(\omega_1 + \omega_2 + \omega_3)$

- Quality Measures (zero mean)

Variance:  $\gamma_2^x = E[x^2(n)] = c_2^x(0)$

Skewness:  $\gamma_3^x = E[x^3(n)] = c_3^x(0, 0)$

Kurtosis:  $\gamma_4^x = E[x^4(n)] - 3[\gamma_2^x]^2 = c_4^x(0, 0, 0)$

Normalized Kurtosis:  $\gamma_4^x / [\gamma_2^x]^2$

- Coherency

Biocoherency:  $P_3^x(\omega_1, \omega_2) = \frac{C_3^x(\omega_1, \omega_2)}{\sqrt{C_2^x(\omega_1)C_2^x(\omega_2)C_2^x(\omega_1 + \omega_2)}}$

Tricoherency:  $P_4^x(\omega_1, \omega_2, \omega_3) = \frac{C_3^x(\omega_1, \omega_2)}{\sqrt{C_2^x(\omega_1)C_2^x(\omega_2)C_3^x(\omega_3)C_2^x(\omega_1 + \omega_2 + \omega_3)}}$

The latter measures are useful in discriminating linear processes from nonlinear ones. A signal is said to be a linear non-Gaussian process of order  $n$  if the magnitude of the  $n^{\text{th}}$ -order coherency function is constant for all frequencies.