

Q1. Multi-input, multi-output systems and coherence

Consider a linear system L with m inputs and n outputs. It can be characterized by an array of impulse responses, $L_{mn}(\tau)$, which specify the response of the n th output to an impulse on the m th input, or,

equivalently, an array of transfer functions $\hat{L}_{mn}(\omega) = \int_0^{\infty} e^{-i\omega t} L_{mn}(t) dt$ that specify the sinusoidal component at ω on the n th output produced by a unit sinusoid on the m th input. We can also denote the array of transfer functions by $\hat{L}(\omega)$.

- A. Given two such systems in series, say A with m inputs and n outputs and transfer functions $\hat{A}_{nm}(\omega)$, and B , which takes these n outputs as its input and produces p outputs, with transfer functions $\hat{B}_{pn}(\omega)$, what are the transfer functions $\tilde{L}_{pm}(\omega)$ of the composite system consisting of A followed by B ?
- B. For systems with the same number of inputs and outputs (i.e., $m = n = p$ above), does the order of composition matter?
- C. With A as above (m inputs, n outputs, transfer functions $\hat{A}_{nm}(\omega)$): How does the cross-spectral matrix of the output $P_{Y_j, Y_k}(\omega)$ relate to the cross-spectral matrix of the input, $P_{X_j, X_k}(\omega)$? What if the inputs consist of independent Gaussian noises with unit spectral density?
- D. Consider two m -input, m -output systems, A and B . For what conditions on A are the cross-spectral matrices of B , and of L , consisting of A followed by B , identical?

Q2. Hermite polynomials and generating functions

Hermite polynomials – orthogonal polynomials with respect to a Gaussian -- play a major role in extending input-output analysis to nonlinear systems. This is because of both the Central Limit Theorem and Price's Theorem (see Question 3). Question 4 illustrates this extension, and can be done without first doing Q2 and Q3.

First, we establish the orthogonality of Hermite polynomials and then prove Price's Theorem using generating functions. If you haven't seen generating functions, they are a good thing to have in your toolkit.

Demonstrating orthogonality of the Hermite polynomials are the "warm-up exercise."

In our standardization, the m th Hermite polynomial $h_m(x)$ is defined as the coefficient of t^m in $\exp(xt - \frac{t^2}{2})$,

specifically, $\sum_{m=0}^{\infty} h_m(x) \frac{t^m}{m!} = \exp(xt - \frac{t^2}{2})$. In this standardization, $h_m(x)$ has a leading coefficient 1.

Show that the Hermite polynomials are orthogonal with respect to a unit Gaussian, namely, that

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} h_m(x)h_n(x)e^{-x^2/2} dx = \begin{cases} 0, & m \neq n \\ m!, & m = n \end{cases}.$$

A. With $I_{m,n} = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} h_m(x)h_n(x)e^{-x^2/2} dx$, express $I(s,t) = \sum_{m,n=0}^{\infty} I_{m,n} \frac{s^m}{m!} \frac{t^n}{n!}$ as an integral of an exponential using the generating-function definition of the Hermites.

B. Integrate $I(s,t)$ (complete the square in the exponent and use $\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-x^2/2} dx = 1$).

C. Equate the expressions in A and B for $I(s,t)$ term-by-term to determine $I_{m,n}$.

Q3. Price's Theorem

Price's Theorem states that if two variables are drawn from a correlated Gaussians (say, x and y , each with zero mean and unit variance, and correlation $\rho = \langle xy \rangle$), then, for any Hermite polynomials h_m and h_n (defined below), $\langle h_m(x)h_n(y) \rangle = 0$ if $m \neq n$, and $\langle h_m(x)h_m(y) \rangle = m! \rho^m$. This is crucial to extending the cross-correlation approach to nonlinear systems.

First, we set up correlated unit-variance, zero-mean Gaussian variables. Let u and v be uncorrelated unit-mean Gaussian variables, and $x = u \cos \theta + v \sin \theta$, $y = u \sin \theta + v \cos \theta$ (note, not a rotation).

A. Determine $\langle x^2 \rangle$, $\langle y^2 \rangle$, and $\rho = \langle xy \rangle$

B. We want to calculate $J_{m,n} = \langle h_m(x)h_n(y) \rangle$. Rather than integrate over a pair of correlated Gaussians in x and y , we use the underlying uncorrelated Gaussians in u and v . So

$$J_{m,n} = \langle h_m(x)h_n(y) \rangle = \left(\frac{1}{\sqrt{2\pi}} \right)^2 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h_m(u \cos \theta + v \sin \theta) h_n(u \sin \theta + v \cos \theta) e^{-u^2/2} e^{-v^2/2} dudv.$$

Write the generating function $J(s,t) = \sum_{m,n=0}^{\infty} J_{m,n} \frac{s^m}{m!} \frac{t^n}{n!}$ using the generating function for the Hermites, integrate, and equate term-by-term to demonstrate the claim of Price's Theorem.

Q4. Application to input-output analysis of nonlinear systems.

Consider a composite system consisting of a linear filter followed by a static nonlinear system. Specifically, the linear system L has an impulse-response $L(\tau)$, that produces a response $q(t)$ to an input $s(t)$, and is followed by a static nonlinear system N whose response to q is given by a nonlinear function $r = N(q)$. That is, the response of N at any given time depends only on its input at that time, and not on previous values of the input.

We analyze the response of this composite system when its input is a Gaussian noise of unit variance.

A. Let V be the variance of $q(t)$ when the input $s(t)$ is a unit-variance Gaussian. Provided that

$$\int_{-\infty}^{\infty} (N(q))^2 e^{-q^2/2V} dq \text{ is finite, } N(q) \text{ can be expanded in terms of Hermite polynomials as}$$

$N(q) = \sum_{k=0}^{\infty} \alpha_k h_k \left(\frac{q}{\sqrt{V}} \right)$. Using the orthogonality of the Hermite polynomials (Q2), namely, that

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} h_m(x) h_n(x) e^{-x^2/2} dx = \begin{cases} 0, & m \neq n \\ m!, & m = n \end{cases}, \text{ determine } \alpha_k. \text{ Hint: Consider } N(q) \text{ as a vector in a Hilbert}$$

space with inner product $(f, g)_H = \frac{1}{\sqrt{2\pi V}} \int_{-\infty}^{\infty} f(q) \overline{g(q)} e^{-q^2/2V} dq$. Then think of projecting it onto the

one-dimensional subspace spanned by $h_m \left(\frac{q}{\sqrt{V}} \right)$, using that inner product. In other words, the α_k are

the coordinates of $N(q)$ in the basis set consisting of the $h_m \left(\frac{q}{\sqrt{V}} \right)$.

- B. Using Price's theorem (Q3), determine the cross-correlation between a Hermite polynomial function of the stimulus and system's response, i.e., $Z_n(\tau) = \langle r(t) h_n(s(t-\tau)) \rangle$.