Lecture 16

Last time: Wide band input

Notice that

$$R_{xy}(\tau) = S_0 w(\tau)$$
$$= \overline{x(t)y(t+\tau)}$$

so if t is the current time in a <u>real time</u> situation, we cannot compute $R_{xy}(\tau)$ for $\tau > 0$ which is necessary since $w_x(\tau)$ is nonzero only for $\tau > 0$.

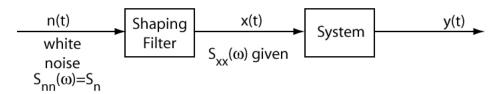
But we showed earlier that

$$R_{xy}(\tau) = R_{yx}(\tau)$$
$$= \overline{y(t)x(t-\tau)}$$

This can be computed in real time by averaging the product of the current output, y(t), and the delayed input, $x(t-\tau)$, for $\tau > 0$.

The Shaping Filter

A shaping filter is a filter that produces an output process with a white noise input, which has a specified spectrum. If a given system is driven by a colored noise, we can precede it with the appropriate shaping filter and always treat the augmented system as driven by white noise.



We usually work with <u>rational spectra</u> – so consider $S_{xx}(\omega)$ to be a rational function. That is, it is a ratio of polynomials in ω .

Since $S_{xx}(\omega)$ is always an <u>even function</u>, it is a ratio of polynomials having only even powers of ω .

$$S_{xx}(\omega) = \frac{a_0 + a_1 \omega^2 + a_2 \omega^4 + \dots}{b_0 + b_1 \omega^2 + b_2 \omega^4 + \dots}$$

We wish to characterize the filter in terms of the Laplace variable s. The relation between s and ω is $s = j\omega$.

Rewrite S_{xx} in terms of s. Since only even powers of ω appear there will be no imaginary terms.

$$\omega^2 = -s^2$$

$$\omega^4 = +s^4$$

$$\omega^{6} = -s^{6}$$

etc.

$$S_{xx}(s) = \frac{a_0 - a_1 s^2 + a_2 s^4 - a_3 s^6 \dots}{b_0 - b_1 s^2 + b_2 s^4 - b^3 s^6 \dots}$$

Factor the numerator and denominator in terms of s^2 .

$$S_{xx}(s) = c \frac{(s^2 - c_1)(s^2 - c_2)...}{(s^2 - d_1)(s^2 - d_1)...}$$

The poles and zeroes are located at:

Zeroes at $s = \pm \sqrt{c_i}$, where i = 1, 2, ...

Poles at $s = \pm \sqrt{d_i}$, where i = 1, 2, ...

If c_i is real and positive, then $\sqrt{c_i}$ is real

If c_i is real and negative, then $\sqrt{c_i}$ is purely imaginary

This kind of root always appears with even multiplicity.

$$(s^2-c_i) \rightarrow (-\omega^2-c_i)$$

Notice that in terms of ω , this must be true to avoid a change of sign of $S_{xx}(\omega)$.

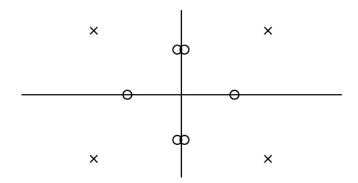
For small ω , $(-\omega^2 - c_i)$ is positive.

For large ω , $(-\omega^2 - c_i)$ is negative.

If c_i is complex, then c_i^* is another root.

The poles or zeroes are at:

$$s = \begin{cases} +\sqrt{c_i} = +a + jb \\ -\sqrt{c_i} = -a - jb \\ +\sqrt{c_i^*} = +a - jb \\ -\sqrt{c_i^*} = -a + jb \end{cases}$$



Collect the factors which define poles or zeroes in the left-half plane together with the <u>square root of any factorable constant</u> (\sqrt{c}). Call the product $S_{xx}(s)_L$ - the left hand factor of the spectrum.

Collect the remaining factors, which define poles or zeroes in the right-half plane, together with the square root of the factorable constant. Call the product $S_{xx}(s)_R$ - the right hand factor of the spectrum.

Then:

$$S_{xx}(s) = S_{xx}(s)_L S_{xx}(s)_R$$

From the symmetry of the pole-zero pattern, we can see that $S_{xx}(s)_R = S_{xx}(-s)_L$

Call the shaping filter transfer function F(s). The input-output relation for power spectral density functions in terms of s is $S_{xx}(s) = F(s)F(-s)S_{yy}(s)$

But with $S_{xx}(s)$ factored and n(t) white,

$$S_{xx}(s)_L S_{xx}(s)_R = F(s)F(-s)S_n$$

$$S_{xx}(s)_L S_{xx}(-s)_L = F(s)F(-s)S_n$$

One solution for F(s) which produces the desired output spectrum is

$$F(s) = \frac{1}{\sqrt{S_n}} S_{xx}(s)_L$$

We can choose S_n to be 1 if we wish.

By construction, this filter is <u>stable</u> and <u>minimum phase</u>.

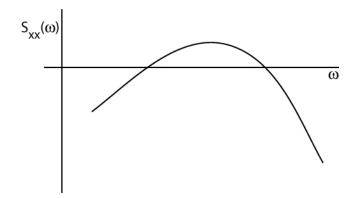
If $S_{nn}(\omega)$ is not broadband relative to the system,

16.322 Stochastic Estimation and Control, Fall 2004 Prof. Vander Velde

$$F(s) = \frac{S_{xx}(s)_L}{S_{nn}(s)_L}$$

Example

Suppose a disturbance process has a spectrum with the following shape.



Suppose this function is well approximated by

$$S_{xx}(\omega) = \frac{10\omega^2}{\omega^4 + 5\omega^2 + 4}$$

Then:

$$S_{xx}(s) = \frac{-10s^2}{s^4 - 5s^2 + 4}$$

$$= \frac{\sqrt{10s(-\sqrt{10s})}}{(s^2 - 1)(s^2 - 4)}$$

$$= \left[\underbrace{\frac{\sqrt{10s}}{(s+1)(s+2)}} \right] \underbrace{\left[\frac{\sqrt{10s}}{(s-1)(s-2)} \right]}_{S_{xx}(s)_{t}}$$

With <u>unit intensity noise</u> driving the shaping filter, the filter transfer function is

$$F(s) = \frac{\sqrt{10}s}{(s+1)(s+2)} = \frac{\sqrt{10}s}{s^2 + 3s + 2}$$

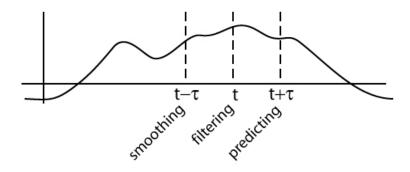
If you are going to work in terms of a state space model, you would now have to define a state variable realization of this transfer, and augment the state space model of the original system with it.

General Introduction to Filtering and Control System Design

Suppose you have input data up to time t, and at that time you want to produce a smoothed estimate of the signal (which is a real time operation). This called <u>filtering</u>.

Suppose you want an estimate of a signal at time $t + \tau$ in the future. This is called <u>predicting</u>.

Suppose you want an estimate of a signal at time $t - \tau$ in the past, using data available up to time t. This is called <u>smoothing</u>.



Ground rules

We shall treat in a fairly general manner the problem of optimum design of systems under the following conditions:

- 1. The system is assumed to be linear and physically realizable.
- 2. The inputs to the system (e.g., signals, noise, disturbances), are members of stationary random processes.
- 3. The desired operation of the system is a linear time-invariant operation on the signal.
- 4. The optimum system is defined to be that which yields the minimum steady state mean squared error between actual output and derived output.

With stationary inputs, the input correlation functions are invariant with time; these are the invariants which are the basis for prediction.

Thus we shall be finding the realizable linear system which operating on the signal and noise yields an output which best approximates in the mean squared sense the output of an ideal linear operator, <u>not necessarily realizable</u>, operating on the signal only. We shall find optimum systems in two senses:

- 1. Configuration fixed find optimum parameter values
- 2. Configuration semi-free find best linear realizable compensator
 - a. Configuration free find best linear realizable system