

Start class #9 in 1999 10/18/99

Auto-Spectral Analysis

- stationary
- random (from a normally distributed process)
- ergodic
- zero mean

Define Auto-covariance:  $R_x(t, \tau) = E[x(t) \cdot x(t+\tau)]$

stationarity:  $= E[x(t_1) \cdot x(t_2 + \tau)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x(t) \cdot x(t+\tau) p(x(t), x(t+\tau)) dx(t) dx(t+\tau)$

ergodicity:  $= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x(t) x(t+\tau) dt$

$= R_x(\tau)$  in practice we just have this, i.e., call it  $\hat{R}_x(\tau)$

Properties

(1)  $R_x(\tau) = R_x(-\tau)$  even fctn.

(2)  $|R_x(\tau=0)| \geq |R_x(\tau)|$   
true variance      lagged auto-covariance

(3) as the record length  $T \rightarrow \infty$  the estimate of the auto-covariance

$\hat{R}_x(\tau) \equiv \frac{1}{T} \int_0^T x(t) x(t+\tau) dt$   $t+\tau < T$

becomes asymptotically unbiased (Proof follows)

Proof of (3): asymptotically unbiased estimate

$$E[\hat{R}_T(\tau)] = E\left[\frac{1}{T} \int_0^T x(t)x(t+\tau) dt\right] \quad \begin{matrix} 2009 \\ t+\tau < T \end{matrix}$$

$$= \frac{1}{T} \int_0^T E[x(t)x(t+\tau)] dt$$

$$= \frac{1}{T} \int_0^{T-|\tau|} E[x(t)x(t+\tau)] dt \quad \left. \begin{matrix} \text{finite record length} \\ \text{as } x(t)=0 \text{ for } t < 0 \end{matrix} \right\} \begin{matrix} t > T \\ t < 0 \end{matrix}$$

$$= \frac{1}{T} \int_0^{T-|\tau|} \frac{E[\cdot]}{T} R_x(t) dt = \frac{1}{T} \int_0^T E[\cdot] dt - \frac{1}{T} \int_0^{|\tau|} E[\cdot] dt$$

$$= R_x(\tau) - \frac{|\tau|}{T} \frac{1}{|\tau|} \int_0^{|\tau|} E[\cdot] dt$$

$$= R_x(\tau) (1 - \tau/T)$$

→ Biased estimate !

$$\rightarrow = R_x(\tau) - \frac{|\tau|}{T} R_x(\tau)$$

Definition: Auto-Spectrum  $S_x(f)$

$$S_x(f) = \int_{-\infty}^{\infty} R_x(\tau) e^{-j2\pi f\tau} d\tau = \mathcal{F}(R_x(\tau))$$

$$= 2 \int_0^{\infty} R_x(\tau) e^{-j2\pi f\tau} d\tau$$

because  $R_x(\tau) = R_x(-\tau)$ ,  $R_x(\tau)$  is even fctn.  
and it also follows that

$$S_x(-f) = S_x^*(f) = S_x(f)$$

Hence  $S_x(f)$  is real !

Hence

$$S_{x^*}(f) = 2 \int_0^{\infty} R_{x^*}(\tau) \cos(2\pi f\tau) d\tau$$

and

~~$R_{x^*}(\tau)$~~   $R_{x^*}(\tau) = 2 \int_0^{\infty} S_x(f) \cos(2\pi f\tau) df$

are Fourier Transform pairs if  $\int_{-\infty}^{\infty} |R(\tau)| d\tau < \infty$

which in practice is always true because of finite record lengths.

for  $\tau = 0$

$$R_x(\tau=0) = 2 \int_0^{\infty} S_x(f) df$$

But it is also

(from its definition)

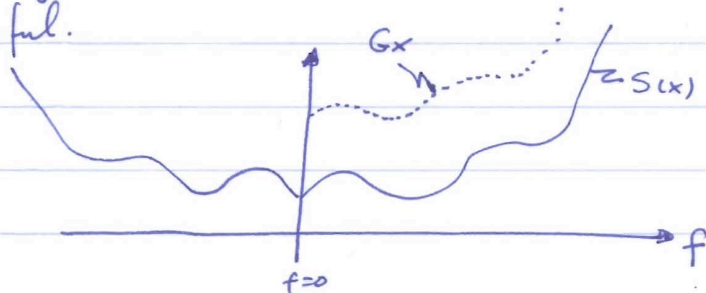
$$R_x(\tau=0) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x^2(t) dt = \int_{-\infty}^{\infty} S_x(f) df$$

$S_x(f)$

PARTITION

OF VARIANCE  
IN FREQUENCY  
SPACE

Hence we can interpret  $S_x$  as an energy (or variance) density function in the sense that the area under the curve  $S_x(f)$  represents an energy or a variance. It is this physical (as opposed to the "messy" statistical) interpretation that makes the spectrum so useful.



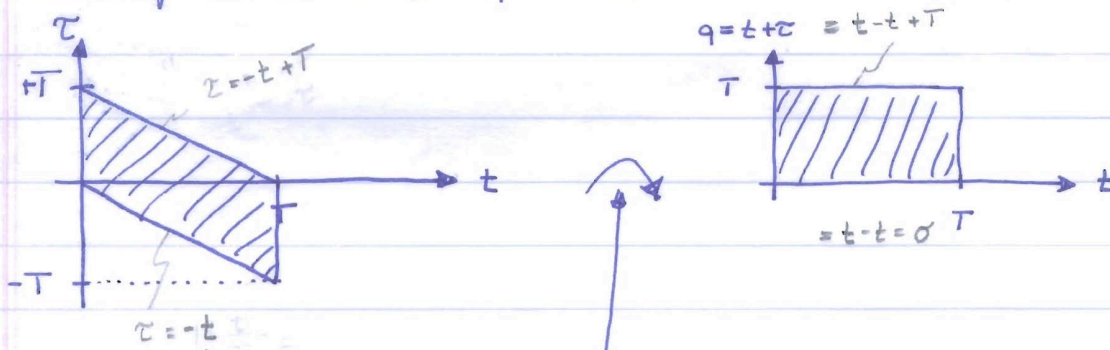
~~$G_x(f) = 2$~~   
 $G_x(f) = \begin{cases} 2S_x & f \geq 0 \\ 0 & f < 0 \end{cases}$

Spectral density <sup>from</sup> of a finite record :  $T S_x(f)$

An estimate of that is

$$\begin{aligned}
 T \hat{S}_x(f) &= \int_{-T}^T \hat{T}R_x(\tau) e^{-j2\pi f\tau} d\tau \quad (5.64) \\
 &= \int_{-T}^T \left[ \frac{1}{T} \int_0^T x(t) x(t+\tau) dt \right] e^{-j2\pi f\tau} d\tau \\
 &= \frac{1}{T} \int_0^T x(t) \int_{-T}^T x(t+\tau) e^{-j2\pi f\tau} d\tau dt
 \end{aligned}$$

Now remember that  $\hat{T}R_x = 0$  for  $|\tau| > T$ , hence the actual integration in  $(\tau, t)$  space is



Some "trick" use in proving the convolution theorem on p. 16 1/2/21

change of variables

$$q = t + \tau$$

$$q = t + \tau \text{ or } \tau = q - t$$

$$dq = d\tau$$

$$\begin{aligned}
 &e^{-j2\pi f\tau} \\
 \text{so } &e^{-j2\pi f(q-t)} \\
 &= e^{-j2\pi fq} e^{+j2\pi ft}
 \end{aligned}$$

$$T \hat{S}_x(f) = \frac{1}{T} \int_0^T x(t) e^{+j2\pi ft} \int_0^q x(q) e^{-j2\pi fq} dq dt$$

$$\underline{T \hat{S}_x(f)} = \frac{1}{T} \left| \int_0^T x(t) e^{-j2\pi ft} dt \right|^2$$

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$$\int_{-T}^T x(f) = \frac{1}{T} X(f) \cdot X^*(f)$$

where

$$X(f) = \int_0^T x(t) e^{-j2\pi f t} dt$$

and

$$X^*(f) = \int_0^T x(t) e^{+j2\pi f t} dt$$

is the complex conjugate of  $X(f)$

↑ end class #9

## SUMMARY

1. (a) stationary but not necessarily ergodic process  $\{x_2(t)\}$   $n=1,2,\dots$

$$R_x(\tau) = E[x_2(t)x_2(t+\tau)] = E[x_1 x_2] \quad \text{where } x_1 = x_2(t) \\ x_2 = x_2(t+\tau)$$

→ missing step

$$= \iint_{-\infty}^{\infty} x_1 x_2 p(x_1, x_2) dx_1 dx_2$$

"ensemble mean"

"expected value"

"average over all possible outcomes"

$$= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{k=1}^N x_2(t) x_2(t+\tau)$$

(b) ergodic process

$$= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x(t) x(t+\tau) dt$$

$$2. E[\hat{R}_x(\tau)] = R_x(\tau) \left(1 + \frac{|\tau|}{T}\right)$$

$$3. \hat{S}_x(f) = \int_{-T}^T \hat{R}_x(\tau) e^{-j2\pi f\tau} d\tau$$

$$= \frac{1}{T} \left| \int_0^T x(t) e^{-j2\pi ft} dt \right|^2$$

$$\hat{S}_x(f) = \frac{1}{T} X(f) \cdot X^*(f)$$

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