

10/2/24

REVIEW of PSD, RANDOM PROCESSES THROUGH LINEAR FILTERS, INTRO TO PARAMETER ESTIMATION

[OS 75 8.3-8.7, 11.0-11.2; NOTES 4.5-5.3]

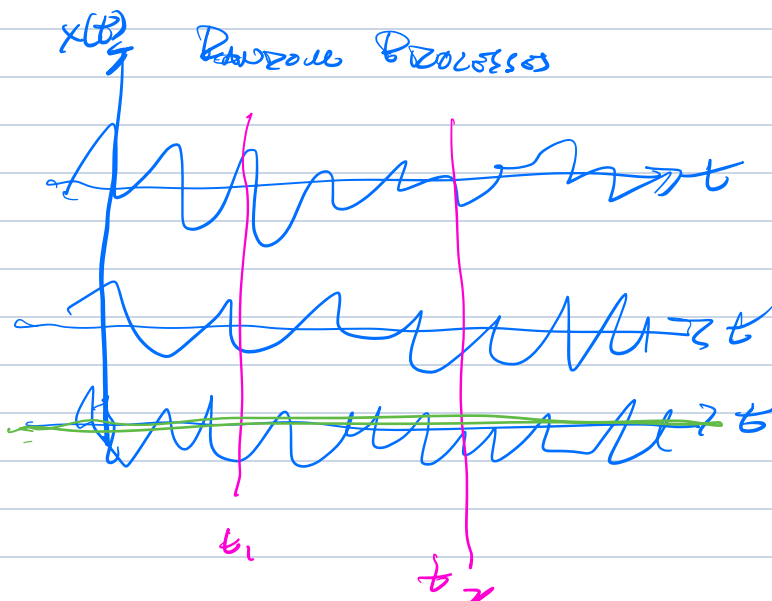
QUIZ 1 10050 WEEK, 11/9

COVERAGE THROUGH PS 5; NO RPS

CLOSED BOOK, ONE SHEET OF NOTES

FRIDAY REVIEW OF Q1 '23

SUNDAY GENERAL REVIEW 2:30



IF $E[x(t)] = \text{CONST.}$
 μ_x

$$\phi_{xx}(t_1, t_2) = E[x(t_1)x(t_2)]$$

DEP. ONLY ON $t_2 - t_1 = \tau$

THEN $x(t)$ WSS

$$E[g(x(t_1))] = \int_{-\infty}^{\infty} g(\alpha) \phi_{x(t_1)}(\alpha) d\alpha$$

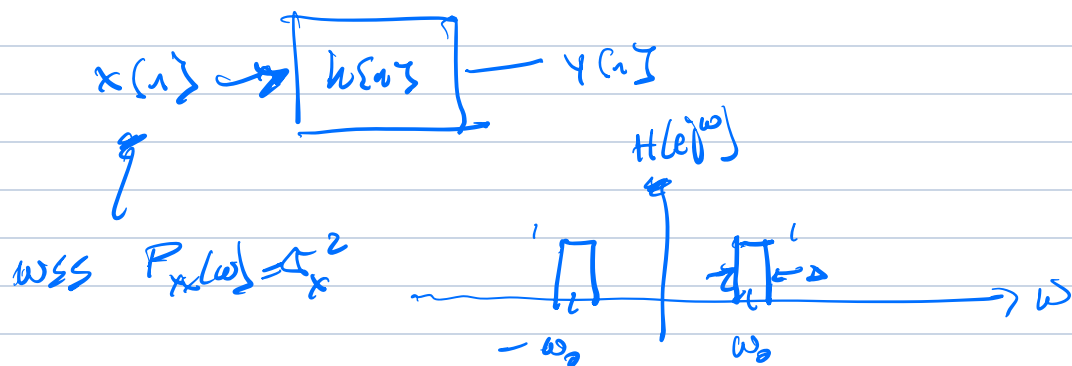
IF $x(t)$ STATIONARY

$$CT \langle g(x(t)) \rangle = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T g(x(t)) dt$$

$$DT \langle g(x[n]) \rangle = \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^N g(x[n])$$

IF $E[g(x(n))]$ = $\langle g(x(n)) \rangle$, RP ERGODIC

PHYSICAL
DEF



$$\lim_{N \rightarrow \infty} \text{output power } \{Y[n]\} = \sigma_x^2 \cdot \frac{2\pi}{2\pi} P_{xx}(\omega) \Big|_{\omega_0}$$

SAMPLE
F.D. DEF.

RP $x(n)$

$$x_N(n) = x(n), |n| \leq N$$

0, ELSE

$$P_{xx}(\omega) = \lim_{N \rightarrow \infty} \frac{1}{2N+1} E \left[\left| \sum_{n=-N}^N x(n) e^{-j\omega n} \right|^2 \right]$$

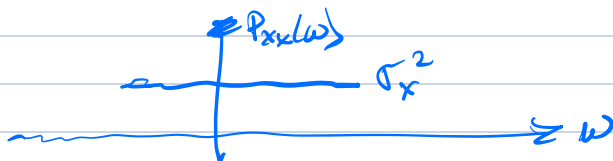
AUTO-CORRELATION F.D.
DEF

$$\text{let } \phi_{xx}[m] = E[x(n)x(n+m)]$$

Wiener-
Khinchin
THEOREM

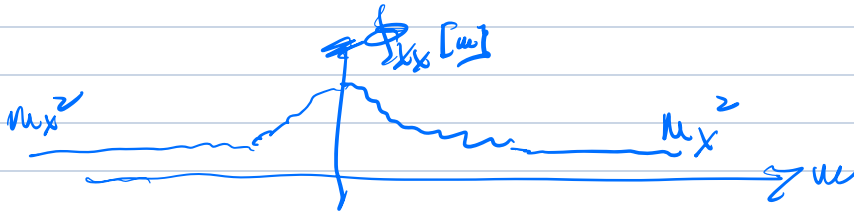
$$P_{xx}(\omega) = \sum_{m=-\infty}^{\infty} \phi_{xx}[m] e^{-j\omega m}$$

IF $P_{xx}(\omega)$ IS CONST. THEN $x(n)$ IS "WHITE" ^{RP}



$$\text{Then } \phi_{xx}[m] = \sigma_x^2 \delta[m]$$

Consider $x[n]$, $m_x \neq 0$



$$\text{Let } x'[n] = x[n] - m_x, \quad m_{x'} = 0$$



IF $x[n]$ IS WHITE, $m_x = 0$

consider $E[x[n]x[n+m]]$

For $m=0$ $E[x[n]x[n+m]] = \phi_{xx}[0] = \sigma_x^2$

For $m \neq 0$ $E[x[n]x[n+m]] = \phi_{xx}[m] = 0$

$$E[x[n]x[n+m]] = E[x[n]]E[x[n+m]]$$

PROPERTIES of AUTO CORRELATION FUN:

UNCORRELATED;
INDEP OF OBSERVATION

* IF $x[n]$ REAL, $\phi_{xx}[m]$ EVEN

$x[n]$ COMPLEX, $\phi_{xx}[m]$ HERMITIAN SYMMETRIC

* IF $x[n]$ IS NOT PERIODIC

$$\phi_{xx}[0] \neq \phi_{xx}[m]$$

★ Line $\phi_{xx}[m] = m_x^2$
 $|m| \rightarrow \infty$

PROPERTIES of PSD FUNCTIONS

* IF $x[n]$ REAL, $P_{xx}(\omega)$ REAL, EVEN

* $P_{xx}(\omega) \geq 0$

* TOTAL POWER $E[x^2[n]] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P_{xx}(\omega) d\omega$

* IF $m_x = E[x[n]] \neq 0$, $P_{xx}(\omega)$ WILL HAVE A TERM $\propto \delta(\omega)$

RANDOM PROCESSES THROUGH LINEAR FILTERS

$$\text{WSS } x[n] \xrightarrow{h[n]} y[n]$$

KNOWN $m_x, \phi_{xx}[m], \phi_{xx}[\omega], P_{xx}(\omega), h[n]$

FIND $m_y, \phi_{yy}[m], P_{yy}(\omega)$

$$y[n] = \sum_{k=-\infty}^{\infty} h[k] x[n-k]$$

$$m_y = E[y[n]] = E\left[\sum_k h[k] x[n-k]\right] = \sum_{k=-\infty}^{\infty} h[k] \underbrace{E[x[n-k]]}_{m_x}$$

$$m_y = m_x \sum_{k=-\infty}^{\infty} h[k]$$

$$\phi_{yy}[m] \equiv E[y[n] y[n+m]]$$

$$= E\left[\underbrace{\sum_{k=-\infty}^{\infty} h[k] x[n-k]}_{y[n]} \underbrace{\sum_{l=-\infty}^{\infty} h[l] x[n+m-l]}_{y[n+m]}\right]$$

$$\phi_{yy}[m] = \sum_{k=-\infty}^{\infty} h[k] \sum_{l=-\infty}^{\infty} h[l] \underbrace{E[x[n-k]x[n+m-l]]}_{\phi_{xx}[m+k-l]}$$

let $r = l - k, l = r + k$

$$\phi_{yy}[m] = \sum_{k=-\infty}^{\infty} h[k] \sum_{r=-\infty}^{\infty} h[r+k] \phi_{xx}[m-r]$$

$$= \sum_{r=-\infty}^{\infty} \phi_{xx}[m-r] \sum_{k=-\infty}^{\infty} h[k] h[r+k]$$

$$\phi_{hh}[r] = h[r] * h[-r]$$

$$= \sum_{r=-\infty}^{\infty} \phi_{xx}[m-r] \phi_{hh}[r] = \phi_{hh}[\omega] * \phi_{xx}[\omega]$$

$$\phi_{yy}[\omega] = \phi_{xx}[\omega] * \phi_{hh}[\omega] = \phi_{xx}[\omega] * h[\omega] * h^*[\omega]$$

\Downarrow

$$P_{yy}(\omega) = P_{xx}(\omega) \cdot |H(e^{j\omega})|^2 = P_{xx}(\omega) |H(e^{j\omega})|^2$$

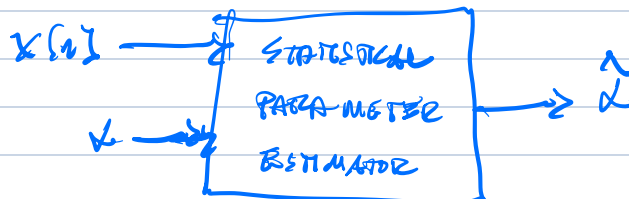
For $h[n]$ DETERMINISTIC.

$$\sum_{n=-\infty}^{\infty} |h[n]|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |H(e^{j\omega})|^2 d\omega$$

WTRD TO PARAMETER ESTIMATION

WSS $x[n]$

Random
PARAMETER α



CHARACTERISTICS of "Good" ESTIMATORS

1. Bias $B = \alpha - E[\hat{\alpha}]$ (want $B=0$)

2. Variance $E[(\hat{\alpha} - E[\hat{\alpha}])^2] = \sigma_{\hat{\alpha}}^2 = E[\hat{\alpha}^2] - \mu_{\hat{\alpha}}^2$

3. Mean-Square Error (MSE) $E[(\hat{\alpha} - \alpha)^2] = \sigma_{\hat{\alpha}}^2 + B^2$

4. Consistency $\hat{\alpha}$ is consistent if
consistent N samples of $x(t)$

$$\lim_{N \rightarrow \infty} B = 0, \quad \lim_{N \rightarrow \infty} \sigma_{\hat{\alpha}}^2 = 0$$