

EE538

DSP I

Module 25

Outline

- Extrapolation of the autocorrelation for a SoS signal
- Basics of Autoregressive (AR) spectral estimation

Test 3 - Session 26

Same format as previous exams., e.g. open book

Problems Topics:

1. Radix-2 FFT
2. DFT and time-domain aliasing
3. Windows and symmetric linear-phase FIR Filter

- problem with nonparametric spectral estimators:
inherently assume

$$r_{xx}[m] = 0 \text{ for } m > M$$

where: $M \leq N-1$

- in fact, though, for SoS (w/o noise)
 $r_{xx}[m] = \sum_{i=1}^P P_i e^{j\omega_i m}$
- does not decay to zero as
 $m \rightarrow \infty$

- in general, sharp spectral peaks in $S_{xx}(\omega)$ imply $r_{xx}[m]$ takes a "long time" to decay to zero as

$$m \rightarrow \infty$$

- interesting observation re: $r_{xx}[m]$ for SoS model \Rightarrow it satisfies recursive relationship

$$r_{xx}[m] = - \sum_{k=1}^p a_k r_{xx}[m-k] \quad \text{for } m \geq P$$

• where: a_k , $k=1, \dots, P$ are the coefficients of the polynomial

$$(z - e^{j\omega_1})(z - e^{j\omega_2}) \dots (z - e^{j\omega_P})$$

$$= z^P + a_1 z^{P-1} + a_2 z^{P-2} + \dots + a_{P-1} z + a_P$$

$$+ a_1 z^{-1} + a_2 z^{-2} + \dots + a_{P-1} z^{-(P-1)} + a_P z^{-P} = 0$$

$$\text{for } z_i = e^{j\omega_i}$$

$$\left\{ \begin{array}{l} 1 + a_1 e^{j\omega_i(-1)} + a_2 e^{j\omega_i(-2)} \\ \vdots \\ + \dots + a_{p-1} e^{j\omega_i(-(p-1))} + a_p e^{j\omega_i(-p)} \end{array} \right\} = 0$$

Multiply both sides by $P_i e^{j\omega_i m}$

• Sum over all i (since $\sum \alpha = 0$)

$$\sum_{i=1}^P P_i e^{j\omega_i m} + a_1 \sum_{i=1}^P P_i e^{j\omega_i (m-1)} + a_2 \sum_{i=1}^P P_i e^{j\omega_i (m-2)} + \dots + a_{P-1} \sum_{i=1}^P P_i e^{j\omega_i (m-(P-1))} + a_P \sum_{i=1}^P P_i e^{j\omega_i m} = 0$$

- ultimately:

$$r_{xx}[m] = - \sum_{k=1}^P a_k r_{xx}[m-k] \quad \text{for } m \geq P$$

- End of Proof

- Thus : if :

- if we estimate $r_{xx}[m]$ for

- $m = 0, 1, \dots, P$ AND

- determine the coefficients

- $a_k, k = 1, \dots, P$

- we could extrapolate $r_{xx}[m]$ for all m , thereby averting a windowing effect
- ostensibly use recursion to determine $r_{xx}[m]$ for $m \geq P$ and form $S_{xx}(\omega) = \sum_{m=-\infty}^{\infty} r_{xx}[m] e^{-j\omega m}$
- however, in case of SoS, typically interested in frequencies, ω_i , which may be determined from roots of $z^P + a_0 z^{P-1} + \dots + a_{P-1} z + a_P$

- note: $r_{xx}[-m] = r_{xx}^*[m]$

- How do we estimate

$a_k, k=1, \dots, P$ given

$r_{xx}[m], m=0, 1, \dots, P$

- $r_{xx}[m] = - \sum_{k=1}^P a_k r_{xx}[m-k]$

- write out P eqns. in

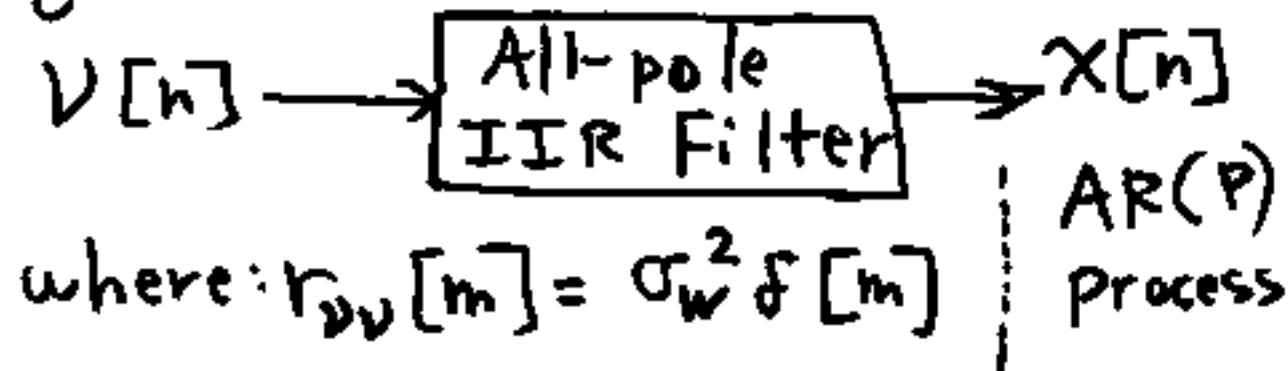
P unknowns $\Rightarrow m=1, 2, \dots, P$

$$\begin{bmatrix} r_{xx}[0] & r_{xx}[-1] \dots r_{xx}[-(P-1)] \\ r_{xx}[1] & r_{xx}[0] \dots r_{xx}[-(P-2)] \\ \vdots & \vdots \dots \vdots \\ r_{xx}[P-1] & r_{xx}[P-2] \dots r_{xx}[0] \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_p \end{bmatrix} = \begin{bmatrix} r_{xx}[1] \\ r_{xx}[2] \\ \vdots \\ r_{xx}[P] \end{bmatrix}$$

$$\underbrace{\begin{matrix} R \\ P \times P \end{matrix}}_{\text{P} \times \text{P}} \quad \underbrace{\begin{matrix} \underline{a} \\ P \times 1 \end{matrix}}_{\text{P} \times 1} = - \underbrace{\begin{matrix} \underline{r} \\ P \times 1 \end{matrix}}_{\text{P} \times 1} \quad \left. \begin{array}{l} \text{See} \\ \text{SoS extrop.m} \\ \text{at web} \\ \text{site} \end{array} \right\}$$

- these observations for the SOS model are the basis for the following spectral estimation techniques:
 - Auto Regressive (AR) Method
 - MUSIC (Multiple Signal Classification)
 - Sect. 12.5.3
 - ESPRIT (Estimation of Signal Parameters by Rotational Invariance Techniques)
 - Sect. 12.5.4

- AR Spectral Estimation
- for sole purpose of matching the spectral density of a WSS random process, $X[n]$, we model $X[n]$ as having been generated as:



$$\bullet x[n] = - \sum_{k=1}^P a_k x[n-k] + v[n]$$

• can show that $r_{xx}[m]$ satisfies

$$r_{xx}[m] = - \sum_{k=1}^P a_k r_{xx}[m-k] + \sigma_w^2 \delta[m]$$

- in freq. domain:

$$X(\omega) = H(\omega) V(\omega)$$

- where: $H(\omega) = \frac{1}{1 + \sum_{k=1}^p a_k e^{-j k \omega}}$

- from basis random process theory

$$S_{xx}(\omega) = |H(\omega)|^2 S_{vv}(\omega)$$

- where: $S_{yy}(\omega) = \text{DTFT}\{r_{yy}[m]\}$
 $= \text{DTFT}\{\sigma_w^2 \delta[m]\}$
 $= \sigma_w^2 \text{ for all } \omega$

- thus:

$$S_{xx}(\omega) = \text{DTFT}\{r_{xx}[m]\}$$
 $= \frac{\sigma_w^2}{\left| 1 + \sum_{k=1}^P a_k e^{-jk\omega} \right|^2}$