Lecture 12

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Agenda today

- How to determine the ACF/PSD for the signal models
- What type of ACF/PSD can these models describe

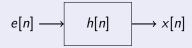
Learning objectives: L12

After today's lecture you should be able to

- Sketch the PSD of an AR, MA and ARMA model given poles and zeros of the transfer function.
- Calculate the ACF for an MA process.
- Derive the Yule-Walker equations used to compute the ACF of an AR process
- Estimate an AR process from data.

Signal models: AR, MA and ARMA processes

A stochastic model for x[n]



where e[n] is WSS, white noise, $\Rightarrow E\{e[n]\} = 0$ and

$$r_e[k] = \mathsf{E}\{e[n]e[n+k]\} = \delta[k]\sigma_e^2 = \begin{cases} \sigma_e^2 & \text{if } k = 0\\ 0 & \text{otherwise.} \end{cases}$$

It follows that

•
$$P_e(e^{j\omega}) = DTFT\{r_e[k]\} = \sigma_e^2 DTFT\{\delta[k]\} = \sigma_e^2$$
.

•
$$H(e^{j\omega}) = DTFT\{h[k]\}$$

•
$$P_{x}(e^{j\omega}) = |H(e^{j\omega})|^{2}\sigma_{e}^{2}$$

The filter H(z) shape the spectrum of the process x[n].

Definition of ARMA

x[n] is an ARMA(p,q)-process if

$$H(z) = \frac{B(z)}{A(z)} = \frac{1 + b_1 z^{-1} + \dots + b_q z^{-q}}{1 + a_1 z^{-1} + \dots + a_p z^{-p}} = \frac{b(z)}{a(z)}$$

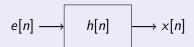
or, equivalently, if

$$x[n] + \cdots + a_p x[n-p] = e[n] + \cdots + b_q e[n-q] \quad (1)$$

- Important special cases:
 - 1) x[n] is AR(p) if B(z) = 1,
 - 2) x[n] is MA(q) if A(z) = 1

PSD of ARMA

The PSD of an ARMA



• We get

$$P_{x}\left(e^{j\omega}\right) = P_{e}\left(e^{j\omega}\right)\left|H\left(e^{j\omega}\right)\right|^{2} = \sigma_{e}^{2}\left|H\left(e^{j\omega}\right)\right|^{2}$$

 \rightsquigarrow the filter H(z) shapes the PSD of x[n].

• Note: e[n] is white noise

$$\Rightarrow$$
 $r_e[n] = \delta[n]\sigma_e^2 \longleftrightarrow P_e\left(e^{j\omega}\right) = \sigma_e^2$.

ACF of ARMA

- Techniques to compute the ACF:
 - 1) Find the PSD as above and compute

$$r_{x}[n] = IDTFT \left\{ P_{x} \left(e^{j\omega} \right) \right\}$$

- → general solution, but often rather complicated.
- 2) Derive $h[n] = IDTFT \{H(e^{j\omega})\}$ and compute

$$r_{\mathsf{x}}[n] = \sigma_{\mathsf{e}}^2 h[n] \star h[-n] = \sum_{k=-\infty}^{\infty} h[k] h[k-n]$$

- → another possible, but complicated, alternative.
- 3) Derive linear equations from the difference equation (1)

 - Only works for AR and MA processes.
 - Second Enables us to estimate AR-parameters from data.

ACF of AR-processes

Consider an AR(p)-process

$$x[n] + a_1x[n-1] + \cdots + a_px[n-p] = e[n]$$

• Multiply by x[n-k], $k \ge 0$, and take expectations:

Yule-Walker (YW) equations

$$r_x[k] + a_1 r_x[k-1] + \dots + a_p r_x[k-p] = \sigma_e^2 \delta[k]$$

Note: 1) YW are linear in the autocorrelation function, r_x[n]
2) by changing k, we can collect any number of equations.
⇒ easy to find r_x[n] using YW