

# *Discrete-Time random processes*

*Andreas Polydoros*  
*University of Athens*  
*Dept. of Physics*  
*Electronics Laboratory*

# Contents

- Review continuous-time processes
- Generation of discrete-time process
- LTI filtering – Examples
- Z-domain description – Examples

# Review of C-T R.P.

- ◆ Second order statistical description of a *real* continuous-time (C-T) random process (R.P.)  $X(t; u)$ :
  - Mean:  $m_X(t) \triangleq \mathcal{E}\{X(t; u)\}$
  - Auto-correlation:  $R_X(t_1, t_2) \triangleq \mathcal{E}\{X(t_1)X(t_2)\}$
- ◆ For a WSS random process:

$$\begin{aligned} R_X(t_1, t_2) &= R_X(t_1 - t_2) \\ &= R_X(\tau) \xleftrightarrow{\mathcal{F}} S_X(f) \end{aligned}$$

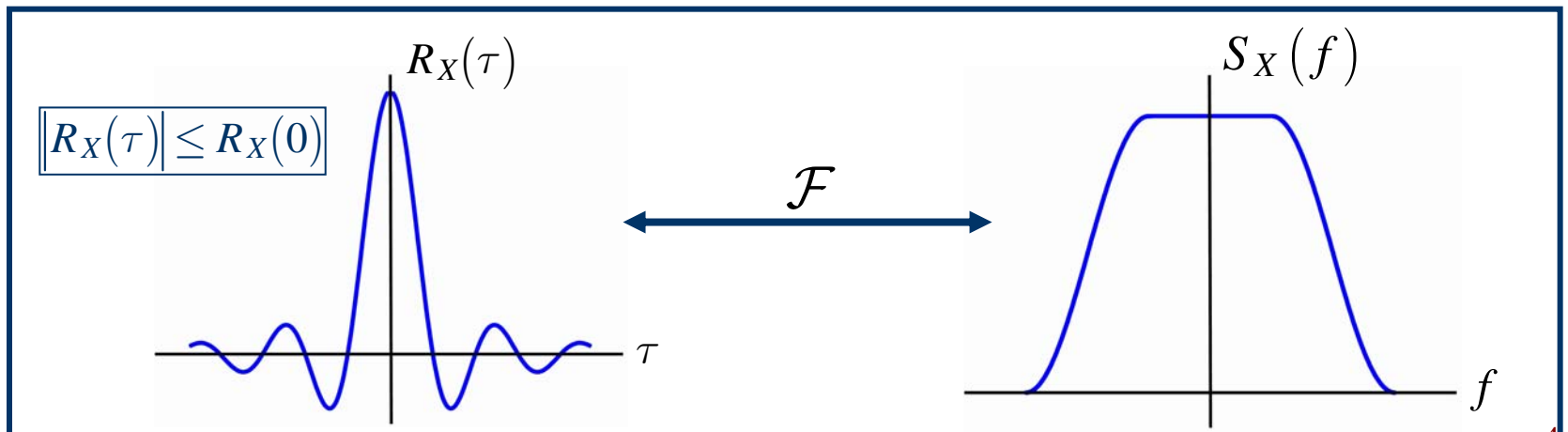
# Properties of $R_X(\tau)$ and $S_X(f)$

- ◆  $R_X(\tau)$  is symmetric and positive definite

$$R_X(\tau) = R_X(-\tau); \quad \sum_i \sum_j a_i a_j R_X(t_1 - t_2) \geq 0$$

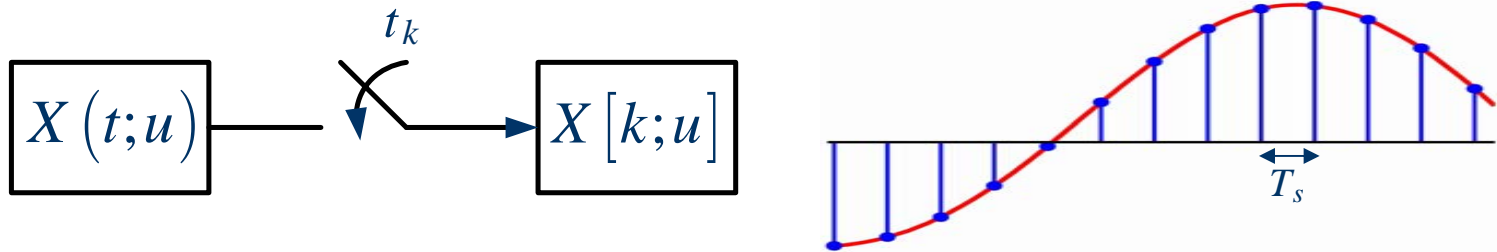
- ◆  $S_X(f)$  is symmetric and positive

$$S_X(f) = S_X(-f); \quad S_X(f) \geq 0$$



# Generation of a D-T R.S. by Sampling

- ◆ A discrete-time (D-T) random *sequence* (R.S.)  $\{X[k;u]\}$  can be generated by sampling a C-T R.P.  $X(t;u)$  every  $T_s$  (i.e., at sampling instants  $t_k = kT_s$ ;  $k = 0, \pm 1, \pm 2, \dots$ )

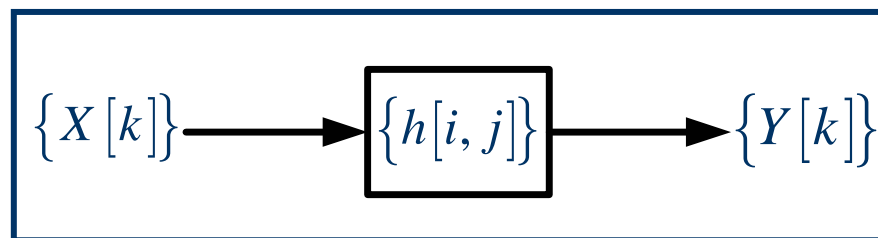


- ◆ The second order statistical description of  $\{X[k;u]\}$  is:
  - $m[k] \triangleq \mathcal{E}\{X[k]\} = \mathcal{E}\{X(kT_s)\} = m_X(kT_s)$
  - $R_X[i, j] \triangleq \mathcal{E}\{x[i]x[j]\} = R_X(iT_s, jT_s)$   
 $\stackrel{\text{W.S.S.}}{=} R_X((i-j)T_s)$   
 $\triangleq R_X[m] \quad (m = i - j)$

# LTI Filtering of D-T R.S. (1/2)

- ◆ In general, a linear filtering operation on a D-T R.S. is described by:

$$Y[k;u] = \sum_{j=-\infty}^{\infty} h[k, j] X[j;u]$$



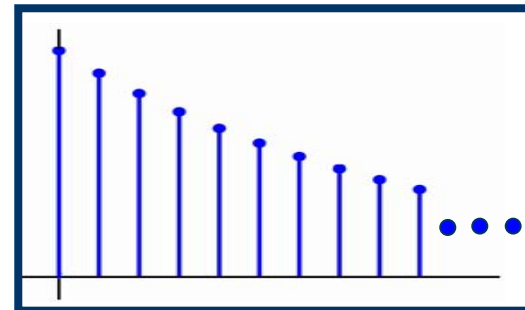
- ◆ If the system is LTI (i.e.,  $h[i, j] = h[i - j]$ )

$$\begin{aligned} Y[k] &= \sum_{j=-\infty}^{\infty} h[k - j] X[j] \\ &= \sum_{n=-\infty}^{\infty} h[n] X[k - n] \\ &= h[k] * X[k] \end{aligned}$$

# LTI Filtering of D-T R.S. (2/2)

- ◆ If the system is causal (i.e.,  $h[n] = 0; n < 0$ )

$$\begin{aligned} Y[k] &= \sum_{n=0}^{\infty} h[n] X[k-n] \\ &= h[0] X[k] + h[1] X[k-1] + \dots + h[l] X[k-l] + \dots \end{aligned}$$



- ◆ If the system is causal and finite (i.e.,  $h[n] = 0; n \notin [0, L]$ )

$$Y[k] = \sum_{n=0}^L h[n] X[k-n]$$

# Second order statistics of the output R.S.

- ◆ Starting from  $Y[k] = h[k] * X[k]$ , it can be shown that

- $m_Y[k] = h[k] * m_X[k]$

- $R_Y[m] = h[m] * R_X[m] * h[-m]$

or  $R_Y[m] = R_X[m] * r_h[m]$

where  $r_h[m] \triangleq h[m] * h[-m] = \sum_{j=-\infty}^{\infty} h[j]h[-(m-j)]$

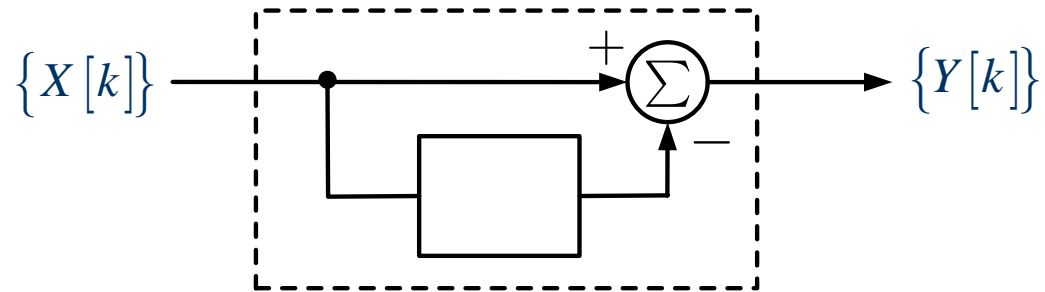
$$= \sum_{j=-\infty}^{\infty} h[j]h[j-m]$$

is the *deterministic* autocorrelation of  $\{h[m]\}$

- ◆ Note that  $r_h[0] = \sum_{j=-\infty}^{\infty} h^2[j] = \text{energy of } \{h[k]\}$   
and  $r_h[m] = r_h[-m]$



# Example 1 (1/2)

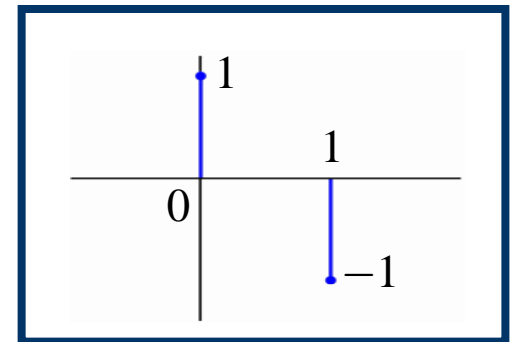


- ◆ The system is described by the differential equation

$$Y[k] = X[k] - X[k-1] \quad \forall k$$

- ◆ or by the impulse response

$$h[k] = \delta[k] - \delta[k-1] = \begin{cases} 1 & ; \quad k = 0 \\ -1 & ; \quad k = 1 \\ 0 & ; \quad \text{otherwise} \end{cases}$$



# Example 1 (2/2)

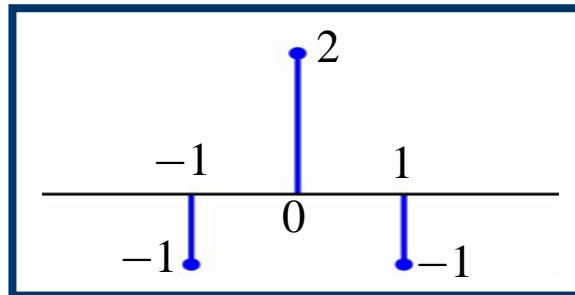
- ◆ The deterministic autocorrelation of  $\{h[m]\}$  is

$$r_h[m] = h[m] * h[-m]$$

$$= (\delta[m] - \delta[m-1]) * (\delta[m] - \delta[m+1])$$

$$= 2\delta[m] - \delta[m-1] - \delta[m+1]$$

Recall that  
 $\delta[-k] = \delta[k]$

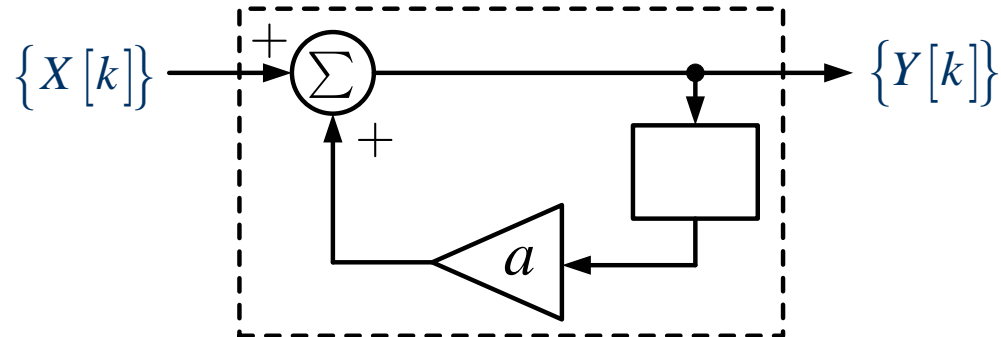


- ◆ Therefore,  $R_Y[m]$  is

$$R_Y[m] = R_X[m] * r_h[m]$$

$$= 2R_X[m] - R_X[m-1] - R_X[m+1]$$

## Example 2 (1/2)

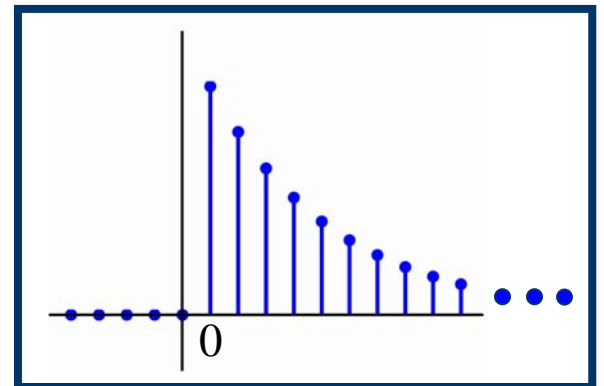


- ◆ Here, the equation describing the system is

$$Y[k] = aY[k-1] + X[k] \quad \forall k$$

- ◆ and the impulse response is

$$h[k] = \begin{cases} 0 & ; k < 0 \\ a^k & ; k \geq 0 \end{cases}$$



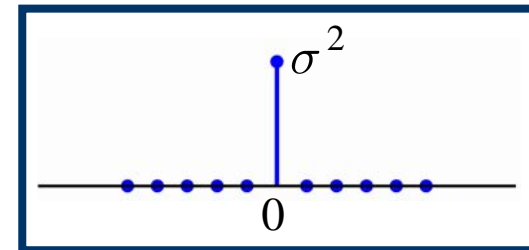
## Example 2 (2/2)

- ◆ The deterministic autocorrelation is (prove it):

$$r_h[m] = a^{|m|} / (1 - a^2)$$

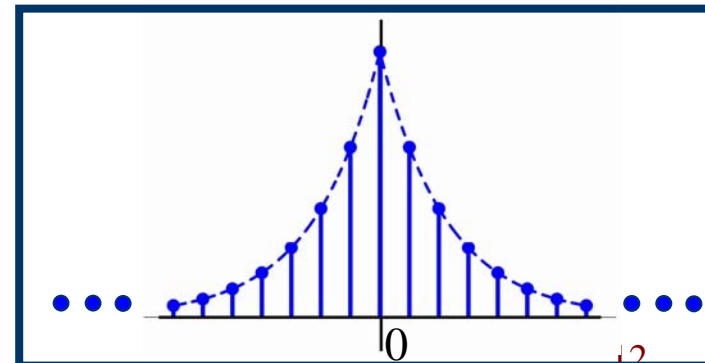
- ◆ If the input sequence  $\{X[k]\}$  is white, i.e.,

$$R_X[m] = \begin{cases} \sigma^2 & ; m = 0 \\ 0 & ; m \neq 0 \end{cases}$$



then, the output sequence autocorrelation is

$$R_Y[m] = \frac{\sigma^2}{1 - a^2} a^{|m|}$$



# Autoregressive processes

- ◆ The R.P.  $\{Y[k]\}$  described by

$$Y[k] + a_1Y[k-1] + \cdots + a_MY[k-M] = W[k]$$

where  $\{W[k]\}$  is a white sequence, is called an autoregressive process of order  $M$ .

# Z-domain Description of WSS R.S.

- ◆ Given a WSS R.S.  $\{X[k]\}$ , we form the Z-transform  $S_X(z)$  of the autocorrelation  $R_X[m]$ :

$$S_X(z) = \mathcal{Z}\{R_X[m]\} \triangleq \sum_{m=-\infty}^{\infty} R_X[m] z^{-m}$$

- ◆ The power spectrum of  $\{X[k]\}$  is  $S_X(e^{j\omega})$  (i.e.,  $S_X(z)$  evaluated on the unit circle)
- ◆ Question: What is the Z-domain relation between a system's input and output sequences

# Z-Domain Filtering

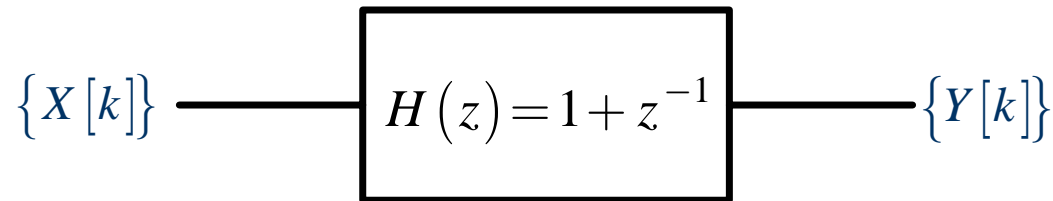
- ◆ The impulse response has a Z-transform  $\mathcal{Z}\{h[k]\} = H(z)$
- ◆ The Z-transform of  $r_h[m]$  is

$$\begin{aligned}\mathcal{Z}\{r_h[m]\} &= \mathcal{Z}\{h[m]*h[-m]\} = \mathcal{Z}\{h[m]\} \cdot \mathcal{Z}\{h[-m]\} \\ &= H(z)H(z^{-1}) \quad (\text{real}\{h[k]\})\end{aligned}$$

- ◆ Thus, applying the Z-transform on both sides of the equation  $R_Y[m] = R_X[m]*r_h[m]$ , we have

$$\begin{aligned}\mathcal{Z}\{R_Y[m]\} &= \mathcal{Z}\{R_X[m]\} \mathcal{Z}\{r_h[m]\} \\ \Rightarrow \boxed{S_Y(z) &= S_X(z)H(z)H(z^{-1})}\end{aligned}$$

## Example 3



- ◆ We want to find  $S_Y(z)$ . The system shown above is similar to the one in example 1. It can be found that

$$R_Y[m] = 2R_X[m] + R_X[m-1] + R_X[m+1]$$
$$\xrightarrow{z} S_Y(z) = 2S_X(z) + z^{-1}S_X(z) + zS_X(z)$$

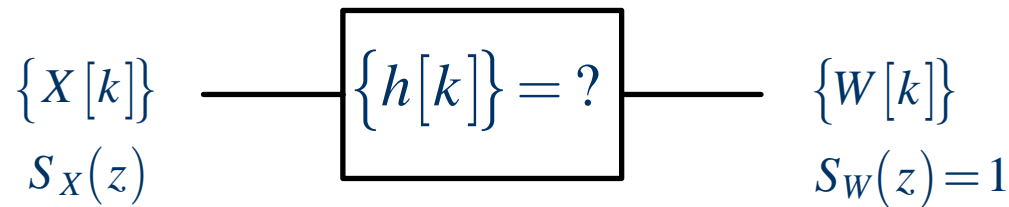
- ◆ Another way to find  $S_Y(z)$  would be as

$$S_Y(z) = S_X(z)H(z)H(z^{-1})$$
$$= S_X(z)(1+z)(1+z^{-1})$$

which is of course, the same result



## Example 4 (1/3)



- ◆ **Whitening Problem:** What is the required impulse response  $\{h[k]\}$  so that the output is white?
- ◆ **Solution:**  $S_X(z)$  is factorized as  $S_X(z) = G(z)G(z^{-1})$ . We then have:

$$1 = S_X(z)H(z)H(z^{-1})$$

$$1 = G(z)G(z^{-1})H(z)H(z^{-1})$$

$$1 = (G(z)H(z))(G(z^{-1})H(z^{-1})) \Rightarrow H(z) = \frac{1}{G(z)}$$

## Example 4 (2/3)

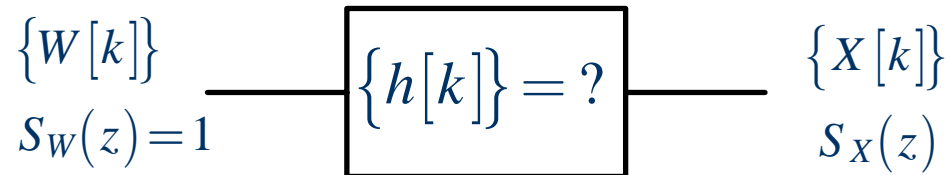
- ◆ Consider the case where  $S_X(z) = 1/(2 + z + z^{-1})$ . In order to find a whitening filter,  $S_X(z)$  must be factorized. This can be accomplished as follows:

$$\begin{aligned} S_X(z) &= \frac{1}{z^{-1}(2z + z^2 + 1)} = \frac{1}{z^{-1}(z+1)^2} \\ &= \frac{1}{z^{-1}(z+1)(z+1)} \\ &= \frac{1}{(1+z^{-1})(1+z)} \\ &= \frac{1}{1+z^{-1}} \cdot \frac{1}{1+z} \end{aligned}$$

## Example 4 (3/3)

- ◆ There are two possible choices for  $G(z)$  (and also for  $H(z)$ )
- ◆ For  $G(z) = 1/(1 + z^{-1}) \Rightarrow H(z) = 1/G(z) = 1 + z^{-1}$   
 $\Rightarrow h[k] = \delta[k] + \delta[k - 1]$
- ◆ For  $G(z) = 1/(1 + z) \Rightarrow H(z) = 1/G(z) = 1 + z$   
 $\Rightarrow h[k] = \delta[k] + \delta[k + 1]$
- ◆ Both choices are valid. The first filter is causal, whereas the second is non-causal.

# Example 5



- ◆ **Coloring Problem:** What is the required impulse response  $\{h[k]\}$  so that for a white input the output has spectrum  $S_X(z)$ ?

- ◆ **Solution:**  $S_X(z)$  is factorized as  $S_X(z) = G(z)G(z^{-1})$ . We then have:  $S_X(z) = 1 \cdot H(z)H(z^{-1})$

$$G(z)G(z^{-1}) = H(z)H(z^{-1})$$

$$G(z)\underbrace{U(z)U(z^{-1})}_{=1}G(z^{-1}) = H(z)H(z^{-1})$$

- ◆ The required filter is  $H(z) = G(z) = U(z)G(z)$