

Lecture 4

Note Title

8/4/2008

Discrete-time signals: $x[n]$ $-\infty < n < \infty$

↓
sequence of complex numbers

Finite-energy sequences: $\sum_{n=-\infty}^{\infty} |x[n]|^2 < \infty$

Convolution:

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

DTFT

$$X(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x[n] e^{-jn\omega}$$

ZT

$$X(z) = \sum_{n=-\infty}^{\infty} x[n] z^{-n}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{jn\omega} d\omega$$

LTI system :
(linear filter)



$H(e^{j\omega}), H(z)$

freq. resp.

transfer function

→ Roc

$h[n]$: right-sided sequence

$$h[n] = 0 \quad \text{for } n \leq M$$

→ right-sided and stable filters



$$\sum |h[n]|^2 < \infty$$

monic & causal

$$h[0] = 1$$

→ minimum-phase: poles & zeros are inside unit-circle.

Folded spectrum:

$$q(t) \xleftrightarrow{FT} Q(f)$$

$$q[n] = q(nT) \quad f_s = 1/T$$

$$q[n] \xleftrightarrow{DTFT} \tilde{Q}(e^{j\omega}) = \frac{1}{T} \sum_{m=-\infty}^{\infty} Q\left(f - \frac{m}{T}\right)$$

$$\omega = 2\pi fT = 2\pi \frac{f}{f_s}$$

No-aliasing condition:

$$f_s = \frac{1}{T} > 2 \text{BW}(q(t))$$

Random Signals

Probability & Random Variables:

→ Random variable, CDF, PDF
(PMF)

→ Mean, Expectation, Variance

→ Joint PDF, Vectors of RVs

→ Conditional probability:

→ discrete & continuous

→ Conditional pdf $X & Y$: RVs

$$f_{X|Y}(x|y) = \frac{f_{XY}(x,y)}{f_Y(y)}$$

Gaussian RV
(Normal)

$$X \sim N(\mu, \sigma^2)$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$X \sim N(0, 1)$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

$$E[X] = \mu$$

$$E[X^2] = \mu^2 + \sigma^2$$

$$Q(x) = \int_x^{\infty} \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt = \Pr(X \geq x)$$

$$Q(x) \leq \frac{1}{\sqrt{2\pi}x} e^{-x^2/2}$$

$X, Y \sim$ Jointly Gaussian

$$f_{XY}(x, y) = \frac{1}{2\pi\sigma^2\sqrt{1-\rho^2}} e^{-\frac{(x^2 - 2\rho xy + y^2)}{2\sigma^2(1-\rho^2)}}$$

$$X \sim N(0, \sigma^2) \quad Y \sim N(0, \sigma^2)$$

$$\rho = \frac{E[XY]}{\sigma^2}$$

$$\underline{X} = [x_1 \ x_2 \ \dots \ x_n]^T \quad X_i \sim N(0, 1), (\text{iid})$$

$$f_{\underline{X}}(\underline{x}) = \frac{1}{(2\pi\sigma^2)^{n/2}} e^{-\frac{\|\underline{x}\|^2}{2\sigma^2}}$$

General case: $\underline{m} = [E[x_1] \dots E[x_n]]^T$

$$\underline{C} = E[(\underline{x} - \underline{m})(\underline{x} - \underline{m})^T]$$

$$[C]_{ij} = E[(x_i - m_i)(x_j - m_j)]$$

$$f_{\underline{x}}(\underline{x}) = \frac{1}{(2\pi)^{n/2} |\underline{C}|^{1/2}} e^{-\frac{1}{2}(\underline{x} - \underline{m})^T \underline{C}^{-1}(\underline{x} - \underline{m})}$$

→ Linear combinations of jointly Gaussian RVs
are jointly Gaussian

Random Processes:

DT

$\{X_k\}$: sequence of
RVs

CT

$\{X(t)\}$:

$X(t)$: RV for each t

Finite distributions:

joint pdf/pmf of a finite collection of RVs

$X_{k_1}, X_{k_2}, \dots, X_{k_n}$

$\forall N$ and k_i

$X(t_1), X(t_2), \dots, X(t_N)$

$\forall N, t_i$

Ex: $\sum A_k \cos(2\pi f_k t + \theta)$