

## EEL 5544 Noise in Linear Systems Lecture 34

CONDITIONAL DENSITIES

- We will use the notation  $f_{X,n}(\cdot|\cdot)$  to denote a general conditional density, where  $f_{X,n}(\cdot)$  is a joint density

**EX:** The **conditional density** for  $X(t_1)$  given  $X(t_2)$ ,  $t_1 \neq t_2$  is

$$f_{X,2}(x_1; t_1 | x_2; t_2) = \frac{f_{X,2}(x_1, x_2; t_1, t_2)}{f_{X,1}(x_2; t_2)}.$$

GAUSSIAN RPS

**DEFN**  $X(t)$  is a **Gaussian random process** if for each  $n \in \mathbf{Z}^+$ , its  $n$ -dimensional distribution function is a **Gaussian distribution**.

- Note that each of the  $n$ -dimensional distribution functions is completely specified by the appropriate **means** and **covariances**.

Alternate definition:

**DEFN**  $X(t)$  is a **Gaussian random process** if:

- for each  $n \in \mathbf{Z}^+$ ,
- any real coefficients  $\{a_k : 1 \leq k \leq n\}$
- and any sampling times  $\{t_k \in \mathbf{T} : 1 \leq k \leq n\}$ ,

the random variable

$$a_1 X(t_1) + a_2 X(t_2) + \dots + a_n X(t_n)$$

is a **Gaussian random variable**.

MEAN, AUTOCORRELATION, AND  
AUTOCOVARANCE FUNCTIONS

- It is generally not practical, and often unnecessary to determine  $n$ -dimensional distribution or density functions for RPs
- In many problems, it is sufficient to know about \_\_\_\_\_ statistics and distribution functions

**DEFN** A random process  $X(t)$  is a \_\_\_\_\_ if  
 $E \{[X(t)]^2\} < \infty$  for each  $t$ .

For second-order RPs, we define the following functions:

**DEFN** The \_\_\_\_\_ (or just \_\_\_\_\_) is given by

$$\mu_X(t) =$$

**DEFN** The \_\_\_\_\_ is given by

$$R_X(t_1, t_2) =$$

**DEFN** The \_\_\_\_\_ is given by

$$C_X(t_1, t_2) =$$

**Interpretation:**

- At each  $t$ ,  $\mu_X(t)$  is the expected value of the random variable  $X(t)$ 
  - $\mu_X(t)$  depends only on the one-dimensional density  $f_{X,1}(x; t)$
- $R_X(t_1, t_2)$  and  $C_X(t_1, t_2)$  measure the statistical “coupling” or “dependence” between  $X(t_1)$  and  $X(t_2)$ 
  - $R_X(t_1, t_2)$  and  $C_X(t_1, t_2)$  depend only on the one- and two-dimensional densities
  - if  $X(t_1)$  and  $X(t_2)$  are statistically independent, then  $C_X(t_1, t_2) = 0$

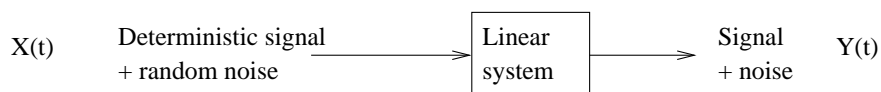
- if  $C_X(t_1, t_2) = 0$ , then  $X(t_1)$  and  $X(t_2)$  are uncorrelated

- Note that

$$C_X(t_1, t_2) = R_X(t_1, t_2) - \mu_X(t_1)\mu_X(t_2),$$

so generally only need to know one of  $C_X$  and  $R_X$

### Application preview: Noise in Linear Systems



What is the noise power at input and output?

- Input: Noise power =  $E \{[X(t)]^2\} = R_X(t, t)$
- Output: Noise power = function of  $R_X(t_1, t_2)$  and **not** just  $R_X(t, t)$

Given  $X(t)$  with finite power ( $X(t)$  a second-order RP), are  $\mu_X$ ,  $R_X$ , and  $C_X$  all finite?

**DEFN** The *normalized autocovariance* of a random process  $X(t)$  is given by

$$K_X(t_1, t_2) = \frac{C_X(t_1, t_2)}{\sigma_1 \sigma_2},$$

where  $\sigma_i = \sqrt{\text{Var}[X(t_i)]}$ .

- Note that

$$-1 \leq K_X(t_1, t_2) \leq 1$$

### Examples