Chapter 2 Stochastic Processes

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Chapter 2: Stochastic Processes

- 2.1 Basic Concepts
- 2.2 Stationary Stochastic Processes
- 2.3 Properties of Correlation Functions
- 2.4 Some Important Stochastic Processes

2.3 Properties of Correlation Functions

OUTLINE:

- 2.3.1 Properties of $R_{XX}(t_1,t_2)$
- 2.3.2 Properties of $R_{XY}(t_1, t_2)$

2.3.1 Properties of $R_{XX}(t_1,t_2)$

1.Definition of correlation function for a stochastic process $\{X(t), t \in T\}$ recall

Review

4. Correlation(autocorrelation) functions

- For a second-order moment process, the correlation function is $R_{XX}(t_1,t_2)=E[X(t_1)X(t_2)]$ or $R_{XX}(t,t+\tau)=E[X(t)X(t+\tau)]$
 - Calculate correlation function for discrete stochastic processes and continuous stochastic processes respectively:

$$R_{XX}(t_1, t_2) = E[X(t_1)X(t_2)]$$

$$= \sum_{k=1}^{n} \sum_{j=1}^{n} kj P(x(t_1) = k, x(t_2) = j)$$

$$R_{XX}(t_1, t_2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x(t_1)x(t_2)f_X(x_1, x_2; t_1, t_2)dx_1dx_2$$

2.3.1 Properties of $R_{XX}(t_1,t_2)$

Relationship with other moments

$$R_{XX}(t,t) = E[X^{2}(t)]$$

$$C_{XX}(t_1, t_2) = R_{XX}(t_1, t_2) - \overline{x}(t_1)\overline{x}(t_2)$$

$$|R_X(t_1,t_2)| \leq \frac{E[X(t_1)]^2 + E[X(t_2)]^2}{2}$$

$$|R_{XX}(t_1, t_2)| \le \frac{R_{XX}(t_1, t_1) + R_{XX}(t_2, t_2)}{2}$$

2.3.1 Properties of $R_{XX}(t_1,t_2)$

2. Properties of $R_{XX}(t_1,t_2)$

i)
$$R_{XX}(t,t) \ge 0$$

ii)
$$R_{XX}(t_1,t_2) = R_{XX}(t_2,t_1)$$

iii)
$$|R_X(t_1, t_2)| \leq \sqrt{R_X(t_1, t_1)R_X(t_2, t_2)}$$

iv) Nonnegative definite (positive semi-definite).

$$\sum_{i=1}^n \sum_{k=1}^n \alpha_i \alpha_k^* R_X(t_i, t_k) \ge 0$$

Nonnegative definiteness implies i), ii), iii).

If a function R(*,*) is nonnegative definite, it can be a correlation function for some second-order random processes.

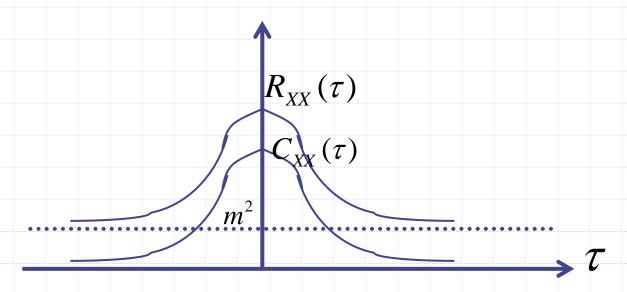
2.3.1 Properties of $R_{XX}(t_1, t_2)$

3. Correlation function of weakly stationary S.P.

$$R_{XX}(t_1, t_2) = R_{XX}(\tau)$$

$$R_{XX}(0) = E[X^2(t)]$$

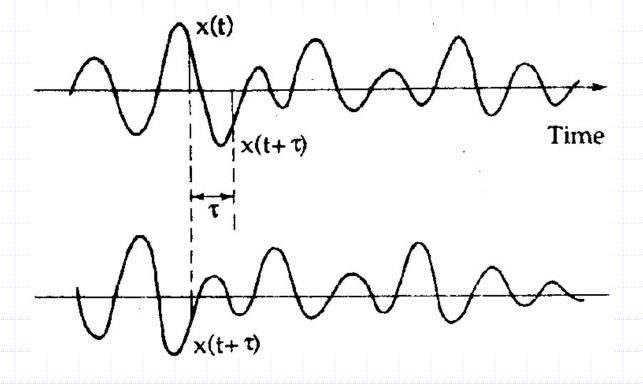
$$C_{XX}(\tau) = R_{XX}(\tau) - m^2$$



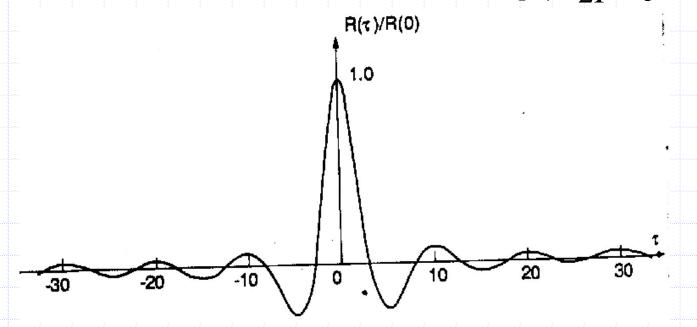
2.3.1 Properties of $R_{XX}(t_1,t_2)$

If X(t) is ergodic, then

$$R_{XX}(\tau) = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} x(t)x(t+\tau)dt$$



- i) An even function: $R_{XX}(\tau) = R_{XX}(-\tau)$
- ii) $R_{XX}(0) \ge |R_{XX}(\tau)|$
- iii) $R_{XX}(0) = E[X^2(t)] \ge 0$
- iv) If X(t) is ergodic, then $R_{XX}(0) = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} x^2(t) dt$



V) If X(t)=X(t+T), then
$$R_{XX}(\tau) = R_{XX}(\tau+T)$$

Example 1 Random phase processes

 $X(t)=Acos(w_0t+\varepsilon)$, t>0, whereas A and w_0 are constants and ε is random variable uniformly distributed between $-\pi$ and π .

$$E[x(t)] = 0$$

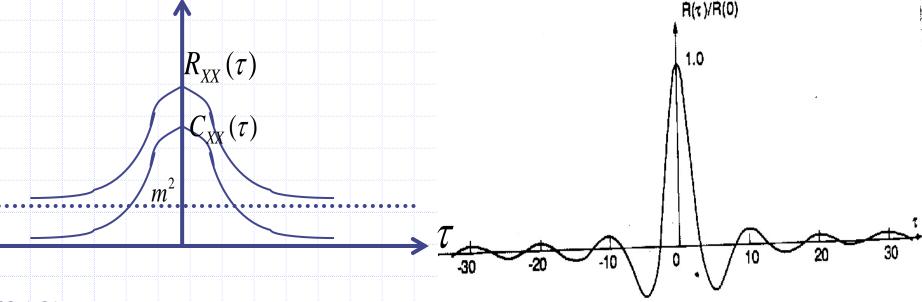
$$C_{XX}(\tau) = R_{XX}(\tau) = \frac{A^2}{2} \cos w_0 \tau$$

Example 2. Random amplitude processes $X(t)=Y\cos(wt)+Z\sin(wt)$, t>0, whereas Y and Z are independent random variables, and EY=EZ=0, $VarY=VarZ=\sigma^2$.

$$E[x(t)] = 0$$

$$C_{XX}(\tau) = R_{XX}(\tau) = \sigma^2 \cos w \tau$$

vi) If X(t) is a non-period process, as $|\tau| \to \infty$ $X(t) \text{ and } X(t+\tau) \text{ are independent, that is}$ $\lim_{|\tau \to \infty|} R_{XX}(\tau) = m^2$



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2.3.1 Properties of $R_{XX}(t_1,t_2)$

Example 3. Pulse Code Modulation (binary case)

- A source generates the symbols 1 or 0 independently with probabilities 1-p and p, respectively.
- The symbol 1 is transmitted by sending a pulse with constant amplitude A and duration T.
- The symbol 0 is transmitted by sending nothing during an interval of length 7.

Example 3.

A signal modulated in this way is represented by the stochastic process

$$X(t) = \sum_{n=-\infty}^{\infty} A_n h(t - nT), \ nT \le t < (n+1)T$$

Where the A_n ; $n = 0,\pm 1,\pm 2,...$ are independent binary random variables defined by

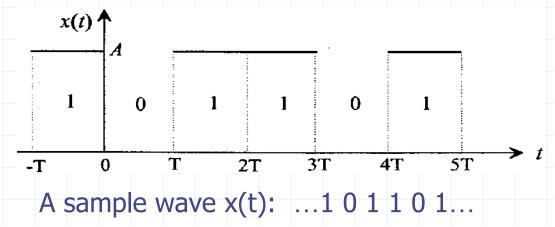
$$A_n = \begin{cases} 0 & \text{with probability } p \\ A & \text{with probability } 1-p \end{cases}$$

And h(t) is given by

$$h(t) = \begin{cases} 1 & \text{for } 0 \le t < T \\ 0 & \text{otherwise} \end{cases}$$

Example 3.

In particular, for any
$$t$$
, $X(t) = \begin{cases} 0 & \text{with probability } p \\ A & \text{with probability } 1-p \end{cases}$



Commonly, the time point t=0 has been chosen in such a way that is coincides with the beginning of a new transmission period.

Obtain: E[X(t)], $R_{XX}(t_1,t_2)$

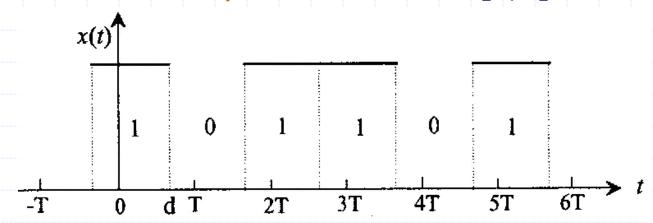
Discuss: Is X(t) a stationary process?

Example 4. Random Delayed Pulse Code

Using the stochastic process X(t) defined in example 3, define the process:

$$Y(t) = X(t-D)$$

where D is uniformly distributed over [0,T].



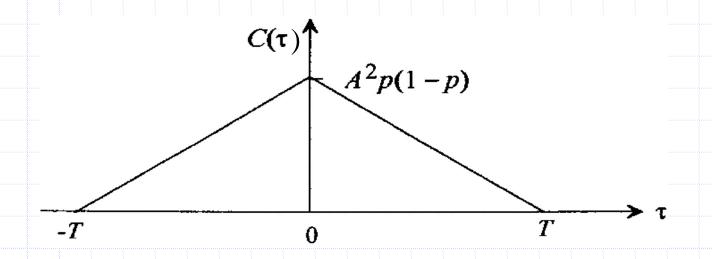
Obtain: E[Y(t)], $R_{YY}(t_1,t_2)$

Discuss: Is Y(t) a stationary process?

Example 4. Random Delayed Pulse Code

Discuss: Is Y(t) a stationary process?

$$C_{YY}(\tau) = \begin{cases} A^{2}p(1-p)(1-\frac{|\tau|}{T}) & \text{for } |\tau| \leq T \\ 0 & \text{otherwise} \end{cases}$$



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2.3.1 Properties of $R_{XX}(t_1,t_2)$

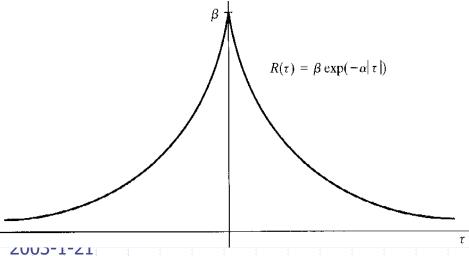
Examples of correlation functions for weakly stationary processes

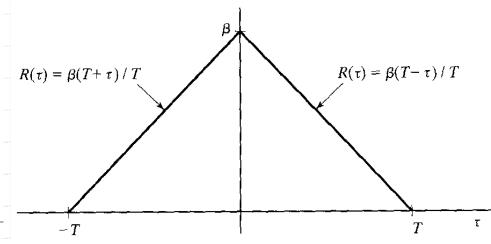
1.
$$R(\tau) = \beta \exp(-\alpha |\tau|)$$

2.
$$R(\tau) = \begin{cases} \beta(T - |\tau|)/T, & -T \le \tau \le T \\ 0, & \text{otherwise} \end{cases}$$

3.
$$R(\tau) = \beta \exp(-\alpha |\tau|) \cos(\omega_0 \tau)$$

4.
$$R(\tau) = 2W\{\sin(2\pi W\tau)/2\pi W\tau\}$$





2.3 Properties of Correlation Function

OUTLINE:

- 2.3.1 Properties of $R_{XX}(t_1,t_2)$
- 2.3.2 Properties of $R_{XY}(t_1,t_2)$

1.Definition of cross-correlation function for two stochastic processes $\{X(t), t \in T\}$ and $\{Y(t), t \in T\}$

Review

6. Cross-covariance functions

For two second-order moment processes
 X(t) and Y(t), their cross-covariance
 function is

$$C_{XY}(t_1, t_2) = Cov[X(t_1), Y(t_2)]$$

$$= E[[X(t_1) - \overline{x}(t_1)][Y(t_2) - \overline{y}(t_2)]]$$

$$= E[X(t_1)Y(t_2)] - \overline{x}(t_1)\overline{y}(t_2)$$

Review

7. Cross-correlation functions

 For two second-order moment processes X(t) and Y(t), their cross-correlation function is

$$R_{XY}(t_1, t_2) = E[X(t_1)Y(t_2)]$$

$$C_{XY}(t_1, t_2) = R_{XY}(t_1, t_2) - \overline{x}(t_1)\overline{y}(t_2)$$

$$R_{XY}(t_1, t_2) = E[Y(t_2)X(t_1)] = R_{YX}(t_2, t_1)$$

$$|R_{X,Y}(t,s)| \leq \sqrt{R_X(t,t)R_Y(s,s)}$$

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Review

Mutually uncorrelated

For two second-order moment processes X(t) and Y(t), if

$$C_{XY}[t_1, t_2] = 0$$
 $t_1, t_2 \in T$

then X(t) and Y(t) are mutually uncorrelated

$$R_{XY}(t_1, t_2) = \overline{x}(t_1)\overline{y}(t_2)$$

Independent random processes:

The processes $\{X(t), t \in T\}$ and $\{Y(t), t \in T\}$ are said to be independent random processes if, for each positive integer n and each choice of $(t_1, t_2, ..., t_n)$, $t_i \in T$, $(s_1, s_2, ..., s_n)$, $s_i \in T$, the two random vectors $\mathbf{X} = (X(t_1), X(t_2), ..., X(t_n))$ and $\mathbf{Y} = (Y(s_1), Y(s_2), ..., Y(s_n))$ are independent.

Two random vectors \mathbf{X} and \mathbf{Y} are independent if and only if $F_{\mathbf{X},\mathbf{Y}}(\mathbf{X},\mathbf{y})=F_{\mathbf{X}}(\mathbf{X})F_{\mathbf{Y}}(\mathbf{y})$.

The independent random processes are also uncorrelated. Uncorrelated random processes need not be independent.

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jointly Gaussian random processes

The processes $\{X(t), t \in T\}$ and $\{Y(t), t \in T\}$ are said to be jointly Gaussian random processes if, for each positive integer n and each choice of $(t_1, t_2, ..., t_n)$, $t_i \in T$, $(s_1, s_2, ..., s_n)$, $s_i \in T$, the two random vectors $\mathbf{X} = (X(t_1), X(t_2), ..., X(t_n))$ and $\mathbf{Y} = (Y(s_1), Y(s_2), ..., Y(s_n))$ are jointly Gaussian .

The vector (X, Y) has 2n-dimensional Gaussian density.

Jointly Gaussian random processes that are uncorrelated are also independent random processes.

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- 2. Jointly Stationary Processes
- For two stationary processes X(t) and Y(t), if

$$R_{XY}(t_1, t_2) = E[X(t_1)Y(t_2)] = R_{XY}(\tau)$$

$$R_{YX}(t_1, t_2) = E[Y(t_1)X(t_2)] = R_{YX}(\tau)$$

then X(t) and Y(t) are called Jointly Stationary Processes.

$$R_{XY}(\tau) = R_{YX}(-\tau)$$

Review

e.g.4.

Given: X(t) and Y(t) are two second-order moment processes. W(t)=X(t)+Y(t).

Obtain: E[W(t)] and $R_{WW}(t_1,t_2)$

$$E[W(t)] = E[X(t)] + E[Y(t)]$$

$$R_{WW}(t_1, t_2) = R_{XX}(t_1, t_2) + R_{YY}(t_1, t_2) + R_{XY}(t_1, t_2) + R_{YX}(t_1, t_2)$$

If X(t) and Y(t) are Jointly Stationary Processes, then,

$$E[W(t)] = m_X + m_Y = m_W$$

$$R_{WW}(t_1, t_2) = R_{XX}(\tau) + R_{YY}(\tau) + R_{XY}(\tau) + R_{YX}(\tau) = R_{WW}(\tau)$$

e.g.4.

If $m_X = m_Y = 0$, and X(t) and Y(t) are mutually uncorrelated, then

$$m_W = 0$$

$$R_{WW}(\tau) = R_{XX}(\tau) + R_{YY}(\tau)$$

The sum of any two correlation functions is a valid correlation function.

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e.g. The product of any two correlation functions is a valid correlation function.

Given: X(t) and Y(t) are two independent second-order moment processes. W(t)=X(t)Y(t).

Obtain: E[W(t)] and $R_{WW}(t_1,t_2)$

$$E[W(t)] = E[X(t)Y(t)] = E[X(t)]E[Y(t)]$$

$$R_{WW}(t_1, t_2) = E[X(t_1)Y(t_1) \bullet X(t_2)Y(t_2)]$$

$$= E[X(t_1)X(t_2)] \bullet E[Y(t_1)Y(t_2)]$$

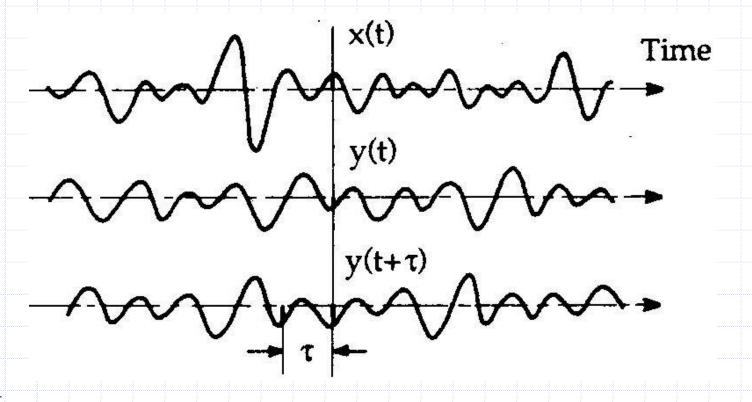
$$= R_{XX}(t_1, t_2)R_{YY}(t_1, t_2)$$

If X(t) and Y(t) are each Stationary Processes, then,

$$E[W(t)] = m_X m_Y = m_W \qquad R_{WW}(t_1, t_2) = R_{WW}(\tau) = R_{XX}(\tau) R_{YY}(\tau)$$

If X(t) and Y(t) are Jointly Stationary Processes and X(t),Y(t) are ergodic, then

$$R_{XY}(\tau) = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} x(t)y(t+\tau)dt$$



4. Properties of $R_{XY}(\tau)$

i)
$$R_{YX}(au)=R_{XY}(- au)$$

$$|ii)|R_{XY}(\tau)| \leq \sqrt{R_X(0)R_Y(0)} \leq \frac{1}{2}[R_{XX}(0) + R_{YY}(0)]$$

$$|R_{YX}(\tau)| \le \sqrt{R_X(0)R_Y(0)} \le \frac{1}{2}[R_{XX}(0) + R_{YY}(0)]$$

iii) $R_{XY}(\tau)$ is not always maximum at $\tau = 0$

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Homework 2.8 2.16 2.17