Stationarity

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March 23, 2021

First-Order Stationary Processes Correlation and Covariance Functions

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Based on Probability, Random Variables and Random Signal Principles, P.Z. Peebles, Jr. and B. Shi

Outline

First-Order Stationary Processes

2 Correlation and Covariance Functions

First Order Stationary

$f_X(x;t)$

if X(t) is to be a first-order stationary

$$f_X(x_1;t_1) = f_X(x_1;t_1 + \Delta)$$

must be true for any time t_1 and any real number \triangle

the first order density function does not change with a shift in time origin

Consequences of stationarity

$f_X(x;t)$

- f_X(x,t₁) is independent of t₁
 the first order density function
 does not change with a shift in time origin
- the process mean value is a constant

$$m_X(t) = \overline{X} = constant$$

the process mean value

$$m_X(t) = \overline{X} = constant$$

$$m_X(t_1) = \int_{-\infty}^{\infty} x f_X(x; t_1) dx$$

$$m_X(t_2) = \int_{-\infty}^{\infty} x f_X(x; t_2) dx$$

let
$$t_2 = t_1 + \Delta$$

$$m_X(t_1) = m_X(t_1 + \Delta)$$

Second-Order Stationary Process

$f_X(x_1,x_2;t_1,t_2)$

if X(t) is to be a second-order stationary

$$f_X(x_1,x_2;t_1,t_2) = f_X(x_1,x_2;t_1+\Delta,t_2+\Delta)$$

must be true for any time t_1 , t_2 and any real number \triangle

the second order density function does not change with a shift in time origin

Second-Order Stationary Process

$f_X(x_1,x_2;t_1,t_2)$

- f_X(x₁,x₂; t₁,t₂) is independent of t₁ and t₂ the second order density function does not change with a shift in time origin
- the autocorrelation function

$$R_{XX}(t, t+\tau) = E[X(t)X(t+\tau)] = R_{XX}(\tau)$$

Nth-order Stationary Processes

$f_X(x_1,\cdots,x_N;t_1,\cdots,t_N)$

if X(t) is to be a N^{th} -order stationary

$$f_X(x_1,\dots,x_N;t_1,\dots,t_N) = f_X(x_1,\dots,x_N;t_1+\Delta,\dots,t_N+\Delta)$$

must be true for any time $t_1,...,t_N$ and any real number Δ

the N^{th} order density function does not change with a shift in time origin

Stationary Process

joint probability distribution

a stationary process is a stochastic process whose <u>unconditional</u> joint probability distribution does not change when shifted in time.

Consequently, parameters such as **mean** and **variance** also do not change over time.

Stationary Process - nomenclature

nomenclature

- stationary process
- strictly stationary process
- strongly stationary process
- strict sense stationary (SSS) process

Strict Sense Stationary Process

for all natural number

if X(t) is to be a strict sense stationary (SSS) process

$$f_X(x_1,\dots,x_N;t_1,\dots,t_N) = f_X(x_1,\dots,x_N;t_1+\Delta,\dots,t_N+\Delta)$$

must be true for any time $t_1,...,t_N$ and any real number Δ and for all natural number N

 white noise is the simplest example of a strictly stationary process.

Wide Sense Stationary Process

1st and 2nd moments

Wide Sense Stationary (WSS) random processes only require that

1st moment (i.e. the mean) and autocovariance
do not vary with respect to time and that
the 2nd moment is finite for all times.

- $E[X(t_1)] = E[X(t_2)] = \overline{X} = constant$ for all t_1 and t_2
- $C_{XX}(t_1,t_2) = C_{XX}(t_1-t_2,0) \triangleq C_{XX}(\tau)$ for all t_1 and t_2
- $E[|X(t)|^2] < \infty$ for all t

Wide Sense Stationary Process - nomenclature

nomenclature

- weak sense stationary (WSS) process
- wide sense stationary (WSS) process

WSS - auto-covariance & auto-correlation

mean, auto-covariance, auto-correlation

$$m_{X}(t) = \overline{X} = constant$$

$$C_{XX}(t_{1}, t_{2}) = E\left[\left\{X(t_{1}) - m_{X}(t_{1})\right\}\left\{X(t_{2}) - m_{X}(t_{2})\right\}\right]$$

$$= E\left[\left\{X(t_{1}) - \overline{X}\right\}\left\{X(t_{2}) - \overline{X}\right\}\right]$$

$$= E\left[X(t_{1})X(t_{2})\right] - \overline{X}^{2}$$

$$\triangleq C_{XX}(\tau)$$

$$\triangleq R_{XX}(\tau) - \overline{X}^{2}$$

$$R_{XX}(t_{1}, t_{2}) \triangleq R_{XX}(\tau)$$

Wide Sense Stationary Process

$m_X(t), R_{XX}(\tau)$

WSS random processes only require that 1st moment (i.e. the mean) and autocorrelation do not vary with respect to time

$$E[X(t)] = m_X(t) = \overline{X} = constant$$

$$E[X(t)X(t+\tau)] = R_{XX}(\tau)$$

Wide Sense Stationary Process

$m_X(t), R_{XX}(\tau)$

- the 2nd order stationarity is sufficient for wide sense stationarity
- if $f_X(x_1; t_1)$ is independent of t_1 then E[X(t)] = constant
- if $f_X(x_1, x_2; t_1, t_2)$ is independent of t_1 and t_2 then $E[X(t)X(t+\tau)] = R_{XX}(\tau)$

The properties of autocorrelation functions (1)

$$|R_{XX}(\tau)|, R_{XX}(-\tau), R_{XX}(0)$$

$$|R_{XX}(\tau)| \le R_{XX}(0)$$

$$R_{XX}(-\tau) = R_{XX}(\tau)$$

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

$$P[|X(t+\tau) - X(t)| > \varepsilon] = \frac{2}{\varepsilon^{2}} (R_{XX}(0) - R_{XX}(\tau))$$

The properties of autocorrelation functions (2)

$R_{NN}(\tau), R_{XX}(\tau)$

if
$$X(t) = \overline{X} + N(t)$$

where N(t) is WSS, is **zero-mean**, and

has autocorrelation function $R_{NN}(\tau) \to 0$ as $|\tau| \to \infty$, then

$$\lim_{|\tau|\to\infty}R_{XX}(\tau)=\overline{X}^2$$

The properties of autocorrelation functions (3)

$R_{NN}(\tau), R_{XX}(\tau)$

if X(t) is mean square periodic, i.e, there exists a $T \neq 0$ such that $E\left[\{X(t+T)-X(t)\}^2\right]=0$ for all t, then $R_{XX}(t)$ will have a **periodic** component with the same period

The properties of autocorrelation functions (4)

$R_{NN}(\tau), R_{XX}(\tau)$

 $R_{XX}(\tau)$ cannot have an arbitrary shape any arbitrary function cannot be an autocorrelation function. power density spectrum

 $R_XX(\tau)$ is related to the power density spectrum through the Fourier transform and the form of the spectrum is not arbitrary

Crosscorrelation functions (1)

$$R_{XY}(t_1,t_2), R_{XY}(t,t+\tau)$$

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

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

if

$$R_{XY}(t,t+\tau)=0$$

then X(t) and Y(t) are called **orthogonal processes**

Crosscorrelation functions (2)

$R_{XY}(t, t+\tau), R_{XY}(\tau)$

if X(t) and Y(t) are statistically independent

$$R_{XY}(t,t+\tau) = E[X(t)Y(t+\tau)] = m_X(t)m_Y(t+\tau)$$

if X(t) and Y(t) are stistically independent and are at least WSS.

$$R_{XY}(\tau) = \overline{XY}$$

which is constant

The properties of crosscorrelation functions (1)

$$R_{XY}(\tau), |R_{XY}(\tau)|$$

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

$$|R_{XY}(\tau)| = \sqrt{R_{XX}(0)R_{YY}(0)}$$

$$|R_{XY}(\tau)| \leq \frac{1}{2} \left[R_{XX}(0) + R_{YY}(0) \right]$$

The properties of crosscorrelation functions (2)

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

$$R_{YX}(-\tau) = E[Y(t)X(t-\tau)] = E[Y(s+\tau)X(s)] = R_{XY}(\tau)$$

$$E\left[\left\{Y(t+\tau)+\alpha X(t)\right\}^2\right]\geq 0$$

the **geometric mean** of two positive numbers cannot exceed their **arithmetic mean**

The properties of crosscorrelation functions (3)

$|R_{XY}(\tau)|$

$$|R_{XY}(\tau)| \leq \frac{1}{2} \left[R_{XX}(0) + R_{YY}(0) \right]$$

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

Covariance Functions

$$C_{XX}(t,t+\tau), C_{XY}(t,t+\tau)$$

$$C_{XX}(t,t+\tau) = E\left[\left\{X(t) - m_X(t)\right\}\left\{X(t+\tau) - m_X(t+\tau)\right\}\right]$$

$$C_{XY}(t,t+\tau) = E\left[\left\{X(t) - m_X(t)\right\}\left\{Y(t+\tau) - m_Y(t+\tau)\right\}\right]$$

$$C_{XX}(t,t+\tau) = R_{XX}(t,t+\tau) - m_X(t)m_X(t+\tau)$$

$$C_{XY}(t,t+\tau) = R_{XY}(t,t+\tau) - m_X(t)m_Y(t+\tau)$$
at least jointly WSS

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

$$C_{XY}(\tau) = R_{XY}(\tau) - \overline{XY}$$

The properties of covariance functions

$C_{XX}(0)$

For a WSS process, variance does not depend on time and if $\tau = 0$

$$C_{XX}(0) = R_{XX}(0) - \overline{X}^2$$

$$\sigma_X^2 = E\left[\left\{X(t) - E\left[X(t)\right]\right\}^2\right] = C_{XX}(0)$$

it the two random processes uncorrelated

$$C_{XY}(t,t+\tau) = R_{XY}(t,t+\tau) - m_X(t)m_Y(t+\tau) = 0$$

$$R_{XY}(t, t+\tau) = m_X(t)m_Y(t+\tau)$$

Discrete-Time Processes and Sequences (1)

$$R_{XX}[n, n+k], R_{YY}[n, n+k], C_{XX}[n, n+k], C_{YY}[n, n+k]$$

$$m_X[n] = \overline{X}, m_Y[n] = \overline{Y}$$

$$R_{XX}[n, n+k] = R_{XX}[k]$$

$$R_{YY}[n, n+k] = R_{YY}[k]$$

$$C_{XX}[n, n+k] = R_{XX}[k] - \overline{X}^2$$

$$C_{YY}[n, n+k] = R_{YY}[k] - \overline{Y}^2$$

Discrete-Time Processes and Sequences (2)

$$R_{XY}[n, n+k], R_{YX}[n, n+k], C_{XY}[n, n+k], C_{YX}[n, n+k]$$

$$m_X[n] = \overline{X}, m_Y[n] = \overline{Y}$$

$$R_{XY}[n, n+k] = R_{XY}[k]$$

$$R_{YX}[n, n+k] = R_{YX}[k]$$

$$C_{XY}[n, n+k] = R_{XY}[k] - \overline{XY}$$

$$C_{YX}[n, n+k] = R_{YX}[k] - \overline{YX}$$