

## (12.5) Correlation Structures of Non-stationary Data

Consider a pair of non-stationary processes  $\{x(t)\}$  and  $\{y(t)\}$ . The mean values at any fixed time  $t$  are

$$\mu_x(t) = E[x(t)] \quad \text{and} \quad \mu_y(t) = E[y(t)]$$

The correlation functions at any pair of fixed times  $t_1$  and  $t_2$  are defined by their expected values

$$R_{xx}(t_1, t_2) = E[x(t_1)x(t_2)]$$

$$R_{yy}(t_1, t_2) =$$

$$R_{xy}(t_1, t_2) = E[x(t_1)y(t_2)]$$

} for stationary data only  $(t_2 - t_1) \neq \tau$  mattered.

It follows that

$$R_{xx}(t_1, t_2) = R_{xx}(t_2, t_1)$$

$$R_{yy}(t_1, t_2) = R_{yy}(t_2, t_1)$$

$$R_{xy}(t_1, t_2) = R_{xy}(t_2, t_1)$$

and

$$|R_{xy}(t_1, t_2)|^2 \leq R_{xx}(t_1, t_2) \cdot R_{yy}(t_1, t_2)$$



177

As before we estimate these functions by using  $N$  samples of  $x_i(t)$  and compute an ensemble average

$$\hat{R}_{xx}(t_1, t_2) = \frac{1}{N} \sum_{i=1}^N x_i(t_1) \cdot x_i(t_2)$$

or rewrite with

$$t_1 = t \quad \text{and} \quad t_2 = t - \tau \quad (\text{or } \tau = t - t_2 = t_1 - t_2)$$

$$\hat{R}_{xx}(t, t - \tau) = \frac{1}{N} \sum_{i=1}^N x_i(t) x_i(t - \tau)$$

or rewrite with

$$t_1 = t - \tau/2 \quad t_2 = t + \tau/2$$

then

$$\tau = t_2 - t_1$$

$$t = (t_1 + t_2) / 2$$

time difference  
lag

center time between  $t_1$  and  $t_2$

And

$$\begin{aligned} R_{xy}(t_1, t_2) &= R_{xy}(t - \tau/2, t + \tau/2) \\ &= E[x(t - \tau/2) \cdot y(t + \tau/2)] \\ &= R_{xy}(\tau, t) \end{aligned}$$



For a zero lag  $\tau = 0$

$$R_{xx}(\tau=0, t) = E[x^2(t)] = \sigma_x^2 \quad \text{Variance}$$

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

$$R_{xy}(\tau=0, t) = E[x(t)y(t)] = \sigma_{xy} \quad \text{Co-Variance}$$

And symmetries are

$$R_{xx}(\tau, t) = R_{xx}(-\tau, t)$$

$$R_{yy}(\tau, t) =$$

$$R_{xy}(\tau, t) = R_{yx}(-\tau, t)$$

If we form an average over time, we get

$$\overline{R}_{xy}(\tau) = \lim_{T \rightarrow \infty} \int_0^T R_{xy}(\tau, t) dt$$

which is a real-valued, even function of  $\tau$  representing the usual cross-correlation of stationary random data



# (12.6) Spectral Structure of Non-Stationary Data

Assume  $x(t)$  and  $y(t)$  real-valued data from non-stationary processes  $\{x(t)\}$  and  $\{y(t)\}$  with finite Fourier Transforms

$$X(f, T) = \int_0^T x(t) e^{-j 2\pi f t} dt$$

$$Y(f, T) =$$

Dropping the record-length  $T$  notation, we define spectral density functions as

$$S_{xx}(f_1, f_2) = E[X^*(f_1) \cdot X(f_2)]$$
$$S_{yy}(f_1, f_2) =$$
$$S_{xy}(f_1, f_2) = E[X(f_1) \cdot Y(f_2)]$$

These are called "double frequency (generalized) spectral density functions." Complex for both  $f_1$  and  $f_2$  in the range  $(-\infty, +\infty)$ .

As for the time-domain correlations, we have

$$|S_{xy}(f_1, f_2)|^2 \leq S_{xx}(f_1, f_2) \cdot S_{yy}(f_1, f_2)$$



and  $S_{xx} = S_{xx}^*$ ,  $S_{yy} = S_{yy}^*$  are real + even  
 $S_{xy} = S_{yx}^*$  is complex for all  $(f_1, f_2)$

Now  $X^*(f_1) \cdot Y(f_2) = \int x(t_1) e^{-j2\pi f_1 t_1} dt_1 \cdot \int y(t_2) e^{+j2\pi f_2 t_2} dt_2$

$\Downarrow$   
 $S_{xy}(f_1, f_2) = E[X^* Y] = \iint R_{xy}(t_1, t_2) e^{+j2\pi(f_1 t_1 - f_2 t_2)} dt_1 dt_2$

$\neq \iint R_{xy}(t_1, t_2) e^{-j2\pi(f_1 t_1 + f_2 t_2)} dt_1 dt_2$

Non-stationary cross-spectra NOT EQUAL double Fourier Transform of non-stationary cross-covariance

Instead, the non-stationary cross-spectra  $S_{xy}^*$  is the inverse Fourier Transform of  $R_{xy}$  followed by the Fourier Transform over  $t_2$ .

Also

$R_{xy}(t_1, t_2) = \iint S_{xy}(f_1, f_2) e^{-j2\pi(f_1 t_1 - f_2 t_2)} df_1 df_2$

This is the direct Fourier Transform of  $S_{xy}(f_1, f_2)$  over  $f_1$  followed by the inverse Fourier Transform over  $f_2$ .



Instead of  $(f_1, f_2)$  consider  $(f, g)$  independent frequencies

$$f_1 = f - g/2$$

$$f_2 = f + g/2$$

or

$$f = (f_1 + f_2)/2$$

$$g = f_2 - f_1$$

average frequency

difference frequency

Then

$$S_{xy}(f_1, f_2) = E[X^*(f - g/2) \cdot \bar{Y}(f + g/2)] \equiv S_{xy}(f, g)$$

For  $g = 0$  (that is,  $f_1 = f_2$ )

$$S_{xx}(f, 0) = E[|X(f)|^2]$$

$$S_{yy}(f, 0) =$$

$$S_{xy}(f, 0) = E[X^*(f) \cdot \bar{Y}(f)]$$

standard auto-spectral  
and

cross-spectral density

for stationary processes

Further



where for  $g = \sigma$  ( $\Rightarrow f_1 = f_2$ )

$$S_{xx}(f, \sigma) = E[|X(f)|^2]$$

$$S_{yy}(f, \sigma) =$$

$$S_{xy}(f, \sigma) = E[X^*(f)Y(f)]$$

standard  
auto- and  
cross-spectral  
density functions  
for stationary processes

Further development shows that

$$\int R_{xy}(\tau, t) e^{-j2\pi f\tau} d\tau = \int S_{xy}(f, g) e^{+j2\pi g t} dg$$

Fourier Transform of  $R_{xy}(\tau, t)$

over  $\tau$  holding  $(f, t)$  fixed

Inverse Fourier Transform of

$S_{xy}$  over  $g$  holding

$(f, t)$  fixed

recall

$$t = (t_1 + t_2) / 2$$

$$\tau = t_2 - t_1$$

$$f = (f_1 + f_2) / 2$$

$$g = f_2 - f_1$$

Either of these operations defines a time-frequency function

$$W_{xy}(f, t) = \int R_{xy}(\tau, t) e^{-j2\pi(f\tau + \cancel{g t})} d\tau$$

Do NOT confuse  
~~with~~

$$\text{with } S_{xy}(f, g) = \iint R_{xy}(\tau, t) e^{-j2\pi(f\tau + g t)} d\tau dt$$

that is the double Fourier Transform of  $R_{xy}(\tau, t)$



For stationarity data, we have

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

and

$$R_{xy}(t, \tau) = R_{xy}(\tau)$$

Consider the non-stationary

$$(p. 180) \quad R_{xy}(t, t) = \iint S_{xy}(f_1, f_2) \underbrace{e^{-j2\pi t(f_1 - f_2)}}_{t=0} df_1 df_2$$

This now becomes

$$R_{xy}(t, t) = R_{xy}(0) = \iint S_{xy}(f_1, f_2) \cdot 1 df_1 df_2$$

↑  
stationarity

$$= \int S_{xy}(f_1) df_1$$

where

$$S_{xy}(f_1) = \int S_{xy}(f_1, f_2) df_2$$

Hence for stationary data we have

$$S_{xy}(f_1, f_2) = S_{xy}(f_1) \cdot \delta_1(f_2 - f_1)$$

where

$$\delta_1(f) = \begin{cases} T & -1/2T \leq f < 1/2T \\ 0 & \text{otherwise} \end{cases}$$



## Frequency - Time Spectra

$$R_{xy}(\tau, t) = E \left[ x(t - \tau/2) y(t + \tau/2) \right]$$

non-stationary  
cross-covariance

and

$$S_{xy}(f, g) = E \left[ X^*(f - g/2) \cdot Y(f + g/2) \right]$$

non-stationary  
cross-spectra

The Fourier Transform of  $R_{xy}(\tau, t)$  with respect to  $\tau$  holding  $t$  fixed gives

$$W_{xy}(f, t) = \int R_{xy}(\tau, t) e^{-j2\pi f\tau} d\tau$$

Defines the frequency-time spectral density which is also called the "instantaneous spectrum"

Note that

$$W_{xy}(f, t) = \int S_{xy}(f, g) e^{+j2\pi g t} dg$$

So the frequency-time spectra is also the inverse Fourier Transform of  $S_{xy}(f, g)$  with respect to  $g$  while holding  $f$  constant.



The reverse holds, too

$$R_{xy}(\tau, t) = \int W_{xy}(f, t) e^{j2\pi f\tau} df$$

and/or

$$S_{xy}(f, g) = \int W_{xy}(f, t) e^{-j2\pi gt} dt$$

↕

$$S_{xy}(f, g) = \iint R_{xy}(\tau, t) e^{-j2\pi(f\tau + gt)} d\tau dt$$

$$R_{xy}(\tau, t) = \iint S_{xy}(f, g) e^{+j2\pi(f\tau + gt)} df dg$$

are Double Fourier Transform Pairs.

Special cases occur for non-stationary auto-correlations

$R_{xx}(\tau, t)$  for which

$$W_{xx}(f, t) = \int R_{xx}(\tau, t) e^{-j2\pi f\tau} d\tau = \int R_{xx}(\tau, t) \cos(2\pi f\tau) dt$$

This is a real and even function  $W_{xx}(f, t) = W_{xx}^*(f, t)$

$$W_{xx}(f, t) = W_{xx}(-f, t)$$

And for  $\tau = 0$

$$R_{xx}(\tau=0, t) = \int W_{xx}(f, t) df = E[x^2(t)] = \sigma_x^2(t)$$