

Lecture 23

- ACF of a Sum Process

$$Z(t) = X(t) + Y(t)$$

Question: $R_Z(\tau)$?

Answer:

$$\begin{aligned} R_Z(\tau) &= E[Z(t)Z(t+\tau)] \\ &= E\{(X(t) + Y(t)) \cdot [X(t+\tau) + Y(t+\tau)]\} \\ &= E[X(t)X(t+\tau)] + E[X(t)Y(t+\tau)] + E[Y(t)X(t+\tau)] + E[Y(t) \cdot Y(t+\tau)] \\ &= R_X(\tau) + R_{XY}(\tau) + R_{YX}(\tau) + R_Y(\tau) \end{aligned}$$

where

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

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

- Cross-Correlation Functions

In general, $R_{XY}(\tau) \neq R_{YX}(\tau)$

Exceptions:

(1) If $\tau=0$, $R_{XY}(0) = E[X(t) \cdot Y(t+0)] = E[Y(t) \cdot X(t)] = R_{YX}(0)$

(2) If $X(t)$ and $Y(t)$ are mutually independent

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

$$R_{YX}(\tau) = E[Y(t) \cdot X(t+\tau)] = E[Y(t)] \cdot E[X(t+\tau)] = \bar{Y} \cdot \bar{X}$$

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

And if $X(t)$ and $Y(t)$ are independent,

$$Z(t) = X(t) + Y(t)$$

$$\begin{aligned} R_Z(\tau) &= R_X(\tau) + \underbrace{R_{XY}(\tau)} + \underbrace{R_{YX}(\tau)} + R_Y(\tau) \\ &= R_X(\tau) + \bar{X} \cdot \bar{Y} + \bar{Y} \cdot \bar{X} + R_Y(\tau) \end{aligned}$$

Furthermore, if at least one of \bar{X} and \bar{Y} is zero (zero mean)

$$R_Z(\tau) = R_X(\tau) + R_Y(\tau)$$

- Cross Correlation Functions between a Sum and Difference Processes

$$\begin{cases} U(t) = X(t) + Y(t) \\ V(t) = X(t) - Y(t) \end{cases}$$

Question: $R_{UV}(\tau)$, and $R_{VU}(\tau)$?

$$\begin{aligned}
R_{uv}(\tau) &= E [u(t) \cdot v(t+\tau)] \\
&= E \left\{ [X(t) + Y(t)] \cdot [X(t+\tau) - Y(t+\tau)] \right\} \\
&= E [X(t)X(t+\tau)] - E [X(t)Y(t+\tau)] + E [Y(t)X(t+\tau)] - E [Y(t)Y(t+\tau)] \\
&= \underbrace{R_X(\tau)} - R_{XY}(\tau) + R_{YX}(\tau) - R_Y(\tau) \quad \dots \dots (1)
\end{aligned}$$

$$\begin{aligned}
R_{vu}(\tau) &= E [v(t) \cdot u(t+\tau)] \\
&= E \left\{ [X(t) - Y(t)] \cdot [X(t+\tau) + Y(t+\tau)] \right\} \\
&= E [X(t)X(t+\tau)] + E [X(t)Y(t+\tau)] - E [Y(t)X(t+\tau)] - E [Y(t)Y(t+\tau)] \\
&= R_X(\tau) + R_{XY}(\tau) - R_{YX}(\tau) - R_Y(\tau) \quad \dots \dots (2)
\end{aligned}$$

If $X(t)$ and $Y(t)$ are independent, and both $X(t)$ and $Y(t)$ have zero mean, then

$$R_{uv}(\tau) = R_{vu}(\tau) = R_X(\tau) - R_Y(\tau).$$

Previously, estimation of autocorrelation function.

```
%Example of autocorrelation calculation
rand('seed', 1000); % use seed to make repeatable
x=10*randn(1, 1001); % generate random samples
t = 0: 0.001 : 1; % sampling index
[t1, R] = corb(x, x, 1000) ; % autocorrelation
subplot(2, 1, 1);
plot(t,x);
xlabel('TIME'); ylabel('X');
subplot(2, 1, 2);
plot(t1,R);
xlabel('LAG'); ylabel('RX');
function [ndt, R] = corb(a, b, f)
% Correlation functioning using convolution
N = length(a);
R = conv(a, fliplr(b))/(N + 1);
ndt = -(N - 1): (N - 1)/f;
end
```

```
function [ndt, R] = corb(a, b, f)
% Correlation function using
convolution
N = length(a);
R = conv(a, fliplr(b))/(N + 1);
ndt = -(N - 1): (N - 1)/f;
end
```

fliplr

Flip array left to right

```
>> x = 1: 5
```

```
x =
```

```
    1    2    3    4    5
```

```
>> fliplr(x)
```

```
ans =
```

```
    5    4    3    2    1
```

In general,

Correlation:
$$z[n] = \sum_{k=-\infty}^{\infty} x[k] \cdot h[n+k]$$

Convolution:
(sum)
$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k] \cdot h[n-k]$$

Use Convolution to implement Correlation by first flipping $h[n]$.

http://www.ece.uah.edu/~dwpan/course/ee385/code/ccf_example.m

% Example of CCF calculation to **identify signal from noise**

```
clear all;
noise = 3*randn(1, 1001); % noise
theta = random('Uniform', 0, 2*pi);
t = 0: 0.001 : 1; % sampling index
x = cos(100*pi*t + theta); % random signal
y = x + noise;
[t1, Ry] = corb(y, y, 1000); % autocorrelation
xd = cos(100*pi*t); % deterministic signal for detection
[t2, Rxy] = corb(xd, y, 1000); % cross-correlation
subplot(3, 1, 1);
plot(t,y);
xlabel('TIME'); ylabel('Y');
subplot(3, 1, 2);
plot(t1,Ry);
xlabel('LAG'); ylabel('R_Y');
subplot(3, 1, 3);
plot(t2,Rxy);
xlabel('LAG'); ylabel('R_{XY}');
```

