

TTT4120 Digital Signal Processing Fall 2021

Lecture: Correlation and Energy Spectral Density

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Lecture in course book*

- Proakis, Manolakis Digital Signal Processing, 4th Ed.
 - 2.6.1 Crosscorrelation and autocorrelation sequences
 - 2.6.2 Properties of crosscorrelation and autocorrelation...
 - 2.6.4 Input-output correlation sequences
 - 4.2.5 Energy density spectrum of aperiodic signals
 - 5.3.1 Input-output correlation functions and spectra

*Level of detail is defined by lectures and problem sets

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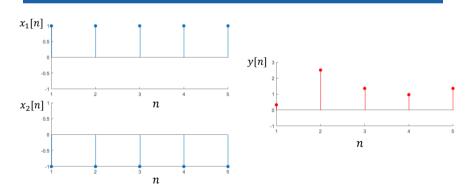
Contents and learning outcomes

- Cross- and autocorrelation sequences
- Properties of cross- and autocorrelation sequences
- Linear time-invariant systems
- Energy spectral density

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Introduction



- Which of the sequences $x_1[n]$ and $x_2[n]$ resembles y[n]?
- How to measure similarity between signal sequences

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Introduction...

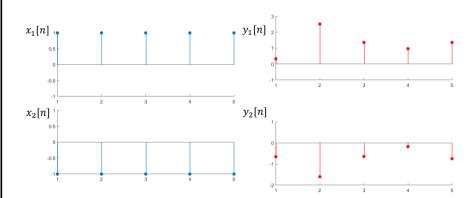
- Signals transmitted over some medium, e.g., wireless channels, experience delays, echoes, and noise
 - Difficult to recognize/or detect the signals at the receiving end
- Suppose that $x_1[n]$ or $x_2[n]$ is transmitted and y[n] is received
 - If y[n] is more similar to $x_1[n]$ than to $x_2[n]$, we decide that $x_1[n]$ was transmitted
 - If y[n] is more similar to $x_2[n]$ than to $x_1[n]$, we decide that $x_2[n]$ was transmitted
- Correlation is a measure of similarity

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Introduction...

• Digital communication example: $y_i[n] = x_i[n] + w[n]$

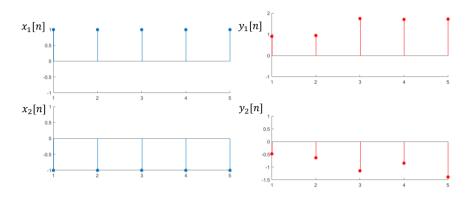


Noise can make received signal fluctuate significantly

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Introduction...

• Digital communication example: $y_i[n] = x_i[n] + w[n]$

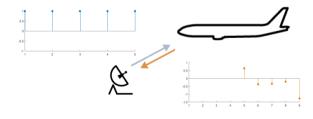


• Noise can make received signal fluctuate significantly

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Introduction...

• Radar example: $y[n] = \alpha x[n-D] + w[n]$, find D?



• Here we need a similarity measure that gives a maximum for *D*, considering all possible delays

Crosscorrelation and autocorrelation

• Crosscorrelation of real-valued sequences x[n] and y[n]

$$r_{xy}[l] = \sum_{n=-\infty}^{\infty} x[n]y[n-l]$$

= $\sum_{n=-\infty}^{\infty} x[n+l]y[n], l = \pm 1, \pm 2, ...$

- Measure of similarity between signals x[n] and y[n]
- Reverse role $r_{yx}[l] \neq r_{xy}[l]$

$$r_{yx}[l] = \sum_{n=-\infty}^{\infty} y[n]x[n-l] = r_{xy}[-l]$$

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Crosscorrelation and autocorrelation...

• Similarity to convolution of x[n] and y[n]

$$x[l] * y[l] = \sum_{k=-\infty}^{\infty} x[k]y[l-k]$$

$$r_{xy}[l] = \sum_{k=-\infty}^{\infty} x[k]y[k-l] = x[l] * y[-l]$$

- Relation can be exploited for efficient computation
- Autocorrelation sequence (self-similarity), y[n] = x[n]

$$r_{xx}[l] = \sum_{n=-\infty}^{\infty} x[n]x[n-l]$$
$$= \sum_{n=-\infty}^{\infty} x[n+l]x[n]$$

Crosscorrelation and autocorrelation...

• Finite causal sequences $x[n] = y[n] = 0, n < 0, n \ge N$

$$x[n] = \{\underline{x[0]}, x[1], x[2], \dots x[N-1]\}$$

$$y[n] = \{\underline{y[0]}, y[1], y[2], \dots y[N-1]\}$$

$$r_{xy}[l] = \sum_{n=-\infty}^{\infty} x[n]y[n-l]$$

• Try a couple of values of l and search for pattern \Rightarrow

$$r_{xy}[l] = \sum_{n=l}^{N-1} x[n]y[n-l]$$
 , $l \ge 0$

$$r_{xy}[l] = \sum_{n=0}^{N-|l|-1} x[n]y[n-l]$$
 , $l < 0$

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Properties of autocorrelation

• Energy of sequences x[n]

$$E_{x} = \sum_{n=-\infty}^{\infty} x^{2}[n] = r_{x}[0] \ge 0$$

Autocorrelation is maximum at lag l=0

$$|r_{xx}[l]| \leq r_{xx}[0] = E_x$$

• Autocorrelation is even \Rightarrow only compute values for $l \ge 0$

$$r_{xy}[l] = r_{xy}[-l] \Rightarrow r_{xx}[l] = r_{xx}[-l]$$

Properties of autocorrelation...

Normalized versions

$$\varrho_{xx}[l] = \frac{r_{xx}[l]}{r_{xx}[0]} \Rightarrow |\varrho_{xx}[l]| \le 1$$

$$\varrho_{xy}[l] = \frac{r_{xy}[l]}{\sqrt{r_{xx}[0]r_{yy}[0]}} \Rightarrow |\varrho_{xy}[l]| \le 1$$

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Properties cross- and autocorrelation...

- Example: Compute the autocorrelation of $x[n] = \alpha^n u[n]$
- Solution:

$$r_{xx}[l] = \sum_{n=-\infty}^{\infty} x[n+l]x[n]$$

$$= \sum_{n=-\infty}^{\infty} \alpha^{n+l}u[n+l]\alpha^n u[n]$$

$$= \alpha^l \sum_{n=0}^{\infty} \alpha^{2n} = \frac{\alpha^l}{1-\alpha^2}, l \ge 0$$
Since $x = [l] = x$. [I] we get the final α

Since $r_{xx}[-l] = r_{xx}[l]$, we get the final expression $r_{xx}[l] = \frac{\alpha^{|l|}}{1 - \alpha^2}, \forall l$

Example 1

- Let y[n] = Ax[n D]. Show that $D = \arg \max_{l} |r_{yx}[l]|$
- Solution:

$$r_{yx}[l] = \sum_{n=-\infty}^{\infty} y[n]x[n-l] = \sum_{n=-\infty}^{\infty} y[n+l]x[n]$$
$$= \sum_{n=-\infty}^{\infty} Ax[n+l-D]x[n] = Ar_{xx}[D-l]$$

From properties of autocorrelation sequences, we know

$$|r_{xy}[l]| = |A||r_{xx}[D-l]| \le |A||r_{xx}[0]|, \forall l$$

 $||r_{xy}[l]||$ reach its maximum for l = D

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Example 2

• Let y[n] = x[n] + Rx[n - D], i.e., received signal contains an echo. How to estimate R, D using the autocorrelation of y[n]?

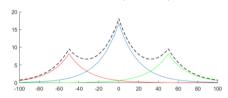
Example 3

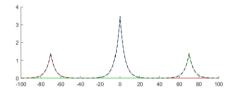
• Let $x[n] = \alpha^n u[n]$ and y[n] = x[n] + Rx[n - D]

$$\Rightarrow r_{yy}[l] = (1 + R^2)r_{xx}[l] + Rr_{xx}[l + D] + Rr_{xx}[l - D]$$

$$\alpha = 0.95, R = 0.8, D = 50$$

$$\alpha = 0.8, R = 0.5, D = 70$$





• Shape of $r_{yy}[l]$ depends on R, D

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Example 3...

- Details on how to obtain $r_{yy}[l]$ and R in previous slide
- From Slide 14: $r_{xx}[l] = \frac{\alpha^{|l|}}{1-\alpha^2}$

•
$$r_{yy}[l] = \sum_{n=-\infty}^{\infty} y[n+l]y[n]$$

$$= \sum_{n=-\infty}^{\infty} (x [n+l] + Rx [n+l-D])(x [n] + Rx [n-D])$$

$$= \sum_{n=-\infty}^{\infty} x [n+l]x [n] + R \sum_{n=-\infty}^{\infty} x [n+l-D]x [n]$$

$$+R\sum_{n=-\infty}^{\infty}x\left[n+l\right]x\left[n-D\right]+R^{2}\sum_{n=-\infty}^{\infty}x\left[n-D\right]x\left[n+l-D\right]$$

$$=r_{xx}[l]+Rr_{xx}[l-D]+Rr_{xx}[l+D]+R^2r_{xx}[l]$$

• Look at the following values (corresponding to the peaks in figure)

$$r_{yy}[0] = (1 + R^2)r_{xx}[0] + Rr_{xx}[-D] + Rr_{xx}[D] \approx (1 + R^2)r_{xx}[0]$$

$$r_{yy}[D] = (1 + R^2)r_{xx}[D] + Rr_{xx}[0] + Rr_{xx}[2D] \approx Rr_{xx}[0]$$

• Given values $r_{yy}[0]$ and $r_{yy}[D]$, we can solve for R

Example 2...

```
Matlab

1 = (-100:100);
a=0.8; % decay rate
R = 0.5; % echo strength
D = 70; % delay of echo

rxx_1 = a.^abs(1)/(1-a^2);
rxx_lpD = a.^abs(1+D)/(1-a^2);
rxx_lmD = a.^abs(1-D)/(1-a^2);
ryy = (1+R^2)*rxx_1+R*rxx_lpD+R*rxx_lmD;

figure
plot(1,(1+R^2)*rxx_1); hold on
plot(1,R*rxx_lpD,'r');
plot(1,R*rxx_lmD,'g')
plot(1,ryy,'k--','LineWidth',1)
```

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Energy spectral density

• Quantity $S_{xx}(\omega) \ge 0$ is the energy density spectrum of x[n]

$$r_{xx}[l] = x[l] * x[-l] \stackrel{\mathcal{F}}{\leftrightarrow} S_{xx}(\omega) = X(\omega)X^*(\omega) = |X(\omega)|^2$$

• Energy of complex-valued sequence x[n]

$$E_{x} = r_{xx}[0] = \sum_{n=-\infty}^{\infty} |x[n]|^{2}$$

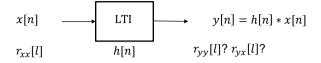
= $\frac{1}{2\pi} \int_{-\pi}^{\pi} S_{xx}(\omega) d\omega = \frac{1}{2\pi} \int_{-\pi}^{\pi} |X(\omega)|^{2} d\omega$

• Quantity $S_{xy}(\omega)$ is the cross-energy density spectrum

$$r_{xy}[l] = x[l] * y[-l] \stackrel{\mathcal{F}}{\leftrightarrow} S_{xy}(\omega) = X(\omega)Y^*(\omega)$$

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Input-output correlations



Input-output correlations

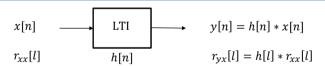
$$\begin{split} r_{yx}[l] &= h[l] * r_{xx}[l] \\ r_{yy}[l] &= r_{hh}[l] * r_{xx}[l] \\ E_y &= r_{yy}[0] = \sum_{k=-\infty}^{\infty} r_{hh}[k] r_{xx}[k] \end{split}$$

• Crosscorrelation between x[n] and y[n] can be seen as the output signal of an LTI system when input signal is $r_{xx}[n]$

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Input-output correlations and energy spectrum



• In z-transform domain

$$r_{yy}[l] = r_{hh}[l] * r_{xx}[l]$$

$$h[l] * h[-l] \stackrel{Z}{\leftrightarrow} H(z)H(z^{-1})$$

$$r_{yx}[l] = h[l] * r_{xx}[l] \stackrel{Z}{\leftrightarrow} H(z)S_{xx}(z)$$

$$r_{yy}[l] = r_{hh}[l] * r_{xx}[l] \stackrel{Z}{\leftrightarrow} H(z)H(z^{-1})S_{xx}(z)$$

· Output- and cross-energy density spectra

$$S_{yy}(\omega) = |H(\omega)|^2 S_{xx}(\omega) = |H(\omega)|^2 |X(\omega)|^2$$
$$S_{yx}(\omega) = H(\omega) S_{xx}(\omega)$$

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Input-output correlations and energy ...

• We have the following relation Fourier transform pair

$$r_{yy}[l] = \frac{1}{2\pi} \int_{-\pi}^{\pi} S_{yy}(\omega) e^{j\omega l} d\omega$$

• Energy of output sequence (of an LTI system)

$$r_{yy}[0] = \frac{1}{2\pi} \int_{-\pi}^{\pi} S_{yy}(\omega) d\omega = \frac{1}{2\pi} \int_{-\pi}^{\pi} |H(\omega)|^2 S_{xx}(\omega) d\omega$$

• Determine impulse response by signal with flat spectrum

$$h[n] = \frac{1}{S_{xx}} r_{yx}[n]$$

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Summary

Today:

- Crosscorrelation and autocorrelation sequences
- Linear time invariant systems
- Energy spectrum

Next:

Inverse z-transform

Example 2 (modified)

• Let y[n] = x[n] + Rx[n - D] be an audio signal corrupted by an echo. We would like to estimate R, D using the autocorrelation of y[n], and design a filter to remove the echo.

$$x[n] \longrightarrow b[n] \qquad y[n] = -\sum_{k=1}^{N} a_k y[n-k] + \sum_{k=0}^{M} b_k x[n-k]$$

$$X(z) \qquad Y(z) = H(z)X(z)$$

Matlab files on BB: Correlation.m

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Example 2 (modified)...

Model the problem using an LTI system

$$x[n] \longrightarrow h[n] \qquad y[n] = x[n] + Rx[n - D]$$

$$X(z) \qquad Y(z) = H(z)X(z) = (1 + Rz^{-D})X(z)$$

- Estimate R and D using autocorrelation sequence $r_{vv}[n]$
- Find the inverse system $H_I(z)$ such that (see previous lecture)

$$h[n] * h_I[n] = \delta[n] \stackrel{\mathcal{Z}}{\leftrightarrow} H(z)H_I(z) = 1$$
$$\Rightarrow H_I(z) = \frac{1}{1 + Rz^{-D}}$$

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Example 2 (modified)...

- If R < 1, H(z) is minimum phase and so is $H_I(z)$
- We can find a causal and stable filter $h_I[n]$

$$x[n] \longrightarrow h[n] \xrightarrow{y[n]} h_I[n] \xrightarrow{x[n]}$$

- We get D from inspecting the peaks of $r_{yy}[l]$
- We obtain an estimate *R* from the relation

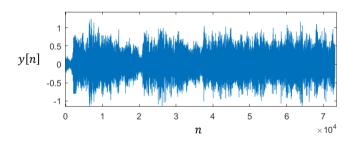
$$\frac{r_{yy}[0]}{r_{yy}[D]} = \frac{1+R^2}{R}$$

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Example 2 (modified)...

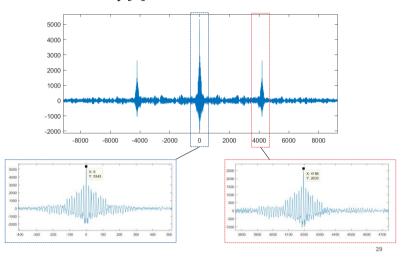
• Plot of received signal y[n] (R = 0.98, D = 4196):



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• Autocorrelation of y[n]:



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Example 2 (modified)...

• From figure we get:

$$D = 4196, \ \frac{r_{yy}[0]}{r_{yy}[D]} = \frac{1+R^2}{R} = \frac{5343}{2633} \Rightarrow R = 0.8430$$

• Delay is correct but parameter estimate of *R* is not exact. Listen to the equalized signal and judge whether the echo is removed (or suppressed)