

Autocorrelation is an effective way to observe periodicities in a signal corrupted by noise.

Let $s[n]$ be a periodic signal of interest.

Let $v[n]$ denote a zero-mean noise waveform that is uncorrelated with $s[n]$.

Let $z[n] = s[n] + v[n]$

$$\begin{aligned} r_{zz}[l] &= \lim_{M \rightarrow \infty} \frac{1}{2M+1} \sum_{n=-M}^M \overbrace{(s[n] + v[n])}^{z[n]} \overbrace{(s^*[n-l] + v^*[n-l])}^{z^*[n-l]} \\ &= \lim_{M \rightarrow \infty} \frac{1}{2M+1} \sum_{n=-M}^M s[n]s^*[n-l] + s[n]v^*[n-l] + v[n]s^*[n-l] + v[n]v^*[n-l] \end{aligned}$$

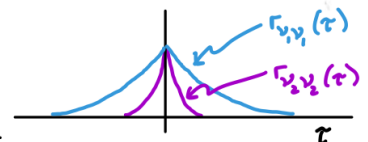
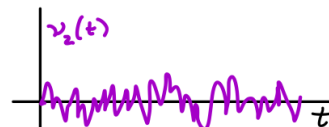
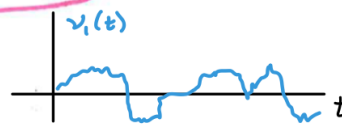
$$= r_{ss}[l] + r_{sv}[l] + r_{vs}[l] + r_{vv}[l]$$

if $s[n]$ is periodic with period N , then so will be $r_{ss}[l]$.
The max value of $r_{ss}[l]$ occurs at $l=0$, $l=\pm N$, $l=\pm 2N, \dots$

both of these terms will be equal to zero because $v[n]$ is zero-mean and uncorrelated with $s[n]$

the elements of a random noise process are only correlated over short time intervals (depending on the frequency content of the process).

$$\Rightarrow \lim_{l \rightarrow \infty} r_{vv}[l] = 0$$



$$r_{zz}[0] = r_{ss}[0] + r_{vv}[0]$$

= Signal power + noise power

assumes $s[n]$ & $v[n]$ are uncorrelated

Autocorrelation is an effective way of revealing periodicities in noisy signals.

Let $z_l[n] = \begin{cases} s[n] + v[n], & 0 \leq n \leq 500 \\ 0, & \text{otherwise} \end{cases}$

where $s[n] = \cos\left(2\pi \frac{1}{10}(n - 5)\right)$ is a sinusoid of amplitude 1 and period 10 samples

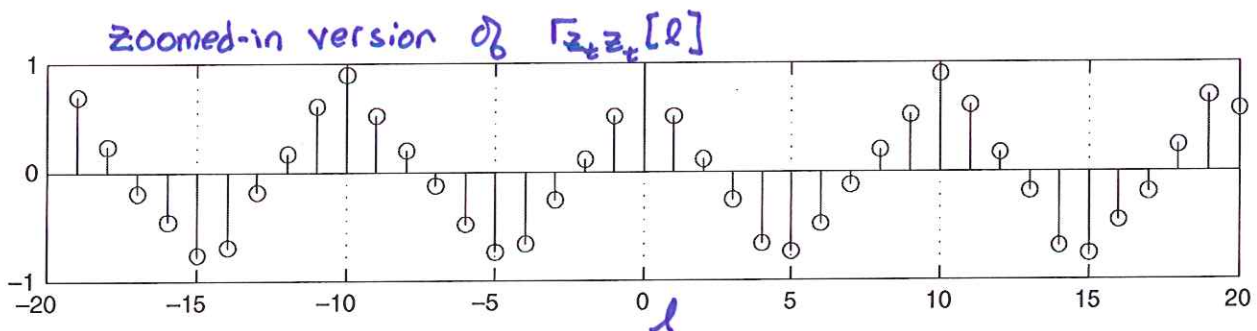
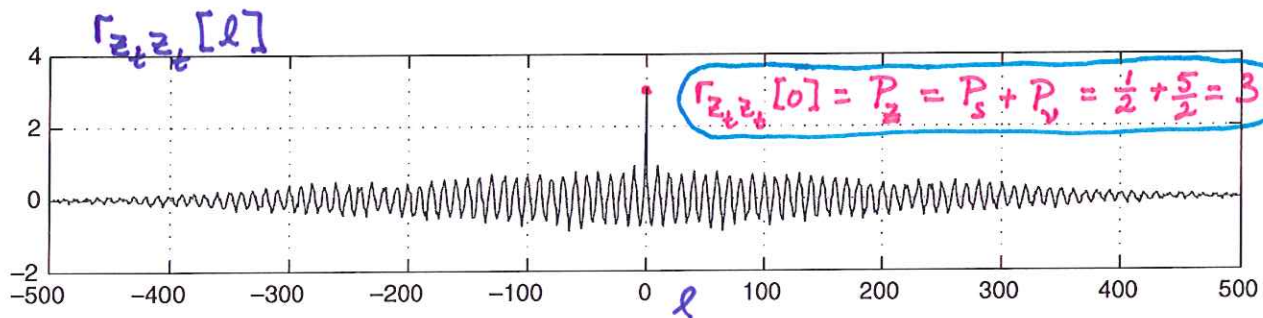
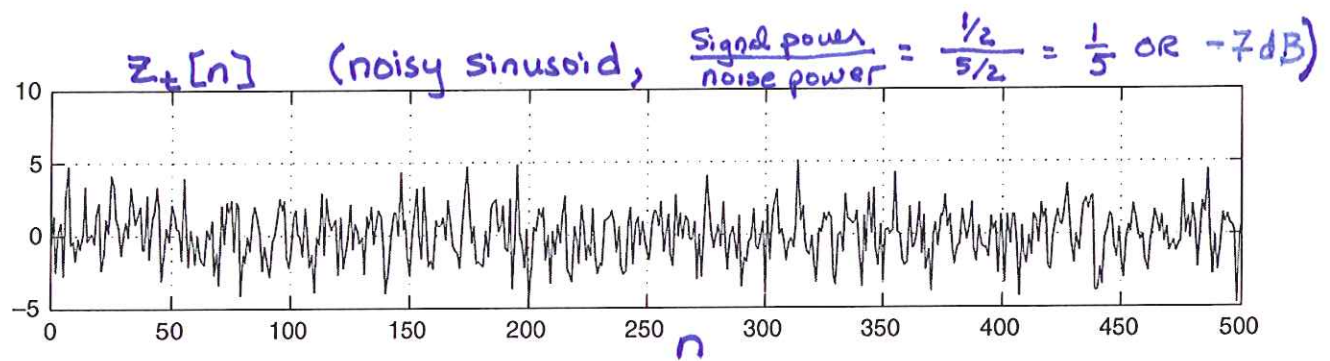
and $v[n]$ is a zero-mean white gaussian noise with variance 5/2

$$\Rightarrow \text{SNR} = \frac{1}{5}$$

$$\text{or } 10 \log_{10}\left(\frac{1}{5}\right) = -7 \text{ dB}$$

Then can use matlab to compute: $r_{z_l z_l}[\ell] = \frac{1}{501} \sum_{n=0}^{500} x_l[n] y_l[n - \ell]$

Plots of $z_l[n]$, $r_{z_l z_l}[\ell]$, and a zoomed version of $r_{z_l z_l}[\ell]$ are shown below.



Note

- autocorrelation function has the same period as $s[n]$
- the autocorrelation function does not retain phase information

Correlations involving the input and output of an LTI system

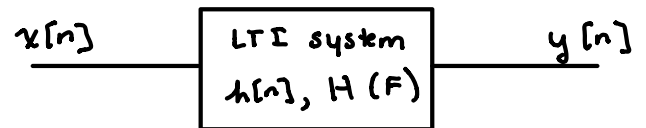
Let $x[n]$ denote an energy signal with autocorrelation function:

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

The energy spectral density of $x[n]$ is denoted by $S_{xx}(f)$ and may be found as the DTFT of $r_{xx}[l]$. $\Rightarrow S_{xx}(F) = X(F) X^*(F)$

Exercise

Given that $y[n]$ is the response of an LTI system with impulse response $h[n]$ when the system's input is $x[n]$.



Find the relationship between $r_{yx}[l]$ and $r_{xx}[l]$.

Solution (Hint: start by expressing $r_{yx}[l]$ as a convolution of two signals)

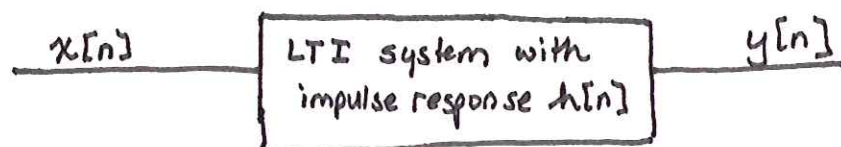
$$r_{yx}[l] =$$

The cross spectral density $S_{yx}(F)$ is the DTFT of $r_{yx}[l]$.

$$\Rightarrow S_{yx}(F) =$$

$$\Rightarrow H(F) =$$

Correlations involving the input, $x[n]$, and the output, $y[n]$, of an LTI system with impulse response $h[n]$.



Problem Given that $x[n]$ has autocorrelation $\Gamma_{xx}[l]$

- a) Find the relation between $\Gamma_{yx}[l]$ and $\Gamma_{xx}[l]$
- b) Find the relation between $\Gamma_{yy}[l]$ and $\Gamma_{xx}[l]$
- c) Find the relation between $\Gamma_{xy}[l]$ and $\Gamma_{xx}[l]$

The relationships determined in this problem are very useful for system identification.

Solution:

$$a) \quad r_{yx}[l] = y[l] * x^*[-l]$$

Using relation of cross correlation and convolution

$$= \underbrace{h[l] * x[l]}_{r_{xx}[l]} * x^*[-l]$$

Output of LTI system is the convolution of system's impulse response and the input

$$= h[l] * r_{xx}[l]$$

using relation between cross correlation and convolution

$$b) \quad r_{yy}[l] = y[l] * y^*[-l]$$

expressing cross-correlation in terms of convolution

$$= \underbrace{h[l] * x[l]}_{y[l]} * \underbrace{x^*[-l] * h^*[-l]}_{y^*[-l]}$$

input-output relation of LTI system + property of convolution shown below.

$$= h[l] * h^*[-l] * \underbrace{x[l] * x^*[-l]}_{r_{xx}[l]}$$

convolution is commutative and associative

$$= h[l] * h^*[-l] * r_{xx}[l]$$

using relation between cross correlation and convolution

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

$$c) \quad r_{xy}[l] = r_{yx}^*[-l]$$

using property of cross-correlations

$$= h^*[-l] * r_{xx}^*[-l]$$

using result of part (a) together with the convolution property below

$$= h^*[-l] * r_{xx}[l]$$

auto correlation functions are Hermitian

Useful property of convolution:

$$\text{If } z[l] = w[l] * v[l] \quad \text{then } z[-l] = w[-l] * v[-l]$$

From which it follows that

$$\text{if } z[l] = w[l] * v[l] \quad \text{then } z^*[-l] = w^*[-l] * v^*[-l]$$

$$\text{If } z[l] = w[l] * v[l] \text{ then } z[-l] = w[-l] * v[-l]$$

Proof

$$z[l] = \sum_{k=-\infty}^{\infty} w[k] v[l-k]$$

using definition of convolution

$$\Rightarrow z[-l] = \sum_{k=-\infty}^{\infty} w[k] v[-l-k]$$

replaced l by $-l$ on both sides of equation

$$= \sum_{m=-\infty}^{\infty} w[-m] v[-l+m]$$

change of variables:
 $k = -m$

$$\text{let } w_f[m] = w[-m]$$

$$\text{let } v_f[m] = v[-m] \Rightarrow v_f[l-m] = v[m-l]$$

$$= \sum_{m=-\infty}^{\infty} w_f[m] v_f[l-m]$$

using previously defined relationships

$$= w_f[l] * v_f[l]$$

definition of convolution

$$= w[-l] * v[-l]$$

relations between
 $w_f[\cdot] \neq w[\cdot]$, $v_f[\cdot] \neq v[\cdot]$

Q.E.D.

$$\text{If } z[l] = w[l] * v[l] \text{ then } z^*[-l] = w^*[-l] * v^*[-l]$$

Proof

From 3rd last equality in the proof above, we have that:

$$z[-l] = \sum_{m=-\infty}^{\infty} w_f[m] v_f[l-m]$$

$$\Rightarrow z^*[-l] = \sum_{m=-\infty}^{\infty} w_f^*[m] v_f^*[l-m]$$

$$= w_f^*[l] * v_f^*[l]$$

$$= w^*[-l] * v^*[-l]$$

Q.E.D.