

ECE 251A: Digital Signal Processing I

Wide Sense Stationary Processes II

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Power Spectrum

Power Spectral Density (PSD)

$$\begin{aligned}R_x(e^{j\omega}) &= \sum_{m=-\infty}^{\infty} r[m]e^{-j\omega m} \\r[m] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} R_x(e^{j\omega})e^{j\omega m}d\omega\end{aligned}$$

Notation: $r[m] = r_x[m] = r_{xx}[m]$.

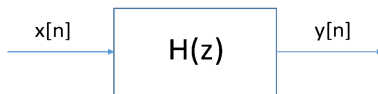
$$R_x(z) = \sum_{m=-\infty}^{\infty} r[m]z^{-m} \quad \text{ROC} = \{|z| : r_L < |z| < r_R\}$$

The ROC includes the unit circle

Properties of PSD $R_x(e^{j\omega})$

- ① PSD is a real function of ω , i.e. $R_x(e^{j\omega}) = R_x^*(e^{j\omega})$. Follows from the Hermitian symmetry of $r[m]$.
- ② $R_x(e^{j\omega}) \geq 0 \Leftrightarrow r[m]$ is a positive semidefinite sequence.
- ③ $r[0] = \frac{1}{2\pi} \int_{-\pi}^{\pi} R_x(e^{j\omega}) d\omega = E(|x[n]|^2) \geq 0$.
- ④ If $x[n]$ is a real RP, then $R_x(e^{j\omega}) = R_x(e^{-j\omega})$.

RP and LTI Systems



Questions of interest: Properties of $y[n]$ if $x[n]$ is WSS and joint properties of $x[n]$ and $y[n]$.

Applications

- 1 Spectral Shaping (Spectral Factorization): Generating a WSS process with desired autocorrelation sequence. (input a white noise sequence and choosing $H(z)$ appropriately).
- 2 Filtering/Prediction: Enhancing $x[n]$ in the presence of noise or predicting another RP given $x[n]$.

Example I ($a = 0.5$)

$x[n] = ax[n-1] + w[n]$, where $w[n]$ is a zero mean, variance one, i.i.d. Gaussian sequence. $H(z) = \frac{1}{1-az^{-1}}$.

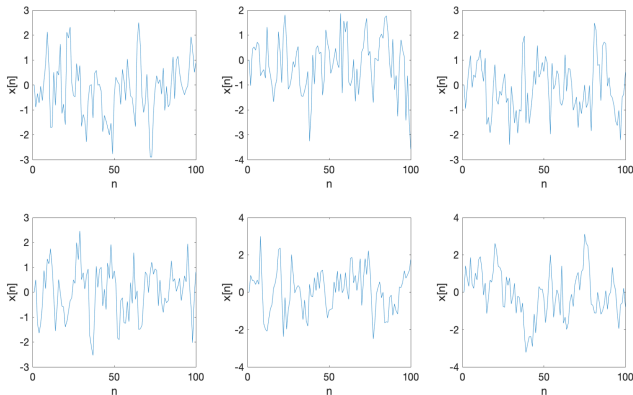


Figure: Six realizations for $a = 0.5$

Example II ($a = -0.9$)

$x[n] = ax[n-1] + w[n]$, where $w[n]$ is a zero mean, variance one, i.i.d. Gaussian sequence. $H(z) = \frac{1}{1-az^{-1}}$.

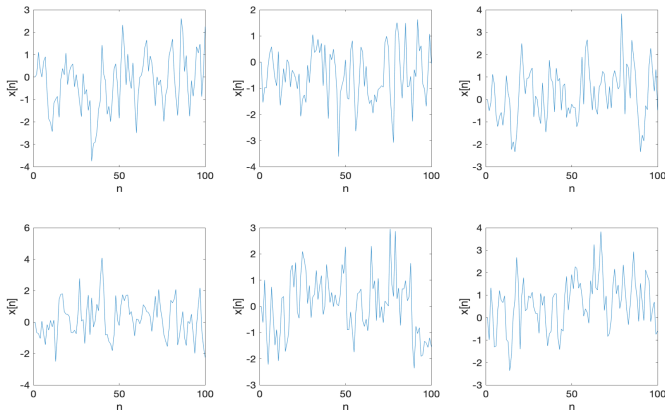


Figure: Six realizations for $a = -0.9$

WSS and LTI Systems

WSS process $x[n]$ is input to a LTI system with transfer function $H(z)$ and output is $y[n]$.

Is $y[n]$ WSS? Answer: Yes

Are $x[n]$ and $y[n]$ jointly WSS? need to check if the cross-correlation $E(x[n+m]y^*[n])$ is also a function of time difference m , i.e.

$r_{xy}[m] = E(x[n+m]y^*[n])$. Answer: Yes

Check:

Mean: $E(y[n]) = E(\sum_l h[l]x[n-l]) = \sum_l h[l]E(x[n-l]) = \sum_l h[l]\mu_x = H(e^{j0})\mu_x = H(1)\mu_x$

Cross-Correlation

$$\begin{aligned} E(x[n+m]y^*[n]) &= E(x[n+m] \sum_l h^*[l]x^*[n-l]) \\ &= \sum_l h^*[l]E(x[n+m]x^*[n-l]) = \sum_l h^*[l]r_{xx}[m+l] \\ &= \sum_l h^*[-l]r_{xx}[m-l] = r_{xx}[m] * h^*[-m] \end{aligned}$$

Conclusion: $r_{xy}[m] = r_{xx}[m] * h^*[-m]$. Function of only the time difference.

Autocorrelation of $y[n]$

$$\begin{aligned} E(y[n+m]y^*[n]) &= E\left(\sum_l h[l]x[n+m-l]y^*[n]\right) = \sum_l h[l]r_{xy}[m-l] \\ &= h[m] * r_{xy}[m] = h[m] * h^*[-m] * r_{xx}[m]. \end{aligned}$$

Conclusion: $y[n]$ is WSS with mean $\mu_y = H(1)\mu_x$ and autocorrelation sequence $r_{yy}[m] = h[m] * h^*[-m] * r_{xx}[m]$.

In the z -domain $R_{yy}(z) = \mathcal{Z}(r_{yy}[m]) = R_{xx}(z)H^*\left(\frac{1}{z^*}\right)$ and simplifies to $R_{xx}(z)H(z^{-1})$ for real systems.

$$R_{yy}(z) = R_{xx}(z)H(z)H^*\left(\frac{1}{z^*}\right) \text{ or } R_{yy}(e^{j\omega}) = R_{xx}(e^{j\omega})|H(e^{j\omega})|^2$$

Spectral Factorization: Given $R_{yy}(z)$ and with white noise input, $R_{xx}(z) = 1$, find $H(z)$ or equivalently factor $R_{yy}(z)$ such that $R_{yy}(z) = H(z)H^*\left(\frac{1}{z^*}\right)$.

Positivity of $R_{xx}(e^{j\omega})$

Proof by Contradiction: Suppose $R_{xx}(e^{j\omega}) < 0$ for $\omega_1 \leq \omega \leq \omega_2$. Let $H(z)$ be a bandpass filter over this region

$$H(e^{j\omega}) = \begin{cases} 1, & \omega_1 \leq \omega \leq \omega_2 \\ 0 & \text{otherwise} \end{cases}$$

If $x[n]$ is the input to $H(z)$, then the output $y[n]$ has a $r_{yy}[0]$ given by

$$r_{yy}[0] = \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(e^{j\omega})|^2 R_{xx}(e^{j\omega}) d\omega = \frac{1}{2\pi} \int_{\omega_1}^{\omega_2} R_{xx}(e^{j\omega}) d\omega < 0$$

This contradicts the fact that $r_{yy}[0] \geq 0$. This suggests the initial assumption of negative PSD is not true.

PSD estimation via power at output of a bandpass filter

$$r_{yy}[0] = \frac{1}{2\pi} \int_{\omega_0 - \delta\omega}^{\omega_0 + \delta\omega} R_{xx}(e^{j\omega}) d\omega \approx \frac{2\delta\omega}{2\pi} R_{xx}(e^{j\omega_0}) \text{ or } R_{xx}(e^{j\omega_0}) \approx \frac{2\pi}{2\delta\omega} r_{yy}[0]$$

Verifying if a Sequence is a Valid Autocorrelation Sequence

Let $\{q[m]\}_{-\infty}^{\infty}$ be a sequence, and we wish to know if it is a valid autocorrelation sequence.

Checking for maximum at the origin and symmetry is easy.

Checking for positive semi-definiteness is difficult.

Compute the DTFT of $q[m]$

$$Q(e^{j\omega}) = \sum_{m=-\infty}^{\infty} q[m]e^{-j\omega m}$$

Main Result: $\{q[m]\}_{-\infty}^{\infty}$ is a valid autocorrelation sequence if and only if

$$Q(e^{j\omega}) \geq 0.$$

- Compute $p[m] = h[m] * h^*[-m] * q[m] = \sum_l \sum_p h[l] h^*[p] q[m + p - l]$ and $P(e^{j\omega}) = |H(e^{j\omega})|^2 Q(e^{j\omega})$
- Let $h[m] = a_m^*$. Then $p[m] = \sum_l \sum_p a_l^* a_p q[m + p - l]$ and $P(e^{j\omega}) = |A(e^{j\omega})|^2 Q(e^{j\omega})$, where $A(e^{j\omega}) = \sum_l a_l^* e^{-j\omega l}$.
- From Fourier theory $p[m] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P(e^{j\omega}) e^{j\omega m} d\omega$
- $p[0] = \sum_l \sum_p a_l^* a_p q[p - l]$ and $p[0] = \frac{1}{2\pi} \int_{-\pi}^{\pi} P(e^{j\omega}) d\omega = \frac{1}{2\pi} \int_{-\pi}^{\pi} |A(e^{j\omega})|^2 Q(e^{j\omega}) d\omega$,
- $p[0] \geq 0$ iff $Q(e^{j\omega}) \geq 0$.

Ergodicity

For WSS, one needs to compute the mean $\mu = E(x[n])$ and autocorrelation $r[m] = E(x[n]x^*[n-m])$. This is a challenge because the expectation operation implies an ensemble average, average over a collection of sequences, i.e.

$$\hat{\mu}_N = \frac{1}{N} \sum_{l=0}^{N-1} x[n, \zeta_l] \text{ and autocorrelation } \hat{r}_N[m] = \frac{1}{N} \sum_{l=0}^{N-1} x[n, \zeta_l] x^*[n-m, \zeta_l].$$

In practice, we have one realization and may use time averages

$$\hat{\mu} = \frac{1}{N} \sum_{n=0}^{N-1} x[n, \zeta] \text{ and } \hat{r}[m] = \frac{1}{N} \sum_{n=0}^{N-1} x[n, \zeta] x^*[n-m, \zeta].$$

Ergodicity relates to when these time averages converge to the ensemble average.

A RP is mean ergodic when the time average $\hat{\mu} = \frac{1}{N} \sum_{n=0}^{N-1} x[n] \rightarrow \mu$ in the mean squared sense, i.e. $E(|\hat{\mu} - \mu|^2) \rightarrow 0$ as $N \rightarrow \infty$.

A RP is correlation ergodic when the time average $\hat{r}[m] \rightarrow r[m]$ in the mean squared sense, i.e. $E(|\hat{r}[m] - r[m]|^2) \rightarrow 0$ as $N \rightarrow \infty$.

Example I

$x[n] = A \cos(\omega_0 n + \phi)$, where ϕ is a random variable uniform between $[-\pi, \pi]$.

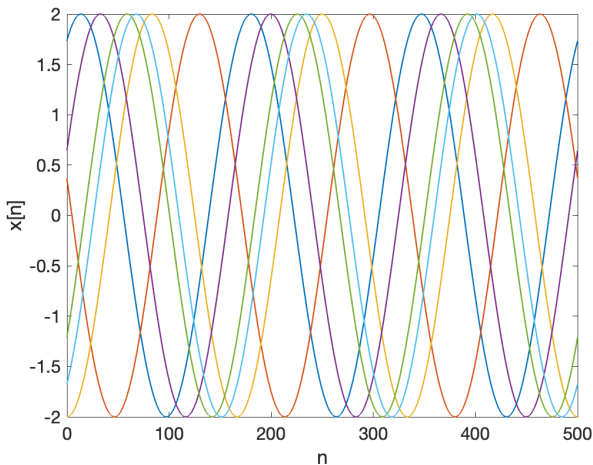


Figure: Six realizations

Examples Revisited

- ❶ $x[n] = A \cos(\omega_0 n + \phi)$, where ϕ is a random variable uniform between $[-\pi, \pi]$. WSS: Yes ($\mu = 0$, and $r[m] = \frac{A^2}{2} \cos(\omega_0 m)$.)
- ❷ $x[n] = w[n]$, where $w[n]$ is a i.i.d. sequence of Gaussian random variable with mean zero and variance 1. WSS: Yes ($\mu = 0$, and $r[m] = \delta[m]$.)
- ❸ $x[n] = ax[n-1] + w[n]$, $|a| < 1$, where $w[n]$ is a i.i.d. sequence of Gaussian random variable with mean zero and variance 1. $x[n]$ is the output of a LTI system $H(z) = \frac{1}{1-az^{-1}}$ with input $w[n]$. WSS: Yes ($\mu = 0$, and $r[m]$ shaped by $H(z)$)
- ❹ $x[n] = A$, where A is a random variable uniform between $[-1, 1]$. WSS: Yes ($\mu = 0$, and $r[m] = E(A^2) = \frac{1}{3}$.)
- ❺ $x[n] = A \cos(\omega_0 n + \phi)$, where ϕ is a random variable uniform between $[-\pi, \pi]$ and A is a random variable independent of ϕ that is uniform between $[-1, 1]$. WSS: Yes ($\mu = 0$, and $r[m] = \frac{E(A^2)}{2} \cos(\omega_0 m)$.)

Examples Revisited

- ❶ $x[n] = A \cos(\omega_0 n + \phi)$, where ϕ is a random variable uniform between $[-\pi, \pi]$. Mean and Correlation Ergodic: Yes
- ❷ $x[n] = w[n]$, where $w[n]$ is a i.i.d. sequence of Gaussian random variables with mean zero and variance 1. Mean and Correlation Ergodic: Yes
- ❸ $x[n] = ax[n-1] + w[n]$, $|a| < 1$, where $w[n]$ is a i.i.d. sequence of Gaussian random variable with mean zero and variance 1. $x[n]$ is the output of a LTI system $H(z) = \frac{1}{1-az^{-1}}$ with input $w[n]$. Mean and Correlation Ergodic: Yes
- ❹ $x[n] = A$, where A is a random variable uniform between $[-1, 1]$. Mean and Correlation Ergodic: No
- ❺ $x[n] = A \cos(\omega_0 n + \phi)$, where ϕ is a random variable uniform between $[-\pi, \pi]$ and A is a random variable independent of ϕ that is uniform between $[-1, 1]$. Mean and Correlation Ergodic: Mean Yes, Correlation No