

2012 Second Midterm for Random Processes

1. Suppose the power spectrum of a (discrete) regular WSS process $\mathbf{x}[t]$ is

$$S_{xx}[\omega] = \frac{5 - 4 \cos(\omega)}{10 - 6 \cos(\omega)}.$$

- (a) (10%) Find its innovation filter $L[z]$ that is analytic for $|z| > 1$.
 Hint: (i) $\cos(\omega) = (z + z^{-1})/2$, and (ii) $S_{xx}[z] = L[z]L[1/z]$ and (iii) roots inside the unit circle.
- (b) (10%) Find the impulse response $l[\tau]$ of the innovation filter $L[z]$ in (a).
 Hint: $l[\tau] = \frac{1}{2\pi} \int_{-\pi}^{\pi} L[\omega] e^{j\omega\tau} d\omega$
- (c) (10%) Determine $S_{xx}^+[z]$ if we rewrite $S_{xx}[z]$ as

$$S_{xx}[z] = S_{xx}^+[z] + S_{xx}^+[1/z] - \sum_{i=1}^n \alpha_i,$$

where

$$R_{xx}^+[\tau] = \frac{1}{2\pi} \int_{-\pi}^{\pi} S_{xx}^+[e^{j\omega}] e^{j\omega\tau} d\omega = 0 \text{ for } \tau < 0,$$

and $\alpha_k = \gamma_k L[z_k^{-1}]$.

Hint: Start from $L[z] = \frac{\gamma_i}{1 - z_i z^{-1}}$ and observe that

$$\begin{aligned} S_{xx}[z] &= L[z]L[z^{-1}] = \left(\sum_{i=1}^n \frac{\gamma_i}{1 - z_i z^{-1}} \right) \left(\sum_{k=1}^n \frac{\gamma_k}{1 - z_k z} \right) \\ &= \sum_{i=1}^n \sum_{k=1}^n \frac{\gamma_i \gamma_k}{(1 - z_i z^{-1})(1 - z_k z)} \\ &= \sum_{i=1}^n \sum_{k=1}^n \left(\frac{-}{1 - z_i z^{-1}} + \frac{-}{1 - z_k z} - \dots \right) \end{aligned}$$

Definition (discrete regular processes) A (discrete) process $\mathbf{x}[t]$ is regular if there exists a $L[z]$ such that

$$S_{xx}[\omega] = |L[e^{j\omega}]|^2$$

where $L[z]$ ($z = e^{j\omega}$) is analytic for $|z| > 1$.

Solution.

$$\begin{aligned} \text{(a) } S_{xx}[z] &= \frac{5 - 2(z + z^{-1})}{10 - 3(z + z^{-1})} = \frac{2(z - 1/2)(z - 2)}{3(z - 1/3)(z - 3)} = \frac{2(z - 1/2)}{3(z - 1/3)} \cdot \frac{2(z^{-1} - 1/2)}{3(z^{-1} - 1/3)} \\ &\Rightarrow L[z] = \frac{2(z - 1/2)}{3(z - 1/3)} \end{aligned}$$

(b) From (a), $L[\omega] = \frac{2(e^{j\omega} - 1/2)}{3(e^{j\omega} - 1/3)}$. Then,

$$\begin{aligned}
1[\tau] &= \frac{1}{2\pi} \int_{-\pi}^{\pi} L[\omega] e^{j\omega\tau} d\omega \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{2(e^{j\omega} - 1/2)}{3(e^{j\omega} - 1/3)} e^{j\omega\tau} d\omega \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(1 - \frac{e^{j\omega}}{3e^{j\omega} - 1}\right) e^{j\omega\tau} d\omega \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} \left(1 - \frac{1/3}{1 - (1/3)e^{-j\omega}}\right) e^{j\omega\tau} d\omega \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 - 3^{-1} [1 + 3^{-1}e^{-j\omega} + 3^{-2}e^{-j2\omega} + \dots]) e^{j\omega\tau} d\omega \\
&= \begin{cases} 0, & \tau < 0 \\ 1 - 3^{-1}, & \tau = 0 \\ -3^{-(1+\tau)}, & \tau > 0 \end{cases}
\end{aligned}$$

(c) $L[z]$ can be expanded into partial fractions as:

$$L[z] = \sum_{i=1}^n \frac{\gamma_i}{1 - z_i z^{-1}} = \frac{2[1 - (1/2)z^{-1}]}{3[1 - (1/3)z^{-1}]} = \frac{\gamma_1}{1 - z_1 z^{-1}} + \frac{\gamma_2}{1 - z_2 z^{-1}},$$

where $z_1 = 0$, $z_2 = 1/3$, $\gamma_1 = 1$ and $\gamma_2 = -\frac{1}{3}$. We can then derive:

$$\begin{aligned}
S_{xx}[z] &= L[z]L[z^{-1}] \\
&= \left(1 + \frac{(-1/3)}{1 - (1/3)z^{-1}}\right) \left(1 + \frac{(-1/3)}{1 - (1/3)z}\right) \\
&= 1 \times 1 + 1 \times \frac{(-1/3)}{1 - (1/3)z} + \frac{(-1/3)}{1 - (1/3)z^{-1}} \times 1 + \frac{(-1/3)}{1 - (1/3)z^{-1}} \times \frac{(-1/3)}{1 - (1/3)z} \\
&= (1 \times 1 + 1 \times 1 - 1 \times 1) + \left(1 \times \left(-\frac{1}{3}\right) + 1 \times \frac{(-1/3)}{1 - (1/3)z} - 1 \times \left(-\frac{1}{3}\right)\right) \\
&+ \left(\frac{(-1/3)}{1 - (1/3)z^{-1}} \times 1 + \left(-\frac{1}{3}\right) \times 1 - \left(-\frac{1}{3}\right) \times 1\right) \\
&+ \left(\frac{(-1/3)(-1/3)/[1 - (1/3)(1/3)]}{1 - (1/3)z^{-1}} + \frac{(-1/3)(-1/3)/[1 - (1/3)(1/3)]}{1 - (1/3)z} - \frac{(-1/3)(-1/3)}{1 - (1/3)(1/3)}\right) \\
&= \left(1 \times 1 + 1 \times \left(-\frac{1}{3}\right) + \frac{(-1/3)}{1 - (1/3)z^{-1}} \times 1 + \frac{(-1/3)(-1/3)/[1 - (1/3)(1/3)]}{1 - (1/3)z^{-1}}\right) \\
&+ \left(1 \times 1 + 1 \times \frac{(-1/3)}{1 - (1/3)z} + \left(-\frac{1}{3}\right) \times 1 + \frac{(-1/3)(-1/3)/[1 - (1/3)(1/3)]}{1 - (1/3)z}\right) \\
&- \left(1 \times 1 + 1 \times \left(-\frac{1}{3}\right) + \left(-\frac{1}{3}\right) \times 1 + \frac{(-1/3)(-1/3)}{1 - (1/3)(1/3)}\right) \\
&= \left(\frac{2}{3} + \frac{(-5/24)}{1 - (1/3)z^{-1}}\right) + \left(\frac{2}{3} + \frac{(-5/24)}{1 - (1/3)z}\right) - \frac{11}{24}
\end{aligned}$$

Hence,

$$S_{xx}^+[z] = \frac{2}{3} + \frac{(-5/24)}{1 - (1/3)z^{-1}}.$$

Alternatively, you can use the formula to derive:

$$\alpha_k = \gamma_k \mathbf{L}[z_k^{-1}] = \begin{cases} \gamma_1 \mathbf{L}[z_1^{-1}], & k = 1 \\ \gamma_2 \mathbf{L}[z_2^{-1}], & k = 2 \end{cases} = \begin{cases} \gamma_1(\gamma_1 + \gamma_2), & k = 1 \\ \gamma_2 \left(\frac{\gamma_1}{1 - z_1 z_2} + \frac{\gamma_2}{1 - z_2^2} \right), & k = 2 \end{cases} = \begin{cases} \frac{2}{3}, & k = 1 \\ -\frac{5}{24}, & k = 2 \end{cases}$$

which gives that

$$S_{xx}^+[z] = \sum_{i=1}^2 \frac{\alpha_i}{1 - z_i z^{-1}} = \alpha_1 + \frac{\alpha_2}{1 - z_2 z^{-1}} = \frac{2}{3} + \frac{(-5/24)}{1 - (1/3)z^{-1}}$$

2. (a) (10%) Find the autocorrelation function of a WSS process $\mathbf{x}(t)$ that is the output due to unit-power zero-mean white input $\mathbf{i}(t)$ and an innovation filter

$$\mathbf{L}(s) = \frac{7 + 5s}{(1 + s)(3 + s)}.$$

Hint: $\mathbf{L}(s) = \sum_{i=1}^n \frac{\gamma_i}{s - s_i}$ and $S_{xx}^+(s) = \sum_{i=1}^n \frac{\alpha_i}{s - s_i}$ and $R_{xx}(\tau) = R_{xx}^+(|\tau|)$.
where $\alpha_k = \gamma_k \mathbf{L}(-s_k)$ for $1 \leq k \leq n$.

- (b) (10%) Repeat (a) for the innovation filter below.

$$\tilde{\mathbf{L}}(s) = \frac{7 - 5s}{(1 - s)(3 - s)}.$$

I.e., find the autocorrelation function of a WSS process $\tilde{\mathbf{x}}(t)$ that is the output due to unit-power zero-mean white input $\mathbf{i}(t)$ and the innovation filter $\tilde{\mathbf{L}}(s)$.

Hint: Think of the relation between $S_{\tilde{x}\tilde{x}}(s)$ and $\tilde{\mathbf{L}}(s)$.

Solution.

- (a) With

$$\mathbf{L}(s) = \sum_{i=1}^n \frac{\gamma_i}{s - s_i} = \frac{1}{s + 1} + \frac{4}{s + 3},$$

we have $s_1 = -1$, $s_2 = -3$, $\gamma_1 = 1$, $\gamma_2 = 4$, $\alpha_1 = \gamma_1 \mathbf{L}[-s_1] = \frac{3}{2}$ and $\alpha_2 = \gamma_2 \mathbf{L}[-s_2] = \frac{11}{3}$.
Hence, from

$$\begin{aligned} R_{xx}^+(\tau) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}^+(\omega) e^{j\omega\tau} d\omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(\sum_{i=1}^n \frac{\alpha_i}{j\omega - s_i} \right) e^{j\omega\tau} d\omega \\ &= \begin{cases} \sum_{i=1}^n \alpha_i e^{s_i \tau}, & \tau > 0 \\ 0, & \tau < 0 \end{cases} \end{aligned}$$

we have

$$R_{xx}(\tau) = R_{xx}^+(|\tau|) = \sum_{i=1}^2 \alpha_i e^{s_i |\tau|} = \frac{3}{2} e^{-|\tau|} + \frac{11}{3} e^{-3|\tau|}$$

(b) Since $S_{\tilde{x}\tilde{x}}(s) = \tilde{L}(s)\tilde{L}(-s) = L(-s)L(s) = S_{xx}(s)$, the autocorrelation function of $\tilde{\mathbf{x}}(t)$ should be the same as that of $\mathbf{x}(t)$.

3. (a) (10%) Does the white WSS process $\mathbf{x}(t)$ with $S_{xx}(\omega) = \frac{N_0}{2}$ satisfy the Paley-Wiener condition?

Lemma (Paley-Wiener condition) A process $\mathbf{x}(t)$ is regular if the Paley-Wiener condition holds, i.e.,

$$\int_{-\infty}^{\infty} \frac{|\log S_{xx}(\omega)|}{1 + \omega^2} d\omega < \infty.$$

- (b) (10%) Is the Fourier transform $\mathbf{X}(t)$ of $\mathbf{x}(t)$ in (a) a regular process? A pure Yes/No answer can only earn you half of the credits. Some justification is needed in order to receive the full credits.

Definition (minimum-phase system) A system is called *minimum-phase* if both $L(\omega)$ and $1/L(\omega)$ are causal and have finite energy.

Definition (causal filter) A causal filter is one whose output depends only on past and present inputs.

A process that can be so represented (i.e., as a response of a *minimum-phase* system $L(\omega)$ with a white input $\mathbf{i}(t)$ of unit power) is called *regular*.

Solution.

(a)

$$\int_{-\infty}^{\infty} \frac{|\log S_{xx}(\omega)|}{1 + \omega^2} d\omega = \int_{-\infty}^{\infty} \frac{|\log(N_0/2)|}{1 + \omega^2} d\omega = \pi |\log(N_0/2)| < \infty$$

Hence, it satisfies the Paley-Wiener condition.

- (b) Since $S_{xx}(f_1, f_2) = N_0\pi\delta(f_1 - f_2)$, we have

$$R_{XX}(\tau) = R_{XX}(u + \tau, u) = S_{xx}(u + \tau, -u) = \pi N_0\delta(\tau).$$

Hence,

$$S_{XX}(\omega) = \pi N_0.$$

Once choice that satisfies $S_{XX}(\omega) = |L(\omega)|^2$ is $L(\omega) = \sqrt{\pi N_0}$, which gives $1/L(\omega) = 1/\sqrt{\pi N_0}$. This indicates both $L(\omega)$ and $1/L(\omega)$ are causal filters. Hence, $\mathbf{X}(t)$ is regular.

4. (10%) Given a set of orthonormal functions $\{\varphi_n(t)\}_{n=-\infty}^{\infty}$ over $[0, T]$, define

$$\hat{\mathbf{x}}(t) = \sum_{n=-\infty}^{\infty} \mathbf{c}_n \varphi_n(t) \quad \text{and} \quad \mathbf{c}_n = \int_0^T \mathbf{x}(t) \varphi_n^*(t) dt.$$

Prove that

$$E[|\hat{\mathbf{x}}(t) - \mathbf{x}(t)|^2] = 0 \quad \text{for } 0 < t < T$$

if

$$\int_0^T R_{xx}(t, s) \varphi_n(s) ds = \lambda_n \varphi_n(t)$$

for some **real** λ_n for every n , and $R_{xx}(t, t) = \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2$ for $0 \leq t < T$.

Hint: Compute $E[|\hat{\mathbf{x}}(t)|^2]$ and $E[\hat{\mathbf{x}}(t) \mathbf{x}^*(t)]$.

Solution. Observe that

$$\begin{aligned} E[|\hat{\mathbf{x}}(t)|^2] &= E \left[\left(\sum_{n=-\infty}^{\infty} \mathbf{c}_n \varphi_n(t) \right) \left(\sum_{m=-\infty}^{\infty} \mathbf{c}_m^* \varphi_m^*(t) \right) \right] \\ &= \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} E[\mathbf{c}_n \mathbf{c}_m^*] \varphi_n(t) \varphi_m^*(t) \\ &= \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} \lambda_n \delta[m - n] \varphi_n(t) \varphi_m^*(t) \quad (\text{because } \{\mathbf{c}_n\} \text{ orthogonal}) \\ &= \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2 \end{aligned}$$

and

$$\begin{aligned} E[\hat{\mathbf{x}}(t) \mathbf{x}^*(t)] &= E \left[\left(\sum_{n=-\infty}^{\infty} \mathbf{c}_n \varphi_n(t) \right) \mathbf{x}^*(t) \right] \\ &= \sum_{n=-\infty}^{\infty} E[\mathbf{c}_n \mathbf{x}^*(t)] \varphi_n(t) \\ &= \sum_{n=-\infty}^{\infty} E \left[\left(\int_0^T \mathbf{x}(s) \varphi_n^*(s) ds \right) \mathbf{x}^*(t) \right] \varphi_n(t) \\ &= \sum_{n=-\infty}^{\infty} \left(\int_0^T R_{xx}(t, s) \varphi_n(s) ds \right)^* \varphi_n(t) \\ &= \sum_{n=-\infty}^{\infty} \lambda_n^* \varphi_n^*(t) \varphi_n(t) \\ &= \sum_{n=-\infty}^{\infty} \lambda_n^* |\varphi_n(t)|^2 = \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2 \quad (\lambda_n \text{ real and non-negative}) \end{aligned}$$

Similarly,

$$E[\hat{\mathbf{x}}^*(t) \mathbf{x}(t)] = \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2.$$

Hence,

$$\begin{aligned}
 E[|\hat{\mathbf{x}}(t) - \mathbf{x}(t)|^2] &= E[|\hat{\mathbf{x}}(t)|^2] - E[\hat{\mathbf{x}}(t)\mathbf{x}^*(t)] - E[\hat{\mathbf{x}}^*(t)\mathbf{x}(t)] + E[|\mathbf{x}(t)|^2] \\
 &= \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2 - \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2 - \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2 + R_{xx}(t, t) \\
 &= R_{xx}(t, t) - \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2,
 \end{aligned}$$

which equals zero by

$$R_{xx}(t, t) = \sum_{n=-\infty}^{\infty} \lambda_n |\varphi_n(t)|^2.$$

□

5. (10%) Form the linear predictor of $\mathbf{x}[t]$ based on observation $\mathbf{z}[t]$ as

$$\hat{\mathbf{x}}[t] = \sum_{k=1}^{\infty} a_k \mathbf{z}[t - k],$$

and let the error process be $\mathbf{e}[t] = \mathbf{x}[t] - \hat{\mathbf{x}}[t]$, where $\{a_k\}_{k=1}^{\infty}$ satisfy

$$E\{\mathbf{e}[t]\mathbf{z}^*[t - m]\} = 0 \text{ for any } m \geq 1.$$

Prove that $\mathbf{e}[t]$ is a white process.

Solution. For $\tau > 0$,

$$E\{\mathbf{e}[t + \tau]\mathbf{e}^*[t]\} = \underbrace{E\{\mathbf{e}[t + \tau]\hat{\mathbf{x}}^*[t]\}}_{=0} - \sum_{m=1}^{\infty} a_m \underbrace{E\{\mathbf{e}[t + \tau]\mathbf{z}^*[t - m]\}}_{=0} = 0.$$

For $\tau < 0$,

$$E\{\mathbf{e}[t + \tau]\mathbf{e}^*[t]\} = E^*\{\mathbf{e}[t]\mathbf{e}^*[t + \tau]\} = 0.$$

Hence, $\mathbf{e}[t]$ is white.

6. (10%) Suppose $\mathbf{x}(t)$ is a WSS process, and define

$$\mathbf{X}(u) \triangleq \int_{-\infty}^{\infty} \mathbf{x}(t)e^{-jut} dt.$$

Let $R_{xx}(\tau)$ and $S_{xx}(\omega)$ be the autocorrelation function and power spectrum of $\mathbf{x}(t)$, respectively. Also let $R_{XX}(u_1, u_2)$ and $S_{XX}(\lambda_1, \lambda_2)$ be the autocorrelation function and two-dimensional power spectrum of $\mathbf{X}(t)$, respectively. Then,

$$R_{XX}(u_1, u_2) = S_{xx}(u_1, -u_2) \quad \text{and} \quad S_{XX}(\lambda_1, \lambda_2) = 4\pi^2 R_{xx}(-\lambda_1, \lambda_2).$$

Proof:

$$\begin{aligned}R_{XX}(u_1, u_2) &= E[\mathbf{X}(u_1)\mathbf{X}^*(u_2)] \\&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} E[\mathbf{x}(t_1)\mathbf{x}^*(t_2)]e^{-j(u_1t_1 - u_2t_2)} dt_1 dt_2 \\&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{xx}(t_1, t_2)e^{-j[u_1t_1 + (-u_2)t_2]} dt_1 dt_2 \\&= S_{xx}(u_1, -u_2)\end{aligned}$$

and

$$\begin{aligned}S_{XX}(\lambda_1, \lambda_2) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{XX}(u_1, u_2)e^{-j(\lambda_1u_1 + \lambda_2u_2)} du_1 du_2 \\&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} S_{xx}(u_1, -u_2)e^{-j(\lambda_1u_1 + \lambda_2u_2)} du_1 du_2 \\&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} S_{xx}(u_1, u'_2)e^{j[(-\lambda_1)u_1 + \lambda_2u'_2]} du_1 du'_2 \\&= 4\pi^2 R_{xx}(-\lambda_1, \lambda_2).\end{aligned}$$

□