

① Let $X(t)$ and $Y(t)$ be independent wide-sense stationary random processes, and define

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

(a) Show that $Z(t)$ is wide-sense stationary.

$$\begin{aligned} \mathbb{E}[Z(t)] &= \mathbb{E}[X(t)Y(t)] \\ &= \mathbb{E}[X(t)] \mathbb{E}[Y(t)] \\ &= \mu_X \mu_Y \quad (X, Y \text{ WSS}) \end{aligned}$$

$$\begin{aligned} R_Z(t_1, t_2) &\triangleq \mathbb{E}[Z(t_1)Z(t_2)] \\ &= \mathbb{E}[X(t_1)X(t_2)Y(t_1)Y(t_2)] \\ &= R_X(t_1 - t_2) R_Y(t_1 - t_2) \\ &= R_X(\tau) R_Y(\tau) \end{aligned}$$

$$\Rightarrow Z(t) \text{ WSS}$$

(b) Find $R_Z(\tau)$ and $S_Z(f)$ in terms of $R_X(\tau)$, $R_Y(\tau)$, and $S_X(f)$, $S_Y(f)$.

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

$$S_Z(f) = \mathcal{F}\{R_Z(\tau)\} = \mathcal{F}\{R_X(\tau) R_Y(\tau)\} = S_X(f) * S_Y(f)$$

(2) Let $\{X_n\}_{n=-\infty}^{+\infty}$ be a stationary Gaussian process with zero mean and autocorrelation function

$$R_x(k) = \begin{cases} 1, & \text{if } k=0 \\ 1/2, & \text{if } |k|=1 \\ 0, & \text{if } |k|>1 \end{cases}$$

(a) Compute the power spectral density of $\{X_n\}$.

$$\begin{aligned} S_x(f) &\triangleq \sum_{k=-\infty}^{+\infty} R_x(k) e^{-j2\pi f k} & (-\tfrac{1}{2} < f \leq \tfrac{1}{2}) \\ &= 1 + \tfrac{1}{2} e^{-j2\pi f} + \tfrac{1}{2} e^{j2\pi f} \\ &= 1 + \cos(2\pi f) \end{aligned}$$

(b) Compute $E[X_{n+1} | X_n, X_{n-1}]$

$$\text{Let } \underline{X} = \underline{X_{n+1}}, \quad \underline{Y} = \begin{bmatrix} X_n \\ X_{n-1} \end{bmatrix}$$

$$E[X|Y] = E[X] + \text{Cov}(X, Y) \text{Cov}^{-1}(Y) (Y - E[Y])$$

$$E[X] = [0]$$

$$E[Y] = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\text{Cov}(Y) = \begin{pmatrix} 1 & 1/2 \\ 1/2 & 1 \end{pmatrix}$$

$$\text{Cov}^{-1}(Y) = \frac{4}{3} \begin{pmatrix} 1 & -1/2 \\ -1/2 & 1 \end{pmatrix}$$

$$\text{Cov}(\bar{X}, \bar{Y}) = \begin{pmatrix} 1/2 & 0 \end{pmatrix}$$

$$\begin{aligned} \rightarrow \mathbb{E}[X|Y] &= \begin{pmatrix} 1/2 & 0 \end{pmatrix} \begin{pmatrix} 4/3 & -2/3 \\ -2/3 & 4/3 \end{pmatrix} \begin{pmatrix} x_n \\ x_{n-1} \end{pmatrix} \\ &= \begin{pmatrix} 2/3 & -1/3 \end{pmatrix} \begin{pmatrix} x_n \\ x_{n-1} \end{pmatrix} = \frac{2}{3}x_n - \frac{1}{3}x_{n-1} \end{aligned}$$

(c) Find the impulse response of a discrete-time LTI filter with the property that if the input to the filter is $\{x_n\}_{n=-\infty}^{+\infty}$, then the output y_n at time n is $\mathbb{E}[x_{n+1}|x_n, x_{n-1}]$.

$$y_n = \frac{2}{3}x_n - \frac{1}{3}x_{n-1} \text{ desired}$$

$$\begin{array}{c} x_n \rightarrow \boxed{h_n} \rightarrow \frac{2}{3}x_n - \frac{1}{3}x_{n-1} \end{array} \quad y_n = \sum_{l=-\infty}^{+\infty} h_l x_{n-l}$$

$$h_n = \frac{2}{3}\delta_n - \frac{1}{3}\delta_{n-1}$$

(d) Compute the transfer function of the filter found in (c).

$$H_n = \mathcal{F}\left\{\frac{2}{3}\delta_n - \frac{1}{3}\delta_{n-1}\right\} = \frac{2}{3} - \frac{1}{3}e^{-j2\pi f}$$

(e) Consider the process

$$Z_n = X_{n+1} - Y_n.$$

Show that this process is wide-sense stationary and determine its autocorrelation. What is the mean square error?

$$Z_n = X_{n+1} - \frac{2}{3}X_n + \frac{1}{3}X_{n-1}$$

$$E[Z_n] = E[X_{n+1}] - \frac{2}{3}E[X_n] + \frac{1}{3}E[X_n] = 0 - 0 + 0 = 0$$

$$R_Z(n_1, n_2) \triangleq E[Z_{n_1} Z_{n_2}]$$

$$= E\left[\left(X_{n_1+1} - \frac{2}{3}X_{n_1} + \frac{1}{3}X_{n_1-1}\right)\left(X_{n_2+1} - \frac{2}{3}X_{n_2} + \frac{1}{3}X_{n_2-1}\right)\right]$$

$$\begin{aligned} &= R_X(n_2 - n_1) - \frac{2}{3}R_X(n_2 - n_1 - 1) + \frac{1}{3}R_X(n_2 - n_1 - 2) - \frac{2}{3}R_X(n_2 - n_1 + 1) + \frac{1}{3}R_X(n_2 - n_1 + 2) \\ &\quad + \frac{4}{9}R_X(n_2 - n_1) - \frac{2}{9}R_X(n_2 - n_1 - 1) - \frac{2}{9}R_X(n_2 - n_1 + 1) \\ &\quad + \frac{1}{9}R_X(n_2 - n_1) \end{aligned}$$

$$\text{Let } k = n_2 - n_1$$

$$= \frac{14}{9}R_X(k) - \frac{8}{9}(R_X(k-1) + R_X(k+1)) + \frac{1}{3}(R_X(k-2) + R_X(k+2))$$

$\Rightarrow Z_n$ WSS

$$\begin{aligned} \text{MSE} = R_Z(0) &= \frac{14}{9}R_X(0) - \frac{8}{9}(R_X(-1) + R_X(1)) + \frac{1}{3}(R_X(-2) + R_X(2)) \\ &= \frac{14}{9} - \frac{8}{9} = \frac{6}{9} = \frac{2}{3} \end{aligned}$$

③ Let X_n be a zero-mean, first order autoregressive process with autocorrelation function

$$R_X(k) = \alpha^{|k|},$$

where $|\alpha| < 1$. Let

$$Y_n = X_n + \beta X_{n-1}$$

(a) Compute $R_{Y,X}(k)$, and $S_{Y,X}(f)$.

$$R_{Y,X}(k) \triangleq \mathbb{E}[X_n Y_{n+k}]$$

$$= \mathbb{E}[X_n (X_{n+k} + \beta X_{n+k-1})]$$

$$= \mathbb{E}[X_n X_{n+k} + \beta X_n X_{n+k-1}]$$

$$= \alpha^{|k|} + \alpha^{|k-1|} \beta$$

$$S_{Y,X}(f) = \mathcal{F}\{\alpha^{|k|} + \alpha^{|k-1|} \beta\}$$

$$= (1 + \beta e^{-j2\pi f}) \frac{1 - \alpha^2}{1 - 2\alpha \cos(2\pi f) + \alpha^2}$$

↙ $\alpha^{|k|}$

(b) Compute $S_Y(f)$, $R_Y(k)$, and $E[Y_n^2]$.

$$\begin{aligned} R_Y(k) &\triangleq E[Y_n Y_{n+k}] \\ &= E[(X_n + \beta X_{n-1})(X_{n+k} + \beta X_{n+k-1})] \\ &= E[X_n X_{n+k} + \beta X_n X_{n+k-1} + \beta X_{n+k} X_{n-1} + \beta^2 X_{n+k-1} X_{n-1}] \\ &= \alpha^{|k|} + \beta \alpha^{|k-1|} + \beta \alpha^{|k+1|} + \beta^2 \alpha^{|k|} \end{aligned}$$

$$\begin{aligned} S_Y(f) &= (1 + \beta^2 + \beta e^{-j2\pi f} + \beta e^{j2\pi f}) \frac{1 - \alpha^2}{1 - 2\alpha \cos(2\pi f) + \alpha^2} \\ &= (1 + \beta^2 + 2\beta \cos(2\pi f)) \frac{1 - \alpha^2}{1 - 2\alpha \cos(2\pi f) + \alpha^2} \end{aligned}$$

(c) For what value of β is $\{Y_n\}$ a white noise process?

Need to get rid of oscillating term.

If $\beta = -\alpha$, then

$S_Y(f) = 1 - \alpha^2$ is constant as desired.

$$\begin{aligned} E[Y_n^2] &= R_Y(0) = 1 + \beta\alpha + \beta\alpha + \beta^2 \\ &= 1 + \beta^2 + 2\beta\alpha \end{aligned}$$