Time Series Analysis

Quiz I-Solutions

1. Answer is given below:

part (a) Assuming e[k] to be white noise. Given the process

$$v[k] - 0.8v[k-1] + 0.15v[k-2] = e[k]$$
(1)

Taking Expectations both the sides

$$\sigma_{vv}[0] - 0.8\sigma_{vv}[1] + 0.15\sigma_{vv}[2] = \sigma_{ee}[0]$$
(2)

writing $\sigma_{vv}[k]$ as σ_k

$$\sigma_1 - 0.8\sigma_0 + 0.15\sigma_1 = 0 \tag{3}$$

$$\sigma_2 - 0.8\sigma_1 + 0.15\sigma_0 = 0 \tag{4}$$

(5)

(6)

solving the above equation gives

$$\rho_1 = 0.696 \quad where \quad \rho_1 = \frac{\sigma_1}{\sigma_0}, \quad ACF$$

 $\rho_2 = 0.406$, using these values we get $\sigma_0 = 1.985$

thus we get

$$\rho_0 = 1$$

 $\rho_1 = 0.696$

$$\rho_2 = 0.406$$

similarly we get ACF for lags 3 and 4

$$\rho_3 = 0.221$$

$$\rho_4 = 0.116$$

2. Answer is given below:

Part A:

$$f_{\Theta}(\theta) = \begin{cases} \frac{1}{2\pi} & -\pi < \theta < \pi \\ 0 & \text{otherwise} \end{cases}$$

$$\mu_{X}(t) = \frac{A}{2\pi} \int_{-\pi}^{\pi} \cos(\omega t + \theta) d\theta = 0$$
 (5.97)

Then

Setting $s = t + \tau$ in Eq. (5.7), we have

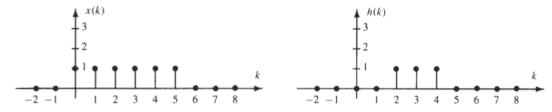
$$R_{XX}(t, t + \tau) = \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \cos(\omega t + \theta) \cos[\omega(t + \tau) + \theta)] d\theta$$

$$= \frac{A^2}{2\pi} \int_{-\pi}^{\pi} \frac{1}{2} \left[\cos \omega \tau + \cos(2\omega t + 2\theta + \omega \tau)\right] d\theta$$

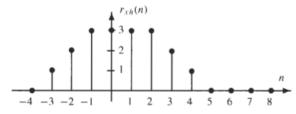
$$= \frac{A^2}{2} \cos \omega \tau$$
(5.98)

Since the mean of X(t) is a constant and the autocorrelation of X(t) is a function of time difference only, we conclude that X(t) is WSS.

Part B (a)



Denoting the correlation by $r_{vh}(n)$, it is clear that for n=0 the correlation is equal to 3. In fact, this will be the value of $r_{vh}(n)$ for $-1 \le n \le 2$. For n=3, x(k) and h(3+k) only overlap at two points, and $r_{vh}(3)=2$. Similarly, because x(k) and h(4+k) only overlap at one point, $r_{vh}(4)=1$. Finally, $r_{vh}(n)=0$ for n>4. Proceeding in a similar fashion for n<0, we find that $r_{vh}(-2)=2$, and $r_{vh}(-3)=1$. The correlation is shown in the figure below.



(b)

$$r_x(n) = x(n) \star x(n) = x(n) \star x(-n) = \sum_{k=-\infty}^{\infty} x(k)x(n+k)$$

In addition observe that $r_x(n)$ is an even function of n:

$$r_{\mathbf{x}}(-n) = \sum_{k=-\infty}^{\infty} x(k)x(-n+k) = \sum_{k'=-\infty}^{\infty} x(k'+n)x(k') = r_{\mathbf{x}}(n)$$

Therefore, it is only necessary to find the values of $r_x(n)$ for $n \ge 0$. For $n \ge 0$, we have

$$r_x(n) = \sum_{k=-\infty}^{\infty} \alpha^k u(k) \alpha^{n+k} u(n+k) = \alpha^n \sum_{k=0}^{\infty} \alpha^{2k} = \frac{1}{1-\alpha^2} \alpha^n \qquad n \ge 0$$

Using the symmetry of $r_v(n)$, we have, for n < 0,

$$r_{\rm r}(n) = \frac{1}{1 - \alpha^2} \alpha^{-n} \qquad n \le 0$$

Combining these two results together, we finally have

$$r_{\mathbf{v}}(n) = \frac{1}{1 - \alpha^2} \alpha^{|n|}$$

3) Be judicious. The answer is from lecture recordings and presentation.

4). Answer is given below:

Part A

A continuous-time sinusoid

$$x_a(t) = \cos(\Omega_0 t) = \cos(2\pi f_0 t)$$

that is sampled with a sampling frequency of f_s results in the discrete-time sequence

$$x(n) = x_a(nT_s) = \cos\left(2\pi \frac{f_0}{f_s}n\right)$$

However, note that for any integer k,

$$\cos\left(2\pi \frac{f_0}{f_s}n\right) = \cos\left(2\pi \frac{f_0 + kf_s}{f_s}n\right)$$

Therefore, any sinusoid with a frequency

$$f = f_0 + kf_s$$

will produce the same sequence when sampled with a sampling frequency f_s . With $x(n) = \cos(n\pi/8)$, we want

$$2\pi \frac{f_0}{f_s} = \frac{\pi}{8}$$

or

$$f_0 = \frac{1}{16} f_s = 625 \text{ Hz}$$

Therefore, two signals that produce the given sequence are

$$x_1(t) = \cos(1250\pi t)$$

and

$$x_2(t) = \cos(21250\pi t)$$

Part B

(a) The Nyquist rate is equal to twice the highest frequency in $x_a(t)$. If

$$y_a(t) = \frac{dx_a(t)}{dt}$$

then

$$Y_a(j\Omega) = j\Omega X_a(j\Omega)$$

Thus, if $X_a(j\Omega) = 0$ for $|\Omega| > \Omega_0$, the same will be true for $Y_a(j\Omega)$. Therefore, the Nyquist frequency is not changed by differentiation.

(b) The signal $y_a(t) = x_a(2t)$ is formed from $x_a(t)$ by compressing the time axis by a factor of 2. This results in an expansion of the frequency axis by a factor of 2. Specifically, note that

$$Y_a(j\Omega) = \int_{-\infty}^{\infty} y_a(t)e^{-j\Omega t} dt = \int_{-\infty}^{\infty} x_a(2t)e^{-j\Omega t} dt$$
$$= \int_{-\infty}^{\infty} \frac{1}{2}x_a(\tau)e^{-j\Omega\tau/2} d\tau = \frac{1}{2}X_a\left(\frac{j\Omega}{2}\right)$$

Consequently, if the Nyquist frequency for $x_a(t)$ is Ω_s , the Nyquist frequency for $y_a(t)$ will be $2\Omega_s$.

(c) When two signals are multiplied, their Fourier transforms are convolved. Therefore, if

$$y_a(t) = x_a^2(t)$$

then

$$Y_a(j\Omega) = \frac{1}{2\pi} X_a(j\Omega) * X_a(j\Omega)$$

Thus, the highest frequency in $y_a(t)$ will be twice that of $x_a(t)$, and the Nyquist frequency will be $2\Omega_s$.

(d) Modulating a signal by $\cos(\Omega_0 t)$ shifts the spectrum of $x_a(t)$ up and down by Ω_0 . Therefore, the Nyquist frequency for $y_a(t) = \cos(\Omega_0 t) x_a(t)$ will be $\Omega_s + 2\Omega_0$.

5. Answer is given below:

Part A:

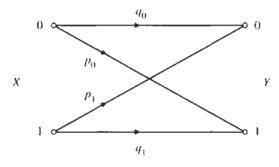


Fig. 1-15

(a) We note that

$$P(x_1) = 1 - P(x_0) = 1 - 0.5 = 0.5$$

$$P(y_0 \mid x_0) = q_0 = 1 - p_0 = 1 - 0.1 = 0.9$$

$$P(y_1 \mid x_1) = q_1 = 1 - p_1 = 1 - 0.2 = 0.8$$

$$P(y_0) = P(y_0 \mid x_0)P(x_0) + P(y_0 \mid x_1)P(x_1) = 0.9(0.5) + 0.2(0.5) = 0.55$$

$$P(y_1) = P(y_1 \mid x_0)P(x_0) + P(y_1 \mid x_1)P(x_1) = 0.1(0.5) + 0.8(0.5) = 0.45$$

(b) Using Bayes' rule (1.42), we have

$$P(x_0 | y_0) = \frac{P(x_0)P(y_0 | x_0)}{P(y_0)} = \frac{(0.5)(0.9)}{0.55} = 0.818$$

(c) Similarly,

$$P(x_1 \mid y_1) = \frac{P(x_1)P(y_1 \mid x_1)}{P(y_1)} = \frac{(0.5)(0.8)}{0.45} = 0.889$$

(d) The probability of error is

$$P_e = P(y_1 | x_0)P(x_0) + P(y_0 | x_1)P(x_1) = 0.1(0.5) + 0.2(0.5) = 0.15.$$

Part B:

(a) From the results of Prob. 1.52, we found that

$$P(X = 1) = 1 - P(X = 0) = 0.5$$

 $P(Y = 0 \mid X = 0) = 0.9$ $P(Y = 1 \mid X = 1) = 0.8$

Then by Eq. (1.41), we obtain

$$P(X = 0, Y = 0) = P(Y = 0 | X = 0)P(X = 0) = 0.9(0.5) = 0.45$$

 $P(X = 0, Y = 1) = P(Y = 1 | X = 0)P(X = 0) = 0.1(0.5) = 0.05$
 $P(X = 1, Y = 0) = P(Y = 0 | X = 1)P(X = 1) = 0.2(0.5) = 0.1$
 $P(X = 1, Y = 1) = P(Y = 1 | X = 1)P(X = 1) = 0.8(0.5) = 0.4$

Hence, the joint pmf's of (X, Y) are

$$p_{XY}(0, 0) = 0.45$$
 $p_{XY}(0, 1) = 0.05$
 $p_{XY}(1, 0) = 0.1$ $p_{XY}(1, 1) = 0.4$

(b) By Eq. (3.20), the marginal pmf's of X are

$$p_X(0) = \sum_{y_j} p_{XY}(0, y_j) = 0.45 + 0.05 = 0.5$$
$$p_X(1) = \sum_{y_j} p_{XY}(1, y_j) = 0.1 + 0.4 = 0.5$$

By Eq. (3.21), the marginal pmf's of Y are

$$p_{Y}(0) = \sum_{x_i} p_{XY}(x_i, 0) = 0.45 + 0.1 = 0.55$$
$$p_{Y}(1) = \sum_{x_i} p_{XY}(x_i, 1) = 0.05 + 0.4 = 0.45$$

(c) Now

$$p_X(0)p_Y(0) = 0.5(0.55) = 0.275 \neq p_{XY}(0, 0) = 0.45$$

Hence X and Y are not independent.