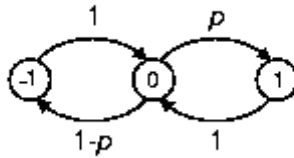


## Solutions to chapter 9

### Problem 9.5



a) The states of the system can be values of the voltage between switches. Hence there are three states, namely -1, 0, and +1. With this representation, the process is a random walk with reflecting boundaries. The Corresponding state transition matrix is

$$P = \begin{bmatrix} 0 & 1 & 0 \\ q & 0 & p \\ 0 & 1 & 0 \end{bmatrix}$$

b)

$$P^2 = \begin{bmatrix} q & 0 & p \\ 0 & 1 & 0 \\ q & 0 & p \end{bmatrix}$$

$$P^3 = \begin{bmatrix} 0 & 1 & 0 \\ q & 0 & p \\ 0 & 1 & 0 \end{bmatrix}$$

$$P^4 = \begin{bmatrix} q & 0 & p \\ 0 & 1 & 0 \\ q & 0 & p \end{bmatrix}$$

We know that a *recurrent state*  $i$  is a *periodic state* if  $p_{i,i}^{(n)} > 0$  for  $n > 1$ . And also state  $i$  is recurrent if  $P_{i,i}^{(n)} = 1$  for some  $n \geq 1$ . In above states, the only recurrent state is state 0 and two other states are transient. Therefore they can't be periodic and we have just to determine if state 0 is periodic and if so what its period is. For this process  $P_{0,0}^{(n)} > 0$  only if  $n$  is even. Therefore, the process is periodic with period 2.

c) For  $(i-1)t_s \leq t < it_s$  and  $i$  odd,  $x(t)=0$ . For  $it_s \leq t < (i+1)t_s$  and  $i$  odd,

$$\Pr(X(t) = 1) = p$$

$$\Pr(X(t) = -1) = q = 1 - p$$

### Problem 9.9

(a)

$$\mathbf{P} = \begin{bmatrix} 0.25 & 0.5 & 0.25 \\ 0.4 & 0.6 & 0 \\ 1 & 0 & 0 \end{bmatrix} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^{-1}.$$

Using MATLAB we found,

$$\mathbf{Q} = \begin{bmatrix} -0.4197 & 0.5774 & 0.2792 \\ 0.1570 & 0.5774 & -0.3981 \\ 0.8940 & 0.5774 & 0.8738 \end{bmatrix}, \quad \mathbf{\Lambda} = \begin{bmatrix} -0.4695 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0.3195 \end{bmatrix}.$$

The limiting form of the  $k$ -step transition probability matrix is then found as follows:

$$\lim_{k \rightarrow \infty} \mathbf{A}^k = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

$$\lim_{k \rightarrow \infty} \mathbf{P}^k = \mathbf{Q} \lim_{k \rightarrow \infty} \mathbf{A}^k \mathbf{Q}^{-1} = \begin{bmatrix} 0.4 & 0.5 & 0.1 \\ 0.4 & 0.5 & 0.1 \\ 0.4 & 0.5 & 0.1 \end{bmatrix}.$$

The steady-state distribution is then

$$\pi = [0.4 \quad 0.5 \quad 0.1].$$

(b) Using MATLAB to calculate  $\mathbf{P}^{100}$ , we get the same matrix found in part (a):

$$\mathbf{P}^{100} = \begin{bmatrix} 0.4 & 0.5 & 0.1 \\ 0.4 & 0.5 & 0.1 \\ 0.4 & 0.5 & 0.1 \end{bmatrix}.$$

$$\Rightarrow P_{1,3}^{(100)} = 0.1.$$

The interpretation of this result is that for all practical purposes, the process has reached steady state after 100 steps.

(c)

$$\pi(3) = \pi(0)\mathbf{P}^3 = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix} \begin{bmatrix} 0.25 & 0.5 & 0.25 \\ 0.4 & 0.6 & 0 \\ 1 & 0 & 0 \end{bmatrix}^3 = [0.4274 \quad 0.4808 \quad 0.0919]$$

$$\text{Pr}(\text{in state 3 after 3rd step}) = 0.0919.$$

### Problem 9.11

(a)

$$\mathbf{P} = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 & 5 & 6 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{5}{12} & 0 & \frac{7}{12} & 0 & 0 & 0 & 0 \\ 0 & \frac{5}{12} & 0 & \frac{7}{12} & 0 & 0 & 0 \\ 0 & 0 & \frac{5}{12} & \frac{1}{6} & \frac{5}{12} & 0 & 0 \\ 0 & 0 & 0 & \frac{7}{12} & 0 & \frac{5}{12} & 0 \\ 0 & 0 & 0 & 0 & \frac{7}{12} & 0 & \frac{5}{12} \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} \end{matrix}$$

(b)

$$\Pr(\text{loses in 3 tosses}) = \Pr(3 \rightarrow 2 \rightarrow 1 \rightarrow 0) = (5/12)^3 = 0.0723.$$

(c)

$$\Pr(\text{loses in 4 tosses}) = \Pr(3 \rightarrow 3 \rightarrow 2 \rightarrow 1 \rightarrow 0) = \frac{1}{6} \left(\frac{5}{12}\right)^3 = 0.0121.$$

$$\begin{aligned} \Pr(\text{loses in 5 tosses}) &= \Pr(3 \rightarrow 3 \rightarrow 3 \rightarrow 2 \rightarrow 1 \rightarrow 0) \\ &+ \Pr(3 \rightarrow 4 \rightarrow 3 \rightarrow 2 \rightarrow 1 \rightarrow 0) \\ &+ \Pr(3 \rightarrow 2 \rightarrow 3 \rightarrow 2 \rightarrow 1 \rightarrow 0) \\ &+ \Pr(3 \rightarrow 2 \rightarrow 1 \rightarrow 2 \rightarrow 1 \rightarrow 0) \\ &= \left(\frac{1}{6}\right)^2 \left(\frac{5}{12}\right)^3 + 3 \frac{7}{12} \left(\frac{5}{12}\right)^4 = 0.0548. \end{aligned}$$

$$\Pr(\text{loses in 5 or fewer tosses}) = 0.0723 + 0.0121 + 0.0548 = 0.1392.$$

We can verify this solution using MATLAB by noting that the probability of interest is found by finding the entry in the 4th row and 1st column of  $\mathbf{P}^5$ .

## Solutions to supplementary exercises

## Problem 1

For averaged periodograms or Bartlett's method we have:

$$\begin{aligned}\frac{\Delta f}{f_s} &= \frac{0.89}{L} && \text{Frequency Resolution} \\ \text{var}\left[\hat{P}_B(e^{j\omega})\right] &\approx \frac{1}{K} P_X^2(e^{j\omega}) && \text{Variance} \\ N &= KL\end{aligned}$$

Where L is length of samples per a time interval, K is number of time intervals and N is total number of samples. Therefore

$$v = \text{Variability} = \text{Relative Variance} = \frac{\text{var}\left[\hat{P}_B(e^{j\omega})\right]}{P_X^2(e^{j\omega})} \approx \frac{1}{K} = 0.01 \rightarrow K = 100$$

$$\frac{\Delta f}{f_s} = \frac{0.89}{L} \rightarrow \frac{1}{1000} = \frac{0.89}{L} \rightarrow L = 890$$

$$N = KL = 89000$$

$$t = \frac{N}{f_s} = \frac{89000}{1000} = 89 \text{ [s]}$$

## Problem 2

$$v = \frac{\text{var}[\hat{P}_B(e^{j\omega})]}{P_X^2(e^{j\omega})} \approx \frac{1}{K}$$

$$\frac{\Delta f}{f_s} = \frac{0.89}{L} \rightarrow$$

$$\left\{ \begin{array}{l} \frac{10}{10000} = \frac{0.89}{L_{Emilia}} \rightarrow L_{Emilia} = 890 \rightarrow t = \frac{N}{f_s} = \frac{N}{10000} = 10 \rightarrow N_{Emilia} = 10^5 \\ \frac{10}{100000} = \frac{0.89}{L_{Emil}} \rightarrow L_{Emil} = 8900 \rightarrow t = \frac{N}{f_s} = \frac{N}{100000} = 10 \rightarrow N_{Emil} = 10^6 \end{array} \right.$$

$$N = KL \rightarrow \left\{ \begin{array}{l} K_{Emilia} = \frac{10^5}{890} = \frac{10^4}{89} \rightarrow v_{Emilia} = 0.0089 \\ K_{Emil} = \frac{10^6}{8900} = \frac{10^4}{89} \rightarrow v_{Emil} = 0.0089 \end{array} \right.$$

As we can see there is no difference between normalized variances.

### Problem 3

It is easy to recognize the underlying method is Blackman-Tukey method with following properties:

$$P_{BT}(e^{j\omega}) = \sum_{-M}^M f_X(k) w(k) e^{-j\omega k}$$

$$\text{var}[P_{BT}(e^{j\omega})] \approx P_X(e^{j\omega}) \frac{1}{N} \sum_{-M}^M w^2(k) \rightarrow v \approx \frac{1}{N} \sum_{-M}^M w^2(k)$$

Therefore, we will have

$$N = 10000$$

$$M = 100$$

$$w(n) = \begin{cases} e^{-0.1|n|} & ; |n| \leq 100 \\ 0 & ; |n| > 100 \end{cases}$$

$$v \approx \frac{1}{N} \sum_{-M}^M w^2(k) = \frac{1}{10^4} \sum_{k=-100}^{100} e^{-0.2|k|} = \frac{1}{10^4} \left( 2 \sum_{k=0}^{100} e^{-0.2k} - 1 \right)$$

$$= \frac{1}{10^4} \left( 2 \left( \frac{e^{-20.2} - 1}{e^{-0.2} - 1} \right) - 1 \right)$$

$$= 10.0333 * 10^{-4}$$

## **Problem 4**

**a)**



$$\begin{aligned}
x[n] + 0.6x[n-1] - 0.2x[n-2] &= e[n] \\
x[n] &= -0.6x[n-1] + 0.2x[n-2] + e[n] \\
R[k] &= E[x[n-k]x[n]] = E[x[n-k](-0.6x[n-1] + 0.2x[n-2] + e[n])] \\
&= -0.6R[k-1] + 0.2R[k-2] + E[x[n-k]e[n]] \\
k=0: E[x[n-k]e[n]] &= E[(-0.6x[n-1] + 0.2x[n-2] + e[n])e[n]] = 1 \\
k > 0: E[x[n-k]e[n]] &= 0
\end{aligned}$$

$$\begin{aligned}
k=0: R[0] &= -0.6R[1] + 0.2R[2] + 1 \rightarrow R[0] + 0.6R[1] - 0.2R[2] = 1 \\
k=1: R[1] &= -0.6R[0] + 0.2R[1] \rightarrow 0.6R[0] - 0.8R[1] = 0 \\
k=2: R[2] &= -0.6R[1] + 0.2R[0] \rightarrow -0.2R[0] + 0.6R[1] + R[2] = 0 \\
|k| > 2: R[k] &= -0.6R[k-1] + 0.2R[k-2]
\end{aligned}$$

$$\begin{bmatrix} 1 & 0.6 & -0.2 \\ 0.6 & -0.8 & 0 \\ 0.2 & 0.6 & 1 \end{bmatrix} \begin{bmatrix} R[0] \\ R[1] \\ R[2] \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} R[0] \\ R[1] \\ R[2] \end{bmatrix} = \begin{bmatrix} 2.4 \\ -1.8 \\ 1.6 \end{bmatrix}$$

**Attention Please!**  $P[\text{output}] = H[z]H^*[\frac{1}{z^*}]P[\text{input}]$

$$\begin{aligned}
H[z] &= \frac{1}{1 + 0.6z^{-1} - 0.2z^{-2}} \\
P_x[z] &= H[z]H^*[\frac{1}{z^*}]P_E[z] \\
&= \left( \frac{1}{1 + 0.6z^{-1} - 0.2z^{-2}} \right) \left( \frac{1}{1 + 0.6z - 0.2z^2} \right) P_E[z] \\
&= \frac{1}{1.4 - 0.6(z + z^{-1}) - 0.2(z^2 + z^{-2})} \\
P_x(e^{j\omega}) &= \frac{1}{1.4 - 1.2\cos(\omega) - 0.4\cos(2\omega)}
\end{aligned}$$

b)

$$x[n] = e[n] + 0.8e[n-1] + 0.2e[n-2]$$

$$\begin{aligned} R[k] &= E[x[n-k]x[n]] = E\left[\left(e[n-k] + 0.8e[n-k-1] + 0.2e[n-k-2]\right)\left(e[n] + 0.8e[n-1] + 0.2e[n-2]\right)\right] \\ &= 0.2R_{EE}(k+2) + 0.96R_{EE}(k+1) + 1.68R_{EE}(k) + 0.96R_{EE}(k-1) + 0.2R_{EE}(k-2) \end{aligned}$$

$$k = 0: R[0] = 1.68R_{EE}[0] = 1.68$$

$$k = 1: R[1] = R[-1] = 0.96R_{EE}[0] = 0.96$$

$$k = 2: R[2] = R[-2] = 0.4R_{EE}[0] = 0.2$$

$$|k| > 2: R[k] = 0$$

$$\begin{aligned} P_X(e^{j\omega}) &= \sum_{n=-\infty}^{\infty} R_{XX}[n]e^{-j\omega n} = \sum_{n=-2}^2 R_{XX}[n]e^{-j\omega n} = 1.68 + 0.96(e^{j\omega} + e^{-j\omega}) + 0.2(e^{2j\omega} + e^{-2j\omega}) \\ &= 1.68 + 1.92\cos(\omega) + 0.2\cos(2\omega) \end{aligned}$$

or we could simply write

$$x[n] = e[n] + 0.8e[n-1] + 0.2e[n-2]$$

$$\rightarrow X[z] = E[z](1 + 0.8z^{-1} + 0.2z^{-2})$$

$$\rightarrow H[z] = 1 + 0.8z^{-1} + 0.2z^{-2}$$

$$P_X[z] = H[z]H^*\left[\frac{1}{z^*}\right]P_E[z]$$

$$= (1 + 0.8z^{-1} + 0.2z^{-2})(1 + 0.8z^1 + 0.2z^2)1$$

$$= 1.68 + 0.96(z + z^{-1}) + 0.2(z^2 + z^{-2})$$

$$P_X(e^{j\omega}) = 1.68 + 1.92\cos(\omega) + 0.4\cos(2\omega)$$

### Problem 6

$$x[n] + a_1 x[n-1] + a_2 x[n-2] = e[n]$$

Multiplying by  $x[n-k]$ ;  $k \geq 0$  and taking the expectation leads to the relation

$$r[k] + a_1 r[k-1] + a_2 r[k-2] = E[x[n-k]e[n]]$$

$$\begin{array}{l}
 k=0: \\
 k=1: \\
 k=2: \\
 k=3: \\
 k=4:
 \end{array}
 \begin{array}{l}
 \begin{bmatrix} r[0] & r[1] & r[2] \end{bmatrix} \\
 \begin{bmatrix} r[1] & r[0] & r[1] \end{bmatrix} \\
 \begin{bmatrix} r[2] & r[1] & r[0] \end{bmatrix} \\
 \begin{bmatrix} r[3] & r[2] & r[1] \end{bmatrix} \\
 \begin{bmatrix} r[4] & r[3] & r[2] \end{bmatrix}
 \end{array}
 \begin{array}{l}
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix}
 \end{array}
 = \begin{array}{l}
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}
 \end{array}$$

$$\begin{array}{l}
 \begin{bmatrix} 7.73 & 6.8 & 4.75 \\ 6.8 & 7.73 & 6.8 \\ 4.75 & 6.8 & 7.73 \\ 2.36 & 4.75 & 6.8 \\ 0.23 & 2.36 & 4.75 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}
 \end{array}
 = \begin{array}{l}
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \\
 \begin{bmatrix} \sigma_e^2 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}
 \end{array}$$

The above equation which has the form of  $Ax=b$ , can be solved by

$$\mathbf{x} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

$$\rightarrow \begin{bmatrix} 1 \\ a_1 \\ a_2 \end{bmatrix} = \sigma_e^2 \begin{bmatrix} 1.1061 \\ -1.6451 \\ 0.7655 \end{bmatrix} \rightarrow \begin{cases} \sigma_e^2 \approx 0.9041 \\ a_1 \approx -1.4873 \\ a_2 \approx 0.6921 \end{cases}$$

### Problem 8

$$d[n] - 0.8d[n-1] = e[n]$$

Multiplying by  $d[n-k]$ ;  $k \geq 0$  and taking the expectation leads to the relation

$$r[k] - 0.8r[k-1] = E[d[n-k]e[n]]$$

$$\begin{aligned} k=0: & \begin{bmatrix} 1 & -0.8 \end{bmatrix} \begin{bmatrix} r[0] \\ r[1] \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ k=1: & \begin{bmatrix} -0.8 & 1 \end{bmatrix} \begin{bmatrix} r[0] \\ r[1] \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \end{aligned}$$

$$\rightarrow \begin{cases} r[0] = 2.7778 \\ r[k] = 0.8r[k-1]; k \geq 1 \end{cases}$$

Also we have

$$D[z](1 + 0.5z^{-1}) + V[z] = X[z] \rightarrow d[n] + 0.5d[n-1] + v[n] = x[n]$$

If we estimate  $\hat{d}[n] = w_0x[n] + w_1x[n-1]$ , then the squared error can be estimated as follows

$$\begin{aligned} E[e^2] &= E[(d[n] - \hat{d}[n])^2] \\ &= E\left[ \left( d[n] - w_0(d[n] + 0.5d[n-1] + v[n]) + w_1(d[n-1] + 0.5d[n-2] + v[n-1]) \right)^2 \right] \\ &= w_0^2(1.25r[0] + r[1] + 0.1) + w_1^2(1.25r[0] + r[1] + 0.1) + w_0((w_1 - 2)r[0] + (2.5w_1 - 1)r[1] + w_1r[2]) \\ &\quad + w_1(-2r[1] - r[2]) + r[0] \\ &= 5.79445w_0^2 + 5.79445w_1^2 + 10.1111w_0w_1 - 7.7778w_0 - 6.2222w_1 + 2.7778 \end{aligned}$$

Taking the derivative of  $E[e^2]$  according to  $w_0$  and  $w_1$ , we get

$$\frac{\partial E[e^2]}{\partial w_0} = 11.5889w_0 + 10.1111w_1 - 7.7778 = 0$$

$$\frac{\partial E[e^2]}{\partial w_1} = 11.5889w_1 + 10.1111w_0 - 6.2222 = 0$$

$$\rightarrow \begin{bmatrix} 11.5889 & 10.1111 \\ 10.1111 & 11.5889 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} = \begin{bmatrix} 7.7778 \\ 6.2222 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} = \begin{bmatrix} 0.8489 \\ -0.2037 \end{bmatrix}$$