

Solutions to supplementary exercises

Problem 1

For averaged periodograms or Bartlett's method we have:

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

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

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

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

$$N = KL = 89000$$

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

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:

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

Therefore, we will have

$$N = 10000$$
$$M = 100$$
$$w(n) = \begin{cases} e^{-0.1|n|} & ; |n| \leq 100 \\ 0 & ; |n| > 100 \end{cases}$$
$$v \approx \frac{1}{N} \sum_{-M}^M w^2(k) = \frac{1}{10^4} \sum_{k=-100}^{100} e^{-0.2|k|} = \frac{1}{10^4} \left(2 \sum_{k=0}^{100} e^{-0.2k} - 1 \right)$$
$$= \frac{1}{10^4} \left(2 \left(\frac{e^{-0.2 \cdot 100} - 1}{e^{-0.2} - 1} \right) - 1 \right)$$
$$= 10.0333 * 10^{-4}$$

Problem 4

a)

$$x[n] + 0.6x[n-1] - 0.2x[n-2] = e[n]$$

$$x[n] = -0.6x[n-1] + 0.2x[n-2] + e[n]$$

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

$$k = 0: E[x[n-k]e[n]] = E[(-0.6x[n-1] + 0.2x[n-2] + e[n])e[n]] = 1$$

$$k > 0: E[x[n-k]e[n]] = 0$$

$$k = 0: R[0] = -0.6R[1] + 0.2R[2] + 1 \rightarrow R[0] + 0.6R[1] - 0.2R[2] = 1$$

$$k = 1: R[1] = -0.6R[0] + 0.2R[1] \rightarrow 0.6R[0] - 0.8R[1] = 0$$

$$k = 2: R[2] = -0.6R[1] + 0.2R[0] \rightarrow -0.2R[0] + 0.6R[1] + R[2] = 0$$

$$|k| > 2: R[k] = -0.6R[k-1] + 0.2R[k-2]$$

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

Attention Please!

$$P[\text{output}] = H(z) = [W^*]^{-1} P[\text{input}]$$

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

b)

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

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

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

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

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

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

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

or we could simply write

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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