

SOLUTIONS

E 100      **Digital Signalbehandling,**      2E1340

Final Examination 2006–12–19,    08.00–13.00

1. (a) Total effective noise at the input of the lattice is  $e(n) = e_1(n) + e_2(n) - k_2 e_3(n)$  so the variance is

$$\sigma_e^2 = \sigma^2 + \sigma^2 + k_2^2 \sigma^2 = (2 + k_2^2) \frac{2^{-2(b-1)}}{12}$$

The transfer function of the lattice filter

$$H(z) = \frac{1}{1 + (k_1 k_2 + k_3)z^{-1} + k_2 z^{-2}} = \frac{1}{(1 - 5/9 z^{-1})(1 - 3/5 z^{-1})} = \frac{27/2}{1 - 3/5 z^{-1}} - \frac{25/2}{1 - 5/9 z^{-1}}$$

and

$$h(n) = \left[ \frac{27}{2} \left(\frac{3}{5}\right)^n - \frac{25}{2} \left(\frac{5}{9}\right)^n \right] u(n)$$

so the total round-off noise power at the output of the lattice is

$$P_Q = \sigma_e^2 \sum_{n=-\infty}^{\infty} |h(n)|^2 \approx 4.52 \sigma_e^2 = 0.0031$$

- (b) This lattice is stable, which follows directly from the pole locations found above. Using the Schur-Cohn Stability test (the reflection coefficients are  $[-0.8667, 0.3333]$ ) is overkill here.

2. The general conditions for perfect reconstruction with delay  $L$  are

$$G_0(z)H_0(-z) + G_1(z)H_1(-z) = 0 \quad (1)$$

$$G_0(z)H_0(z) + G_1(z)H_1(z) = 2z^{-L} \quad (2)$$

Inserting the given filters in (1) gives

$$a + (b - a)z^{-1} - bz^{-2} + c + (c + d)z^{-1} + dz^{-2} = 0$$

i.e.

$$\begin{cases} a + c = 0 \\ b - a + c + d = 0 \\ d - b = 0 \end{cases}$$

which gives  $a = b = -c = d$ . Inserting these into (2) gives

$$2z^{-L} = a + (a + a)z^{-1} + az^{-2} - a + (a + a)z^{-1} - az^{-2} = 4az^{-1}$$

so the only solution is  $L = 1$  and  $a = b = -c = d = 1/2$ .

3. a) The only common spectral estimate that we can derive from  $|X(k)|$  is the periodogram,  $\hat{P}^{\text{Periodogram}}(f) = |X(k)|^2/N$ , so  $\hat{P}^{\text{Periodogram}}(f) \leq P_{\max}(f)$  corresponds to  $|X(k)| \leq \sqrt{NP_{\max}}$ . The cutoff frequency  $f = 0.2$  corresponds to  $k = 0.2 \cdot N \approx 410$ . To summarize, we should verify if

$$|X(k)| \leq \begin{cases} \sqrt{2048 \cdot 1} \approx 45 & 0 \leq k \leq 410 \\ \sqrt{2048 \cdot 0.1} \approx 14 & 410 < k \leq 1024 \end{cases}$$

This is clearly not the case in Figure 4. However, since the periodogram has a large variance, it may very well happen that the true spectrum is below the mask. Therefore, a reasonable answer is “*The figure does not provide sufficient information.*”.

- b) Here we have even less information. Since  $|\text{Re}[X(k)]| \leq |X(k)|$ , we cannot say much, even if  $\text{Re}[X(k)]^2/N$  happens to be below  $P_{\max}(f)$ . Again, the most reasonable answer is “*The figure does not provide sufficient information.*”.

- c) If we assume that the signal is Gaussian distributed, we know that the variance of the Bartlett spectral estimate is about  $P^2(f)/L$ , and here, the number of segments is  $L = 16$ , so the standard deviation is 1/4 of the true value. Since all but the highest peaks (which we hope are above and not below the true spectrum) are clearly below 80% of the mask, we can claim “*Yes, the power spectral density of the signal is below the mask, with high probability.*”

4. The MSE is given by

$$J = E[(s(n) - \hat{s}(n))^2] = E[(s(n) - ay(n) - by(n-1))^2]$$

To minimize  $J$  we should find  $a$  and  $b$  such that  $\frac{\partial J}{\partial a} = 0$  and  $\frac{\partial J}{\partial b} = 0$ .

This yields

$$\begin{bmatrix} r_y(0) & r_y(1) \\ r_y(1) & r_y(0) \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} r_{sy}(0) \\ r_{sy}(1) \end{bmatrix}$$

Since  $e(n)$  and  $s(n)$  are uncorrelated, we have

$$\begin{aligned} P_y(z) &= H(z)H(z^{-1})P_s(z) + P_e(z) = \sigma_s^2(1 + z^{-1})(1 + z) + \sigma_e^2 \\ &= \sigma_s^2 z^{-1} + 2\sigma_s^2 + \sigma_e^2 + \sigma_s^2 z \end{aligned}$$

$$P_{sy}(z) = H(z)P_s(z) = \sigma_s^2(1 + z^{-1})$$

Thus, we have

$$\begin{bmatrix} 2\sigma_s^2 + \sigma_e^2 & \sigma_s^2 \\ \sigma_s^2 & 2\sigma_s^2 + \sigma_e^2 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} \sigma_s^2 \\ 0 \end{bmatrix}$$

Therefore  $a = \frac{\sigma_s^2(2\sigma_s^2 + \sigma_e^2)}{(\sigma_s^2 + \sigma_e^2)(3\sigma_s^2 + \sigma_e^2)}$  and  $b = \frac{-\sigma_s^4}{(\sigma_s^2 + \sigma_e^2)(3\sigma_s^2 + \sigma_e^2)}$ .

When  $\sigma_e^2 \ll \sigma_s^2$ ,  $a = \frac{2}{3}$  and  $b = \frac{-1}{3}$ . They are independent of  $\sigma_s^2$ .

5. Since the frequency axis will be scaled by 4/3, the highest frequency that can be represented without aliasing will be  $f_{\max} = 3/4$  and we will see below that it is possible to preserve all frequencies below  $f_{\max} = 3/4$ .

We can use the ordinary structure of rate conversion, as shown in Fig. 1, but use a filter of the form

$$H(f) = \begin{cases} C & 0 \leq f < f_1 \\ 0 & \text{otherwise in } [0, 1[ \end{cases}$$

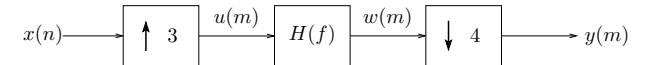


Figure 1: Implementation of rate conversion.

As can be seen in the following example, we can then use  $f_1 = 1/4$ , corresponding to  $f_{\max} = 3/4$  and similarly to the ordinary rate conversion structure, it makes sense to use  $C = 3$ , corresponding to a scaling by  $3/4$  of  $X(f)$  (the extra factor  $3/4$  that you see in the example, appears since the low pass filtering cuts the frequencies of the signal that have highest power in this specific example). Based on

$$U(f) = X(3f)$$

$$W(f) = H(f)U(f)$$

$$Y(f) = \frac{1}{4} \sum_{k=0}^3 W\left(\frac{f-k}{4}\right)$$

we can easily draw Figure 2.

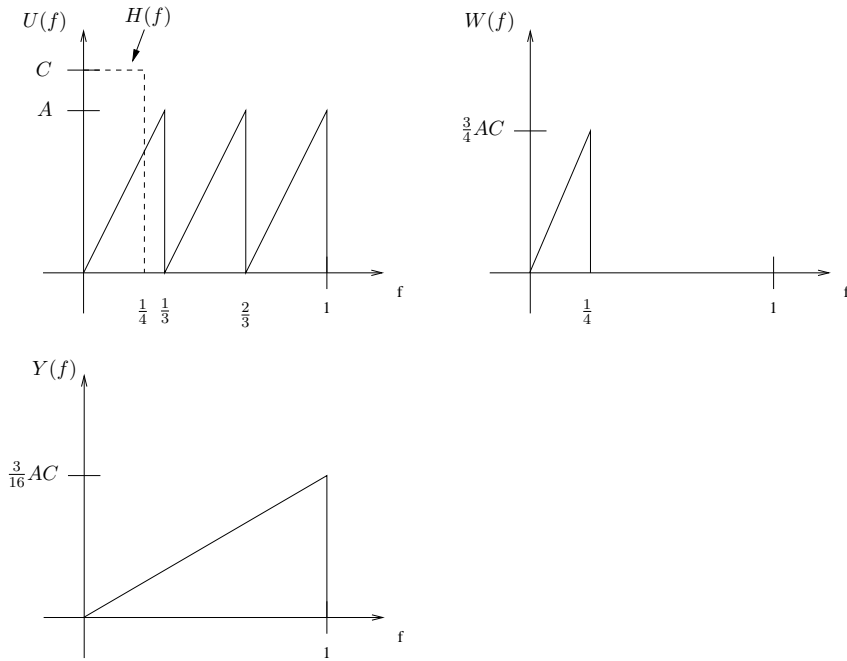


Figure 2: Fourier transforms of the signals.