

KTH, SIGNAL PROCESSING LAB  
SCHOOL OF ELECTRICAL ENGINEERING

**Digital Signal Processing**      EQ2300/ 2E1340

Final Examination 2011–12–20, 14.00–19.00

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- Literature:**
- Hayes: *Statistical Digital Signal Processing and Modeling*  
or  
Proakis, Manolakis: *Digital Signal Processing*
  - Bengtsson: *Complementary Reading in Digital Signal Processing*
  - Bengtsson and Jaldén: *Summary slides*
  - *Beta – Mathematics Handbook*
  - *Collection of Formulas in Signal Processing, KTH*
  - Unprogrammed pocket calculator.

- Notice:**
- Answer in English or Swedish.
  - At most one problem should be treated per page.
  - Answers without motivation/justification carry no rewards.
  - Write your name and *personnummer* on each page.
  - Write the number of solution pages on the cover page.

The exam consists of five problems with a maximum of 10 points each.  
For a passing grade, 24 points are normally required.

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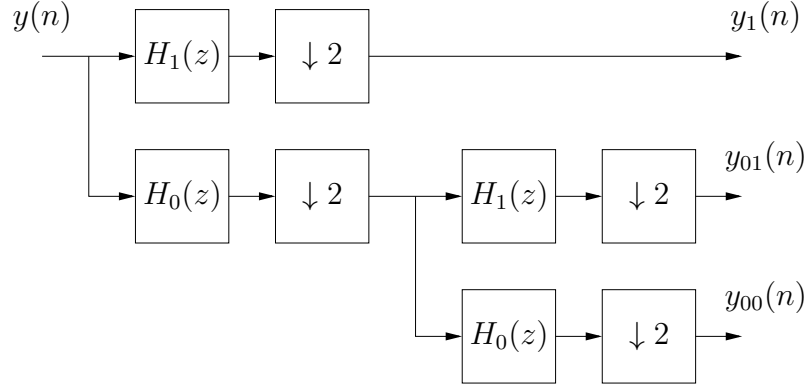
**Results:** Will be reported within three working weeks on “My pages”.

**Solutions:** Will be available on the course homepage after the exam.

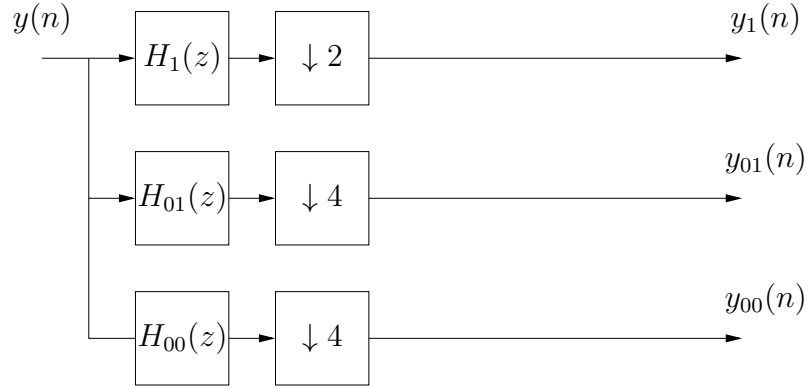
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***Good luck!***

1. In project 2 you used the cascaded 2-step analysis filter bank below to split the input signal  $y(n)$  into three different frequency bands.

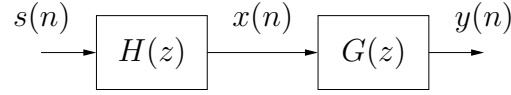


The filters in the filter bank were given by  $H_0(z) = -\frac{1}{8} + \frac{1}{4}z^{-1} + \frac{3}{4}z^{-2} + \frac{1}{4}z^{-3} - \frac{1}{8}z^{-4}$  and  $H_1(z) = \frac{1}{2} - z^{-1} + \frac{1}{2}z^{-2}$ . An alternative implementation of the analysis filter bank is as a 3 branch 1-step filter bank as given below, where  $H_{01}(z)$  and  $H_{00}(z)$  are suitably chosen filters.



- Obtain the filter  $H_{01}(z)$  so that the two implementations are equivalent, i.e., so that  $y_{01}(n)$  is the same in both cases. (4p)
- Considering direct implementations of all filters, how many multiplications are required to obtain a single sample of  $y_{01}(n)$  in each of the two implementations? Which implementation is most efficient? (Only consider the multiplications required to obtain  $y_{01}(n)$ , not  $y_0(n)$  and  $y_{00}(n)$ ) (3p)
- Show how to implement the filter  $H_{01}(n)$  as a polyphase filter. Specify all polyphase component filters and compute the number of multiplications required to compute a single sample of  $y_{01}(n)$ . How do the number of multiplications per sample compare to the direct implementations? (3p)

2. Consider the following communication link: The transmitter sends signal  $s(n)$ , which is distorted by the channel  $H(z)$  during transmission. The received signal is  $x(n)$ , and the data sequence after equalizer  $G(z)$  is denoted as  $y(n)$ .



Assume we know that the channel distortion can be modeled as

$$H(z) = 1 + 0.5z^{-1}.$$

- (a) Design a causal equalizer  $g(n)$  so that  $y(n) = s(n)$ . (1p)
- (b) Assume we have received  $N = 200$  samples at the receiver. The FIR filter  $\hat{g}(n)$  is an 20-tap FIR filter which approximates  $g(n)$  with its first 20 taps (i.e.,  $\hat{g}(n)$  is equal to  $g(n)$  for  $n = 0, \dots, 19$ , and is zero elsewhere). Now we want to filter  $x(n)$  with  $\hat{g}(n)$ .
  - i. A  $X$ -point DFT is performed on  $x(n)$  and  $\hat{g}(n)$  respectively, then the two DFTs are multiplied and an inverse is performed to compute their circular convolution. How do you need to choose  $X$  to make sure the circular convolution is the same as the linear convolution of  $x(n)$  and  $\hat{g}(n)$ ? (1p)
  - ii. If we instead filter  $x(n)$  with  $\hat{g}(n)$  using the overlap-save method. What is then the best choice of the FFT block-length  $M$ ? (Consider only the case  $M = 2^p$ , where  $p$  is a positive integer). (3p)
- (c) Now we want to estimate the power spectrum of the data sequence  $y(n)$ ,  $n = 0, \dots, 199$ . The estimation should fulfill the following requirements:
  - i. 3 dB resolution bandwidth (BW)  $\Delta f \leq 0.013$ , where  $f \in [-\frac{1}{2}, \frac{1}{2}]$  is the normalized frequency.
  - ii. Sidelobe level  $\leq -20\text{dB}$
  - iii. The variance  $\text{Var}(\hat{P}(f)) \leq \frac{1}{2}P^2(f)$ , where  $P(f)$  is the true power spectrum.
 How do you choose your windowing and estimation method? (5p)

Window	Sidelobe Level (dB)	3 dB BW $\Delta f$
Rectangular	-13	$0.89/N$
Bartlett	-27	$1.28/N$
Hanning	-32	$1.44/N$
Hamming	-43	$1.30/N$
Blackman	-58	$1.68/N$

Table 1: Properties of a few commonly used windows. Each window is assumed to be of length  $N$ .

3. The autocorrelation sequence of an autoregressive (AR) process  $x(n)$  is given by

$$r_x(k) = \alpha^{|k|}, \quad |\alpha| < 1.$$

(a) Using the Levinson-Durbin recursion, determine the order of the AR model. Find the transfer function of an all-pole filter which is applied to unit variance white noise to generate this AR process, i.e.,

$$H(z) = \frac{b(0)}{A_p(z)} = \frac{b(0)}{1 + \sum_{k=1}^p a_p(k)z^{-k}}$$

where  $b(0) = r_x(0) + \sum_{l=1}^p a_p(l)r_x(l)$ . (4p)

(b) Suppose that the available measurements of  $x(n)$  are noisy and the observed process  $y(n)$  is

$$y(n) = x(n) + w(n)$$

where  $w(n)$  is the white noise with variance  $\sigma_w^2$  and uncorrelated with  $x(n)$ . It is here known that  $y(n)$  is a second-order AR process. Find the AR parameters for  $y(n)$ , i.e.,  $\mathbf{a}_2 = [1 \ a_2(1) \ a_2(2)]^T$ . (4p)

(c) How will the AR parameters change if  $\sigma_w^2 \rightarrow 0$ ? Interpret your answer. (2p)

4. Consider an AR(1) process  $d(n)$ , described by the relation  $d(n) = 0.3d(n-1) + e(n)$ , where  $e(n)$  is white noise with variance  $\sigma_e^2 = 1$ .

Assume that  $d(n)$  goes through a filter with impulse response  $h(n) = \delta(n) - \delta(n-1)$  and that the output, say  $f(n)$ , is corrupted by an additive noise source  $w(n)$  producing the signal

$$x(n) = f(n) + w(n) = d(n) * h(n) + w(n)$$

The noise source  $w(n)$  is assumed white, zero-mean with variance  $\sigma_w^2 = 1$  (independent of  $e(n)$  and  $d(n)$ ). We would like to build a one-step predictor for  $d(n)$ , of the form

$$\hat{d}(n+1) = c + w_0x(n) + w_1x(n-1)$$

so as to minimize the the MSE

$$J = \mathbb{E} \left\{ \left[ \hat{d}(n+1) - d(n+1) \right]^2 \right\}$$

with respect to  $c, w_0$  and  $w_1$ . (Note that  $c$  is a constant, but we need to design it as well).

a) Find the autocorrelation function (ACF) of  $x(n)$  (3p)

b) Use the previous calculation to find the optimal values for  $c, w_0, w_1$ , i.e., the values that minimize  $J$  (7p)

5. Consider a continuous-time signal  $x_a(t)$  with power spectrum

$$X_a(F) = \begin{cases} 1 - |F|/F_N, & |F| \leq F_N \\ 0, & |F| > F_N \end{cases}$$

Suppose that we oversample  $x_a(t)$  by a factor of  $M$ , i.e., assume that we sample  $x_a(t)$  with frequency  $F_s = 2F_N M$  (which leads to a sampling period  $T = \frac{1}{2F_N M}$ ). Finally, suppose that we pass the resulting discrete-time signal  $x(n) = x_a(nT)$  through a quantizer  $Q$  and a proper decimator, as shown in Figure 1.

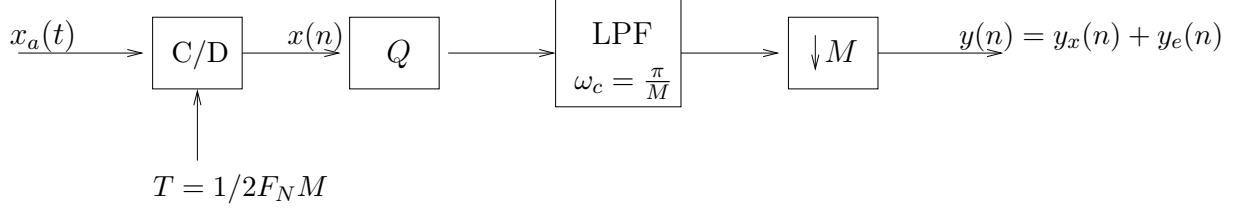


Figure 1: Oversampling followed by direct quantization and decimation

The continuous-to-discrete time converter (C/D) is ideal, i.e., implemented with infinite precision. The quantizer  $Q$  is scalar and uniform, with the quantization step  $\Delta$  sufficiently small in order for the usual assumptions regarding the quantization error  $e(n)$  to be valid. The filter LPF is an ideal lowpass filter with frequency response

$$H_D(e^{j\omega}) = \begin{cases} 1, & |\omega| \leq \pi/M \\ 0, & \pi/M < |\omega| \leq \pi \end{cases}$$

- Calculate the signal-to-quantization noise ratio (SQNR) at the output  $y(n)$  (5p)
- Assume that we add a feedback loop around the quantizer, as shown in Figure 2 below, where we choose  $H(z) = 1/(1 - z^{-1})$ . Calculate the new SQNR at the output  $y(n)$ . Assume that  $M$  is sufficiently large so that  $\sin\left(\frac{\omega}{2M}\right) \approx \frac{\omega}{2M}$  to simplify the calculation of the integrals. Hint: You may also find the following trigonometric identity useful:  $2\sin^2(\omega/2) = 1 - \cos(\omega)$  (3p)
- What have we accomplished by adding the feedback loop in Figure 2? If you cannot solve part (b), try to answer part (c) by reasoning (2p)

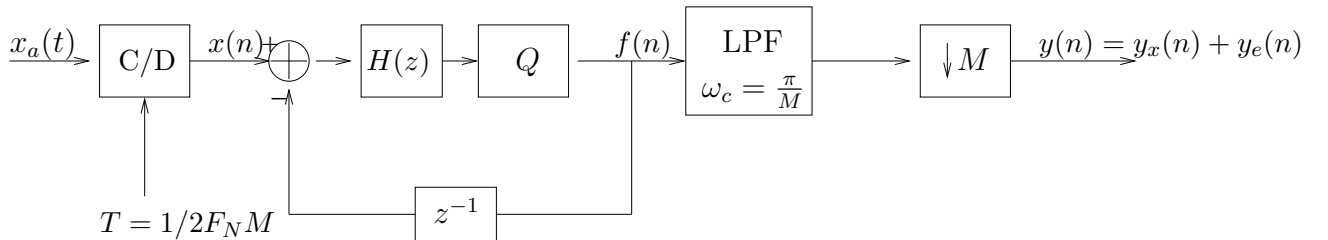


Figure 2: Adding a feedback loop around the quantizer