

# Homework 10

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## ADSI Problem 6.2: The autocorrelation of an AR(1) process (proof)

Consider an AR(1) process given by  $y(n) = -ay(n-1) + x(n)$   
with  $-1 < a < 1$  and  $x(n) \sim WN(0, \sigma_x^2)$ .

1. Show that the autocorrelation of the AR(1) process is given by

$$r_{yy}(l) = \frac{\sigma_x^2}{1-a^2}(-a)^{|l|}$$

Hint: Use equation (13.138) and (13.140).

### ADSI Problem 6.3: Wiener FIR Filtering, minimum square error

Consider a signal  $x(n) = s(n) + w(n)$  where  $s(n)$  is an AR(1) process that satisfies the difference equation

$$s(n) = 0.8s(n-1) + v(n)$$

where  $\{v(n)\}$  is a white noise sequence with variance  $\sigma_v^2 = 0.49$  and  $\{w(n)\}$  is a white noise sequence with variance  $\sigma_w^2 = 1$ . The processes  $\{v(n)\}$  and  $\{w(n)\}$  are uncorrelated.

#### 1) Determine the autocorrelation sequences for $x(n)$ and $s(n)$

Determine the autocorrelation sequences  $\{r_s(l)\}$  and  $\{r_x(l)\}$ .

#### 2) Design a Wiener filter of length $M=2$ to estimate $s(n)$

Design a Wiener filter of length  $M = 2$  to estimate  $\{s(n)\}$ .

#### 3) Determine the minimum mean square error for $M=2$

### ADSI Problem 6.4: Linear interpolation, estimate missing samples

Sometimes it happens that a datapoint is missing from some signal acquisition due to sensor failure, transmission errors etc. Assume that we have a long stationary sequence  $\{x[n]\}_{n=0}^{N-1}$  where the  $j$ 'th sample is missing i.e.

$$\{x[n]\} = \{x[0], x[1], \dots, x[j-1], x[j+1], x[j+2], \dots, x[N-2], x[N-1]\}$$

We want estimate the missing datapoint as a linear combination of the two neighbouring samples

$$\hat{x}[j] = c_1 x[j-1] + c_2 x[j+1]$$

1. Use our standard mean square error approach to derive equations for  $c_1$  and  $c_2$  based on the autocorrelation  $r_{xx}(l)$ .

## 1) Use mean square error to derive coefficients based on the ACRS

## ADSI Problem 6.5: Levinson-Durbin by Hand

*The aim of this problem is to get a finger-tip feeling of the flow of the Levinson-Durbin recursion. Assume that the following autocorrelation function values have been determined from an unknown random process  $\{x(n)\}$*

$l$	$r_{xx}(l)$
0	5
1	4
2	3
3	2
4	1

Work through the Levinson-Durbin recursion by hand and find the optimum linear predictors for  $m=1, 2$  and  $3$  as well as the corresponding minimum mean square errors  $J_m$ 's and reflection coefficients  $k_m$ 's.

## 1) Work through Levinson-Durbin by hand to find the optimum linear predictors for $m=1,2,3$

## 2) Find the corresponding minimum mean square errors

## 3) Find the corresponding reflection coefficients

## ADSI Problem 6.6: Levinson-Durbin and linear prediction

The autocorrelation function of an AR(2) process with two complex conjugated poles at  $p = r_p e^{\pm j\omega_p}$  can be calculated analytically and is given by

$$r_{xx}(l) = \frac{r_p^l (\sin((l+1)\omega_p) - r_p^2 \sin((l-1)\omega_p))}{(1 - r_p^2) \sin(\omega_p) (1 - 2r_p^2 \cos(2\omega_p) + r_p^4)} \quad \text{for } l \geq 0$$

Assume that  $r_p = 0.9$  and  $\omega_p = \pi/16$ .

### 1) Plot the autocorrelation function.

### 2) Compute reflection coefficients for m'th order optimum linear predictors

Use the above autocorrelation function and the Levinson-Durbin recursion to calculate reflection coefficients and minimum mean square errors for  $m$ 'th order optimum linear predictors for  $m = 1$  to  $m = 6$ . Are the results in agreement with your anticipations and Eq. (14.149)?

$$J_{m+1} = J_m + \beta_{m+1} k_{m+1} = (1 - k_{m+1}^2) J_m. \quad (14.149)$$

## ADSI Problem 6.7: Linear prediction

This problem addresses linear prediction on a simple harmonic signal where the results can be compared with our intuitive understanding.

Let a discrete time signal be given by

$$x(n) = \sqrt{2} \sin(\omega_0 n + \phi)$$

Where the phase  $\phi$  is uniformly distributed between 0 and  $2\pi$ .

### 1) Determine the autocorrelation function for $x(n)$

### 2) Determine the 2nd order forward linear prediction filter

Write down the normal equation for the forward linear prediction filter and determine the filter coefficients for a 2nd order filter. For mathematical convenience we set  $\omega_0 = \frac{\pi}{3}$ .

### 3) Find the system function for the filter and locate the zeros

Find the system function  $H(z)$  for the filter and locate the zeros.

### 4) Determine the frequency response, plot it and comment on the result

Determine the frequency response  $H(\omega)$ . Plot it and comment on the result.

### 5) Calculate the prediction error

## ADSI Problem 6.8: Autocorrelation function and linear prediction

*Assume that for a given sequence of data  $\{x(n)\}$  the autocorrelation function has been calculated and used to solve the normal equations so that the optimum  $p$ 'th order linear predictor was found. Now, an amplifier is placed in the signal chain so that the signal is  $\{c \cdot x(n)\}$ . How does the autocorrelation function and the linear predictor change?*

# Functions