

ASSIGNMENT #6

Due: Wednesday, March 28, 2001 in class

1. The technique used in class to derive the optimum coefficients for LPC is just one example of the least squares approach to approximation/prediction. The following problem illustrates other uses of the least squares approach.

Suppose we wish to approximate the waveform $x(t) = e^{-1.5t}$ over the interval $0 \leq t \leq 3$ by the function $\hat{x}(t) = a_0 + a_1t + a_2t^2$.

- a) Find the values for the coefficients that minimize

$$E = \int_0^3 [x(t) - \hat{x}(t)]^2 dt$$

- b) Compare your coefficients with the first 3 terms of a Taylor series for $e^{-1.5t}$ expanded about the point $t = 0$.
- c) Using matlab, plot on the same axes $x(t)$, $\hat{x}(t)$, and the Taylor series approximation.
- d) Compare the mean-squared error E for $\hat{x}(t)$ and the 3 term Taylor series.
2. Show that if the predictor coefficients satisfy

$$R_n(l) = \sum_{k=1}^p R_n(k-l), \quad 1 \leq l \leq p,$$

then the prediction error simplifies to

$$E_n = R_n(0) - \sum_{k=1}^p R_n(k).$$

3. Consider the signal

$$x(n) = \{2^{-n} + 4^{-n}\}u(n)$$

- a) Calculate the autocorrelation function

$$R(l) = \sum_{n=-\infty}^{\infty} x(n)x(n-l)$$

for this signal.

- b) Use the Levinson-Durbin recursion to find the coefficient $K_1^{(1)}$ of a first order predictor:

$$\hat{x}(n) = K_1^{(1)} x(n-1).$$

- c) Calculate the error $E^{(1)}$ using the formula

$$E^{(1)} = (1 - K_1^2) E^{(0)}$$

- d) Continue the Levinson-Durbin recursion to find the coefficients $K_1^{(2)}$ and $K_2^{(2)}$ for a second order predictor:

$$\hat{x}(n) = K_1^{(2)} x(n-1) + K_2^{(2)} x(n-2).$$

- e) Calculate $E^{(2)}$ using the formula

$$E^{(2)} = E^{(1)} (1 - K_2^2).$$

- f) Calculate the prediction sequence, $\hat{x}(n)$, $-\infty < n < \infty$ for the second order predictor.

- g) Calculate the error sequence

$$e(n) = x(n) - \hat{x}(n), \quad -\infty < n < \infty$$

for the second order predictor.

- h) Using your answer to part g), calculate the mean-squared error

$$E^{(2)} = \sum_{n=-\infty}^{\infty} [x(n) - \hat{x}(n)]^2$$

- i) Compare your answers to parts e) and h).

4. Given the discrete-time signal

$$x(n) = \{2^{-n} + 4^{-n}\} u(n)$$

we consider predicting $x(n)$ with a second-order predictor ($p = 2$)

$$\hat{x}(n) = {}^{(2)}_1 x(n-1) + {}^{(2)}_2 x(n-2).$$

$^{(2)}_1$ and $^{(2)}_2$ are to be determined according to the Covariance Method as those coefficients that minimize the mean square error defined below with $p = 2$.

$$E^{(p)} = \sum_{n=p} \{x(n) - \hat{x}(n)\}^2$$

a) Compute the covariance below for arbitrary k and l , note $p = 2$.

$$c(k, l) = \sum_{n=p} x(n-k)x(n-l)$$

- b) Set up the appropriate equations for determining the optimum $^{(2)}_1$ and $^{(2)}_2$. Solve for $^{(2)}_1$ and $^{(2)}_2$.
- c) What is the numerical value of $E^{(2)}$ given the $^{(2)}_1$ and $^{(2)}_2$ determined in (b)?
- d) Repeats parts (a) thru (c) for a third order predictor where $p=3$ and

$$\hat{x}(n) = {}^{(3)}_1 x(n-1) + {}^{(3)}_2 x(n-2) + {}^{(3)}_3 x(n-3)$$

- i. Compare the optimum coefficients $^{(3)}_1$, $^{(3)}_2$, and $^{(3)}_3$ with $^{(2)}_1$ and $^{(2)}_2$ computed in part (b).
 - ii. Compare the minimum $E^{(3)}$ with the minimum $E^{(2)}$ computed in part (c).
- e) Discuss the results for this problem and the previous problem involving the Autocorrelation Method.