6.433 Recursive Estimation6.435 System Identification

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Problem Set 7

Problem 1.

Pseudo-random Binary Sequences (PRBS)

A pseudo-random binary sequence taking the values 0 and 1 can be generated by

$$x(t+1) = \begin{pmatrix} a_1 & \cdots & a_n \\ 1_1 & 0 & \vdots \\ & \ddots & \\ 0 & & 10 \end{pmatrix} x(t)$$
(1)

 $u(t) = (0, \dots 01)x(t)$

Here $a_i \in \{0, 1\}$ and addition \oplus is mod 2. The initial state can be arbitrary excepting $x(0) \equiv 0$ is not allowed.

- (i) Draw a shift-register realization corresponding to the state-space representation Eq. (1).
- (ii) Show that the maximum period of shift register with *n*-states is $M = 2^n 1$. Such a PRBS is called maximum length.
- (iii) Show that the PRBS can be generated by

$$H(q)u(t) = 0 \tag{2}$$

for an appropriate polynomial H(q). (Notational convention as in book.)

- (iv) Eq. (2) has a solution of period $M = 2^n 1$ iff (a) H(q) is irreducible and (b) H(q) is a factor of $q \oplus q^{-M}$ but not a factor of $q \oplus q^{-p}$ for any p < M.
- (v) If u(t) is a maximum length PRBS of period $M = 2^n 1$, then within one period it contains (M+1)/2 ones and (M-1)/2 zeroes.

(vi) For k = 1, ..., M - 1

$$u(t) \oplus u(t-k) = u(t-\ell)$$
 for $\ell \in [1, M-1]$

that depends on k.

- (vii) Let u(t) be PRBS that shifts between a and -a and has period M. Compute its covariance function and spectral density.
- (viii) In what sense does a PRBS behave similar to white noise? Discuss this by comparing

$$\int_{-\pi}^{\pi} f(\omega) \Omega(\omega) \mathrm{d}\omega$$

and f being a continuous function and when Ω corresponds to the spectral density of a white noise process (variance 2) and a PRBS which shifts between a and -a with period M.

(ix) Carry out this comparison for a filtered PRBS and a filtered white noise corresponding to the filter

$$y_1(t) - 0.9y_1(t-1) = u_1(t)$$
 (or $u_2(t)$)

 $(u_1(\cdot))$ white noise with variance λ^2 and $u_2(t)$ is a PRBS of amplitude λ and period M) by comparing the covariance functions in the two cases for M = 500, 200, 100, 50 and 20.

Problem 2.

Consider the class of scalar models

$$A(q()Y_k = B(q)U_k + \xi_k , \quad k = 0, ..., N$$

where the disturbances ξ_k are generated by

$$F\xi_k = C(q)E_k \qquad (q = z^{-1})$$

and the E_k 's are a white Gaussian sequence and $Y_k = U_k = \xi_k = 0$, k < 0. The polynomials A(q), B(q), C(q) are given by

$$A(q) = 1 + a_1q + \dots + a_nq^n$$

$$B(q) = b_1q + \dots + b_nq^n$$

$$C(q) = 1 + c_1q + \dots + c_nq^n$$

Let $\theta = (a_1, \ldots, a_n; b_1, \ldots, b_n; c_1, \ldots, c_n)$. A parameter value θ which minimizes

$$\frac{1}{N}\sum_{k=1}^{N}\epsilon_{k}^{2}(\theta)$$

is sought where ϵ_k is the prediction error. If the c_i or the pair (a_i, b_i) are fixed then the minimization problem corresponds to a standard least squares problem.

- (a) Derive a generalized recursive least squares algorithm for the general case of correlated disturbances.
- (b) Note that if the past disturbances are known then one could derive a recursive least squares algorithm for the parameters: Now derive an algorithm for the determination of the parameters where the past disturbances are estimated.
- (c) Finally, note that is C(q) is a known stable polynomial then the determination of the parameters $(a_i, b_i)_{i=1,...,n}$ is a least-squares problem. Derive a three-stage least squares algorithm for parameter estimation.

Discuss a modification to this algorithm when the system order is unknown.