UNIVERSITY OF CALIFORNIA AT BERKELEY

Department of Mechanical Engineering ME233 Advanced Control Systems II

Final Examination Spring 2008

Closed book and closed ME233 class notes.

Notes allowed: Lecture Notes (slides), 8 sheets of handwritten notes and Laplace and Z-transform tables from ME232 class notes.

Your Name:			

Please answer all questions.

Problem:	1	2	3	4	5	Total
Max. Grade:	20	20	20	20	20	100
Grade:						

1 Problem

A set of random variables, X, Y and Z are Gaussian

$$\left[\begin{array}{c} X \\ Y \\ Z \end{array}\right] \sim \mathcal{N} \left(\left[\begin{array}{ccc} 1 \\ 1 \\ 1 \end{array}\right], \left[\begin{array}{ccc} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{array}\right]\right)$$

i.e.

$$\begin{bmatrix} \hat{x} \\ \hat{y} \\ \hat{z} \end{bmatrix} = E \left\{ \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} \right\} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \qquad E \left\{ \begin{bmatrix} X - \hat{x} \\ Y - \hat{y} \\ Z - \hat{z} \end{bmatrix} \begin{bmatrix} X - \hat{x} \\ Y - \hat{y} \\ Z - \hat{z} \end{bmatrix}^T \right\} = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

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1. Determine the value of the parameter $\theta \in \mathcal{R}$ that minimizes

$$J = E\{(X - \theta Y)^2\}.$$

2. Determine the value of the parameter $\theta \in \mathcal{R}$ that minimizes

$$J = E\{(X - \theta Y)^2 | Y = 0.5\}.$$

3. Determine the value of the parameter $\theta \in \mathcal{R}$ that minimizes

$$J = E\{(X - \theta Y)^2 | Y = 0.5, Z = -0.5\}.$$

Consider the design of a Kalman filter for a LTI stationary discrete time system that satisfies the standard assumptions, except for the fact that the input and output noises are correlated:

$$x(k+1) = Ax(k) + Bu(k) + Bw(k)$$
$$y(k) = Cx(k) + v(k)$$

where

• $x(k) \in \mathbb{R}^n$ is the state and

$$E\{x(0)\} = x_o$$
 $E\{(x(0) - x_o)(x(0) - x_o)^T\} = X_o$

- $u(k) \in \mathbb{R}^m$, $m \le n$ is the control input.
- $y(k) \in \mathbb{R}^p$, $p \le n$ is the measured output.
- w(k) and v(k) are stationary Gaussian, zero-mean, white noises that satisfy

$$E\left\{\left[\begin{array}{c}w(k)\\v(k)\end{array}\right]\left[\begin{array}{c}w^T(k+L)&v^T(k+L)\end{array}\right]\right\}=\left[\begin{array}{cc}W&S\\S^T&V\end{array}\right]\delta(L)$$

for all integers L, where

$$\left[\begin{array}{cc} W & S \\ S^T & V \end{array}\right] = \left[\begin{array}{cc} W^T & S \\ S^T & V^T \end{array}\right] \succeq 0 \qquad V \succ 0$$

and

$$E\{(x(0) - x_o) [w^T(k) v^T(k)]\} = 0$$

To design a Kalman filter for this system, we will first use the following output injection control law

$$u(k) = r(k) - Ky(k)$$

where K is a constant gain to be determined, and r(k) becomes the new exogenous input to the system.

1. Show that the resulting closed loop system can be expressed as follows

$$x(k+1) = A_c x(k) + B r(k) + B w_c(k)$$

$$y(k) = C x(k) + v(k)$$

and obtain expressions for A_c and $w_c(k)$.

2. Determine K so that

$$E\{w_c(k)v^T(k)\} = 0$$

and obtain the resulting expressions for the variance

$$W_c = E\{w_c(k)w_c^T(k)\}\$$

- 3. Write down the Kalman filter equations for estimating the state conditional mean $\hat{x}(k) = E\{x(k)|Y_k\}$, where $Y_k = \{y(0), \cdots, y(k)\}$.
- 4. Using prior known steady-state Kalman filter results, show that (C, A) detectable guarantees the convergence of the a-priori state estimation error covariance to a positive semi definite matrix that satisfies an algebraic Riccati equation.
- 5. Using prior known steady-state Kalman filter results, list the conditions that guarantee the existence of a unique asymptotically stabilizing stationary Kalman filter for this system.

Consider the first order system

$$x(k+1) = x(k) + u(k) + w(k)$$

where

- $E\{x(0)\} = x_o = 1$
- $E\{\tilde{x}^2(0)\} = X_o = 1$, where $\tilde{x}(0) = x(0) x_o$.
- $w(k) \sim \mathcal{N}(0,1)$ i.e. w(k) is zero-mean, normal with variance 1.
- $E\{w(k)x(0)\}=0$ for all $k\geq 0$

Assume that the state x(k) is directly measurable.

1. Determine the optimal control

$$U^o = \{u^o(0), u^o(1)\}$$

that minimizes the performance index

$$J = \sum_{k=1}^{2} E \left\{ 2^{k} x^{2}(k) + u^{2}(k-1) \right\} .$$

2. Determine the value of the optimal performance index.

Consider the discrete time system

$$A(q^{-1}) y(k) = q^{-1} B(q^{-1}) [u(k) + d(k)]$$

where q^{-1} is the one-step delay operator, u(k) is the controlling input, y(k) is the measured output.

d(k) is a sinusoidal disturbance of the form

$$d(k) = d_o \sin(\frac{\pi}{2} k + \phi_o)$$

where d_o and ϕ_o are **unknown**.

Assume that the polynomials

$$A(q^{-1}) = 1 + 0.9 q^{-1}$$

$$B(q^{-1}) = 0.5 + 0.5 q^{-1}$$

are known.

Design a control system that satisfies the following requirements:

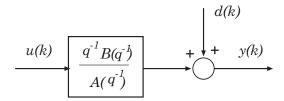
- The system output should asymptotically track an **arbitrary** desired output $y_d(k)$, which is known two steps in advance, with zero phase error.
- The disturbance d(k) should be rejected.
- The closed loop poles of the feedback system should be the roots of

$$A_c^*(q) = q(q - 0.5)$$

Clearly indicate all steps of your design process. ¹

¹ If your answer requires the solution of a Diophantine equation, write down the linear equation Mx = b, that must be solved in order to obtain the solution of the Diophantine equation. This includes defining the elements of the unknown vector x and all of the coefficients of the matrix x and the vector x and ont need to solve this linear equation.

Consider the identification of an open loop asymptotically stable plant with a sinusoidal output disturbance, as depicted below



where q^{-1} is the one-step delay operator, u(k) is the controlling input, y(k) is the measured output.

d(k) is a sinusoidal disturbance of the form

$$d(k) = d_0 \sin(\omega_d k + \phi_0)$$

where the frequency ω_d is **known** but d_o and ϕ_o are **unknown**.

The plant polynomials

$$A(q^{-1}) = 1 + a_1 q^{-1} + a_2 q^{-2}$$
 $B(q^{-1}) = b_o + b_1 q^{-1}$

are of known order but their coefficients are unknown.

The following algorithm is proposed for identifying the plant parameters:

(i) The control input is

$$u(k) = U_1 \sin(\omega_1 k) + U_2 \sin(\omega_2 k)$$

where U_1 and U_2 are constants and $\omega_1 \neq \omega_2 \neq \omega_d$.

(ii) The a-priori output and error estimates are

$$\hat{y}^{o}(k) = \phi(k)^{T} \hat{\theta}(k-1)$$
 $e^{o}(k) = y(k) - \hat{y}^{o}(k)$

where the regressor $\phi(k)$ is given by

$$\phi(k) = \begin{bmatrix} -y(k-1) & -y(k-2) & u(k-1) & u(k-2) & \sin(\omega_d k) & \cos(\omega_d k) \end{bmatrix}^T$$

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and $\hat{\theta}(k) \in \mathcal{R}^6$ is the parameter estimate, which is updated by a standard RLS PAA with forgetting factor, as shown below.

(iii) RLS PAA with forgetting factor:

$$e(k+1) = \frac{\lambda_1(k)e^o(k+1)}{\lambda_1(k) + \phi^T(k+1)F(k)\phi(k+1)}$$

$$\hat{\theta}(k+1) = \hat{\theta}(k) + \frac{1}{\lambda_1(k)}F(k)\phi(k+1)e(k+1)$$

$$F(k+1) = \frac{1}{\lambda_1(k)} \left[F(k) - \lambda_2(k) \frac{F(k)\phi(k+1)\phi^T(k+1)F(k)}{\lambda_1(k) + \lambda_2(k)\phi^T(k+1)F(k)\phi(k+1)} \right]$$

where
$$0<\underline{\lambda}_1\leq \lambda_1(k)\leq 1$$
 and $0\leq \lambda_2(k)\leq \overline{\lambda}_2<2.$

Perform a stability analysis to determine if $\lim_{k\to\infty} e^o(k) = 0$.