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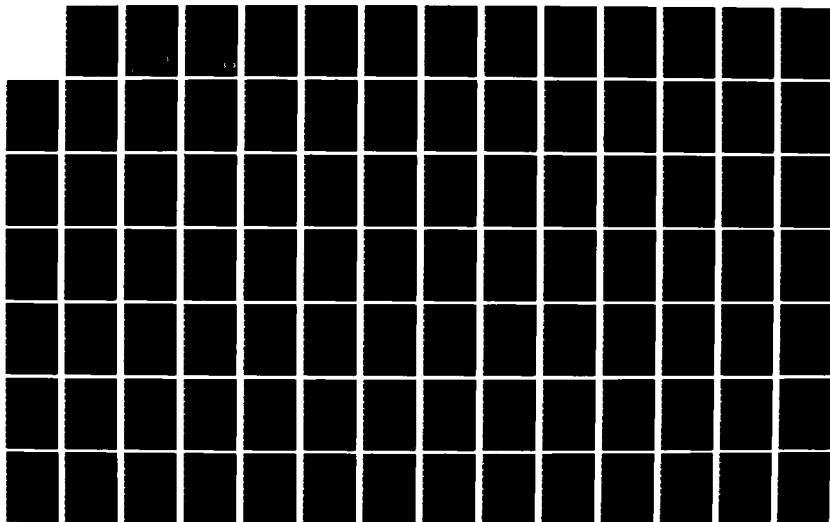
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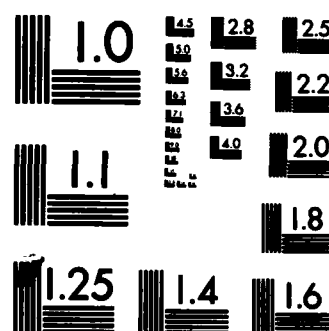
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THESIS

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Jean M. Howey  
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ROBUST FLIGHT CONTROLLERS

THESIS

Presented to the Faculty of the School of Engineering  
of the Air Force Institute of Technology  
Air University  
in Partial Fulfillment of the  
Requirements for the Degree of  
Master of Science

by

Jean M. Howey, B.S.

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USAF

Graduate Aeronautical Engineering

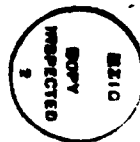
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## Preface

In the past two decades, one of the major stumbling blocks in applying optimal control theory to flight control problems has been that measurements of all the states of the system are not available. Thus, it is necessary to use some type of observer or filter to reconstruct the states of the aircraft. However, when this is done, all guarantees of desirable stability robustness properties are lost. This thesis addresses this problem and evaluates the success of some techniques to recover the good stability robustness properties associated with full-state feedback.

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## Abstract

This study examines the concept of robustifying a controlled system against differences which may exist between the real world system and the low-order design model upon which the controller design is based. The types of controllers considered are based upon the Linear system model, Quadratic cost, and Gaussian (LQG) noise process methodology of optimal control theory. It is assumed that full-state feedback is not available and a Kalman filter is employed to provide state estimates to the controller. Both continuous-time and sampled-data controllers are considered.

Two robustification techniques are considered. The first is the method of injecting zero-mean white Gaussian noise into the design model at the point of entry of the control inputs during the process of tuning the Kalman filter. The second method is an extension of the first, where the white noise is replaced by a time-correlated noise. This allows the primary strength of the noise to be concentrated only in the frequency range where robustification is desired. Comparing the results of applying the two methods allows a designer to make a trade-off between the amount of desired robustification and the performance degradation at the design conditions which occurs when the techniques are applied.

Both methods are found to improve substantially the robustness properties of the controllers considered. For the specific flight control problem considered in this thesis, the technique of injecting white input noise into the design model produced the desired degree of robustification without prohibitively degrading performance at the design conditions.

The second method, though effective, did not yield substantial enough performance benefits over the first to warrant use in actual implementation.

## ROBUST FLIGHT CONTROLLERS

### I. Introduction

#### 1.1 Background

A vital element of a flight control system is that it be "robust". Robustness implies that the controller provides adequate (i.e., stable closed-loop) performance over a wide range of operating conditions and system parameters. For example, finite-dimension models of the dynamics of an aircraft are typically generated by linearizing the aircraft equations of motion about a limited number of specific equilibrium flight conditions. Controllers are then designed for these "trim" conditions. However, at flight conditions other than the specific trim condition used for controller design, the closed-loop performance of the controller system may be inadequate or unstable. That is, at the off-design flight condition, the aircraft parameters (upon which the design is based) have changed sufficiently to make performance of the controller inadequate. One method used in the past has been to schedule feedback gains for the controller so as to compensate for the change in system parameters. However, it is desirable to have enlarged regions about nominal design conditions within which a specific controller design will yield adequate performance. This would allow linear perturbation techniques to be used with more confidence and may also reduce the number of required design conditions about which such perturbations are defined. It may even be desirable to have a fixed controller which is sufficiently robust to produce acceptable performance for all conditions in the aircraft's flight envelope.



Robustness is also a concern when controllers are designed using purposefully reduced order models. A robust controller will provide stable closed-loop performance even when states of the real-world system have been ignored in the design model. A third robustness issue is survivability. Robustness can also imply the ability to maintain closed-loop stability if, for instance, part of the flight control system is lost due to ground fire: i.e., the "true" system is vastly different than the system model upon which the controller was designed. Such a robust controller will provide a stable, though degraded, aircraft performance while adaptation algorithms attempt to discern what system elements have been lost and to determine an appropriate modification of controller characteristics for future use.

In the 1960's, modern control theory methods, such as optimal control theory, showed promise in application to flight control problems. One drawback of applying optimal control theory techniques to flight control problems, however, is that the resulting controllers require full-state feedback, but measurements of all states are generally not available. Thus, it becomes necessary to include a filter or observer in the controlled system to estimate the states. However, once an observer has been inserted into the loop, stability robustness becomes a major concern (Ref 5).

Recently, efforts to improve the robustness properties of observer-based controllers have included a technique, developed by J.C. Doyle and G. Stein (Ref 6), which injects white noise into a controller system model at the point of entry of the control inputs during the process of tuning the Kalman filter. It is claimed that, as the strength of the input noise

is increased, the filter-based controller will asymptotically recover the good robustness properties of a full-state feedback system in the continuous-time case. A disadvantage of this technique is that the additional noise can degrade the performance of the system at the design conditions.

A natural extension to the idea of injecting white input noise into a system model is to consider time-correlated noise. This allows the robustification technique to be applied only over a desired frequency range rather than over all frequencies as in the original Doyle and Stein method. Thus, the degradation in performance due to the additional noise can be reduced (Ref 15;16;28).

The techniques described above are applied in this thesis to a specific aircraft flight control problem. Two types of controllers are considered. The first is an optimal Linear Quadratic Gaussian (LQG) regulator. Available software allows the design of a continuous-time and a digital controller. The second type of controller is an optimal LQG-based Proportional-plus-Integral digital controller. For both types of controller, a Kalman filter is implemented to provide estimates of the states of the system. The success of applying the robustification techniques is determined by designing a Kalman filter and controller for one trim condition, then evaluating the performance of the controller at an off-design flight condition, as well as designing the filter and controller on the basis of purposely reduced-order models for computational loading reasons. Time histories of the mean and standard deviations of the aircraft states and controls are examined. Additionally, it is desired to compare the performance of the controller at the design conditions with

no input noise, white noise and time-correlated noise used for filter retuning (robustification) to determine how much the performance is degraded by the input of the noise, and also to determine if performance benefits can be gained by using time-correlated noise as opposed to white noise.

## 1.2 Problem

The primary objectives of this thesis are:

1. To apply the robustification techniques of injecting white or time-correlated noise into a system model at the control entry points during filter tuning to a flight control problem for a high performance aircraft.
2. To extend the techniques for LQG regulators to LQG-based PI controllers.
3. To evaluate the robustness properties of the controller designs by performing covariance analyses for the controlled systems at both the design flight condition and other off-design conditions within the aircraft's operational flight envelope, using a purposefully reduced-order design model. Robustified designs are to be compared to unrobustified designs and also to full-state feedback designs when possible.

## 1.3 Sequence of Presentation

The body of this thesis is contained in Chapter II-V. Chapter II presents the equations used for designing and evaluating optimal deterministic LQG regulators for both the continuous- and discrete-time case.

Also included is the application of the Doyle and Stein technique to continuous-time systems and several extensions of the technique to discrete-time systems.

Chapter III presents the equations necessary to design and evaluate optimal LQG-based PI controllers for the discrete-time case. Then the Doyle and Stein technique is applied to systems employing PI controllers. As pointed out in the text, robustification by this technique is a procedure which affects only the Kalman filter design; thus, the same method is used for PI controllers as for regulators in this thesis.

Chapter IV examines the idea of injecting time-correlated noise rather than white noise into a controlled system model during tuning of the filter, to improve the tradeoff of the controller's robustness properties versus performance degradation at design conditions. First, a stochastic process model (shaping filter) is developed for the noise process which is then augmented with the system state differential equations. Then, the types of time-correlated noise of interest to this thesis are considered, including a discussion of the specific shaping filters to generate the desired noise process.

Chapter V presents the model of a high-performance aircraft to be used in this thesis. A design flight condition is chosen and linearized perturbation equations of motion are developed for the aircraft about that operating point. The design model for the Kalman filter and controller is purposefully reduced in order from the full set of linearized equations so that this aspect of robustness can be examined. Finally, models at other flight conditions are listed with which the robustness of the system to parameter changes will be evaluated.

The findings of this thesis, results and conclusions are presented in Chapter VI. Recommendations for further research are made in Chapter VII.

Four appendices are included. The first presents a generic format for a controller into which the types of controllers in Chapters II and III can be rearranged. The usefulness of the format becomes apparent in a performance analysis, when the same set of equations can be used to evaluate the performance of any controller in the standard format.

Appendix B lists the source code for a Fortran program used in designing LQG regulators. Included are a discussion of how the program is executed and modifications from a previous version to allow the input of time-correlated noise (Ref 21).

Appendix C lists the modification to the software of Reference 13 and 30 to allow the input of white and time-correlated noise into the system model. The software is an interactive program used for designing PI regulators with a Command Generator Tracker in the feed-forward loop. This thesis will exploit only the PI regulator design capabilities.

Appendix D includes further performance analysis results in addition to the findings in Chapter VI. It was stated in Reference 6 that the Doyle and Stein technique is not guaranteed to improve robustness if the design model is non-minimum phase (i.e., there are transmission zeroes in the right-half  $s$ -plane). Appendix D shows a case where the design model was non-minimum phase and where the addition of input noise to the system model has actually drive an initially stable closed-loop system to be unstable.

## II. LQG REGULATORS

### 2.1 Introduction

This chapter presents the equations used in designing optimal Linear Quadratic Gaussian (LQG) controllers for a system modelled as linear, time-invariant, and driven by zero-mean, white Gaussian noise and deterministic inputs, subject to quadratic costs for defining optimality criteria. The model used for this thesis is described in Chapter V. The methods used are taken primarily from Reference 24 unless otherwise stated.

In the first section, the controller equations for a continuous-time system having continuous-time measurements are given. The structures of controllers assuming perfect access to all the states of a system versus controllers with a Kalman filter to estimate states are examined. Next, the equations needed to evaluate the performance of controller designs are given. Finally, a technique developed by Doyle and Stein (Ref 6) to robustify the Kalman filter for continuous-time systems is presented.

The fourth section contains the controller equations for a continuous-time system having sampled-data measurements. Next the performance analysis equations for the sampled-data system are developed.

The final sections contain alternative methods for applying the Doyle and Stein technique to sampled-data systems. Three possible basic approaches are simply to discretize the controller designed for the continuous system or to modify the technique in one of two ways so that it applies directly to a discrete system.

All methods described in this chapter are incorporated in a Fortran program originally written by Captain Eric Lloyd and modified to some extent by this author. For source codes and information about the program, see Reference 21. Modifications and additions to the program are listed in Appendix B.

## 2.2 Continuous-Time Controllers

The state description for an important class of continuous-time systems is given by the linear stochastic differential equation

$$\dot{\underline{x}}(t) = F(t) \underline{x}(t) + B(t) \underline{u}(t) + G(t) \underline{w}(t) \quad (2-1)$$

where  $\underline{w}(t)$  is a white Gaussian noise with statistics

$$E\{\underline{w}(t)\} = 0 \quad (2-2a)$$

$$E\{\underline{w}(t) \underline{w}^T(t+\tau)\} = Q(t)\delta(\tau) \quad (2-2b)$$

An optimal LQG controller for the system minimizes the cost functional

$$J_c = E \left\{ \frac{1}{2} \underline{x}^T(t_f) X_f \underline{x}(t_f) \right\} + \frac{1}{2} \int_{t_0}^{t_f} \left\{ \begin{bmatrix} \underline{x}(t) \\ \underline{u}(t) \end{bmatrix}^T \begin{bmatrix} W_{xx}(t) & W_{xu}(t) \\ W_{xu}^T(t) & W_{uu}(t) \end{bmatrix} \begin{bmatrix} \underline{x}(t) \\ \underline{u}(t) \end{bmatrix} \right\} dt \quad (2-3)$$

where  $W_{xx}(t)$  is a positive semi-definite weighting matrix associated with the system states,  $\underline{x}(t)$ ;  $W_{uu}(t)$  is a positive definite weighting matrix associated with the controls,  $\underline{u}(t)$ , applied, to the system;  $W_{xu}(t)$  is a cost-weighting matrix associated with cross terms of  $\underline{x}(t)$  and  $\underline{u}(t)$  and is

chosen so that the composite matrix of Equation (2-3) is positive semi-definite; and  $X_f$  is a positive semi-definite weighting matrix associated with the states at the final time,  $\underline{x}(t_f)$ .

The optimal control to be applied at time  $t$  which minimizes the above cost function is described as

$$\underline{u}^*(t) = - G_c^*(t) \underline{x}(t) \quad (2-4)$$

with the gain matrix,  $G_c^*(t)$ , given by

$$G_c^*(t) = W_{uu}^{-1}(t) B^T(t) K_c(t) \quad (2-5)$$

and  $K_c(t)$  is calculated using the backward Riccati differential equation

$$\begin{aligned} \dot{-K}_c(t) = & F^T(t) K_c(t) + K_c(t) F(t) + W_{xx}(t) \\ & - K_c(t) B(t) W_{uu}^{-1}(t) B^T(t) K_c(t) \end{aligned} \quad (2-6)$$

subject to the final condition

$$K_c(t_f) = X_f \quad (2-7)$$

Notice that Equation (2-6) assumes that the cross weighting matrix  $W_{xu}(t)$  is zero. If this is not the case, an appropriate variable transformation can be made which will account for the non-zero cross terms and allow equation (2-6) to be used (Ref 24:202-203; 19:79-86).

Up to this point, the controller equations have been derived allowing for time-varying systems, cost-weighting matrices and feedback gains. However, as described in Chapter V, the model used for this thesis is time-invariant with stationary noises (i.e., the covariance kernel



$E\{w(t)w^T(t+\tau)\}$  is a function of only the time difference  $\tau$  (Ref 23:139-140)) and constant weighting matrices. Thus a steady-state constant gain controller can be used, ignoring terminal transients in the feedback gain

$$\underline{u}^*(t) = -G_c^* \underline{x}(t) = - \left[ W_{uu}^{-1} B^T \bar{K}_c \right] \underline{x}(t) \quad (2-8)$$

where  $\bar{K}_c$  is now the solution to the steady-state Riccati equation

$$-\dot{\bar{K}}_c = 0 = F^T \bar{K}_c + \bar{K}_c F + W_{xx}^{-1} \bar{K}_c B W_{uu}^{-1} B^T \bar{K}_c \quad (2-9)$$

Henceforth, the assumption of a time-invariant system with stationary noises and constant weighting matrices is made, and the time argument is omitted unless needed to avoid confusion.

The optimal control law of Equation (2-4) is given as a gain matrix times the state values at a particular time. If, however, perfect knowledge of the states is not available, then an estimate of the state,  $\hat{\underline{x}}(t)$ , must be used in place of  $\underline{x}(t)$ . In this case, a Kalman filter is employed to evaluate the conditional mean,  $\hat{\underline{x}}(t)$ , and conditional covariance,  $P(t)$ , of the states of the system.

Instead of the actual values of the states, assume that what are accessible from the system are noise-corrupted measurements of the time-invariant form

$$\underline{z}(t) = H \underline{x}(t) + \underline{v}(t) \quad (2-10)$$

where  $\underline{v}(t)$  is a stationary white Gaussian noise assumed independent of the dynamics driving noise  $\underline{w}(t)$  and with statistics

$$E\{\underline{v}(t)\} = \underline{0} \quad (2-11a)$$

$$E\{\underline{v}(t) \underline{v}^T(t+\tau)\} = R\delta(\tau) \quad (2-11b)$$

Then the Kalman filter equations yield an estimate of the states and the covariance via (Ref 23:257,259)

$$\dot{\underline{\hat{x}}}(t) = \underline{F}\underline{\hat{x}}(t) + \underline{B}\underline{u}(t) + \underline{P}\underline{H}^T \underline{R}^{-1} [\underline{z}(t) - \underline{H}\underline{\hat{x}}(t)] \quad (2-12a)$$

$$\dot{\underline{P}}(t) = \underline{F}\underline{P}(t) + \underline{P}(t)\underline{F}^T + \underline{G}\underline{Q}\underline{G}^T - \underline{P}(t) \underline{H}^T \underline{R}^{-1} \underline{H}\underline{P}(t) \quad (2-12b)$$

These differential equations are solved forward in time subject to the initial conditions

$$E\{\underline{x}(t_0)\} = \bar{\underline{x}}_0 \quad (2-13a)$$

$$E\{[\underline{x}(t_0) - \bar{\underline{x}}_0] [\underline{x}(t_0) - \bar{\underline{x}}_0]^T\} = \underline{P}_0 \quad (2-13b)$$

which are obtained from an a priori Gaussian density function for the states at the initial time.

The precomputable Kalman filter gain for the continuous-time system is expressed in Equation (2-12a) as

$$\underline{K}(t) = \underline{P}(t) \underline{H}^T \underline{R}^{-1} \quad (2-14)$$

If it is assumed that the initial transients of Equation (2-12b) are short compared to the total time of interest, the steady-state covariance,  $\bar{\underline{P}}$ , can be computed as the solution to

$$\dot{\bar{\underline{P}}}(t) = 0 = \bar{\underline{F}}\bar{\underline{P}} + \bar{\underline{P}}\bar{\underline{F}}^T + \bar{\underline{G}}\bar{\underline{Q}}\bar{\underline{G}}^T - \bar{\underline{P}}\bar{\underline{H}}^T \bar{\underline{R}}^{-1} \bar{\underline{H}}\bar{\underline{P}} \quad (2-15)$$

and the constant Kalman filter gain is given by

$$\underline{K} = \bar{\underline{P}}\bar{\underline{H}}^T \bar{\underline{R}}^{-1} \quad (2-16)$$

Do not confuse the Kalman Filter gain,  $K$ , with  $K_c$  of Equation (2-6) and (2-9).

Figure (2-1) shows the form of the continuous-time controller.

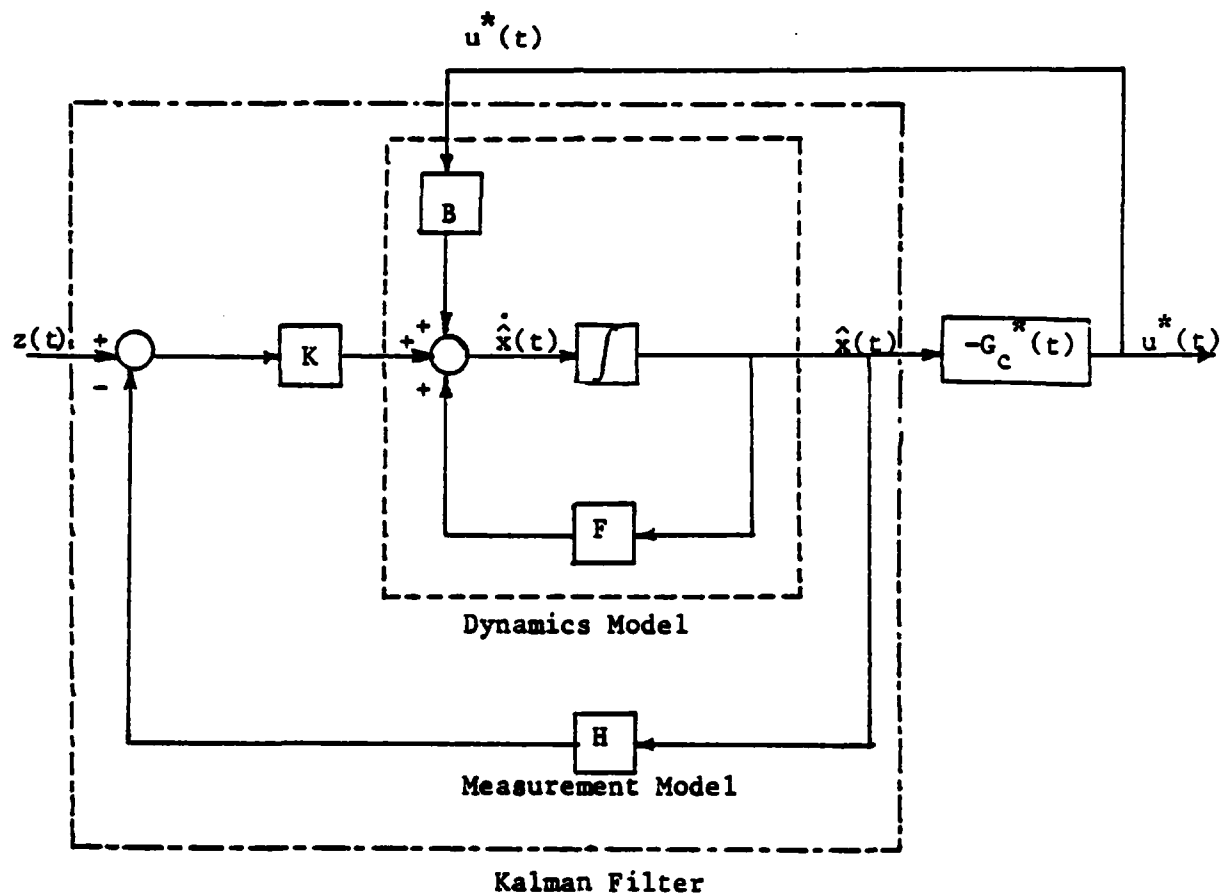


Figure 2-1: Continuous Measurement LQG Optimal Stochastic Controller

### 2.3 Continuous-Time Performance Analysis

The performance analysis for a continuous-time LQG controller is based on the development for a sampled-data controller given in Reference 23 and derived in Reference 21.

First, a "truth model" is developed for a given system which is judged to represent adequately the response of a system to inputs and disturbances encountered in the real world. Then, controllers for the system can be designed, generally based on lower order, simplified models. The performance is determined by examining the statistical characteristics of the truth model states,  $\underline{x}_t(t)$ , and the controls,  $\underline{u}(t)$ , for each proposed controller inserted into the real world simulation provided by the truth model. This is depicted in Figure (2-2).

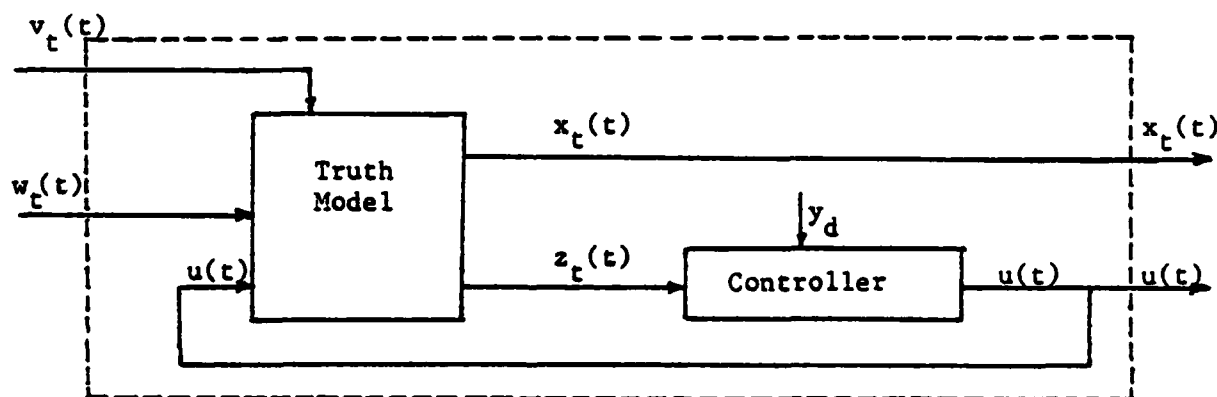


Figure 2-2: Performance Evaluation for Linear Sampled-Data Controller

The performance evaluation determines the statistical characteristics of  $\underline{x}_t(t)$ , the truth model states, rather than the states of the simplified controller model. It is important to determine the effect of applying controls from a reduced order controller on the actual system response.

Additionally, the statistical characteristics of the controls are examined to ensure that they do not exceed physical limits or design specifications.

The assumed linear differential equation describing the continuous-time truth model system and available measurements are

$$\dot{\underline{x}}_t(t) = F_t \underline{x}(t) + B_t \underline{u}(t) + G_t \underline{w}_t(t) \quad (2-17)$$

$$\underline{z}_t(t) = H_t \underline{x}(t) + \underline{v}_t(t) \quad (2-18)$$

where initial conditions and statistics of the noise are given by

$$E\{\underline{x}_t(t_0)\} = \bar{\underline{x}}_{t_0} \quad (2-19)$$

$$E\{[\underline{x}_t(t_0) - \bar{\underline{x}}_{t_0}][\underline{x}_t(t_0) - \bar{\underline{x}}_{t_0}]^T\} = P_{t_0} \quad (2-20)$$

$$E\{\underline{w}_t(t)\} = 0 \quad (2-21)$$

$$E\{\underline{w}_t(t) \underline{w}_t^T(t+\tau)\} = Q_t \delta(\tau) \quad (2-22)$$

$$E\{\underline{v}_t(t)\} = 0 \quad (2-23)$$

$$E\{\underline{v}_t(t) \underline{v}_t^T(t+\tau)\} = R_t \delta(\tau) \quad (2-24)$$

Note that  $\underline{v}_t(t)$  and  $\underline{w}_t(t)$  are assumed to be independent of each other and that the subscript  $t$  refers to the truth model.

A useful form of expressing the control law of Equation (2-4) and other more general linear control algorithms, and an equation to propagate the internal states of the controller, are given by

$$\underline{u}(t) = G_{cx-c} \underline{x}_c(t) + G_{cz} \underline{z}_t(t) + G_{cy} \underline{y}_d(t) \quad (2-25)$$

$$\dot{\underline{x}}_c(t) = F_c \underline{x}_c(t) + B_{cz} \underline{z}_t(t) = B_{cy} \underline{y}_d(t) \quad (2-26)$$

where  $\underline{y}_d(t)$  refers to a command input for the system controlled variable,  $\underline{y}(t)$ , to track, where

$$\underline{y}(t) = C\underline{x}(t) + D_y \underline{u}(t) \quad (2-27)$$

The gain matrices in Equations (2-25) and (2-26) are evaluated explicitly in Appendix A. Putting the control law into the generic form allows direct comparison of different types of linear control laws.

Determination of time histories are desired of the mean and covariance of an augmented vector

$$\underline{y}_a(t) = \begin{bmatrix} \underline{x}_t(t) \\ \underline{u}(t) \end{bmatrix} \quad (2-28)$$

All quantities of interest in a performance analysis are assumed to be components or linear combinations of components of this vector. If  $q_k$  is a scalar quantity of interest, i.e.,

$$q_k = \underline{q}_k^T \underline{y}_a(t) \quad (2-29)$$

then the statistics of  $q_k$  are given by

$$\text{mean}\{q_k(t)\} = \underline{q}_k^T \underline{m}_{y_a}(t) \quad (2-30a)$$

$$\text{cov}\{q_k(t)\} = \underline{q}_k^T P_{y_a y_a}(t) \underline{q}_k \quad (2-30b)$$

where  $\underline{m}_{y_a}(t)$  is the mean and  $P_{y_a y_a}(t)$  is the covariance of the augmented vector  $\underline{y}_a(t)$ .

To evaluate the first two moments of  $\underline{y}_a(t)$ , it is first necessary to form another augmented vector composed of the internal states of the truth model and the controller model

$$\underline{x}_a(t) = \begin{bmatrix} \underline{x}_t(t) \\ \underline{x}_c(t) \end{bmatrix} \quad (2-31)$$

Using Equation (2-18) and (2-25),  $\underline{u}(t)$  and  $\underline{z}(t)$  are eliminated from Equation (2-17)

$$\begin{aligned} \dot{\underline{x}}_t(t) = & \{ F_t + B_t G_{cz} H_t \} \underline{x}_t(t) + B_t G_{cx} \underline{x}_c(t) \\ & + B_t G_{cy} \underline{y}_d(t) + B_t G_{cz} \underline{v}_t(t) + G_{t-t} \underline{w}_t(t) \end{aligned} \quad (2-32)$$

Similar substitution into Equation (2-26) yields

$$\dot{\underline{x}}_c(t) = F_c \underline{x}_c(t) + B_{cy} \underline{y}_d(t) + B_{cz} [H_t \underline{x}_t(t) + \underline{v}_t(t)] \quad (2-33)$$

Form the augmented noise vector

$$\underline{w}_a(t) = \begin{bmatrix} \underline{w}_t(t) \\ \underline{v}_t(t) \end{bmatrix} \quad (2-34)$$

as a zero-mean white Gaussian noise with covariance kernel

$$E\{\underline{w}_a(t) \underline{w}_a^T(t + \tau)\} = Q_a \delta(\tau), \text{ where}$$

$$Q_a = \begin{bmatrix} Q_t & 0 \\ 0 & R_t \end{bmatrix} \quad (2-35)$$

Non-zero cross terms can easily handle the case of  $\underline{w}_t(t)$  and  $\underline{v}_t(t)$  being correlated. Now an augmented system equation can be written as

$$\dot{\underline{x}}_a(t) = F_a \underline{x}_a(t) + B_a \underline{y}_d(t) + G_a \underline{w}_a(t) \quad (2-36)$$

where

$$F_a = \begin{bmatrix} F_t + B_t G_{cz} H_t & B_t G_{cx} \\ B_{cz} & F_c \end{bmatrix} \quad (2-37)$$

$$B_a = \begin{bmatrix} B_t G_{cy} \\ B_{cy} \end{bmatrix} \quad (2-38)$$

$$G_a = \begin{bmatrix} G_t & B_t G_{cz} \\ 0 & B_{cz} \end{bmatrix} \quad (2-39)$$

The initial conditions for the augmented vector are

$$E\{\underline{x}_a(t_0)\} = \begin{bmatrix} \bar{\underline{x}}_{to} \\ \bar{\underline{x}}_{co} \end{bmatrix} = \bar{\underline{x}}_{ao} \quad (2-40)$$



$$P_a(t_0) = \begin{bmatrix} P_0 & 0 \\ 0 & 0 \end{bmatrix} \quad (2-41)$$

Initial conditions on the controller states are often assumed to be zero:

$$\bar{x}_{co} = 0.$$

The mean and covariance of the states in Equation (2-36) propagate as

$$\dot{\bar{x}}_a(t) = F_a \bar{x}_a(t) + B_a y_d(t) \quad (2-42)$$

$$\dot{P}_{x_a x_a}(t) = F_a P_{x_a x_a}(t) + P_{x_a x_a}(t) F_a^T + G_a Q_a G_a^T \quad (2-43)$$

or solving the differential equation yields

$$\bar{x}_a(t) = \phi_a(t, t_0) \bar{x}_{ao} + \int_{t_0}^t \phi_a(t, \tau) B_a y_d(\tau) d\tau \quad (2-44)$$

$$P_{x_a x_a}(t) = \phi_a(t, t_0) P_{ao} \phi_a^T(t, t_0) + \int_{t_0}^t \phi_a(t, \tau) G_a Q_a G_a^T \phi_a^T(t, \tau) d\tau \quad (2-45)$$

where  $\phi_a(t, t_0)$  is the state transition matrix associated with  $F_a$ , i.e.,

$$\phi_a(t, t_0) = \phi_a(t - t_0) = \exp \{F_a(t - t_0)\}.$$

However, the requirement is to solve for the statistics of the vector  $y_a(t)$  defined in Equation (2-32). This vector is related to  $x_a(t)$  by

$$\underline{y}_a(t) = \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} \begin{bmatrix} \underline{x}_t(t) \\ \underline{x}_c(t) \end{bmatrix} + \begin{bmatrix} 0 \\ G_{cy} \end{bmatrix} \underline{y}_d(t) + \begin{bmatrix} 0 \\ G_{cz} \end{bmatrix} \underline{v}_t(t) \quad (2-46)$$

The vector is seen to be a linear combination of jointly Gaussian variables with known statistics. Thus, the desired statistics can be generated from Equation (2-39) using the method shown in Reference 23:112.

$$\underline{m}_{y_a}(t) = \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} \underline{m}_{x_a}(t) + \begin{bmatrix} 0 \\ G_{cy} \end{bmatrix} \underline{y}_d(t) \quad (2-47)$$

$$\begin{aligned} P_{y_a y_a}(t) &= \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} P_{x_a x_a}(t) \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix}^T \\ &+ \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} P_{x_a v_t}(t) \begin{bmatrix} 0 & G_{cz}^T \end{bmatrix} \\ &+ \begin{bmatrix} 0 \\ G_{cz} \end{bmatrix} P_{x_a v_t}^T(t) \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix}^T \\ &+ \begin{bmatrix} 0 \\ G_{cz} \end{bmatrix} P_{v_t v_t}(t) \begin{bmatrix} 0 & G_{cz}^T \end{bmatrix} \end{aligned} \quad (2-48)$$

In the last term, note that  $P_{v_t v_t}(t) = R_t \delta(0)$  and that  $\delta(0) = \infty$ . This term gives an infinite value to the lower right-hand partition of the covariance matrix. However, it is still of interest to look at the contribution of the other three terms separately. To evaluate Equation (2-48) fully, it is necessary to find an expression for  $P_{x_a v_t}(t)$ . This can be written as

$$P_{x_a v_t} = E\{[\underline{x}_a(t) - \underline{m}_{x_a}(t)][\underline{v}_t(t) - \underline{m}_{v_t}(t)]^T\} \quad (2-49)$$

Expand this, noting that  $\underline{m}_{v_t}(t) = 0$

$$P_{x_a v_t}(t) = E\{\underline{x}_a(t)\underline{v}_t^T(t) - \underline{m}_{x_a}(t)\underline{v}_t^T(t)\} \quad (2-50)$$

$$P_{x_a v_t} = E\{\underline{x}_a(t)\underline{v}_t^T(t)\} - \underline{m}_{x_a}(t)E\{\underline{v}_t^T(t)\} \quad (2-51)$$

but  $E\{\underline{v}_t^T(t)\} = \underline{m}_{v_t}^T(t) = \underline{0}^T$ , therefore

$$P_{x_a v_t}(t) = E\{\underline{x}_a(t)\underline{v}_t^T(t)\} \quad (2-52)$$

Replace  $\underline{x}_a(t)$  with the solution form for Equation (2-36)

$$\begin{aligned} P_{x_a v_t}(t) = & E\{\phi_a(t, t_0)\underline{x}_a(t_0)\underline{v}_t^T(t) \\ & + \int_{t_0}^t \phi_a(t, \tau) B_{a-d}(\tau) \underline{v}_t^T(t) d\tau \\ & + \int_{t_0}^t \phi_a(t, \tau) G_{a-a}(\tau) \underline{v}_t^T(t) d\tau\} \end{aligned} \quad (2-53)$$

Making the assumption that  $\underline{x}_a(t_0)$  and  $\underline{v}_t(t)$  are independent (and thus uncorrelated), the first term in Equation (2-53) is zero since  $E\{\underline{v}_t^T(t)\} = \underline{0}^T$ . Substituting in the augmented matrices, the remaining terms are

$$P_{\underline{x}_a \underline{v}_t}(t) = E \left\{ \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} B_t & G_{cy} \\ & B_{cy} \end{bmatrix} \underline{y}_d(\tau) \underline{v}_t^T(\tau) d\tau \right. \\ \left. + \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} G_t & B_t G_{cz} \\ 0 & B_{cz} \end{bmatrix} \begin{bmatrix} \underline{w}_t(\tau) \\ \underline{v}_t(\tau) \end{bmatrix} \underline{v}_t^T(\tau) d\tau \right\} \quad (2-54)$$

which can be rewritten as

$$P_{\underline{x}_a \underline{v}_t}(t) = E \left\{ \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} B_t G_{cy} \\ & B_{cy} \end{bmatrix} \underline{y}_d(\tau) \underline{v}_t^T(\tau) d\tau \right. \\ \left. + \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} G_t \underline{w}_t(\tau) + B_t G_{cz} \underline{v}_t(\tau) \\ B_{cz} \underline{v}_t(\tau) \end{bmatrix} \underline{v}_t^T(\tau) d\tau \right\} \quad (2-55)$$

Again, the first term is zero since  $\underline{v}_t(t)$  is zero-mean, and  $\underline{y}_d(\tau)$  is deterministic. This leaves

$$P_{x_a v_t}(t) = E \left\{ \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} G_{t \leftarrow t}(\tau) + B_t G_{cz \leftarrow t}(\tau) \\ B_{cz \leftarrow t}(\tau) \end{bmatrix} \underline{v}_t^T(\tau) d\tau \right\} \quad (2-56)$$

Noting that  $\underline{w}_t(\tau)$  and  $\underline{v}_t(\tau)$  are assumed independent and

$$E\{\underline{v}_t(\tau) \underline{v}_t^T(t)\} = R_t \delta(t-\tau) \quad (2-57)$$

the expected value operation can be moved inside the integral, and Equation (2-56) further reduces to

$$P_{x_a v_t}(t) = \int_{t_0}^t \phi_a(t, \tau) \begin{bmatrix} B_t G_{cz} \\ B_{cz} \end{bmatrix} R_t \delta(t-\tau) d\tau \quad (2-58)$$

Applying the Dirac delta sifting property to the above equation, noting that  $t$  is the upper limit of the integration and that it also appears in the delta function argument, yields

$$P_{x_a v_t}(t) = \phi_a(t, t) \begin{bmatrix} B_t G_{cz} \\ B_{cz} \end{bmatrix} \frac{1}{2} R_t \quad (2-59)$$

The state transition matrix  $\phi_a(t, t)$  is the identity matrix  $I$ . The value of  $\frac{1}{2}$  appears when integrating the Dirac delta function between the time limits  $t_0$  and  $t$  since its argument goes to zero at  $\tau = t$ , i.e., at its upper limit. The resulting form is

$$P_{x_a v_t} = \frac{1}{2} \begin{bmatrix} B_t G_{cz} \\ B_{cz} \end{bmatrix} R_t \quad (2-60)$$

At this point, all quantities needed to obtain the performance analysis for a linear, time-invariant, continuous-time system and controller have been defined in this section. The primary result is that, for a truth model given by Equations (2-17) and (2-24) and a controller of the form (2-25) and (2-26), the statistics of desired outputs are given by (2-47) and (2-48), using (2-42) through (2-45) and (2-60).

#### 2.4 Improving Robustness in Continuous-Time Controllers

The stability robustness of LQG controllers is guaranteed assuming that full-state feedback is available. However, once a Kalman filter is inserted into the loop, all guarantees of robustness, such as minimum gain and phase margins, are lost (Ref 6). J.C. Doyle demonstrated this for a simple case of an observer-based controller in Reference 5.

In Reference 6, Doyle and Stein introduced a method for improving the robustness of a control system that employs an observer or state estimator to generate estimates of states when full-state feedback is unavailable. In many practical systems, this is the case. Their method assumes that the linear, time-invariant system to be controlled is observable, controllable and has no transmission zeros in the right-half s-plane, (i.e., it is minimum phase).

Figure (2-3) shows the structures for a full-state feedback controller and an observer-based controller. Doyle and Stein claim that if

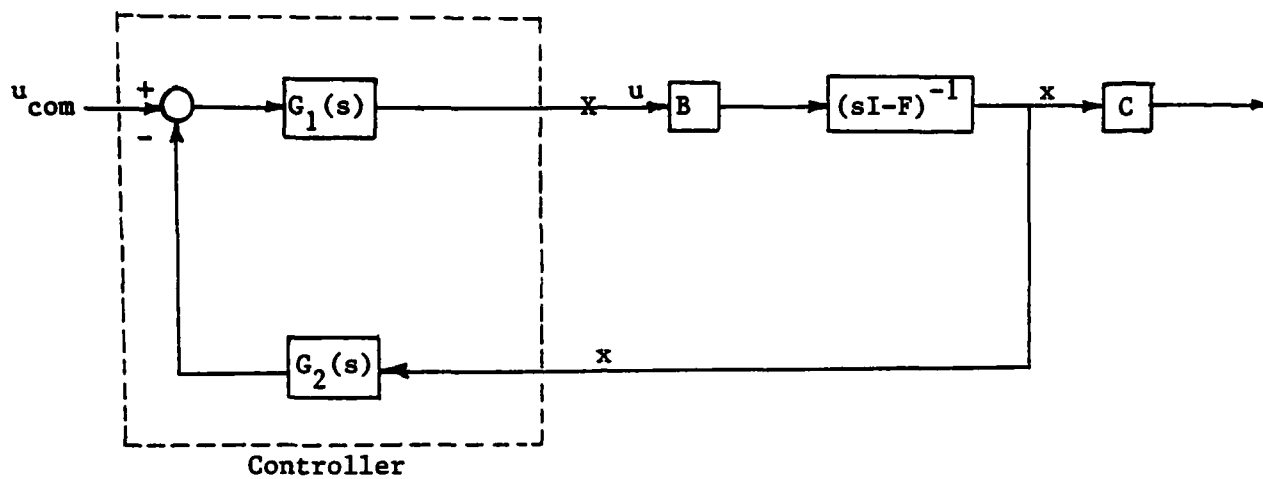


Figure 2-3a: Full-State Feedback System

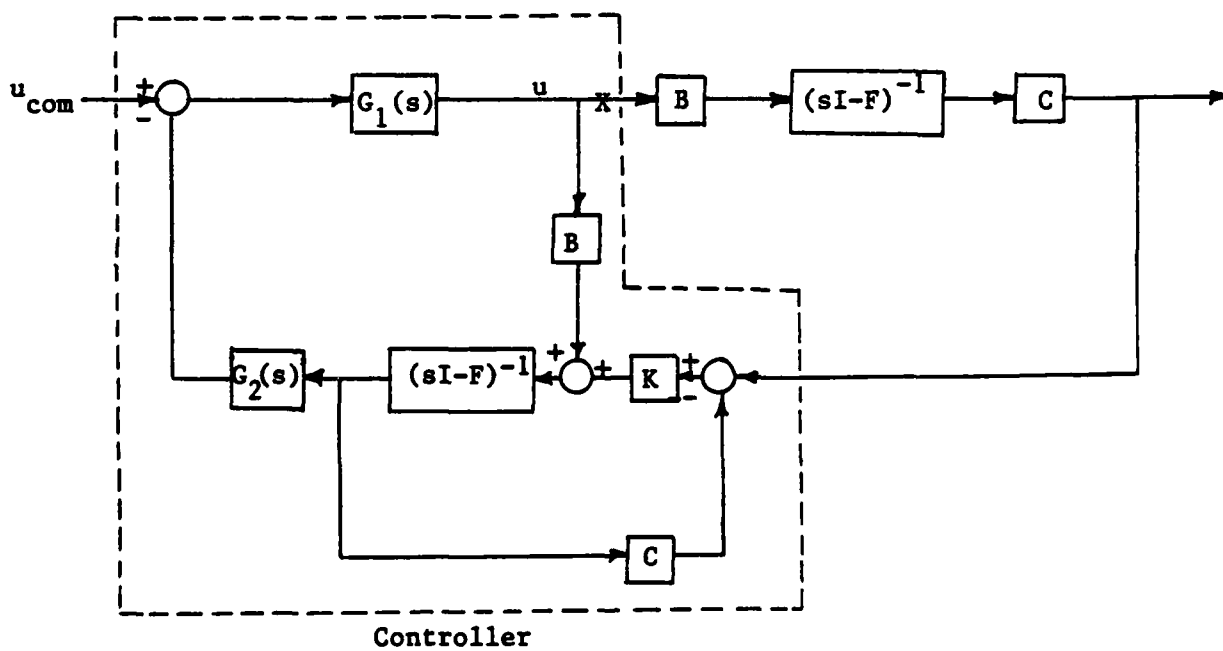


Figure 2-3b: Observer Based Feedback System

the corresponding return-difference mappings are asymptotically equal when the control loops are broken at point  $x$  (the point of entry of the control inputs), then the robustness properties of the observer-based controller will asymptotically approach those of the full-state feedback controller.

The return-difference mappings of Figure (2-3) are equal if the observer satisfies the following equation

$$K[I + H(sI - F)^{-1}K]^{-1} = B[H(sI - F)^{-1}B]^{-1} \quad (2-61)$$

where  $H, F$ , and  $B$  are system matrices and  $K$  is the observer gain. Let  $K$  be parameterized as a function of the scalar variable,  $q$ . Equation (2-61) is satisfied asymptotically as  $q$  approaches infinity if

$$\frac{K(q)}{q} \rightarrow BW \quad (2-62)$$

where  $W$  is any nonsingular matrix. If the observer used is Kalman filter, then  $K(q)$  becomes the Kalman filter gain

$$K(q) = \bar{P}(q)H^T R^{-1} \quad (2-63)$$

and  $\bar{P}(q)$  replaces in Equation (2-16).

To implement the Doyle and Stein technique, the value of  $GQG^T$  in Equation (2-16) must be altered. Let  $Q_0$  be the matrix  $GQG^T$  of the original system (i.e., let  $G = I$ ) and  $Q(q)$  be the modified matrix after the robustification technique is applied. The modified matrix is

$$Q(q) = Q_0 + q^2 BVB^T \quad (2-64)$$



where  $q$  is the design parameter chosen to reflect the amount of desired robustification and  $V$  is a positive definite symmetric matrix. Note that the second term of Equation (2-64) implies adding additional pseudonoise to the system at the point of entry of the control inputs,  $\underline{u}(t)$ , rather than the point of entry of the dynamic driving noise,  $\underline{w}(t)$ . As  $q$  approaches infinity, the observer-based controller recovers the robustness properties of a full-state feedback controller. Note that if  $q$  is zero,  $Q(q)$  is the  $GQG^T$  matrix of the original system.

## 2.5 Sampled-Data Controller

For a continuous-time system, represented by Equation (2-1), having sampled-data measurements, an equivalent discrete-time stochastic difference equation (Ref 23:170) can be written as

$$\underline{x}(t_{i+1}) = \phi(t_{i+1}, t_i) \underline{x}(t_i) + B_d(t_i) \underline{u}(t_i) + G_d(t_i) \underline{w}_d(t_i) \quad (2-65)$$

where  $\underline{w}_d(t_i)$  is a discrete-time white Gaussian noise with statistics

$$E\{\underline{w}_d(t_i)\} = \underline{0} \quad (2-66a)$$

$$E\{\underline{w}_d(t_i) \underline{w}_d^T(t_j)\} = Q_d(t_i) \delta_{ij} \quad (2-66b)$$

and  $\delta_{ij}$  is the Kronecker delta, equal to one if  $i=j$  and zero if  $i \neq j$ .

The matrix  $\phi(t_{i+1}, t_i)$  is the state transition matrix for the system over a single sample period and  $B_d(t_i)$  and  $Q_d(t_i)$  are defined in terms of the continuous-time system matrices to be

$$B_d(t_i) = \int_{t_i}^{t_{i+1}} \phi(t_{i+1}, \tau) B(\tau) d\tau \quad (2-67)$$

$$Q_d(t_i) = \int_{t_i}^{t_{i+1}} \phi(t_{i+1}, \tau) G(\tau) Q(\tau) G^T(\tau) \phi^T(t_{i+1}, \tau) d\tau \quad (2-68)$$

Note that Equation (2-67) inherently assumes  $\underline{u}(t)$  is held constant over a sample period, i.e.,  $\underline{u}(t) = \underline{u}(t_i)$  for all  $t_i$  in the interval  $[t_i, t_{i+1})$ .  $G_d(t_i)$  is defined to be an identity matrix  $I$ . If, as in the continuous-time controller case, a time invariant system model and stationary noises are assumed, and if in addition a fixed sample period is assumed, the integrations of Equation (2-67) and (2-68) need only be performed once and  $B_d$  and  $Q_d$  are constant matrices.

A discrete-time cost function similar to that of Equation (2-3) can be generated as

$$J_c = E \left\{ \frac{1}{2} \underline{x}^T(t_{n+1}) X_f \underline{x}(t_{n+1}) + \sum_{i=0}^n \frac{1}{2} \begin{bmatrix} \underline{x}(t_i) \\ \underline{u}(t_i) \end{bmatrix}^T \begin{bmatrix} X(t_i) & S(t_i) \\ S^T(t_i) & U(t_i) \end{bmatrix} \begin{bmatrix} \underline{x}(t_i) \\ \underline{u}(t_i) \end{bmatrix} \right\} \quad (2-69)$$

In the above equation,  $t_n$  is the last time at which a control is applied. It is assumed that a zero-order-hold is used to interface with the continuous-time system.  $X_f$  and  $X(t_i)$  are positive semi-definite weighting matrices and  $U(t_i)$  is a positive definite weighting matrix.  $S(t_i)$  is chosen so that the composite matrix of Equation (2-69) is positive semi-definite. The weighting matrices are again assumed constant henceforth, in order to obtain constant gain control laws eventually.

The LQ optimal full-state feedback control law is given by

$$u^*(t_i) = -G_c^*(t_i) \underline{x}(t_i) \quad (2-70)$$

where  $G_c^*(t_i)$  is the solution to the equation

$$G_c^*(t_i) = [U + B_d^T K_c(t_{i+1}) B_d]^{-1} [B_d^T K_c(t_{i+1}) \phi + S^T] \quad (2-71)$$

and  $K_c(t_i)$  satisfies the backward recursive Riccati difference equation

$$K_c(t_i) = X + \phi^T K_c(t_{i+1}) \phi - [B_d^T K_c(t_{i+1}) \phi + S^T]^T G_c^*(t_i) \quad (2-72)$$

subject to the final condition

$$K_c(t_{n+1}) = X_f \quad (2-73)$$

As in the case of a continuous-time system having continuous-time measurements, perfect access to all the states usually is not available. Instead, sampled-data measurements may be available in the form of

$$\underline{z}(t_i) = H \underline{x}(t_i) + \underline{v}_d(t_i) \quad (2-74)$$

where  $\underline{v}_d(t_i)$  is a discrete-time white Gaussian noise assumed independent of dynamics driving noise  $\underline{w}_d(t_i)$  in Equation (2-65), with statistics

$$E \{ \underline{v}_d(t_i) \} = \underline{0} \quad (2-75a)$$

$$E \{ \underline{v}_d(t_i) \underline{v}_d^T(t_j) \} = R_d \delta_{ij} \quad (2-75b)$$

Again, a Kalman filter is employed to generate an estimate of the

conditional mean,  $\hat{x}(t_i^+)$ , and conditional covariance,  $P(t_i^+)$ , of the states of the system.

The mean and covariance are propagated between sampled times by these equations (Ref 23:217)

$$\hat{x}(t_{i+1}^-) = \phi \hat{x}(t_i^+) + B_d u(t_i) \quad (2-76a)$$

$$P(t_{i+1}^-) = \phi P(t_i^+) \phi^T + G_d Q_d G_d^T \quad (2-76b)$$

Then, at the sample times when measurements are taken, the mean and covariance are updated by (Ref 23:217)

$$\hat{x}(t_i^+) = \hat{x}(t_i^-) + K(t_i) [z(t_i) - H \hat{x}(t_i^-)] \quad (2-77)$$

$$P(t_i^+) = P(t_i^-) - K(t_i) H P(t_i^-) \quad (2-78)$$

where  $K(t_i)$  is the Kalman filter gain, given by

$$K(t_i) = P(t_i^-) H^T [H P(t_i^-) H^T + R_d]^{-1} \quad (2-79)$$

The superscripts, - and +, refer to quantities just before and just after the measurements at the samples times are taken, respectively.

The mean and covariance are propagated and updated forward in time by the above discrete-time Kalman filter equations subject to the initial conditions

$$E\{\underline{x}(t_0)\} = \bar{x}_0 \quad (2-80a)$$

$$E\{[\underline{x}(t_0) - \bar{\underline{x}}_0][\underline{x}(t_0) - \bar{\underline{x}}_0]^T\} = P_0 \quad (2-80b)$$

which are obtained from an a priori Gaussian density function of the states at the initial time, as discussed below Equation (2-13).

As in the continuous-measurement case, it is desired to generate a constant-gain, steady-state LQG controller. Thus, steady-state solutions from Equations (2-72), (2-76b) and (2-78) are desired to generate a constant Kalman filter gain matrix,  $K$ , and a constant feedback gain matrix,  $G_c^*$ .

The form of the sampled-data controller is shown in Figure (2-4).

## 2.6 Sampled-Data Performance Analysis

The performance analysis for a sampled-data LQG controller is based on the development in Reference 24:Ch 14.

Similar to Section 2.3, a truth model for a given system is developed which has the following form

$$\dot{\underline{x}}_t(t) = F_t \underline{x}_t(t) + B_t \underline{u}(t) + G_t \underline{w}_t(t) \quad (2-81a)$$

$$\underline{z}_t(t_i) = H_t \underline{x}_t(t_i) + \underline{v}_{dt}(t_i) \quad (2-81b)$$

The measurements are now in sampled-data form, and  $\underline{v}_{dt}(t_i)$  is a discrete time white Gaussian noise of covariance  $R_{dt}$ .

As shown in Figure (2-5), proposed sampled-data controllers can be inserted into the truth model simulation of the real world and their performance evaluated.

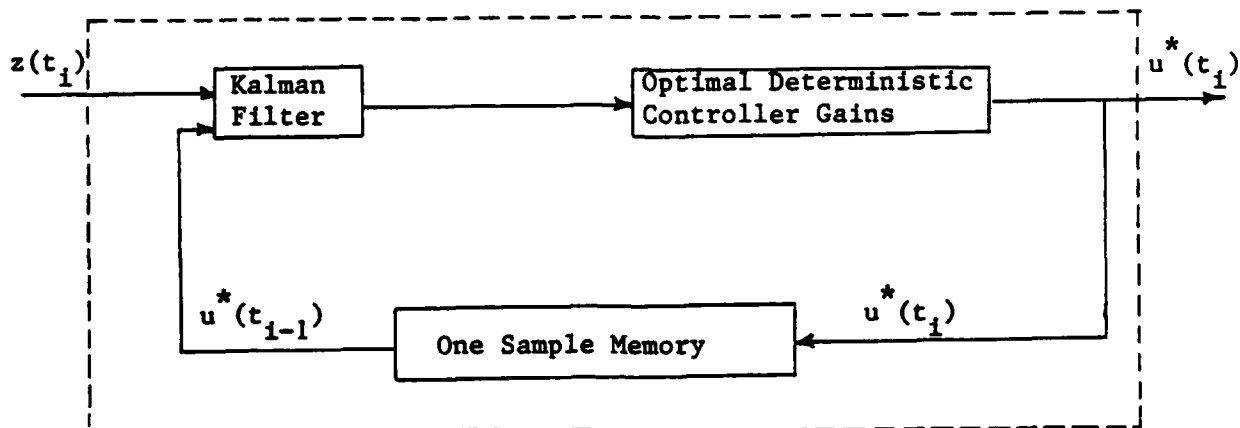


Figure 2-4a: Overall Structure of Sampled-Data LQG Regulator

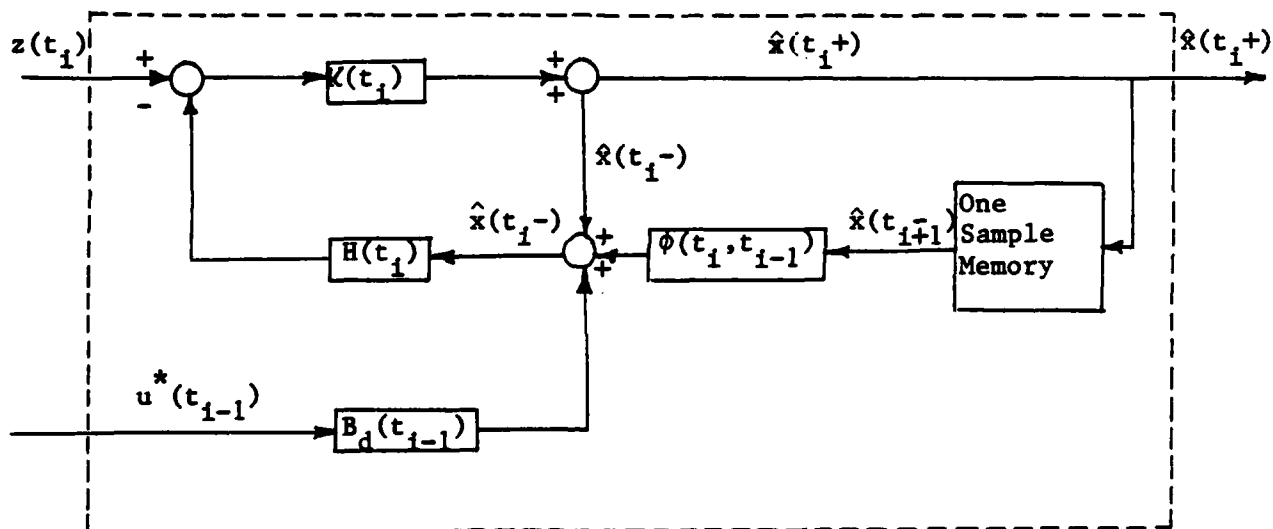


Figure 2-4b: Kalman Filter Block in (A)

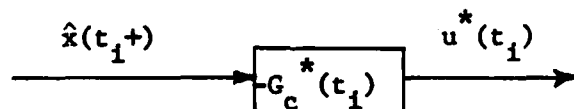


Figure 2-4c: Optimal Deterministic Controller Block in (A)

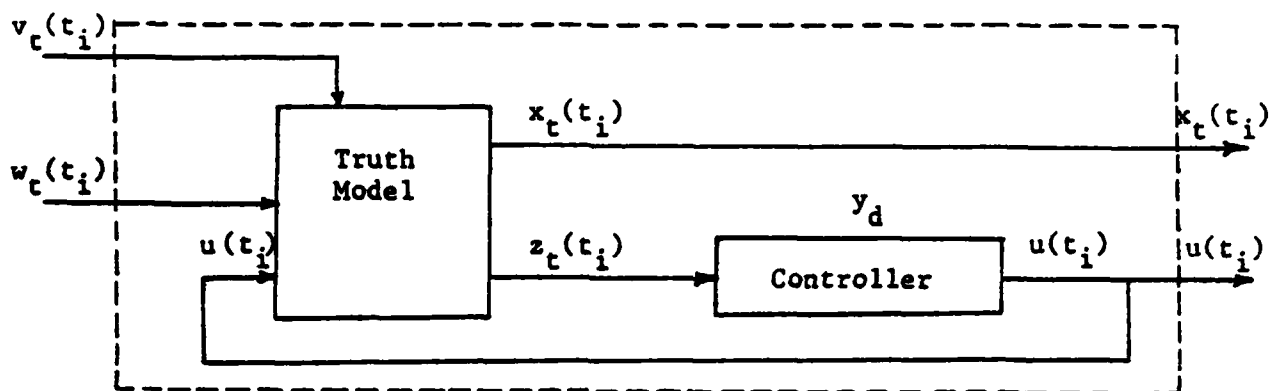


Figure 2-5: Performance Evaluation for Linear Sampled-Data Controller

For each controller, it is desired to evaluate a control law of the generic form

$$\underline{u}(t_i) = G_{cx} \underline{x}_c(t_i) + G_{cz} \underline{z}_t(t_i) + G_{cy} \underline{y}_d(t_i) \quad (2-82)$$

where  $\underline{u}(t_i)$  uses measurements up to and including the  $i$ th measurement and is held constant over the  $i$ th sample period. The states of the controller can be propagated by the linear difference equation

$$\underline{x}_c(t_{i+1}) = \phi_{c-c} \underline{x}_c(t_i) + B_{cz} \underline{z}_t(t_i) + B_{cy} \underline{y}_d(t_i) \quad (2-83)$$

Note the analogous form of the above two equations to that of Equation (2-25) and (2-26). Expressions for the gains multiplying  $\underline{x}_c(t_i)$ ,  $\underline{z}_t(t_i)$  and  $\underline{y}_d(t_i)$  are given in Appendix A. Initial conditions for the controller states are usually assumed zero.

The performance of a controller is evaluated by generating time histories of the mean and covariance of the augmented Gaussian vector

$$\underline{y}_a(t_1) = \begin{bmatrix} \underline{x}_t(t_1) \\ \underline{u}(t_1) \end{bmatrix} \quad (2-84)$$

and the corresponding vector between sample times. This is developed subsequent to a description pertaining to sample times only.

First, as in Section 2.4, it is necessary to evaluate the statistics of a second augmented vector

$$\underline{x}_a(t_1) = \begin{bmatrix} \underline{x}_t(t_1) \\ \underline{x}_c(t_1) \end{bmatrix} \quad (2-85)$$

By performing the integrations indicated in Equations (2-67) and (2-68), an equivalent discrete-time equation can be generated for Equation (2-81a). Then by substituting Equations (2-81b) and (2-82) into the equations for the truth model and controller states, Equations (2-81a) and (2-83) can be rewritten as

$$\begin{aligned} \underline{x}_t(t_{i+1}) = & [\phi_t + B_{dt} G_{cz} H_t] \underline{x}_t(t_i) \\ & + B_{dt} G_{cx} \underline{x}_c(t_i) + B_{dt} G_{cy} \underline{y}_d(t_i) \\ & + I \underline{w}_{dt}(t_i) + B_{dt} G_{cz} \underline{v}_{dt}(t_i) \end{aligned} \quad (2-86)$$

$$\begin{aligned} \underline{x}_c(t_{i+1}) = & \phi_c \underline{x}_c(t_i) + B_{cz} H_t \underline{x}_t(t_i) \\ & + B_{cy} \underline{y}_d(t_i) + B_{cz} \underline{v}_{dt}(t_i) \end{aligned} \quad (2-87)$$



The subscript d refers to a discrete quantity,

Form the augmented, stationary, zero-mean white Gaussian noise vector

$$\underline{w}_{da}(t_i) = \begin{bmatrix} \underline{w}_{dt}(t_i) \\ \underline{v}_{dt}(t_i) \end{bmatrix} \quad (2-88)$$

with covariance

$$Q_{da} = \begin{bmatrix} Q_{dt} & 0 \\ 0 & R_{dt} \end{bmatrix} \quad (2-89)$$

If  $\underline{w}_{dt}(t_i)$  and  $\underline{v}_{dt}(t_i)$  are correlated when off-diagonal terms can be added to Equation (2-89).

Now an augmented system equation can be written by combining Equations (2-86) and (2-87)

$$\underline{x}_a(t_{i+1}) = \phi_a \underline{x}(t_i) + B_{da} \underline{y}_d(t_i) + G_{da} \underline{w}_{da}(t_i) \quad (2-90)$$

where

$$\phi_a = \begin{bmatrix} \phi_t + B_{dt} G_{cz} H_t & B_{dt} G_{cx} \\ B_{cz} H_t & \phi_c \end{bmatrix} \quad (2-91)$$

$$B_{da} = \begin{bmatrix} B_{dt} G_{cy} \\ B_{cy} \end{bmatrix} \quad (2-92)$$

$$G_{da} = \begin{bmatrix} I & B_{dt} G_{cz} \\ 0 & B_{cz} \end{bmatrix} \quad (2-93)$$

The initial conditions for the augmented system are given by

$$E\{\underline{x}_a(t_0)\} = \begin{bmatrix} \underline{\bar{x}}_0 \\ 0 \end{bmatrix} = \underline{\bar{x}}_{ao} \quad (2-94)$$

$$E\{[\underline{x}_a(t_0) - \underline{\bar{x}}_{ao}][\underline{x}_a(t_0) - \underline{\bar{x}}_{ao}]^T\} = \begin{bmatrix} P & 0 \\ 0 & 0 \end{bmatrix} \quad (2-95)$$

If desired, a cost function for the augmented system can be formed, as shown in References 24:Ch 14;12..

The mean and covariance of the augmented vector,  $\underline{x}_a(t_1)$  can be propagated by

$$\underline{m}_{\underline{x}_a}(t_1) = \phi_a \underline{m}_{\underline{x}_a}(t_1) + B_{da} \underline{y}_d(t_1) \quad (2-96)$$

$$P_{\underline{x}_a \underline{x}_a}(t_1) = \phi_a P_{\underline{x}_a \underline{x}_a} \phi_a^T + G_{da} Q_{da} G_{da}^T \quad (2-97)$$

The vector of interest,  $\underline{y}_a(t_1)$ , is related to  $\underline{x}_a(t_1)$  by the following equation

$$\underline{y}_a(t_1) = \begin{bmatrix} \underline{x}_t(t_1) \\ \underline{u}(t_1) \end{bmatrix} = \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cz} \end{bmatrix} \begin{bmatrix} \underline{x}_t(t_1) \\ \underline{x}_c(t_1) \end{bmatrix} + \begin{bmatrix} 0 \\ G_{cy} \end{bmatrix} \underline{y}_d(t_1) + \begin{bmatrix} 0 \\ G_{cz} \end{bmatrix} \underline{v}_{dt}(t_1) \quad (2-98)$$

and the desired statistics are therefore generated by

$$\underline{m}_{y_a}(t_1) = \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} \underline{m}_{x_a}(t_1) + \begin{bmatrix} 0 \\ G_{cy} \end{bmatrix} \underline{y}_d(t_1) \quad (2-99)$$

$$\begin{aligned} P_{y_a y_a}(t_1) &= \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix} P_{x_a x_a}(t_1) \begin{bmatrix} I & 0 \\ G_{cz}H_t & G_{cx} \end{bmatrix}^T \\ &+ \begin{bmatrix} 0 \\ G_{cz} \end{bmatrix} R_{dt} \begin{bmatrix} 0 & G_{cz}^T \end{bmatrix} \end{aligned} \quad (2-100)$$

Equations (2-99) and (2-100) provide an efficient means for evaluating statistics at the sample times via the equivalent discrete-time system model. However, it may be desired to obtain more complete results between sample times, to ensure that adequate sample period choice and

system control have been achieved. Between sample instants, a differential equation for  $\underline{y}_a(t)$  can be written as

$$\dot{\underline{y}}_a(t) = \begin{bmatrix} F_t & B_t \\ 0 & 0 \end{bmatrix} \underline{y}_a(t) + \begin{bmatrix} G_t \\ 0 \end{bmatrix} \underline{w}_t(t) \quad (2-101)$$

and the mean and covariance can be propagated from  $t_i$  to  $t_{i+1}$ , where  $\underline{u}(t_i)$  undergoes a step change, using the initial conditions given by Equations (2-90) and (2-91). The propagation equations are

$$\dot{\underline{m}}_{y_a}(t) = \begin{bmatrix} F_t & B_t \\ 0 & 0 \end{bmatrix} \underline{m}_{y_a}(t) \quad (2-102)$$

$$\begin{aligned} \dot{P}_{y_a y_a}(t) = & \begin{bmatrix} F_t & B_t \\ 0 & 0 \end{bmatrix} P_{y_a y_a}(t) + P_{y_a y_a}(t) \begin{bmatrix} F_t & B_t \\ 0 & 0 \end{bmatrix}^T \\ & + \begin{bmatrix} GQG^T & 0 \\ 0 & 0 \end{bmatrix} \end{aligned} \quad (2-103)$$

At this point, all quantities of interest have been defined to evaluate the performance of a sampled-data controller. The primary results are that for a truth model given by (2-81) and a controller of the form (2-82) and (2-83), the statistics of desired outputs at the sample times are given by (2-99) and (2-100). Between sample times the statistics are given by (2-102) and (2-103).

## 2.7 Improving Robustness in Discrete-Time Systems

Three approaches are taken to apply the robustification technique described in Section 2.3 to sampled-data controllers. The first is simply to discretize the LQG controller designed for the continuous-time, continuous-measurement system. The second approach is to extend the technique of adding pseudonoise to the points of entry of  $\underline{u}$  to a sampled-data system by making a first order (or better) approximation to the modified Q matrix. The third is to apply the fundamental conditions under which full-state feedback characteristics are obtained asymptotically from a controller with a filter or observer in the loop.

It has been observed in Reference 21 that, unlike the continuous-measurement case, the scalar parameter  $q$  of Equation (2-64) cannot be adjusted arbitrarily upwards for the discrete-time case. Rather, there is a finite range of values for  $q$  that will robustify the Kalman filter, and beyond that range the closed-loop system is unstable.

### 2.7.1 Discretizing the Continuous-Time LQG Controller

The required format for the discretized controller is given by Equation (2-82) which requires values for  $G_{cx}(t_i)$ ,  $G_{cz}(t_i)$  and  $G_{cy}(t_i)$ , and by Equation (2-83) which requires values for  $\phi_c(t_{i+1}, t_i)$ ,  $B_{cz}(t_i)$  and  $B_{cy}(t_i)$ . Recall that the continuous-time controller equations are expressed as

$$\underline{u}(t) = G_{cx}(t) \underline{x}_c(t) + G_{cz}(t) \underline{z}(t) + G_{cy}(t) \underline{y}_d(t) \quad (2-104)$$

$$\dot{\underline{x}}_c(t) = F_c(t) \underline{x}_c(t) + B_{cz}(t) \underline{z}(t) + B_{cy}(t) \underline{y}_d(t) \quad (2-105)$$

In Appendix A, it is shown that for a steady-state constant-gain LQG regulator

$$G_{cx}(t) = G_c^* \quad (2-106a)$$

$$G_{cz}(t) = 0 \quad (2-106b)$$

$$G_{cy}(t) = 0 \quad (2-106c)$$

$$F_c(t) = F - BG_c^* - KH \quad (2-106d)$$

$$G_{cz}(t) = K \quad (2-106e)$$

$$B_{cy}(t) = 0 \quad (2-106f)$$

For this problem,  $G_c^*(t)$  is a constant value  $G_c^*$ , and thus the discretized control law is given by

$$\underline{u}(t_i) = -G_c^* \underline{x}_c(t_i) \quad (2-107)$$

To obtain a discrete propagation equation for the controller states, first-order approximations for  $\phi_c(t_{i+1}, t_i)$ ,  $B_{cz}(t_i)$  and  $B_{cy}(t_i)$  are obtained by

$$\phi_c(t_{i+1}, t_i) = [I + F_c(t_i)\Delta t] \quad (2-108a)$$

$$B_{czd}(t_i) = B_{cz}(t_i)\Delta t \quad (2-108b)$$

$$B_{cyd}(t_i) = B_{cy}(t_i)\Delta t \quad (2-108c)$$

Again, for this thesis, the system is assumed time invariant and  $F_c$ ,  $B_{cz}$ , and  $B_{cy}$  are constant matrices. They are evaluated explicitly in Appendix A. The quantity  $\Delta t$  is the sample period for the sampled-data system. Higher order approximations are also possible, but if  $\Delta t$  is not sufficiently small compared to the transient times of the system, this basic idea breaks down anyway, so its use is confined to problems with short sample periods

In addition, a discrete approximation for  $R_{dt}(t_i)$  is needed for use in the performance analysis described in Section 2.6. If the sample time is sufficiently small to allow the approximations in Equation (2-108), then it is reasonable to make a first-order approximation for the discrete measurement noise covariance  $R_{dt}(t_i)$ . This is given by

$$R_{dt}(t_i) = R_t(t_i)/\Delta t \quad (2-109)$$

Notice that as  $\Delta t \rightarrow 0$ , the discrete-time white noise  $\underline{v}_{dt}(t_i)$  of Equation (2-81b) converges to the continuous-time white noise described by (2-18) and (24) if (2-109) is satisfied throughout the limiting process.

### 2.7.2 Doyle and Stein Technique for Sampled-Data Systems

The second approach to enhancing robustness is to extend the Doyle and Stein technique and apply it directly to a sampled-data controller. Several ways of accomplishing this are considered.

If the sampled period of the system is sufficiently small, a first-order approximation for the discrete-time noise strength matrix  $Q_d(q)$ ,

can be made which is the discrete counterpart for Equation (2-64).

This yields

$$Q_d(q) + Q_{do} + q^2 BVB^T \Delta t \quad (2-110)$$

where  $Q_{do}$  is

$$Q_{do} = Q_d(t_1) \quad (2-111)$$

which is defined in Equation (2-68). Again, this implies that  $G_d = I$ .

For larger sample times, a first-order approximation may not be sufficient. If this is the case, subintervalling may be considered a higher order approximation for  $Q_d(q)$ , such as

$$Q_d(q) = \frac{1}{2} [\phi Q(q) + Q(q) \phi^T] \Delta t \quad (2-112)$$

as given in Reference 24:172.

Another alternative is to make the return-difference mapping for the observer-based controller asymptotically equal to that of the full-state feedback controller in the discrete-time case. The two configurations are shown in Figure (2-6). The observer-based system is based upon the sub-optimal control law

$$\underline{u}(t_1) = -G_c^* \hat{\underline{x}}(t_1^-) \quad (2-113)$$

rather than the law given by Section 2.5 using  $\hat{\underline{x}}(t_1^+)$ .

To recover the robustness properties of the full-state feedback system,  $[\phi K]$  must be found such that

$$\phi K [I + H(zI - \phi)^{-1} \phi K]^{-1} = B_d [H(zI - \phi)^{-1} B_d]^{-1} \quad (2-114)$$



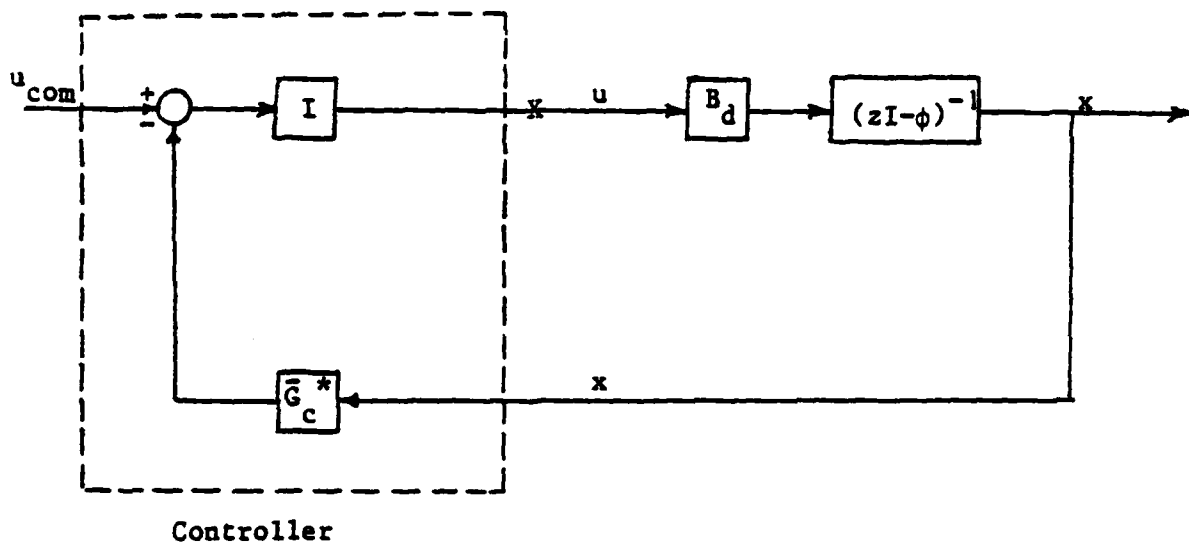


Figure 2-6a: Full-State Feedback System

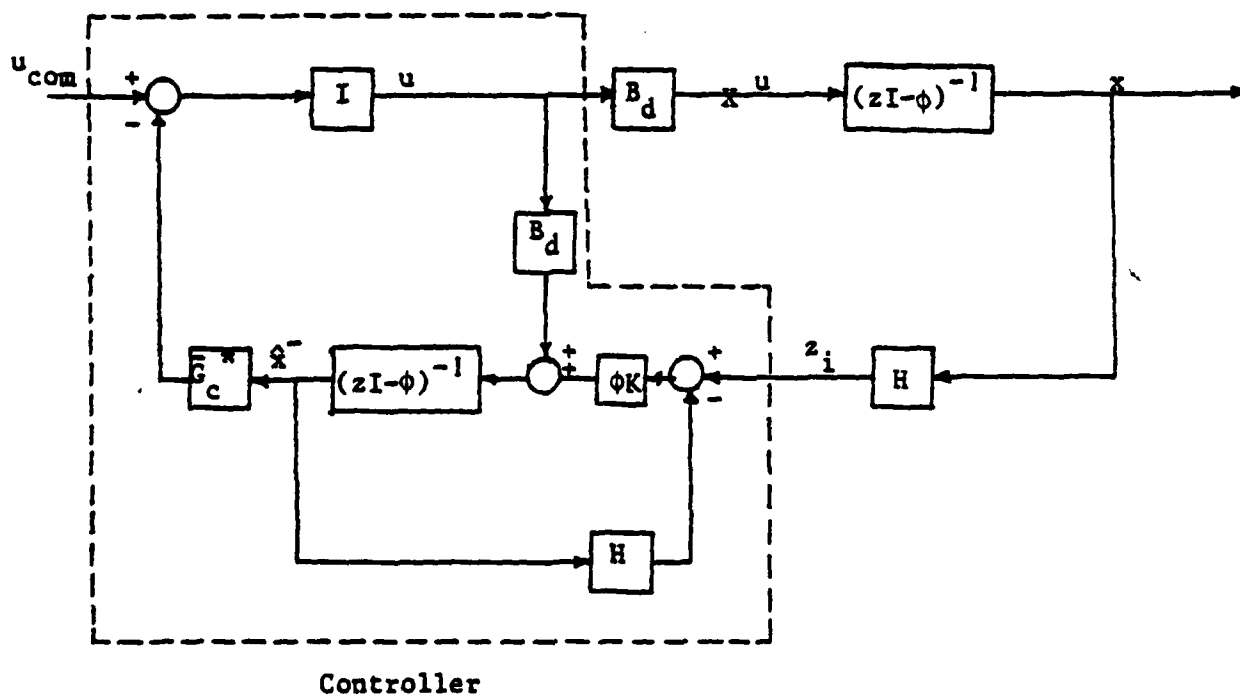


Figure 2-6b: Observer-Based Feedback System

where the analysis is carried out in the z-domain. Equation (2-114) is satisfied asymptotically when the Kalman filter gain,  $K$ , is parameterized as  $K(q)$ , and

$$\lim_{q \rightarrow \infty} \frac{\phi K}{q} = B_d W \quad (2-115)$$

where  $W$  is a nonsingular  $m \times m$  matrix. Therefore, for finite values of  $q$ ,  $K$  is given by

$$K = q \phi^{-1} B_d W \quad (2-116)$$

Unfortunately, this does not lend itself to an interpretation of retuning via pseudonoise addition, as in the continuous-time case.

Reference 24:113-114 suggests looking at the dual state equations to select  $W$ . This yields

$$W = [H \phi^{-1} B_d]^{-1} \quad (2-117)$$

which assigns  $m$  eigenvalues of the closed loop dual system to the origin and the remaining  $(n-m)$  eigenvalues to the invariant zeros of the system where  $n$  is the number of states and  $m$  is the number of controls.

To use the performance analysis equations, the sub-optimal control must be put into the generic format of Equations (2-82) and (2-83). This is shown explicitly in Appendix A.

## 2.8 Summary

The preceding sections have presented the equations for designing and evaluating optimal LQG regulators. Three applications are considered:

a continuous-time controller, a discretized continuous-time controller and a sampled-data controller. The structure of these controllers was examined with and without a Kalman filter in the loop to estimate states.

For the case where a Kalman filter is necessary to estimate states, a technique was presented to robustify a controlled system against differences that exist between the controller design model and the real world system. The technique was developed for a continuous-time system and several extensions to discrete-time systems were presented.

A disadvantage of the type of controller considered in this chapter is that it will not regulate the states to zero in the face of unmodeled disturbances. Also, it only regulates the deviations in the states from an equilibrium position and does not allow the system to track a desired piecewise constant non-zero input. Chapter III presents the development for a type of controller that does exhibit these characteristics, based on an LQG methodology.

### III. PI CONTROLLERS

#### 3.1 Introduction

This chapter presents the equations used for designing Proportional-plus-Integral (PI) controllers for a system modeled as linear, time-invariant, and driven by stationary white Gaussian noise and deterministic inputs. The model used for this thesis is described in Chapter V.

The advantage of a PI controller over the regulator described in Chapter II is that, in the absence of stochastic inputs, it will maintain the system output at some nonzero commanded value with zero steadystate error, even in the presence of unmodeled constant disturbances. This is known as a "Type-1" property (Ref 4). This property is achieved by integration of the error in the controlled variables, i.e., integration of the difference between achieved and desired values of the controlled variables. In discrete-time feedback control, the PI effect is achieved by performing summation (or pseudo-integration, often interpreted as Euler integration) on either the difference between the actual and desired system outputs (regulation error). In this thesis, this structure is obtained via LQG synthesis applied to an augmented system composed of the original system states and the controls (Ref 24:141-150). Two possible implementations are "position form" and "incremental form". In position form, the control  $\underline{u}(t_i)$  is specified in terms of the current position of the system state,  $\underline{x}(t_i)$ , as in the LQ regulator solution of the previous chapter. In incremental form, only changes in states and commands from the previous values are used to generate increments in control relative to the value at the previous sample time. For implementation, an incremental form PI controller is generally preferable because it is not necessary to provide initial

values for the controller states as in the position form. Also it lends itself more readily to relinearizations about new nominal values and to anti-windup compensation (Ref 24).

The optimal deterministic (LQ) PI controller is developed in the first section of this chapter for a sampled-data system. Next, the Doyle and Stein technique is applied to a PI controller. Then the controller is put into the generic format presented in Chapter II, with a Kalman filter in the loop to provide estimates of the states to the controller. Once the generic form is available, the performance of the controller may be evaluated using the method of the previous chapter.

The software developed to design optimal PI controllers is described in Reference 13. Modifications to the software are described in References 26, 27 and 29. The performance analysis software is described in Reference 29. The above references deal primarily with the design of controllers that employ a PI regulator in the inner feedback loop and a Command Generator Tracker (CGT) to provide inputs to the system (Ref 12). This chapter contains the development of only the inner loop PI regulator.

### 3.2 Sampled-Data PI controller

The following development for a PI controller, based on augmentation to the original system state relations of equations for pseudo-integration of the control input rates, is found in Reference 24:Ch 14.

#### 3.2.1 Control-Rate Pseudo Integration

Recall from Chapter II that an equivalent discrete-time difference equation can be generated for the continuous-time system model in the form of

$$\underline{x}(t_{i+1}) = \phi \underline{x}(t_i) + B_d \underline{u}(t_i) \quad (3-1)$$

where, by the certainty equivalence principle (Ref 24:Ch 13), the noise vector was deleted.

Define the perturbation state and control variables to be

$$\delta \underline{x}(t_i) = \underline{x}(t_i) - \underline{x}_0 \quad (3-2a)$$

$$\delta \underline{u}(t_i) = \underline{u}(t_i) - \underline{u}_0 \quad (3-2b)$$

where  $\underline{x}_0$  and  $\underline{u}_0$  are the nominal values of the states and the controls to maintain the system at its equilibrium "trim" operating condition such that controlled variables assume the desired values. If an equation for the output of the system is given by

$$\underline{y}(t_i) = C \underline{x}(t_i) + D_y \underline{u}(t_i) \quad (3-3)$$

then the nominal control,  $\underline{u}_0$ , to hold the system at that equilibrium operating point is found as the solution to

$$\underline{x}_0 = \phi \underline{x}_0 + B_d \underline{u}_0 \quad (3-4a)$$

$$\underline{y}_d = C \underline{x}_0 + D_y \underline{u}_0 \quad (3-4b)$$

or

$$\begin{bmatrix} \underline{0} \\ \underline{y}_d \end{bmatrix} = \begin{bmatrix} (\phi - I) & B_d \\ C & D_y \end{bmatrix} \begin{bmatrix} \underline{x}_0 \\ \underline{u}_0 \end{bmatrix} \quad (3-5)$$

where  $\underline{y}_d$  is the desired system output. Then, (3-5) can be solved as

$$\begin{bmatrix} \underline{x}_0 \\ \underline{u}_0 \end{bmatrix} = \begin{bmatrix} (\phi - I) & B_d \\ C & D_y \end{bmatrix}^{-1} \begin{bmatrix} \underline{0} \\ \underline{y}_d \end{bmatrix} = \begin{bmatrix} \pi_{11} & \pi_{12} \\ \pi_{21} & \pi_{22} \end{bmatrix} \begin{bmatrix} \underline{0} \\ \underline{y}_d \end{bmatrix} \quad (3-6)$$

or

$$\underline{x}_0 = \pi_{12} \underline{y}_d \quad \underline{u}_0 = \pi_{22} \underline{y}_d \quad (3-7)$$

The difference in perturbation control variable is

$$\delta \underline{u}(t_{i+1}) - \delta \underline{u}(t_i) = [\underline{u}(t_{i+1}) - \underline{u}_0] - [\underline{u}(t_i) - \underline{u}_0] \quad (3-8)$$

or, equivalently

$$\delta \underline{u}(t_{i+1}) = \delta \underline{u}(t_i) + [\underline{u}(t_{i+1}) - \underline{u}(t_i)] \quad (3-9)$$

The above equation has the form of an update relation for  $\delta \underline{u}(t_{i+1})$ . If the second term is thought of as an Euler integration of the time rate-of-change of the control input  $\underline{u}$  at time  $t_i$ , then the change in  $\underline{u}$ , i.e., the bracketed term in Equation (3-9), can be written as

$$\Delta \underline{u}(t_i) = [\underline{u}(t_{i+1}) - \underline{u}(t_i)] = \dot{\underline{u}}(t_i) \Delta t \quad (3-10)$$

where  $\Delta t$  is the sample time of the controller. Thus,  $\Delta \underline{u}(t_i)$  is termed the control pseudo-rate and Equation (3-9) becomes

$$\delta \underline{u}(t_{i+1}) = \delta \underline{u}(t_i) + \Delta \underline{u}(t_i) \quad (3-11)$$

At this point, an augmented state description can be written using Equations (3-1) and (3-11) for the perturbation variables  $\delta \underline{x}(t_i)$  and  $\delta \underline{u}(t_i)$ :

$$\begin{bmatrix} \delta \underline{x}(t_{i+1}) \\ \delta \underline{u}(t_{i+1}) \end{bmatrix} = \begin{bmatrix} \phi & B_d \\ 0 & I \end{bmatrix} \begin{bmatrix} \delta \underline{x}(t_i) \\ \delta \underline{u}(t_i) \end{bmatrix} + \begin{bmatrix} 0 \\ I \end{bmatrix} \Delta \underline{u}(t_i) \quad (3-12)$$

The control pseudo-rate is now considered to be the input to this augmented perturbation state equation.

### 3.2.2 Optimal Regulator Solution

The optimal regulator solution described in Section 2.5 can now be applied to the augmented system above, subject to a quadratic cost criterion

$$J_c = \frac{1}{2} \begin{bmatrix} \delta \underline{x}(t_{N+1}) \\ \delta \underline{u}(t_{N+1}) \end{bmatrix}^T \begin{bmatrix} X_f & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \delta \underline{x}(t_{N+1}) \\ \delta \underline{u}(t_{N+1}) \end{bmatrix} + \frac{1}{2} \sum_{i=1}^N \begin{bmatrix} \delta \underline{x}(t_i) \\ \delta \underline{u}(t_i) \\ \Delta \underline{u}(t_i) \end{bmatrix}^T \begin{bmatrix} X_{11} & X_{12} & S_1 \\ X_{12}^T & X_{22} & S_2 \\ S_1^T & S_2^T & U \end{bmatrix} \begin{bmatrix} \delta \underline{x}(t_i) \\ \delta \underline{u}(t_i) \\ \Delta \underline{u}(t_i) \end{bmatrix} \quad (3-13)$$

where  $X_{11}$  places a weighting on state deviations away from the nominal  $\underline{x}_0$ ,  $X_{22}$  places a weighting on control deviations away from  $\underline{u}_0$ , and  $U$  places a weighting on the control pseudo-rates,  $\Delta \underline{u}(t_i)$ . Notice that this term allows the designer to place a weighting on the rate of change of control inputs, i.e., to guard against commanding actuator rates that are beyond the physical limits of the actuators.



The term  $X_f$  applies a weighting on deviations of the states at the final time. The cross-terms  $X_{12}$ ,  $S_1$  and  $S_2$  arise as the continuous-time cost is converted to a discrete-time cost, as demonstrated in Reference 24 and 12. In Equation (3-13), notice that the index for the summation begins at -1. This places a weighting on the potentially large control differences which may occur at the initial time due to a change in the setpoint (Ref 24:142).

To apply the optimal regulator solution to Equations (3-12) and (3-13), redefine the matrices as follows

$$\phi_\delta = \begin{bmatrix} \phi & B_d \\ 0 & I \end{bmatrix} \quad (3-14)$$

$$B_d = \begin{bmatrix} 0 \\ I \end{bmatrix} \quad (3-15)$$

$$\delta \underline{x}_\delta = \begin{bmatrix} \delta \underline{x} \\ \delta \underline{u} \end{bmatrix} \quad (3-16)$$

$$X_{f\delta} = \begin{bmatrix} X_f & 0 \\ 0 & 0 \end{bmatrix} \quad (3-17)$$

$$X_{\delta} = \begin{bmatrix} X_{11} & X_{12} \\ X_{12}^T & X_{22} \end{bmatrix} \quad (3-18)$$

$$S_{\delta} = \begin{bmatrix} S_1 \\ S_2 \end{bmatrix}$$

Thus, the state equation and cost function become

$$\delta \underline{x}_{\delta}(t_{i+1}) = \phi_{\delta} \delta \underline{x}_{\delta}(t_i) + B_{\delta} \Delta u(t_i) \quad (3-20)$$

$$J_c = \frac{1}{2} \delta \underline{x}_{\delta}^T(t_{N+1}) X_{f\delta} \delta \underline{x}_{\delta}(t_{N+1}) + \frac{1}{2} \sum_{i=0}^N \begin{bmatrix} \delta \underline{x}_{\delta}(t_i) \\ \Delta u(t_i) \end{bmatrix}^T \begin{bmatrix} X_{\delta} & S_{\delta} \\ S_{\delta}^T & U \end{bmatrix} \begin{bmatrix} \delta \underline{x}_{\delta}(t_i) \\ \Delta u(t_i) \end{bmatrix} \quad (3-21)$$

Then, the optimal, constant-gain LQ regulator solution is given by

$$\Delta u(t_i) = -G_c^* \delta \underline{x}_{\delta}(t_i) \quad (3-22)$$

where

$$G_c^* = [G_{c1}^* \quad G_{c2}^*] = [U + B_{\delta}^T \bar{K}_c B_{\delta}]^{-1} [B_{\delta}^T \bar{K}_c \phi_{\delta} + S_{\delta}^T] \quad (3-23)$$

and  $\bar{K}_c$  is the steady-state solution to the backward Riccati difference equation

$$K_c(t_i) = X_\delta + \phi_\delta^T K_c(t_{i+1}) [\phi_\delta - B_\delta G_c^*] - S_\delta G_c^* \quad (3-24)$$

solved backwards from the terminal condition

$$K_c(t_{N+1}) = X_{f\delta} \quad (3-25)$$

The optimal control input is given by Equation (3-22) in partitioned form as

$$\Delta \underline{u}^*(t_i) = -G_{c_1}^* \delta \underline{x}(t_i) - G_{c_2}^* \delta \underline{u}(t_i) \quad (3-26)$$

Combining this with Equation (3-11) yields

$$\delta \underline{u}^*(t_{i+1}) = \delta \underline{u}^*(t_i) - [G_{c_1}^* \quad G_{c_2}^*] \begin{bmatrix} \delta \underline{x}(t_i) \\ \delta \underline{u}^*(t_i) \end{bmatrix} \quad (3-27)$$

Unfortunately, the above control law does not achieve the desired Type 1 property.

### 3.3 Achieving Type-1 Control

The desired integral characteristics for the controller can be achieved by manipulation into the form of a continuous-time PI controller (Ref 1)

$$\underline{u}^*(t) = -K_x \underline{x}(t) + K_z \int_{t_0}^t [\underline{y}_d - \underline{y}(\tau)] d\tau \quad (3-28)$$

where  $\underline{y}_d$  is the desired output and  $\underline{y}(t)$  is the actual system output defined by

$$\underline{y}(t) = C\underline{x}(t) + D_y \underline{u}(t) \quad (3-29)$$

A discrete-time equivalent is given by

$$\underline{u}^*(t_i) = -K_x \underline{x}(t_i) + K_z \sum_{j=-1}^{i-1} [\underline{y}_d - \underline{y}(t_j)] \quad (3-30)$$

In incremental form, the above equation is expressed as

$$\underline{u}^*(t_{i+1}) = \underline{u}^*(t_i) - K_x [\underline{x}(t_{i+1}) - \underline{x}(t_i)] + K_z [\underline{y}_d - \underline{y}(t_i)] \quad (3-31)$$

Define a perturbation output equation as

$$\delta \underline{y}(t_i) = \underline{y}(t_i) - \underline{y}_d = C \delta \underline{x}(t_i) + D_y \delta \underline{u}(t_i) \quad (3-32)$$

and the control law as

$$\begin{aligned} \delta \underline{u}^*(t_{i+1}) = & \delta \underline{u}^*(t_i) - K_x [\delta \underline{x}(t_{i+1}) - \delta \underline{x}(t_i)] \\ & - K_z [C \delta \underline{x}(t_i) + D_y \delta \underline{u}(t_i)] \end{aligned} \quad (3-33)$$

The above equation can be rewritten using Equation (3-9) as

$$\delta \underline{u}^*(t_{i+1}) = \delta \underline{u}^*(t_i) - [K_x \quad K_z] \begin{bmatrix} (\phi - I) & B_d \\ C & D_y \end{bmatrix} \begin{bmatrix} \delta \underline{x}(t_i) \\ \delta \underline{u}(t_i) \end{bmatrix} \quad (3-34)$$

Equating Equations (3-34) and (3-27), it is seen that

$$[K_x \quad K_z] = [G_{c1}^* \quad G_{c2}^*] \begin{bmatrix} (\phi - I) & B_d \\ C & D_y \end{bmatrix}^{-1} \quad (3-35)$$

$$\begin{bmatrix} K_x & K_z \end{bmatrix} = \begin{bmatrix} G_{c1}^* & G_{c2}^* \end{bmatrix} \begin{bmatrix} \pi_{11} & \pi_{12} \\ \pi_{21} & \pi_{22} \end{bmatrix} \quad (3-36)$$

The values for the feedback gains  $K_x$  and  $K_z$  are thus defined by

$$K_x = G_{c1}^* \pi_{11} + G_{c2}^* \pi_{21} \quad (3-37a)$$

$$K_z = G_{c1}^* \pi_{12} + G_{c2}^* \pi_{22} \quad (3-37b)$$

once  $G_{c1}^*$  and  $G_{c2}^*$  are available from the augmented state regulator solution and  $\pi_{11}$ ,  $\pi_{12}$ ,  $\pi_{21}$ , and  $\pi_{22}$  are obtained from Equation (3-35) and (3-36). The final form for the PI controller which achieves Type-1 control, implemented in incremental form is

$$\delta \underline{u}^*(t_{i+1}) = \delta \underline{u}^*(t_i) - K_x [\delta \underline{x}(t_{i+1}) - \delta \underline{x}(t_i)] \quad (3-38)$$

(3-38)

$$+ K_z [\underline{y}_d(t_{i+1}) - \underline{y}(t_i)]$$

Notice that the different time arguments on the  $\underline{y}$  terms are correct (Ref 24:147).

### 3.4 Doyle and Stein Technique for PI Controllers

The technique of injecting white input noise into a controlled system is accomplished for a sampled-data PI controller in a manner similar to that described in Section 2.7.2 for a sampled-data LQ regulator.

The equivalent discrete-time, stochastic difference equation to describe the system to be controlled is given by

$$\underline{x}(t_{i+1}) = \phi \underline{x}(t_i) + B_d \underline{u}(t_i) + G_d \underline{w}_d(t_i) \quad (3-39)$$

where the noise vector has now been included. The discrete-time input vector,  $B_d$ , and the noise covariance,  $Q_d$ , are formed by solving the following equations

$$B_d = \int_{t_i}^{t_{i+1}} \phi(t_{i+1} - \tau) B \, d\tau \quad (3-40)$$

$$Q_d = \int_{t_i}^{t_{i+1}} \phi(t_{i+1} - \tau) G G^T \phi^T(t_{i+1} - \tau) \, d\tau \quad (3-41)$$

where  $Q_d = E(\underline{w}_d(t_i) \underline{w}_d^T(t_i))$ .  $G_d$  is now defined to be the identity matrix,  $I$ .

Recall from Chapter II that input noise was added to the system by modifying the  $Q_d$  matrix via

$$Q_d(q) = Q_{do} + q^2 B V B^T \Delta t \quad (3-42)$$

where  $Q_{do}$  is now the original noise covariance given in Equation (3-40),  $B$  is the input vector from the continuous-time state differential equation,  $q^2$  is a scalar design parameter which adjusts the strength of the input noise, and  $\Delta t$  is the sample time of the discrete controller. This, in effect, makes a first-order approximation to the discrete-time input noise.

However, the discrete-time input vector is available from Equation (3-40) and is approximated to first order by  $B_d = B \Delta t$ , and so it will be

used to modify the  $Q_d$  matrix for PI controllers. The resulting modified noise covariance matrix is

$$Q_d(q) = Q_{do} + q^2 B_d V B_d^T / \Delta t \quad (3-43)$$

where  $V$  is a nonsingular matrix, chosen to be the identity matrix for this thesis. Notice that the Kalman filter design is the same regardless of the type of controller implemented so that robustification would be accomplished in the same manner as described in Chapter II. The structure of the sampled-data PI controller is shown in Figure(3-1). The Kalman filter of Figure (3-1) is the same as for the LQG regulator shown in Figure (2-4).

### 3.5 Performance Analysis for PI Controllers

The performance analysis for a PI controller is accomplished by following the same procedure given in Section 2.6 for LQ regulators. As given in Equation (3-38), the incremental form for the optimal, deterministic PI control law is given by

$$\begin{aligned} \delta \underline{u}^*(t_{i+1}) = & \delta \underline{u}^*(t_1) - K_x [\delta \underline{x}(t_{i+1}) - \delta \underline{x}(t_1)] \\ & + K_z [\underline{y}_d(t_{i+1}) - \underline{y}(t_1)] \end{aligned} \quad (3-44)$$

Again, the differing time arguments on the  $\underline{y}$  terms are correct.

When a Kalman filter is employed to estimate states for the controller, the conditional mean and covariance of the states are propagated between sample times by Equations (2-76a) and (2-76b). At the sample times, when measurements become available, the conditional mean and

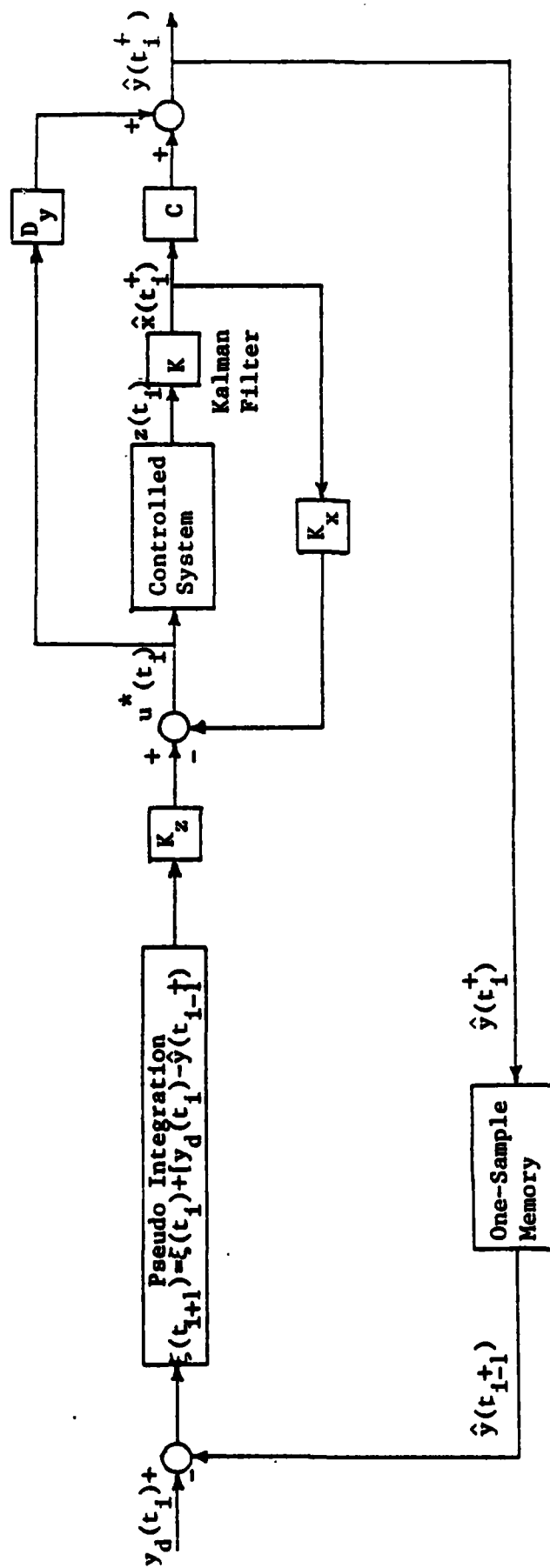


Figure 3-1: Sampled-Data PI Controller



covariance are updated by Equations (2-77) and (2-78), where the Kalman filter gain is given in Equation (2-79). Using the control law of Equation (3-44) and the Kalman filter equations, the PI controller can be put into the generic format introduced in Equation (2-82) and (2-83):

$$\underline{u}^*(t_{i+1}) + G_{cx} \underline{x}_c(t_i) + G_{cz} \underline{z}(t_i) + G_{cy} \underline{y}_d(t_i) \quad (3-45a)$$

$$\underline{x}_c(t_{i+1}) = \phi_c \underline{x}_c(t_i) + B_{cz} \underline{z}(t_i) + B_{cy} \underline{y}_d(t_i) \quad (3-45b)$$

where the subscript c refers to states of the controller. The gain matrices of Equations (3-45a) and (3-45b) are evaluated explicitly in Appendix A.

At this point, the performance of the PI controller may be evaluated using the method of Section 2.6, evaluating the statistical characteristics of an augmented vector

$$\underline{y}_a(t_i) = \begin{bmatrix} \underline{x}_t(t_i) \\ \underline{u}(t_i) \end{bmatrix} \quad (3-46)$$

which is composed of the states of the truth model and the controls generated by the PI controller.

### 3.6 Summary

The chapter has presented the equations for designing and evaluating PI controllers based on an LQG methodology. The advantage of a PI controller over the LQG regulator discussed in Chapter II is that it will track a desired output with zero steady-state error (in a deterministic

setting, or in zero steady-state mean error in a stochastic environment) even in the face of unmodeled constant disturbance. The type chosen for implementation was an incremental form of PI controller that is based on augmentation of system state equations with relations involving pseudointegration of the control rates.

The Doyle and Stein stability robustness enhancement technique of inputting white noise into the system model at the control entry points during filter tuning was applied to PI controllers. The following chapter will extend this method to allow time-correlated input noise for the case where robustification is desired over only a limited frequency range.

## Chapter IV. Time-Correlated Input Noise

### 4.1 Introduction

Section 2.4 introduced the idea of robustifying a Kalman filter by adding pseudonoise to a system at the point of entry of the control inputs,  $\underline{u}$ . By increasing the intensity of this pseudonoise, the stability robustness properties of a full-state feedback system could be asymptotically recovered by the observer-based system.

One disadvantage of this technique is that the noise addition degrades the performance of the system at the design conditions. The white noise adds uncertainty to the model with the same intensity throughout the frequency spectrum. It may be desired to robustify the Kalman filter only over a certain frequency range where confidence in the adequacy of the controller design model may be low. Thus, a natural extension to the Doyle and Stein technique (Ref 6) is to add time-correlated (colored) input noise where the strength of the noise is highest for the frequency range of interest. This chapter develops this idea and describes how to apply it to the types of controllers discussed in Chapter II and III (Ref 15;16;18).

In the first section, a stochastic model is developed for the fictitious colored noise process. Then, this is augmented with the model used to describe the system. Finally, the colored noise processes to be used in this thesis are described, including the shaping filters to generate them.

### 4.2 Stochastic Model

Stationary, time-correlated input noise is represented as the

steady-state output of a linear time-invariant system driven by white Gaussian noise as

$$\dot{\underline{x}}_u(t) = F_u \underline{x}_u(t) + G_u \underline{w}_u(t) \quad (4-1a)$$

$$\underline{n}_u(t) = H_u \underline{x}_u(t) + D_u \underline{w}_u(t) \quad (4-1b)$$

The subscript  $u$  refers to uncertainty parameters. The statistics of the stationary white noise,  $\underline{w}_u(t)$ , are given by

$$E\{\underline{w}_u(t)\} = \underline{0} \quad (4-2a)$$

$$E\{\underline{w}_u(t) \underline{w}_u^T(t+\tau)\} = Q_u \delta(\tau) \quad (4-2b)$$

The colored noise process,  $\underline{n}_u(t)$ , is added to the system model, given by

$$\dot{\underline{x}}(t) = F \underline{x}(t) + B \underline{u}(t) + G \underline{w}(t) \quad (4-3a)$$

$$\underline{z}(t) = H \underline{x}(t) + \underline{v}(t) \quad (4-3b)$$

at the point of entry of  $\underline{u}$ . This results in

$$\dot{\underline{x}}(t) = F \underline{x}(t) + B \left[ \underline{u}(t) + \underline{n}_u(t) \right] + G \underline{w}_u(t) \quad (4-4)$$

Define an augmented vector to be the system states and the uncertainty model states

$$\underline{x}_g(t) = \begin{bmatrix} \underline{x}(t) \\ \underline{x}_u(t) \end{bmatrix} \quad (4-5)$$

Next, augment the uncertainty state differential equations with those of the original system. This yields an augmented system equation, given by

$$\dot{\underline{x}}_s(t) = F_s \underline{x}_s(t) + B_s \underline{u}(t) + G_s \underline{w}_s(t) \quad (4-6a)$$

$$\underline{z}_s(t) = H_s \underline{x}_s(t) + \underline{v}(t) \quad (4-6b)$$

The above matrices are described by

$$F_s = \begin{bmatrix} F & BH_u \\ 0 & F_u \end{bmatrix} \quad (4-7)$$

$$B_s = \begin{bmatrix} B \\ 0 \end{bmatrix} \quad (4-8)$$

$$G_s = \begin{bmatrix} G & BD_u \\ 0 & G_u \end{bmatrix} \quad (4-9)$$

$$H_s = \begin{bmatrix} H & 0 \end{bmatrix} \quad (4-10)$$

The noise vector,  $\underline{w}_s(t)$ , is given by

$$\underline{w}_s(t) = \begin{bmatrix} \underline{w}(t) \\ \underline{w}_u(t) \end{bmatrix} \quad (4-11)$$

with covariance kernel  $E\{\underline{w}_s(t)\underline{w}_s^T(t+\tau)\} = Q_s \delta(\tau)$ , where

$$Q_s = \begin{bmatrix} Q & 0 \\ 0 & Q_u \end{bmatrix} \quad (4-12)$$

The augmented system model described in Equations (4-5) through (4-12) is used to design either the continuous-time Kalman filter in Equations (2-12a) and (2-12b) or the sampled-data Kalman filter in Equations (2-76) through (2-79). However, it is not necessary to feed back the uncertainty model states in the controller algorithm; they are used only to modify the filter within the overall controller. Therefore, the original system model is used to design the LQ regulator of Chapter II or the PI controller of Chapter III. However, the dimensions of the Kalman filter and the controller are now incompatible. This is solved by augmenting the controller gain matrix with zeros

$$G_{c_s}^* = \left[ G_c^* \mid 0 \right] \quad (4-13)$$

Thus, the order of the original system has been increased to reflect the adding of uncertainty model states, but these are fed back through the controller with zero gains.

At this point, the control law is put into the continuous-time generic form of Equations (2-25) and (2-26) or the sampled-data form of Equations (2-82) and (2-83). then, the performance of the robustified system is evaluated using the performance analysis described in Chapter II.

### 4.3 Shaping Filter Design

The time-correlated noise processes to be examined in this thesis are generated by inputting white noise to first- or second-order shaping filters as described in Reference 23:180-186. The following sections describe the form of the shaping filters.

#### 4.3.1 First-Order Colored Noise Processes

A noise process which has low intensity at low frequencies and high intensity at high frequencies (or high intensity at low frequencies and low intensity at high frequencies) can be generated as the output of a system with a transfer function of the form

$$\frac{x_u(s)}{w_u(s)} = \frac{s + a}{s + b} \quad (4-14)$$

In Chapter V, it is shown that the model used for this thesis is adequate for low frequencies but may not be for higher frequencies. Thus, for this case, the strength of the noise should be kept low at low frequencies and then increased where the adequacy of the model is in doubt. Figure (4-1) shows the power spectral density function for this process.

Using the standard controllable form given in Reference 23:Ch 2, the state-space representation for Equation (4-14) is

$$\dot{x}_u(t) = -b x_u(t) + w_u(t) \quad (4-15a)$$

$$n_u(t) = (a - b) x_u(t) + w_u(t) \quad (4-15b)$$

For this first-order shaping filter, the uncertainty model matrices are thus scalar and given by

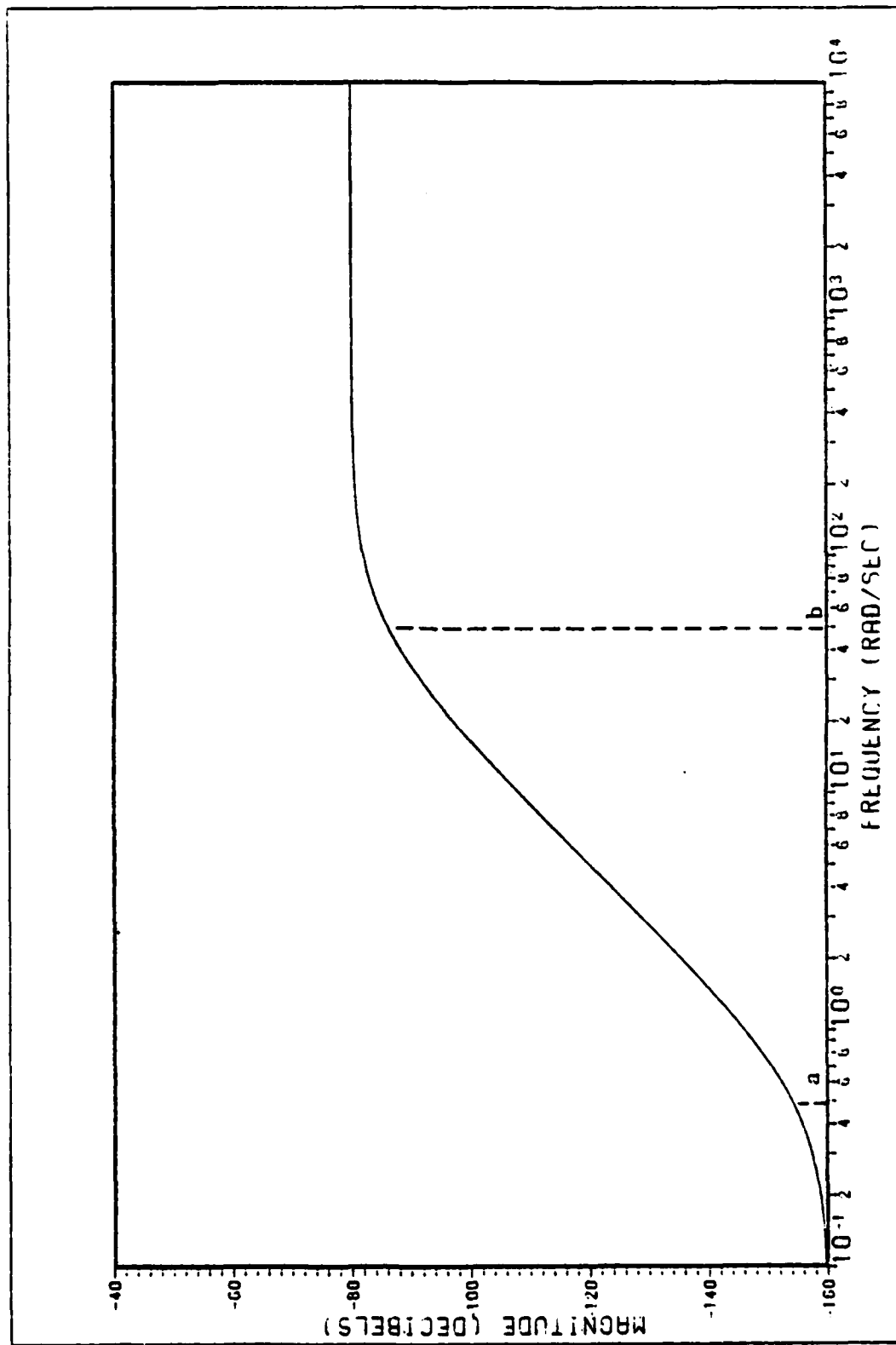


Figure 4-1: Power Spectral Density Function for a First-Order Shaping Filter



$$F_u = -b \quad (4-16a)$$

$$B_u = 1 \quad (4-16b)$$

$$H_u = (a - b) \quad (4-16c)$$

$$D_u = 1 \quad (4-16d)$$

Note that a noise process is generated for each control applied to the system. Thus, the dimension of the augmented system is that of the basic system plus the number of controls.

#### 4.3.2 Second-Order Colored Noise Processes

A noise process which has high intensity over only a limited frequency range can be generated by passing white noise through a linear system model that has a transfer function of the form

$$\frac{x_u(s)}{w_u(s)} = \frac{s + c}{(s + d)(s + e)} \quad (4-17)$$

If it is the case where the adequacy of the design model is in doubt over a limited frequency range, then a second-order shaping filter as in Equation (4-17) would be appropriate. The power spectral density function for this process is shown in Figure (4-2).

A state-space representation for this process, using standard controllable form is given by

$$\begin{bmatrix} \dot{x}_{u_1}(t) \\ \dot{x}_{u_2}(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -de & -(d + e) \end{bmatrix} \begin{bmatrix} x_{u_1}(t) \\ x_{u_2}(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} w_u(t) \quad (4-18a)$$

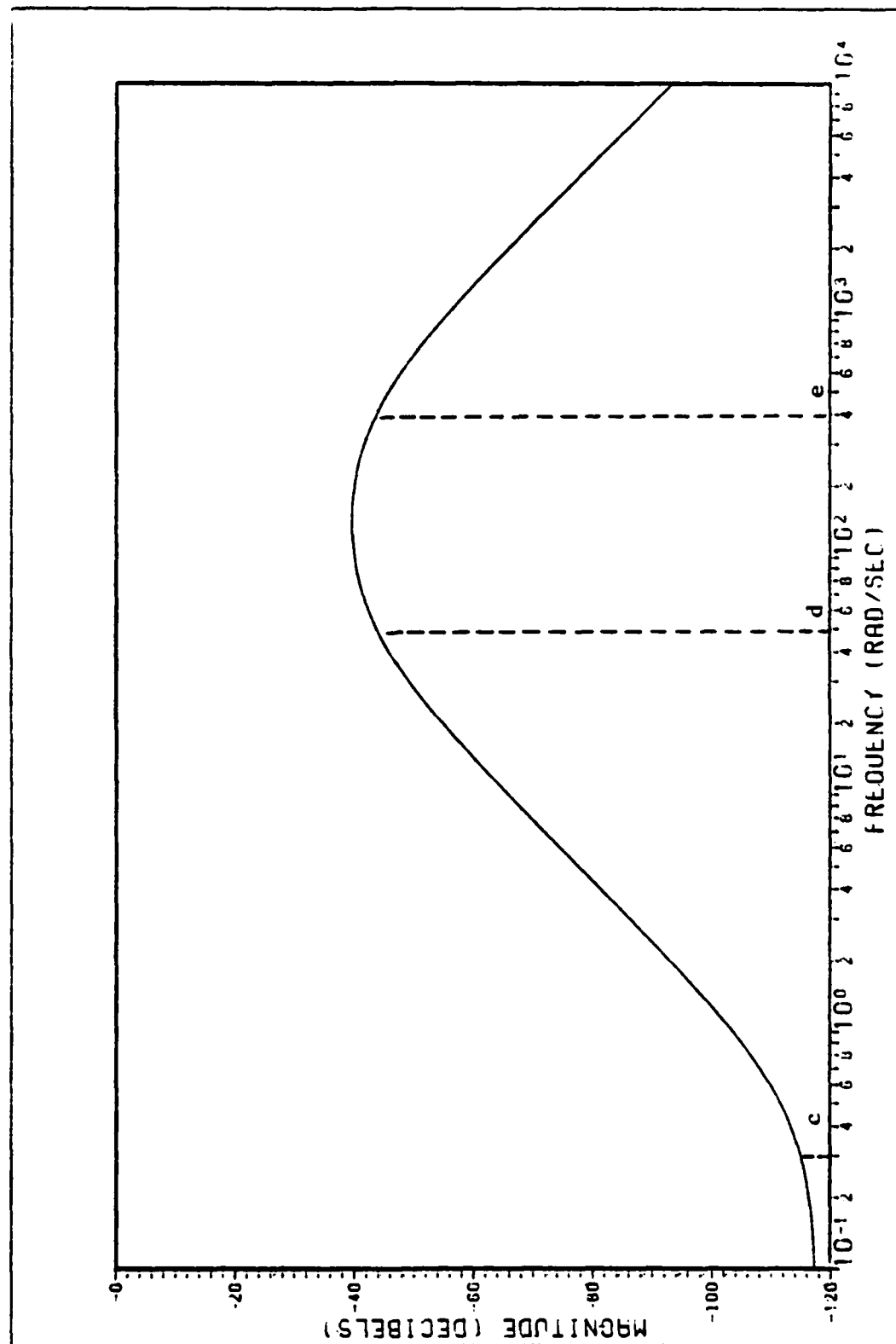


Figure 4-2: Power Spectral Density Function for a Second-Order Colored Noise Process

$$n_u(t) = \begin{bmatrix} c & 1 \end{bmatrix} \begin{bmatrix} x_{u1}(t) \\ x_{u2}(t) \end{bmatrix} \quad (4-18b)$$

For this second-order shaping filter, the uncertainty model matrices are thus given by

$$F_u = \begin{bmatrix} 0 & 1 \\ -de & -(d + e) \end{bmatrix} \quad (4-19a)$$

$$B_u = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad (4-19b)$$

$$H_u = \begin{bmatrix} c & 1 \end{bmatrix} \quad (4-19c)$$

$$D_u = \begin{bmatrix} 0 \end{bmatrix} \quad (4-19d)$$

For this case, the dimensionality of the original system is increased by two times the number of controls when these shaping filter states are augmented to the system states.

The above technique for robustifying the Kalman filter in a control system is incorporated into two subroutines added to the LQG regulator design program described in Reference 21. Source code and instructions for running the program with the additional routines are listed in Appendix B.

Similarly, for PI controllers, the user must input the original design model, before augmenting the shaping filter states, to the program described in Reference 13. The original model is used to design the PI controller. Then, the shaping filter states are augmented with the model for the Kalman filter design. Finally, the PI controller feedback gain matrix is augmented with zeros to make the dimensions of the two solutions compatible. Modifications to the software in Reference 13 are listed in Appendix C.

#### 4.4 Summary

This chapter has presented an extension to the Doyle and Stein robustness method of injecting white input noise into a controlled system during the process of tuning the Kalman filter. The extension allows time-correlated noise to be injected into the system model, which is accomplished by augmenting shaping filter states to the state differential equations of the system to be controlled. Thus, the strength of the colored noise can be concentrated at frequencies where robustification is desired and attenuated elsewhere.

Two types of shaping filters to generate the colored noise were examined. The first concentrated the strength of the noise at higher frequencies. The second concentrated the strength of the noise over a limited frequency range. Then, the method was applied to the types of controllers discussed in this thesis.

At this point, the types of controllers which are designed for this thesis have been discussed. In addition, techniques to enhance the robustness characteristics of the controllers have been presented. The following chapter presents the linear, time-invariant model to which the above mentioned methods are applied in this thesis.

## Chapter V. AFTI/F-16 Flight Control Design

### 5.1 Introduction

This chapter presents the longitudinal dynamic equations for the Advanced Fighter Technology Integration aircraft, an F-16 aircraft (AFTI/F-16) modified for advanced controls research. For longitudinal control, the aircraft is equipped with a decoupled horizontal tail and trailing edge flap which make it possible for the AFTI/F-16 to perform such maneuvers as pitch pointing. A pitch-pointing maneuver allows the aircraft's attitude angle to be changed while holding the flight path angle constant. The controllers to be designed lend themselves readily to multiple-input multiple-output (MIMO) system methods such as those described in this thesis. The aircraft equations are adequately portrayed as linear and time invariant by linearization about specified trim conditions in the aircraft's flight envelope. The controllers described in earlier chapters will be designed for the AFTI/F-16, applying the techniques of tuning the Kalman filter for robustness purposes by injecting fictitious white or time-correlated noise into the system model at the point of entry of the control inputs.

Data used to form the AFTI/F-16 longitudinal equations was taken from References 11 and 12. Reference 11 lists dimensional stability derivatives in a body-axis coordinate system for the aircraft, but does not include data for the trailing-edge flap. This information is given in Reference 12 in a stability-axis coordinate system. The necessary transformations were performed to convert all values to the body-axis frame.

The development of a linear perturbation model for an aircraft's longitudinal dynamics is found in References 7;8;9;10;25. The model for clear air turbulence is developed in Reference 17 and used in Reference 12. Both these models are incorporated to form the AFTI/F-16 equations described later in this chapter.

The first section of this chapter describes a thirteen-state linear perturbation truth model for the AFTI/F-16. This thirteen-state model is unobservable because a measurement is not available for all four aircraft states. Therefore, the unobservable state is deleted from the full model to leave a twelve-state truth model. This model, described in a second section, will be used to evaluate the performance of lower order controller models. The lower order (eight-state) controller model is described in a third section. The final section gives the data and longitudinal equations for one twelve-state truth model at an off-design flight condition. This will be used to evaluate the robustness of the controllers designed.

## 5.2 AFTI/F-16 Truth Model

The longitudinal aircraft truth model consists of four aircraft states, three turbulence (gust) states, and six additional states to describe the third order dynamics for each of the horizontal tail and trailing-edge-flap actuators. The aircraft states are:  $\theta$ , attitude angle;  $\alpha$ , angle of attack,  $q$ , pitch rate, and;  $u$ , forward perturbation velocity. There are two angle of attack gust states and one pitch-rate gust state. Noise-corrupted measurements are available for  $\theta$ ,  $\alpha$ , and  $q$ , and the measurement for  $\alpha$  includes an angle-of-attack gust state. The actuator and gust models will be described first.

### 5.2.1 Actuator Dynamics

The actuators for the horizontal tail and trailing-edge-flap are modeled as third-order systems, yielding delivered angle  $\delta$  in response to commanded angle  $\delta_{com}$ , of the form

$$\frac{\delta(s)}{\delta_{com}(s)} = \frac{(1/T_s) (\omega_n^2)}{(s + 1/T_s) (s^2 + 2\xi \omega_n s + \omega_n^2)} \quad (5-1)$$

Multiplying the numerator and denominator terms yields

$$\frac{\delta(s)}{\delta_{com}(s)} = \frac{a_0}{s^3 + a_2 s^2 + a_1 s + a_0} \quad (5-2)$$

Using the standard observable state space representation given in Reference 23:Ch 2, the corresponding model is

$$\begin{bmatrix} \dot{\delta}(t) \\ \dot{\delta}_1(t) \\ \dot{\delta}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -a_0 & -a_1 & -a_2 \end{bmatrix} \begin{bmatrix} \delta(t) \\ \delta_1(t) \\ \delta_2(t) \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \delta_{com}(t) \quad (5-3a)$$

$$\delta(t) = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} \delta(t) \\ \delta_1(t) \\ \delta_2(t) \end{bmatrix} \quad (5-3b)$$

where the  $b_i$ 's arise when taking the Laurent series for  $G(s)$ . Notice that  $\delta_{com}$  is the command surface deflection and  $\delta$  is the surface deflection output from the actuators. Two intermediate states,  $\delta_1$  and  $\delta_2$ , are needed to describe fully the actuator dynamics.

The values chosen for the parameters in Equation (5-1) are specified in Reference 27 as

$$\frac{\delta(s)}{\delta_{com}(s)} = \frac{(20.2)(71.4)^2}{(s + 20.2)(s^2 + 2(.736)(71.4)s + 71.4^2)} \quad (5-4)$$

The same actuator equations are used for both the horizontal tail and trailing-edge-flap. Expanding Equation (5-4) yields

$$\frac{\delta(s)}{\delta_{com}(s)} = \frac{102980}{(s^3 + 125.3s^2 + 7221s + 102980)} \quad (5-5)$$

Thus the resulting state equation is

$$\begin{bmatrix} \dot{\delta}(t) \\ \dot{\delta}_1(t) \\ \dot{\delta}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -102980 & -7221 & -125.3 \end{bmatrix} \begin{bmatrix} \delta(t) \\ \delta_1(t) \\ \delta_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 102980 \end{bmatrix} \delta_{com}(t) \quad (5-6)$$

### 5.2.2 Turbulence State Equations

The model used for clear air turbulence is found in References 12 and 17. The state equations for the three gust states are given by

$$\dot{\alpha}'_g = a'_{g1} \alpha'_g + n'_\alpha \quad (5-7a)$$

$$\dot{\alpha}'_g = a'_{g2} \alpha'_g + a'_{g3} \alpha'_g = a'_n n'_\alpha \quad (5-7b)$$

$$\dot{q}'_g = a'_{gd} \dot{\alpha}'_g + q'_{g1} q'_g \quad (5-7c)$$



The noise source,  $n_\alpha$ , is modeled as a white Gaussian noise with statistics

$$E\{n_\alpha(t)\} = 0 \quad (5-8a)$$

$$E\{n_\alpha(t) n_\alpha^T(t + \tau)\} = \delta(\tau) \quad (5-8b)$$

The coefficients in Equations (5-7) are computed using the following equations

$$a_{g1}' = \frac{-V_t}{L_w} \quad (5-9a)$$

$$a_{g2} = \frac{\sigma_w}{L_w} (1 - \sqrt{3}) \sqrt{\frac{V_t}{L_w}} \quad (5-9b)$$

$$a_g = \frac{-V_t}{L_w} \quad (5-9c)$$

$$a_n = \frac{\sigma_w}{V_t} \sqrt{\frac{3V_t}{L_w}} \quad (5-9d)$$

$$a_{gd} = \frac{\pi V_t}{4b} \quad (5-9e)$$

$$q_{g1} = \frac{-\pi V_t}{4b} \quad (5-9f)$$

$V_t$  is aircraft velocity and  $b$  is the wing span. The quantities  $\sigma_w$  and  $L_w$  are given in Reference 3 as functions of altitude, where  $\sigma_w$  is the root mean square intensity of the clear air turbulence in feet per

second and  $L_w$  is a reference scale length in feet.

### 5.2.3 System Equations

The thirteen states for the unobservable truth model are listed in Equation (5-10) on the following page.

As demonstrated in Reference 12, it is convenient to define the truth model equations in two stages. First, define the system matrix,  $F'_t$  of Equation (2-25), neglecting all  $\dot{\alpha}$  stability derivative terms and also the  $\dot{\alpha}_g$  term in Equation (5-7c). Then modify the matrix to incorporate the neglected terms. This is done for mathematical convenience as the  $\dot{\alpha}$  terms are neglected in a later truth model. Defining the system matrix in two stages does not change the definition of the states; the second stage is merely a more adequate model. The initial matrix is given in Equation (5-11) on the following page.

To incorporate the neglected terms, rows 2,3,4 and 13 are modified with the following equations (using standard double-index array notation  $F_t(I,J)$  for the I-th row and J-th column of  $F_t$ ).

$$F_t(2,J) = F'_t(2,J) / (1 - Z_{\dot{\alpha}}) \quad (5-12a)$$

$$F_t(3,J) = F'_t(3,J) + M_{\dot{\alpha}} \times F_t(2,J) \quad (5-12b)$$

$$F_t(4,J) = F'_t(4,J) + X_{\dot{\alpha}} \times F_t(2,J) \quad (5-12c)$$

$$F_t(13,J) = F'_t(13,J) + a_{g_d} \times F'_t(12,J) \quad (5-12d)$$

The column index J ranges from 1 to 13.

$\underline{x}_t =$ 

$$\begin{bmatrix} \theta(t) \\ \alpha(t) \\ q(t) \\ u(t) \\ \delta_{HT}(t) \\ \delta_{HT_1}(t) \\ \delta_{HT_2}(t) \\ \delta_{TEF}(t) \\ \delta_{TEF_1}(t) \\ \delta_{TEF_2}(t) \\ \alpha'_g(t) \\ \alpha_g(t) \\ q_g(t) \end{bmatrix}$$

(5-10)

[illegible]

1

In Equations (5-11) and (5-12), the dimensional stability derivatives "X", "Z", and "M" refer to body forces acting along the X- and Z-axes (forward and down axes) and the pitching moment about the Y-axis, (axis emanating from aircraft center of mass toward the right wing) respectively. The subscript identifies the state with respect to which the derivative is taken. The term "g" is the acceleration due to gravity. The subscripts HT and TEF refer to deflections of the horizontal tail and trailing-edge-flap, respectively. The terms  $\alpha_0$ ,  $u_0$  and  $w_0$  refer to trim values of angle of attack, and velocity components along the X- and Z-axes, respectively. A trim condition is a steady-state level flight condition about which the longitudinal dynamic equations are linearized. For this case, the trim condition is chosen to be steady level flight at a Mach number of 0.6 and an altitude of 10,000 feet.

Table (5-1) gives the values needed to define the thirteen-state truth model. Note that the value of  $\alpha_w$  given in the gust model coefficients is for "Level 1" or light to moderate air turbulence. This level of air turbulence has a high probability of being encountered in flight.

From Chapter II, the truth model is of the form

$$\dot{\underline{x}}_t(t) = F_t \underline{x}_t(t) + B_t \underline{u}(t) + G_t \underline{w}_t(t) \quad (5-13)$$

The control vector,  $\underline{u}$ , is given by

$$\underline{u} = \begin{bmatrix} \delta_{HT_{com}}(t) \\ \delta_{TEF_{com}}(t) \end{bmatrix} \quad (5-14)$$

Table 5-1

AFTI/F-16 Data for  $M = 0.6$ ,  $H = 10,000$  Feet

TRIM CONDITIONS	ACTUATOR PARAMETERS	ADDITIONAL PARAMETERS
$V_T = 646.4$ Ft/Sec	$a_o = 102980.0$	$g = 32.174$ Ft/Sec <sup>2</sup>
$\alpha_o = 2.174$ Deg	$T_S = 20.2$ /Sec	$b = 30$ Ft
$u_o = 645.9$ Ft/Sec	$\xi = 0.736$	$\sigma_w = 5.0$ Ft/Sec
$w_o = 24.52$ Ft/Sec	$\omega_n = 71.4$ /Sec	$L_w = 1750.0$ Ft

## LONGITUDINAL STABILITY DERIVATIVES

Z - BODY FORCES	X - BODY FORCES	PITCHING MOMENTS
$Z_{\alpha} = -1.355$ /Sec	$X_{\alpha} = 7.56$ Ft/Sec <sup>2</sup>	$M_{\alpha} = 4.95 \times 10^{-2}$ /Sec <sup>2</sup>
$Z_{\dot{\alpha}} = 2.241 \times 10^{-3}$	$X_{\dot{\alpha}} = -5.5 \times 10^{-2}$ Ft/Sec	$M_{\dot{\alpha}} = -0.1556$ /Sec
$Z_q = -5.435 \times 10^{-3}$	$X_q = 0.1334$ Ft/Sec	$M_q = -0.5376$ /Sec
$Z_u = -1.215 \times 10^{-4}$ /Ft	$X_u = -6.47 \times 10^{-3}$ /Sec	$M_u = -6.55 \times 10^{-4}$ /Sec/Ft
$Z\delta_{HT} = -0.1335$ /Sec/Rad	$X\delta_{HT} = 0.8404$ Ft/Sec <sup>2</sup> /Rad	$M\delta_{HT} = -0.1332$ /Sec <sup>2</sup> /Rad
$Z\delta_{TEF} = -0.2365$ /Sec/Rad	$X\delta_{TEF} = -2.38$ Ft/Sec <sup>2</sup> /Rad	$M\delta_{TEF} = -1.32$ /Sec <sup>2</sup> /Rad

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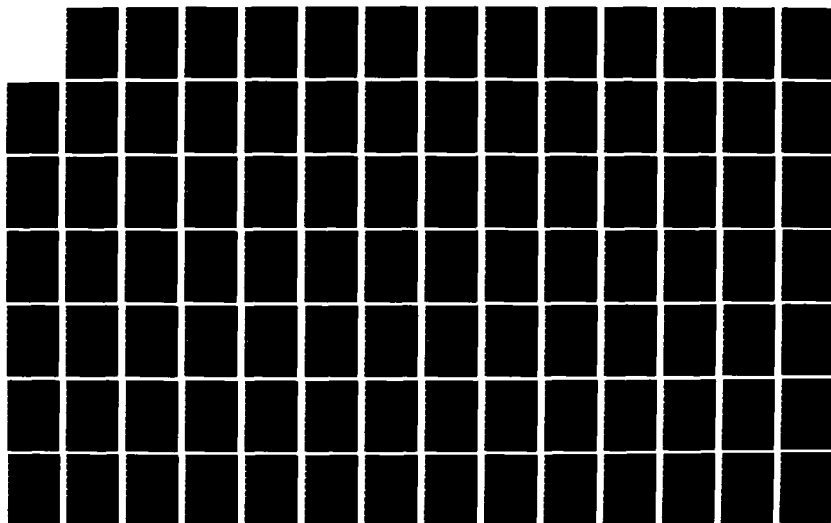
ROBUST FLIGHT CONTROLLERS(U) AIR FORCE INST OF TECH  
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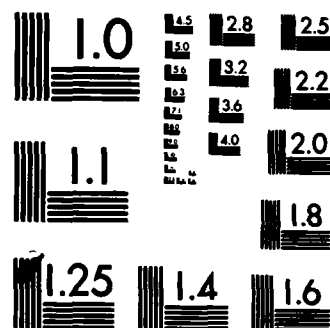
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The  $B_t$  and  $G_t$  matrices are given by Equations (5-15) and (5-16) on the following page. The last element of the  $G_t$  matrix is defined to be

$$q_n = a_{g_d} \times a_n \quad (5-17)$$

where  $a_n$  and  $a_{g_d}$  were defined previously in Equations (5-9d) and (5-9e).

Employing the data of Table (5-1) and Equations (5-11) and (5-12), the  $F_t$  matrix is given by Equation (5-18). Equations (5-19) through (5-22) give the other system matrices with their numerical values inserted.

Recall the continuous-time measurement equation is of the form

$$\underline{z}_t(t) = H_t \underline{x}_t + \underline{v}_t(t) \quad (5-23)$$

The covariance kernel description,  $R_t$ , for the measurement noise,  $\underline{v}_t(t)$ , (Equation (2-28)) was established using the values given in Reference 12. The results are listed in Equation (5-24).

$$R_t = \begin{bmatrix} 9.52 \times 10^{-8} & 0 & 0 \\ 0 & 2.44 \times 10^{-7} & 0 \\ 0 & 0 & 6.44 \times 10^{-7} \end{bmatrix} \quad (5-24)$$

Recall from Equation (2-120) that a first order approximation for a discrete-time  $R_{dt}$ , to be used to design a sampled-data controller is given by

$$R_{dt} = R_t / \Delta t \quad (5-25)$$

The truth model described above will hereafter be referred to as (T13,10,0.6). "T" indicates truth model, 13 refers to the number of

$$B_t = \begin{bmatrix} 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ a_o & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & a_o \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \end{bmatrix} \quad (5-15)$$

$$G_t = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \\ a_n \\ q_n \end{bmatrix} \quad (5-16)$$

[illegible]

$$B_t = \begin{bmatrix} 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 102980. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 102980 \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \end{bmatrix} \quad (5-19)$$

$$G_t = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \\ 8.142 \times 10^{-3} \\ 0.1378 \end{bmatrix} \quad (5-20)$$

$$Q_t = [1] \quad (5-21)$$

$$H_t = \begin{bmatrix} 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. \\ 0. & 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 1. & 0. \\ 0. & 0. & 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. \end{bmatrix} \quad (5-22)$$

states, 10 refers to the altitude in thousands of feet, and 0.6 to the Mach number for which the equations were computed.

### 5.3 Reduced Order Truth Model

The truth model of the previous section represents an unobservable system because a measurement of the fourth state,  $u$ , is unavailable. In practice, it is rarely desirable to feed back this state through a controller. Therefore, the fourth state is deleted from the truth model before any further simplification to yield models upon which to base a controller/estimator design. In addition, all terms involving  $\dot{\alpha}$  stability derivations are ignored because their effects on the terms of Equation (5-19) are negligible. Incorporating these derivatives causes the values for the  $F_t$  matrix to change only slightly, except for the terms involving,  $u$ , which were deleted. Thus, the modifications listed in Equations (5-12a), (5-12b) and (5-12c) are not necessary. Equation (5-12d) now becomes

$$F_t(12,J) = F'_t(12,J) + a_{g_d} \times F'_t(11,J) \quad (5-26)$$

The matrices needed to define the twelve-state truth model are given in Equations (5-27) through (5-33). Note that  $R_t$  was defined previously in Equation (5-24).

The twelve-state truth model based on linearization about a trim condition at 10000 feet and Mach 0.6 will be referred to as (T12,10,0.6).

### 5.4 Controller Design Model

The reduced order controller is formed by modelling the AFTI/F-16 actuator dynamics as first-order lags instead of third-order systems as

$$\underline{x}_t = \begin{bmatrix} \theta(t) \\ \alpha(t) \\ q(t) \\ \delta_{HT}(t) \\ \delta_{HT_1}(t) \\ \delta_{HT_2}(t) \\ \delta_{TEF}(t) \\ \delta_{TEF_1}(t) \\ \delta_{TEF_2}(t) \\ \alpha'_g(t) \\ \alpha_g(t) \\ q_g(t) \end{bmatrix} \quad (5-27)$$

$$\underline{u}(t) = \begin{bmatrix} \delta_{HT_{com}}(t) \\ \delta_{TEF_{com}}(t) \end{bmatrix} \quad (5-28)$$

$P_t =$ 

(5-29)

0.	0.	1.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$-1.089 \times 10^{-3}$	-1.355	0.9946	-0.1335	0.	0.	-0.2365	0.	0.	0.	-1.355	$-5.435 \times 10^{-3}$
0.	$4.95 \times 10^{-2}$	-0.5376	-13.32	0.	0.	-1.32	0.	0.	0.	$4.95 \times 10^{-2}$	-0.5376
0.	0.	0.	0.	1.	0.	0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	1.	0.	0.	0.	0.	0.	0.
0.	0.	0.	-102980	-7221.	-125.3	0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	1.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	1.	0.	0.	0.
0.	0.	0.	0.	0.	0.	-102980.	-7221.	-125.3	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	-0.3694	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	$-1.271 \times 10^{-3}$	-0.3694	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	$-2.151 \times 10^{-2}$	-6.25	-16.92

$$B_t = \begin{bmatrix} 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 102980. & 0. \\ 0. & 0. \\ 0. & 0. \\ 0. & 102980. \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \end{bmatrix} \quad (5-30)$$

$$G_t = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \\ 8.142 \times 10^{-3} \\ 0.1378 \end{bmatrix} \quad (5-31)$$

$$Q_t = [1] \quad (5-32)$$

$$H_t = \begin{bmatrix} 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. \\ 0. & 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 1. & 0. \\ 0. & 0. & 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. \end{bmatrix} \quad (5-33)$$



given in Equation (5-1). The transfer function for a first-order lag with a coefficient of  $1/T_s$ , i.e., a lag time of  $T_s$ , is given by

$$\frac{\delta(s)}{\delta_{com}(s)} = \frac{1/T_s}{s + 1/T_s} \quad (5-34)$$

In state space representation, the above equation becomes

$$\dot{\delta}(t) = (-1/T_s) \delta(t) + (1/T_s) \delta_{com}(t) \quad (5-35)$$

where  $T_s$  is the time constant for the actuator. Again,  $\delta_{com}$  is the command surface deflection and  $\delta$  is the surface deflection output from the actuator. A good fit to the frequency response of the actuators yields  $1/T_s = 20.2$  (Ref 27). Equation (5-35) then becomes

$$\dot{\delta}(t) = -20.2 \delta(t) + 20.2 \delta_{com}(t) \quad (5-36)$$

Making this modification to the twelve-state truth model of Section 5.3 yields an eight-state controller design model. The matrices needed to define the controller model are given in Equations (5-37) through (5-43).

This eight-state controller model will be referred to as (C8,10,0.6).

### 5.5 Truth Models at Off-Design Conditions

To evaluate the effectiveness of the techniques previous described for robustifying the Kalman filter, the performance of the controllers are to be evaluated at flight conditions other than that for which the controllers were designed, as well as by using truth models of higher order than the design model. This is accomplished by using a truth model defined at an off-design condition in the performance analysis described in Chapter III. The following sections contain one such truth model.

$$\underline{x}(t) = \begin{bmatrix} \theta(t) \\ \alpha(t) \\ q(t) \\ \delta_{HT}(t) \\ \delta_{TEF}(t) \\ \alpha'_g(t) \\ \alpha_g(t) \\ q_g(t) \end{bmatrix} \quad (5-37)$$

$$\underline{u}(t) = \begin{bmatrix} \delta_{HT_{com}}(t) \\ \delta_{TEF_{com}}(t) \end{bmatrix} \quad (5-38)$$

F =

(5-39)

0.	0.	1.	0.	0.	0.	0.
$-1.889 \times 10^{-3}$	-1.355	0.9946	-0.1335	-0.2365	0.	$-5.435 \times 10^{-3}$
0.	$4.95 \times 10^{-2}$	-0.5376	-13.32	-1.32	0.	$4.95 \times 10^{-2}$
0.	0.	0.	-20.2	0.	0.	0.
0.	0.	0.	0.	-20.2	0.	0.
0.	0.	0.	0.	0.	-0.3694	0.
0.	0.	0.	0.	0.	$-1.271 \times 10^{-3}$	0.
0.	0.	0.	0.	0.	$-2.151 \times 10^{-2}$	-16.92

$$B = \begin{bmatrix} 0. & 0. \\ 0. & 0. \\ 0. & 0. \\ 20.2 & 0. \\ 0. & 20.2 \\ 0. & 0. \\ 0. & 0. \\ 0. & 0. \end{bmatrix} \quad (5-40)$$

$$G = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \\ 8.142 \times 10^{-3} \\ 0.1378 \end{bmatrix} \quad (5-41)$$

$$Q = [1] \quad (5-42)$$

$$H = \begin{bmatrix} 1. & 0. & 0. & 0. & 0. & 0. & 0. & 0. \\ 0. & 1. & 0. & 0. & 0. & 0. & 1. & 0. \\ 0. & 0. & 1. & 0. & 0. & 0. & 0. & 0. \end{bmatrix} \quad (5-43)$$

### 5.5.1 Truth Model (T12,20,0.6)

This truth model is defined for an altitude of 20,000 feet and a Mach number of 0.6. The values of parameters and stability derivatives needed to form the system matrices are given in Table (5-2). The truth model  $G_t$  and  $F_t$  matrices are given in Equation (5-44) and (5-45).

$$G_t = \begin{bmatrix} 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 0. \\ 1. \\ 7.052 \times 10^{-3} \\ .1591 \end{bmatrix} \quad (5-44)$$

### 5.6 Summary

This chapter has presented the model which will be used to design LQG regulators as in Chapter II and PI controllers as in Chapter III for the AFTI/F-16. The robustification methods detailed in Chapters II and IV will

be applied to the model, and the effect of this will be examined by performing covariance analysis as described in Chapter II. The results of this study are presented in the following chapter.

Table 5-2

AFTI/F-16 Data for  $M = 0.6$ ,  $H = 20,000$  Feet

TRIM CONDITIONS	ACTUATOR PARAMETERS	ADDITIONAL PARAMETERS
$V_T = 622.15$ Ft/Sec	$a_o = 102980.$	$g = 32.174$ Ft/Sec <sup>2</sup>
$\alpha_o = 3.152$ Deg	$T_S = 20.2$ /Sec	$b = 30$ Ft
$u_o = 621.2$ Ft/Sec	$\xi = 0.736$	$\sigma_w = 2.5$ Ft/Sec
$w_o = 34.2$ Ft/Sec	$\omega_n = 71.4$ /Sec	$L_w = 1750.$ Ft

## LONGITUDINAL STABILITY DERIVATIVES

Z - BODY FORCES	X - BODY FORCES	PITCHING MOMENTS
$Z_{\alpha} = -1.006$ /Sec	$X_{\alpha} = 23.79$ Ft/Sec <sup>2</sup>	$M_{\alpha} = -1.218$ /Sec <sup>2</sup>
$Z_{\alpha} = 1.66 \times 10^{-3}$	$X_{\alpha} = -5.69 \times 10^{-2}$ Ft/Sec	$M_{\alpha} = -0.1068$ /Sec
$Z_q = -3.97 \times 10^{-3}$	$X_q = 0.136$ Ft/Sec	$M_q = -0.384$ /Sec
$Z_u = -1.48 \times 10^{-4}$ /Ft	$X_u = -4.827 \times 10^{-3}$ /Sec	$M_u = -7.09 \times 10^{-4}$ /Sec/Ft
$Z\delta_{HT} = -9.18 \times 10^{-2}$ /Sec/Rad	$X\delta_{HT} = 0.711$ Ft/Sec <sup>3</sup> /Rad	$M\delta_{HT} = -8.95$ /Sec <sup>2</sup> /RAD
$Z\delta_{TEF} = -0.1689$ /Sec/Rad	$X\delta_{TEF} = -2.165$ Ft/Sec <sup>2</sup> /RAD	$M\delta_{TEF} = -0.7736$ /Sec <sup>2</sup> /Rad

$F_t =$

0.	0.	1.	0.	0.	0.	0.	0.	0.	0.	0.	0.	0.
$-1.510 \times 10^{-4}$	-1.377	0.9952	-0.121	0.	0.	-0.220	0.	0.	0.	0.	-1.397	$-4.83 \times 10^{-3}$
0.	-0.8540	-0.470	-16.10	0.	0.	-4.18	0.	0.	0.	0.	-0.8540	-0.470
0.	0.	0.	0.	1.	0.	0.	0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	1.	0.	0.	0.	0.	0.	0.	0.
0.	0.	0.	-102980.	-7221.	-125.3	0.	0.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	1.	0.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	1.	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	-102980.	-7221.	-125.3	0.	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	-0.4740	0.	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	$-7.200 \times 10^{-4}$	-0.4740	0.	0.
0.	0.	0.	0.	0.	0.	0.	0.	0.	$-1.564 \times 10^{-2}$	-10.29	-21.72	0.

(5-45)



## VI. Results

### 6.1 Introduction

The results of the study of methods to improve the robustness properties of controlled systems are presented in this chapter. The two methods examined are the techniques of tuning the Kalman filter by inputting both stationary white and time-correlated Gaussian noise into the system model at the control entry points.

The results of applying these techniques are presented first for LQG regulators. In the first sections, results are given for a continuous-time system. Then, two approaches for a discrete-time system are examined: discretizing the continuous-time controller using first-order approximations, and designing a sampled-data controller.

Finally, in the last sections, the robustification techniques are applied to a sampled-data PI controller.

### 6.2 Robust LQG Regulators

Two separate issues of robustness are addressed in the following sections. The first is the idea of robustifying a system against differences between the real world and the finite-dimension, low-order model that is chosen for controller and Kalman filter design. As discussed in Chapter V, the model to represent the real world (the truth model) and the lower-order controller model were purposefully established so that the differences between them occur in a specific high frequency range. As proposed by Doyle and Stein (Ref 6), the stability characteristics of a full-state feedback system can be recovered by adding white noise at the control entry points during the process of filter tuning. The validity

of this claim is examined by looking first at a purposefully reduced-order controller evaluated against a truth model of the same dimension with first-order actuator dynamics. That is, the low-order controller model is evaluated as if it were a perfect representation of the real-world system. The performance of this controller is the best that can be expected with a Kalman filter to estimate the states. Then, the higher-order actuator dynamics are introduced into the truth model to demonstrate the effects of ignored states on the performance of the system. Next, the robustification techniques are introduced to attempt to recover the stability characteristics of a reduced order but full-state feedback system. It would be desirable also to examine the performance of a full-state feedback system to evaluate the claim that the stability robustness properties of the filter-based controller will asymptotically approach those of the full-state feedback controller using the Doyle and Stein technique. Unfortunately, the software used to design and evaluate LQG regulators does not include the option of designing an LQ full-state controller. An attempt was made to approximate full-state feedback by setting the measurement noise intensities to small values and assuming a measurement was available for each state, but the results were inconclusive. Therefore, performance comparisons are only presented between designs using the robustified or unmodified Kalman filter.

As discussed in Chapter V, the difference between the controller design model and the truth model against which the performance is evaluated lies in the dynamics model for the actuators. Open loop frequency responses for the third-order model and the first-order model are shown in Figure (6-1). It is seen that the frequency responses differ

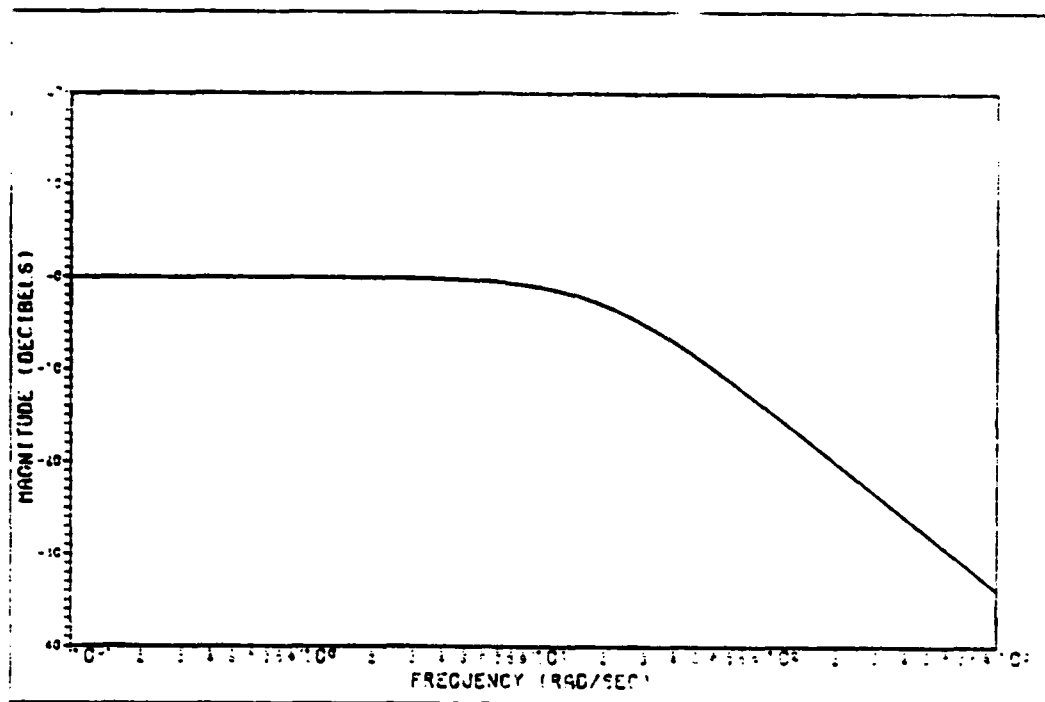


Figure 6-1a: Open Loop Frequency Response for First-Order Actuator Model

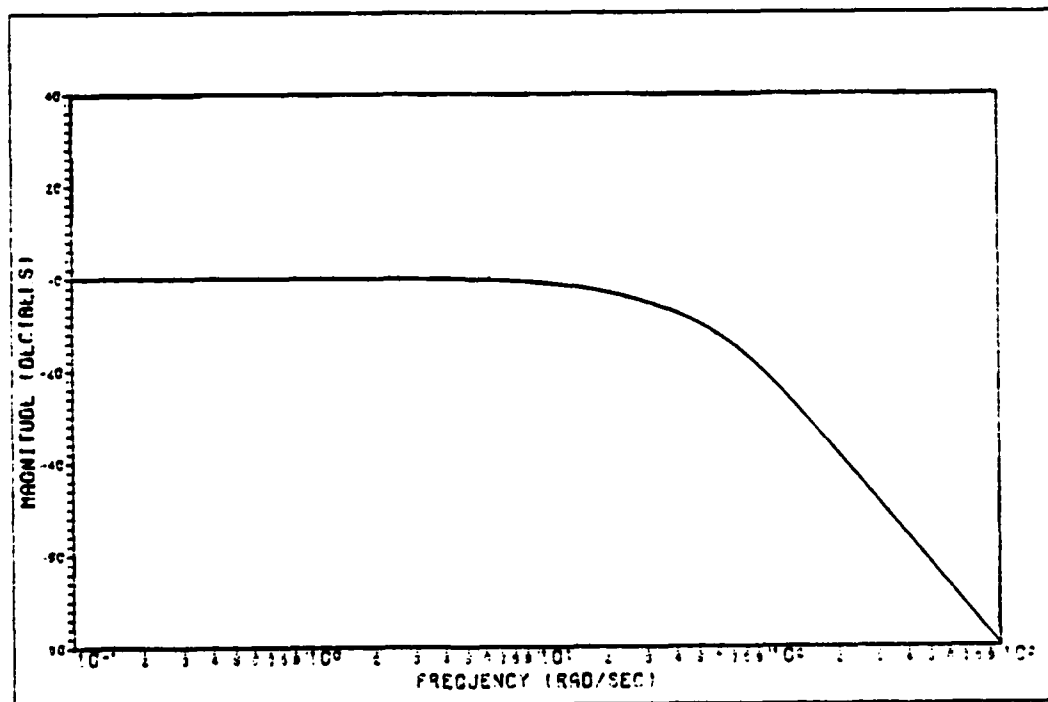


Figure 6-1b: Open-Loop Frequency Response for Third-Order Actuator Model

in the region around 70 rad/sec and beyond, where the complex poles are located in the third-order model. Robustification of the LQG controller based on the reduced-order model can be accomplished by adding white noise to the filter's system model at the point of entry of  $\underline{u}$ , or, by adding time-correlated noise with the primary power concentrated around the region of the model inadequacy, i.e., around 70 rad/sec and higher.

One first-order and one second-order shaping filter transfer function are considered, each of which concentrates the highest strength of the time-correlated noise in the region where the design model and truth-model actuator dynamics differ. The first-order shaping filter is described by the transfer function

$$\frac{x_u}{w_u} = \frac{s + 0.5}{s + 50} \quad (6-1)$$

A power spectral density plot of the time-correlated noise generated by this shaping filter is shown in Figure (6-2a).

The second-order shaping filter is described by the transfer function

$$\frac{x_u}{w_u} = \frac{s + 0.5}{(s + 50)(s + 400)} \quad (6-2)$$

Figure (6-2b) shows the power spectral density function for the time-correlated noise generated by the above shaping filter. The values in Equations (6-1) and (6-2) were chosen by examining power spectral density plots of the time-correlated noise with the poles and zeroes of the shaping filters in different locations. The chosen values generate the

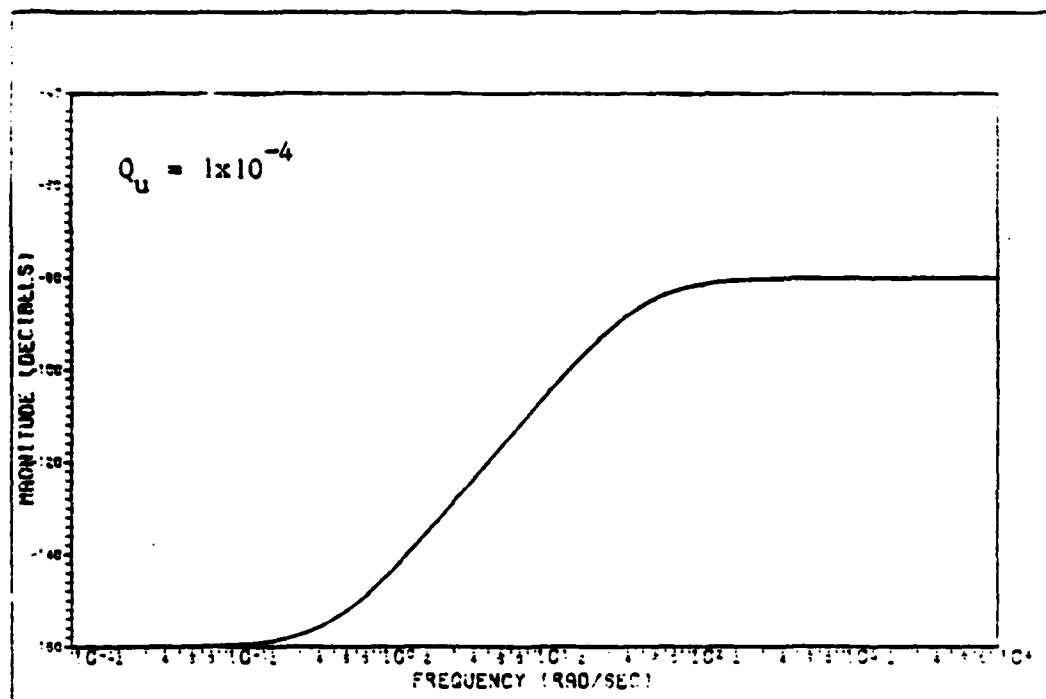


Figure 6-2a: First-Order Colored Noise Process  
Power Spectral Density

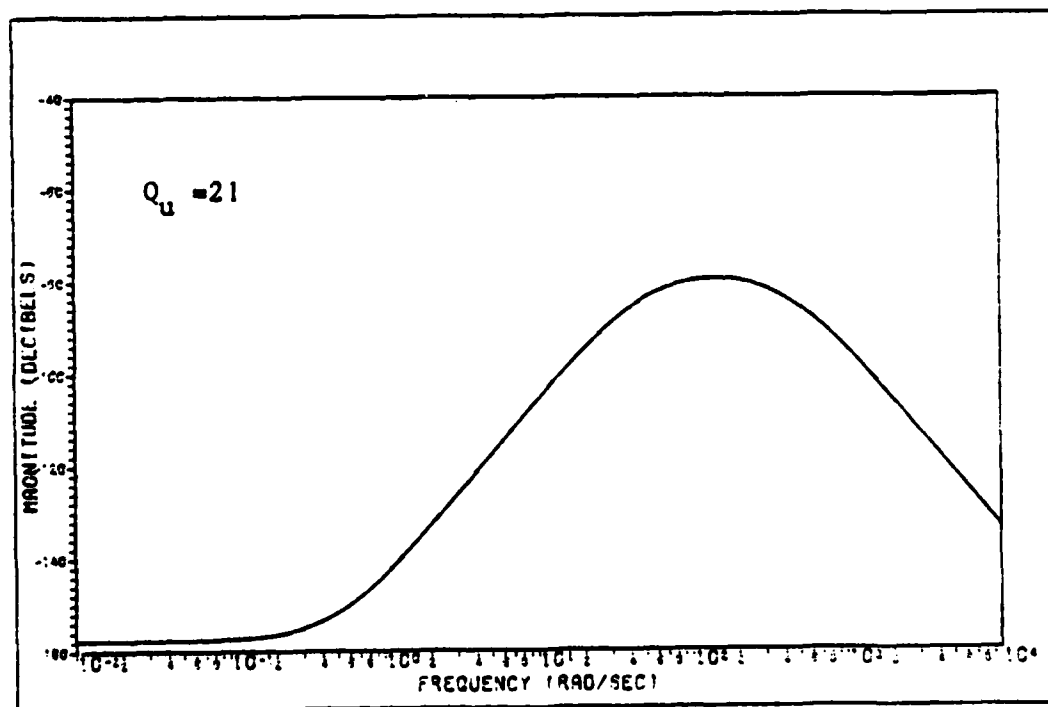


Figure 6-2b: Second-Order Colored Noise Process  
Power Spectral Density

desired noise, the strength of which is highest in the frequencies where the truth and design model frequency responses differ.

The second robustness issue is concerned with operation of the controlled system at flight conditions other than the design condition. The differences between the real world and the reduced-order design model now occur in other frequency ranges besides those given by the actuator dynamics models. Since these frequency ranges are now unknown, the use of white input noise would be motivated since it injects equal uncertainty into the system model over the entire frequency spectrum. The additional performance degradation or instabilities that arise in going from the design condition to an off-design condition with a Kalman filter in the loop are examined to determine the success of the Kalman filter robustification techniques. That is, if the full-state feedback system (based on a reduced-order design model) is stable, then the LQG system (which may well yield an unstable closed-loop system with the unrobustified filter in the loop) can be stabilized with the addition of white input noise (as claimed by Ref 6) and possibly by colored noise if the robustification is applied over the appropriate frequency ranges.

The cost-weighting matrices used in Equation (2-3) are the same for the continuous-time, discretized, and sampled-data regulators (discretizing (2-3) to yield (2-69) is accomplished as in Ref 24). The values used are based on those given in References 12 and 22, with some iteration on the control weightings to achieve designs with commanded control surface deflections that do not exceed physical limits of the horizontal tail and tailing-edge flap. The state, control and cross-weighting matrices are given by

$$W_{xx} = \begin{bmatrix} 50 & & -0- \\ & 50 & \\ -0- & & 150 \\ & & & -0- \end{bmatrix} \quad (6-3)$$

$$W_{xu} = [0] \quad (6-4)$$

$$W_{uu} = \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix} \quad (6-5)$$

It is desired to make a direct comparison between the performance of the controlled system with white and time-correlated input noise added to the model upon which the Kalman filter is based, where the maximum intensity of the time-correlated noise is equal to the intensity of the white noise. By equating the maximum magnitudes of the white and time-correlated noise, it can be determined if there are performance benefits in choosing one type of noise over another by comparing the degree of robustification achieved by each. This is accomplished using the frequency domain shaping filter design techniques in Reference 23, which state that

$$PSD_0(s) = G(s) G(-s) PSD_1(s) \quad (6-6)$$

where  $PSD_1$  and  $PSD_0$  are the power spectral densities of the input and output of a shaping filter, respectively.  $G(s)$  is the transfer function of the shaping filter. It is desired to have the maximum magnitude of

the right-hand side of Equation (6-6) be equal the magnitude of the white input noise, (the scalar white noise parameter,  $q^2$  of Equation (2-64)). Thus, by setting  $PSD_o$  equal to  $q^2$  and solving for the value of  $PSD_1$  which makes this true, the value of  $Q_u$  in Equation (4-12) can be found for the time-correlated noise. Thus, it is the intensity,  $Q_u$ , of the white driving noise,  $w_u(t)$  to the shaping filter which will generate a time-correlated noise with a maximum intensity of  $q^2$ . Table (6-1) lists the values of  $Q_u$  which make the maximum intensity of the

Table 6-1

Strength of Dynamic Driving Noise to Shaping Filters

$G(s) = 1$	$G(s) = \frac{s + 0.5}{s + 50}$	$G(s) = \frac{s + 0.5}{(s + 50)(s + 400)}$
$q^2$	$Q_u$	$Q_u$
$1 \times 10^{-6}$	$1 \times 10^{-6}$	0.21
$1 \times 10^{-4}$	$1 \times 10^{-4}$	21.
$1 \times 10^{-2}$	$1 \times 10^{-2}$	2100.

time-correlated noise equal to that of the white noise. The values were arrived at using the software described in Reference 20.

#### 6.2.1 Continuous-Time LQG Regulators at Design Condition

Figure (6-3) shows time histories of the mean and standard deviation of the aircraft state,  $\Theta$ , for an eight-state controller evaluated against a truth model of the same dimension (the truth model and controller model are identical as given in Section 5.4). A perfectly known initial condition of one degree (or 0.0175 radians) was placed on  $\Theta$ , and the



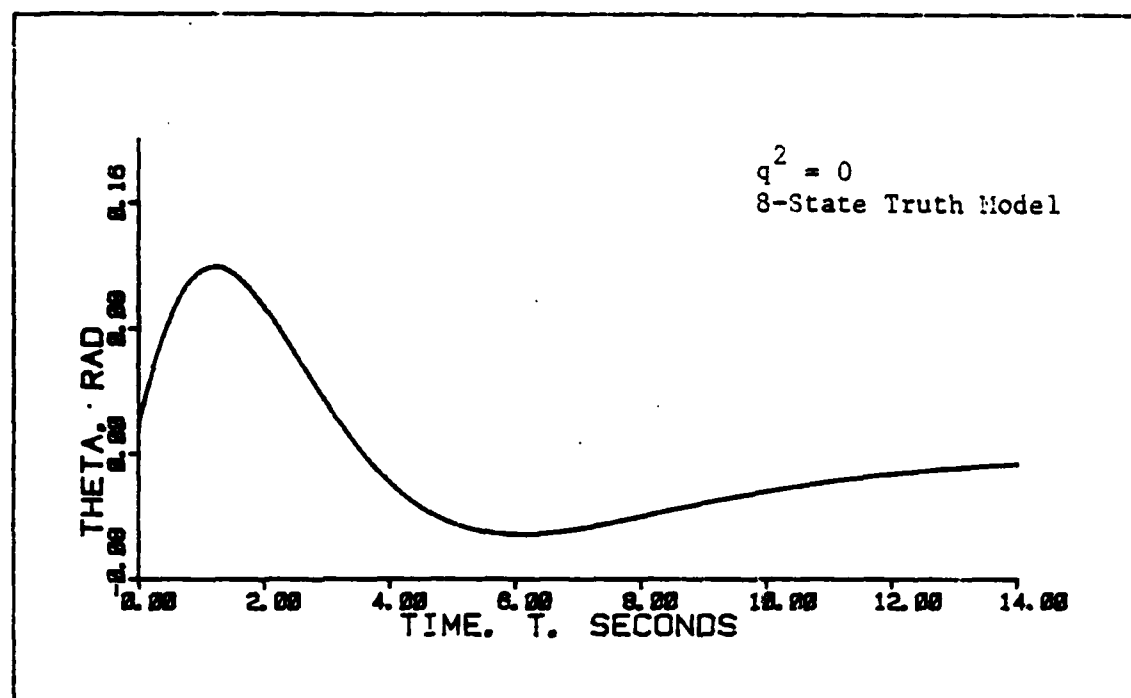


Figure 6-3a: Mean of  $\Theta$  at the Design Condition

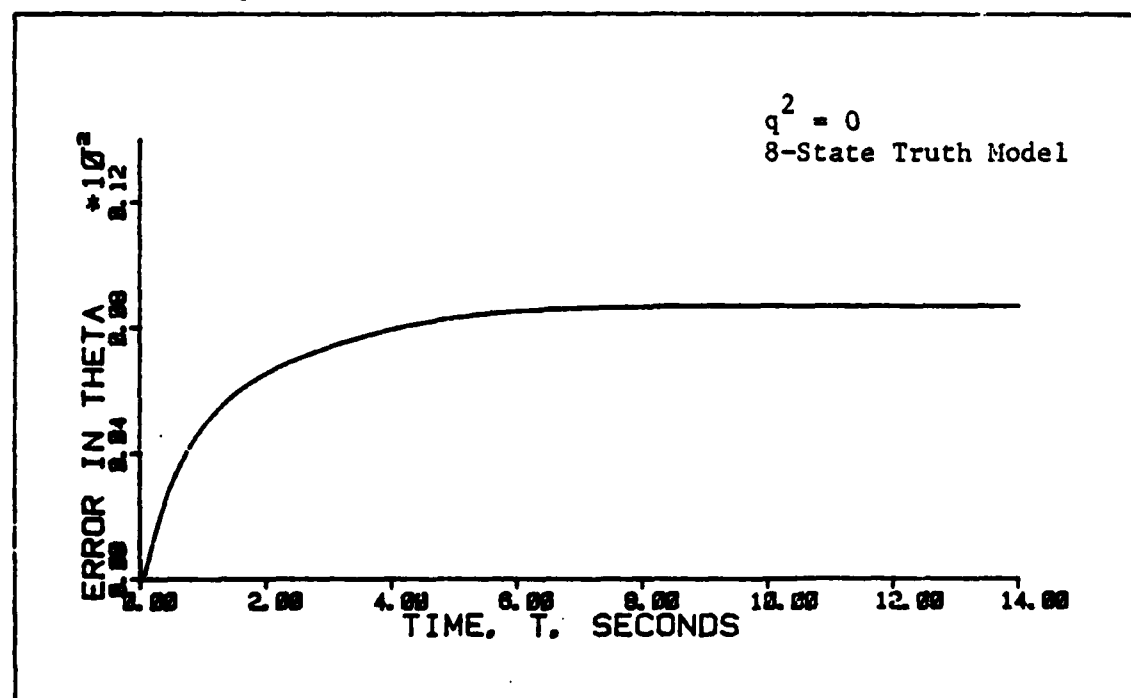


Figure 6-3b: Standard Deviation of  $\Theta$  at the Design Condition

plot shows the response of  $\Theta$  to the initial perturbation and to the dynamic driving noise built into the model. The trends are similar for all three aircraft states, therefore only one will be examined in detail in this chapter.

The time response is stable, although slow. The slow response is typical of systems with a Kalman filter to estimate states. The added dynamics of the filter tend to slow down the response. However, the controller does respond to the perturbation and regulate the state back to zero. Figure (6-4) shows the response of the same controller when specific ignored states are now accounted for in the truth model. This figure is the result of a performance analysis when third-order actuator dynamics are included in the twelve-state truth model, while only first-order dynamics are used in the eight-state design model. As can be seen, the system is still stable, but the mean of  $\Theta$  response has degraded substantially, especially in steady-state.

Figures (6-5a) and (6-5b) show the response of the same state with a white Gaussian noise of strength  $q^2 = 1 \times 10^{-4}$  injected into the system model. The transient time, the overshoot, and the error in the final mean value of  $\Theta$  have all been substantially improved with the noise addition. The standard deviation of  $\Theta$  has not changed noticeably from the previous case. The trend for lower and higher  $q^2$  values is shown in Figures (6-5c,e,d,f). It is seen that noise of a very small intensity ( $1 \times 10^{-6}$ ) enhances the robustness properties dramatically, and increasing the intensity of the noise only serves to change the transient characteristics of the mean time response and the magnitude of the standard deviation.

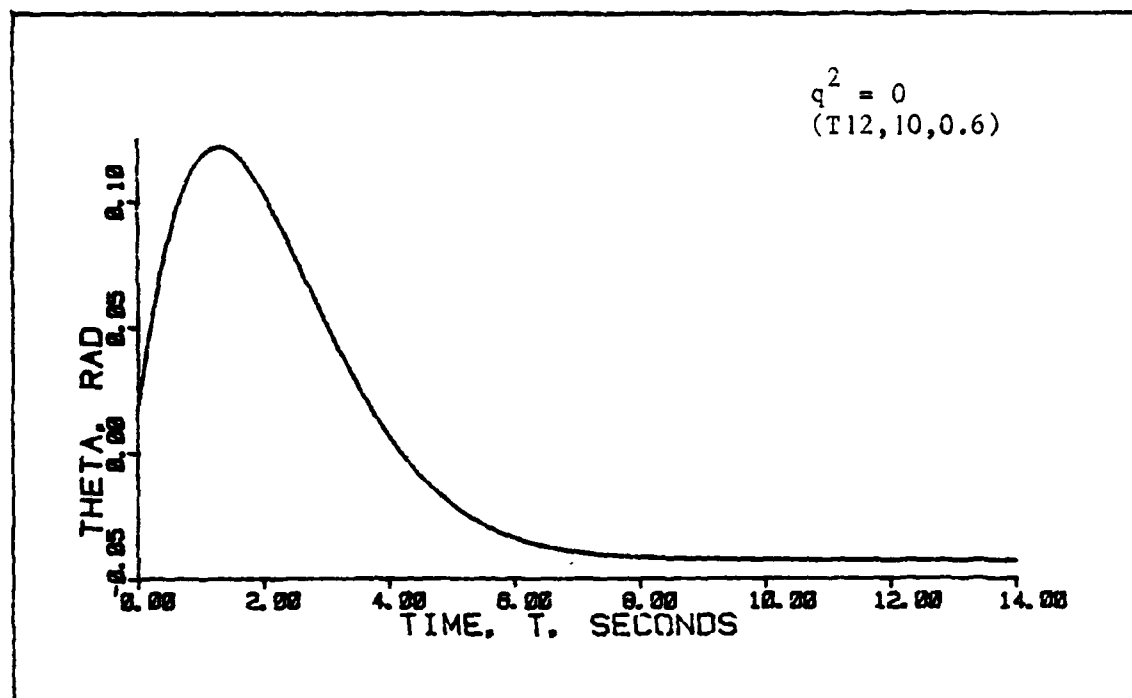


Figure 6-4a: Mean of  $\Theta$  With no Noise Addition

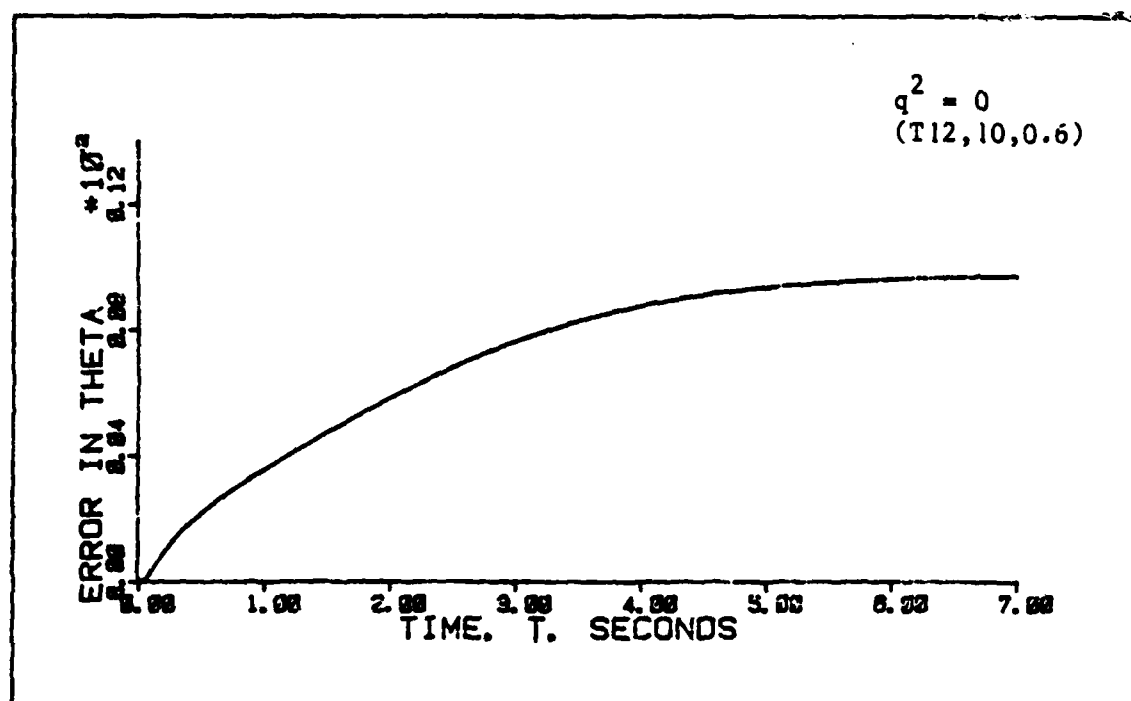


Figure 6-4b: Standard Deviation of  $\Theta$  With No Noise Addition

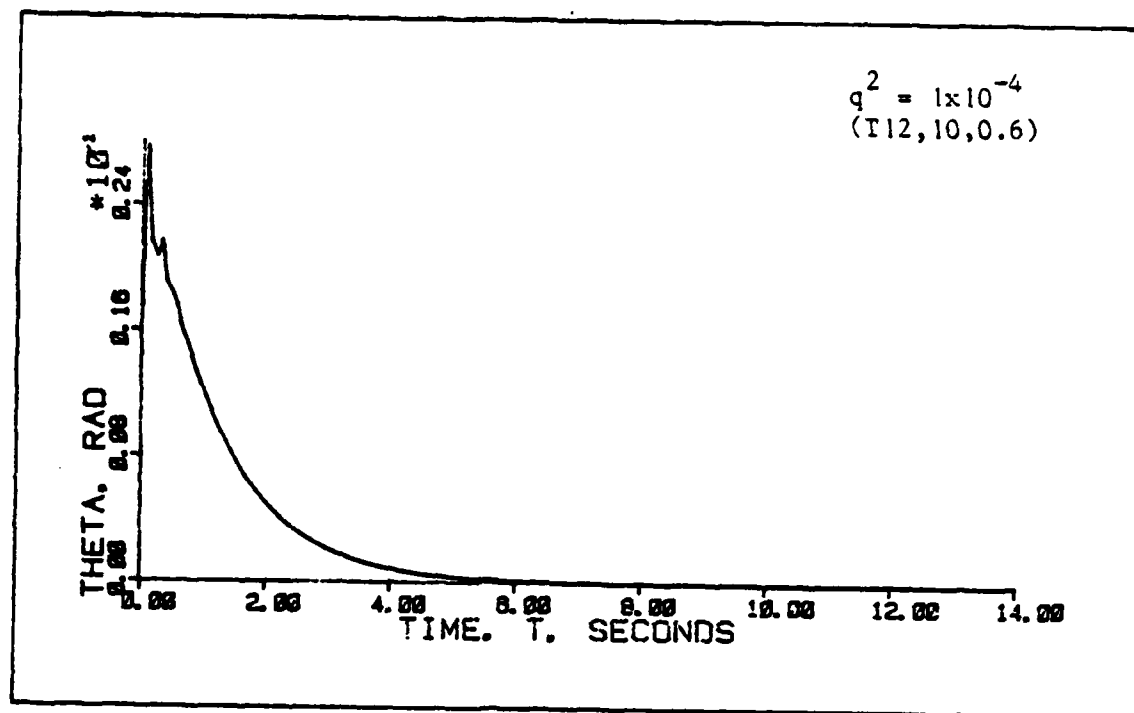


Figure 6-5a: Mean of  $\theta$  With White Noise Addition

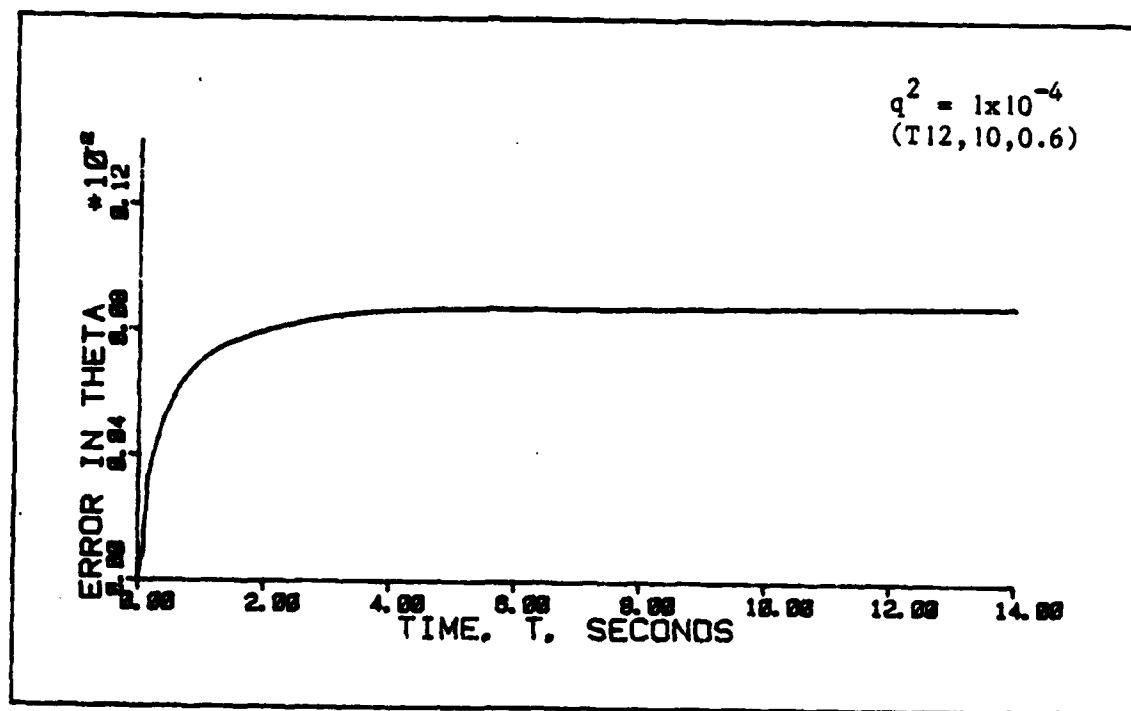


Figure 6-5b: Standard Deviation of  $\theta$  With White Noise Addition

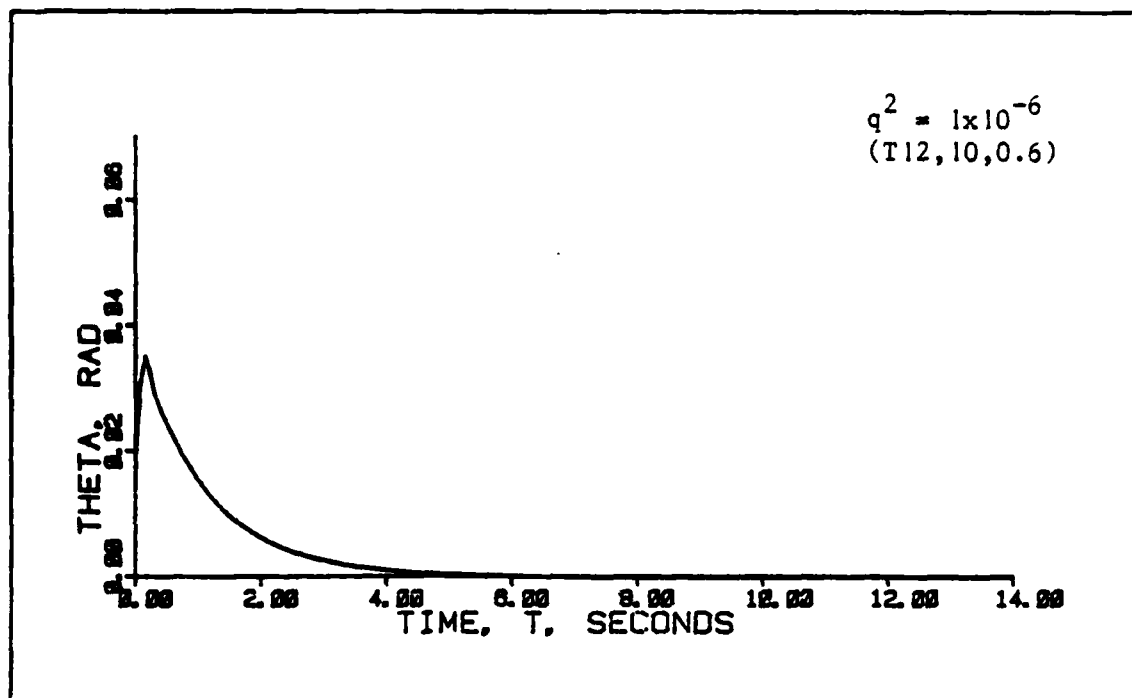


Figure 6-5c: Mean of  $\theta$  With White Noise Addition

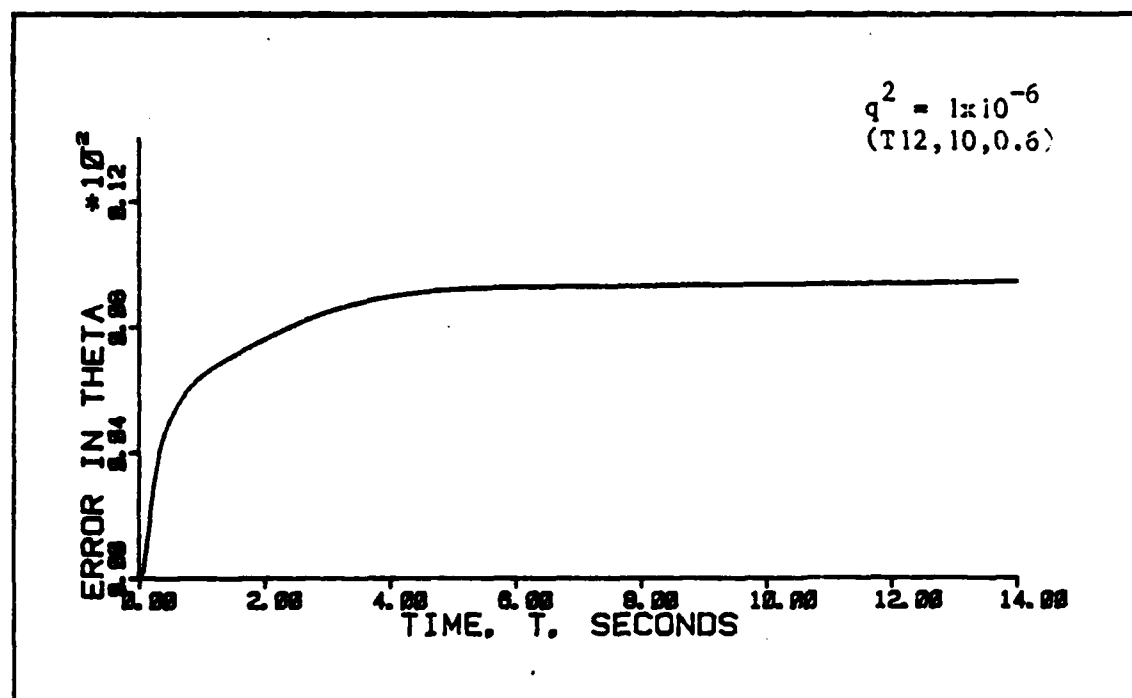


Figure 6-5d: Standard Deviation of  $\theta$  With White Noise Addition

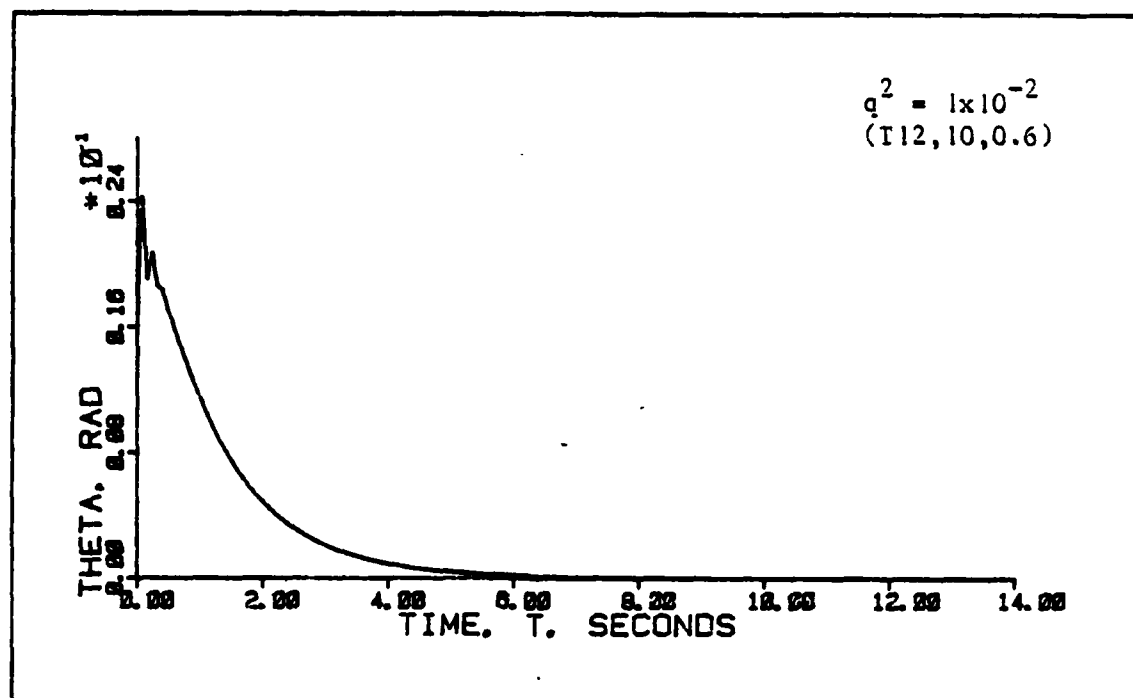


Figure 6-5e: Mean of  $\theta$  With White Noise Addition

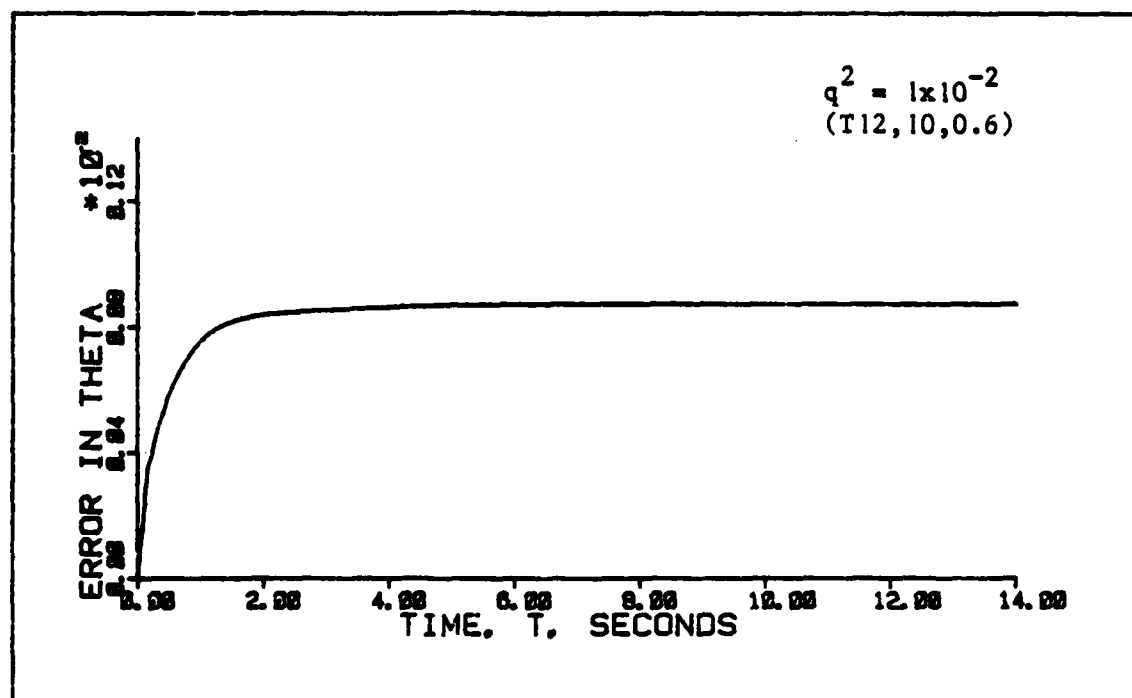


Figure 6-5f: Standard Deviation of  $\theta$  With White Noise Addition

Then, as shown in Figure (6-6), time-correlated noise generated by a first-order shaping filter is injected into the system model. The maximum power spectral density of the noise is  $1 \times 10^{-4}$ . The figure demonstrates that the improvement in the transient time, overshoot, and final value of  $\Theta$  is as dramatic as for the white noise case. Again, the standard deviation has not changed enough to be discernible in the figure.

Finally, time-correlated noise generated by a second-order shaping filter is injected into the design model. As can be seen in Figure (6-7), the transient time is approximately the same as for the original system, but much slower than the cases above. However, the state is converging to zero, and the overshoot has been reduced to about half of the original value. Any change in the steady-state value of the standard deviation is not noticeable in the figure.

Thus, Figures (6-3) through (6-7) demonstrate that the robustification techniques can recover the stability robustness characteristics that would be expected from a full-state feedback system. The methods of injecting white and first-order colored noise produce very similar improvements, while the second-order colored noise produced less but still noticeable benefits in the time response.

However, it was stated in previous chapters that the addition of noise into the design model will degrade the performance (as measured by the standard deviation of the states or how well known the states are) of the system at the design conditions. Additionally, it was claimed that the performance degradation will be less for colored noise than for white noise. To evaluate this claim, Table (6-2) lists the steady-state values of the standard deviations of all three aircraft states with no input

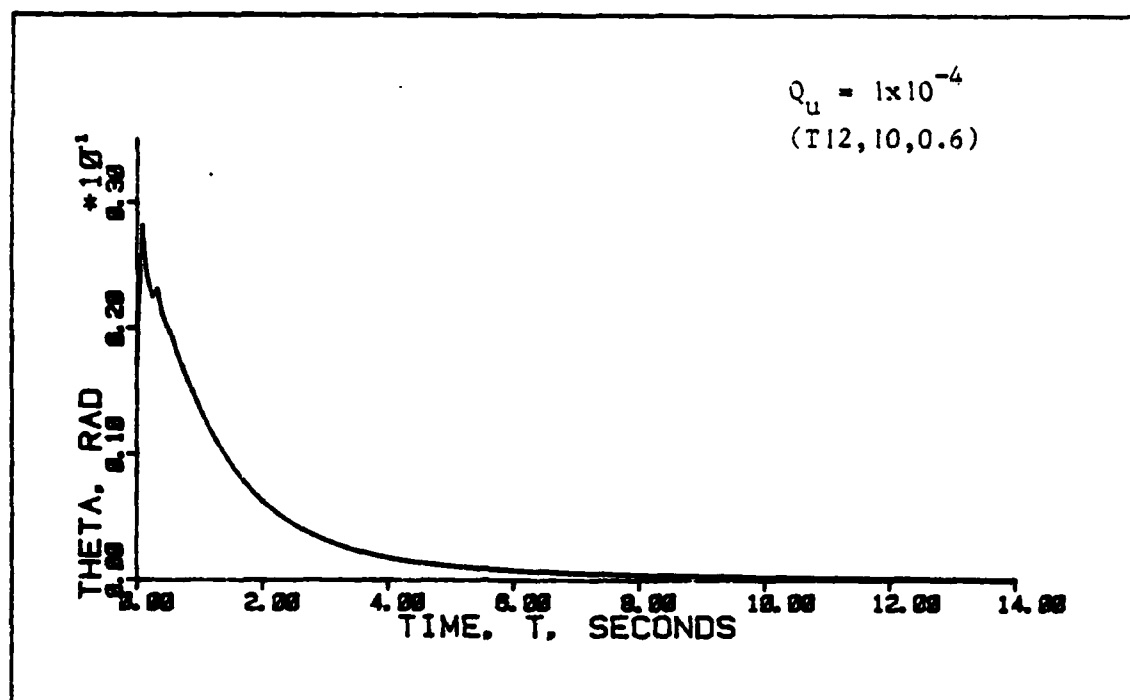


Figure 6-6a: Mean of  $\theta$  With First-Order Colored Noise Addition

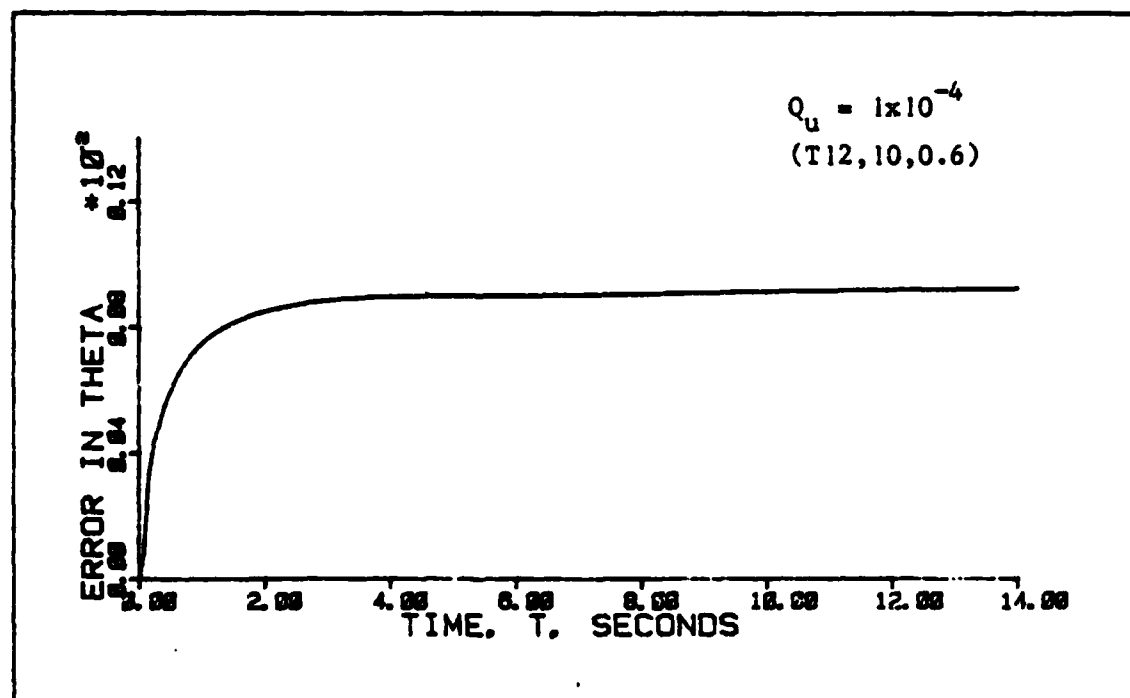


Figure 6-6b: Standard Deviation of  $\theta$  With First-Order Colored Noise Addition



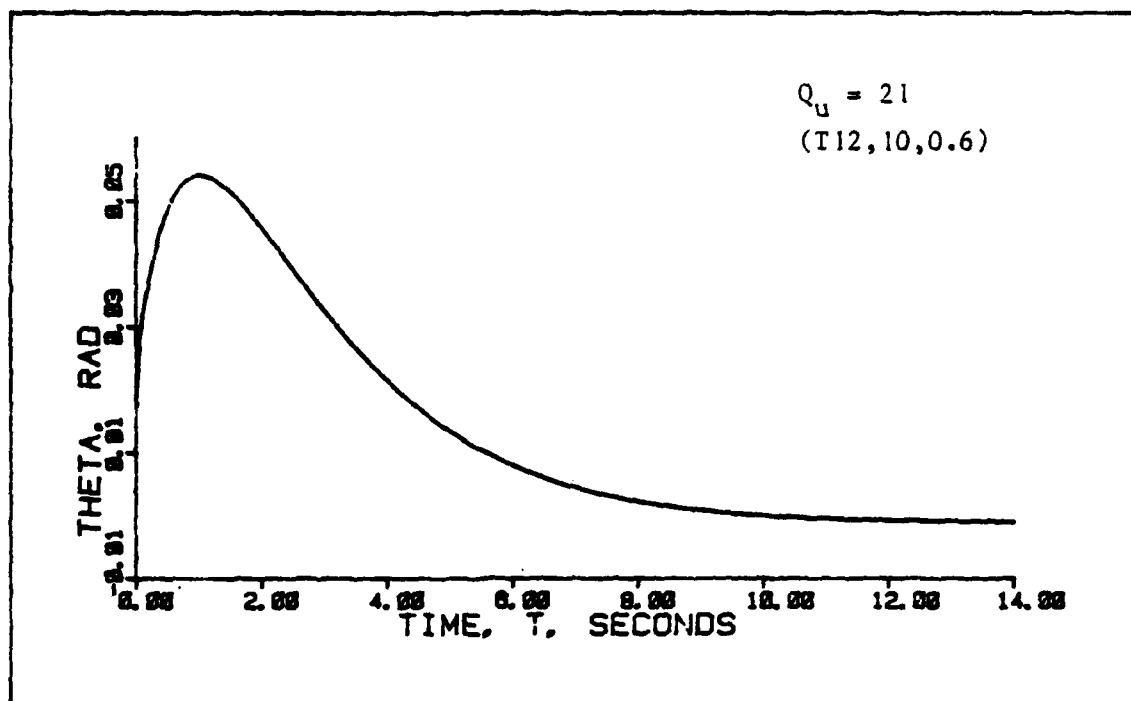


Figure 6-7a: Mean of  $\theta$  With Second-Order Colored Noise Addition

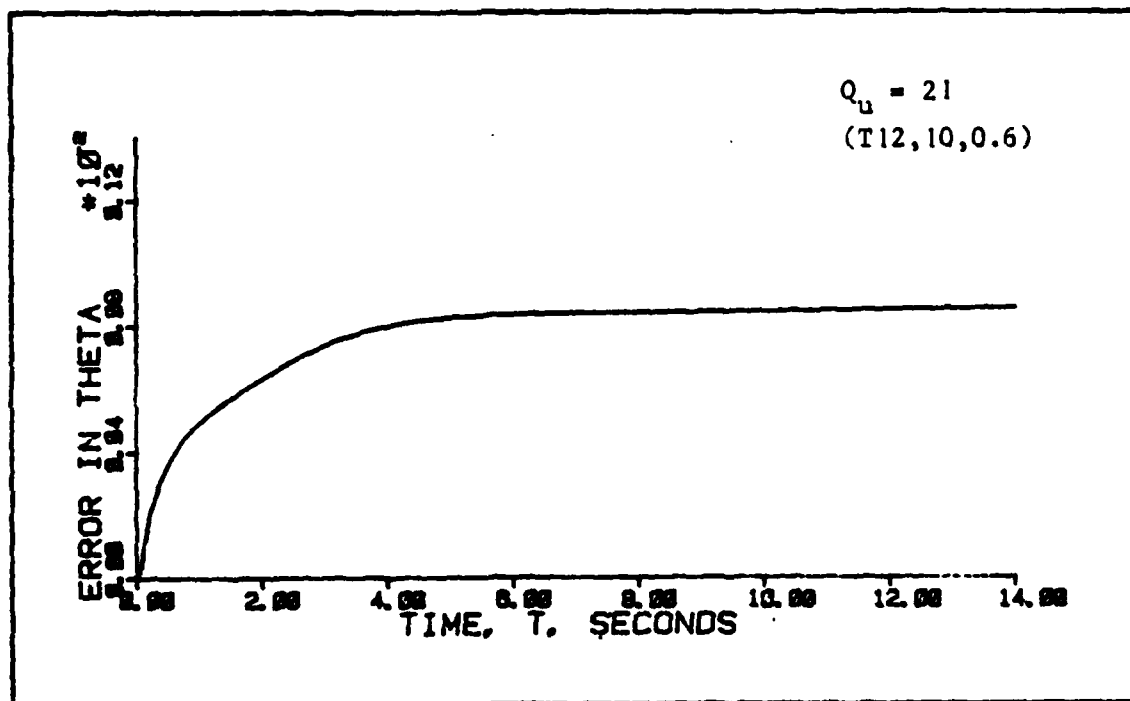


Figure 6-7b: Standard Deviation of  $\theta$  With Second-Order Colored Noise Addition

noise, white noise, and first-order and second-order time-correlated noise for one value of noise intensity. The values listed are the results of a performance evaluation where the unmodified controller design model and truth model are identical (both are of dimension eight at the design condition). The results shown in the table only partially substantiate the claim that time-correlated noise can minimize the performance degradation. The addition of white noise to the system model increased the standard deviation of all three aircraft states as expected.

It was expected that the addition of first-order colored noise would reduce the standard deviation of the aircraft states to values less than those for white noise, but still larger than the case with no noise addition. Table (6-2) shows that this is not the case. Rather, the standard deviations have increased, if only slightly for  $\theta$  and  $\alpha$ . Only the pitch rate,  $q$ , has decreased as expected. Recall from Figures (6-5a) and (6-6a) that white and first-order colored noise produce similar mean of theta responses. That is, the robustness enhancement is very similar for the two cases. However, even though the added uncertainty is applied over a

Table 6-2

Comparison of Steady-State Deviations of Aircraft States at the Design Condition for a Continuous Time System

	$q^2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No noise	0	-	$8.705 \times 10^{-4}$	$5.035 \times 10^{-3}$	$1.200 \times 10^{-3}$
White Noise	$1 \times 10^{-4}$	-	$9.590 \times 10^{-4}$	$5.306 \times 10^{-3}$	$1.978 \times 10^{-3}$
1st-order Shaping Filter	-	$1 \times 10^{-4}$	$1.006 \times 10^{-3}$	$5.823 \times 10^{-3}$	$1.761 \times 10^{-3}$
2nd-order Shaping Filter	-	21	$8.704 \times 10^{-4}$	$5.039 \times 10^{-3}$	$1.201 \times 10^{-3}$

more limited frequency range when first-order colored noise is injected into the model, the standard deviations have increased slightly. This is not well understood, except to say that time-correlated noise actually changes the structure of the Kalman filter design model. The added complexity of the design model may overcome any benefits that are realized because the noise was not applied over all frequencies as in the white noise case.

As expected, the performance at design conditions is degraded less by using a second-order filter as opposed to a first-order filter or white noise. However, the greater complexity of the Kalman filter design model (four additional states in this case) is not justified by the performance benefits seen in Table (6-2). In this instance, white noise injected into the system design model provides the desired robustification while not degrading the performance at the design condition substantially.

Table (6-3) presents similar steady-state standard deviation information about the three aircraft states as shown in Table (6-2). However, the values listed here are the results of a performance evaluation of the eight-state controller evaluated against the twelve-state truth model, accounting for higher-order actuator dynamics. An important difference between Tables (6-2) and (6-3) is the effect of adding white input noise on the standard derivation of  $\Theta$ . At the design condition, adding white noise detunes the filter and results in a performance degradation. However, when third-order actuator dynamics are included in the truth model, adding white noise improves the tuning for the  $\Theta$  channel, and the standard deviation decreases for this state. Thus, when the performance

of the system is evaluated in an environment different from the design condition, the noise addition can result in a performance enhancement. It is seen that for this case, second-order time correlated noise again accomplishes improved filter tuning over first-order and white noise. The improvement, though, is still not substantial enough to justify the added complexity of the Kalman filter design model. White input noise, as stated above, accomplishes the desired robustness enhancement without adding states to the design model.

Table 6-3

Comparison of Steady-State Standard Deviations of Aircraft States with Higher-Order Actuator Dynamics for a Continuous-Time System.

	$q_2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No noise	0	-	$8.984 \times 10^{-4}$	$4.894 \times 10^{-3}$	$1.189 \times 10^{-3}$
White Noise	$1 \times 10^{-4}$	-	$8.850 \times 10^{-4}$	$5.689 \times 10^{-3}$	$4.979 \times 10^{-3}$
1st Order Shaping Filter	-	$1 \times 10^{-4}$	$9.232 \times 10^{-4}$	$5.813 \times 10^{-3}$	$5.092 \times 10^{-3}$
2nd Order Shaping Filter	-	21	$8.571 \times 10^{-4}$	$5.168 \times 10^{-3}$	$1.969 \times 10^{-3}$

### 6.2.2 Continuous-Time LQG Regulators at Off-Design Condition

Figure (6-8) shows the results of a performance analysis for the unrobustified system with a Kalman filter at an off-design flight condition (T12,20,0.6) for the eight-state controller design model. As shown in the figure, the system is unstable. The mean of  $\theta$  is diverging, and the standard deviation is growing with time. In addition, the inputs generated by the controller are growing with time beyond the actual limits of the control surfaces ( $\pm 23$  degrees for the horizontal tail and  $\pm 20$  degrees for the trailing-edge flap: Ref 27). This is demonstrated in the mean plots of Figure (6-9).

Figures (6-10a) and (6-10b) show the system responses at the same off-design flight condition, except now white noise of strength  $q^2 = 1 \times 10^{-4}$  has been injected into the filter's system model at the control entry points. The addition of this noise is sufficient to stabilize the system, driving the aircraft state towards zero and the standard deviation of the state to a finite value. Figures (6-10c,d,e,f) demonstrate the trend for a lower and higher value of  $q$ . As noted before, stability is recovered with a very low noise strength. Higher values change only transient characteristics and the magnitude of the standard deviation. Figure (6-11) demonstrates that the mean of the commanded controls are no longer exceeding the physical limits of the control surfaces. However, the large initial changes in the command inputs do exceed the actuator rate limits of the control surfaces because there is no weighting on input rates in the cost function for an LQG regulator.

For a time-correlated noise generated by a first-order shaping filter with a maximum intensity of  $1 \times 10^{-4}$ , the results are similar to those of

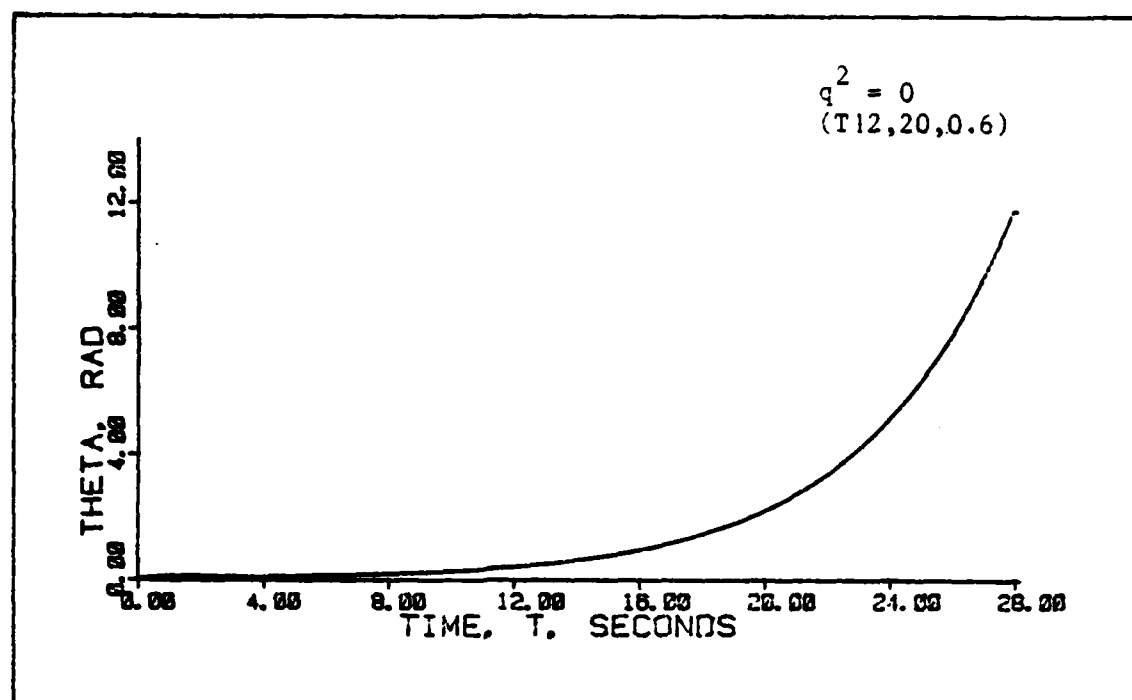


Figure 6-8a: Mean of Theta at Off Design Flight Condition

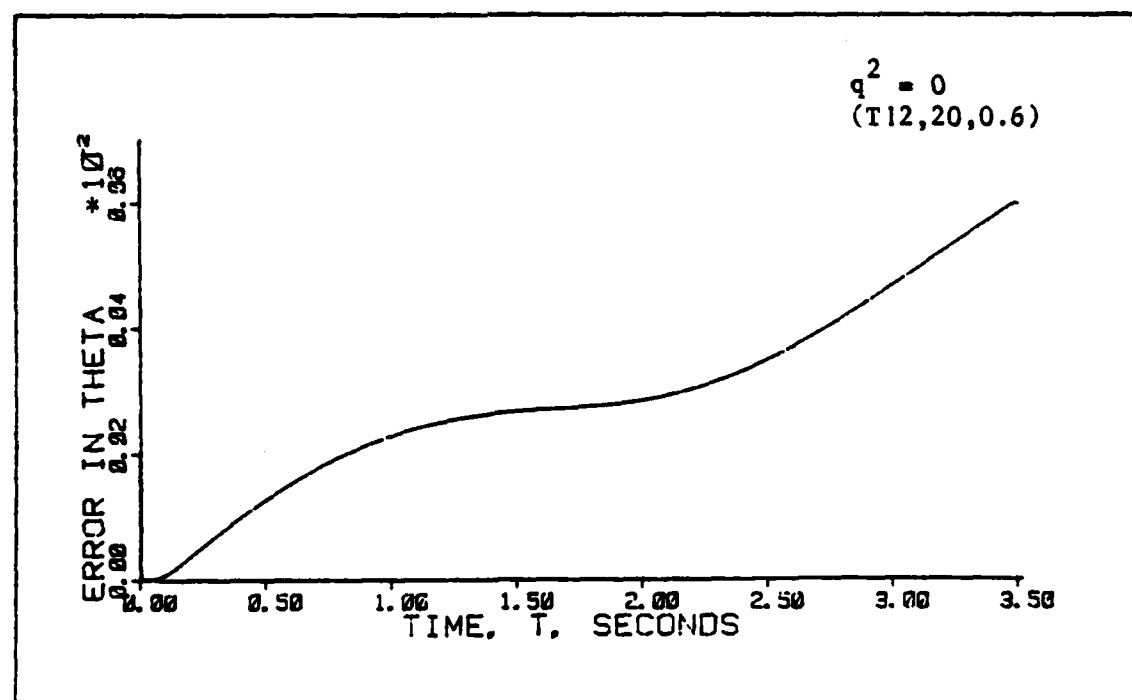


Figure 6-8b: Standard Deviation of  $\theta$  at Off-Design Flight Condition

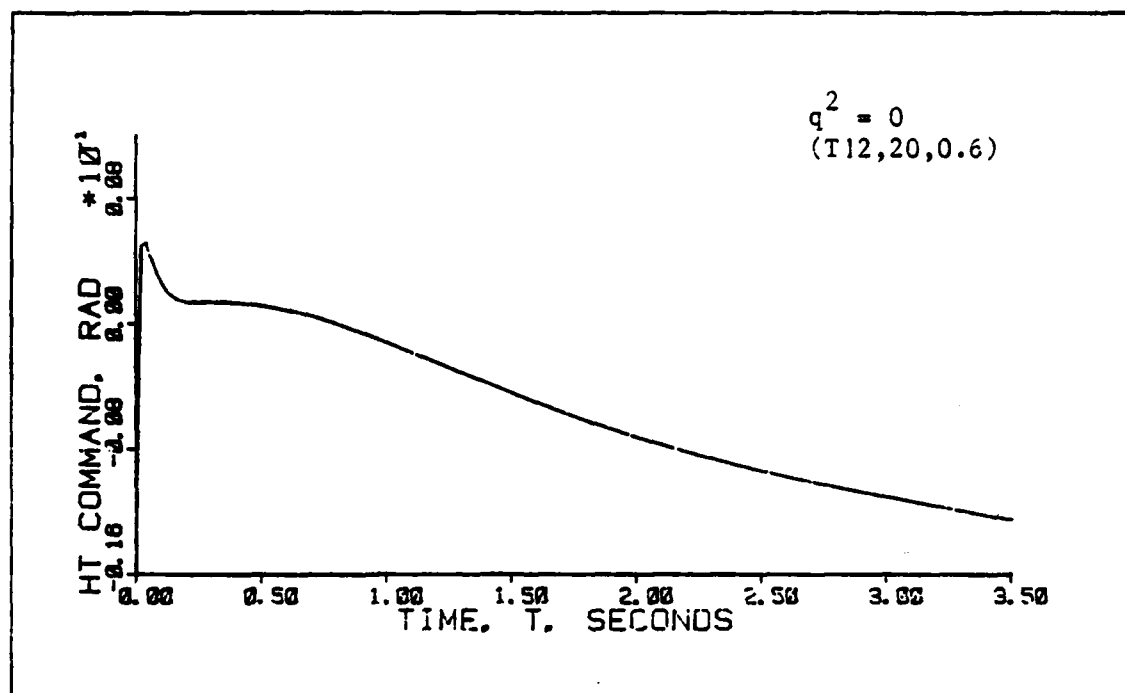


Figure 6-9a: Mean of Horizontal Tail Commanded Deflection

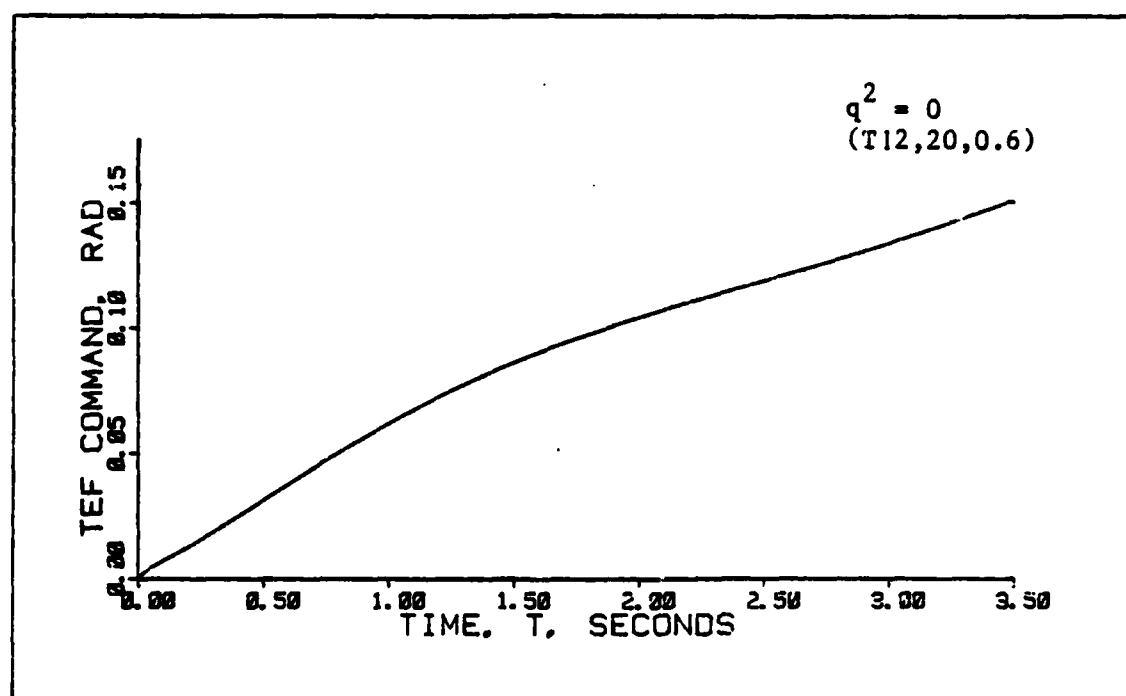


Figure 6-9b: Mean of Trailing Edge Flap Commanded Deflection

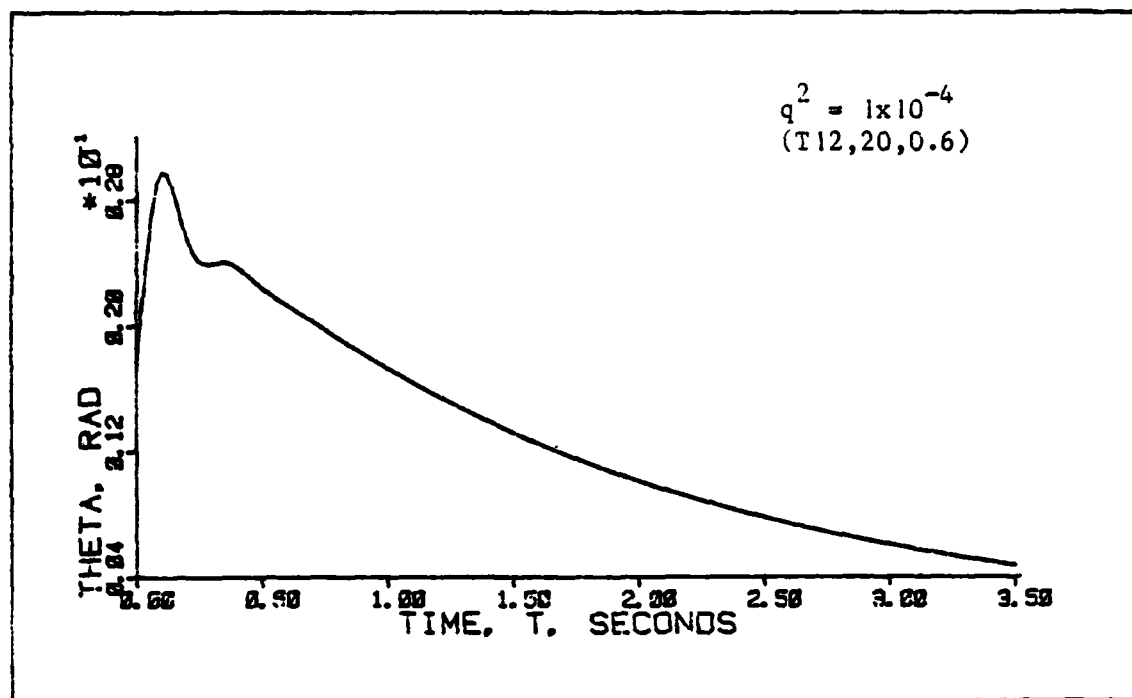


Figure 6-10a: Mean of  $\theta$  With White Noise Addition at Off-Design Flight Condition

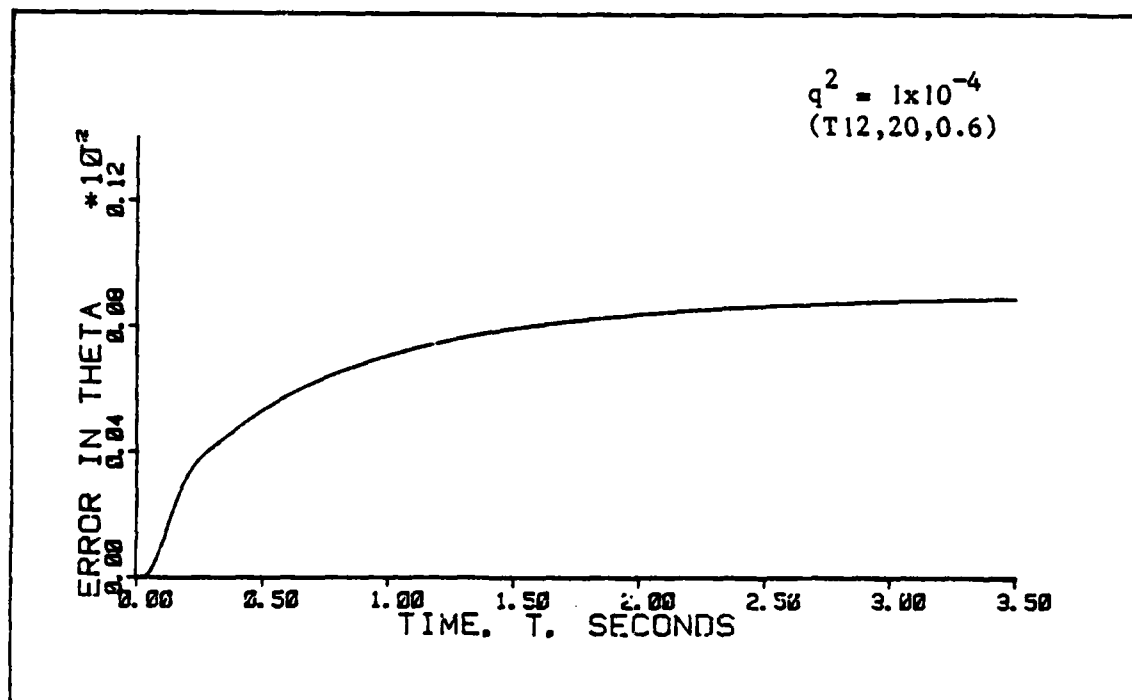


Figure 6-10b: Standard Deviation of  $\theta$  With White Noise Addition at Off-Design Flight Condition



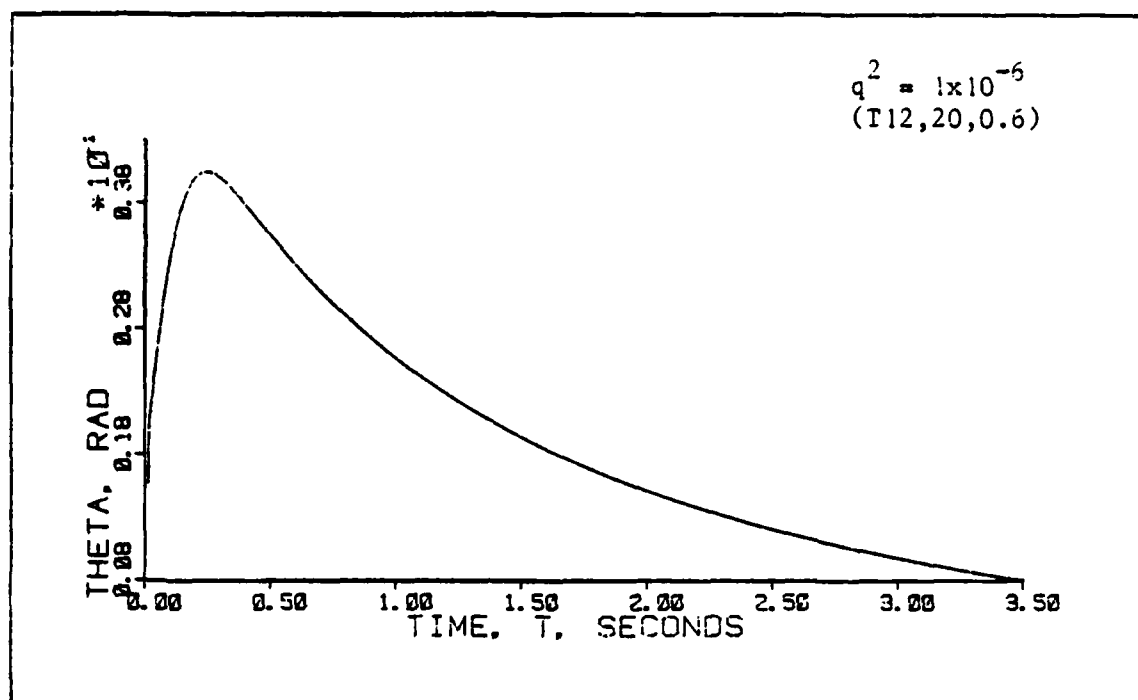


Figure 6-10c: Mean of  $\Theta$  With White Noise Addition with an Off-Design Flight Condition

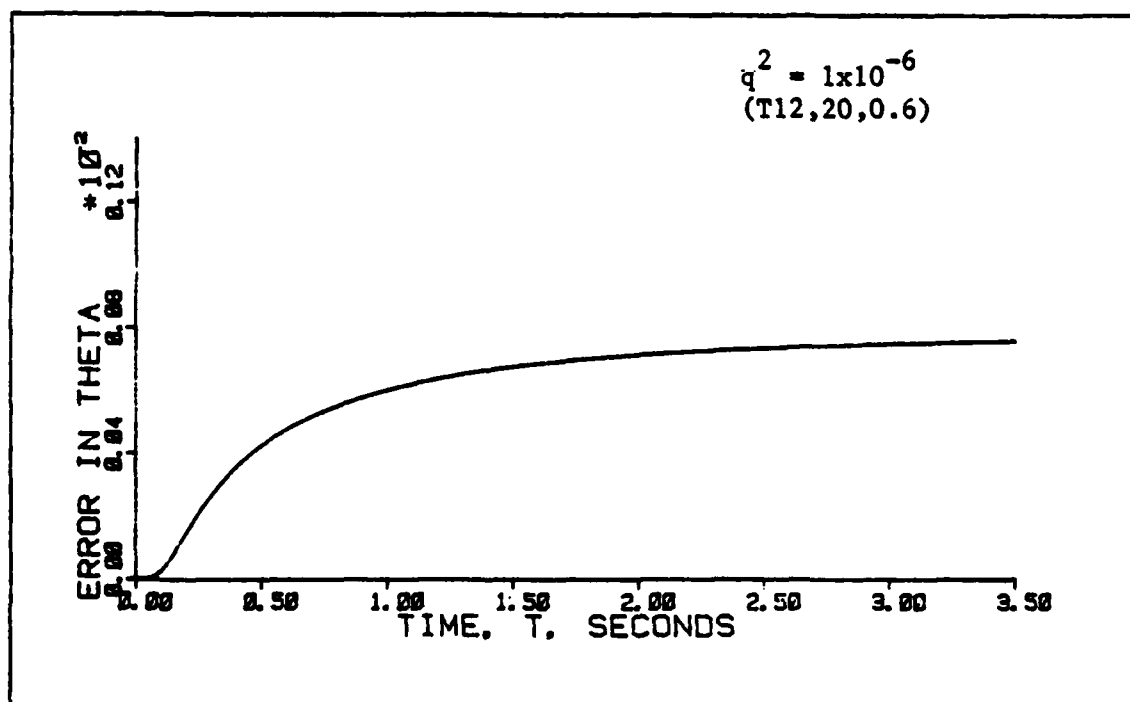


Figure 6-10d: Standard Deviation of  $\Theta$  With White Noise Addition at An Off-Design Flight Condition

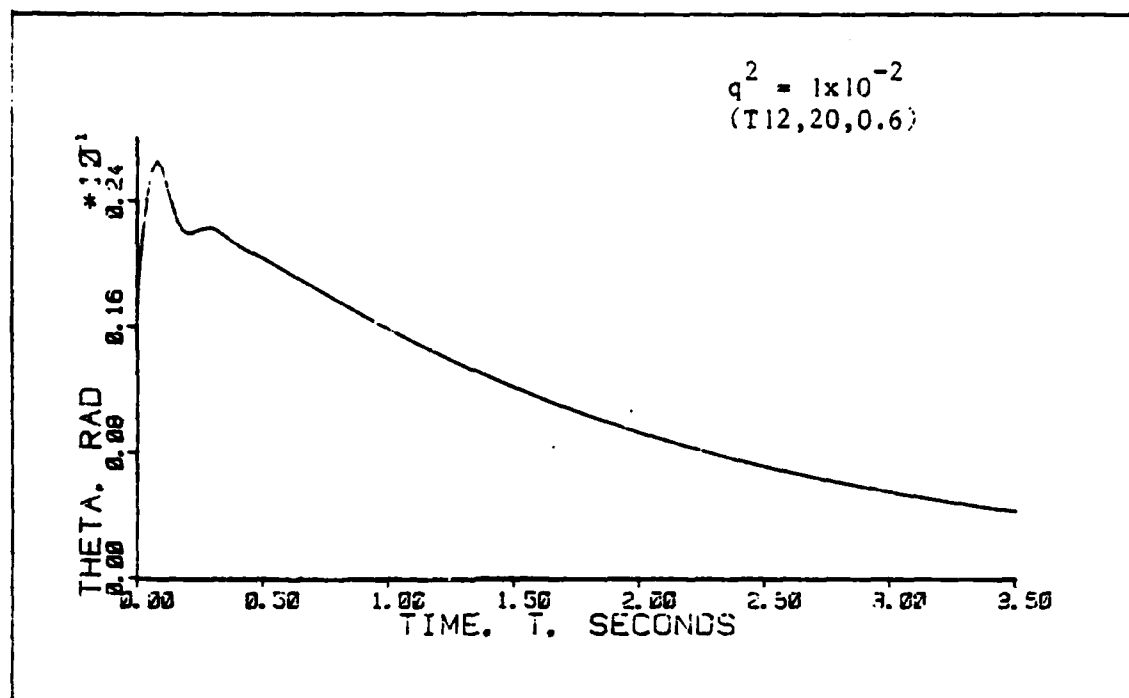


Figure 6-10e: Mean of  $\Theta$  With White Noise Addition at Off-Design Flight Condition

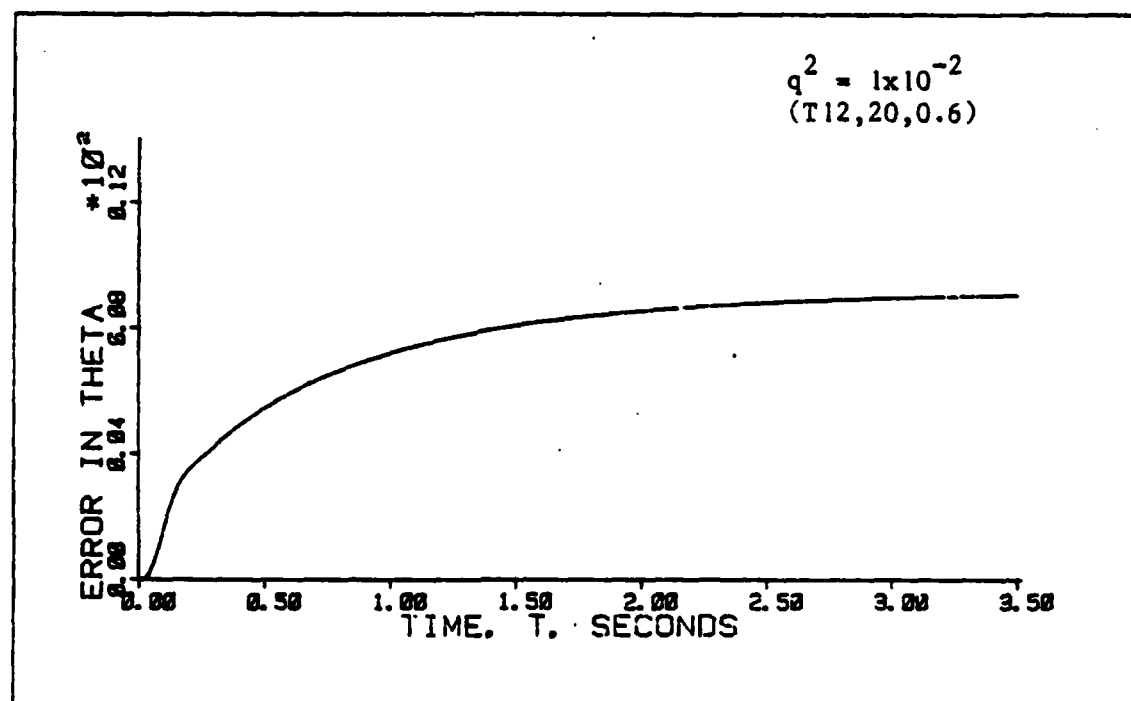


Figure 6-10f: Standard Deviation of  $\Theta$  With White Noise Addition at Off-Design Flight Condition

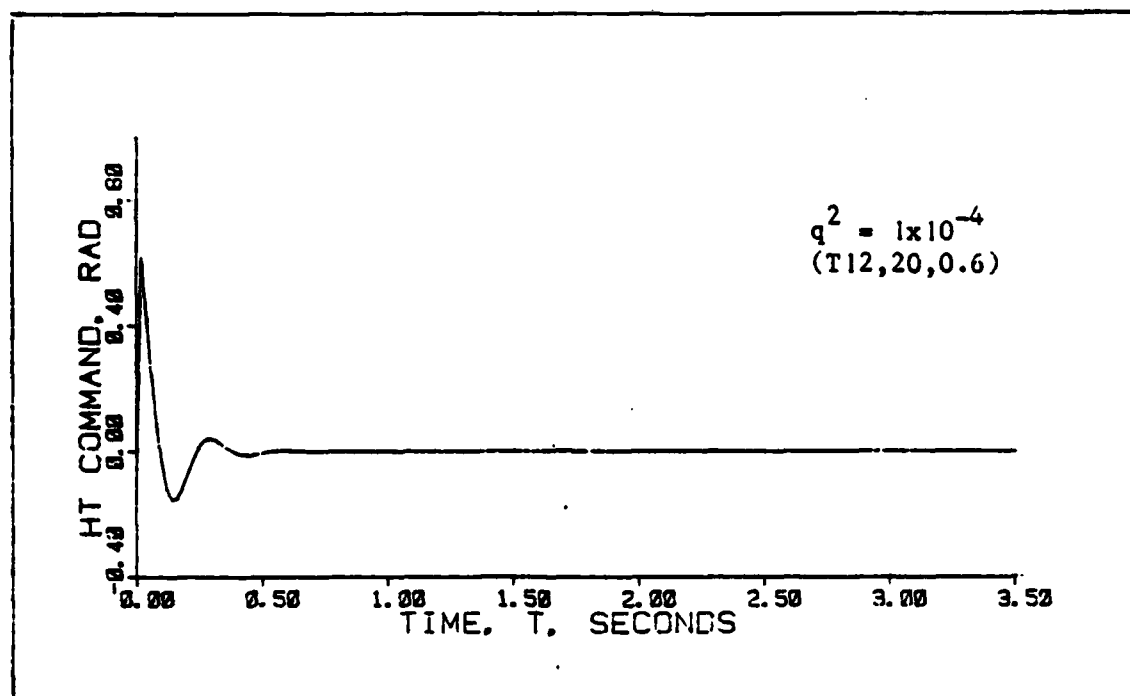


Figure 6-11a: Mean Horizontal Tail Commanded Deflection  
With White Noise Addition

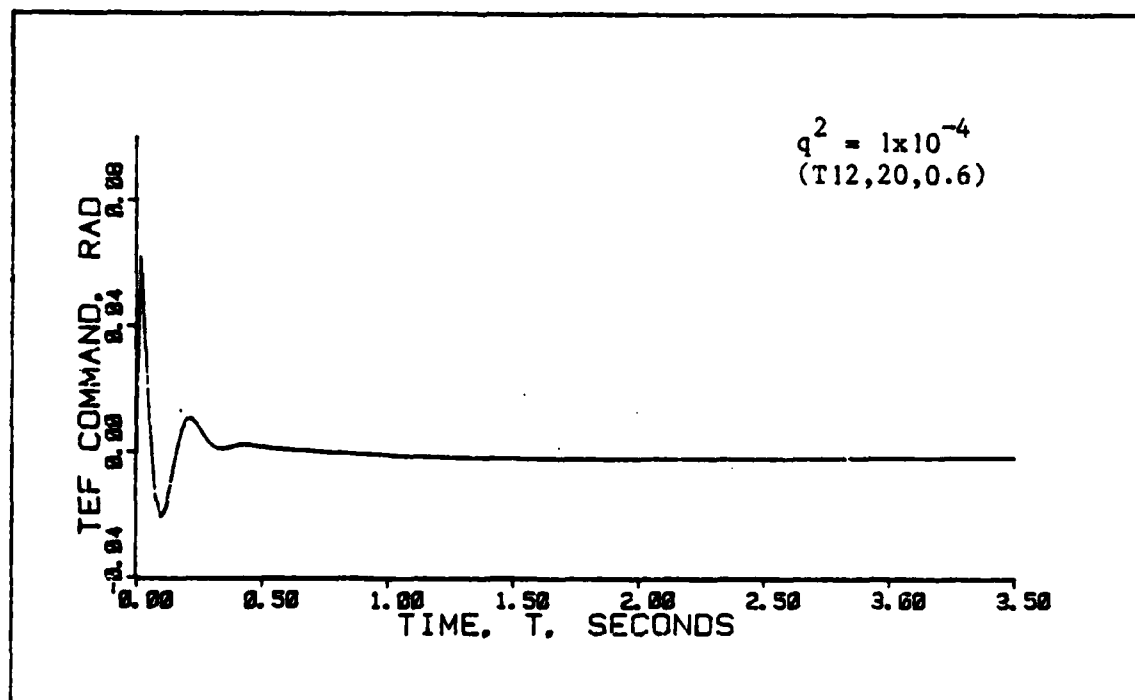


Figure 6-11b: Mean Trailing Edge Flap Commanded  
Deflection With White Noise Addition

the white noise. This is shown in Figure (6-12). At the off-design flight condition which was initially unstable with no noise addition, the response is stable with the mean of  $\theta$  approaching zero, and the standard deviation approaching a finite value.

As shown in Figure (6-13), the addition of colored noise generated by a second-order shaping filter with a maximum intensity of  $1 \times 10^{-4}$  is not sufficient to stabilize the system. The standard deviation of  $\theta$  is growing with time, although the mean is approaching zero. However, increasing the intensity to  $1 \times 10^{-2}$ , the standard deviation does approach a finite value as shown in Figure (6-14).

Thus, Figures (6-10) through (6-14) demonstrate that the stability robustness of the controlled system is enhanced by the addition of all three types of input noise, such that the response of the system is stable even in the face of parameter changes in the real world. White and first-order colored noise of the same maximum intensity produce very similar results. Robustness can also be improved using second-order colored noise, but a greater maximum noise intensity is necessary to achieve a comparable response to the previous two cases. Figure (6-15) shows a power spectral density plot of the second-order time-correlated noise for the case of  $Q_u = 2100$ . The magnitude of the time-correlated noise at low frequencies (at 0.1 Hz and below) was found to be  $1.3 \times 10^{-6}$ , after converting from decibels to a linear magnitude scale. Inputting a white noise of this approximate intensity is sufficient to recover the stability of the system as is demonstrated in Figure (6-16). This implies that the robustness improvement is not primarily due to the time-correlated noise concentrated around 70 rad/sec; rather the magnitude of the noise

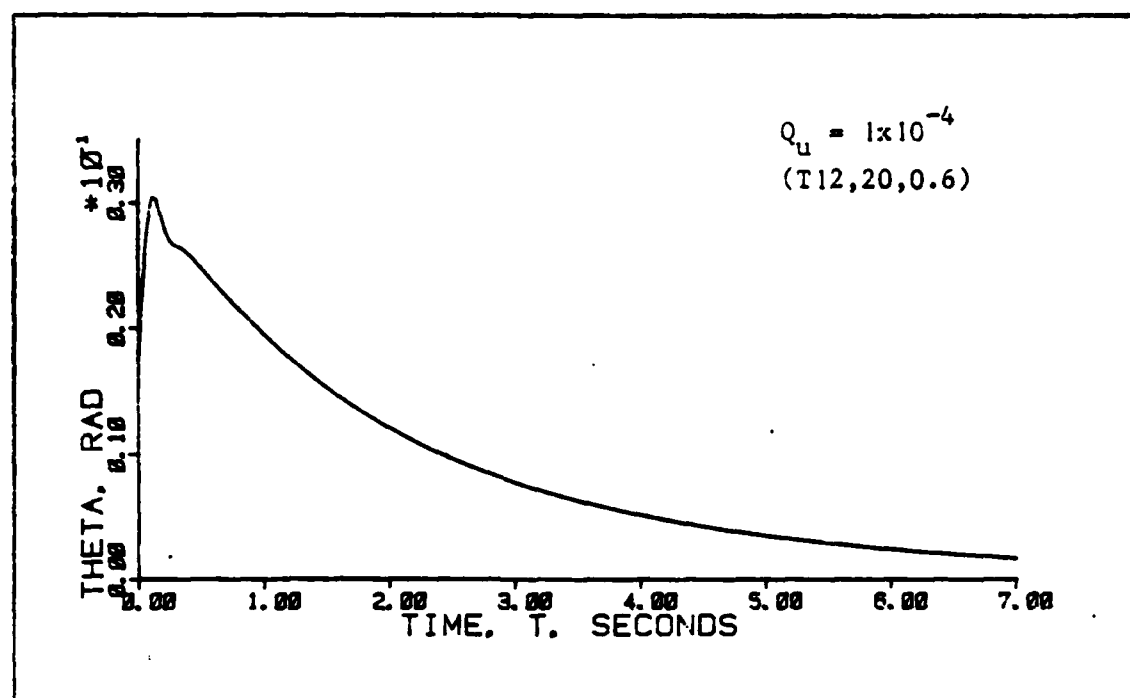


Figure 6-12a: Mean of  $\Theta$  With First-Order Colored Noise Addition at Off-Design Flight Condition

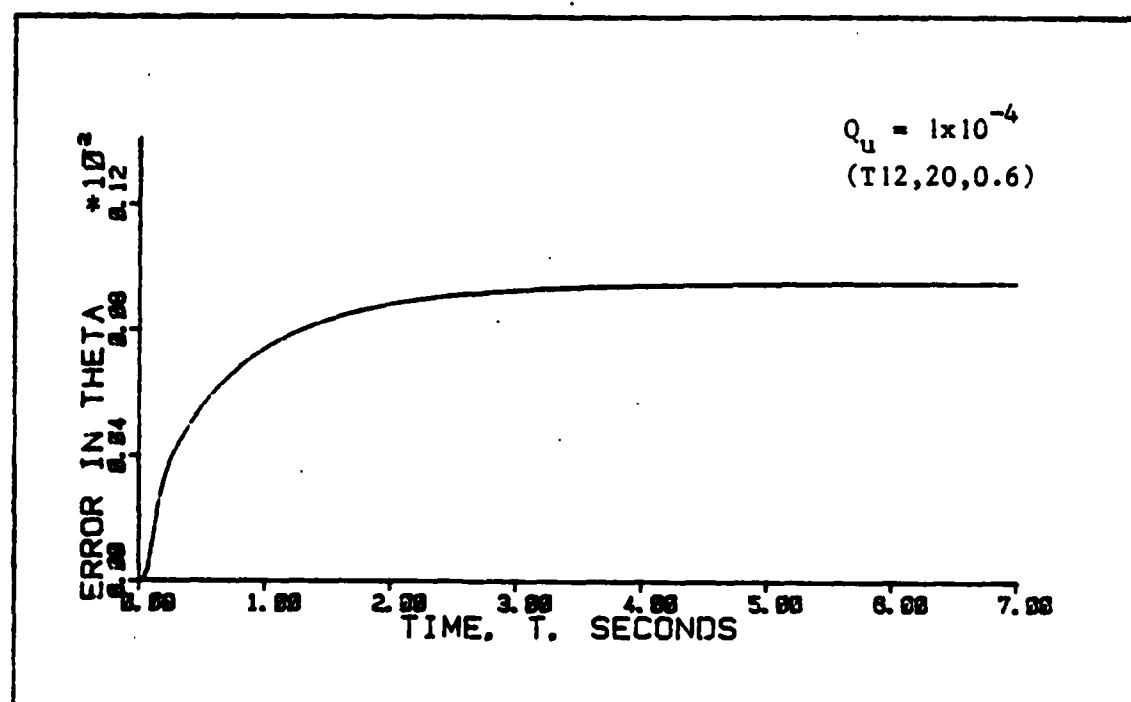


Figure 6-12b: Standard Deviation of  $\Theta$  With First-Order Colored Noise Addition at Off-Design Flight Condition

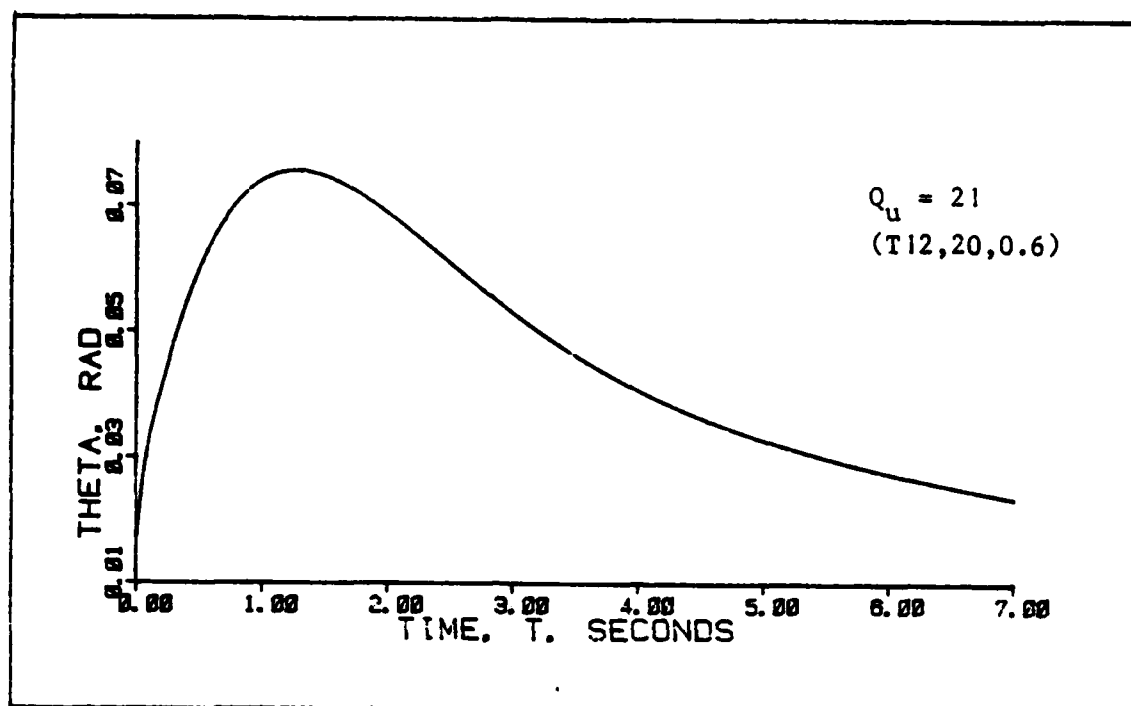


Figure 6-13a: Mean of  $\Theta$  With the Second-Order Colored Noise Addition at Off-Design Flight Condition

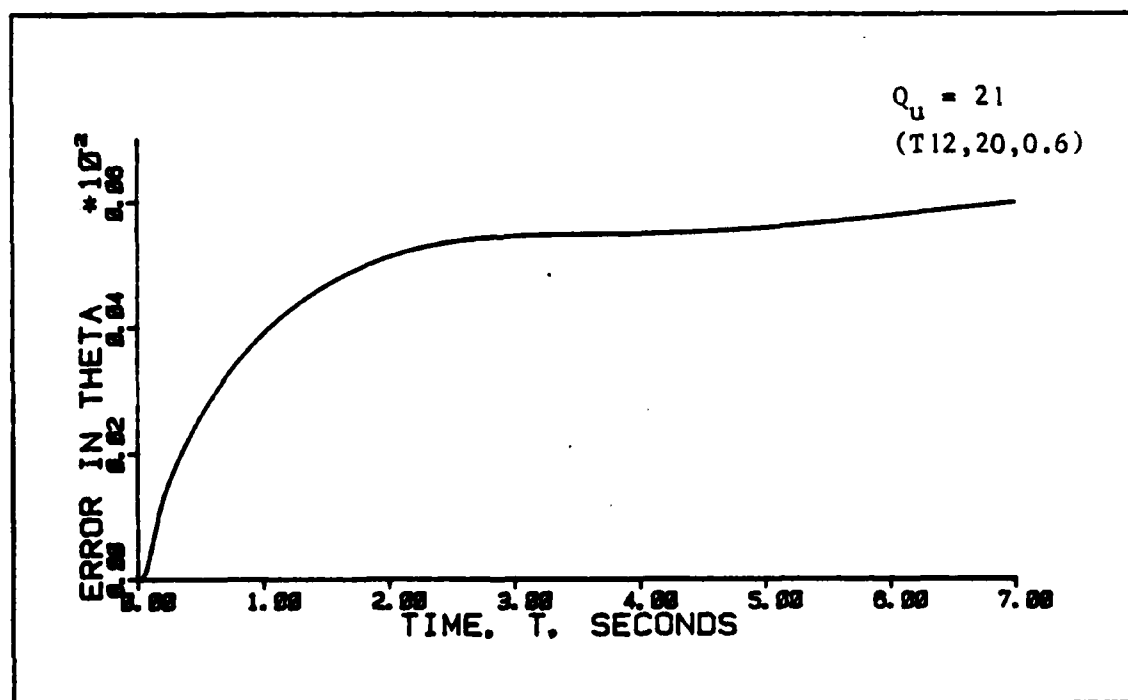


Figure 6-13b: Standard Deviation of  $\Theta$  With Second-Order Colored Noise Addition at Off-Design Flight Condition

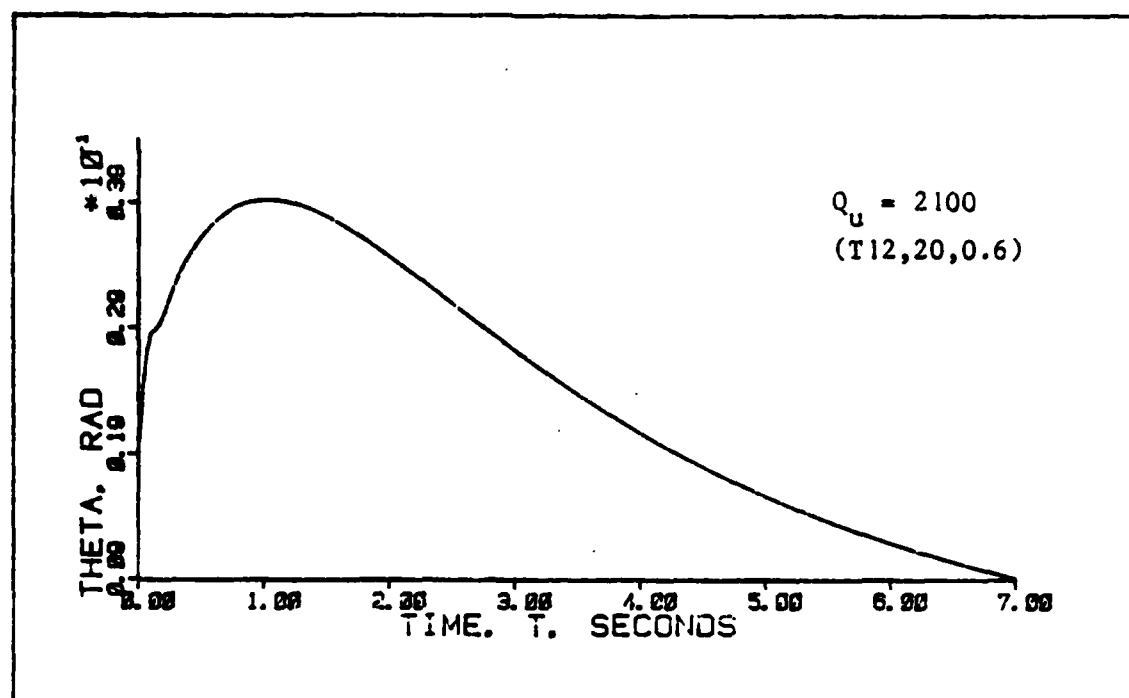


Figure 6-14a: Mean of  $\Theta$  With Second-Order Colored Noise Addition at Off-Design Flight Condition

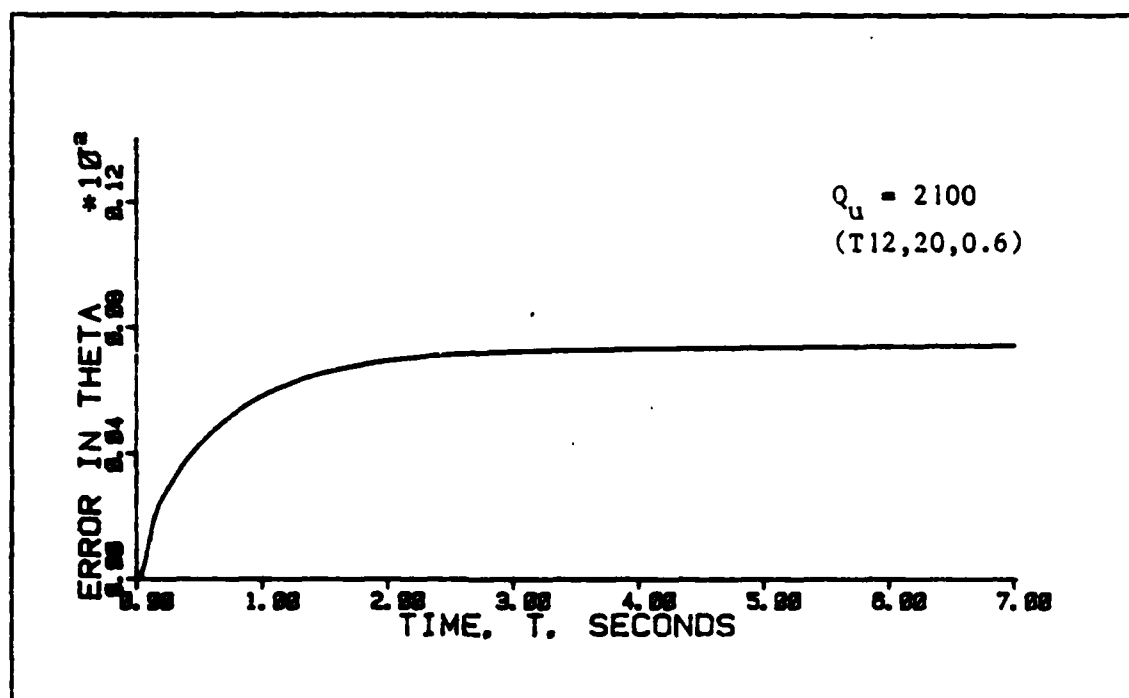


Figure 6-14b: Standard Deviation of  $\Theta$  With Second-Order Colored Noise Addition at Off-Design Flight Condition

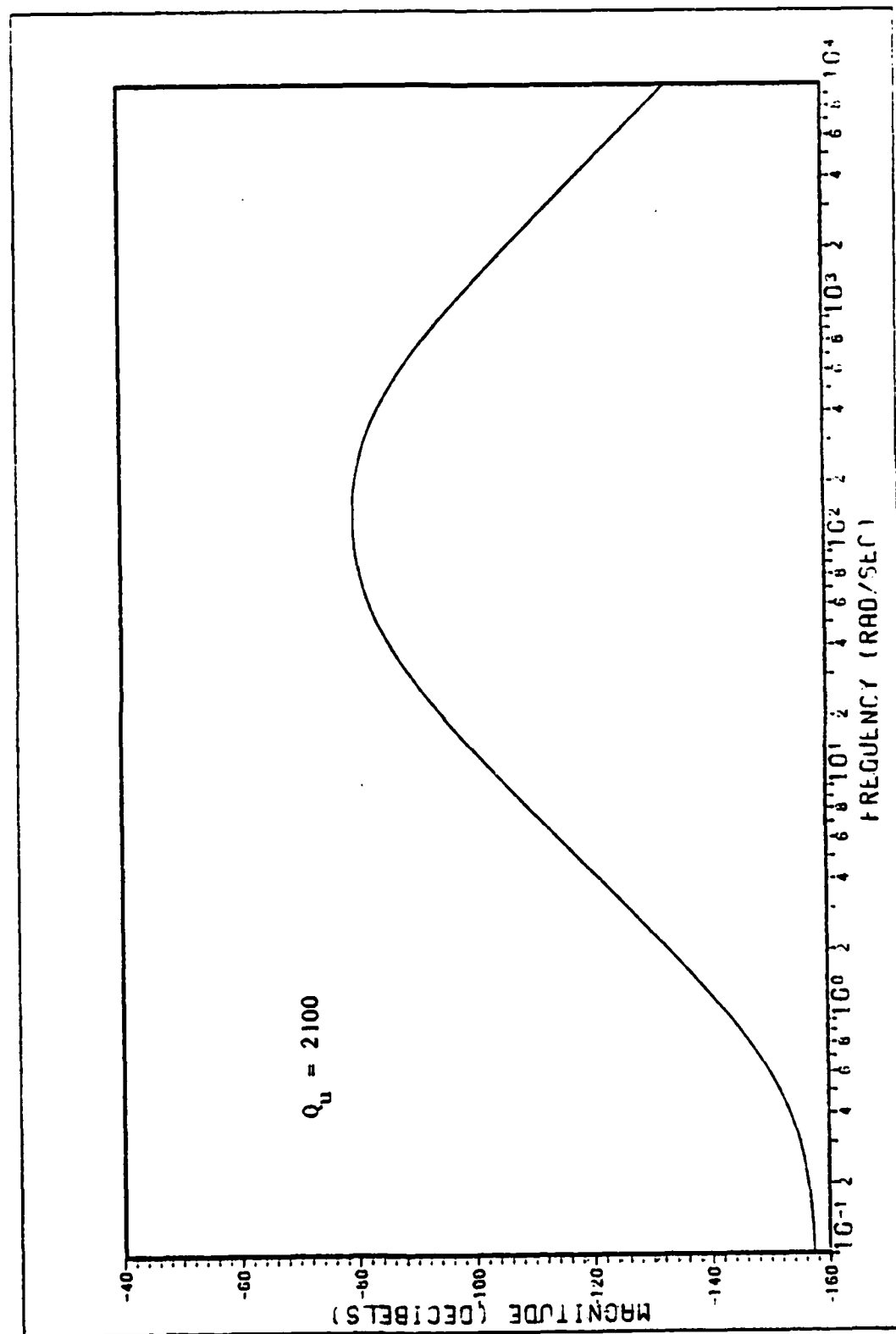


Figure 6-15: Power Spectral Density Function for a Second-Order Colored Noise Process



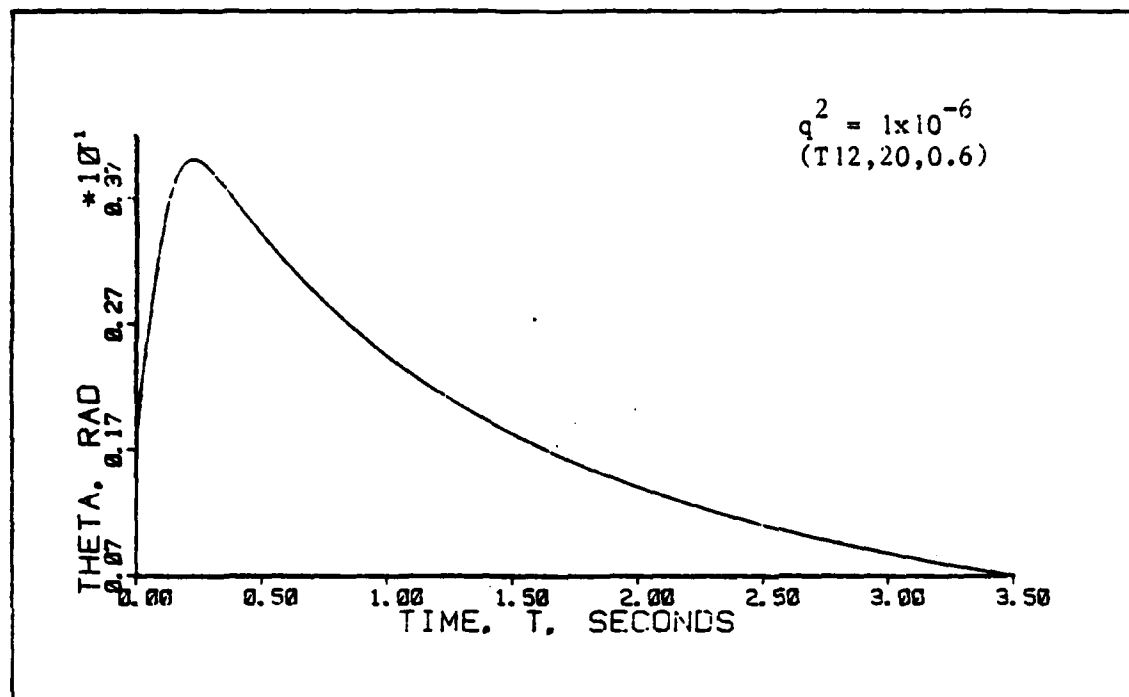


Figure 6-16a: Mean of  $\Theta$  With White Noise Addition at Off-Design Flight Condition

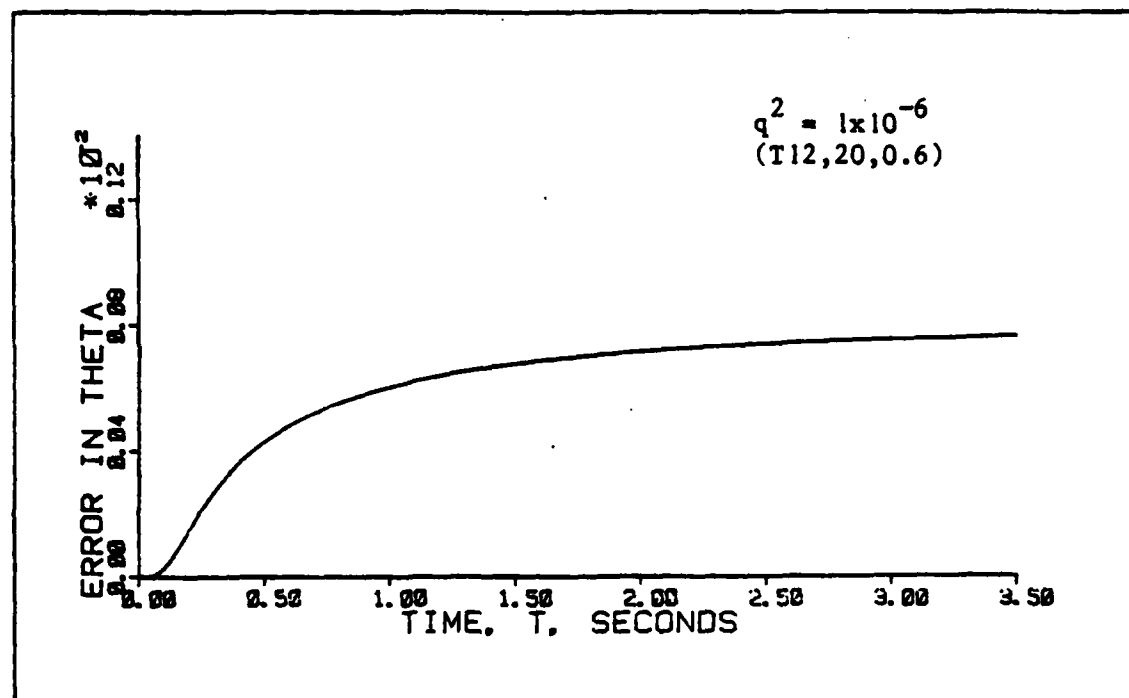


Figure 6-16b: Standard Deviation of  $\Theta$  With White Noise Addition at Off-Design Flight Condition

elsewhere is sufficiently high to achieve the desired robustification.

### 6.2.3 Discretized Continuous-Time LQG Regulators at Design Condition

This method of extending the robustification techniques to discrete-time systems was introduced in Section 2.7.1. A continuous-time controller identical to that of Sections 6.2.1 and 6.2.2 is designed, then discrete controller equations are formed by making first-order approximations to the discrete controller gain matrices. The sample rate of the discretized controller is 50 Hertz. The design model matrices are given in Section 6.4.

The time histories of the mean and standard deviation of  $\theta$  for the eight-state controller evaluated against a truth model of the same dimension are shown in Figure (6-17). Again, the state was given a perfectly known initial condition of one degree. The performance of the discretized controller is very similar to that for the continuous-time case; the state converges to zero fairly slowly. This similarity is as expected since the sample period is short compared to natural system transients. Figure (6-18) shows the response of the same controller evaluated against a twelve-state truth model with higher-order actuator models. As can be seen, changing the environment in which the controller is evaluated introduces a slightly larger initial overshoot and steady-state error in the mean of  $\theta$  response. Again, the unmodified system does not display good robustness properties.

Figure (6-19) shows the response of the same controller with the filter robustified by adding a white noise of strength  $q = 1 \times 10^{-6}$  to the design model. The improvement in transient time, initial overshoot, and

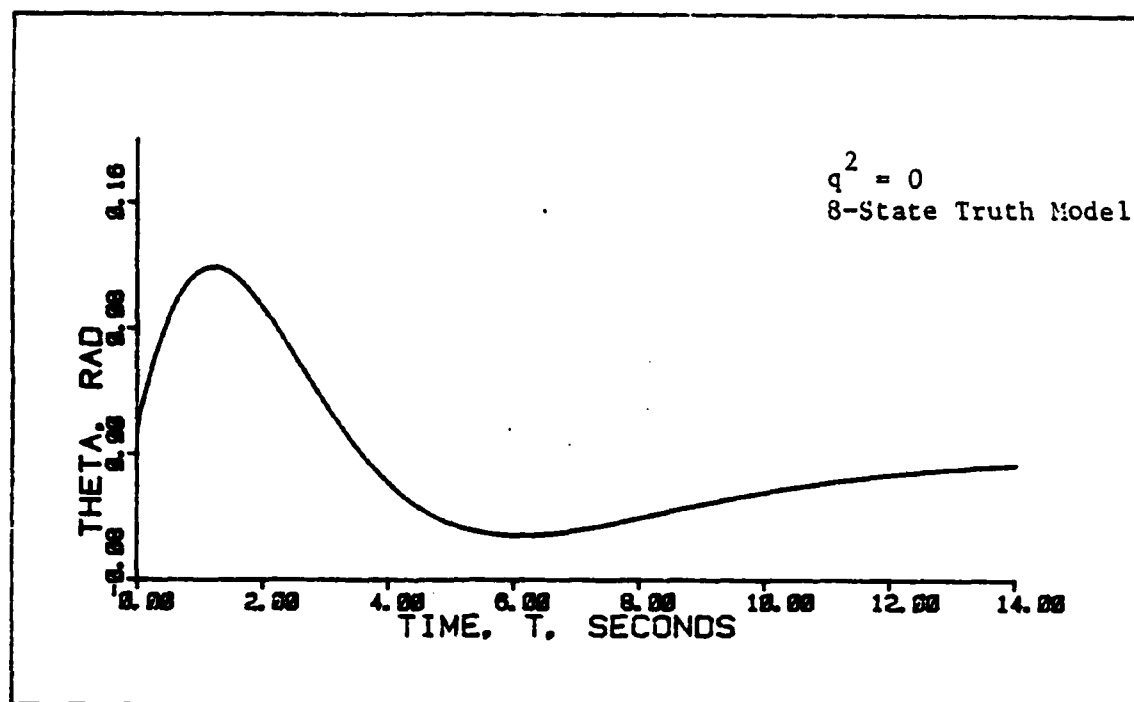


Figure 6-17a: Mean of  $\Theta$  at the Design Condition

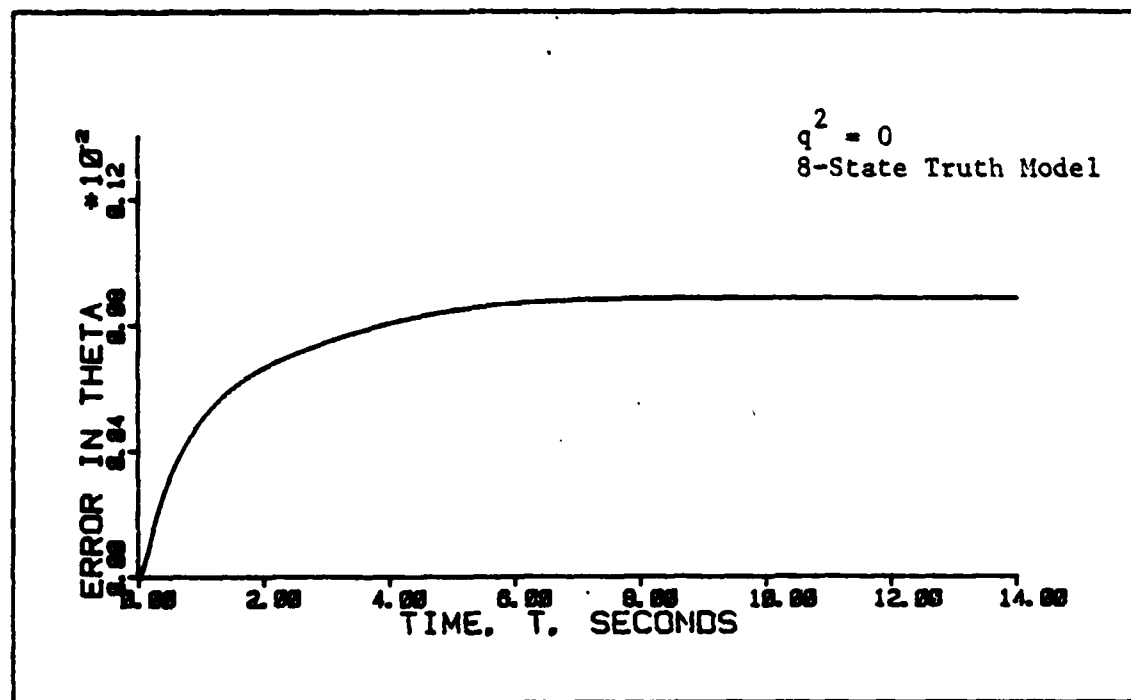


Figure 6-17b: Standard Deviation of  $\Theta$  at the Design Condition

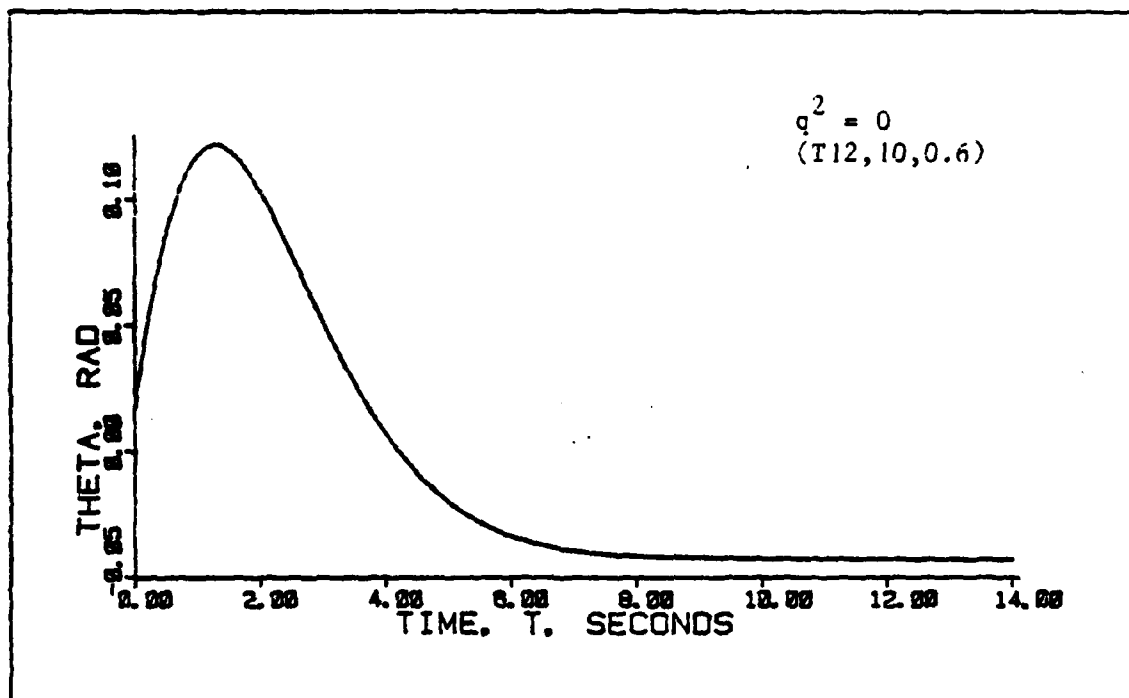


Figure 6-18a: Mean of  $\Theta$  With No Noise Addition

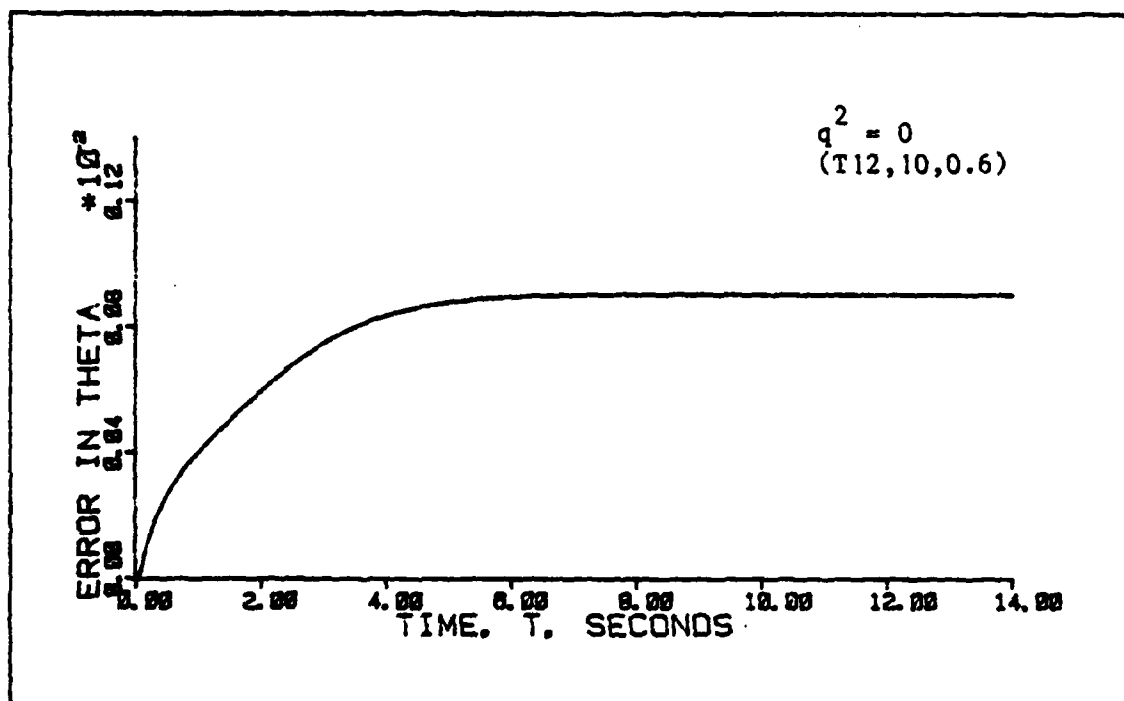


Figure 6-18b: Standard Deviation of  $\Theta$  With No Noise Addition

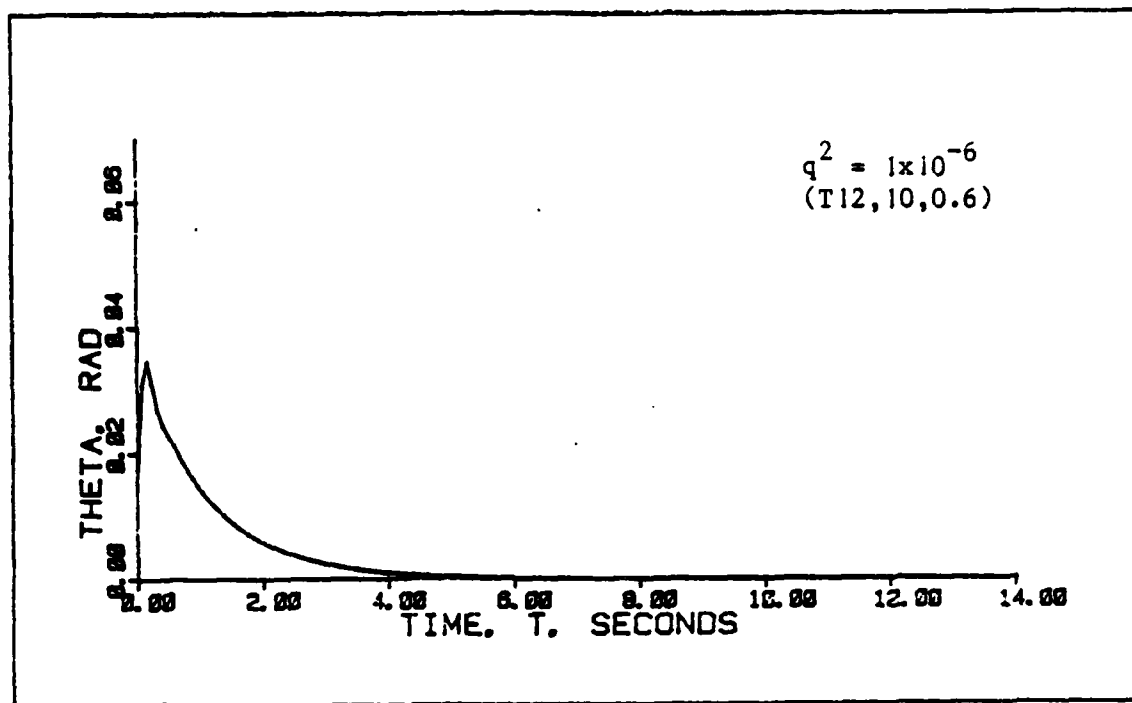


Figure 6-19a: Mean of  $\Theta$  With White Noise Addition

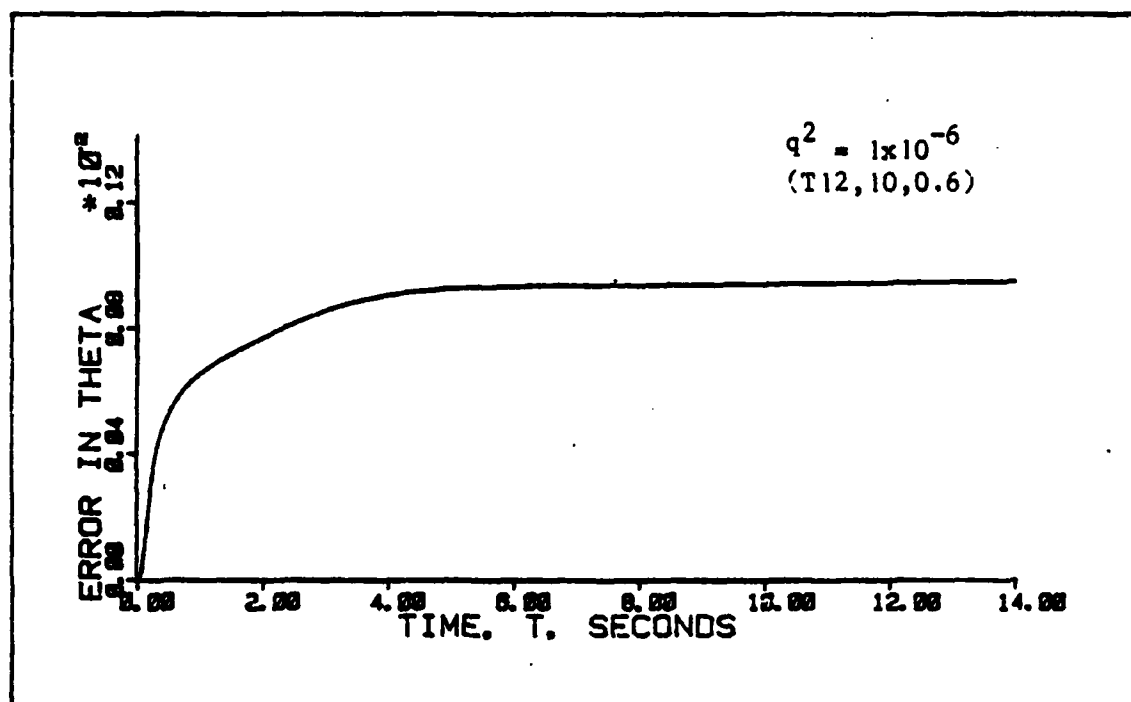


Figure 6-19b: Standard Deviation of  $\Theta$  With White Noise Addition

steady-state value is comparable to the continuous-time case with no noticeable change in the standard deviation.

The results of applying colored noise generated by a first-order shaping filter are shown in Figure (6-20). The overshoot is larger than for the white noise addition, and there is still a slight steady-state mean error, but the improvement over the unrobustified case is still substantial.

It was found that augmenting second-order shaping filter states with the design model, designing a continuous-time controller, than discretizing the controller produced an unstable closed-loop system for any value of  $Q_u$ . This characteristic was not observed in the continuous-time case.

In addition, as described previously in Reference 21, it was observed that the strength of the white and colored noise could not be adjusted arbitrarily upwards. It was found that, with the twelve-state truth model (T12,10,0.6),  $q^2$  and  $Q_u$  could be adjusted between zero and  $1 \times 10^{-6}$ , and the robustification improvement was similar to the continuous-time case. However, noise intensities beyond this value ( $Q_u = 2 \times 10^{-6}$ ) would actually drive closed-loop system eigenvalues outside of the z-domain unit circle.

A comparison of the steady-state values of the standard deviations of the aircraft states are given in Table (6-4). The table includes values for a system with no noise addition, white and first-order colored noise of the same maximum intensity. The results are similar to the continuous-time case. It is seen that the addition of white noise increases the values of the standard deviations of the states. As in the previous case, colored noise addition generated by a first-order shaping filter did not yield lower standard deviations than white noise

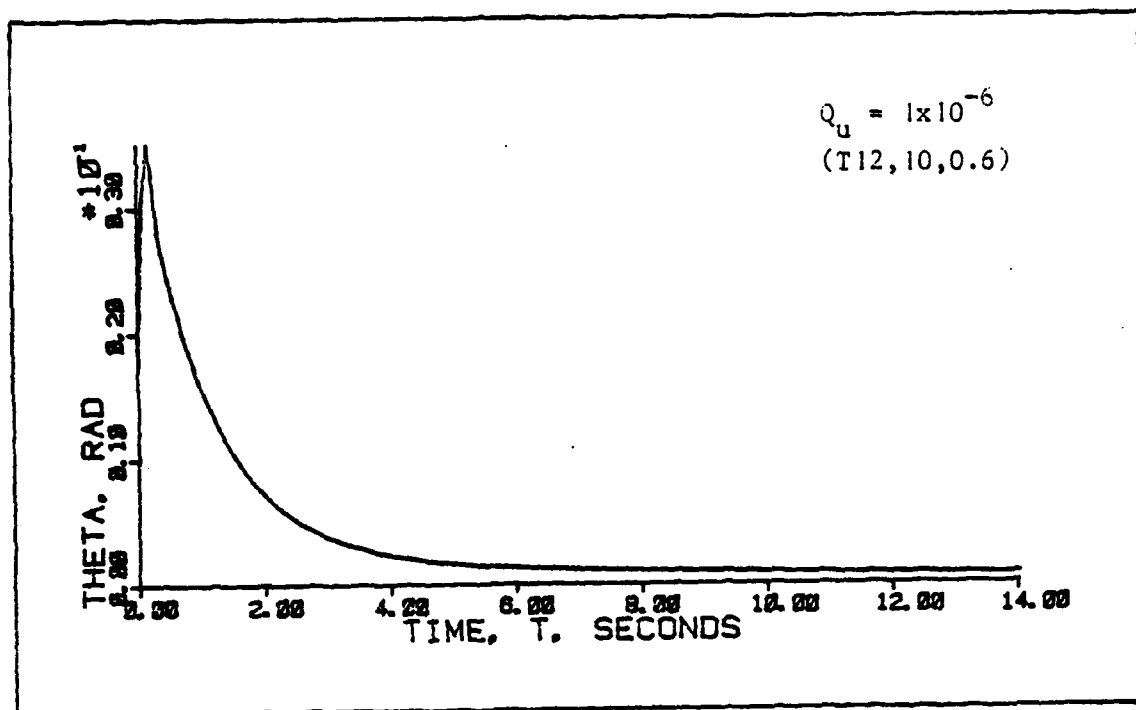


Figure 6-20a: Mean of  $\Theta$  With First-Order Colored Noise Addition

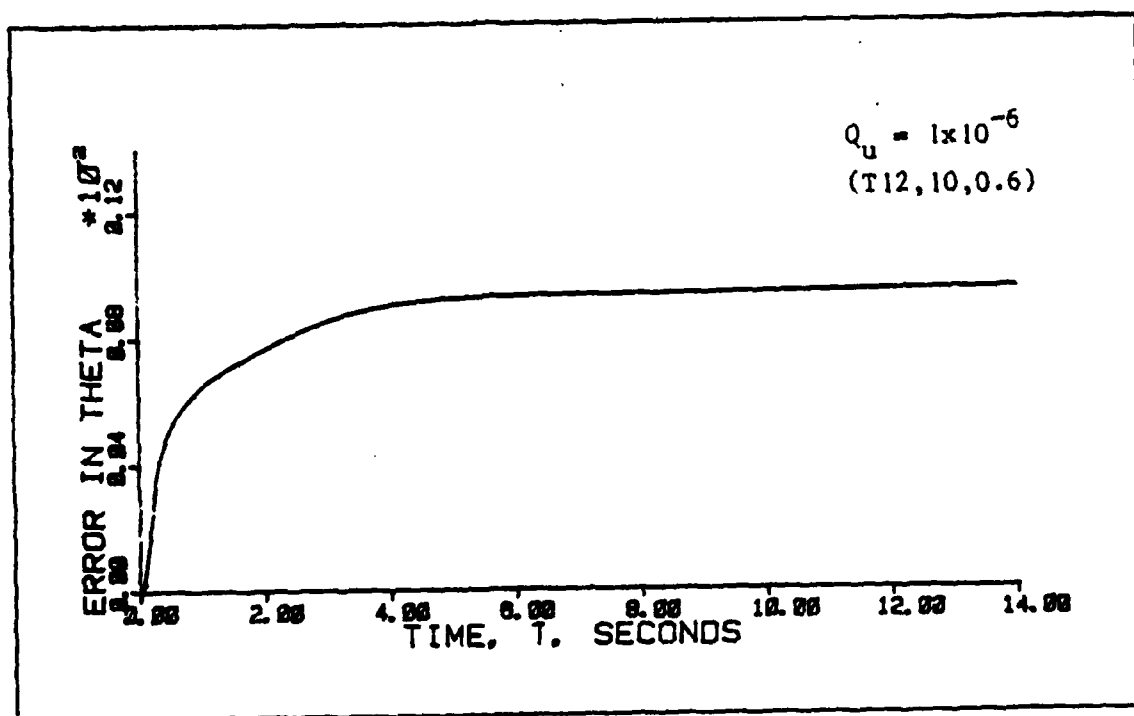


Figure 6-20b: Standard Deviation of  $\Theta$  With First-Order Colored Noise Addition

Table 6-4

Comparison of Steady-State Standard Deviations  
of Aircraft States at the Design Condition for a  
Discretized Continuous-Time System

	$q^2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No Noise	0	-	$8.877 \times 10^{-4}$	$5.035 \times 10^{-3}$	$1.258 \times 10^{-3}$
White Noise	$1 \times 10^{-4}$	-	$9.645 \times 10^{-4}$	$5.310 \times 10^{-3}$	$2.113 \times 10^{-3}$
1st Order Shaping Filter	-	$1 \times 10^{-4}$	$1.005 \times 10^{-3}$	$5.821 \times 10^{-3}$	$1.817 \times 10^{-4}$
2nd Order Shaping Filter	-	-	Unstable	Unstable	Unstable

Table 6-5

Comparison of Steady-State Standard Deviations  
of Aircraft States With Higher-Order Actuators  
For a Discretized Continuous-Time System

	$q^2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No Noise	0	-	$9.053 \times 10^{-4}$	$4.899 \times 10^{-3}$	$1.257 \times 10^{-3}$
White Noise	$1 \times 10^{-6}$	-	$9.511 \times 10^{-4}$	$5.198 \times 10^{-4}$	$2.825 \times 10^{-4}$
1st Order Shaping Filter	-	$1 \times 10^{-6}$	$9.524 \times 10^{-4}$	$5.212 \times 10^{-3}$	$2.758 \times 10^{-4}$
2nd Order Shaping Filter	-	-	Unstable	Unstable	Unstable



addition except for the pitch rate,  $q$ . Table (6-5) indicates identical trends for the case where higher-order dynamics are included in the truth model. The lower maximum noise intensity than that used for Table (6-4) reflects the fact that, with a higher-dimension truth model, the closed-loop system is driven unstable for a lower value of  $q^2$ .

Thus, it is seen that the desired robustness enhancement can be gained by adding white input noise to the system model. Time-correlated noise is not appropriate for this problem because it does not yield performance benefits and only serves to increase the complexity of the design model.

#### 6.2.4 Discretized Continuous-Time LQG Regulators at Off-Design Condition

Figure (6-21) demonstrates that the system exhibits an unstable response when the flight condition is changed to an altitude of 20000 feet and a Mach number of 0.6 (T12,20,0.6).

Again, it was found that the robustness characteristics could be improved with the injection of white and colored noise for a finite range of  $q^2$  and  $Q_u$ . At the off-design condition, for values of  $q^2$  and  $Q_u$  beyond  $2.5 \times 10^{-5}$ , the closed-loop system was unstable.

The response of the system with the maximum value of  $q^2$  is shown in Figures (6-22a) and (6-22b). As can be seen, the stability has been recovered, indicating that the full-state feedback system was stable. Thus, by the addition of white input noise, the characteristics approach that of a full-state feedback controller. Figures (6-22c) and (6-22d) show that stability is recovered with a lower value of  $q^2$ . Increasing it beyond this value changes only the transients of the mean and the magnitude of the standard deviation.

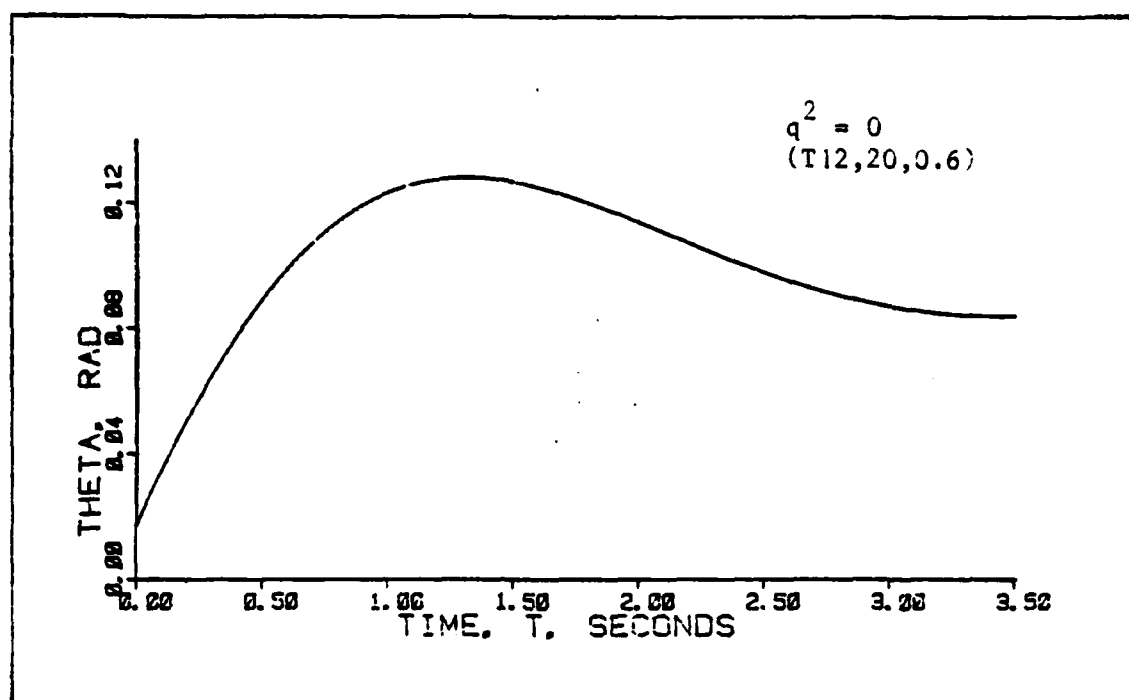


Figure 6-21a: Mean of  $\Theta$  With No Noise Addition at Off-Design Flight Condition

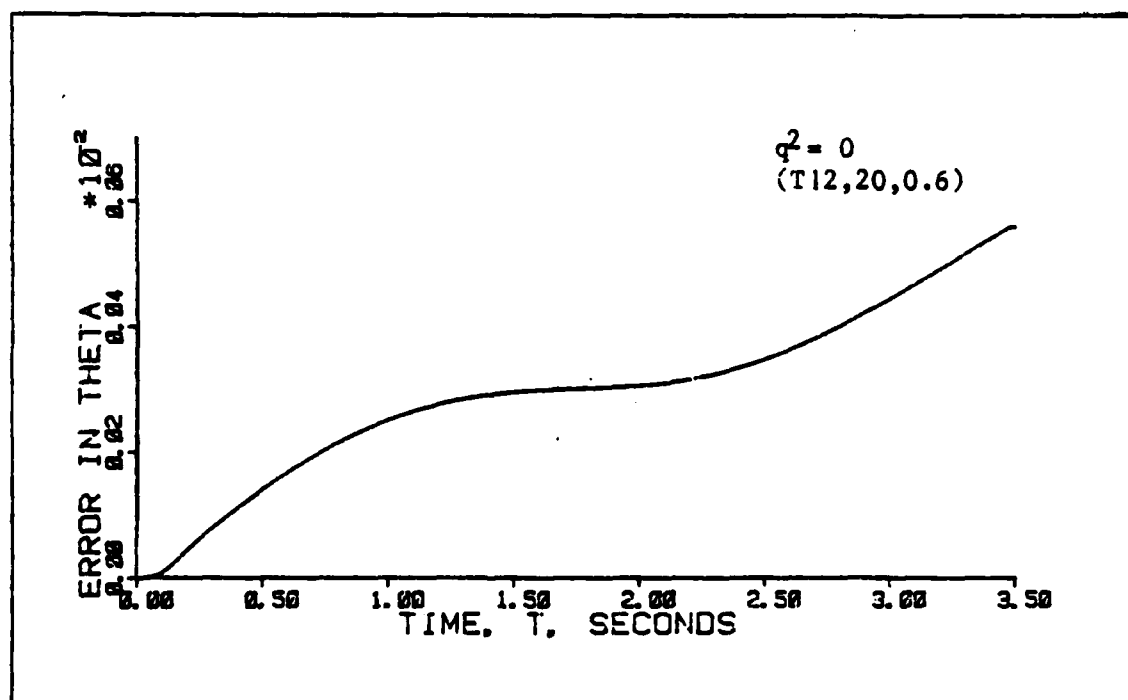


Figure 6-21b: Standard Deviation of  $\Theta$  With No Noise Addition at the Off-Design Flight Condition

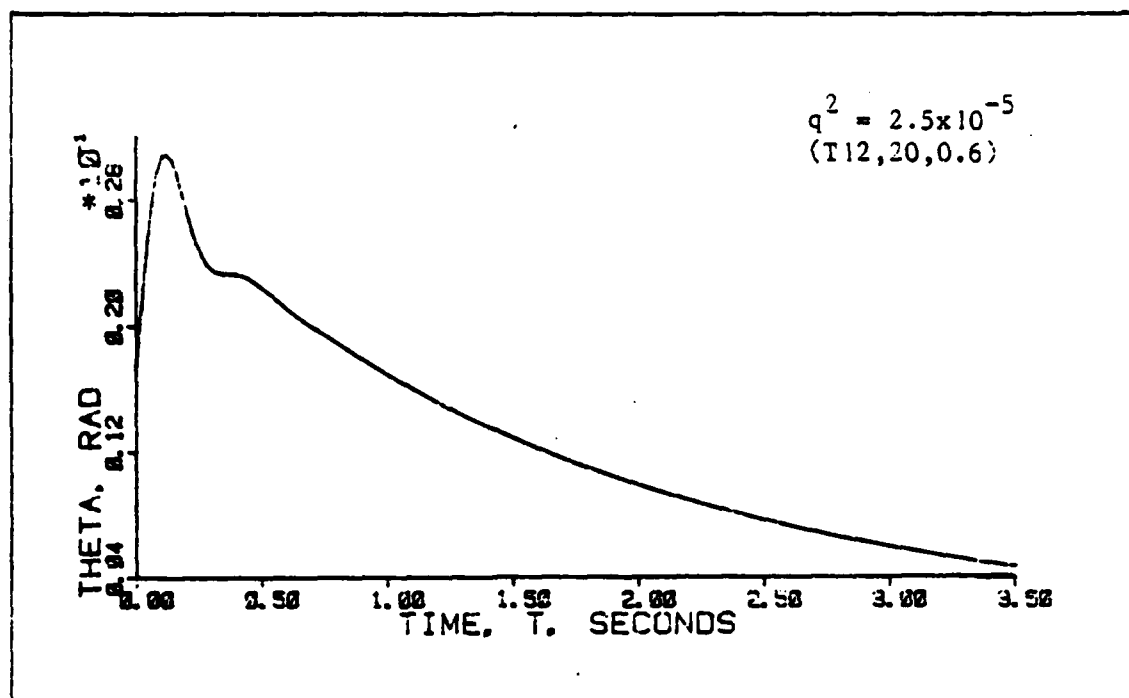


Figure 6-22a: Mean of  $\Theta$  With White Noise Addition at Off-Design Flight Condition

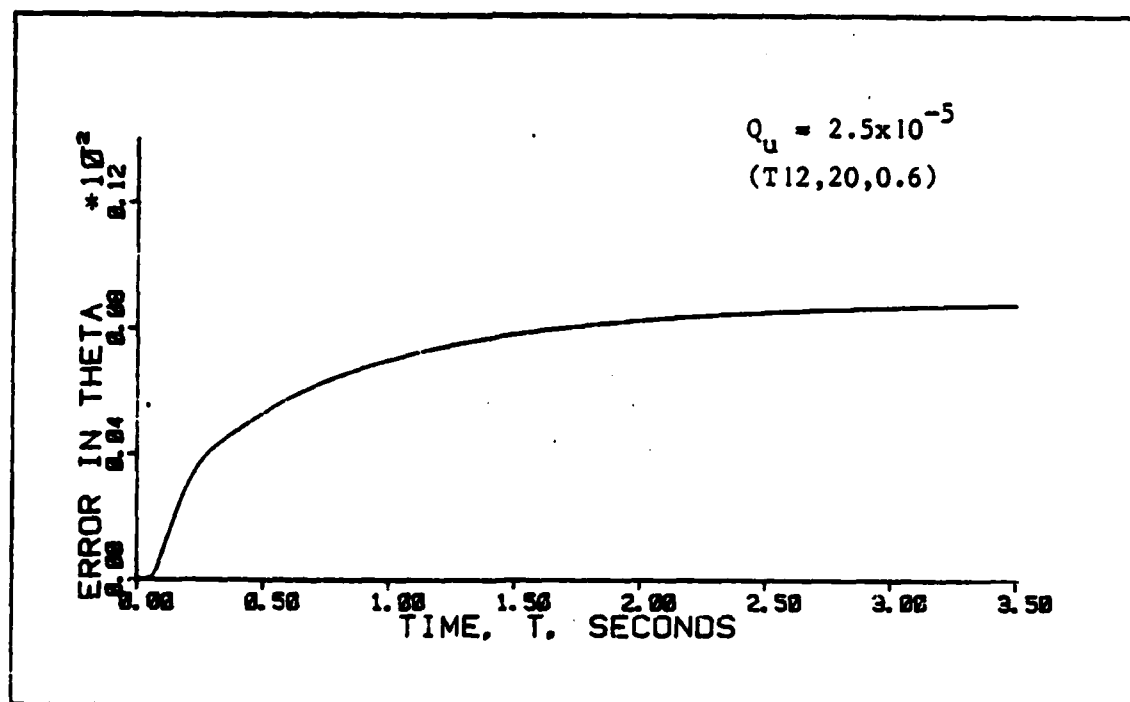


Figure 6-22b: Standard Deviation of  $\Theta$  With White Noise Addition at the Off-Design Flight Condition

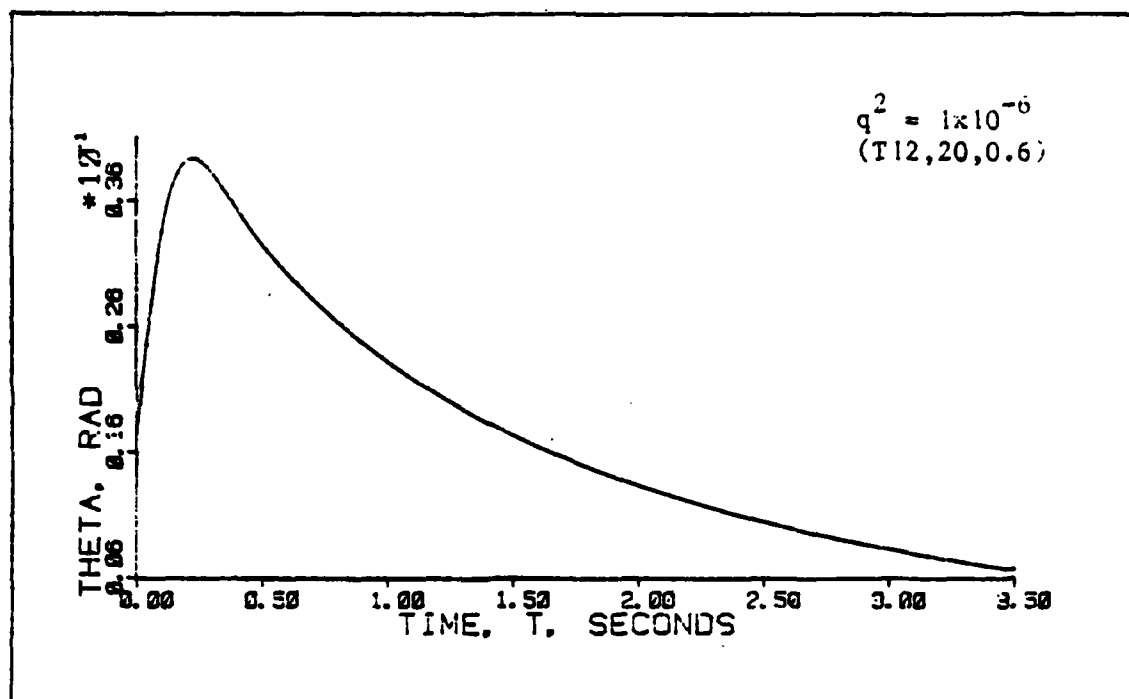


Figure 6-22c: Mean of  $\Theta$  With White Noise Addition at an Off-Design Flight Condition

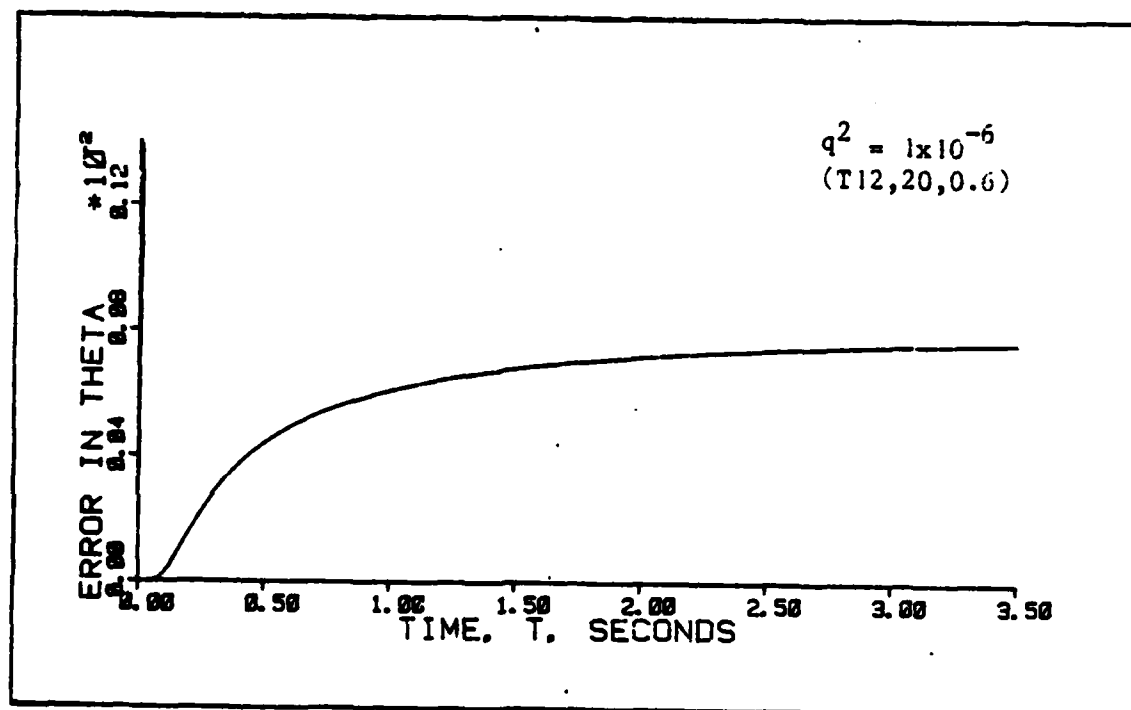


Figure 6-22d: Standard Deviation of  $\Theta$  With White Noise Addition at an Off-Design Flight Condition

Figure (6-23) demonstrates that adding colored noise generated by a first-order shaping filter achieves the same degree of robustness enhancement with only an increase in the initial overshoot of  $\theta$  compared to that of Figure (6-22a). The steady-state standard deviations of Figures (6-22b) and (6-23b) are not significantly different.

Augmenting second-order shaping filter states with the system model again resulted in an unstable closed-loop system. For any non-zero value of  $Q_u$ , some of the closed-loop system eigenvalues were driven outside the unit circle in the  $z$ -domain.

Because of the sensitivity of the system to slight changes in the strength of the added noise, it is felt that this method is inappropriate for this problem. The range of admissible values for  $q^2$  and  $Q_u$  is very slight, and the loss of closed-loop system stability is very abrupt when the range is exceeded. In addition, there is no range of  $Q_u$  which improves the robustness characteristics of the controlled system when colored-noise generated by a second-order filter is added to the model.

#### 6.2.5 Sampled-Data LQG Regulators at Design Condition

This method of extending the robustification techniques to sampled-data systems was introduced in Section 2.7.2. For this case, the continuous-time system equations are discretized, then a sampled-data Kalman filter and controller are designed from the onset.

The results of performance analyses for the sampled-data controller with a sample rate of 50 Hertz are found to be extremely similar to those for the continuous-time controller.

Figure (6-24) shows the response of  $\theta$  to an initial condition of one degree. This is the result of a performance analysis where the dimensions

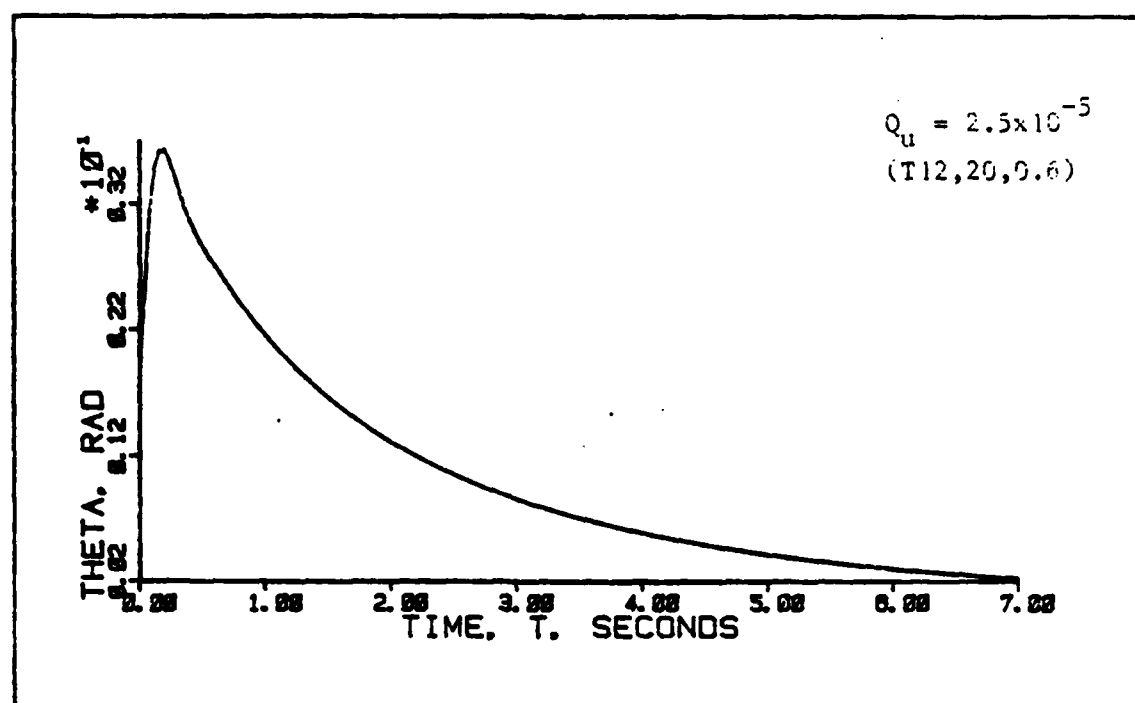


Figure 6-23a: Mean of  $\Theta$  With First-Order Colored Noise Addition at Off-Design Flight Condition

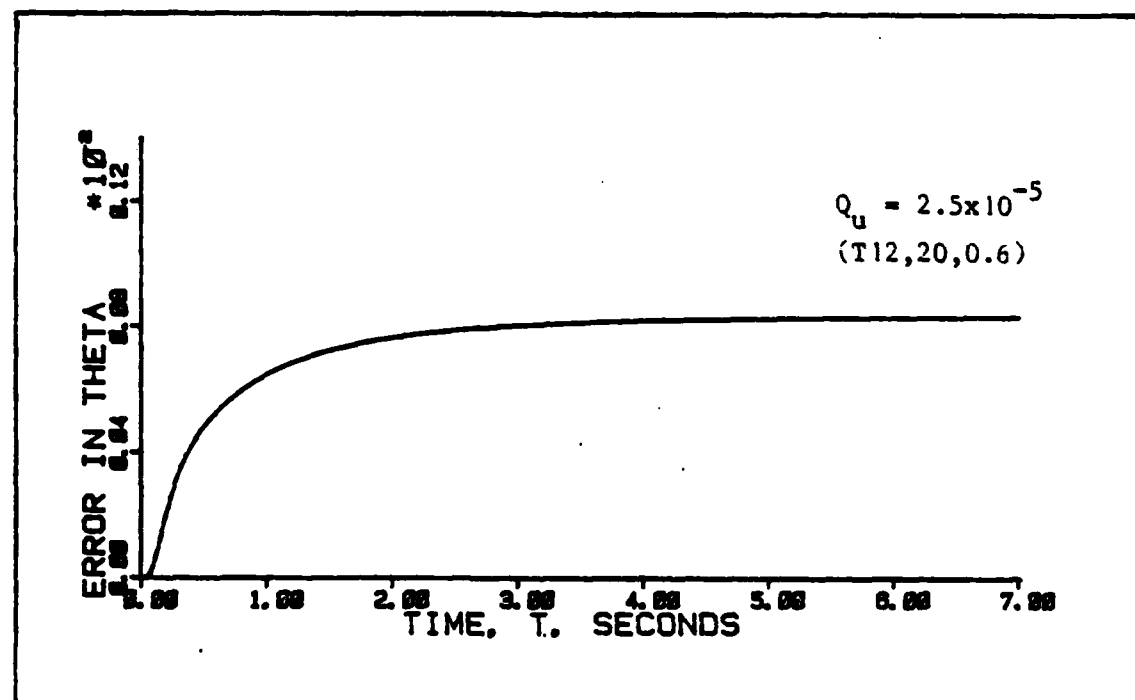


Figure 6-23b: Standard Deviation of  $\Theta$  With First-Order Colored Noise Addition at an Off-Design Flight Condition

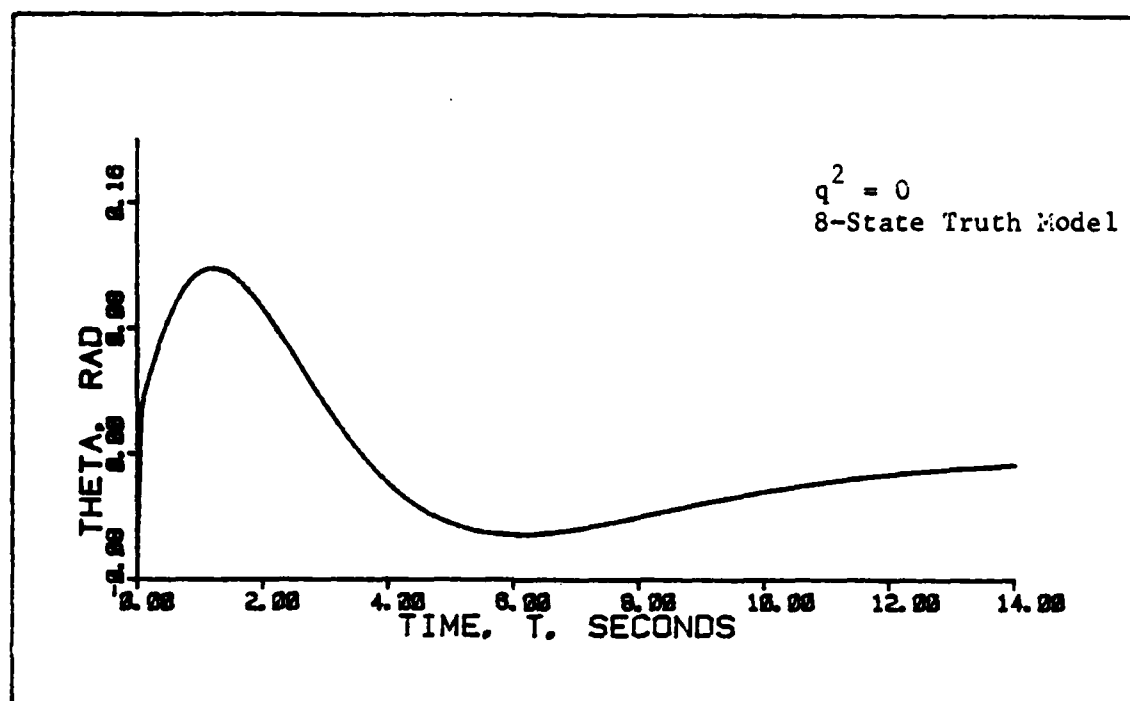


Figure 6-24a: Mean of  $\Theta$  at the Design Condition

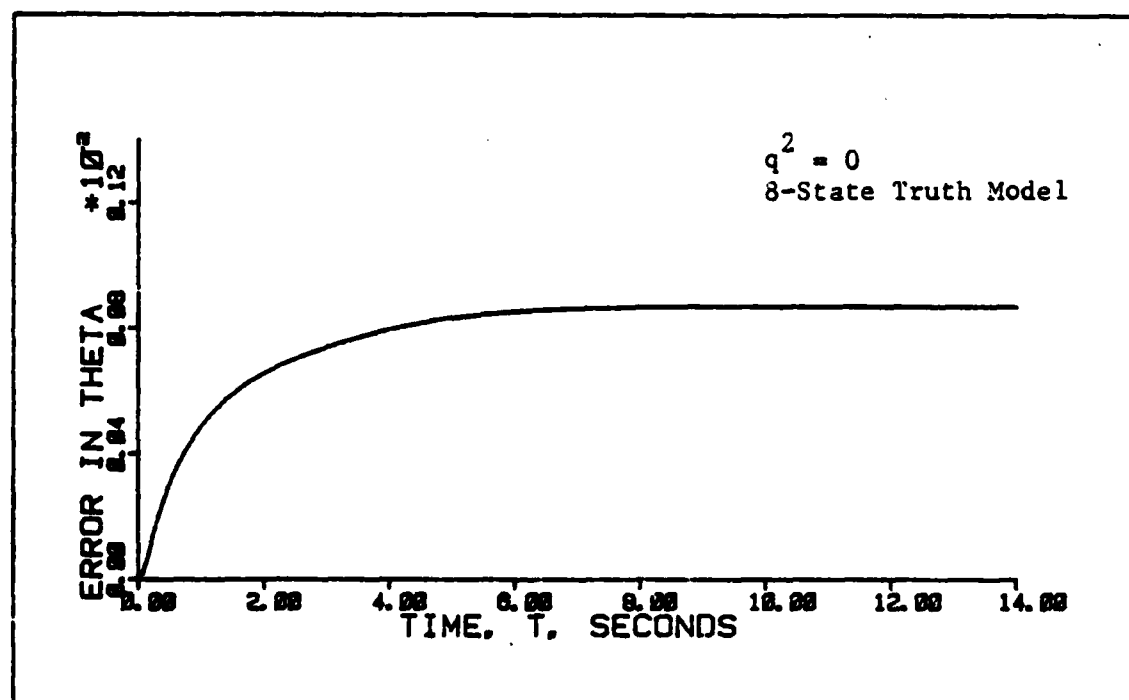


Figure 6-24b: Standard Deviation of  $\Theta$  at the Design Condition

of the truth model and the design model are the same (Section 5.4). As observed previously, the mean of  $\theta$  converges slowly to zero. Figure (6-25) shows the response of the same controller when higher-order actuator dynamics are introduced in the truth model. Again, large initial overshoots and steady-state mean errors are introduced. The unmodified sampled-data controller does not exhibit good robustness properties when certain states are ignored in the design model.

With the addition of white input noise, (as described in Section 2.7.2), the undesirable characteristics of the unrobustified system do not appear in the time response. This is shown in Figures (6-26a) and (6-26b). Figures (6-26c,d,e,f) show the trend for a higher and lower value of  $q^2$ . Increasing the magnitude of the white noise changes the transient mean response and the magnitude of the standard deviation. For this problem (where the evaluation was performed using the truth model (T12,10,0.6), in the range of values of  $q^2$  that were examined ( $0 \leq q^2 \leq 1$ ), the dramatic instabilities observed in Sections 6.2.3 and 6.2.4 do not occur.

Figure (6-27) demonstrates that nearly identical robustness benefits can be gained with the addition of colored noise generated by a first-order shaping filter as with a white noise. However, the addition of colored noise generated by a second-order shaping filter does not have the fast time response of the previous two cases, as shown in Figure (6-28). Nonetheless, the steady-state error and initial overshoot have been substantially improved.

Table (6-6) makes a comparison of the steady-state values of all three aircraft states at the design condition (truth and model dimension equal) for cases with no noise addition, white noise, and first- and



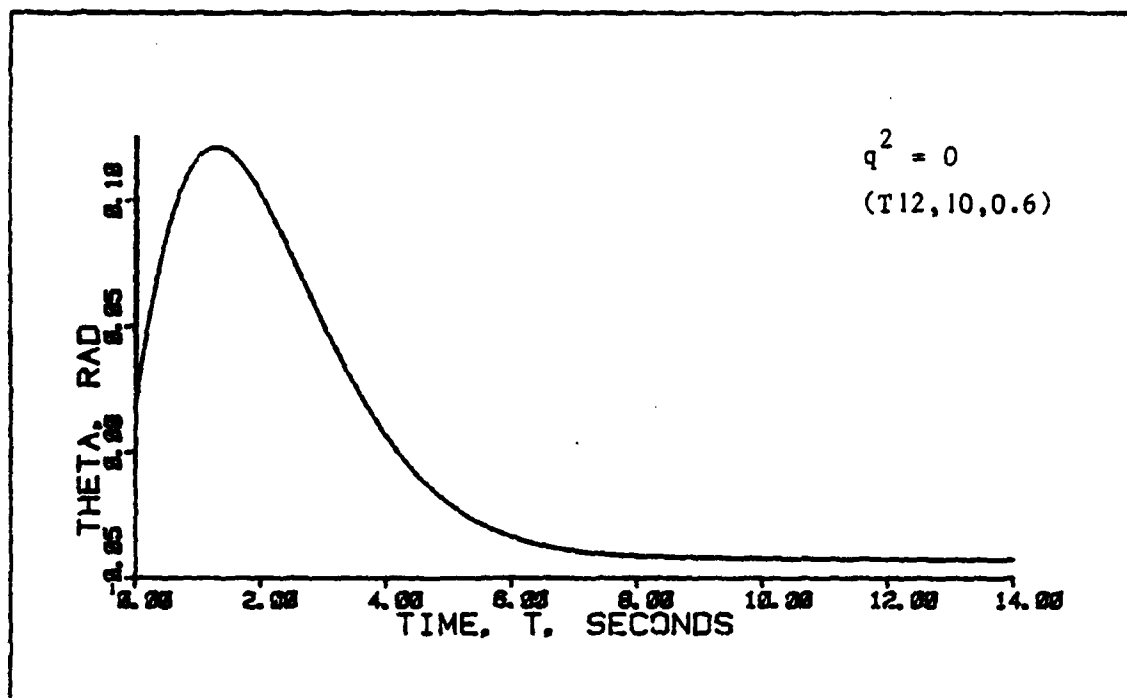


Figure 6-25a: Mean of  $\Theta$  With No Noise Addition

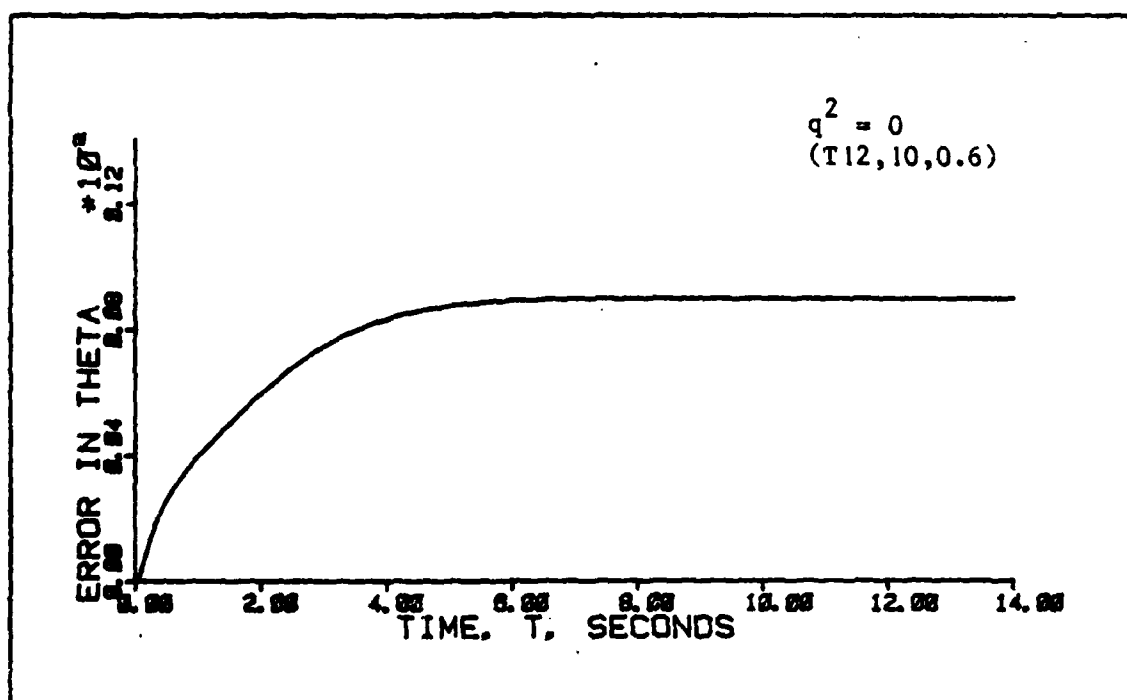


Figure 6-25b: Standard Deviation of  $\Theta$  With No Noise Addition

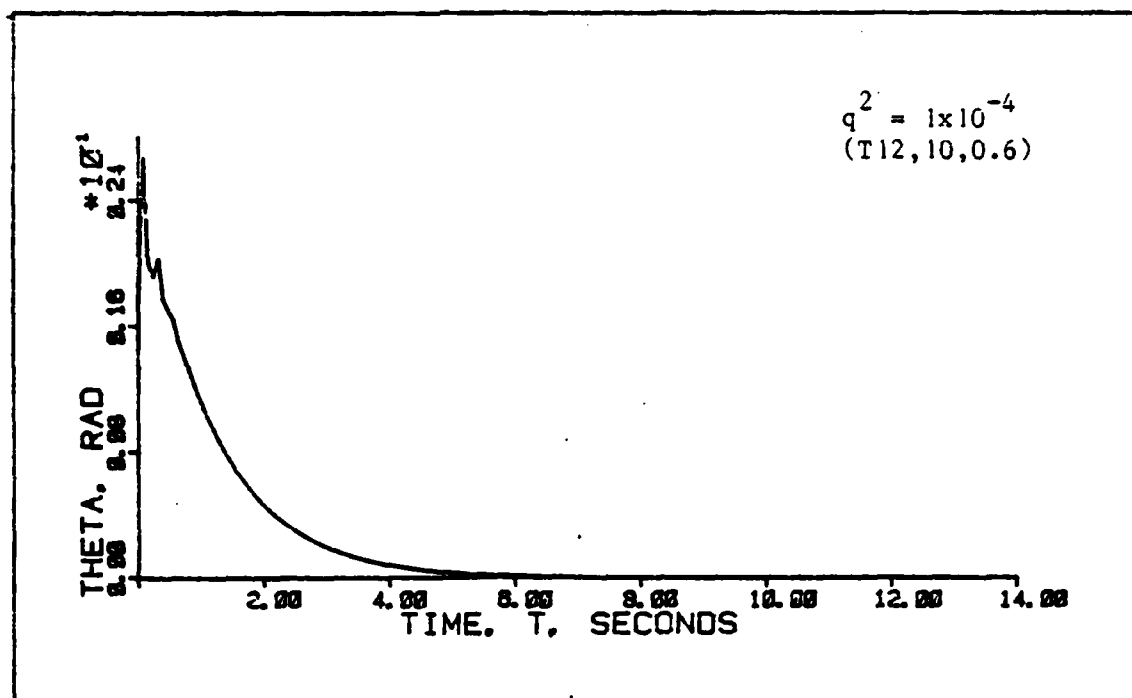


Figure 6-26a: Mean of  $\Theta$  With White Noise Addition

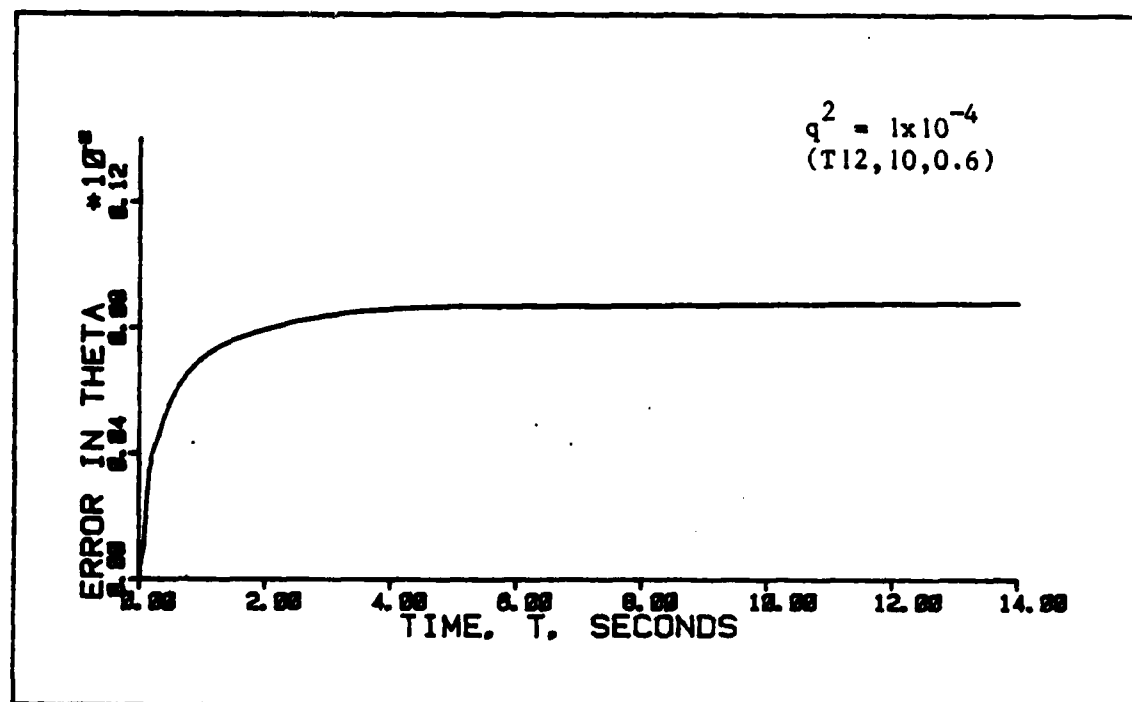


Figure 6-26b: Standard Deviation of  $\Theta$  With White Noise Addition

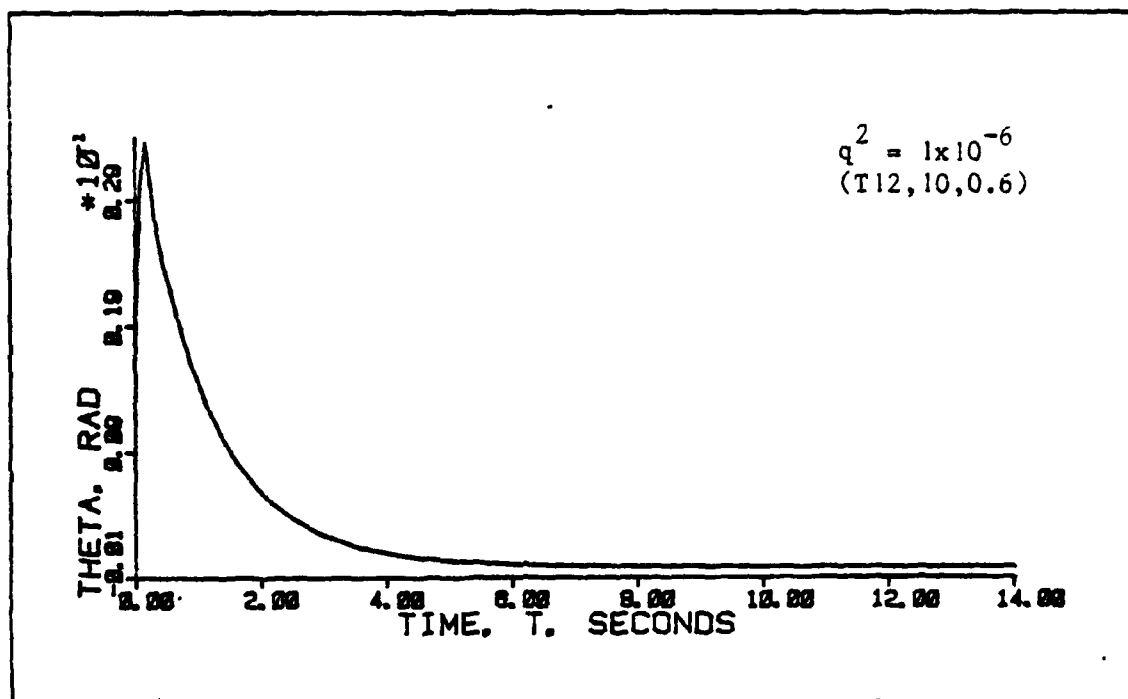


Figure 6-26c: Mean of  $\Theta$  With White Noise Addition

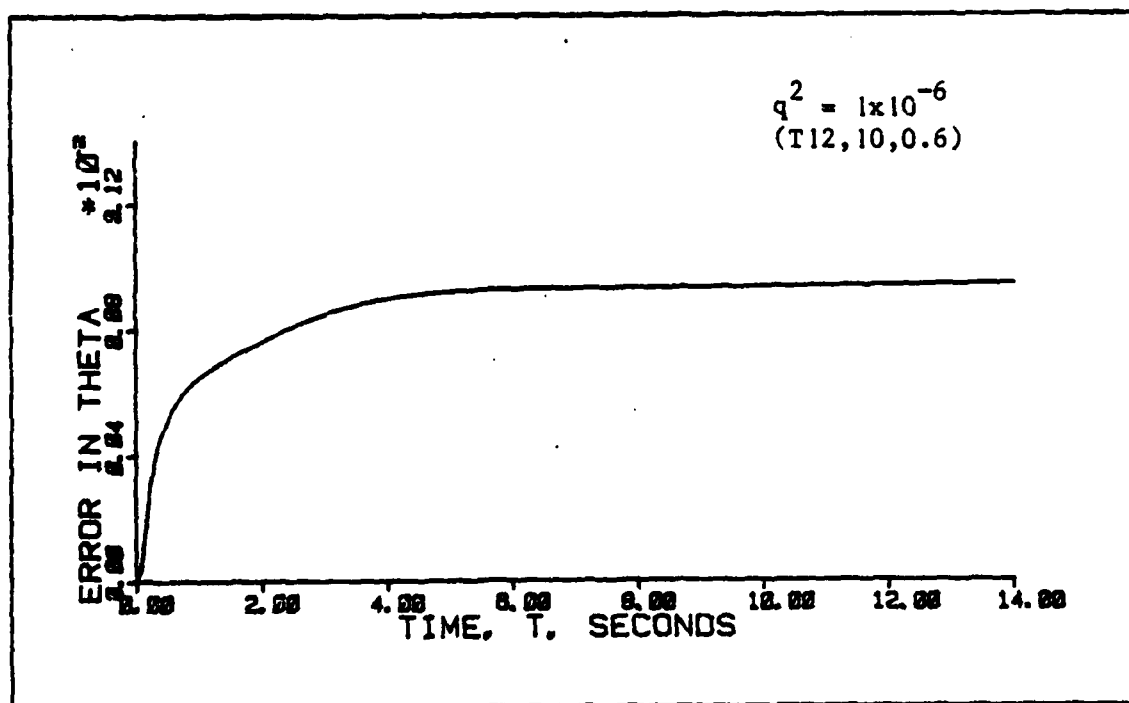


Figure 6-26d: Standard Deviation of  $\Theta$  With White Noise Addition

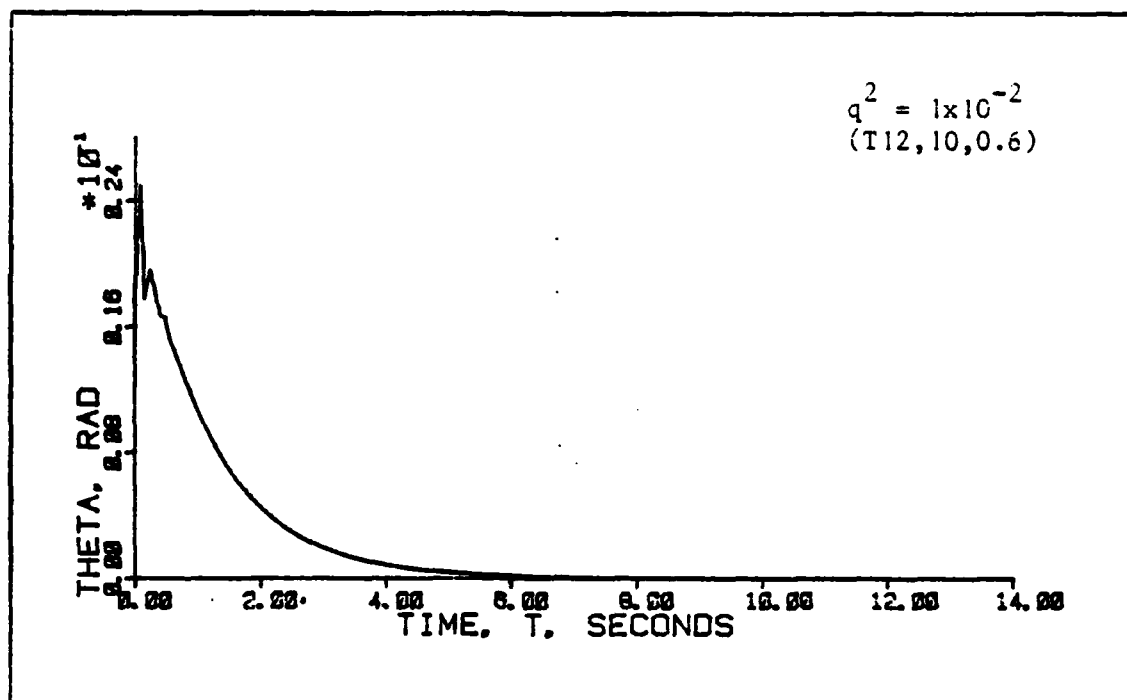


Figure 6-26e: Mean of  $\Theta$  With White Noise Addition

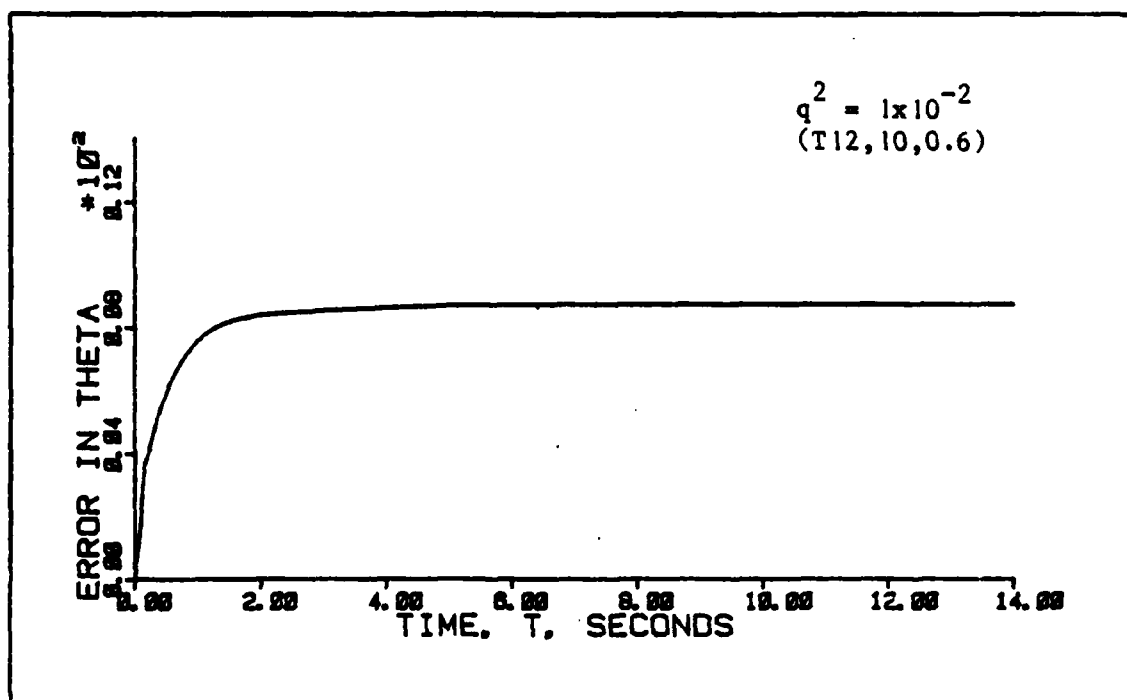


Figure 6-26f: Standard Deviation of  $\Theta$  With White Noise Addition

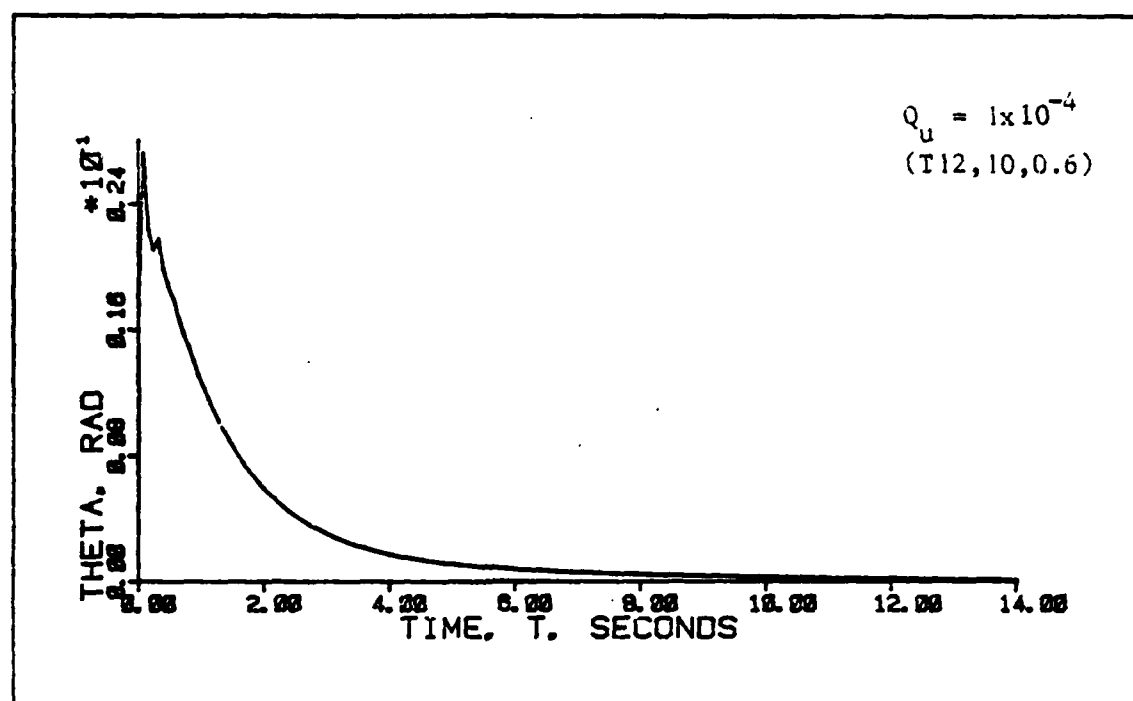


Figure 6-27a: Mean of  $\Theta$  With First-Order Colored Noise Addition

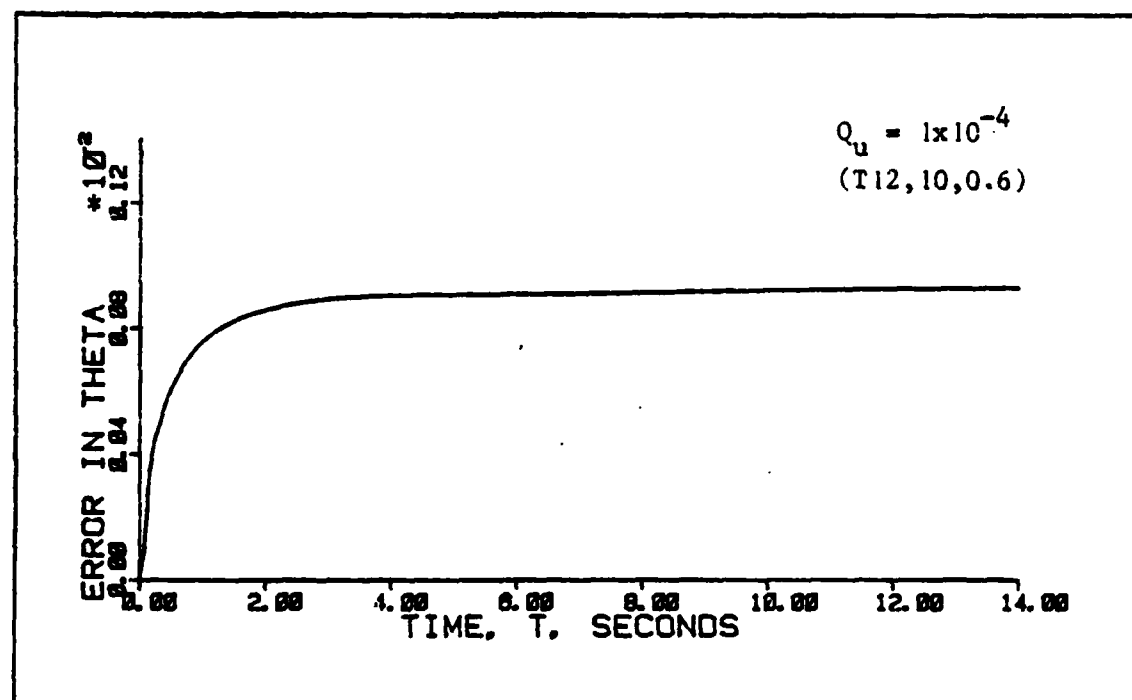


Figure 6-27b: Standard Deviation of  $\Theta$  With First-Order Colored Noise Addition

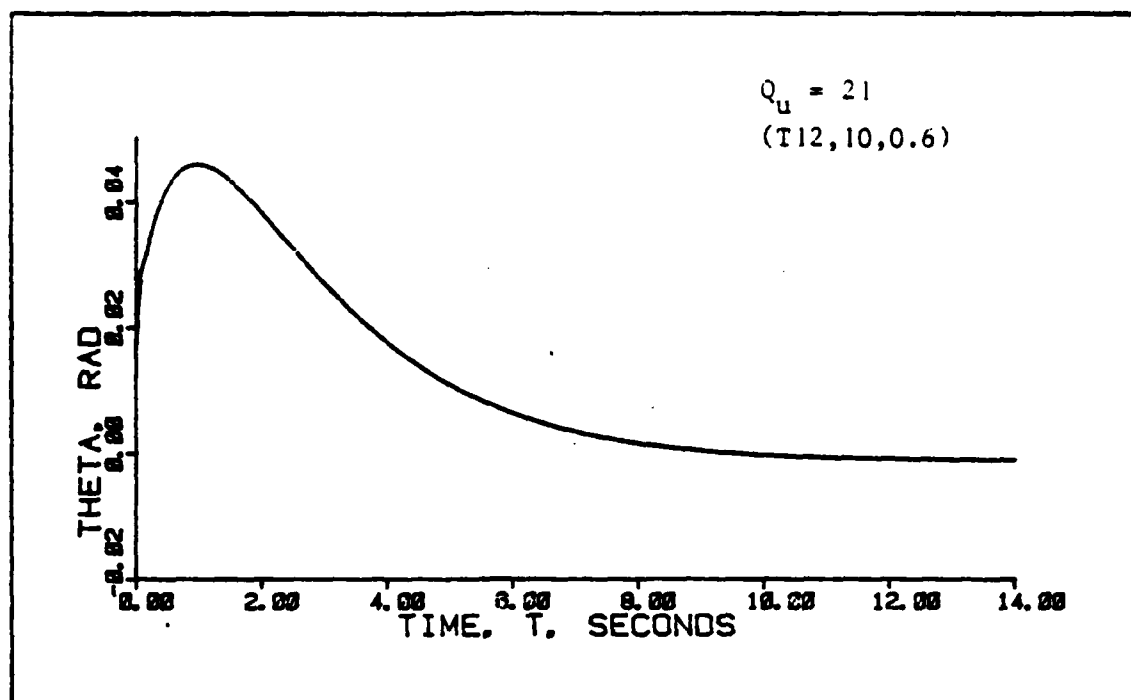


Figure 6-28a: Mean of  $\Theta$  With Second-Order Colored Noise Addition

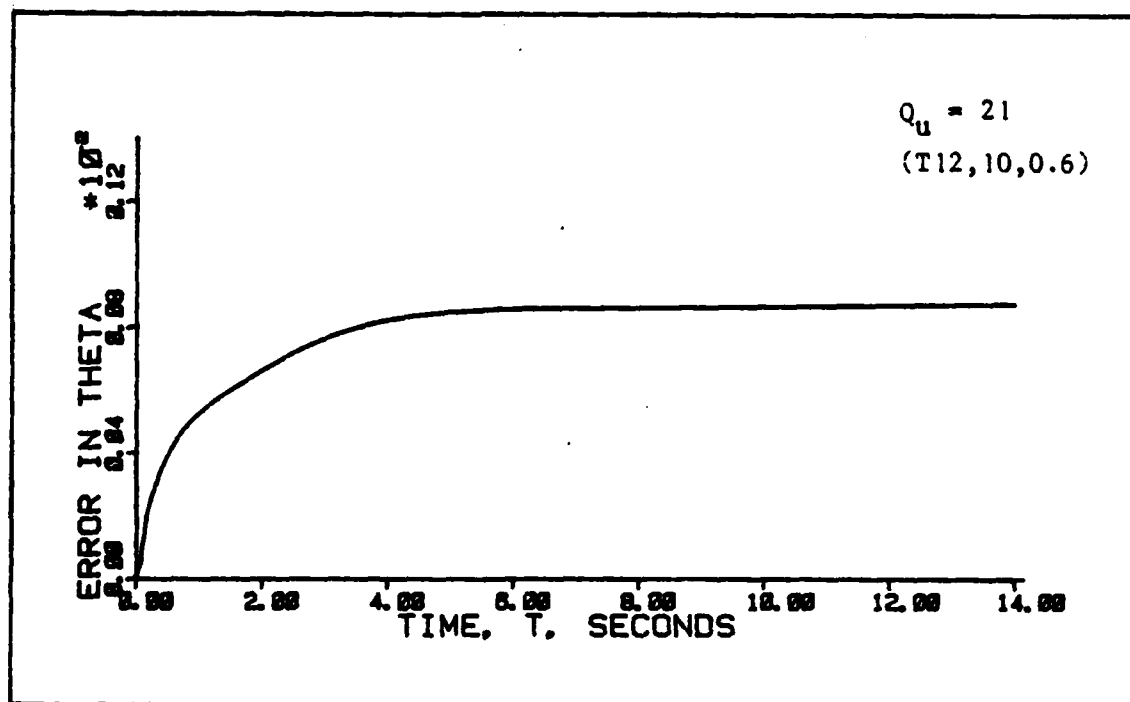


Figure 6-28b: Standard Deviation of  $\Theta$  With Second-Order Colored Noise Addition

Table 6-6

Comparison of Steady-State Standard Deviations  
of Aircraft States at the Design Condition  
For A Sampled-Data System

	$q^2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No Noise	0	-	$8.712 \times 10^{-4}$	$5.036 \times 10^{-3}$	$1.206 \times 10^{-3}$
White Noise	$1 \times 10^{-4}$	-	$9.583 \times 10^{-4}$	$5.307 \times 10^{-3}$	$1.980 \times 10^{-3}$
1st Order Shaping Filter	-	$1 \times 10^{-4}$	$1.002 \times 10^{-3}$	$1.842 \times 10^{-3}$	$1.765 \times 10^{-3}$
2nd Order Shaping Filter	-	21	$8.700 \times 10^{-4}$	$5.040 \times 10^{-3}$	$1.208 \times 10^{-3}$

second-order colored noise. The results are essentially the same as for the continuous-time controller. The same trend is observed in Table (6-7) where higher-order actuator dynamics are included in the truth model. As in the continuous-time case, adding white noise to the model does degrade the performance for all states at the design condition. However, when third-order dynamics are included in the truth model, white input noise decreases the standard deviation  $\theta$ . When the filter is evaluated in an environment other than the design condition, adding noise can improve the filter tuning. Decreases in the standard deviations appear in all states for second-order colored noise, and they are not substantial enough to justify the added complexity of the design model. Thus, the Doyle and Stein technique extended to a sampled-data system provides the desired robustification against ignored states for the problem considered in this thesis. Significant performance benefits are not gained by adding time-correlated noise to the system model as opposed to white noise.

Table 6-7

Comparison of Steady-State Standard Deviations  
Of Aircraft States with Higher-Order Actuator  
Dynamics For A Sampled-Data System

	$q^2$	$Q_u$	$\sigma_\theta$	$\sigma_\alpha$	$\sigma_q$
No noise	0	-	$8.984 \times 10^{-4}$	$4.894 \times 10^{-3}$	$1.197 \times 10^{-3}$
White Noise	$1 \times 10^{-4}$	-	$8.831 \times 10^{-4}$	$5.689 \times 10^{-3}$	$4.859 \times 10^{-3}$
1st Order Shaping Filter	-	$1 \times 10^{-4}$	$9.311 \times 10^{-4}$	$5.815 \times 10^{-3}$	$5.143 \times 10^{-3}$
2nd Order Shaping Filter	-	21	$8.789 \times 10^{-4}$	$5.170 \times 10^{-3}$	$2.342 \times 10^{-3}$

#### 6.2.6 Sampled-Data LQG Regulators at Off-Design Condition

As shown in Figure (6-29), the response of the system is unstable at the off-design flight condition with no noise addition.

White noise addition of strength  $q^2 = 1 \times 10^{-4}$ , as shown in Figure (6-30) is sufficient to recover the stability of the system. First-order colored noise of the same maximum intensity produces a very similar response, as shown in Figure (6-31). However, second-order colored noise of the same maximum intensity is not sufficient to recover stability completely as demonstrated by the divergence of the standard deviation in Figure (6-32). Increasing the maximum noise intensity to  $1 \times 10^{-2}$  achieves the desired stability characteristics as shown in Figure (6-33), but the time response is still much slower than for the previous two cases.

It is observed in Figures (6-29) through (6-33) that all three types of input noise substantially improve the robustness properties of the



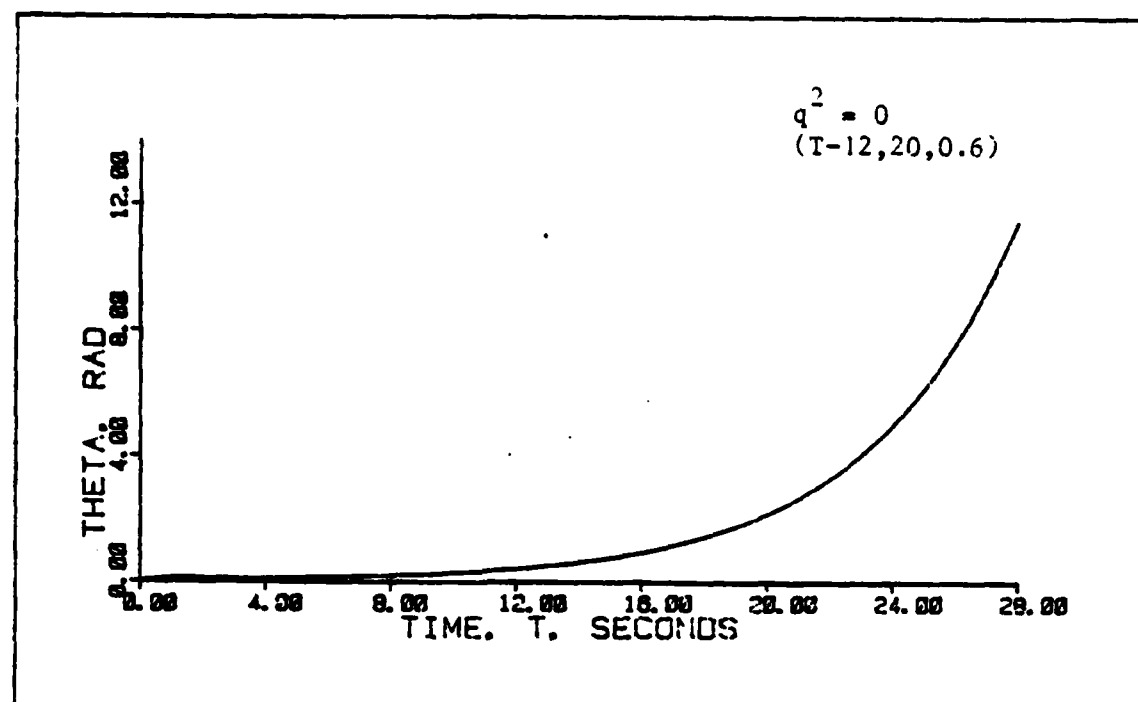


Figure 6-29a: Mean of  $\theta$  With No Noise Addition at the Off-Design Flight Condition

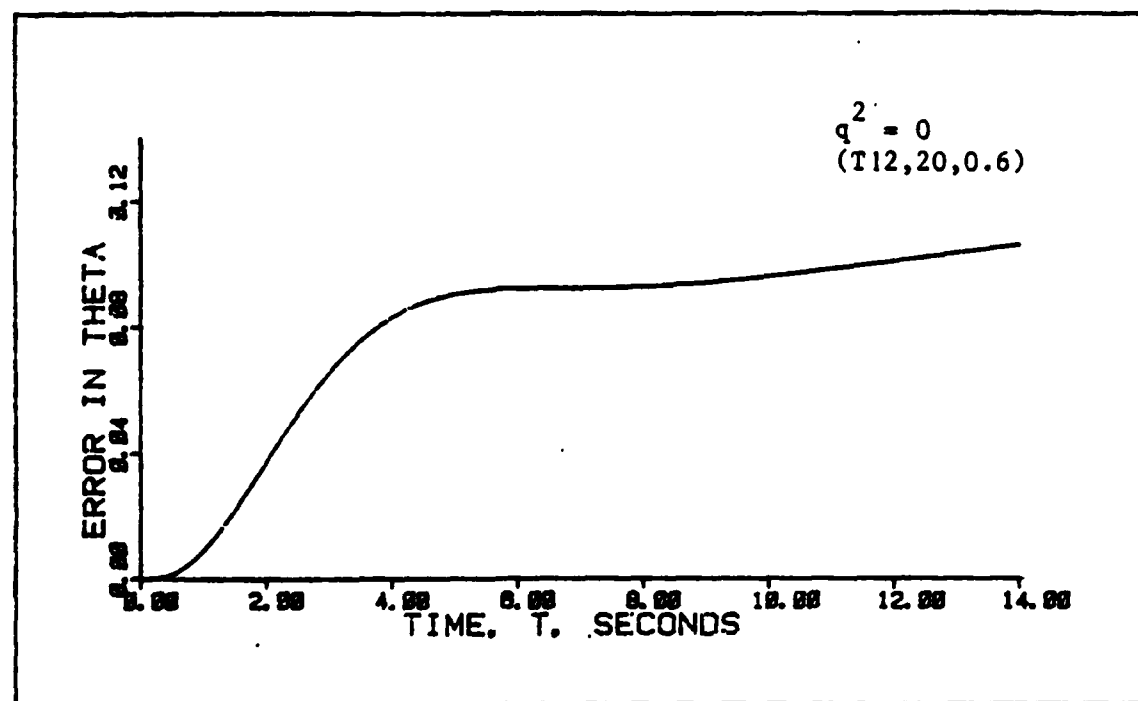


Figure 6-29b: Standard Deviation of  $\theta$  With No Noise Addition at the Off-Design Condition

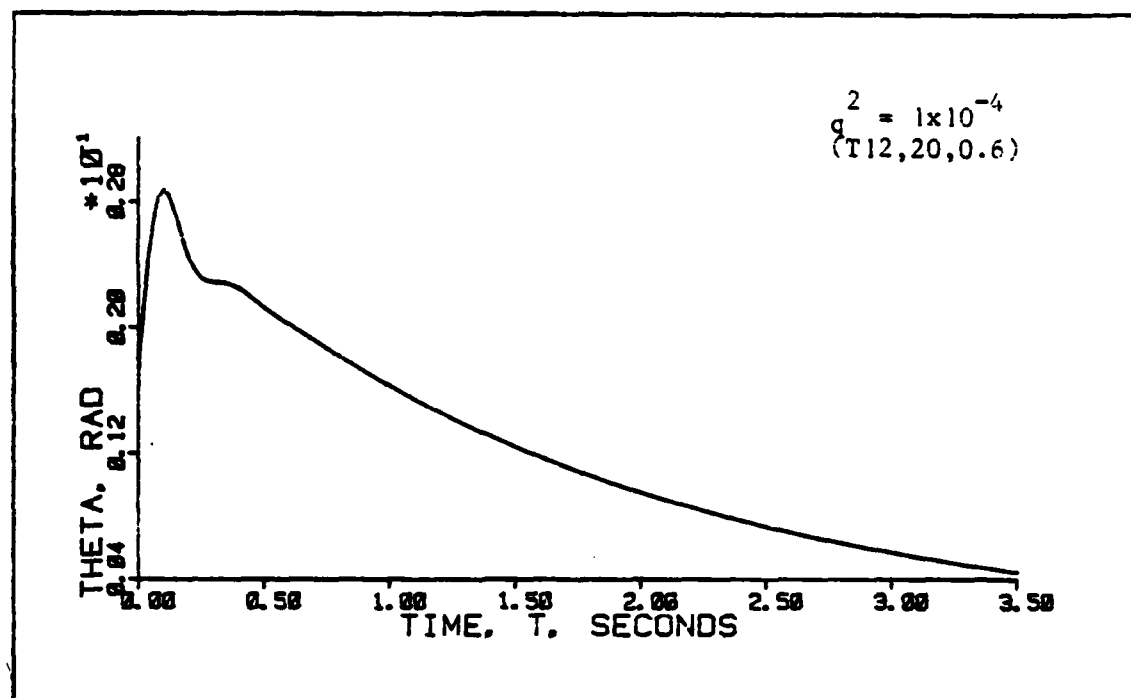


Figure 6-30a: Mean of  $\Theta$  With White Noise Addition  
Off-Design Flight Condition

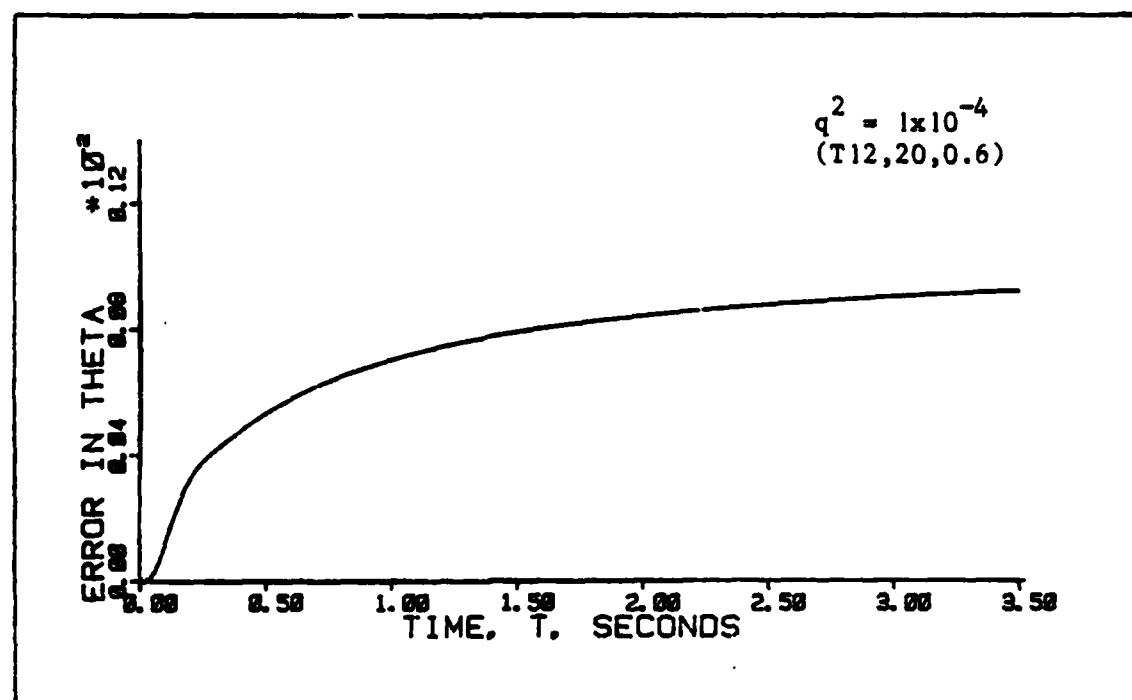


Figure 6-30b: Standard Deviation of  $\Theta$  With White Noise  
Addition at the Off-Design Flight Condition

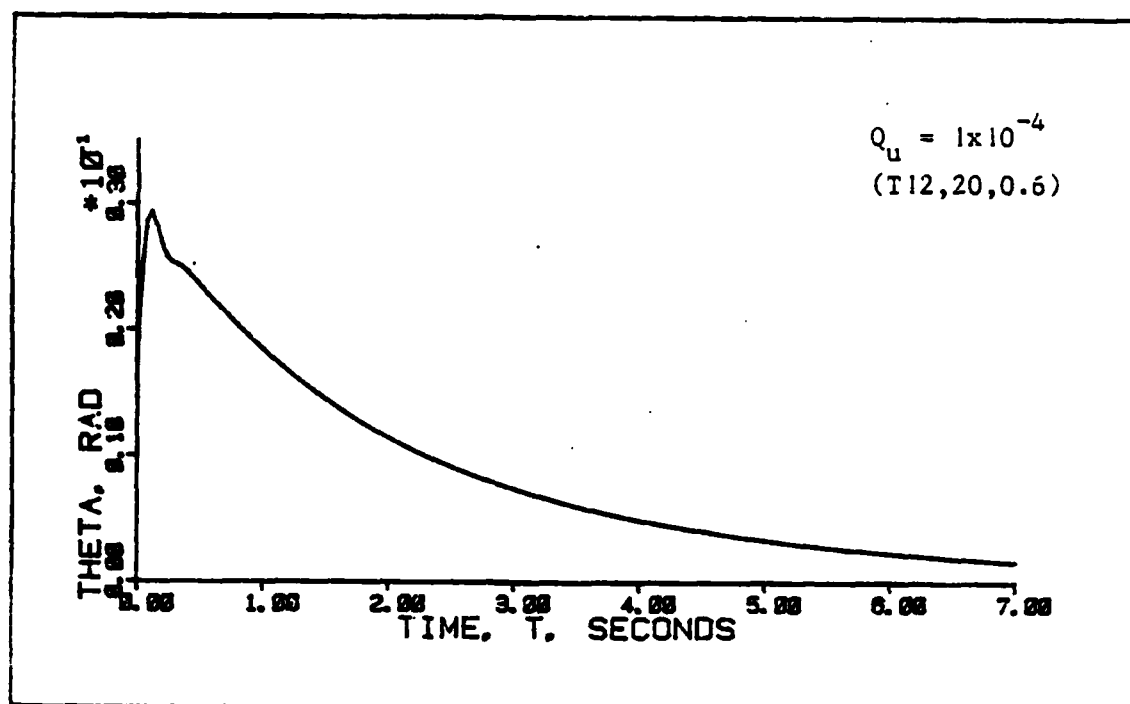


Figure 6-31a: Mean of  $\Theta$  With First-Order Colored Noise Addition at the Off-Design Flight Condition

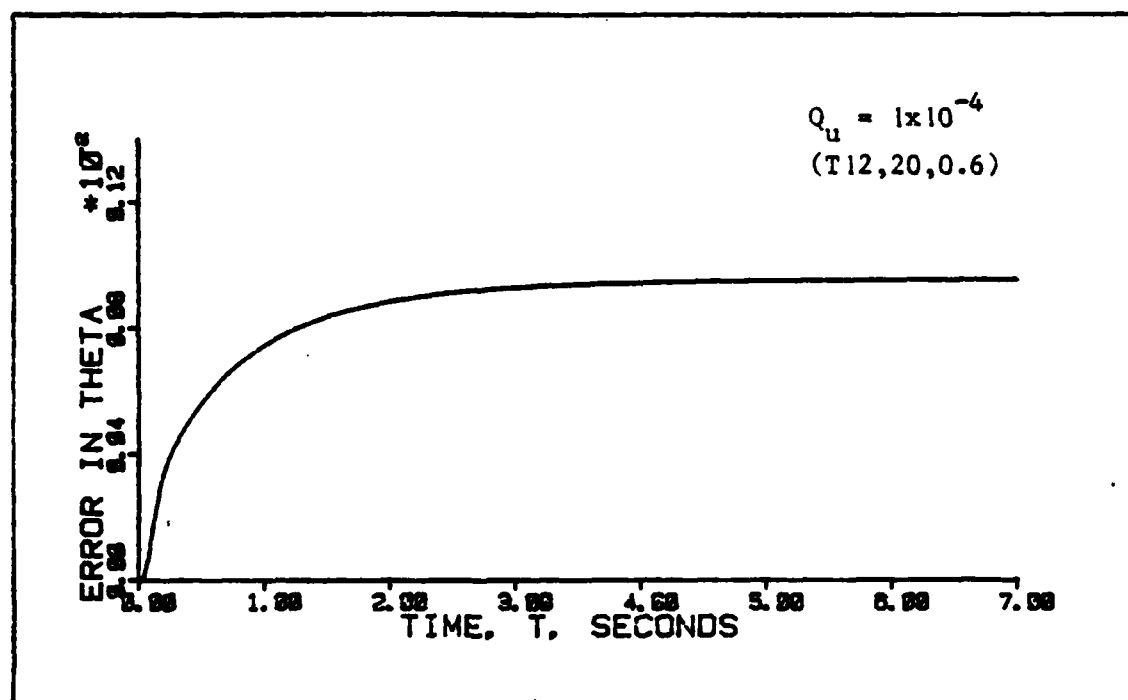


Figure 6-31b: Standard Deviation of  $\Theta$  With First-Order Colored Noise Addition at the Off-Design Flight Condition

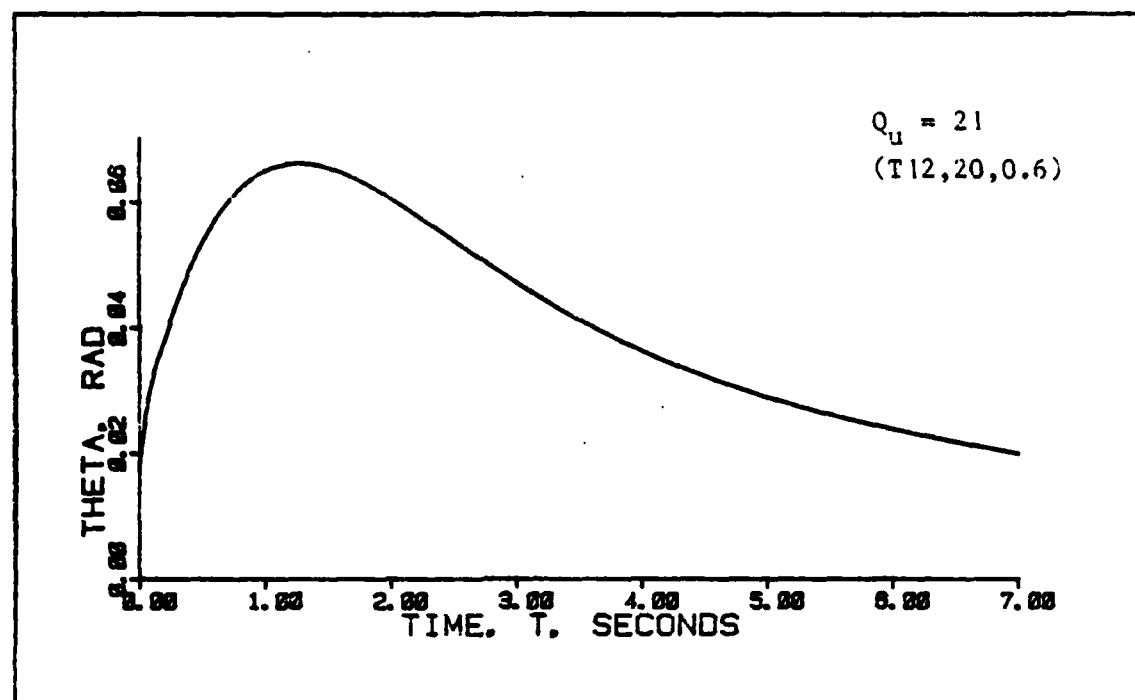


Figure 6-32a: Mean of  $\Theta$  With Second-Order Colored Noise Addition at the Off-Design Flight Condition

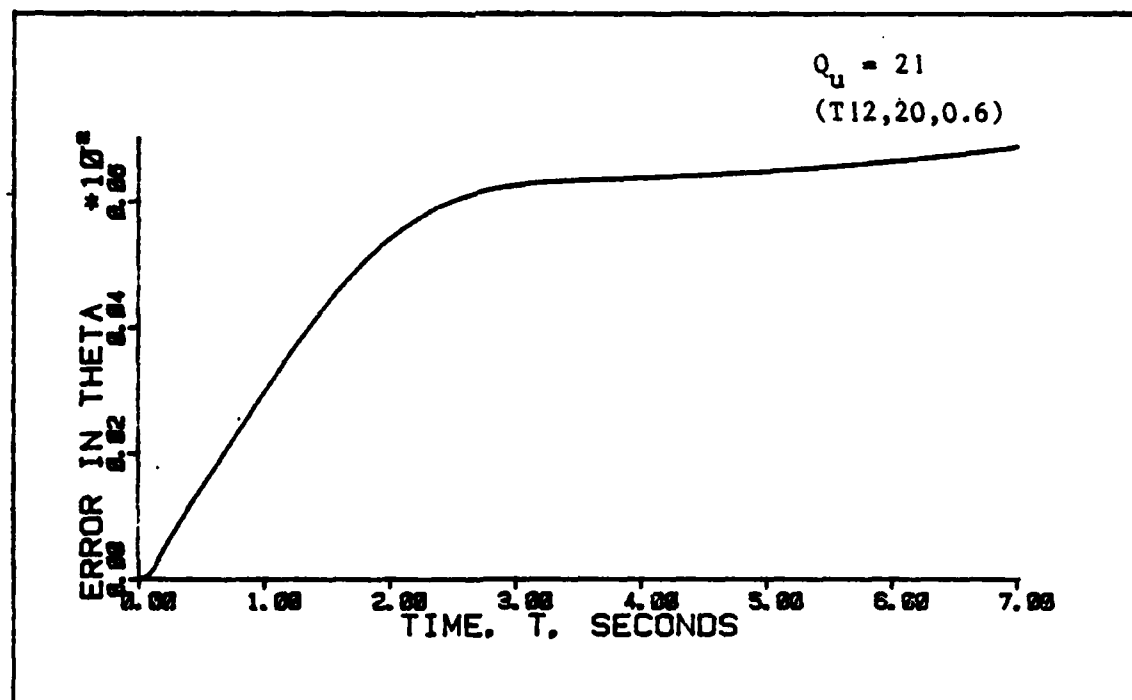


Figure 6-32b: Standard Deviation of  $\Theta$  With Second-Order Colored Noise Addition at the Off-Design Flight Condition

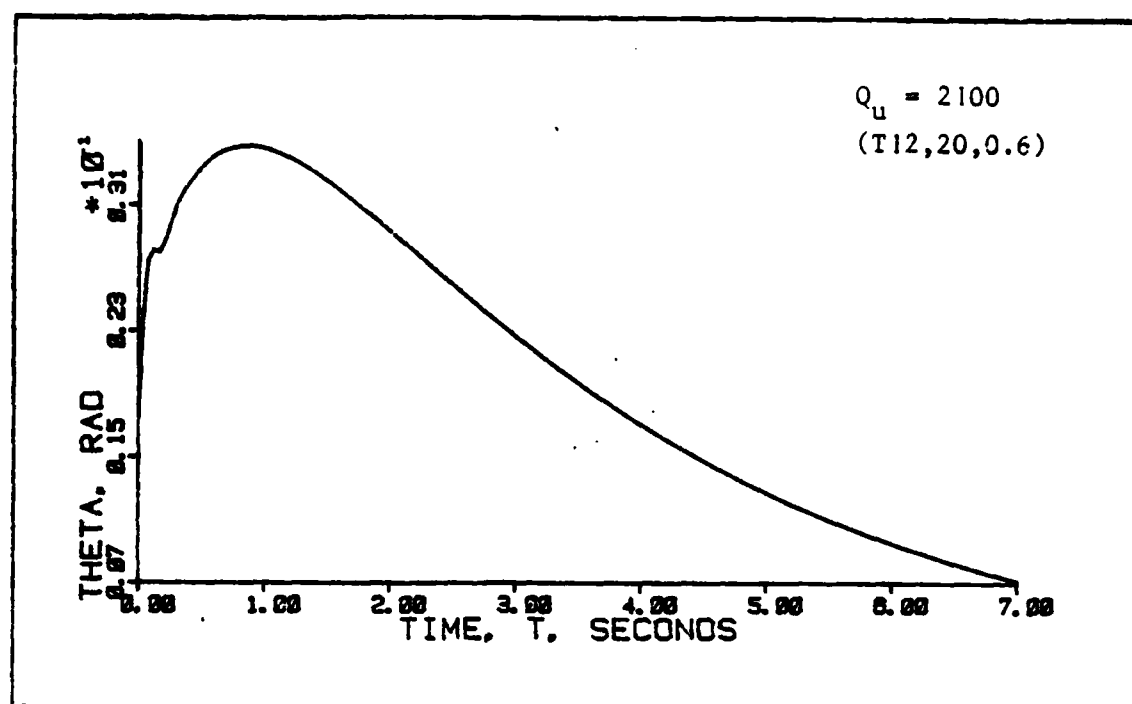


Figure 6-33a: Mean of  $\Theta$  With Second-Order Colored Noise Addition at the Off-Design Flight Condition

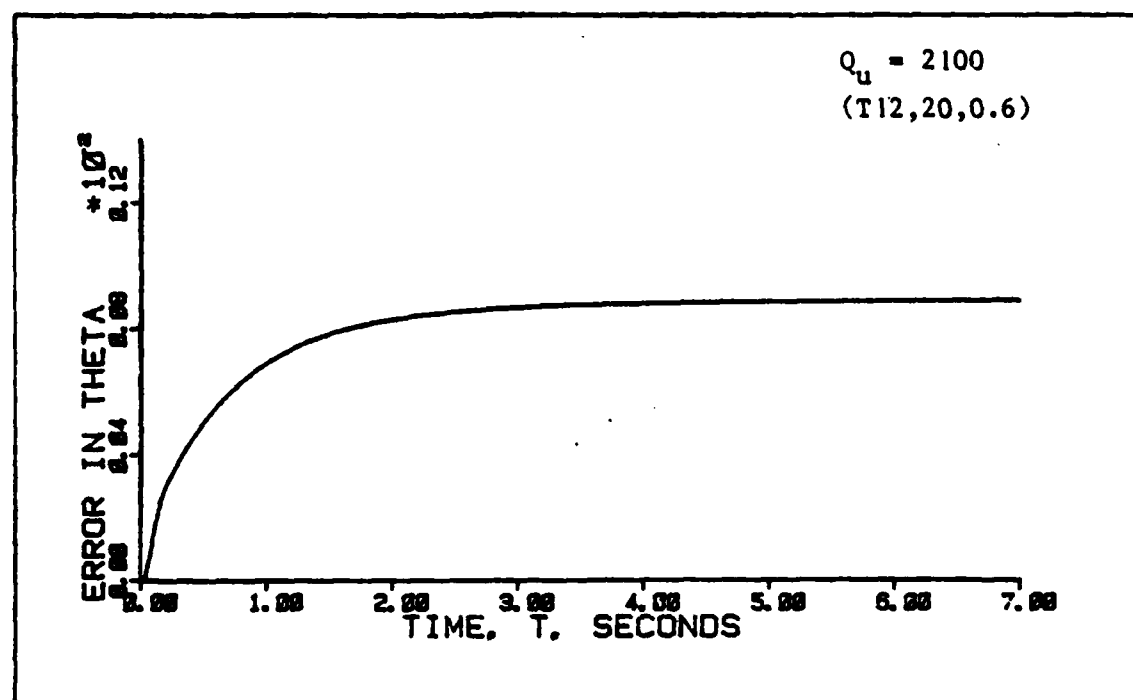


Figure 6-33b: Standard Deviation of  $\Theta$  With Second-Order Colored Noise Addition at the Off-Design Flight Condition

controlled system when it is subjected to changes in real-world parameters. The improvement due to the addition of white and first-order colored noise is very similar. This indicates that even at the off-design flight condition, the design model is still fairly adequate at low frequencies where the power spectral density of the first-order colored noise is low. Robustness can also be enhanced with second-order colored noise to a lesser degree if the maximum intensity of the noise is sufficiently high. As in the continuous-time case, it is felt that this effect is partially due to the magnitude of the noise at other frequencies, not just the peak near 70 radians/second.

### 6.3 Sampled-Data PI Controllers

The results of applying the Doyle and Stein technique to a sampled-data PI controller were inconclusive for this problem. However, some results are presented to show how employing a Kalman filter to estimate the states of a system can adversely affect stability.

The software used to design the PI controller allows the design of a full-state feedback controller or one with a Kalman filter in the loop. The performance analysis program then generates time histories of the standard deviation of the states both with and without a filter.

The weighting on the states, controls and control rates are defined in Section 3.2.2 and are given below:

$$X_{11} = \begin{bmatrix} 20 & -10 & -0- & -0- \\ -10 & 10 & -0- & -0- \\ -0- & -0- & -0- & -0- \\ -0- & -0- & 1 & 0 \\ & & 0 & 1 \end{bmatrix} \quad (6-3)$$

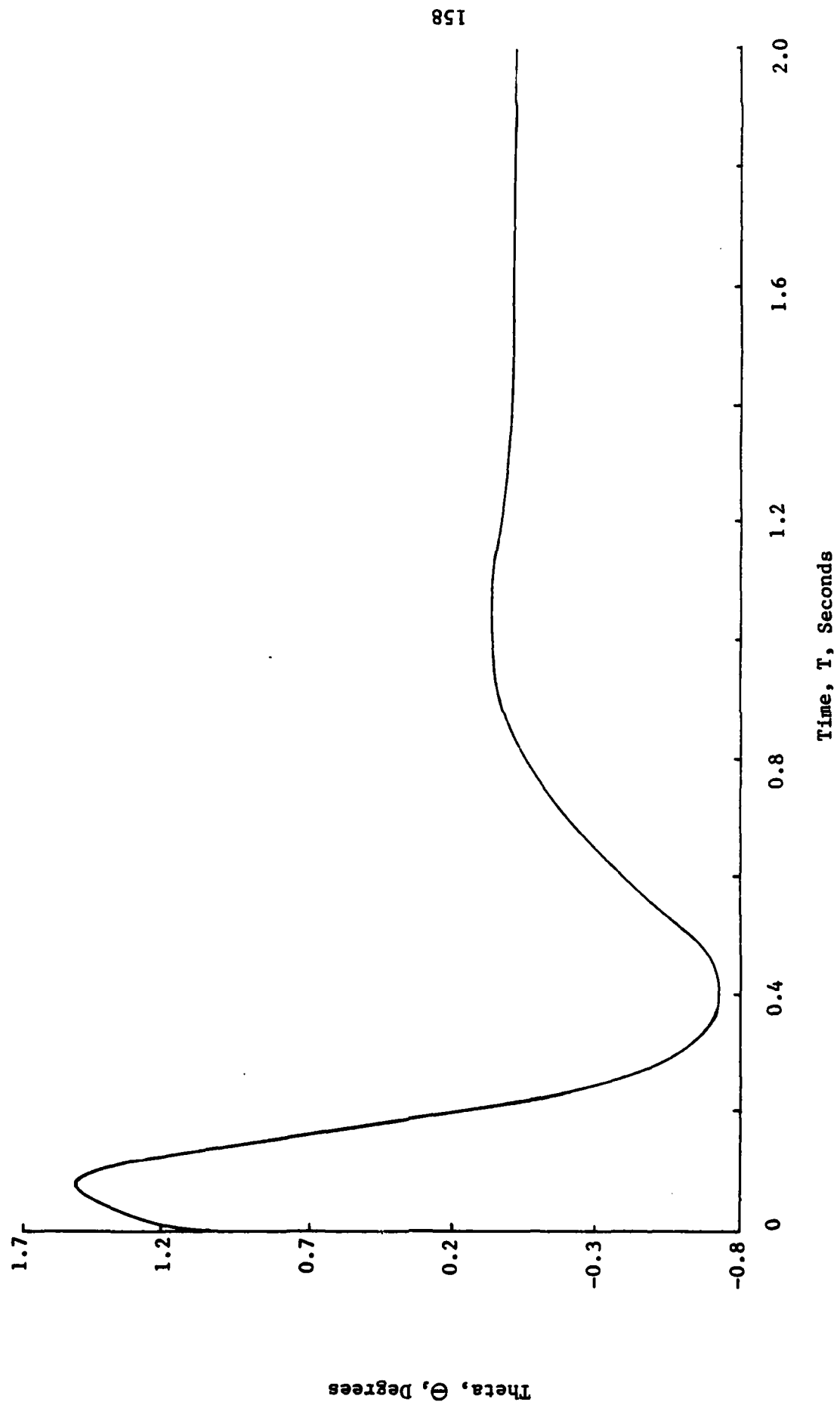


Figure 6-34: Full-State Feedback PI Controller Response of  $\Theta$

5

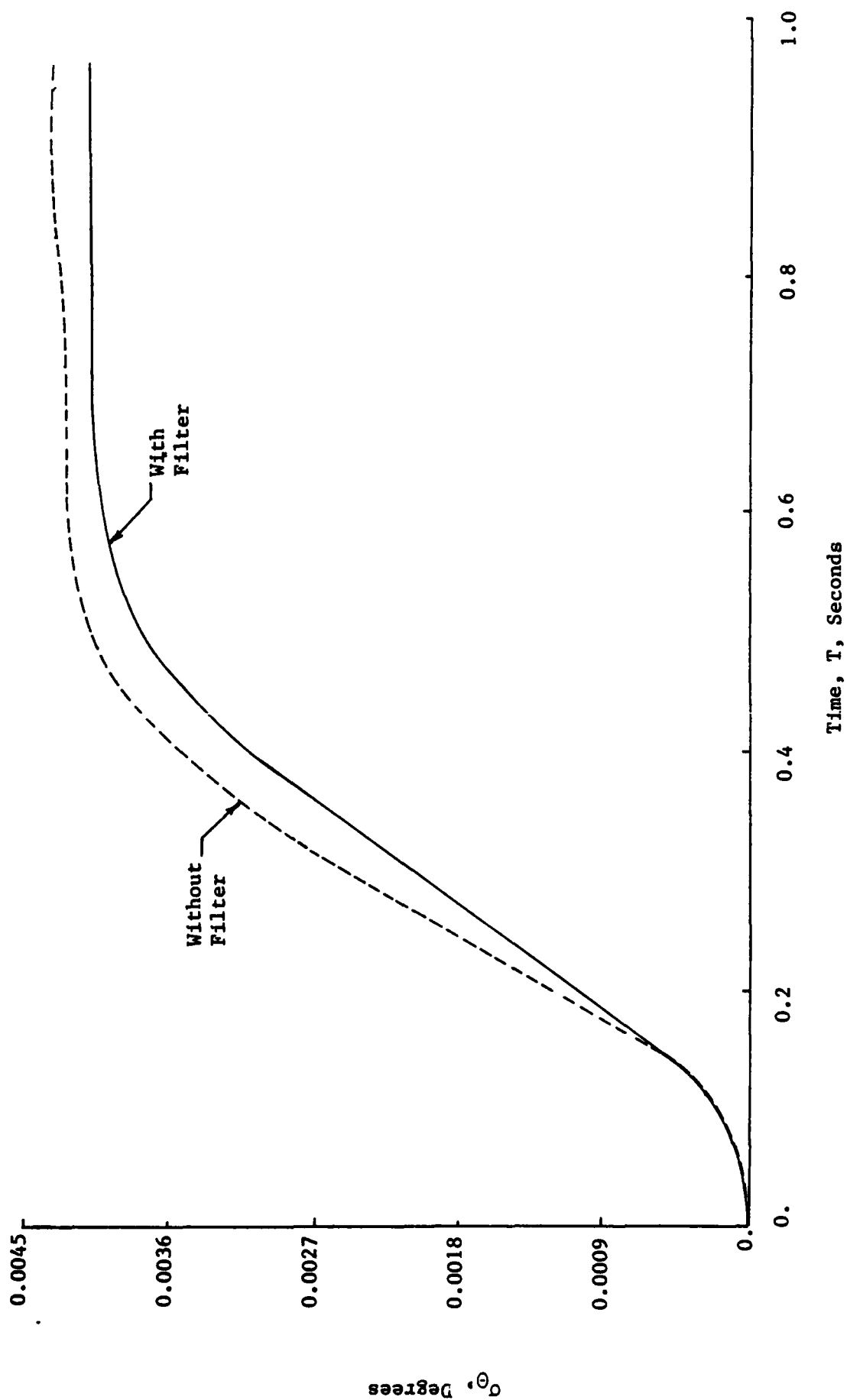


Figure 6-35: Standard Deviation of  $\Theta$  at Design Condition



$$X_{22} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad (6-4)$$

$$U = \begin{bmatrix} 10 & 0 \\ 0 & 10 \end{bmatrix} \quad (6-5)$$

The weighting matrices specified above are given in Reference 27. They are arrived at using a methodology called 'implicit model following'. As mentioned before, the software used to design PI regulators also includes a method for designing a Command Generator Tracker to specify feedforward gains that cause the output of the controlled system to track the output of a command model in which the desired characteristics of the response have been built in (damping ratio, overshoot, settling time, etc.). Implicit model following can be used to affect the feedback gains of the PI controller in a way that makes the controlled system more robust. A more detailed explanation of implicit model following is given in References 27 and 29.

Figure (6-34) shows the time-response of the full-state feedback system with an initial condition of one degree on  $\Theta$  evaluated against a deterministic truth model of the same dimension. It is seen that the response has very nice, well-damped second-order characteristics. Figure (6-35) demonstrates the standard deviations for this controller in a stochastic environment are very similar with and without a filter in the loop. The design model is defined in Section 5.4. In Figure (6-36),

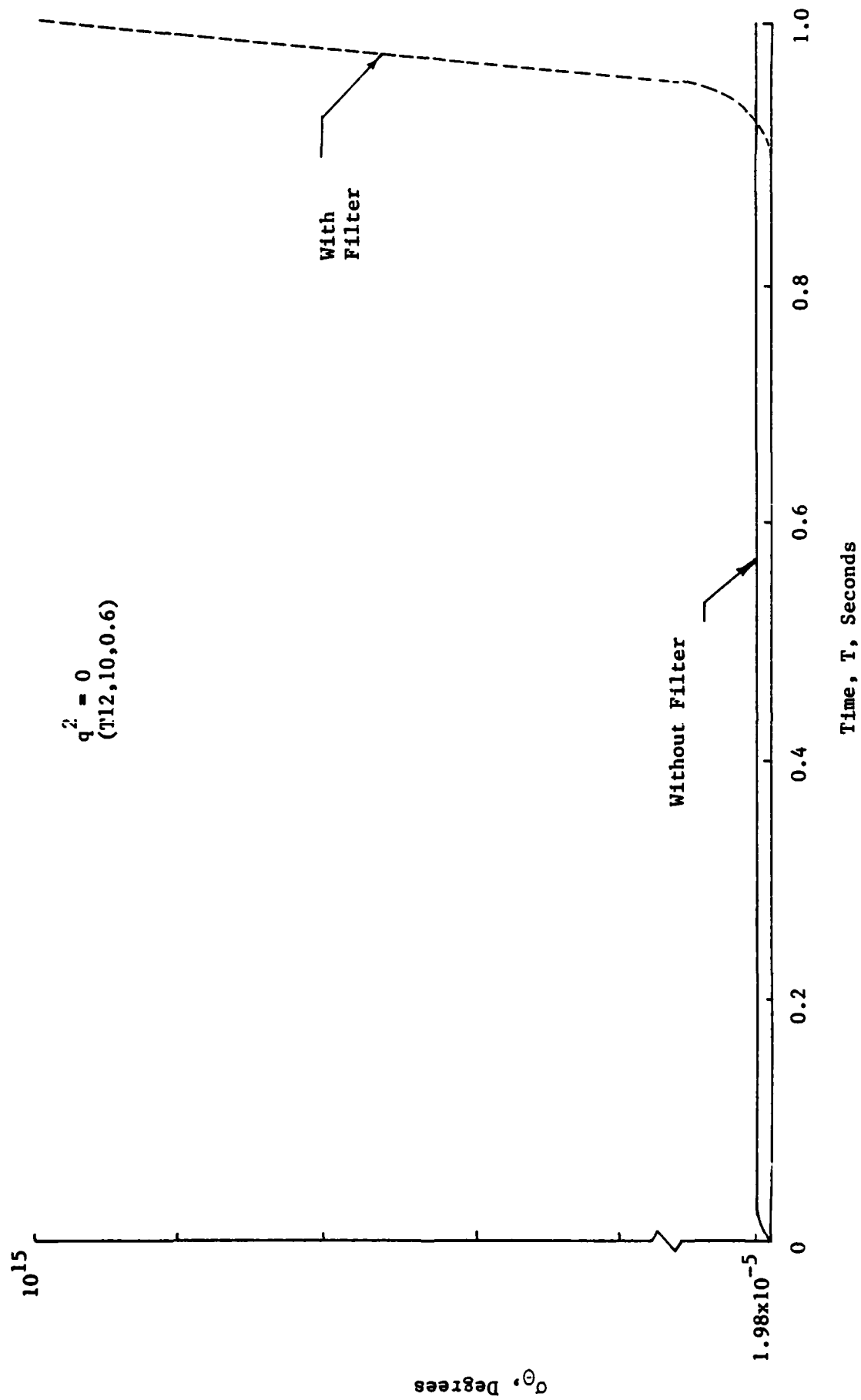


Figure 6-36: Standard Deviation of  $\theta$  Evaluated Against Third-Order Actuator Dynamics

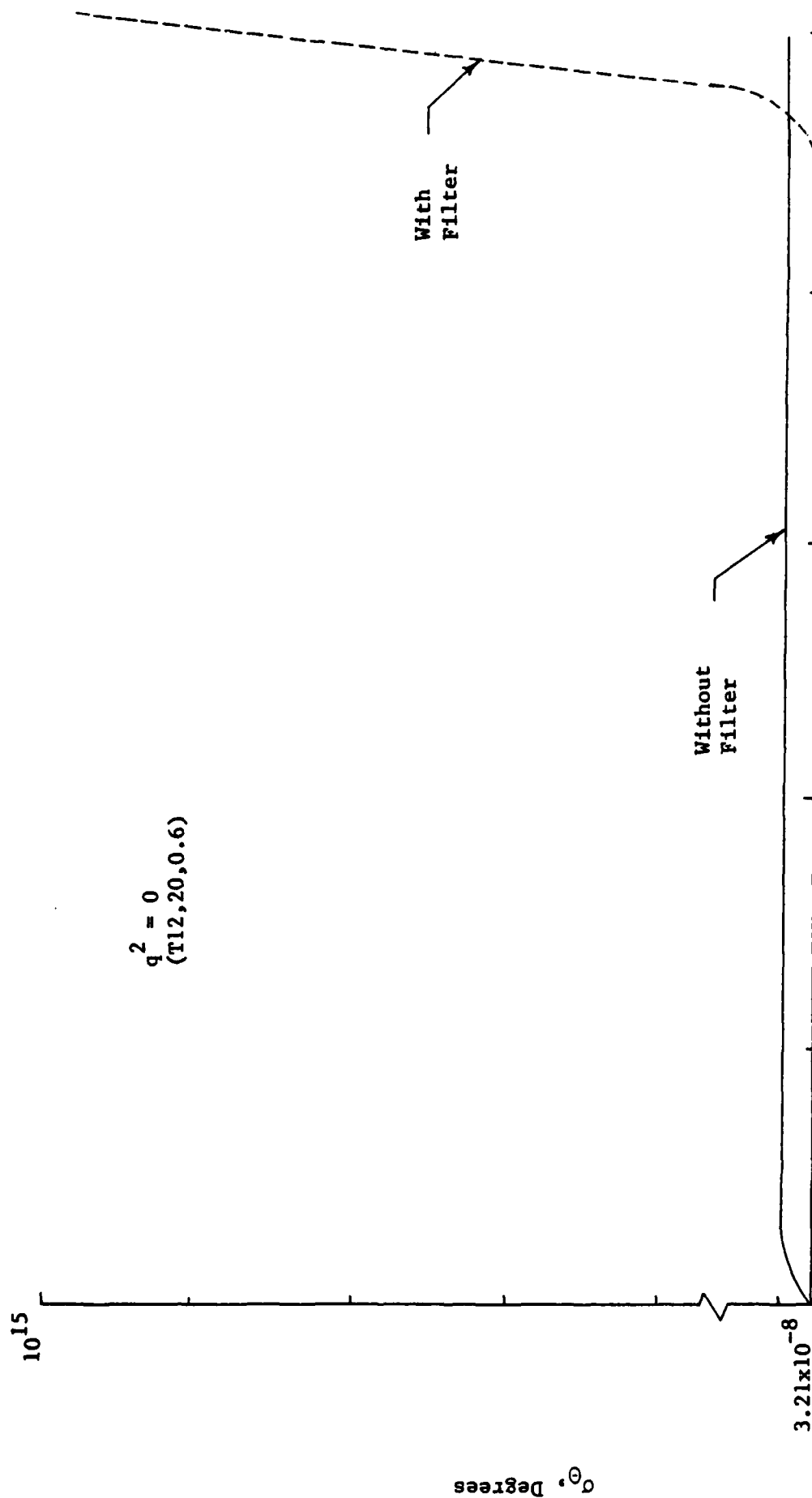


Figure 6-37: Standard Deviation of  $\theta$  At an Off-Design Flight Condition

the results of the performance analysis are shown, where the eight-state controller was evaluated against a truth model with higher order actuator dynamics in the truth model. It is seen that without full-state feedback, stability was lost. The results are the same at an off-design condition (T12,20,0.6) as shown in Figure (6-37). Thus, the figures demonstrate that when actuator states are ignored in the design model, a full-state feedback controller exhibits good stability robustness properties. However, once a filter is inserted into the loop, these properties are lost.

Attempts were made to recover the stability of the controlled system by applying the Doyle and Stein technique. It was found that any non-zero value of  $q^2$  only made the standard deviations diverge more rapidly. This was true for both flight conditions.

Recall that the Doyle and Stein technique was applied to sampled-data LQG regulators by modifying the  $Q_u$  matrix as follows:

$$Q_d(q) = Q_{do} + q^2 B V B^T \Delta t \quad (6-6)$$

using the B matrix of the continuous-time system equation.

The technique was applied to sampled-data PI controllers with the equation

$$Q_d(q) = Q_{do} + q^2 B_d V B_d^T / \Delta t \quad (6-7)$$

using the  $B_d$  matrix of the equivalent discrete-time system equation.

The Doyle and Stein technique was derived by equating the return-difference mappings for a full-state feedback system and a filter-based system. The derivation was based on the controllers being a full-state

feedback regulator and an LQG regulator. It was expected, since the Kalman filter is the same for LQG regulators and PI controllers, that at least some robustness enhancement could be obtained by directly applying the technique to a PI controller. However, this was not found to be true.

#### 6.4 Summary

This chapter has presented the results of applying two techniques to improve the robustness properties of a controlled system. The two methods examined are the techniques of inputting stationary white or time-correlated Gaussian noise into a system model during the process of tuning the Kalman filter.

Two separate issues of robustness were examined. The first was the idea of robustifying a controller so that instabilities do not occur when states are ignored in the controller and Kalman filter design model. For this thesis, the problem considered was an aircraft flight control problem, and the effect of ignoring states that described the actuator dynamics was examined. In this particular instance, the design model misrepresentation was confined to a particular portion of the frequency spectrum; this specifically motivated investigation of adding time-correlated rather than white noise for robustification.

The second robustness issue was the idea of robustifying a controller against changes in the real-world parameters upon which the controller design was based. Here the designed model misrepresentation was not confined to only a particular band of frequencies.

The primary result of this chapter is that the stability robustness of the controlled system can be substantially enhanced by applying both

robustification techniques. However, for the problem considered, the Doyle and Stein technique of inputting white noise into the system model is the most appropriate way to achieve the robustness enhancement.

Conclusions about the results presented in this chapter are made in the following chapter. In addition, some recommendations for further research are made.

## VII. Conclusions and Recommendations

The robustification of LQG regulators by inputting white noise or time-correlated noise generated by a first-order shaping filter into the system model did work well for all three applications of the controllers examined (continuous, discretized, sampled-data). In addition, time-correlated noise generated by a second-order shaping filter improved the controller's robustness properties to a lesser degree for the continuous-time and sampled-data case. For the discretized case, this method produced instabilities in the response of the system and could not be used.

It was noted that in Chapter VI that few performance benefits were gained by using time-correlated noise as opposed to white noise for the problem considered in this thesis. It is felt that the range of frequencies where the design model differed from the chosen representation of the real world was not sufficiently narrow in this case to result in performance benefits. Greater performance benefits may be realized by examining the application of colored noise addition to problems such as gust-load alleviation, flutter suppression, or aeroelastic effects such as including first- and higher-order bending modes in the truth model representation of the real world.

It is felt that the robustification techniques for LQG regulators were examined in great detail in this thesis. However, the extension to PI controllers was only touched upon, and the Doyle and Stein technique was applied unsuccessfully. To examine this more thoroughly, it is suggested that the Doyle and Stein technique be applied more directly to PI controllers. That is, the conditions analogous to those of Doyle and Stein that would make the return-difference mapping asymptotically equal

for a full-state feedback controller and a filter-based PI controller should be derived.

The technique of injecting time-correlated noise into the system model for a PI controller was not examined in this thesis. The existing software was not originally designed to allow augmenting of states with the design model. Therefore, it would be difficult to augment shaping filter states to allow for colored input noise. The modifications shown in Appendix C were made so that a few cases of colored input noise could be examined quickly. If an extensive study were made, the awkwardness of these modifications would become apparent. To examine all of the robustification techniques discussed in this thesis would thus require additional software.

It has been mentioned that a drawback of adding colored-noise to a system model is that it adds states to the Kalman filter model. An alternative method, called residualization (Ref 28), reduces the order of the model back to that of the original system before the shaping filter states were added. It would be useful to examine this implementation and compare it to the performance of the implementation used in this thesis for a problem where colored input noise is appropriate.

Finally, as noted in Sections 6.2.3 and 6.2.4, the technique of adding white or colored noise to a continuous-time controller, then discretizing the controller only improved robustness characteristics for a finite range of  $q^2$  or  $Q_u$ . Beyond that range, the closed-loop system was unstable. This phenomenon is not well understood, and further research would be warranted to determine why this occurs. While it is not expected that an extension of the Doyle and Stein method would have all



the same characteristics of the original technique, it would be useful to determine what occurs during the discretization process to drive closed-loop system poles outside the  $z$ -domain unit circle so abruptly.

The results of the previous chapter have thus demonstrated that the techniques considered can substantially improve the robustness properties of a controlled system. For the particular problem considered white noise added to the system model at the control entry points is the appropriate method for accomplishing the robustification

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## APPENDIX A: Generic Controller Format

### A1. Introduction

In Chapter II, the idea of a standard format for a linear controller was introduced. The format is particularly useful in that any LQG controller synthesized by LQG methods can be rearranged into the standard form. Subsequently, the performance analysis equations presented in Chapter II apply to any controller put in the "generic" form, not just LQG regulators.

This appendix defines explicitly the generic form controller equations for the types of controllers considered in this thesis. In the first section, the continuous-time LQ regulator is examined, that is, a full-state feedback regulator with a control law expressed as:

$$\underline{u}^*(t) = G_c^* \underline{x}(t) \quad (A-1)$$

The full-state feedback generic structure is presented because it is desired to recover the robustness characteristics of the LQ controller with the addition of input noise as covered in Chapters II and IV.

Then, the continuous-time LQG regulator is presented, based on a control law that employs state estimates from a Kalman filter:

$$\underline{u}^*(t) = -G_c^* \hat{\underline{x}}(t) \quad (A-2)$$

Next, the structure of a sampled-data, full-state feedback regulator is examined with a control law expressed as

$$\underline{u}^*(t_i) = -G_c^* \underline{x}(t_i) \quad (A-3)$$

Two forms of a sampled-data LQG regulator are examined. The first is based on the optimal control law given by

$$\underline{u}^*(t_i) = -G_c^* \hat{\underline{x}}(t_i^+) \quad (\text{A-4})$$

The second is based on the suboptimal control law

$$\underline{u}^*(t_i) = -G_c^* \hat{\underline{x}}(t_i^-) \quad (\text{A-5})$$

In the last two sections, the format is presented for a sampled-data PI controller designed on the basis of LQG methodology. First, the full-state feedback case is considered, then a Kalman filter is used to provide state estimates.

#### A.2 Continuous-time Controller Generic Structure

The generic format desired for the controllers considered in this thesis is given by

$$\underline{u}^*(t) = G_{cx} \underline{x}_c(t) + G_{cz} \underline{z}(t) + G_{cy} \underline{y}_d(t) \quad (\text{A-6})$$

$$\underline{x}_c(t) = F_c \underline{x}_c(t) + B_{cz} \underline{z}(t) + B_{cy} \underline{y}_d(t) \quad (\text{A-7})$$

which takes the form of an algebraic relation for the optimal control,  $\underline{u}(t)$  and a propagation equation for the controller states. The controller states,  $\underline{x}_c(t)$ , are defined for the particular type of controller considered, and  $\underline{z}(t)$  and  $\underline{y}_d(t)$  are measurements and desired values of controlled variables, respectively.

For the full-state feedback case, perfect measurements of the states are available, i.e.,

$$\underline{z}(t) = \underline{x}(t) \quad (\text{A-8})$$

and  $\underline{y}_d(t) = \underline{0}$ . Thus, it is seen by comparing Equations (A-1) and (A-6) that

$$G_{cx} = 0 \quad (A-9)$$

$$G_{cz} = -G_c^* \quad (A-10)$$

$$G_{cy} = 0 \quad (A-11)$$

In this instance, there are no internal controller states, so Equation (A-7) is not maintained.

For an LQG regulator,  $\underline{y}_d(t)$  is again zero, therefore  $G_{cy}$  and  $B_{cy}$  can be set to zero. The controller states are defined to be the conditional mean estimates from a Kalman filter,  $\hat{\underline{x}}(t)$ . By the certainty equivalence principle,  $\underline{x}(t)$  in Equation (A-2) can be replaced by  $\hat{\underline{x}}(t)$  which are now the controller states,  $\underline{x}_c(t)$ :

$$\underline{u}^*(t) = -G_c^* \underline{x}_c(t) \quad (A-12)$$

Comparing Equations (A-6) and (A-12), it is seen that

$$G_{cz} = 0 \quad (A-13)$$

$$G_{cz} = -G_c^* \quad (A-14)$$

Recall that the continuous-time Kalman filter equation yields an estimate of the states,  $\hat{\underline{x}}(t)$ , where  $\hat{\underline{x}}(t)$  is now defined as the controller states,  $\underline{x}_c(t)$ . This is given by

$$\dot{\underline{x}}_c(t) = F \underline{x}_c(t) + B \underline{u}^*(t) + K [\underline{z}(t) - H \underline{x}_c(t)] \quad (A-15)$$

Substituting Equation (A-12) into the above and rearranging yields

$$\dot{\underline{x}}_c(t) = \left[ F - BG_c^* - KH \right] \underline{x}_c(t) + K \underline{z}(t) \quad (A-16)$$

Comparing Equations (A-7) and (A-16), it is seen that

$$F_c = \left[ F - BG_c^* - KH \right] \quad (A-17)$$

$$B_{cz} = K \quad (A-18)$$

### A.3 Optimal Sampled-Data Generic Controller Structure

The generic structure of Equations (A-4) and (A-5) is given in sampled-data form by

$$\underline{u}^*(t_i) = G_{cx} \underline{x}_c(t_i) + G_{cz} \underline{z}(t_i) + G_{cy} \underline{y}_d(t_i) \quad (A-19)$$

$$\underline{x}_c(t_{i+1}) = \phi_c \underline{x}_c(t_i) + G_{cz} \underline{z}(t_i) + G_{cy} \underline{y}_d(t_i) \quad (A-20)$$

It is easily seen that for a sampled-data full-state feedback regulator, the generic form is identical to Equations (A-9) through (A-11).

For the optimal LQG regulator, the actual state values are replaced by estimates just after measurements are taken at the sample times. Again,  $\underline{y}_d(t)$  will be zero for a regulator, therefore  $G_{cy}$  and  $B_{cy}$  can be set to zero. Define the controller states to be  $\hat{\underline{x}}(t_i^-)$ , the state estimates just prior to a measurement update. The Kalman filter propagation and update relations for the estimates of the states are given by

$$\underline{x}(t_{i+1}^-) = \phi \underline{x}(t_i^+) + B_d \underline{u}^*(t_i) \quad (A-21)$$



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ROBUST FLIGHT CONTROLLERS(U) AIR FORCE INST OF TECH  
WRIGHT-PATTERSON AFB OH SCHOOL OF ENGINEERRNG  
J M HOWEY DEC 83 AFIT/GAE/EE/83D-2

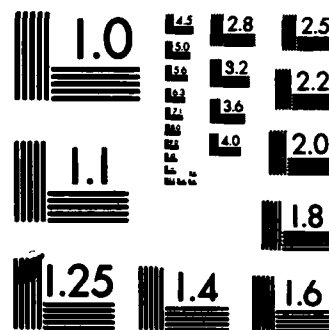
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END



MICROCOPY RESOLUTION TEST CHART  
NATIONAL BUREAU OF STANDARDS-1963-A

$$\hat{\underline{x}}(t_i^+) = \begin{bmatrix} I & -KH \end{bmatrix} \hat{\underline{x}}(t_i^-) + K\underline{z}(t_i) \quad (A-22)$$

Replace  $\hat{\underline{x}}(t_i^-)$  by  $\underline{x}_c(t_i)$  and substitute (A-22) into (A-4) to yield

$$\underline{U}^*(t_i) = -G_c^* \begin{bmatrix} I-KH \end{bmatrix} \underline{x}_c(t_i) - G_c^* K\underline{z}(t_i) \quad (A-23)$$

Comparing (A-19) and (A-23), it is seen that

$$G_{cx} = -G_c^* \begin{bmatrix} I-KH \end{bmatrix} \quad (A-24)$$

$$G_{cz} = -G_c^* K \quad (A-25)$$

Now substitute Equations (A-4) and (A-22) into (A-21), replacing  $\hat{\underline{x}}(t_i^-)$  with  $\underline{x}_c(t_i)$ :

$$\begin{aligned} \underline{x}_c(t_{i+1}) = & \begin{bmatrix} \phi - B_d G_c^* \end{bmatrix} \begin{bmatrix} I - KH \end{bmatrix} \underline{x}_c(t_i) \\ & + \begin{bmatrix} \phi - B_d G_c^* \end{bmatrix} K \underline{z}(t_i) \end{aligned} \quad (A-26)$$

Comparing (A-20) and (A-26), it is seen that

$$\phi_c = \begin{bmatrix} \phi - B_d G_c^* \end{bmatrix} \begin{bmatrix} I-KH \end{bmatrix} \quad (A-27)$$

$$B_{cz} = \begin{bmatrix} \phi - B_d G_c^* \end{bmatrix} K \quad (A-28)$$

#### A.4 Sub-optimal Sampled-Data Generic Controller Structure

For the suboptimal control law, where the state values are replaced by estimates just prior to the sample times, the controller states are defined to be  $\hat{\underline{x}}(t_i^-)$ . The output,  $\underline{y}_d(t_i)$ , is zero, thus  $G_{cy}$  and  $B_{cy}$  can be set to zero. By direct comparison of Equations (A-5) and (A-29), it is seen that

$$G_{cx} = -G_c^* \quad (A-29)$$

$$G_{cz} = 0 \quad (A-30)$$

Now substitute Equation (A-5) and (A-22) into (A-21), replacing  $\hat{x}(t_i^-)$  by  $\underline{x}_c(t_i)$ , to yield

$$\begin{aligned} \underline{x}_c(t_{i+1}) &= \{\phi[I-KH] - B_d G_c^*\} \underline{x}_c(t_i) \\ &+ [\phi K] \underline{z}(t_i) \end{aligned} \quad (A-31)$$

Thus, by comparing Equations (A-20) and (A-31), it is seen that

$$\phi_c = \{\phi[I-KH] - B_d G_c^*\} \quad (A-32)$$

$$B_{cz} = [\phi K] \quad (A-33)$$

#### A.5 Sampled-Data PI Controller Generic Structure

An equivalent form for the PI control law derived in Chapter III which achieves Type-1 control is given by (Ref 1;23).

$$u^*(t_i) = -K_x \underline{x}(t_i) + K_z \underline{\xi}(t_i) + K_z y_d(t_i) \quad (A-34)$$

where  $\underline{\xi}(t_i)$  are termed the pseudo-integral states and are expressed as

$$\underline{\xi}(t_{i+1}) = \underline{\xi}(t_i) + [y_d(t_i) - y(t_i)] \quad (A-35)$$

For the PI controller, a non-zero output,  $y_d(t_i)$ , will be allowed, where  $y(t_i)$ , the actual output of the system, is given by

$$\underline{y}(t_i) = C\underline{x}(t_i) + D_y \underline{u}^*(t_i) \quad (\text{A-36})$$

For the full-state feedback case, the controller states are defined to be  $\underline{\xi}(t_i)$ . Substituting (A-35) and (A-36) into (A-34) and recalling that  $\underline{x}(t_i) = \underline{z}(t_i)$  yields

$$\underline{u}^*(t_i) = K_{z\underline{c}} \underline{x}(t_i) - K_{\underline{x}} \underline{z}(t_i) + K_{z\underline{d}} \underline{y}(t_i) \quad (\text{A-37})$$

Comparing this equation (A-19), it is seen that

$$G_{cx} = K_z \quad (\text{A-38})$$

$$G_{cz} = -K_x \quad (\text{A-39})$$

$$G_{cy} = K_z \quad (\text{A-40})$$

Then, by substituting Equations (A-34) and (A-36) into (A-35) with  $\underline{x}_c(t_i) = \underline{\xi}(t_i)$ :

$$\begin{aligned} \underline{x}_c(t_{i+1}) &= \left[ I - D_y K_z \right] \underline{x}_c(t_i) - \left[ C - D_y K_x \right] \underline{z}(t_i) \\ &\quad + \left[ I - D_y K_z \right] \underline{y}_d(t_i) \end{aligned} \quad (\text{A-41})$$

and comparing this to Equation (A-20) yields

$$\phi_c = \left[ I - D_y K_z \right] \quad (\text{A-42})$$

$$B_{cz} = \left[ C - D_y K_x \right] \quad (\text{A-43})$$

$$B_{cy} = \left[ I - D_y K_z \right] \quad (\text{A-44})$$

With a Kalman filter in the loop,  $\underline{x}(t_i)$  is replaced by  $\hat{\underline{x}}(t_i^+)$  in Equations (A-34) and (A-36).

The controller states,  $\underline{x}_c(t_i)$ , for PI controller are defined to be an augmented vector containing the state estimates just prior to a measurement update and the pseudo-integral states:

$$\underline{x}_c(t_i) = \begin{bmatrix} \hat{\underline{x}}(t_i^-) \\ \underline{\xi}(t_i) \end{bmatrix} \quad (\text{A-45})$$

To generate the generic controller structure, first substitute the Kalman filter update relation (A-22) into Equations (A-21) and (A-34). This yields

$$\hat{\underline{x}}(t_{i+1}^-) = \phi \{ [I - KH] \underline{x}(t_i^-) + K \underline{z}(t_i) \} + B_d u^*(t_i) \quad (\text{A-46})$$

$$u^*(t_i) = -K_x \{ [I - KH] \hat{\underline{x}}(t_i^-) + K \underline{z}(t_i) \} + K_z \underline{\xi}(t_i) \quad (\text{A-47})$$

Substitute (A-47) into (A-46) to yield

$$\begin{aligned} \hat{\underline{x}}(t_{i+1}^-) = & \left[ \phi - B_d K_x \right] \left[ I - KH \right] \hat{\underline{x}}(t_i^-) \\ & + B_d K_z \underline{\xi}(t_i) + \left[ \phi - B_d K_x \right] K \underline{z}(t_i) \\ & + B_d K_z \underline{y}_d(t_i) \end{aligned} \quad (\text{A-48})$$

Next, substitute Equations (A-22) and (A-36) into (A-35) to yield

$$\begin{aligned}
\underline{\xi}(t_{i+1}) &= -\left[C - D_y K_x\right] \left[I - KH\right] \hat{\underline{x}}(t_i^-) \\
&+ \left[I - D_y K_z\right] \underline{\xi}(t_i) - \left[C - D_y K_x\right] K \underline{z}(t_i) \\
&+ \left[I - D_y K_z\right] \underline{y}_d(t_i)
\end{aligned} \tag{A-49}$$

Equations (A-47) and (A-48) in terms of the controller states,  $\underline{x}_c(t_i)$ , are given by

$$\begin{aligned}
\underline{u}^*(t_i) &= \left[ -K_x \{I - KH\} \mid K_z \right] \underline{x}_c(t_i) \\
&- \left[ K_x K \right] \underline{z}(t_i) + K_z \underline{y}_d(t_i)
\end{aligned} \tag{A-50}$$

$$\begin{aligned}
\underline{x}_c(t_{i+1}) &= \left[ \begin{array}{c|c} \left[ \phi - B_d K_x \right] \left[ I - KH \right] & B_d K_z \\ \hline -\left[ C - D_y K_x \right] \left[ I - KH \right] & \left[ I - D_y K_z \right] \end{array} \right] \underline{x}_c(t_i) \\
&+ \left[ \begin{array}{c} \left[ \phi - B_d K_x \right] K \\ \hline -\left[ C - D_y K_x \right] K \end{array} \right] \underline{z}(t_i) + \left[ \begin{array}{c} B_d K_z \\ \hline \left[ I - D_y K_z \right] \end{array} \right] \underline{y}_d(t_i)
\end{aligned} \tag{A-51}$$

Comparing Equations (A-19) and (A-50), and (A-20) and (A-51), it is seen that

$$G_{cx} = \left[ -K_x \{I - KH\} \mid K_z \right] \tag{A-52}$$

$$G_{cz} = \left[ -K_x K \right] \tag{A-53}$$

$$G_{cy} = \left[ K_z \right] \tag{A-54}$$

$$\phi_c = \left[ \begin{array}{c|c} \left[ \phi - B_d K_x \right] \left[ I - KH \right] & B_d K_z \\ \hline - \left[ C - D_y K_x \right] \left[ I - KH \right] & \left[ I - D_y K_z \right] \end{array} \right] \quad (A-55)$$

$$B_{cz} = \left[ \begin{array}{c} \left[ \phi - B_d K_x \right] K \\ - \left[ C - D_y K_x \right] K \end{array} \right] \quad (A-56)$$

$$B_{cy} = \left[ \begin{array}{c} B_d K_z \\ \left[ I - D_y K_z \right] \end{array} \right] \quad (A-57)$$



## APPENDIX B: Modifications and Additions to LQGRP

### A.1 Introduction

This appendix is intended to be used as a supplement to Reference 21, Appendix D, which is the user's guide to a program entitled Linear Quadratic Gaussian Regulator Performance (LQGRP). The program provides the capability of designing and evaluating the performance of continuous-time and sampled-data LQG regulators. In addition, it contains options of enhancing the robustness of both types of regulators via the Doyle and Stein technique.

Three types of changes are made to the original program which are documented herein. The first type involves corrections to errors in the original source code. The second type entails modifications that result in more convenient input and output and a more efficient program. The third type of change generates additions so that the technique of injecting time-correlated noise into the design model while tuning the Kalman filter for robustness purposes can be applied to LQG regulators.

Corrections to the original source code (contained on subsequent pages) are bracketed and marked with a "C". Modifications are bracketed and marked with an "M", and additions are bracketed and marked with an "A".

### A.2 Corrections to LQGRP

The first correction appears on line 6770. The variable name IRF2 replaces IRFM in the original code. IRFM is the dimension of the design model  $F$  matrix, but what should be passed to the subroutine is the

dimension of F minus the number of deterministic states, which is IRF2.

Next, on line 15330, ICBM (the column dimension of the model B matrix) replaces IRFM in the second argument of the subroutine MAT3.

A third correction is made in lines 17120 and 17130. The original source code contains extra arguments in the call to subroutine DDTCON. They do not interfere with execution of the program, but they are not needed and have been deleted in this version.

In line 19430, BM (IDS, IDS) (starting address of the design model B matrix after deterministic states are deleted) replaces BM in the original program. It is intended to pass to the subroutine only the lower portion of the design model B matrix, BM, after the deterministic states (listed first) are deleted. The original program passes the entire B matrix. Making this correction changes the starting address of the array in subroutine DAS2 so that only the desired portion of BM is used.

Lines 19500 and 19510 are inserted into the program to correct an error in line 19520. Originally, RMD appears in place of WM4. RMD is the discrete-time measurement noise covariance matrix. However, the subroutine KFLTR requires a vector containing the diagonal elements of RMD. Thus, the diagonal elements of RMD are stored in the first column of WM4 in lines 15900 and 19510, and WM4 is passed to the subroutine instead of RMD. WM4 is a work-space array not originally being used at that point in the program.

The last correction is made to line 20840. Originally, the line read call MAT3, which calculates the transpose of the desired result. This is corrected by replacing it with CALL MAT3A.

### B.3 Modifications to LQGRP

The matrices and vectors defined in lines 370 through 510 were originally of dimension 5 or 10. All matrices and vectors of dimension 5 are changed to 13, and those of dimension 10 are changed to 26. This change is reflected also in lines 680 through 700. This allows the number of states in the truth or design model to be a maximum number of 13.

The original version of LQGRP had 6 input/output options for entering and printing out truth and design model matrices. The version listed herein has 7. The first 2 options are unchanged. Option 3 now allows the user to change as many elements of a specified matrix as desired by listing the row, column and value of the element being changed. When the user enters a 0, the program exits this option. Options 4 and 5 are also unchanged. Option 6 initially zeros the entire array and then allows the user to enter as many elements as desired by entering the row, column and value of each element. Entering a 0 will exit this option. Option 7 takes the place of option 6 in the previous version. Thus, on a first run through the program, option 6 initializes the array to zero and then only non-zero elements need to be entered. On subsequent runs, changes can be made to any vector or matrix by using option 3.

The changes discussed above require that modifications be made to subroutines MVEC10 and MMAT10, which control the programs input and output. These are listed in lines 12370 to 12380, 12460 to 12530, 12590, 13060, 13130 to 13230 and 13280.

The program stores results of the performance evaluation routines to four data files for plotting. These data files contain the time histories of the following vectors and matrices, defined in Section 2.3:

$\underline{x}_a(t)$ ,  $P_{\underline{x}_a \underline{x}_a}(t)$ ,  $\underline{u}(t)$  and  $P_{uu}(t)$ . The modifications listed in lines 10650 and 10690 calculate and store the square roots of the diagonal elements (standard deviations) of  $P_{\underline{x}_a \underline{x}_a}(t)$  and  $P_{uu}(t)$  rather than storing the diagonal elements (variances). This change also occurs in lines 11570 and 11600. Notice in lines 11560 through 11660 that only the first 7 states of  $\underline{x}_a(t)$  and the upper left (7x7) partition of  $P_{\underline{x}_a \underline{x}_a}(t)$  is stored for plotting. This is an option of the user. Any portion of the vector and matrix may be stored by changing the 7 in lines 10630, 10770, 11550, 11610 and 11650 to the desired value or to IREM (the dimension of the design model F matrix) if the entire vector and matrix are desired.

The four vectors and matrices listed above are printed to the terminal during execution of the program. Currently, only the first 3 states of  $\underline{x}_a(t)$  and the upper-left (3x3) partition of  $P_{\underline{x}_a \underline{x}_a}(t)$  are printed. Any portion may be printed by changing the 3 in lines 10530 and 10570 to a desired value or to IREFM if the entire vector and matrix are desired. Note that all of  $\underline{u}(t)$  and  $P_{uu}(t)$  are stored and listed at the terminal.

The preceding modifications were made to LQGRP to make the program more convenient to use in conjunction with the problem considered in this thesis. None are necessary for execution of the program, as contrasted to the corrections of Section B.2.

#### B.4 Additions to LQGRP

This section describes the additions necessary to the program so that the technique of injecting colored noise into the design model

during filter tuning may be applied. The changes are primarily lines of code inserted into the program. However, a few lines of the original source code are altered slightly, and these will be mentioned.

For a continuous-time system, the input prompt for colored noise is given in line 13870. If colored noise is not desired, the program skips to line 13920 and continues execution. Note that line 13920 is from the original program with the label 2900 inserted in front. If colored noise is desired, the dimension of the model  $F$  matrix, IRFM, is stored in a dummy variable, IHOLD, for later use. Then, in line 13910, subroutine CNOISE is called, which controls the augmentation of shaping filter states with the original design model.

Lines 13950 through 14020 augment zeros to the deterministic controller gain matrix,  $G_c^*$ , if the colored-noise options are executed. Lines 13950 and 13960 are from the original source code, altered by replacing IRFM with IHOLD. Recall, as discussed in Section 4.3, that  $G_c^*$  is calculated using the design model of the unaugmented system, and the dimensionality difference is taken care of by augmenting  $G_c^*$  with the appropriate number of columns of zeros.

For a sampled-data system, the colored noise input prompt should appear between lines 16710 and 16720. It is missing from this listing, but should read

```
WRITE (KOUT,*) 'COLORED INPUT NOISE (Y OR N)>'
```

Lines 16720 through 16760 then control whether the colored-noise routines are executed. Line 16760 is from the original program with a label 25 inserted.

On line 17090, IRFM is again replaced by IHOLD. Lines 17120 and 17130 are also from the original program with an additional argument, IHOLD, in the parameter list. Then, lines 17140 through 17190 check to see if the colored-noise routines were executed. If so,  $G_c^*$  is augmented with the appropriate number of columns of zeros.

In both the continuous-time and sampled-data case, it is desired to use the original design model for controller gain calculations, as mentioned above, in subroutines CDTCON and DDTCON, which calculate deterministic controller gains for the continuous-time and sampled-data case, respectively. These subroutines are essentially unchanged except that IHOLD is substituted for IRFM where indicated in CDTCON (lines 5110 through 5890). In DDTCON, IHOLD is an additional parameter in the argument list and replaces IRFM where indicated (lines 20480 through 21030).

Lines 24770 through 24930 contain subroutine CNOISE which controls what type of shaping filters are augmented to the design model. An input prompt asks if a first- or second-order shaping filter is desired. Subroutine FORDER is called (lines 24940 through 25510) if a first-order shaping filter is desired. If a second-order filter is desired, then subroutine SORDER is called (lines 25520 through 26160). Both subroutines prompt for the shaping filter design parameters described in Chapter IV. Then, the design model is augmented with the shaping filter states and appropriate partitions of the augmented matrices are zeroed. Finally, the dimensions of the design model F and G matrices are altered to reflect the higher dimension after augmenting.

The matrices described above thus modify LQGRP so that shaping filter states can be augmented with the original system design model. It should

be noted that the program will still ask if modification via the Doyle and Stein technique is desired if the colored-noise options are chosen. Either the colored-noise options or the Doyle and Stein options may be chosen, but not both.

Additionally, it is desirable to have the original design and truth models stored on Tape 8 (Ref 21) so that they can be read into the program via input option 18. When the colored-noise options are chosen, the design model is altered. Thus, to make subsequent runs through LQGRP, the original design and truth models must be read back in.





G		000730
G		000740
G	*****MAIN PROGRAM FOLLOWS*****	000750
G		000760
G	----- RAX, ICAX AND DATE OWS COLUMNS OF MATRIX XX	000770
G		000780
G	*****THIS PROGRAM CAN HANDLE UP TO 1000 DIFFERENT COMBINATIONS OF	000790
G	RUNTIME, DELTIM, AND UNSPECIFIED PARAMETERS	000800
	DO 2932 LOG=1,1000	000810
	WRITE(KOUT,11) ' '	000820
11	FORMAT(A1,/)	000830
12	FORMAT(I1)	000840
	WRITE(KOUT,*) 'THIS IS RUN NUMBER ', LOG	000850
	WRITE(KOUT,*) ' '	000860
	WRITE(KOUT,*) ' '	000870
	WRITE(KOUT,*) 'ENTER A DESCRIPTION OF THIS RUN>'	000880
	WRITE(KOUT,*) ' '	000890
	READ(KIN,12) DSCRIPT	000900
	WRITE(KOUT,*) ' '	000910
	WRITE(KOUT,*) ' '	000920
	CALL INPUT(MT, BT, GT, HT, FM, EM, GM, HM, PO, QT, QP, RT, RM, XO,	000930
	1 MUU, MAX, MXU)	000940
	IF (.EQ.0) THEN	000950
	GO TO 2933	000960
	END IF	000970
	CALL RGS(GCSTR, RKFS, GCX, GCY, GCZ, BCY, BCZ, FC, YD,	000980
	1 WM1, WM2, WM3, WM4, WM5, WM6, WM7, WM8, WM9, WM10, WV1, WV2,	000990
	1 WMA, WMB, WMC, WMD, WME, WMF, FT, BT, GT, HT, QT, RT,	001000
	1 FM, EM, GM, HM, OM, RM, XO, PC, WXY, MUU, MXU, FA, GA, QA, GUA,	001010
	1 MXA, PXA, RA, PXVA, GCZA, IRY, IFLGCZ, IFLGSD,	001020
	1 MU, PUMAX, PUMIN, PXTMAX, PXTMIN, MUMAX, MUMIN, MXAMAX, MXAMIN,	001030
	1 PUCUT, PXTOUT, MXTOUT, MUCUT, WV3, WV4)	001040
	WRITE(KOUT,*) ' '	001050
	WRITE(KOUT,*) 'DO YOU WISH TO CALCULATE THE EIGENVALUES OF THE CLOS	001060
	120- COP STATE TRANSITION MATRIX USED IN THE PERFORMANCE ANALYSIS	001070
	1 (APPLICABLE TO BOTH THE CONTINUOUS-TIME AND SAMPLED-DATA CASE)	001080
	1 Y OR N>	001090
	READ(KIN,12) MSG	001100
	IF (MSG.EQ.'Y') THEN	001110
	WRITE(KOUT,*) 'THE CLOSED-LGCP STATE TRANSITION MATRIX EIGENVALUES	001120
	1 ARE...'	001130
	NSAV=NOIM	001140
	NOIM=NOIM2	001150
	NOIM1=NOIM2+1	001160
	CALL MEIG(MMA, WV3, WV4, IRFA, WME)	001170
	NOIM=NSAV	001180
	NOIM1=NSAV+1	001190
	END IF	001200
	WRITE(KOUT,*) ' '	001210
	WRITE(KOUT,*) 'TYPE Y TO PERFORM THE COVARIANCE ANALYSIS, TYPE	001220
	1 N TO SKIP IT>'	001230
	READ(KIN,12) MSG	001240
	IF (MSG.EQ.'N') THEN	001250
	GO TO 2932	001260
	END IF	001270
	CALL PERFALL IRY, IFLGCZ, MXA, GCY, GUA, PXA, PXVA, IFLGSD,	001280
	1 RA, GCZA, YD, WMA, WMB, WME, WMF, WMD, WM1, MUOUT, MXTOUT, PUCUT, PXTOUT,	001290
	1 MXAMIN, MXAMAX, PXTMIN, PXTMAX, MUMIN, MUMAX, PUMAX, MU, WMC,	001300
	1 W/3, WV4)	001310
2932	CONTINUE	001320
2933	WRITE(KOUT,*) 'PROGRAM TERMINATED, NO MORE INPUT DATA'	001330
	END	001340
	* DECK STORED	001350
	SUGGESTION STORED (IWCHRY, ENTIME, DELTIM, IFSTOL, IS12SZ,	001360
	1 IS13SZ, IS14SZ, IS15SZ, STOR12, STOR13, STOR14, STOR15, NOIM)	001370
G	STORE DATA TO PERMANENT FILE	001380

```

C THIS SUBROUTINE STORES PLOT DATA TO LOCAL FILES. TAPE12,TAPE13. 00139
C TAPE14-TAPE15 FOR PERMANENT STORAGE. ISKYSZ IS THE SIZE OF THE 00140
C DATA ARRAYS,STORXX. NOIM IS THE CALLING PROGRAM DIMENSION OF 00141
C THE ARRAYS. NOTE THAT (RUNTIME+DELTIME)*2 GIVES THE TOTAL 00142
C NUMBER OF DATA POINTS IN THE RUN. THE LAST ENTRY IN EACH 00143
C DATA FILE IS THE SCALE FACTOR FOR THE DATA ON THE FILE. THE 00144
C NEXT TO LAST ENTRY IS THE MINIMUM VALUE OF THE DATA IN THE FILE 00145
C DIMENSION STOR12(NOIM6),STOR13(NOIM6),STOR14(NOIM6),STOR15(NOIM6) 00146
COMMON /INOU/ KIN,KOUT,KPUNCH 00147
IF (IFSTCL.EQ.0) THEN 00148
C PUT A RUN HEADER ON THE TAPE 00149
OPEN(UNIT=12,ERR=10,FILE='TAPE12',RECL=80) 00150
OPEN(UNIT=13,ERR=10,FILE='TAPE13',RECL=80) 00151
OPEN(UNIT=14,ERR=10,FILE='TAPE14',RECL=80) 00152
OPEN(UNIT=15,ERR=10,FILE='TAPE15',RECL=80) 00153
WRITE(12,101) INCHRN,RTIME,DELTIME,IS12SZ 00154
WRITE(13,101) INCHRN,RTIME,DELTIME,IS13SZ 00155
WRITE(14,101) INCHRN,RTIME,DELTIME,IS14SZ 00156
WRITE(15,101) INCHRN,RTIME,DELTIME,IS15SZ 00157
101 FORMAT(' ',I11,1X,2E19.6,I13) 00158
END IF 00159
IF (IFSTCL.LT.0) THEN 00160
C CLOSE FILES AND PUT END OF FILE MARKER JM THEN. 00161
CLOSE(12,ERR=10) 00162
CLOSE(13,ERR=10) 00163
CLOSE(14,ERR=10) 00164
CLOSE(15,ERR=10) 00165
RETURN 00166
END IF 00167
WRITE(12,102) (STOR12(I),I=1,IS12SZ) 00168
WRITE(13,102) (STOR13(I),I=1,IS13SZ) 00169
WRITE(14,102) (STOR14(I),I=1,IS14SZ) 00170
WRITE(15,102) (STOR15(I),I=1,IS15SZ) 00171
RETURN 00172
10 WRITE(KOUT,*) 'AN ERROR HAS OCCURRED IN THE STORED ROUTINE' 00173
102 FORMAT(13(' ',E19.6,1),/) 00174
END 00175
* DECK INPUT 00176
SUBROUTINE INPUTM(FT,BT,GT,HT,FP,BH,GH,WH,PO,QT,QV,RT,RH,XC, 00177
INUU,WXX,WXU) 00178
CHARACTER MSG*60,MSG1*50 00179
REAL FT(NOIM,NOIM),BT(NOIM,NOIM),GT(NOIM,NOIM),HT(NOIM,NOIM), 00180
IMH(NOIM,NOIM), 00181
IQM(NOIM,NOIM),RM(NOIM,NOIM),XC(NOIM), 00182
1 WUU(NOIM,NOIM),WXX(NOIM,NOIM),FM(NOIM,NOIM),BH(NOIM,NOIM), 00183
1 GH(NOIM,NOIM),WXU(NOIM,NOIM), 00184
1 PO(NOIM,NOIM),QT(NOIM,NOIM),RT(NOIM,NOIM),COM1(1),COM2(1) 00185
COMMON /MAIN4/NOIM2,NOIM3 00186
COMMON /MAIN2/COM2 00187
COMMON /MAIN1/NOIM,NOIM1,COM1 00188
COMMON /INOU/ KIN,KOUT,KPUNCH 00189
COMMON /MAUN5/ MSG 00190
COMMON /MAIN6/ ICBT,ICBM,ICFA,ICGA,LCGM,ICGT,ICQA,IRFA,IRFM,IRFT, 00191
1 IRMT,IRQA,IO,LOG,IRMP,NUMOTS 00192
MSG1='*****' 00193
ICFT=1-FT,ICFM=IRFM,IRBT=1-BT,IRBM=IRFM,IRGT=1-GT,IRGM=IRFM 00194
IRX=IRF,IRQ=ICQ=ICGM,IRR=ICR=IRMT,IRWXX=ICWXX=IRFM, 00195
IRWUU=ICWUU=ICBM,IRHT=IRFT,IRCHM=IRFM,IRGCZ=ICBT,IRGCZ=IRMT, 00196
IO IS A INPUT ROUTINE PARAMETER--1=READ,2=READ+PRINT, 00197
3= PRINT ONLY, 4=PUNCH 00198
C 00199
NSA1=NOIM1 00200
NOIM1=NOIM 00201
IF (LOG.EQ.1) THEN 00202
WRITE(KOUT,*)MSG1 00203
WRITE(KOUT,*)'THE I/O OPTIONS ARE 0,1,2,3,4,5,6,.....' 00204

```

WRITE(KOUT,*) '1-READ ENTIRE ARRAY/VECTOR, 2-READ AND PRINT'	002150
WRITE(KOUT,*) 'ENTIRE ARRAY/VECTOR, 3-READ AND 4-READ AND PRINT'	002160
WRITE(KOUT,*) 'SELECTED ARRAY/VECTOR ELEMENTS, 5-PRINT ENTIRE'	002170
WRITE(KOUT,*) 'ARRAY/VECTOR, 6 TO PRE-ZERO ARRAY ELEMENTS THEN'	002180
WRITE(KOUT,*) 'ENTER SELECTED ELEMENTS, 7 OR GREATER IF NO'	002190
WRITE(KOUT,*) 'MORE INPUT.'	002200
WRITE(KOUT,*) MSG1	002210
WRITE(KOUT,*) 'SELECT WHICH MATRIX YOU WISH TO ENTER.'	002220
WRITE(KOUT,*) 'BY ENTERING THE APPROPRIATE NUMBER, 1-FT, 2-BT'	002230
WRITE(KOUT,*) '3-GT, 4-HT, 5-FM, 6-BM, 7-GM, 8-HM, 9-PO, 10-OT, 11-RT, >'	002240
WRITE(KOUT,*) '12-GH, 13-RH, 14-XO, 15-XU, 16-WX, 17-WU, 18-EQUATE ALL	002250
1	002260
WRITE(KOUT,*) 'CONTROLLER MODEL MATRICES TO THEIR'	002270
WRITE(KOUT,*) 'TRUTH MODEL COUNTERPARTS, 19----- NO MORE DATA'	002280
WRITE(KOUT,*) 'ENTRIES TO BE MADE, 20--STORE ALL MATRICES ON TAPE7'	002290
WRITE(KOUT,*) '21- READ ALL MATRICES FROM TAPE3'	002300
WRITE(KOUT,*) MSG1	002310
WRITE(KOUT,*) MSG1	002320
WRITE(KOUT,*) ' '	002330
WRITE(KOUT,*) 'FOR SAMPLED DATA MEASUREMENTS, ENTER EITHER A CONTINUO	002340
US RM TO USE TO APPROXIMATE THE DISCRETE TIME RMD(RMD=RM/SAMPLE	002350
1 TIME) OR ENTER THE DISCRETE TIME RMD'	002360
WRITE(KOUT,*) MSG1	002370
WRITE(KOUT,*) MSG1	002380
END IF	002390
DO 992 INPUT=1,1,11	002400
WRITE(KOUT,33333) ' '	002410
33333 FORMAT(A13, /)	002420
WRITE(KOUT,*) 'ENTER CODE FOR WHICH ARRAY/VECTOR TO BE INPUT>'	002430
READ(KIN,*) INCHMA	002440
GO TO(1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21) INCHMA	002450
C	002460
C***TRUTH MODEL INPUT	002470
1 WRITE(KOUT,*) 'ENTER-I/O OPTION, FT MATRIX SIZE>'	002480
READ(KIN,*,END=27) IO,IRFT	002490
MSG='TRUTH MODEL F MATRIX ENTRIES'	002500
CALL MMATIO (FT,IRFT,IRFT,IC,KIN,KOUT,NOIM,NOIM1)	002510
IF (IO.EQ.1) THEN	002520
RETURN	002530
END IF	002540
GO TO 992	002550
2 WRITE(KOUT,*) 'ENTER-I/O OPTIONS, COLUMN SIZE OF BT>'	002560
READ(KIN,*,END=27) IO,ICBT	002570
MSG='TRUTH MODEL B MATRIX ENTRIES'	002580
CALL MMATIO (BT,IRFT,ICBT,IC,KIN,KOUT,NOIM,NOIM1)	002590
IF (IO.EQ.2) THEN	002600
RETURN	002610
END IF	002620
GO TO 992	002630
3 WRITE(KOUT,*) 'ENTER-I/O OPTION, COLUMN SIZE OF GT>'	002640
READ(KIN,*,END=27) IO,ICGT	002650
MSG='TRUTH MODEL G MATRIX ENTRIES'	002660
CALL MMATIO (GT,IRFT,ICGT,IC,KIN,KOUT,NOIM,NOIM1)	002670
IF (IO.EQ.3) THEN	002680
RETURN	002690
END IF	002700
GO TO 992	002710
4 WRITE(KOUT,*) 'ENTER-I/O OPTION, ROW SIZE OF HT>'	002720
READ(KIN,*,END=27) IO,IRHT	002730
MSG='H MATRIX ENTRIES'	002740
CALL MMATIO (HT,IRHT,IRFT,IC,KIN,KOUT,NOIM,NOIM1)	002750
IF (IO.EQ.4) THEN	002760
RETURN	002770
END IF	002780
GO TO 992	002790
C***INPUT CONTROLLER MODEL	002800

5	WRITE(KOUT,*)'ENTER-I/O OPTION, FM MATRIX SIZE>'	002710
	READ(KIN,*,END=27)IO,IRFM	002720
	MSG='CONTROLLER MODEL F MATRIX ENTRIES'	002730
	WRITE(KOUT,*)'ENTER THE NUMBER OF DETERMINISTIC STATES IN THIS MODEL>'	002740
	READ(KIN,*,END=27)NLOSTS	002750
	CALL MMATIO (FM,IRFM,IRFM,IC,KIN,KOUT,NDIM,NDIM1)	002770
	IF (IO.EQ.0) THEN	002780
	RETURN	002790
	END IF	002800
	GO TO 332	002810
6	WRITE(KOUT,*)'ENTER-I/O OPTION, COLUMN SIZE OF GM>'	002820
	READ(KIN,*,END=27)IO,ICGM	002830
	MSG='CONTROLLER MODEL G MATRIX ENTRIES'	002840
	CALL MMATIO (GM,IRFM,ICGM,IC,KIN,KOUT,NDIM,NDIM1)	002850
	IF (IO.EQ.0) THEN	002860
	RETURN	002870
	END IF	002880
	GO TO 332	002890
7	WRITE(KOUT,*)'ENTER-I/O OPTION, COLUMN SIZE OF GN>'	002900
	READ(KIN,*,END=27)IO,ICGN	002910
	MSG='CONTROLLER MODEL G MATRIX ENTRIES'	002920
	CALL MMATIO (GN,IRFM,ICGN,IC,KIN,KOUT,NDIM,NDIM1)	002930
	IF (IO.EQ.0) THEN	002940
	RETURN	002950
	END IF	002960
	GO TO 332	002970
8	WRITE(KOUT,*)'ENTER I/O OPTION, ROW SIZE OF HM>'	002980
	READ(KIN,*,END=27)IO,IRHM	002990
	MSG='THE CONTROLLER MODEL MEASUREMENT MATRIX, HM, IS'	003000
	CALL MMATIO (HM,IRFM,IRFM,IO,KIN,KOUT,NDIM,NDIM1)	003010
	IF (IO.EQ.0) THEN	003020
	RETURN	003030
	END IF	003040
	GO TO 332	003050
9	WRITE(KOUT,*)'IT MUST BE ENTERED THRU OPTION 1 PRIOR TO USING THIS	003060
	1 OPTION.DO YOU WISH TO ABORT THIS OPTION, Y OR N>'	003070
	READ(KIN,37,END=27)MSG	003080
	IF (MSG.EQ.'Y') THEN	003090
	GO TO 332	003100
	END IF	003110
	WRITE(KOUT,*)'ENTER I/O OPTION,IRFT IS ASSUMED SIZE OF PO>'	003120
	READ(KIN,*,END=27)IO	003130
	MSG='THE INITIAL COVARIANCE MATRIX,PO, IS'	003140
	CALL MMATIO (PO,IRFT,IRFT,IC,KIN,KOUT,NDIM,NDIM1)	003150
	IF (IO.EQ.0) THEN	003160
	RETURN	003170
	END IF	003180
	GO TO 332	003190
10	WRITE(KOUT,*)'ENTER I/O OPTION, ICGT IS ASSUMED SIZE OF QT>'	003200
	READ(KIN,*,END=27)IO	003210
	MSG='THE INPUT NOISE STRENGTH MATRIX QT IS'	003220
	CALL MMATIO (QT,ICGT,ICGT,IC,KIN,KOUT,NDIM,NDIM1)	003230
	IF (IO.EQ.0) THEN	003240
	RETURN	003250
	END IF	003260
	GO TO 332	003270
11	WRITE(KOUT,*)'ENTER I/O OPTION,IRHT IS ASSUMED SIZE OF RT>'	003280
	READ(KIN,*,END=27)IO	003290
	MSG='THE MEASUREMENT NOISE STRENGTH MATRIX RT IS'	003300
	CALL MMATIO (RT,IRHT,IRHT,IC,KIN,KOUT,NDIM,NDIM1)	003310
	IF (IO.EQ.0) THEN	003320
	RETURN	003330
	END IF	003340
	GO TO 332	003350
12	WRITE(KOUT,*)'ENTER I/O OPTION,ICGM IS ASSUMED SIZE OF GN>'	003360

	PE1)(KIN,*,END=27)IC	003370
	MSG='CONTROLLER MODEL INPUT NOISE STRENGTH MATRIX, RM'	003371
	CALL MMATIO(RM,ICGM,ICGM,IC,KIN,KOUT,NOIM,NOIM1)	003372
	IF (IO.EQ.3) THEN	003373
	RETURN	003374
	END IF	003375
	GO TO 332	003376
13	WRITE(KOUT,*)'ENTER I/O OPTION, IRHM IS ASSUMED SIZE OF RM>'	003377
	READ(KIN,*,END=27)IC	003378
	MSG='CONTROLLER MODEL MEASUREMENT NOISE STRENGTH MATRIX, RM'	003379
	CALL MMATIO(RM,IRHM,IRHM,IC,KIN,KOUT,NOIM,NOIM1)	003380
	IF (IO.EQ.3) THEN	003381
	RETURN	003382
	END IF	003383
	GO TO 332	003384
14	WRITE(KOUT,*)'FT MUST BE ENTERED THRU OPTION 1 PRIOR TO USING THIS	003385
	1 OPTION. DO YOU WISH TO ABORT THIS OPTION, Y OR N>'	003386
	READ(KIN,37,END=27)MSG	003387
	IF (MSG.EQ.'Y') THEN	003388
	GO TO 332	003389
	END IF	003390
	WRITE(KOUT,*)'ENTER I/O OPTION,IRFT IS ASSUMED SIZE OF XO>'	003391
	READ(KIN,*,END=27)IC	003392
	MSG='THE INITIAL STATE VECTOR, XO, IS'	003393
	CALL MVEGEO(XO,IRFT,IC,KIN,KOUT,NOIM)	003394
	IF (IO.EQ.3) THEN	003395
	RETURN	003396
	END IF	003397
	GO TO 332	003398
15	WRITE(KOUT,*)'WM MUST BE ENTERED THRU OPTION 6 OR 17 PRIOR TO USING	003399
	16 THIS OPTION. DO YOU WISH TO ABORT THIS OPTION, Y OR N>'	003400
	READ(KIN,37,END=27)MSG	003401
	IF (MSG.EQ.'Y') THEN	003402
	GO TO 332	003403
	END IF	003404
	WRITE(KOUT,*)'ENTER I/O OPTION,ICWM IS ASSUMED SIZE OF WUU>'	003405
	READ(KIN,*,END=27)IC	003406
	MSG='THE CONTROL FUNCTION COST WEIGHTING MATRIX, WUU'	003407
	CALL MMATIO(WUU,ICWM,ICWM,IC,KIN,KOUT,NOIM,NOIM1)	003408
	IF (IO.EQ.3) THEN	003409
	RETURN	003410
	END IF	003411
	GO TO 332	003412
16	WRITE(KOUT,*)'FM MUST BE ENTERED THRU OPTION 5 OR 17 PRIOR TO USING	003413
	16 THIS OPTION. DO YOU WISH TO ABORT THIS OPTION, Y OR N>'	003414
	READ(KIN,37,END=27)MSG	003415
	IF (MSG.EQ.'Y') THEN	003416
	GO TO 332	003417
	END IF	003418
	WRITE(KOUT,*)'ENTER I/O OPTION,IRFM IS ASSUMED SIZE OF WXX>'	003419
	READ(KIN,*,END=27)IC	003420
	MSG='THE STATE COST WEIGHTING MATRIX, WXX'	003421
	CALL MMATIO(WXX,IRFM,IRFM,IC,KIN,KOUT,NOIM,NOIM1)	003422
	IF (IO.EQ.3) THEN	003423
	RETURN	003424
	END IF	003425
	GO TO 332	003426
18	WRITE(KOUT,*)'ALL CONTROLLER MODEL MATRICES HAVE BEEN SET EQUAL TO	003427
	1 THEIR TRUTH MODEL COUNTERPARTS'	003428
	WRITE(KOUT,*)'FT,BT,GT,HT,QT,RT MUST BE ENTERED PRIOR TO USING THIS	003429
	18 OPTION. THE NUMBER OF DETERMINISTIC STATES MUST BE ENTERED IN	003430
	1 THIS OPTION(FOR CONTROLLER MODEL).'	003431
	WRITE(KOUT,*)'DO YOU WISH TO ABORT THIS OPTION,Y OR N>'	003432
	READ(KIN,37,END=27)MSG	003433
	IF (MSG.EQ.'Y') THEN	003434
	GO TO 332	003435

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END IF
WRITE(KOUT,*) 'ENTER THE NUMBER OF DETERMINISTIC STATES IN THE COST ROLLER MODEL'
READ(KIN,*,END=27) NUMOTS
IRFM=IRFT
97  FORMAT(11)
    ICBM=ICBT
    ICGM=ICGT
    IRHM=IRHT
    CALL EQUATE(FM,FT,IRFT,IRFT)
    CALL EQUATE(BM,BT,IRBT,ICBT)
    CALL EQUATE(GM,GT,IRGT,ICGT)
    CALL EQUATE(HM,HT,IRHT,IRFT)
    CALL EQUATE(QM,QT,ICGT,ICGT)
    CALL EQUATE(RM,RT,IRHT,IRHT)
    GO TO 332
17  WRITE(KOUT,*) 'NOTE THAT FM AND BM MUST BE ENTERED THROUGH APPROPRIATE
    LATE OPTIONS PRIOR TO EXECUTING THIS OPTION. DO YOU WISH TO ABORT THIS
    THIS OPTION, Y OR N?'
    READ(KIN,97,END=27) MSG
    IF (MSG.EQ.'Y') THEN
        GO TO 332
    END IF
    WRITE(KOUT,*) 'ENTER I/O OPT.CN, IRFM X ICBM ASSSUMED SIZE NXNU'
    READ(KIN,*,END=27) IC
    MSG='THE CROSS (STATE-CONTROL) COST WEIGHTING MATRIX, WAXU'
    CALL MMATIO(WAXU,IRFM,ICBM,IC,KIN,KOUT,NOIN,NOIN1)
    IF (IO.EQ.0) THEN
        RETURN
    END IF
    GO TO 332
20  WRITE(KOUT,*) ' THIS OPTION STORES ALL MATRICES CN TO TAPE7, DO YOU
    WISH TO ABORT THIS OPTION, Y OR N?'
    READ(KIN,97) MSG
    IF (MSG.EQ.'Y') THEN
        GO TO 332
    END IF
    WRITE(7,*) IRFT,ICBT,ICGT,IRHT,IRFM,ICBM,ICGM,IRHM,NUMOTS
    WRITE(7,*) ((FT(I,J),J=1,IRFT),I=1,IRFT), ((BT(I,J),J=1,ICBT),I=1,IRBT),
    ((GT(I,J),J=1,ICGT),I=1,IRGT), ((HT(I,J),J=1,IRHT),I=1,IRHT),
    ((FM(I,J),J=1,IRFM),I=1,IRFM), ((BM(I,J),J=1,ICBM),I=1,IRBM),
    WRITE(7,*) ((GM(I,J),J=1,ICGM),I=1,IRGM), ((HM(I,J),J=1,IRHM),I=1,IRHM),
    ((PM(I,J),J=1,IRPT),I=1,IRPT), ((QM(I,J),J=1,ICQT),I=1,IRQT),
    ((RM(I,J),J=1,IRRT),I=1,IRRT), ((WXX(I,J),J=1,IRFM),I=1,IRFM),
    WRITE(7,*) ((QT(I,J),J=1,ICGT),I=1,ICGT), ((RT(I,J),J=1,IRHT),I=1,IRHT),
    ((QM(I,J),J=1,ICGM),I=1,ICGM), ((RM(I,J),J=1,IRHM),I=1,IRHM),
    WRITE(7,*) (WAXU(I,J),J=1,ICM),I=1,IRFM)
    GO TO 332
21  WRITE(KOUT,*) ' THIS OPTION READS ALL MATRICES FROM TAPES, DO YOU
    WISH TO ABORT THIS OPTION, Y OR N?'
    READ(KIN,97) MSG
    IF (MSG.EQ.'Y') THEN
        GO TO 332
    END IF
    WRITE(KOUT,*) 'DO YOU WISH TO REWIND TAPES BEFORE THE READ, Y OR N?'
    READ(KIN,97) MSG
    IF (MSG.EQ.'Y') THEN
        REWIND(9)
    END IF
    READ(8,*,END=10033) IRFT,ICBT,ICGT,IRHT,IRFM,ICBM,ICGM,IRHM,NUMOTS
    READ(9,*,END=10033) ((FT(I,J),J=1,IRFT),I=1,IRFT), ((BT(I,J),J=1,ICBT),I=1,IRBT),
    ((GT(I,J),J=1,ICGT),I=1,IRGT), ((HT(I,J),J=1,IRHT),I=1,IRHT),
    ((FM(I,J),J=1,IRFM),I=1,IRFM), ((BM(I,J),J=1,ICBM),I=1,IRBM),
    READ(9,*,END=10033) ((GM(I,J),J=1,ICGM),I=1,IRGM), ((HM(I,J),J=1,IRHM),I=1,IRHM),
    ((PM(I,J),J=1,IRPT),I=1,IRPT), ((QM(I,J),J=1,ICQT),I=1,IRQT),
    ((RM(I,J),J=1,IRRT),I=1,IRRT), ((WXX(I,J),J=1,IRFM),I=1,IRFM)

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C
C TRANSFORM SYSTEM SO THAT WKU NOT=0 CAN STILL BE HANDLED BY
C KLEINMAN ROUTINES. SEE KRAKERNAAK AND SI/AN'S BOOK, PAGE 322
CALL PRINT(4M1,WUU,ICBM,GOSTR,WKU,IMOLD,3M,W2,FPRIM,WXX,WXXPRM,
: FM)
C NOW HAVE FPRIM, WXXPRM CAN USE RICCATI SOLVER FOR KCSSPM
MT=1
CALL TRANS2(IMOLD,ICBM,3M,W2)
C W2=BMT ICBM X IMOLD
CALL EQUATE(WM1,WUU,ICBM,ICBM)
C GMINV DESTROYS THE CALLING ARRAY
CALL GMINV(ICBM,ICBM,W1,W3,MR,MT)
C W3=WUUI ICBM X ICBM-----MR IS AN ERROR INDICATOR
IF (MR.NE.ICBM) THEN
PRINT*, 'AN ERROR OCCURRED IN INVERTING WUU, MR=',MR,'ICBM=',ICBM
END IF
CALL EQUATE(4M1,W3,ICBM,ICBM)
C W1=WUUI SAVE FOR LATER COMPUTATIONS
CALL MAT1(WM3,W2,ICBM,ICBM,IMOLD,W4)
C W4=(WUUI)(BMT) ICBM X IMOLD
C NOW CALL RICCATI EQUATION SOLVER
CALL MAT1(3M,4M,IMOLD,ICBM,IMOLD,W3)
C W3=3M(WUUI)(BMT) IMOLD X IMOLD
CALL MRIC(IMOLD,FPRIM,W3,WXXPRM,W2,W4)
C W2=KCSSPM IMOLD X IMOLD
C W3=FM-BM(WUUI)(BMT)(KCSSPM)---I DONT USE THIS RESULT
MSG='KCSSPM FOR THE DETERMINISTIC CONTROLLER IS'
IC=5
NSAV=NOIM1
NOIM1=NOIM
CALL MMATIO(WM2,IMOLD,IMOLD,IC,KIN,KOUT,NOIM,NOIM1)
C NOW CALCULATE OPTIMAL GAIN MATRIX GCSTAR. NOTE I NEED THE
C NEGATIVE OF GCSTAR FOR THE CONTROL LAW GENERATION FROM AN LQG
C CONTROLLER, AND THIS WILL BE THE GCX REQUIRED IN THE PERFORMANCE
C ANALYSIS ROUTINE
NOIM1=NSAV
C NOW HAVE KCSSPM, CALCULATE GCSTR=WUUI(BMT(KCSSPM)+W4UT)
C RECALL WUUI IN W1
C1=1.0
CALL MAT4(3M,W2,ICBM,IMOLD,IMOLD,W4)
CALL TRANS2(IMOLD,ICBM,WKU,W3)
CALL MA001(ICBM,IMOLD,W4,W3,W2,C1)
CALL MAT1(WM1,W2,ICBM,ICBM,IMOLD,GCSTR)
NSAV=NOIM1
NOIM1=NOIM
IC=5
C GCSTR ICBM X IMOLD
C
MSG='THE OPTIMAL STEADY STATE FEEDBACK GAIN MATRIX, GCSTR'
CALL MMATIO(GCSTR,ICBM,IMOLD,IO,KIN,KOUT,NOIM,NOIM1)
NOIM1=NSAV
C
END
* DECK CKFTR
SUBROUTINE CKFTR(FM,GM,P,MM,NUMOTS,RKFSS,G,W1,W2,
: WM3,W4,W5,W6,IRFM,IRHM,ICGM,F2,H2,3M,ICBM)
C CALLING PROGRAM MUST SUPPLY EIGHT WORK SPACE ARRAYS
C
CTHIS ROUTINE CALCULATES THE KALMAN FILTER GAINS WHEN
C GIVEN THE FM,GM, GM AND 4 MATRICES AND THE NUMBER OF
C DETERMINISTIC STATES. THE CONTROLLER MODEL MUST BE
C SPECIFIED SUCH THAT ALL THE DETERMINISTIC STATES APPEAR
C FIRST AND TOGETHER, THAT IS
CO/OT(X1,X2,...,XK,XL,XM,...,XN)T=F1 1 31 -
C (X) + (U) + (W)

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C                                     M      F2      32      G2      006110
C WHERE X1 THROUGH XK ARE THE DETERMINISTIC STATES AND THE 006120
C REMAINING STATES ARE STOCHASTIC. F1 IS K X K, AND F2 IS N-K X 006130
C N-K, AND G1,G2,AND G2 ARE PARTITIONED ACCORDINGLY. 006140
C THIS ROUTINE FIRST STRIPS OFF THE DETERMINISTIC STATES THEN COMPUTES 006150
C AND RETURNS KALMAN FILTER GAINS FOR THE REMAINING STATES. 006160
C THE KALMAN FILTER GAINS FOR THE DETERMINISTIC STATES ARE SET TO ZERO 006170
C AND THE KALMAN FILTER GAIN THAT IS RETURNED IS 006180
C 006190
C WHERE THE DIMENSION OF THE ZERO VECTOR IS K AND THE RKFS IS THE 006200
C STEADY STATE KALMAN FILTER GAIN MATRIX FOR THE N-K STOCHASTIC STATES. 006210
C THIS AUGMENTED MATRIX IS RETURNED IN RKFS 006220
C ALSO NOTE THAT IN ORDER TO GENERATE THE KALMAN FILTER, ONLY 006230
C MEASUREMENTS OF STACHASTIC STATES ARE NEEDED SO THE H MATRIX IS 006240
C REDUCED ACCORDINGLY. 006250
      CHARACTER MSG*60,MSG1*1
      DIMENSION F2(NOIM,NOIM),M2(NOIM,NOIM),FM(NOIM,NOIM),GM(NOIM,NOIM), 006260
      1 R(NOIM,NOIM),PM(NOIM,NOIM),WM1(NOIM,NOIM),Q(NOIM,NOIM), 006270
      1 WM2(NOIM,NOIM),WM3(NOIM,NOIM),WM4(NOIM,NOIM),WM5(NOIM,NOIM) 006280
      1 ,WM6(NOIM,NOIM) 006290
      REAL RKFS(NOIM,NOIM),BM(NOIM,NOIM) 006300
      DIMENSION COM1(1),COM2(1) 006310
      COMMON /MAIN2/COM2 006320
      COMMON /MAIN1/NOIM,NOIM1,COM1 006330
      COMMON /INOUT/ KIN,KOUT,KPUNCH 006340
      COMMON /MAUNS/ MSG 006350
C 006360
C ***KALMAN FILTER STEADY STATE GAIN----MODULE # 2 006370
C 006380
C 006390
C 006400
C 006410
C 006420
C 006430
C 006440
C 006450
C 006460
C 006470
C 006480
C 006490
C 006500
C 006510
C 006520
C 006530
C 006540
C 006550
C 006560
C 006570
C 006580
C 006590
C 006600
C 006610
C 006620
C 006630
C 006640
C 006650
C 006660

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2115 WM(IJ,J)=BM(I,J)                                006670
C WM= #32 IRF2 X ICBM                                006680
      IR42=1-44                                        006690
      CALL MAT3(IRF2,ICBM,WM1,0,WM2)                  006700
C WM2=GM(7)(GYT) IRF2 X IRF2 --USED AS 'Q' IN KLIENMAN RICCATI ROUTINE 006710
      WRITE(KOUT,*) 'QJ YOL WISH TO MODIFY Q BY THE DOYLE AND STEIN TECHNI 006720
      1QUE, Y OR N>'
      READ (KIN,11)MSG1                                006730
11    FORMAT(A1)                                         006740
      IF (MSG1.EQ.'Y') THEN                             006750
        CALL DAS1(A42,WM4,WM1,ICBM,WM3,IRF2)          006760
        END IF                                           006770
      MT=1                                               006780
      CALL EQUATE(WM1,R,IRH2,IRH2)                     006790
C GHINV DESTROYS THE CALLING ARRAY                     006800
      CALL GHINV(IRH2,IRH2,WM1,WM3,MR,MT)              006810
      IF (MR.NE.IRH2) THEN                              006820
        WRITE(KOUT,*) 'MR=',MR,' IRH2=',IRH2          006830
        WRITE(KOUT,*) 'R-INVERSE IS FOULED UP'        006840
        END IF                                           006850
C WM3= RI IRH2 X IRH2                                  006860
      CALL TRANS2(IRH2,IRF2,H2,WM4)                   006870
C WM4= H2T IRF2 X IRH2                                 006880
      CALL MAT1(WM4,WM3,IRF2,IRH2,IRH2,WM5)           006890
C WM5= H2T(RI) IRF2 X IRH2                            006900
      CALL MAT1(WM5,H2,IRF2,IRH2,IRF2,WM3)            006910
C WM3= H2T(RI)(H2) IRF2 X IRF2                       006920
      CALL MRIC(IRF2,F2,WM3,WM2,WME,WM4)              006930
C NOW CALL RICCATI EQUATION SOLVER TO GET PMSS        006940
C WM6=PMSS IRF2 X IRF2                                006950
      CALL MAT1(WM6,WM5,IRF2,IRF2,IRH2,WM1)           006960
C WM1=RKFSS IRF2 X IRH2                               006970
      IO=5                                              006980
C FORM RKFSS WITH ZEROS ADDED FOR DETER. STATES.     006990
      PRINT*, 'NUMOTS=',NUMOTS                        007000
      IF (NUMOTS.NE.1) THEN                             007010
        DO 2119 J=1,IRHM                               007020
        DO 2118 I=1,NUMOTS                             007030
2118   RKFSS(I,J)=0
        DO 2119 I=IDS,IRFM                             007040
        II=I-NUMOTS
2119   RKFSS(I,J)=WM1(II,J)
        ELSE                                           007050
        CALL EQUATE(RKFSS,WM1,IRFP,IRHM)              007060
        END IF                                           007070
      MSG='STEADY STATE KILMAN FILTER GAIN MATRIX,RKFSS' 007080
      CALL MNATIO(RKFSS,IRFM,IRHM,IC,KIN,KOUT,NOIP,NOIN) 007090
      NUMOTS=NUMSAV                                    007100
C                                                       007110
      ENO                                              007120
      * DECK FRMAUG                                     007130
      SUBROUTINE FRMAUG(Q,R,FT,ST,GCZ,MT,GCX,BCZ,FC,GCY,BCY,GT,XO,PC, 007140
      IFA,GA,QA,GUA,WM1,WM2,WYA,WM3,WM4,WM5,WMF,      007150
      IMXA,PXA,RA,PXVA,GCZA,IRY,IFLGCZ,IFLGSD)        007160
C THIS ROUTINE FORMS A SET OF AUGMENTED MATRICES NEEDED BY THE 007170
C PERFORMANCE ANALYSIS ROUTINES                      007180
      DIMENSION Q(NOIP,NOIN),R(NOIP,NOIN),FT(NOIN,NOIN) 007190
      1,ST(NOIN,NOIN),GCZ(NOIN,NOIN),MT(NOIN,NOIN),GCX(NOIN,NOIN), 007200
      1 GCZ(NOIN,NOIN),FC(NOIN,NOIP),BCY(NOIN,NOIN),GT(NOIN,NOIN), 007210
      1 XO(NOIN),PO(NOIN,NOIN),WM1(NOIN,NOIN),WM2(NOIN,NOIN) 007220
      DIMENSION FA(NOIP2,NOIP2),BA(NOIP2,NOIP2),GA(NOIP2,NOIP2), 007230
      1 QA(NOIP2,NOIP2),WMA(NOIP2,NOIP2),WMB(NOIP2,NOIP2), 007240
      1 WMC(NOIP2,NOIP2),WMO(NOIP2,NOIP2),WME(NOIP2,NOIP2),WMF 007250
      1 (NOIN,NOIN),GCY(NOIN,NOIP),PXA(NOIP2,NOIP2),PA(NOIP2,NOIP2), 007260
      1 GCZA(NOIP2,NOIP2),PXVA(NOIP2,NOIP2),GUA(NOIP2,NOIP2) 007270
      REAL WYA(NOIP2)

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INTEGER IFLGSD	007330
CHARACTER MSG*50	007340
DIMENSION COM1(1),COM2(1)	007350
COMMON /MAIN1/NOI42,NOI43	007360
COMMON /MAIN2/COM2	007370
COMMON /MAIN1/NOI4,NOI41,COM1	007380
COMMON /INOUT/ KIN,KOUT,KPUNCH	007390
COMMON /MAUN5/ MSG	007400
COMMON /MAIN6/ IC3T,IC3N,ICFA,ICGA,ICGN,ICGT,ICQA,IRFA,IRFM,IRFT,	007410
IRHT,IRQA,IC,LQG,IRHM,NUMOTS	007420
WRITE(KOUT,*) ' ENTER A 5 IF YOU WANT ALL THE AUGMENTED MATRICES PR	007430
1INTEC OUT, A 7 FOR NO MATRICES TO BE PRINTED>'	007440
READ(KIN,*)IO	007450
IRFA=IRFT+IRFM	007460
NSAV1=NOIH	007470
NSAV2=NOIM1	007480
NSAV3=NOIM2	007490
NSAV4=NOI43	007500
C	007510
C***FORM AUGMENTED MATRICES THAT ARE REQUIRED WHEN FORMING XA	007520
C WA=(NT VT)T IMPLIES THAT GA= 0 J	007530
C J R	007540
C	007550
C FORM QA IRQA X IRQA, IRQA=IRHT+ICGN	007560
C FOR EQUIVALENT DISCRETE TIME SYSTEMS IRQA= IRFT+IRHT	007570
IF (IFLGSD.EQ.3) THEN	007580
IRQ=ICGT	007590
ELSE	007600
IRQ=IRFT	007610
END IF	007620
DO 2703 I=1,IRQ	007630
DO 2703 J=1,IRHT	007640
2703 WM1(I,J)=J	007650
DO 2704 I=1,IRHT	007660
DO 2704 J=1,IRQ	007670
2704 WM2(I,J)=I	007680
IFJRM=1	007690
NOIM3=NSAV3	007700
NOIM2=NSAV1	007710
CALL AUGMAT(WM2,R,WPD,IFORM,IRHT,IRQ,IRHT,IRHT)	007720
CALL AUGMAT(Q,WM1,WPC,IFORM,IRQ,IRQ,IRQ,IRHT)	007730
ICQA=IRQ+IRHT	007740
IRQA=ICQA	007750
IFORM=2	007760
NOIM2=NSAV3	007770
CALL AUGMAT(WMC,WMD,QA,IFORM,IRQ,ICQA,IRHT,IRQA)	007780
MSG=' THE AUGMENTED Q MATRIX IS ,QA'	007790
CALL MATIO(QA,IRQA,IRQA,IO,KIN,KOUT,NSAV3,NSAV3)	007800
IMA=NSAV3-IRFT	007810
C INITIALIZE PXA MXA AND STORAGE VARIABLES	007820
DO 5105 IPXA=1,NSAV3	007830
5105 MXA(IMXA)=0	007840
DO 5106 IPXA=1,IRFT	007850
5106 MXA(IMXA)=XC(IMXA)	007860
MSG=' THE INITIAL XA VECTOR IS'	007870
CALL MVECIO(MXA,IRFA,IO,KIN,KOUT,NSAV3)	007880
DO 5107 IPXA=1,IRFT	007890
DO 5102 JPXA=1,IRFT	007900
5102 PXA(IPXA,JPXA)=PO(IPXA,JPXA)	007910
DO 5101 JPX=1,IMA	007920
JPXA=JPX+IRFT	007930
5101 PXA(IPXA,JPXA)=J	007940
DO 5103 IPX=1,IMA	007950
IPXA=IPX+IRFT	007960
DO 5103 JPXA=1,IRFA	007970
5103 PXA(IPXA,JPXA)=0	007980

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MSG=' THE INITIAL COVARIANCE MATRIX, PXA IS'                                007390
CALL MMATIO(PXA,IRFA,IRFA,IC,KIN,KOUT,NSAV3,NSAV3)                          008390
C***PXA CALCULATION--- REQUIRED ONLY FIRST TIME THROUGH LOOP                008390
C      PXA= BT(GCZ)                                                            008390
C      1/2      R                                                                008390
C      3CZ                                                                    008390
C      TO USE THE KLIENMAN MULTIPLY ROUTINES, THE DECLARED DIMENSION OF    008390
C      ARRAY ARGUMENTS MUST BE THE SAME. THEREFORE IT IS NECESSARY TO      008390
C      FORM GCZA SUCH THAT GCZA(I,J)=GCZ(I,J) FOR I=1,IRHT, AND J=1,        008390
C      ICBM, AND ZERO ELSEWHERE                                              008390
C      THE SAME REASON REQUIRES CALCULATION OF RA                          008390
C      IT=NSAV3-IRHT                                                            008390
C      JT=NSAV3-ICBM                                                            008390
C      DO 6013 IG=1,IRHT                                                        008390
C      DO 6014 JG=1,ICBM                                                        008390
6014 GCZA(IG,JG)=GCZ(IG,JG)                                                  008390
C      DO 6013 JG=1,JT                                                            008390
C      JGA=ICBM+JG                                                            008390
6013 GCZA(IG,JGA)=0                                                            008390
C      DO 6015 IG=1,IT                                                            008390
C      IGI=IG+IRHT                                                            008390
C      DO 6015 JG=1,NSAV3                                                        008390
6015 GCZA(IGI,JG)=J                                                            008390
C      IR=NSAV3-IRHT                                                            008390
C      DO 6017 IRI=1,IRHT                                                        008390
C      DO 6017 JRI=1,IRHT                                                        008390
6017 RA(IRI,JRI)=R(IRI,JRI)                                                  008390
C      DO 6016 JRI=1,IR                                                            008390
C      JRJ=JRI+IRHT                                                            008390
6016 RA(IRI,JRJ)=0                                                            008390
C      DO 6018 IRI=1,IR                                                            008390
C      IRII=IRI+IRHT                                                            008390
C      DO 6018 JRI=1,NSAV3                                                        008390
6018 RA(IRII,JRI)=0                                                            008390
C      RA = R IN UPPER LEFT PARTITION, ZERO ELSEWHERE                      008390
C      IF((IFLGZ.EQ.0).AND.(IFLGSQ.EQ.1)) THEN                                008390
C      CALCULATE PXA ONLY FOR GCZ NOT EQUAL TO ZERO MATRIX AND NOT FOR S=0  008390
C      RECALL THAT (BT(GCZ) 3CZ)T IS THE RIGHT PARTITION OF GA              008390
C      DO 6003 IPXA=1,IRHT                                                        008390
C      IPA=IPXA+ICGT                                                            008390
C      DO 6003 JPXA=1,IRFA                                                        008390
6003 PXA(JPXA,IPXA)=GA(JPXA,IPA)                                              008390
C      C1=.5                                                                    008390
C      NOIM=NSAV3                                                                008390
C      NOIM1=NSAV3+1                                                            008390
C      CALL MSCALE(WMF,PXA,IRFA,IRHT,C1)                                        008390
C      CALL MAT1(WMF,RA,IRFA,IRHT,IRHT,PXA)                                    008390
C      PXA=PXAT IRFA X IRHT                                                    008390
MSG=' CROSS COVARIANCE, PXA IS'                                              008390
CALL MMATIO(PXA,IRFA,IRHT,IC,KIN,KOUT,NSAV3,NSAV3)                          008390
END IF                                                                        008390
C                                                                              008390
C                                                                              008390
C      FA= FA11 FA12 =FT+BT(GCZ)(HT)      3T(GCX)                            008390
C      =FA21 FA22 =3CZ(HT)                FC                                008390
C                                                                              008390
C                                                                              008390
C      WHERE GCZ IS THE GAIN MATRIX THAT ACTS DIRECTLY ON THE MEASUREMENT    008390
C      VECTOR, AND GCX IS THE GAIN MATRIX THATS ACTS ON THE CONTROLLER      008390
C      STATE ESTIMATES.*****THESE MUST BE SUPPLIED BY THE GAIN MATRIX     008390
C      ROUTINE-----C                                                         008390
C                                                                              008390
C***FORM FA                                                                    008390
C      NOIM1=NSAV1+1                                                            008390
C      NOIM=NSAV1                                                                008390
C                                                                              008390

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CALL MAT1(3T,GCZ,IRFT,ICBT,IRHT,WM1)	003653
C WM1=3T(GCZ) IRFT X IRHT	003663
CALL MAT1(4M1,HT,IRFT,IRHT,IRFT,WM2)	003673
C WM2= 3T(GCZ)HT IRFT X IRFT	003683
C1=1.3	003693
CALL M4001(IRFT,IRFT,FT,WM2,WM1,C1)	003703
C WM1= FA11 IRFT X IRFT	003713
IF (ICBM.NE.ICBT) THEN	003720
WRITE(KOUT,*) 'ICBM=',ICBM,' ICBT=',ICBT	003733
WRITE(KOUT,*) 'BT AND BM ARE NOT THE SAME SIZE- WILL CAUSE ERRORS'	003740
END IF	003750
CALL MAT1(3T,GCX,IRFT,ICBT,IRFM,WM2)	003760
C WM2=FA12 IRFT XIRFM	003770
IFORM=1	003783
NOIM2=NSAV1	003793
CALL AUGMAT(WM1,WM2,WM1,IFORM,IRFT,IRFT,IRFT,IRFM)	003803
C WM4= (FA11 FA12) IRFT X IRFT+IRFM	003813
CALL MAT1(8CZ,HT,IRFM,IRHT,IRFT,WM1)	003820
C WM1= FA21 IRFM X IRFT	003833
CALL AUGMAT(WM1,FC,WM3,IFORM,IRFM,IRFT,IRFM,IRFM)	003840
C WM3= (FA21 FA22) IRFM X IRFM+IRFT	003853
NOIM2=NSAV3	003863
IFORM=2	003873
IRFA=IRFT+IRFM	003883
ICFA=IRFA	003893
CALL AUGMAT(WM4,WM3,FA,IFORM,IRFT,IRFA,IRFM,ICFA)	003903
MSG= 'THE AUGMENTED F MATRIX FA IS'	003913
CALL MMATIO(FA,IRFA,IRFA,IO,KIN,KOUT,NSAV3,NSAV3)	003923
C FA IRFA X IRFA	003933
C	003943
C***FORM BA - - - FOR REGULATOR CASE , NOT REQUIRED Y=0	003953
CALL MAT1(3T,GCY,IRFT,ICBT,IRY,WM1)	003963
C WM1= 3T(GCY) IRFT X IRY	003973
IFORM=2	003983
NOIM2=NSAV1	003993
CALL AUGMAT(WM1,BCY,BA,IFORM,IRFT,IRY,IRFM,IRY)	004003
MSG= 'THE AUGMENTED B MATRIX BA IS'	004013
CALL MMATIO(BA,IRFA,IRY,IO,KIN,KOUT,NSAV3,NSAV3)	004023
C BA IRFA X IRY	004033
C	004043
C***FORM GA	004053
C	004063
C GA= GT 3T(GCZ)	004073
C = 0 BCZ	004083
C	004093
CALL MAT1(8T,GCZ,IRFT,ICBM,IRHT,WM1)	004103
C WM1= 8T(GCZ) IRFT X IRHT	004113
IFORM=1	004123
CALL AUGMAT(GT,WM1,4MC,IFORM,IRFT,IRQ,IRFT,IRHT)	004133
C RECALL IRQ=ICGT FOR CONTINUOUS SYS.=IRFT FOR S=0 SYS	004143
C WM4= (GT 3T(GCZ)) IRFT X IRHT+IRQ	004153
DO 3301 IR=1,IRFM	004163
DO 3301 IC=1,IRQ	004173
3301 WM1(IR,IC)=0	004183
CALL AUGMAT(WM1,BCZ,WMO,IFORM,IRFM,IRQ,IRFM,IRHT)	004193
C WMO= (0 BCZ) IRFM X IRQ+IRHT	004203
ICGA=IRQ+IRHT	004213
IFORM=2	004223
NOIM2=NSAV3	004233
CALL AUGMAT(WMC,WMO,GA,IFORM,IRFT,ICGA,IRFM,ICGA)	004243
MSG= 'THE AUGMENTED G MATRIX GA IS'	004253
CALL MMATIO(GA,IRFA,ICGA,IO,KIN,KOUT,NSAV3,NSAV3)	004263
C GA IRFA X ICGA	004273
C	004283
C GUA=GCZ(HT) GCX)	004293
C	004303

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      NOIM=NSAV1
      NOIM1=NSAV1+1
      CALL MAT1(GCZ,HT,ICBT,IRHT,IRFT,WM2)
C WM2= GCZ(WT) ICBT X IRFT
      IFORP=1
      NOIM2=NSAV1
      NOIM3=NSAV3
      CALL AUGMAT(WM2,GCX,GUA,IFORP,ICBT,IRFT,ICBM,IRFM)
      MSG='THE AUGMENTED MATRIX GUA IS'
      CALL MMAT10(GUA,ICBT,IRFA,IC,KIN,KOUT,NSAV3,NSAV3)
C GUA ICBT X IRFA
C*****AUGMENTED SYSTEM MATRICES NOW AVAILABLE FOR COMPUTATION
C
      NOIM=NSAV1
      NOIM1=NSAV2
      NOIM2=NSAV3
      NOIM3=NSAV4
      END
* DECK PERFAL
      SUBROUTINE PERFAL(TRY,IFLGZ,MXA,GCY,GUA,PXA,PXVA,IFLGSD,
      1RA,GCZA,YO,EAT,INTGA,WME,WMF,PUU,WMI,MUOUT,MXTOUT,PUOUT,PXTOUT,
      1MXAMIN,MXAMAX,PXTMIN,PXTMAX,PUMIN,MUMAX,PUMIN,PUMAX,MU,INTBA,
      1WV3,WV4)
      CHARACTER MSG*60
      REAL WMI(NOIM,NOIM),EAT
      1(NOIM2,NOIM2),INTGA(NOIM2,NOIM2),WME(NOIM2,NOIM2),
      1WMF(NOIM2,NOIM2),WV3(NOIM2),WV4(NOIM2),
      1 MXA(NOIM2),PXA(NOIM2,NOIM2),PXVA(NOIM2,
      1NOIM2),MUCUT(NOIM),MXTOUT(NOIM),PXTOUT(NOIM),PUOUT(NOIM),
      1 YO(NOIM3),MXAMIN(NOIM),MXAMAX(NOIM),MUMIN(NOIM),MUMAX(NOIM),
      1 PXTMIN(NOIM),PXTMAX(NOIM),PUMIN(NOIM),PUMAX(NOIM),GCZA
      1 (NOIM2,NOIM2),RA(NOIM2,NOIM2),GUA(NOIM2,NOIM2),GCY(NOIM,
      1 NCIN),INTBA(NOIM2,NOIM2)
      INTEGER IFLGZ
      DIMENSION COM1(1),COM2(1)
      REAL MU(NCIN),PUU(NOIM2,NOIM2)
      COMMON /RTIM/ RTIME,DELTIM
      COMMON /MAIN/ NCIN2,NOIM3
      COMMON /MAIN2/COM2
      COMMON /MAIN1/NOIM,NOIM1,COM1
      COMMON /INOU/ KIN,KOUT,KPUNCH
      COMMON /MAUNS/ MSG
      COMMON /MAIN6/ICBT,ICBM,ICFA,ICGA,ICGN,ICGT,ICGA,IRFA,IRFM,IRFT,
      1IRHT,IRQA,IO,LOG,IRPM,NUMOTS
      NSAV1=NOIM
      NSAV2=NOIM1
      NSAV3=NOIM2
      NSAV4=NOIM3
C
C****PERFORMANCE ANALYSIS ROUTINE
C THIS IS A CONTINUOUS TIME MEASUREMENT PERFORMANCE ANALYSIS
C ROUTINE FOR EVALUATING CONTINUOUS TIME CONTROL SYSTEMS DRIVEN BY
C WHITE GAUSSIAN NOISE. IT COMPUTES THE MEAN AND COVARIANCE OF THE
C OF THE TRUTH MODEL STATES, THE CONTROLLER STATES, AND THE CONTROLS
C GENERATED. A SET OF AUGMENTED MATRICES IS USED TO DO THE
C CALCULATIONS---Y=(XT USTAT), XA=(XT XM)T .. THE PERFORMANCE
C ANALYSIS ROUTINE IS DEVELOPED IN A MASTERS THESES FOR AIR FORCE
C INSTITUTE OF TECHNOLOGY BY ERIC LLOYD, TITLE 'ACBUST CONTROL
C SYSTEM DESIGN'
C
C****MXA,PXA CALCULATION--- THE MEAN AND COVARIANCE OF THE XA VECTOR
C FOUND USING SOLUTION FORMS OF THE PROPAGATION EQUATIONS
C KLEINMAN ROUTINES ARE USED TO PROVIDE THE SOLUTIONS
C IN THE FOLLOWING TWO EQS. THE FIRST OCCURRENCE OF PXA OR MXA
C IS THE VALUE AT TIME T=DELTIM, THE SECOND ---AT TIME T

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C	PX1=EAT(PX1)EATT+INTGA	009973	
C	IX1=EAT(MXA)+INTB1	009983	
C	SEE DEFINITIONS BELOW FOR EAT, INTGA, INTB1	009990	
C		010000	
C	NOTE SINCE THIS PROGRAM CONSIDERS ONLY THE REGULATOR CASE,	010010	
C	Y= THE DESIRED INPUT- IS ASSUMED = ZERO	010020	
C		010030	
C		010040	
C	MX10=(E(X0) CIT =(X0 CIT E=THE EXPECTED VALUE OPERATOR	010050	
C	PX10= PO 3	010060	
C	1 3	010070	
C		010080	
C	EAT= EXP(FA*TIME)	010090	
C	INTB1= INTEG(EAT)BA FOR CONTINUOUS TIME SYSTEMS	010100	
C	= BAD FOR DISCRETE SYSTEMS	010110	
C	INTGA= INT(EAT(GA)GA(GA)EATT) FOR CONT TIME SYSTEMS	010120	
C	= GOA(GO1)GOAT FOR DISCRETE TIME SYSTEMS	010130	
C		010140	
C	LCOUNT=3	010150	
C	IRN=NINT(NTIME/DELTIM)	010160	
C	WRITE(KOUT,*)'ENTER A : IF YOU WANT NO PRINTS OF PX1, PUU, MXA,	010170	
C	1 AND MU MATRICES DURING THE PERFORMANCE ANALYSIS, ELSE ENTER THE	010180	
C	1 NUMBER OF TIME INCREMENTS BETWEEN PRINTS( THERE ARE ',IRN,' TOTAL	010190	
C	1 TIME INCREMENTS IN THIS RUN)>'	010200	
C	READ(KIN,*)IPCNTL	010210	
C	IF (IPCNTL.EQ.0) THEN	010220	
C	IC=0	010230	
C	ELSE	010240	
C	IC=5	010250	
C	END IF	010260	
C	DO 5003 IN=1,NSAV1	010270	
C	PUU(IN,IN)=0	010280	
C	MU(IN)=0	010290	
C	MX1MIN(IN)=0	010300	
C	MX1MAX(IN)=0	010310	
C	PXTMIN(IN)=0	010320	
C	PXTMAX(IN)=0	010330	
C	MUMIN(IN)=0	010340	
C	MUMAX(IN)=0	010350	
C	PUMAX(IN)=0	010360	
C	PUMIN(IN)=0	010370	
C	5003 CONTINUE	010380	
C	WRITE(KOUT,*)'ENTER THE NUMBER OF SAMPLE PERIODS DESIRED BETWEEN	010390	
C	1 PLOT POINTS(MAX 1000 PLOT PCINTS) THERE ARE ',IRN,' SAMPLE PERIOD	010400	
C	1S REQUESTED FOR THIS RUN>'	010410	
C	READ(KIN,*)IPLTPS	010420	
C	DEPLTP=DELTIM*IPLTPS	010430	
C	DO 5000 ITLMP=1,IRN	010440	
C		010450	
C	IF (IC.EQ.3) THEN	010460	
C	JJ=ITLMP-1	010470	
C	IPNT=400(JJ,IPCNTL)	010480	
C	IF ((IPNT.EQ.0).OR.(ITLMP.EQ.IRN)) THEN	010490	
C	TIME=JJ*DELTIM	010500	
C	WRITE(KOUT,*)'TIME= ',TIME	010510	
C	MSG=' PX1'	010520	
C	CALL MWRITE(PXA,3,3,IC,KIN,KOUT,NSAV3,NSAV3)	010530	M
C	MSG=' PUU'	010540	
C	CALL MWRITE(PUU,IC3M,IC3M,IC,KIN,KOUT,NSAV3,NSAV3)	010550	
C	MSG=' MXA'	010560	
C	CALL MWRITE(MXA,3,3,IC,KIN,KOUT,NSAV3)	010570	M
C	MSG=' MU'	010580	
C	CALL MWRITE(MU,IC3M,IC,IC,KIN,KOUT,NSAV1)	010590	
C	END IF	010600	
C	END IF	010610	
C	FOR WANT TO STORE FOR PLOTTING. MXT,PXY,MU,PUU	010620	

DO 123 IWR=1,ICBM	010653	M
MXTOUT(IWR)=M(A(IWR))	010654	
PXTOUT(IWR)=SQRT(PXA(IWR,IWR))	010655	M
123 CONTINUE	010656	
DO 124 IWR=1,ICBM	010657	
MUOUT(IWR)=MU(IWR)	010658	
PUOUT(IWR)=SQRT(PUU(IWR,IWR))	010659	M
124 CONTINUE	010660	
C*****	010661	
C***** NO CROSS CORRELATION TERMS ARE PLOTTED	010662	
C	010663	
NDIM=NSAV1	010664	
IPCTL=MOD(JJ,IPLTPS)	010665	
IF (IPCTL.EQ.0) THEN	010666	
CALL STORED(LOG,RATIME,DELPLT,LCCOUNT,7,7,ICBM,ICBM,	010667	
1 MXTOUT,PXTOUT,MUOUT,PUOUT,NDIM6)	010668	M
LCCOUNT=LCCOUNT+1	010669	
END IF	010670	
C NOTE THAT YD IS RESTRICTED BY VALUE OF LCCOUNT TO BE CONSTANT	010671	
C BETWEEN PLOT POINTS	010672	
IF (LCCOUNT.GT.1000) THEN	010673	
C RESET LCCOUNT	010674	
LCCOUNT=1000	010675	
END IF	010676	
C**UPDATE PXA,MXA	010677	
C	010678	
C	010679	
NDIM=NSAV3	010680	
NDIM1=NSAV3+1	010681	
CALL MAT3(IRFA,IRFA,EAT,PXA,MXE)	010682	
C1=1.0	010683	
CALL MADD1(IRFA,IRFA,MXE,INTGA,PXA,C1)	010684	
C**PXA AT NEW TIME NOW AVAILABLE	010685	
C MXA= EAT(MXA0)+INTEG(EAT(BA))(YD)	010686	
DO 1814 IK=1,IRFA	010687	
1814 WV3(IK)=0	010688	
DO 1815 IK=1,IRFA	010689	
DO 1815 IJ=1,IRFA	010690	
1815 WV3(IK)=WV3(IK)+EAT(IK,IJ)*MXA(IJ)	010691	
DO 1812 IWR=1,IRFA	010692	
1812 WV4(IWR)=INTBA(IWR,1)*YD(LCCOUNT)	010693	
DO 1813 IJ=1,IRFA	010694	
1813 MXA(IJ)=WV3(IJ)+WV4(IJ)	010695	
C**MXA AT NEW TIME NOW AVAILABLE	010696	
C**MU,PUU CALCULATION FOR ZERO MEAN MEASUREMENT NOISE	010697	
C MU=GUA(MXA)+GCZ(MVT)+GCY(YD)....MVT,THE MEAN OF NOISE V ASSUMED +0	010698	
X1=YD(LCCOUNT)	010699	
NDIM=NSAV1	010700	
NDIM1=NSAV1+1	010701	
CALL MSCALE(WM1,GCY,ICBM,IRY,X1)	010702	
C WM1= GCY(YD) ICBM X 1	010703	
NDIM=NSAV3	010704	
NDIM1=NSAV3+1	010705	
DO 1817 IJ=1,ICBM	010706	
1817 WV3(IJ)=0	010707	
DO 1816 IJ=1,ICBM	010708	
DO 1816 IK=1,IRFA	010709	
1816 WV3(IJ)=WV3(IJ)+GUA(IJ,IK)*PXA(IK)	010710	
C WV3=GUA(MXA) ICBM X 1	010711	
C ADDED TO WM1 ABOVE TO GET MU	010712	
DO 329 I=1,ICBM	010713	
329 MU(I)=WV3(I)+WM1(I,1)	010714	
C MU IRFA X 1 ----NOW AVAILABLE-----	010715	
C**PUU CALCULATION,PUU=GUA(PXA)GUAT+GUA(PXVA)GCZT+GCZ(PXVA)GUAT+	010716	
C GCZ(R)GCZT	010717	
CALL MAT3(ICBM,IRFA,GUA,PXA,PUU)	010718	



C PUU=GUA(PXA)GUAT ICBM X ICBM	011290
C*****IFLGZ*****	011300
IF (IFLGZ.EQ.0) THEN	011310
C SINCE GZ NOT EQUAL TO ZERO CALCULATE OTHER TERMS OF PUU	011320
C*****IFLGZ*****	011330
IF (IFLGZ.EQ.0) THEN	011340
C GUNT DO THIS FOR S-D CASE	011350
CALL MAT1(GUA,PXA,ICBM,IRFA,IRHT,MHF)	011360
CALL MAT4(MHF,GCZA,ICBM,IRHT,ICBM,MHE)	011370
C1=1.7	011380
CALL MADD1(ICBM,ICBM,PUU,MHE,MHF,C1)	011390
C MHF=GUA(PXA)GUAT+GUA(PXA)GCZT	011400
CALL MAT4(GCZA,PXA,ICBM,IRHT,IRFA,PUU)	011410
CALL MAT4(PUU,GUA,ICBM,IRFA,ICBM,MHE)	011420
CALL MADD1(ICBM,ICBM,MHE,MHF,PUU,C1)	011430
C PUU=GUA(PXA)GUAT+GUA(PXA)GCZT+GCZ(PXA)GUAT	011440
END IF	011450
CALL MAT3(ICBM,IRHT,GCZA,PA,MHE)	011460
CALL MADD1(ICBM,ICBM,MHE,PUU,MHF,C1)	011470
CALL EQUATE(PUU,MHF,ICBM,ICBM)	011480
END IF	011490
C*** PUU NOW AVAILABLE ICBM X ICBM	011500
C	011510
C	011520
C	011530
9630 CONTINUE	011540
DO 120 INR=1,7	011550
MATOUT(INR)=MXA(INR)	011560
120 PXTOUT(INR)=SORT(PXA(INR,INR))	011570
DO 121 INR=1,ICBM	011580
MUOUT(INR)=MU(INR)	011590
121 PUOUT(INR)=SORT(PUU(INR,INR))	011600
CALL STORED(LOG,RTIME,DELT,LCOUNT,7,7,ICBM,ICBM,	011610
1 MATOUT,PXTOUT,MUOUT,PUOUT,NDIM6)	011620
C CALL TO STORED WITH LCOUNT = 9 INDICATE THIS DATA RUN COMPLETE	011630
LCOUNT=-12	011640
CALL STORED(LOG,RTIME,DELT,LCOUNT,7,7,ICBM,ICBM,	011650
1 MATOUT,PXTOUT,MUOUT,PUOUT,NDIM6)	011660
NDIM=NSAV1	011670
NDIM1=NSAV2	011680
NDIM2=NSAV3	011690
NDIM3=NSAV4	011700
C RESET IO TO SOME NONZERO VALUE TO AVOID TERMINATING THE PROGRAM	011710
C WHEN RETURNING TO MAIN ROUTINE. LOGR	011720
IO=25	011730
END	011740
* DECK MYPLOT	011750
SUBROUTINE MYPLOT	011760
END	011770
* DECK AUGMAT	011780
SUBROUTINE AUGMAT(A1,A2,A3,IFORM,IRA1,ICA1,IRA2,ICA2)	011790
C	011800
C*****NDIM2,NDIM3 MUST BE SET IN THE CALLING PROGRAM BEFORE USING	011810
C*****NDIM2,NDIM3 MUST BE SET IN THE CALLING PROGRAM BEFORE USING	011820
C***** THIS SUBROUTINE. THEY MUST BE DECLARED IN A COMMON BLOCK	011830
C LABELED --MAIN4---	011840
* DECK MVECIO	011850
C THIS SUBROUTINE FORMS AUGMENTED MATRICES OF THE FORM	011860
C IFORM=1 A3=(A1 A2)	011870
C IFORM=2 A3=(A1 A2)T	011880
C	011890
C	011900
C IRA1,IRA2,ARE ROW DIMENSIONS,ICA1,ICA2,ARE COLUMN DIMENSIONS	011910
DIMENSION A1(NDIM2,NDIM2),A2(NDIM2,NDIM2),A3(NDIM3,NDIM3)	011920
COMMON /MAIN4/NDIM2,NDIM3	011930
IF (IFORM.EQ.1) THEN	011940

C FORM THE AUGMENTED MATRIX A3=(A1 A2)	011950	
DO 12 II=1,ICA1	011960	
DO 11 III=1,ICA1	011970	
11 A3(II,III)=A1(II,III)	011980	
DO 12 IV=1,ICA2	011990	
IVI=IV+ICA1	012000	
12 A3(II,IVI)=A2(II,IV)	012010	
RETURN	012020	
END IF	012030	
C FORM AUGMENTED MATRIX A3=(A1 A2)	012040	
IRA3=IRA1+IRA2	012050	
ICA3=ICA1	012060	
DO 14 II=1,ICA3	012070	
DO 13 III=1,IRA1	012080	
13 A3(III,II)=A1(III,II)	012090	
DO 14 IV=1,IRA2	012100	
IVI=IV+IRA1	012110	
14 A3(IVI,II)=A2(IV,II)	012120	
RETURN	012130	
END	012140	
SUBROUTINE MVECIO(A,NUMEL,IC,KIN,KOUT,NDIM)	012150	
C THIS SUBROUTINE READS PRINTS ENTIRE (PORTIONS OF) THE VECTOR	012160	
C A, DEPENDING ON THE VALUE OF ---IO---. IO=1---READ ONLY	012170	
C IO=2---READ AND PRINT. IO=3 READ SELECTED VALUE, IO=4	012180	
C READ AND PRINT SELECTED VALUES	012190	
C IO=5---PRINT ONLY	012200	
C TO USE IO=3 OR 4 THE CALLING PROGRAM MUST INITIALIZE THE VEC.	012210	
C *****THIS ROUTINE SETS IO=3--- WHEN NO DATA IN INPUT FILE	012220	
C	012230	
CREAD IS FROM UNIT SPECIFIED BY CALLING PROGRAM IN KIN,WRITE IS TO	012240	
C KOUT,NDIM IS THE DECLARED DIMENSION OF A IN THE CALLING	012250	
C PROGRAM	012260	
CHARACTER MSG*51	012270	
DIMENSION A(NDIM)	012280	
COMMON /MAUNS/ MSG	012290	
IF ((IO.EQ.1).OR.(IO.EQ.2)) THEN	012300	
C READ ENTIRE VECTOR	012310	
WRITE(KOUT,*) 'ENTER ',NUMEL,'ELEMENTS>'	012320	
READ(KIN,*,END=29) (A(I),I=1,NUMEL)	012330	
END IF	012340	
IF (IO.EQ.1) THEN	012350	
RETURN	012360	
END IF	012370	
IF ((IO.EQ.3).OR.(IC.EQ.4).OR.(IO.EQ.6)) THEN	012380	M
C READ ONLY SELECTED ELEMENTS. THE FIRST NUMBER ON EACH CARD	012390	
C IS THE SUBSCRIPT, THE SECOND IS THE DATA ENTRY	012400	
C*****NOTE ONLY ONE DATA ENTRY PER CARD	012410	
C***** FIRST CARD MUST CONTAIN THE TOTAL NUMBER OF ENTRIES TO BE	012420	
C READ	012430	
IF(IC.NE.6) GO TO 4	012440	
DO 3 I=1,NDIM	012450	
3 A(I)=0.	012460	
4 CONTINUE	012470	
WRITE(KOUT,*) 'ENTER THE ELEMENT NUMBER ,THEN ITS VALUE>'	012480	
5 READ(KIN,*,END=29) I,ENTRY	012490	
IF(I.LE.0) GO TO 21	012500	M
IF(I.GT.NDIM) GO TO 5	012510	
A(I)=ENTRY	012520	
IF (IO.EQ.4) THEN	012530	
WRITE(KOUT,33) ' '	012540	
WRITE(KOUT,*) MSG	012550	
WRITE(KOUT,*) 'ELEMENT NUMBER , ENTRY>'	012560	
WRITE(KOUT,11) I,A(I)	012570	
END IF	012580	
GO TO 5	012590	M
21 CONTINUE	012600	

RETURN	012510
END IF	012520
IF ((IO.EQ.2).OR.(IO.EQ.5)) THEN	012530
C TO GET HERE IO=2 OR 5 SO PRINT OUT ENTIRE VECTOR	012540
WRITE(KOUT,33) ' '	012550
WRITE(KOUT,*) MSG	012560
WRITE(KOUT,*) ' THE VECTOR HAS ', NUMEL, ' ELEMENTS'	012570
WRITE(KOUT,22) (A(I), I=1, NUMEL)	012580
RETURN	012590
END IF	012600
RETURN	012710
29 PRINT*, 'END OF DAT REACHED DURING INPUT IN MVEGIC'	012720
IO=	012730
C *****THIS ROUTINE SETS IO=0---- WHEN NO DATA IN INPUT FILE	012740
33 FORMAT(A10, /)	012750
11 FORMAT(I4, 15X, E12.6)	012760
22 FORMAT(10(1J(1X, E12.6), /))	012770
RETURN	012780
END	012790
* DECK MMATIO	012800
SUBROUTINE MMATIO (A, IR, IC, IC, KIN, KOUT, NDIM, NDIM1)	012810
CHARACTER MSG*63	012820
DIMENSION A (NDIM, NDIM1)	012830
COMMON /MAUNS/ MSG	012840
C THIS SUBROUTINE READS AND/OR PRINTS THE MATRIX A DEPENDING ON THE	012850
C VALUE OF IO. IT READS FROM UNIT SPECIFIED BY KIN AND WRITES TO UNIT	012860
C KOUT. IO=1--READ ENTIRE ARRAY IO=2--READ AND PRINT ENTIRE	012870
C ARRAY. IO=3--READ SELECTED ELEMENTS OF A IO=4--READ AND	012880
C PRINT SELECTED ELEMENTS OF A IO=5 ---PRINT ENTIRE ARRAY	012890
C NDIM, NDM1 ARE THE DIMENSIONS OF A IN THE CALLING PROGRAM	012900
C *****NOTE IF IO=3 OR 4 THE CALLING PROGRAM MUST INITIALIZE	012910
C THE ENTIRE ARRAY BEFORE CALL	012920
C *****THIS ROUTINE SETS IO =0 ---- WHEN THE INPUT FILE IS EMPTY	012930
C	012940
IF ((IO.EQ.1).OR.(IO.EQ.2)) THEN	012950
C READ ENTIRE ARRAY IN FREE FORMAT, ROW MAJOR ORDER	012960
WRITE(KOUT,*) 'ENTER ', (IR*IC), ' ARRAY ELEMENTS IN ROW MAJ ORDER>'	012970
READ(KIN,*, END=29) ((A(I,J), J=1, IC), I=1, IR)	012980
END IF	012990
IF (IO.EQ.1) THEN	013000
RETURN	013010
END IF	013020
IF ((IO.EQ.3).OR.(IO.EQ.4).OR.(IO.EQ.5)) THEN	013030
C READ IN SELECTED ELEMENTS OF A	013040
C THE FIRST CARD IN THE INPUT STREAM MUST CONTAIN THE TOTAL	013050
C NUMBER OF ELEMENTS TO BE READ IN. ONLY ONE ENTRY PER CARD.	013060
C THE FIRST ITEM ON EACH CARD IS THE ROW, THE SECOND IS THE COL THE	013070
C LAST ON EACH CARD IS THE DATA FOR THAT LOCATION	013080
C FREE FORMAT IS USED	013090
IF (IC.NE.6) GO TO 4	013100
DO 45 N=1, IR	013110
OC 45 N=1, IC	013120
45 A(N,N)=0.0	013130
4 CONTINUE	013140
WRITE(KOUT,*) 'ENTER THE ROW AND COLUMN FOLLOWED BY ITS VALUE>'	013150
5 READ(KIN,*, END=29) II, J, ENTRY	013160
IF ((II.LE.0).OR.(J.LE.0)) GO TO 2:	013170
IF ((II.GT. IR).OR.(J.GT. IC)) GO TO 5	013180
A(II, J)=ENTRY	013190
IF (IO.EQ.4) THEN	013200
WRITE(KOUT,33) ' '	013210
WRITE(KOUT,*) MSG	013220
WRITE(KOUT,*) ' (', II, ', ', J, ') = ', A(II, J)	013230
	013240
	013250
	013260

END IF	013270	
GO TO 5	013280	M
20 CONTINUE	013290	
RETURN	013300	
END IF	013310	
IF ((IO.EQ.2).OR.(IC.EQ.5)) THEN	013320	
C IO = 2 OR 5 IF HERE SC PRINT ENTIRE ARRAY	013330	
WRITE(KOUT,33) ' '	013340	
WRITE(KOUT,*) MSG	013350	
WRITE(KOUT,*) ' MATRIX SIZE IS ',IR,' X ',IC	013360	
DO 49 I=1,IR	013370	
49 WRITE(KOUT,48) (A(I,J),J=1,IC)	013380	
33 FORMAT(A1C,/)	013390	
48 FORMAT(5(10(1X,E12.6),1,/) )	013400	
END IF	013410	
RETURN	013420	
29 PRINT *, 'END OF DATA REACHED DURING INPUT'	013430	
C *****THIS ROUTINE SETS IO = 0 ---- WHEN THE INPUT FILE IS EMPTY	013440	
IO=1	013450	
RETURN	013460	
END	013470	
* DECK CLOGRS	013480	
SUBROUTINE CLOGRS(IGCSTR,FP,BP,RKFSS,WM,GCX,GCY,GCZ,	013490	
1BCY,ECZ,FC,YO,RM,QM,FT,BT,GT,OT,RT,MT,IRY,IFLG CZ,WM1	013500	
1 ,WM2,WM3,PO,GM,WM4,WM5,WM6,	013510	
1 WV1,WV2,WUU,WXX,XO,WXU,WM7,WM8)	013520	
C THIS ROUTINE PERFORMS SET UP FOR USING THE CONTINUOUS TIME	013530	
C PERFORMANCE ANALYSIS FOR ANALOG REGULATOR	013540	
DIMENSION GICSTR(NDIM,NDIM),FP(NDIM,NDIM),BM(NDIM,NDIM),PG(NDIM,	013550	
1NDIM),HM(NDIM,NDIM),GCX(NDIM,NDIM),GCY(NDIM,NDIM),GCZ(NDIM,NDIM),	013560	
1BCY(NDIM,NDIM),ECZ(NDIM,NDIM),FC(NDIM,NDIM),YO(NDIM3),WM1(NDIM,	013570	
1NDIM),WM2(NDIM,NDIM),WM3(NDIM,NDIM),FT(NDIM,NDIM),	013580	
1 BT(NDIM,NDIM),GT(NDIM,NDIM),HT(NDIM,NDIM),RM(NDIM,NDIM),	013590	
1 OT(NDIM,NDIM),RT(NDIM,NDIM),QM(NDIM,NDIM),WV1(NDIM),WV2(NDIM	013600	
1),XO(NDIM),GM(NDIM,NDIM),WUU(NDIM,NDIM),	013610	
1 WXX(NDIM,NDIM),WM4(NDIM,NDIM),WM5(NDIM,NDIM),	013620	
1 WM6(NDIM,NDIM),WXU(NDIM,NDIM),WM7(NDIM,NDIM),WM8(NDIM,NDIM)	013630	
DIMENSION COM1(1),COM2(1)	013640	
CHARACTER MSG*63	013650	
INTEGER IFLGCZ	013660	
REAL RKFSS(NDIM,NDIM)	013670	
COMMON /MAIN4/NDIM2,NDIM3	013680	
COMMON /MAUN5/ MSG	013690	
COMMON /MAIN1/ NDIM,NDIM1,COM1	013700	
COMMON /MAIN2/ COM2	013710	
COMMON /RNTM/RNTIME,DELTIME	013720	
COMMON /INOU/ KIN,KOUT,KPUNCH	013730	
COMMON /MAIN6/ICBT,ICBM,ICFA,ICGA,ICGM,ICGT,ICG4,IRFA,IRFM,IRFT,	013740	
1 IRFT,IRQA,IO,LOG,IRHM,NUMOTS	013750	
WRITE(KOUT,*) 'DO YOU WANT TO CALCULATE EIGENVALUES OF THE TRUTH MO	013760	
1DEL AND CONTROLLER MODEL F MATRICES, Y OR N>'	013770	
READ(KIN,23,END=2933) MSG	013780	
23 FORMAT(A1)	013790	
IF (MSG.EQ.'Y') THEN	013800	
WRITE(KOUT,*) ' '	013810	
WRITE(KOUT,*) 'THE EIGENVALUES OF THE TRUTH MODEL F MATRIX'	013820	
CALL MEIGN(FT,WV1,WV2,IRFT,WM1)	013830	
WRITE(KOUT,*) 'THE EIGENVALUES OF THE CONT. MODEL F MATRIX'	013840	
CALL MEIGN(FM,WV1,WV2,IRFM,WM1)	013850	
END IF	013860	
WRITE(KOUT,*) 'COLORED INPUT NOISE (Y OR N) >'	013870	
READ(KIN,23)MSG	013880	
INOL0=IRFM	013890	A
IF (MSG.EQ.'N') GO TO 2960	013900	
CALL CNOISE(FM,GM,WM,BM,QM,IRFM,IRHM,ICGM,ICBM,WM1,WM2,WM3)	013910	
2900 CALL CKFT(FM,GM,WM,WM,NUMOTS,RKFSS,QM,WM1,WM2,WM3,WM4,WM5,WM6,	013920	

1	IRFM,IRFM,ICBM,(CY,GCZ,BM,ICBM)	013930
C	GCZ,GCY, IN CALL TO CKFTR ARE USED AS DUMMY ARRAYS FOR M2 ,F2	013940
	CALL GDTCON(FM,BM,HAX,HUU,GCSTR,ICOLD,ICBM,WM1,WM2,WM3,WM4,WM5,	013950
	WM6,HAI,WM7,WM8)	013960
	IF (IHOLD.EQ.IRFM) GC TO 35	013970
	ICH=IHOLD+1	013980
	DO 25 I=1,ICBM	013990
	DO 25 J=ICH,IRFM	014000
25	GCSTR(I,J)=0.	014010
35	CONTINUE	014020
	WRITE(KOUT,*) 'ENTER THE TOTAL RUN TIME AND THE TIME INCREMENT>'	014030
	READ(KIN,*,END=2933)RTIME,DELTIM	014040
	C1=-1	014050
	CALL IDNT(IRFM,WM1,C1)	014060
C	WM1=-1 IRFM X IRFM	014070
	CALL MAT1(GCSTR,WM1,ICBM,IRFM,IRFM,GCX)	014080
	IO=5	014090
	MSG='GCX FOLLOWS, GCY,GCZ SET=0'	014100
	CALL MMATIO(GCX,ICBM,IRFM,IC,KIN,KOUT,NOIM,NOIM)	014110
C	NOTE***** OTHER GAIN MATRICES,GCZ,AND GCY SHOULD BE CALCULATED IN	014120
C	THIS MODULE FOR USE IN THE PERFORMANCE ANALYSIS ROUTINE.	014130
	DO 2902 III=1,ICBM	014140
	DO 2902 IIII=1,IRFM	014150
2902	GCZ(III,IIII)=0	014160
	IFLGCZ=1	014170
C	IFLGCZ=1 INDICATES GCZ IS SET TO ZERO--PERF ANALYSIS ROUTINE USES IF	014180
C		014190
	DO 2903 I=1,1000	014200
2903	YD(I)=0	014210
	IRY=1	014220
	DO 1723 I=1,ICBM	014230
	DO 1723 J=1,IRY	014240
1723	GCY(I,J)=0	014250
C	FORM BCY	014260
	DO 1702 I=1,IRFM	014270
	DO 1702 J=1,IRY	014280
1702	BCY(I,J)=0	014290
C	YD IS ALLOWED TO ONLY BE A SCALAR AT THIS TIME	014300
C	FORM BCZ	014310
	C1=1.0	014320
	CALL EQUATE(BCZ,RKFSS,IRFM,IRFM)	014330
	MSG='BCZ FOLLOWS,BCY=0'	014340
	CALL MMATIO(BCZ,IRFM,IRFM,IC,KIN,KOUT,NOIM,NOIM)	014350
C	FORM FC	014360
	CALL MAT1(BM,GCX,IRFM,ICBM,IRFM,WM1)	014370
	CALL MADD1(IRFM,IRFM,FM,WM1,WM2,C1)	014380
	WRITE(KOUT,*) ' '	014390
	WRITE(KOUT,*) 'DO YOU WANT TO CALCULATE THE EIGENVALUES OF THE CONT	014400
	INUOUS-TIME LO CONTROLLER? Y OR N>'	014410
	READ(KIN,23)MSG	014420
	IF (MSG.EQ.'Y') THEN	014430
	WRITE (KOUT,*) 'THE EIGENVALUES THAT CORRESPOND TO THE POLES OF	014440
	THE CONTINUOUS-TIME LO CONTROLLER ARE...'	014450
	CALL MEIGN(WM2,WV1,WV2,IRFM,WM1)	014460
	END IF	014470
	C1=-1	014480
	CALL MAT1(RKFSS,WM,IRFM,IRFM,IRFM,WM1)	014490
	WRITE(KOUT,*) ' '	014500
	WRITE(KOUT,*) 'DO YOU WANT TO CALCULATE THE POLES OF THE CONTINU	014510
	OUS-TIME KALMAN FILTER? Y OR N>'	014520
	READ(KIN,23)MSG	014530
	IF (MSG.EQ.'Y') THEN	014540
	CALL MADD1(IRFM,IRFM,FM,WM1,WM7,C1)	014550
	WRITE(KOUT,*) 'THE EIGENVALUES THAT ARE THE POLES OF THE CONTINU	014560
	OUS-TIME KALMAN FILTER ARE....'	014570
	CALL MEIGN(WM7,WV1,WV2,IRFM,WM8)	014580

A

END IF	014590
CALL MADD1(IFFM,IRFM,WM2,WM1,FC,C1)	014600
MSG='FC FOR THE LOG CONTROLLER IS'	014610
CALL MHATIO(FC,IRFM,IRFM,IC,KIN,KOUT,NDIM,NDIM)	014620
WRITE(KOUT,*)'	014630
WRITE(KOUT,*)'DO YOU WISH TO CALCULATE THE EIGENVALUES OF THE LOG	014640
CONTROLLER F MATRIX? Y OR N'	014650
READ(KIN,23)MSG	014660
IF (MSG.EQ.'Y') THEN	014670
WRITE(KOUT,*)'	014680
WRITE(KOUT,*)'THE EIGENVALUES OF THE LOG CONTROLLER F MATRIX ARE'	014690
CALL MEIGN(FC,MV1,MV2,IRFM,WM1)	014700
END IF	014710
RETURN	014720
2933 IO=0	014730
END	014740
* DECK DAS1	014750
SUBROUTINE DAS1(GOGT,BM,V,ICBM,WM3,IRFM)	014760
DIMENSION GOGT(NDIM,NDIM),BM(NDIM,NDIM),V(NDIM,NDIM),	014770
1 WM3(NDIM,NDIM),COM1(1),COM2(1)	014780
C THIS SUBROUTINE MODIFIES GOGT AND RETURNS THE MODIFIED	014790
C VALUE IN GOGT, WHERE GOGT IS USED IN THE KALMAN FILTER	014800
C GAIN CALCULATIONS. THE MODS ARE IN ACCORDANCE WITH THE	014810
C THE TECHNIQUE DEVELOPED BY O'YLE AND STEIN IN "ROBUSTNESS	014820
C WITH OBSERVERS", IEEE TRANS. ON AUTO. CONTROL, VOIL AC24,	014830
C NO. 4, AUG. 79, PGS 607-611.	014840
C	014850
C THE VALUE RETURNED IN GOGT IS QQ, WHERE QQ IS	014860
C	014870
C QQ=GOGT+SQ(SQ)BM(V(BMT))	014880
C	014890
C SQ IS A SCALAR DESIGN PARAMETER, THAT AS IT APPROACHES	014900
C INFINITY, CAUSES THE LOG CONTROLLER TO RECOVER THE ROBUSTNESS	014910
C PROPERTIES OF A FULL STATE FEEDBACK CONTROLLER.	014920
C THE MATRX--V-- IS ALSO A DESIGN, PARAMETER WITH THE REQUIREMENT	014930
C THAT IT BE POSITIVE DEFINITE. BM---- IS THE CONTROLLER MODEL INPUT	014940
C MATRIX. GOGT --- IS THE CONTROLLER MODEL INPUT NOISE STRENGTH	014950
C MATRIX OM, PREMULTIPLIED BY GM AND POST MULTIPLIED BY GMT WHERE GM IS	014960
C THE INPUT NOISE MATRIX.	014970
CHARACTER MSG*60	014980
COMMON /MAIN1/NDIM,NDIM1,COM1	014990
COMMON /MAIN2/ COM2	015000
COMMON /INOU/ KIN,KOUT,KPUNCH	015010
COMMON /MAU5/ MSG	015020
WRITE(KOUT,11)'	015030
WRITE(KOUT,*)'THIS ROUTINE MODIFIES THE VALUE OF GM(OM)GMT	015040
1 USED IN CALCULATING THE KALMAN FILTER GAIN, RKFSS.'	015050
WRITE(KOUT,*)'THE MODIFIED Q IS = GM(OM)GMT+SQ*SQ(BM)V(BMT)WHERE	015060
1 SQ IS A SCALAR DESIGN PARAMETER AND V IS A POSITIVE DEFINITE	015070
1MATRIX DESIGN PARAMETER. THE LARGER SQ, THE MORE ROBUST THE CONTRO	015080
1L SYSTEM WILL BE.'	015090
DO 5 INP=1,1000	015100
WRITE(KOUT,11)'	015110
11 FORMAT(A10,/)	015120
WRITE(KOUT,*)'ENTER 1-TO INPUT SQ, 2-TO INPUT V 3- TO CALCULATE MODC	015130
1IFIED QQ, 4- TO EXIT THIS ROUTINE>'	015140
READ(KIN,*)ISEL	015150
GO TO (1,2,3,4)ISEL	015160
1 WRITE(KOUT,11)'	015170
WRITE(KOUT,*)'ENTER SQ>0-->'	015180
READ(KIN,*)SQ	015190
GO TO 5	015200
2 WRITE(KOUT,11)'	015210
WRITE (KOUT,*)'V IS INIALIZED TO .ZERO UPON ENTRY INTO THIS OPTIO	015220
1N'	015230
DO 7 I=1,NDIM	015240

```

      DO 7 J=1,NOIM
      V(I,J)=C
      WRITE(KOUT,*) 'ENTER I/O OPTION FOR POSITIVE DEF V (SEE INPUT ROUTINE
1E) >'
      READ(KIN,*) IO
      MSG='DESIGN PARAMETER V MATRIX ENTRIES'
      CALL MMATIO(V,IC6H,IC6H,IC,KIN,KOUT,NOIM,NOIM)
      GO TO 5
3     CALL MAT3(IRFM,IC6H,BP,V,WH3)
C     WH3=BM(V)BNT IRFM X IRFM
      SQ1=SQ*SQ
      CALL MADD1(IRFM,IRFM,GOGT,WP3,V,SQ1)
C     V IRFM X IRFM NOW CONTAINS THE MOD. VALUE ---QG
      MSG='THE DOYLE AND STEIN MODIFIED OO MATRIX IS'
      IO=5
      CALL MMATIO(V,IRFM,IRFM,IC,KIN,KOUT,NOIM,NOIM)
5     CONTINUE
4     CONTINUE
C     QO IS ACCEPTABLE SO PUT INTO GOGT
      DO 20 I=1,IRFM
      DO 20 J=1,IRFM
20    GOGT(I,J)=V(I,J)
      RETURN
      END
      SUBROUTINE PYINTG(PHI,INTGA,INTBA,WME,GA,QA,FA,BA,IRFA,ICGA,IRY,
1IROA)
C     THIS SUBROUTINE SETS UP THE NECESSARY INTEGRALS FOR USE BY
C     THE PERFORMANCE ANAL. ROUTINE. THE STATE TRANSITION MATRIX,
C     EXP(FA*TIME)=EAT, INTEG(EAT(GA) QA (GAT) EATT), AND
C     INTEG(EAT (BA)). WME IS A DUMMY WORK SPACE
      REAL PHI(NOIM2,NOIM2),INTGA(NOIM2,NOIM2),INTBA(NOIM2,NOIM2),
1 WME(NOIM2,NOIM2),QA(NOIM2,NOIM2),BA(NOIM2,NOIM2),
1GA(NOIM2,NOIM2),FA(NOIM2,NOIM2)
      DIMENSION COM1(1),COM2(1)
      COMMON/MAIN1/NOIM,NOIM1,COM1
      COMMON/MAIN2/ COM2
      COMMON /INOUT/ KIN,KOUT,KPUNCH
      COMMON /RNTIM/RATIME,DELTIM
      COMMON /MAIN3/NOIM2,NOIM3
C***FORM GA(QA)GAT --NEED FOR KLIENMAN ROUTINE
C
      NSAV1=NOIM
      NSAV2=NOIM1
      NOIM=NOIM2
      NOIM1=NOIM2+1
      CALL DSCRT(IRFA,FA,DELTIM,INTGA,WME,10)
C     WME=INT(EAT)
      CALL MAT1(WME,BA,IRFA,IRFA,IRY,INTBA)
C     INTBA=INT(EAT)BA IRFA XIRY NEEDED IN MXA UPDATE
      CALL MAT3(IRFA,IRQA,GA,QA,PHI)
C     PHI=GA(QA)(GAT) IRFA X IRFA
      CALL INTEG(IRFA,FA,PHI,INTGA,DELTIM)
C     PHI=EXP(FA) IRFA X IRFA
C     INTGA=INTEGRAL ( EXP(FA) (GA) (QA) (GATT) (EXP(FA) T) ) IRFA A IRFA
C
      NOIM=NSAV1
      NOIM1=NSAV2
      RETURN
      END
* DECK DSCRTZ
      SUBROUTINE DSCRTZ(WH1,PHI,BCYD,BCZD,IRFM,DELTIM,FC,BCY,
1 IRY,BCZ,IRHM,PHIT,OTD,OTD,GT,GT,FT,BT,IRFT,ICGT,ICBT,IRMT,
1RTD,RT,GCX,ICBM)
C     THIS ROUTINE DISCRETIZES A CONTINUOUS TIME LOG CONTROLLER USING
C     FIRST ORDER APPROXIMATIONS TO THE REQUIRED INTEGRALS
C     AND PROVIDES AN EQUIVALENT DISCRETE TIME REPRESENTATION OF THE TRUTH

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1	FT(NDIM,NDIM),FT(NDIM,NDIM),WAX(NDIM,NDIM),	01657
1	WUU(NDIM,NDIM),GCSTR(NDIM,NDIM),RKFS(NDIM,NDIM),	01658
1	WU(NDIM,NDIM),WM1(NDIM,NDIM),WM2(NDIM,NDIM),	01659
1	WM3(NDIM,NDIM),WM4(NDIM,NDIM),WM5(NDIM,NDIM)	01660
	REAL Y0(NDIM)	01661
	INTEGER IFLGCZ	01662
	CHARACTER MSG*60,MSG1*1,MSG2*1	01663
	COMMON /MAIN1/NDIM,NDIM1,COM1	01664
	COMMON /MAIN2/COM2	01665
	COMMON/INCU/KIN,KCUT,KPUNCH	01666
	COMMON /MAIN3/NDIM2,NDIM3	01667
	COMMON/MAUN5/MSG	01668
	COMMON /RNTIM/RNTIME,DELTIM	01669
	COMMON/PAING/ICET,ICBM,ICFA,ICGA,ICGM,ICGT,ICQA,IRFA,IRFM,IRFT,	01670
	IRHT,IRQA,IO,LOG,IRHM,NUMOTS	01671
	WRITE(KOUT,*) 'COLORED INPUT NOISE (Y OR N) >'	01672
	READ(KIN,11)MSG	01673
	IMOLD=IRFM	01674
	IF (MSG.EQ.'N') GO TO 25	01675
25	CALL CNOISE(FM,GM,WM,WM,WM,WM,WM,WM,WM,WM,WM,WM,WM,WM,WM,WM)	01676
	WRITE(KOUT,*) 'ENTER THE TOTAL RUN TIME AND SAMPLE TIME>'	01677
	READ(KIN,*)RNTIME,DELTIM	01678
11	FORMAT(A1)	01679
	C CALCULATE EQUIVALENT DISCRETE TIME VERSIONS OF, BM--BMO,GM--GMO=I,	01680
	C G4--GMO, AND PHIN THE STATE TRANSITION MATRIX FOR FM	01681
	C.....	01682
	C SINCE WORKSPACE IS AT A PREMIUM, THE TRUTH MODEL MATRICES	01683
	C PHIT,BTD,BTO,BTO,RTD,RTD,RTD,RTD WILL BE USED FOR THEIR CONTROLLER MODEL	01684
	C COUNTER PARTS DURING THIS ROUTINE BEFORE THE EQUIVALENT DISCRETE	01685
	C TIME TRUTH MODEL IS COMPUTED. AT THAT TIME THERE IS NO LONGER ANY	01686
	C NEED FOR THOSE CONTROLLER MODEL MATRICES SINCE THE CONTROLLER IS PUT	01687
	C INTO THE PERFORMANCE ANALYSIS FORPAT,PHIC BCY,.....GCZ	01688
	C.....	01689
	NSAV3=NDIM2	01690
	NDIM2=NDIM	01691
	CALL MYINTG(PHIT,BTD,BTD,BTD,RTD,RTD,RTD,RTD,WM,WM,WM,WM,WM,WM,WM,WM)	01692
	C RTD IS USED AS DUMMY WORK SPACE IN CALL TO MYINTG	01693
	NDIM2=NSAV3	01694
	CALL IONT(IRFM,BTD,1.0JJ)	01695
	WRITE(KOUT,*) 'ENTER A C IF THE VALUED ENTERED INTO RM IS A CONTINUOUS	01696
	TIME VALUE TO FORM THE BASIS OF AN APPROXIMATE DISCRETE TIME	01697
	1 RM, ENTER A 0 OTHERWISE>'	01698
	READ(KIN,11)MSG1	01699
	IF (MSG1.EQ.'C') THEN	01700
	C APPROXIMATE RMO=RM/SAMPLE TIME	01701
	C1=1/DELTIM	01702
	CALL MSCALE(RTD,RTD,IRHM,IRHM,C1)	01703
	ELSE	01704
	C THE VALUE IN RM IS DISCRETE TIME ALREADY	01705
	CAL EQUATE(RTD,RTD,IRHM,IRHM)	01706
	END IF	01707
	C SET UP A, S, AND U FOR DUTCON	01708
	CALL XSU(GCZ,GCY,PHIC,GCX,BCY,BCZ,RKFSS,WM1,WM2,WM3,WM4,WM5,	01709
	1 FM,WM,IMOLD,ICBM,WXX,WUU,WXU,PHIT)	01710
	C GCZ NOW CONTAINS X, PHIC CONTAINS U, GCY CONTAINS S	01711
	C GCX,BCZ,BCY,RKFSS WERE DUMMY WORK AREAS IN XSU	01712
	CALL DUTCON(PHIT,WXX,GCX,GCZ,BTD,PHIC,RKFSS,BCY,BCZ,GCY,WUU,	01713
	1GCSTR,IMOLD)	01714
	IF (IMOLD.EQ.0) GO TO 35	01715
	ICH=IMOLD+1	01716
	DO 30 I=1,ICBM	01717
	DO 30 J=ICH,IRFM	01718
30	GCSTR(I,J)=0.	01719
35	CONTINUE	01720
	BCY,BCZ,GCY,PHIC,GCZ,GCX,RKFSS ARE USED AS DUMMY WORK SPACE IN DUTCON	01721
	WRITE(KOUT,*) 'DO YOU WISH TO COMPUTE THE KALMAN FILTER GAIN OF PICO	01722

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1K IT DIRECTLY (AS IN MAYBECK SECTION, 14.5) ENTER A C TO COMPUTE 017230
INTE- A P TO PICK IT DIRECTLY>' 017240
READ (KIN,11)MSG2 017250
IF (MSG1.EQ.'C') THEN 017260
CALL DKFTR(PHIT,BTD,GTD,OTD,GCX,RTD,HH,GCY,GCZ,PHIC,RKFSS,BCY,BCZ, 017270
1FM,GM,GM,3M,WM1,WM2) 017280
CGCX,GCY,GCZ,BCZ,BCY ARE USED AS DUMMY WORK SPACE IN CALL TO DKFTR 017290
ELSE 017300
CALL PKDIRC(GCX,PHIT,GCY,GCZ,PM,BTD,RKFSS,ICBM,IRFM,IRHM) 017310
C GCX,GCY,GCZ ARE, DUMMY WORKSPACES IN CALL TO PKDIRC 017320
END IF 017330
C*****IN THE FOLLOWING CALCULATIONS, BCY,GCY,BCZ ARE DUMMY WORKSP 017340
C*****ACES UNTIL THEIR LAST USE WHEN THEY ARE SET EQUAL TO THEIR F 017350
C*****INAL VALUES FOR PERPAL SUBROUTINE 017360
C1=1.0 017370
CALL IDNT(IRFM,BCY,C1) 017380
CALL MAT1(RKFSS,HH,IRFM,IRHM,IRFM,GCY) 017390
C1=-1.0 017400
CALL MADD1(IRFM,IRFM,BCY,GCY,BCZ,C1) 017410
C BCZ = I-RKFSS(HH) 017420
WRITE(KOUT,*) 017430
WRITE(KOUT,*)'DO YOU WISH TO CALCULATE THE SAMPLED-DATA FILTER P 017440
10LES? Y OR N>' 017450
READ(KIN,12)MSG 017460
IF (MSG.EQ.'Y') THEN 017470
CALL MAT1(PHIT,BCZ,IRFM,IRFM,IRFM,PHIC) 017480
WRITE(KOUT,*)'THE EIGENVALUES THAT CORRESPOND TO THE SAMPLED-DATA 017490
1 FILTER POLES ARE.....' 017500
CALL MEIGN(PHIC,WV1,WV2,IRFM,WM1) 017510
END IF 017520
CALL MSCALE(BCY,GCSTR,ICBM,IRFM,C1) 017530
C BCY=-GCSTR ICBM X IRFM 017540
IF (MSG2.EQ.'C') THEN 017550
C FORMULATE THE OPTIMAL CONTROL LAW FOR PERPAL 017560
CALL MAT1(BCY,BCZ,ICBM,IRFM,IRFM,GCX) 017570
C GCX=-GCSTR(I-RKFSS(HH)) ICBM X IRFM 017580
CALL MAT1(BCY,RKFSS,ICBM,IRFM,IRHM,GCZ) 017590
C GCZ=-GCSTR(RKFSS) ICBM X IRHM 017600
IFLGCZ=J 017610
CALL MAT1(BTD,BCY,IRFM,ICBM,IRFM,GCSTR) 017620
C1=1.0 017630
CALL MADD1(IRFM,IRFM,PHIT,GCSTR,BCY,C1) 017640
C BCY= PHIT-BTD(GCTR) IRFM X IRFM 017650
WRITE(KOUT,*) 017660
WRITE(KOUT,*)'DO YOU WISH TO CALCULATE THE POLES OF THE OPTIMAL 017670
1 LO SAMPLED-DATA CONTROLLER? Y OR N>' 017680
READ(KIN,12)MSG 017690
IF (MSG.EQ.'Y') THEN 017700
WRITE(KOUT,*)'THE EIGENVALUES THAT CORRESPOND TO THE POLES OF THE 017710
1E OPTIMAL LO CONTROLLER ARE.....' 017720
CALL MEIGN(BCY,WV1,WV2,IRFM,PHIC) 017730
END IF 017740
CALL MAT1(BCY,BCZ,IRFM,IRFM,IRFM,PHIC) 017750
C PHIC=(PHIT-BTD(GCTR))(I-RKFSS(HH)) IRFM X IRFM 017760
CALL MAT1(BCY,RKFSS,IRFM,IRFM,IRHM,BCZ) 017770
C BCZ=(PHIT-BTD(GCTR))RKFSS IRFM X IRHM 017780
ELSE 017790
C FORMULATE THE SUBOPTIMAL CONTRL LAW USTR=-GCSTR(X AT T SUB I MINUS) 017800
CALL EQUATE(GCX,BCY,ICBM,IRFM) 017810
C GCX=-GCSTR ICBM X IRFM 017820
CALL MAT1(PHIT,BCZ,IRFM,IRFM,IRFM,GCSTR) 017830
C1=1.0 017840
CALL MAT1(BTD,GCX,IRFM,ICBM,IRFM,BCY) 017850
WRITE(KOUT,*) 017860
WRITE(KOUT,*)'DO YOU WISH TO CALCULATE THE POLES OF THE OPTIMAL 017870
1L LO SAMPLED DATA CONTROLLER? Y OR N>' 017880

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READ(KIN,12)MSG	017390
IF (MSG.EQ.'Y') THEN	017390
CALL MADD1(IRFM,IRFM,PHIT,ECY,PHIC,C1)	017391
WRITE(KOUT,*) 'THE EIGENVALUES THAT CORRESPOND TO THE POLES OF THE	017920
10PTIMAL LOG CONTROLLER ARE.....'	017930
CALL MEIGN(PHIC,MV1,MV2,IRFM,WM1)	017940
END IF	017950
CALL MADD1(IRFM,IRFM,GCST,ECY,PHIC,C1)	017960
C PHIC=PHIT(I-RKFST(H))=BTD,GCSTF IRFM X IPFM	017970
CALL MAT1(PHIT,RKFSS,IRFM,IRFM,IPFM,BCZ)	017980
C BCZ=PHIT(RKFSS) IRFM X IRFM	017990
DO 101 I=1,IRFM	018000
DO 101J=1,IRHM	018010
101 GCZ(I,J)=0	018020
IFLGZ=1	018030
END IF	018040
IRY=1	018050
DO 102 I=1,NDIM	018060
DO 102 J=1,IRY	018070
GCY(I,J)=0	018080
102 BCY(I,J)=0	018090
DO 107 I=1,1000	018100
107 YD(I)=0	018110
IO=5	018120
MSG='PHIC FOR THE SAMPLED DATA CONTROLLER IS'	018130
CALL MMAT10(PHIC,IRFM,IRFM,IO,KIN,KOUT,NDIM,NDIM)	018140
WRITE(KOUT,*)	018150
WRITE(KOUT,*) 'DO YOU WANT TO CALCULATE THE EIGENVALUES OF THE LOG	018160
1 CONTROLLER STATE TRANSITION MATRIX, PHIC? Y OR N>'	018170
READ(KIN,12)MSG	018180
IF (MSG.EQ.'Y') THEN	018190
WRITE(KOUT,*) 'THE EIGENVALUES OF THE LOG CONTROLLER STATE TRANSITION	018200
10N MATRIX ARE.....'	018210
CALL MEIGN(PHIC,MV1,MV2,IRFM,WM1)	018220
END IF	018230
MSG='BCZ FOLLOWS, BCY=0'	018240
CALL MMAT10(BCZ,IRFM,IRFM,IO,KIN,KOUT,NDIM,NDIM)	018250
MSG='GCX FOR THE SAMPLED DATA CONTROLLER IS'	018260
CALL MMAT10(GCX,ICBM,IRFM,IO,KIN,KOUT,NDIM,NDIM)	018270
IF (MSG2.EQ.'C') THEN	018280
MSG='GCZ FOR A COMPUTED KFSS'	018290
ELSE	018300
MSG='GCZ FOR KFSS PICKED DIRECTLY'	018310
END IF	018320
CALL MMAT10(GCZ,ICBM,IRFM,IO,KIN,KOUT,NDIM,NDIM)	018330
WRITE(KOUT,*) 'GCY IS SET = 0'	018340
C THIS IS A REGULATOR SO YD IS ALWAYS ZERO	018350
NSAV3=NDIM-2	018360
NDIM2=NDIM	018370
CALL MYINTG(PHIT,QTD,BTD,RTC,GT,QT,FT,3T,IRFT,ICGT,ICBT,ICGT)	018380
C RTD IS USED AS DUMMY WORK SPACE IN CAL TO MYINTG	018390
NDIM2=NSAV3	018400
C PHIT,QTD,BTD ARE EQUIV. DISCRETE TIME REPRESENTATIONS OF TRUTH MODEL	018410
C MATRICE	018420
WRITE(KOUT,*)	018430
WRITE(KOUT,*) 'WAS THE VALUE ENTERED IN RT DURING INPUT A CONTINUOUS	018440
1US TIME OR A DISCRETE TIME VALUE? ENTER A C FOR CONTINUOUS, A	018450
1 D FOR DISCRETE VALUE>'	018460
READ(KIN,12)MSG	018470
12 FORMAT(A1)	018480
IF (MSG.EQ.'C') THEN	018490
C1=1/DELTIM	018500
CALL MSCALE(RTD,RT,IRHT,IRHT,C1)	018510
ELSE	018520
CALL EQUATE(RTD,RT,IRHT,IPHT)	018530
END IF	018540

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C RTD IS THE DISCRETE TIME APPROX OF :T
C1=1.0
CALL IDNT(IRFT,GTO,C1)
C GTO=GTO + I
RETURN
END
* DECK DKFTR
SUBROUTINE DKFTR(PH1P,BMD,GPD,QMD,WM1,RMD,MM,WM3,WM5,WM6,PKFSS,
1 F2,M2,FH,GM,QN,OP,WM2,WM4)
C THIS ROUTINE CALCULATE THE STEADY STATE KALMAN FILTER GAIN MATRIX
C FOR A SAMPLED ATA CONTROLLER
REAL PH1M(NDIM,NDIM),BMD(NDIP,NDIM),QMD(NDIM,NDIM),
1 GMD(NDIM,NDIM),WP1(NDIM,NDIM),RMD(NDIM,NDIM),
1 MM(NDIP,NDIM), WM3(NDIM,NDIP), WM5(NDIM,NDIM),
1 WM6(NDIP,NDIM), PKFSS(NCIP,NDIM),F2(NDIP,NDIM)
1 M2(NDIM,NDIM),FH(NDIM,NDIP),
1 GM(NDIM,NDIM),QN(NCIN,NDIM),OP(NDIM,NDIM),
1 WM2(NDIM,NDIM),WM4(NDIM,NDIP)
REAL COM1(1),COM2(1)
COMMON /MAIN1/NDIM,NDIM1,COM1
COMMON /INOU/ KIN,KOUT,KPUNCH
COMMON /RATIM/ RNTIME,DELTIM
COMMON /PAIN-/NDIM2,NDIM3
COMMON /MAUN3/MSG
COMMON /MAIN2/ COM2
COMMON /MAIN6/ ICET,ICBM,ICFA,ICGA,ICGM,ICGT,ICGA,IRFA,IRFM,IRFT,
1 IRHT,IRQA,IO,LOG,IRHM,NUMOTS
CHARACTER MSG1*1,MSG*60
C DELETE DETERMINISTIC STATES AS IN THE CONTINUOUS TIME CASE
WRITE(KOUT,*) 'IF YOU PLAN TO USE THE DOYLE AND STEIN TECHNIQUE FORG
1 THIS RUN YOU MAY WISH TO MODIFY THE VALUE OF NUMOTS, THE NUMBER
1F DETERMINISTIC STATES. DO YOU WANT TO CHANGE NUMOTS? Y OR N'
READ(KIN,11)MSG1
NUMSAV=NUMOTS
IF (MSG1.EQ.'Y') THEN
WRITE(KOUT,*) 'ENTER THE NEW VALUE OF NUMOTS FOR THIS RUN'
READ(KIN,*)NUMOTS
END IF
WRITE(KOUT,*) 'NUMOTS= ',NUMOTS
IDS=NUMOTS+1
IRF2=IRFM-NUMOTS
IF (NUMOTS.EQ.0) THEN
C STORE SYSTEM MODEL IN INTERMEDIATE MATRICES COMPATIBLE
C WITH THOSE BELOW WHEN THERE ARE DETERMINISTIC STATES REMOVED
CALL EQUATE(F2,PH1M,IRF2,IRF2)
CALL EQUATE(WM1,GMD,IRF2,IRF2)
CALL EQUATE(M2,FH,IRHM,IRF2)
CALL EQUATE(WM5,BMD,IRF2,ICBM)
CALL EQUATE(WM2,QPD,IRF2,IRF2)
ELSE
C DELETE THE DETERMINISTIC STATES FROM THE MODEL USED TO FORM
C THE STEADY STATE KALMAN FILTER GAIN MATRIX
DO 2112 I=IDS,IRF+
II=I-NUMOTS
DO 2112 J=IDS,IRF+
JJ=J-NUMOTS
2112 WM4((II,JJ))=FH(I,J)
DO 2113 I=1,IRHM
DO 2113 J=IDS,IRF+
JJ=J-NUMOTS
2113 M2(I,JJ)=MM(I,J)
C FORM B2D G2D NOW
DO 2115 I=IDS,IRF+

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II=I-NUOTS	019210
DO 2113 J=1,ICB+	019220
2113 WM3(II,J)=BM(I,J)	019230
DO 2114 I=IDS,I=FP	019240
II=I-NUOTS	019250
DO 2114 J=1,ICGM	019260
2114 WM1(II,J)=GM(I,J)	019270
11 FORMAT(A1)	019280
NSAV=NOIM2	019290
NOIM2=NOIM	019300
CALL MYINTG(F2,WM2,WM5,WM6,WP1,QM,WM4,WM3,IRF2,ICGM,ICBM,ICGM)	019310
NOIM2=NSAV	019320
C1=1.0	019330
CALL IONT(IRF2,WM1,C1)	019340
END IF	019350
C CALCULATE QMD (QMD) QMDT	019360
CALL MAT3(IRF2,IRF2,WM1,WM2,RKFSS)	019370
CRKFSS IS CUMMY WORK SPACE AT THIS POINT IN THE PROGRAM	019380
WRITE(KOUT,*) 'DO YOU WISH TO MODIFY THE QMD MATRIX BY THE DOYLE	019390
1 AND STEIN TECHNIQUE FOR CONTINUOUS TIME CONTROLLERS EXTENDED TO	019400
1 THE DISCRETE TIME SYSTEMS, Y OR N>	019410
READ(KIN,1)MSG1	019420
IF (MSG1.EQ.'Y') THEN	019430
CALL DAS2(BM(IDS,IDS),RKFSS,WM1,ICBM,WM3,IRF2,WM6)	019440
C RETURNS A MODIFIED QMD VALUE TO BE USED IN FINDING RKFSS	019450
END IF	019460
C CALCULATE THE KALMAN FILTER GAINS, RKFSS, FOR EITHER THE MODIFIED	019470
C QMD OR THE UNMODIFIED QMD	019480
C QMD IS STORED IN RKFSS	019490
CALL TRANS2(IRHM,IRF2,H2,WM3)	019500
DO 101 I=1,IRHM	019510
101 WM4(I,1)=QMD(I,1)	019520
CALL KFLTR(IRF2,IRHM,F2,WM3,RKFSS,WM4,WM6,WP1,WM5)	019530
C WM6=PHSS,WM5 CLOSED LOOP MEAS MATRIX	019540
C WM1 = RKFSS WITHOUT THE ZEROS FOR THE DETERMINISTIC STATES	019550
C NOW ADD THE ZEROS FOR THOSE STATES	019560
IF (NUOTS.EQ.0) THEN	019570
DO 2029 I=1,IRFM	019580
DO 2029 J=1,IRHM	019590
2029 RKFSS(I,J)=WM1(I,J)	019600
ELSE	019610
DO 2119 J=1,IRHM	019620
DO 2118 I=1,NUOTS	019630
2118 RKFSS(I,J)=0	019640
DO 2119 I=IDS,IRFM	019650
II=I-NUOTS	019660
2119 RKFSS(I,J)=WM1(II,J)	019670
END IF	019680
C NOW WRITE OUT THE RKFSS MATRIX	019690
27 IO=5	019700
MSG='STEADY STATE SAMPLED DATA KALMAN FILTER GAIN MATRIX'	019710
CALL MMATIO(RKFSS,IRFM,IRHM,IC,KIN,KOUT,NOIM,NOIM)	019720
NUMOTS=NUMSAV	019730
RETURN	019740
END	019750
* DECK DAS2	019760
SUBROUTINE DAS2(BM,QMD,V,ICBP,WM3,IRFM,WM1)	019770
REAL BM(NOIM,NOIM),QMD(NOIM,NOIM), V(NOIM,NOIM),	019780
I WM3(NOIM,NOIM), WM1(NOIM,NOIM)	019790
C THIS ROUTINE ALLOWS QMD TO BE MODIFIED IN A MANNER SIMILAR TO THE DOYLE	019800
C AND STEIN TECHNIQUE FOR CONTINUOUS TIME SYSTEMS.	019810
C QMDMOD=QMD+(SQ*SQ)*(EM(V)BNT)*SAMPLE TIME	019820
C WHERE SQ IS A SCALAR DESIGN PARAMETER AND V IS A POSITIVE DEFINITE	019830
C SYMMETRIC MATRIX DESIGN PARAMETER. AS SQ-->TO INFINITY IN THE CONTINUOUS	019840
C TIME CASE, ROBUSTNESS PROPERTIES OF A FULL STATE FEEDBACK	019850
C CONTROLLER ARE RECOVERED. THIS ROUTINE IS BASED ON E. LLOYD MASTERS	019860

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C THESIS ,031, 1FIT                                019470
  CHARACTER MSG*60                                    019480
  REAL COM1(1),COM2(1)                                019490
  COMMON /MAIN1/NOIM,NOIM1,COM1                      019500
  COMMON /RTIM/RTIME,DELTIM                          019510
  COMMON /MAINS/MSG                                    019520
  COMMON /MAIN2/ COM2                                019530
  COMMON /INOU/KIN,KCUT,KPUNCH                       019540
  WRITE(KOUT,11)' '                                    019550
11 FORMAT(A1)                                          019560
  WRITE(KOUT,*) 'THIS ROUTINE MODIFIES THE VALUE OF QMD USED TO CALCU019570
  LATE THE STEADY STATE KALMAN FILTER GAIN. THE MODIFICATION PERFOR019580
  IMED IS SIMILAR TO THE DOYLE AND STEIN TECHNIQUE FOR CONTINUOUS-TIME019590
  IE SYSTEMS. FOR A COMPLETE DESCRIPTION SEE E. LLOYD MASTERS THESI020000
  IS, 1FIT, 031. BRIEFLY THE MODIFICATION ATTEMPTS TO ENHANCE ROBUST020010
  INESS OF THE LQG CONTROLLER AND IS QMDQ=QMD(QMD)GMOT*SQ*SQ*DELTIM020020
  1*(3M(V)3MT))... THE LARGER THE VALUE CHOSEN FOR THE SCALAR SQ TH020030
  IE MORE ROBUSTNESS RECOVERY (COMPARED TO FULLSTATE FEED-BACK), V M020040
  MUST BE A POSITIVE DEF. MATRIX. CHOOSING 1/E 3005 PSEUDONOISE EQUA020050
  ILLY TO ALL CONTROL INPUTS.'                      020060
  WRITE(KOUT,11)' '                                    020070
  WRITE(KOUT,*) 'ENTER 1-- TO INPUT SQ, 2-- TO INPUT V, 3-- TO    020080
  1 COMPUTE MODIFIED Q, 4-- TO EXIT ROUTINE.....NOTE 1,2 MUST BE 020090
  1 ACCOMPLISHED BEFORE 3, AND 3 BEFORE 4, BUT THAT 1,2,3, CAN BE 020100
  1 DONE ANY NUMBER OF TIMES BEFORE USING 4.'          020110
5  WRITE(KOUT,11)' '                                    020120
  WRITE(KOUT,*) 'ENTER OPTIONS'                        020130
  READ(KIN,*) IOPT                                     020140
  GO TO (1,2,3,4) IOPT                                020150
1  WRITE(KOUT,11)' '                                    020160
  WRITE(KOUT,*) 'ENTER SQ'                             020170
  READ(KIN,*) SQ                                       020180
  GO TO 5                                              020190
2  WRITE(KOUT,11)' '                                    020200
  WRITE(KOUT,*) 'V IS INITIALIZED TO THE IDENTITY MATRIX UPON ENTRY T020210
  IO THIS OPTION. IF YOU DESIRE TO CHANGE V ...REMEMBER IT MUST BE 020220
  1 POSITIVE DEFINITE..... ENTER THE I/O OPTION (I/O OPTIONS ARE 020230
  1 PRINTED AT THE BEGINNING OF THE PROGRAM) ELSE ENTER A : >' 020240
  C1=1.0                                              020250
  CALL IDNT(ICBM,V,C1)                                020260
  READ(KIN,*) IO                                       020270
  IF (IO.EQ.0) THEN                                    020280
  GO TO 5                                              020290
  ELSE                                                020300
  MSG='THE CHOSEN V MATRIX IS'                        020310
  CALL MMAT10(V,ICBM,ICBM,IO,KIN,KOUT,NOIM,NOIM)      020320
  END IF                                              020330
  GO TO 5                                              020340
3  WRITE(KOUT,11)' '                                    020350
  CALL MAT4(V,BM,ICBM,ICBM,IRFM,WM3)                 020360
  CALL MAT1(BM,WM3,IRFM,ICBM,IRFM,WM1)                020370
  C1=SQ*SQ*DELTIM                                     020380
  CALL MADD1(IRFM,IRFM,QMD,C1,WM3,C1)                 020390
  IO=5                                                020400
  MSG='MODIFIED Q MATRIX, QMOD='                     020410
  CALL MMAT10(WM3,IRFM,IRFM,IO,KIN,KOUT,NOIM,NOIM)    020420
  GO TO 5                                              020430
  CALL EQUATE(QMD,WM3,IRFM,IRFM)                     020440
C  REPLACE THE VALUE IN QMD WITH QMOD                020450
  RETURN                                              020460
  END                                                020470
* DECK DOTCON                                         020480
SUBROUTINE DOTCON(PHIN,WMX,WM1,X,BMO,U,WM2,PHIRP,      020490
1 XPRIM,S,WUU,GCSTR,IMOLQ)                          020500
  REAL COM1(1),COM2(1), PHIP(ICIM,NOIM),            020510
  1WAX (NOIM,NOIM),WM1(NOIM,NOIM),X(NOIM,NOIM),      020520

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1 BMD(NDIM,NDIM), U(NDIM,NDIM),W42(NDIM,NDIM),	020530
1 PHIP=M(NDIM,NDIM),X=IM(NDIM,NDIM),S(NDIM,NDIM),	020540
1 WUU(NDIM,NDIM),GCSTR(NDIM,NDIM)	020550
CHARACTER MSG*63	020560
COMMON /MAIN1/NDIM,NDIM1,COM1	020570
COMMON /MAIN2/ COM2	020580
COMMON /MAIN4/NDIM2,NDIM3	020590
COMMON /INOU/KIN,KOUT,KPUNCH	020600
COMMON /RATIM/ RNTIME,DELTIM	020610
COMMON /MAIN6/ICBT,ICBM,ICFA,ICGA,ICGM,ICGT,ICOA,IRFA,IRFM,IPFT,	020620
1 IRMT,IRQA,IO,LOG,IRHM,NUPOTS	020630
COMMON /MAUN5/MSG	020640
C THIS SUBROUTINE COMPUTES THE STEADY STATE OPTIMAL FEEDBACK GAIN MATRIX	020650
CX GCSTR, BASED ON A LINEAR QUADRATIC COST CRITERION, FOR A SAMPLED	020660
C DATA CONTROLLER	020670
C SEE MAYBECK CHAP 14 FOR A DETAILED DISCUSSION OF ALGORITHM AND	020680
C EQUATIONS	020690
C*****	020700
MT=1	020710
C	020720
C TRANSFORM SYSTEM SO KLEINMAN RICCATI SOLVER WILL HANDLE S NOT=0	020730
CALL PRIMIT(W42,U,ICBM,GCSTR,S,IMOLD,BMD,W41,PHIPRM,X,XPRIM,	020740
1 PHIP)	020750
C	020760
C W42=UI*ST	020770
CNOW COMPUTE KPRIM FROM RICCATI EQUATION	020780
CALL MAT3(IMOLD,ICBM,BMD,GCSTR,W41)	020790
CALL DRIC(IMOLD,PHIPRM,W41,XPRIM,X,GCSTR)	020800
C GCSTR CONTAINS INFO THAT IS NOT USED	020810
C X CONTAINS **KPRIM** IRFM X IRFM	020820
C NOW COMPUTE GCSTRPRIM	020830
C1=1.0	020840
CALL MAT3A(ICBM,IMOLD,BMD,X,GCSTR)	020850
CALL MADD1(ICBM,ICBM,U,GCSTR,W41,C1)	020860
CALL GPINV(ICBM,ICBM,W41,GCSTR,MR,MT)	020870
IF (MR.NE.ICBM) THEN	020880
PRINT *, 'INVERSE OF U IN DDTCN NOT OF FULL RANK, RANK IS ',	020890
1 MR, ' RANK SHOULD BE =ICBM=',ICBM	020900
END IF	020910
CALL MAT4(GCSTR,BMD,ICBM,ICBM,IMOLD,W41)	020920
CALL MAT1(W41,X,ICBM,IMOLD,IMOLD,GCSTR)	020930
CALL MAT1(GCSTR,PHIPRM,ICBM,IMOLD,IMOLD,W41)	020940
C W41=GCSTRPRIM=((U+BMD(KPRIM)BMD)) (BMD(KPRIM)PHIPRM)	020950
C ICBM X IRFM	020960
CALL MADD1(ICBM,IMOLD,W41,W42,GCSTR,C1)	020970
C GCSTR ICBM X ICBM	020980
IO=5	020990
MSG='THE OPTIMAL STEADY STATE FEEDBACK GAIN MATRIX,GCSTR'	021000
CALL MAT10(GCSTR,ICBM,IMOLD,IO,KIN,KOUT,NDIM,NDIM)	021010
RETURN	021020
END	021030
* DECK PKDIRC	021040
SUBROUTINE PKDIRC(W,PHIP,W41,W42,W4,BMD,RKFSS,ICBM,IRFM,IRHM)	021050
REAL COM1(1),COM2(1),W(NDIM,NDIM),	021060
1 PHIP(NDIM,NDIM), W41(NDIM,NDIM), W42(NDIM,NDIM),	021070
1 W4(NDIM,NDIM), BMD(NDIM,NDIM), RKFSS(NDIM,NDIM)	021080
CHARACTER MSG*63,PSG1*1	021090
COMMON /MAIN1/NDIM,NDIM1,COM1	021100
COMMON /MAIN2/COM2	021110
COMMON /MAUN5/ MSG	021120
COMMON /INOU/ KIN, KOUT,KPUNCH	021130
C AS IN MAYBECK, SECTION 14.5, RKFSS=PHIP(BMD)W *SG. SO IS A	021140
C SCALAR DESIGN PARAMETER AND W IS ANY NONSINGULAR M X M MATRIX.	021150
C MAYBECK SUGGESTS THAT W=(M4(PHIP)BMD) IS A POSSIBLE CHOICE.	021160
C THE RKFSS PICKED AS A RESULT OF THIS ALGORITHM FORMS THE BASIS	021170
C OF A SUBOPTIMAL CONTROL LAW,USTAR=GCSTR*(A(TI-MINUS)	021180

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WRITE(KOUT,12) ' ' 021190
WRITE(KOUT,*) 'THIS ROUTINE CALCULATES THE STEADY STATE KALMAN FILTER GAIN DIRECTLY ( THAT IS WITH OUT USE FMSS FROM THE MATRIX RICCATI EQUATION AS THE BASIS OF KFSS) I EQUATION AS IN SECTION 14.5 .MAYBECK. KFSS=SQ*(PHIM)BMD(M) WHERE THE SCALAR SQ AND THE MATRIX 021200
1 1 ARE DESIGN PARAMETERS...THE LARGER THE SQ THE MORE ROBUSTNESS THE SUBSEQUENT CONTROLLER WILL HAVE. NOTE THERE ARE NO STABILITY CLAIMS FOR THE RESULTING CONTROLLER. SO BE SURE TO CHECK THE EIGENVALUES OF THE SUBSEQUENT CONTROLLER.' 021210
12 WRITE(KOUT,12) ' ' 021220
FORMAT(A1/) 021230
WRITE(KOUT,*) 'THE OPTIONS FOR THIS ROUTINE ARE 1->CHOOSE SQ, 2-> CHOOSE W, 3-> COMPUTE AND PRINT RKFSS, 4-> EXIT ROUTINE.....' 021240
5 WRITE(KOUT,12) ' ' 021250
WRITE(KOUT,*) 'ENTER OPTION>' 021260
READ(KIN,*) IOPT 021270
GO TO (1,2,3,4) IOPT 021280
GO TO 4 021290
1 WRITE(KOUT,12) ' ' 021300
WRITE(KOUT,*) 'ENTER A VALUE FOR SQ. LARGER SQ GIVE BETTER ROBUSTNESS>' 021310
1SS>' 021320
READ(KIN,*) SQ 021330
GO TO 5 021340
2 WRITE(KOUT,12) ' ' 021350
WRITE(KOUT,*) 'DO YOU WANT TO PICK W ARBITRARILY OR DO YOU WANT W TO BE HM(PHIM)BMD --INVERSE AS IN MAYBECK SECTION 14.5. NOTE THAT 1 IRMP MUST BE EQUAL TO ICBM SINCE W MUST BE ICBM X ICBM. ENTER 1 AN A FOR ARBITRARY. W OTHERWISE>' 021360
READ(KIN,11) MSG1 021370
11 FORMAT(A1) 021380
IF (MSG1.EQ.'A') THEN 021390
WRITE(KOUT,*) 'ENTER I/O OPTION FOR W(SEE INPUT ROUTINE FOR EXPLANATION OF INPUT OPTIONS 1,2,3,4,5,6)>' 021400
READ(KIN,*) IO 021410
MSG='ARBITRARY W MATRIX' 021420
CALL MMATIO(W,ICBM,ICBM,IC,KIN,KOUT,NOIM,NOIM) 021430
ELSE 021440
C COMPUTE W AS DESCRIBED ABOVE 021450
IF (ICBM.NE.IRHM) THEN 021460
WRITE(KOUT,*) 'NOTE THAT IRHM MUST EQUAL ICBM TO USE THIS METHOD OF 1 CALCULATING W. ICBM=' ,ICBM, ' IRHM=' ,IRHM 021470
GO TO 5 021480
END IF 021490
CALL EQUATE(WM1,PHIM,IRFM,IRFM) 021500
C GMINV DESTROYS THE CALLING ARRAY 021510
MT=1 021520
CALL GMINV(IRFM,IRFM,WM1,WM2,MR,MT) 021530
CALL MAT1(WM1,WM2,IRHM,IRFM,IRFM,WP1) 021540
CALL MAT1(WM1,BMD,IRHM,IRFM,ICBM,WM2) 021550
CALL GMINV(IRHM,ICBM,WM2,W,MR,MT) 021560
END IF 021570
GO TO 5 021580
3 WRITE(KOUT,12) ' ' 021590
C CALCULATE RKFSS 021600
CALL EQUATE(WM1,PHIM,IRFM,IRFM) 021610
C GMINV DESTROYS CALLING ARRAY 021620
MT=1 021630
CALL GMINV(IRFM,IRFM,WM1,WM2,MR,MT) 021640
CALL MAT1(WM2,BMD,IRFM,IRFM,ICBM,WM1) 021650
CALL MAT1(WM1,W,IRFM,ICBM,ICBM,WM2) 021660
CALL MSKALE(RKFSS,WM2,IRFM,IRFM,SC) 021670
IO=5 021680
MSG='RKFSS, PICKED DIRECTLY IS' 021690
CALL MMATIO(RKFSS,IRFM,IRHM,IC,KIN,KOUT,NOIM,NOIM) 021700
GO TO 5 021710
4 RETURN 021720

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      END
      * DECK PRINT
      SUBROUTINE PRIMIT(WM2,U,ICBM,GCSTR,S,IRFM,BND,WM1,PHIPRM,X,XPRIM,
      1 PHIP)
      C THIS SUBROUTINE COMPUTES THE PRIMED QUANTITIES NEEDED WHEN USING
      C KLEINMAN RICCATI SOLVER WITH ACN ZERO CROSS COST WEIGHTING
      C MATRIX W4U
      REAL WM2(NDIM,NDIM),GCSTR(NDIM,NDIM),S(NDIM,NDIM),
      1BFO(NDIM,NDIM),WM1(NDIM,NDIM),PHIPRM(NDIM,NDIM),
      1X(NDIM,NDIM),XFRM(NDIM,NDIM),PHIM(NDIM,NDIM)
      REAL COM1(1),COM2(1)
      COMMON /MAIN1/ NDIM,NDIM1,COM1
      COMMON /MAIN2/COM2
      COMMON /INOU/KIN,KOUT,KPUNCH
      C NOW COMPUTE XPRIP, PHIPRM
      CALL EQUATE(WM2,U,ICBM,ICBM)
      C GMINV DESTROYS THE CALLING ARRAY
      MT=1
      CALL GMINV(ICBM,ICBM,WM2,GCSTR,MR,MT)
      C GCSTR= UI ICBM X ICBM
      CALL MAT*(GCSTR,S,ICBM,ICBM,IRFM,WM2)
      CWM2= UI(IST) ICBM X IRFM
      CALL MAT1(BMD,WM2,IRFM,ICBM,IRFM,WM1)
      C1=-1.0
      CALL MADD1(IRFM,IRFM,PHIM,WM1,PHIPRM,C1)
      C PHIPRM= PHIM-BND(UI)ST IRFM X IRFM
      CALL MAT1(S,WM2,IRFM,ICBM,IRFM,WM1)
      CALL MADD1(IRFM,IRFM,X,WM1,XPRIP,C1)
      C XPRIP= X-S(UI)ST IRFM X IPFM
      C
      RETURN
      END
      * DECK RGS
      SUBROUTINE RGS(GCSTR,RKFSS,GCA,GCY,GCZ,BCY,BCZ,FC,YD,
      1 WM1,WM2,WM3,WM4,WM5,WM6,WM7,WM8,WM9,WM10,WV1,WV2,
      1 WMA,WMB,MMC,MMD,MME,MMF,FT,ET,GT,HT,OT,RT,
      1 FM,GM,GN,HN,HO,FO,MXA,WUU,WXU,FA,BA,GA,GUA,
      1MAA,FXA,FA,FXA,GCZA,TRY,IFLGZ,IFLGSD,
      1MU,PUMAX,PUMIN,PXTMAX,PXTMIN,MUMAX,MUMIN,MXAMAX,MXAMIN,
      1PUOUT,PXTOUT,MXTOUT,MUOUT,WV3,WV4)
      REAL COM1(1),COM2(1)
      C
      C THIS SUBROUTINE SETS UP A CONTINUOUS, DISCRETIZED CONTINUOUS OF A
      C SAMPLED DATA LOG CONTROLLER BASED ON USERS REQUEST. EACH
      C CONTROLLER IS PUT INTO THE PROPER FORMAT FOR THE PERFORMANCE
      C ANALYSIS SUBROUTINE, PERPAL.
      CHARACTER MSG*60
      REAL FT (NDIM,NDIM),ST(NDIM,NDIM),GT(NDIM,NDIM),
      1 HT (NDIM,NDIM),RM (NDIM,NDIM),FM(NDIM,NDIM),
      1BM(NDIM,NDIM),GM (NDIM,NDIM),XC(NDIM),
      1HM(NDIM,NDIM),OM (NDIM,NDIM),PC(NDIM,NDIM),
      1OT(NDIM,NDIM),RT (NDIM,NDIM),
      1GCSTR(NDIM,NDIM),RKFSS(NDIM,NDIM),
      1WV1(NDIM),WV2(NDIM),WXU(NDIM,NDIM),
      1WM1 (NDIM,NDIM),WM2 (NDIM,NDIM),WM3(NDIM,NDIM),
      1WM4 (NDIM,NDIM),WM5 (NDIM,NDIM),WM6(NDIM,NDIM),
      1WM7 (NDIM,NDIM),WM8 (NDIM,NDIM),WM9(NDIM,NDIM),
      1WM10 (NDIM,NDIM),WMA(NDIM2,NDIM2),WMB(NDIM2,NDIM2),
      1MMC(NDIM2,NDIM2),MMD(NDIM2,NDIM2),MME(NDIM2,NDIM2),
      1MMF(NDIM2,NDIM2),WUU(NDIM,NDIM),WXA(NDIM,NDIM),
      1FA(NDIM2,NDIM2),BA(NDIM2,NDIM2),GA(NDIM2,NDIM2),
      1MAA(NDIM2),GCX(NDIM,NDIM),GCZ(NDIM,NDIM),
      1QA(NDIM2,NDIM2),PXA(NDIM2,NDIM2),FXA(NDIM2,NDIM2),
      1MUOUT(NDIM),MXTOUT(NDIM),PXTOUT(NDIM)
      REAL FUCUT(NDIM),YD(NDIM),GUA(NDIM2,NDIM2),

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1) AMIN(NDIM), MAXMAX(NDIM), MUPIN(NDIM),      022510
1) MU4AX(NDIM), P1TMEN(NDIM), PATMAX(NDIM),      022520
1) PUMIN(NDIM), PUMAX(NDIM), GCZA(NDIM2,NDIM2),    022530
1) RA(NDIM2,NDIM2), BCY(NDIM,NDIM), BCZ(NDIM,NDIM), 022540
1) WV3(NDIM2), WV4(NDIM2),                      022550
1) GCY(NDIM,NDIM), FC(NDIM,NDIM)                022560
-----
INTEGER IFLGCZ,IRY                               022570
REAL MU(NDIM)                                    022580
COMMON /RNTIM/ RNTIME,DELTIM                     022590
COMMON /MAIN2/COM2                                022600
COMMON /MAIN1/NDIM,NDIM1,CCP1                     022610
COMMON / INOU/ KIN,KOUT,KPUNCH                     022620
COMMON /MAIN4/NDIP2,NDIM3                          022630
COMMON /MAUN5/ MSG                                022640
COMMON /MAIN6/ ICBT,ICBM,ICFA,ICGA,ICGM,ICGT,ICQA,IRFA,IRFM,IRFT, 022650
1 IRHT,IRQA,IO,LOG,IRHM,KUPDTS                    022660
-----
C                                                    022670
  WRITE(KOUT,*) 'ENTER A C FOR CONTINUOUS TIME LOG CONTROLLER AND A 022680
1 D FOR A SAMPLED DATA LOG CONTROLLER'          022690
  READ(KIN,12,END=2933)MSG                         022700
  IF (MSG.EQ.'C') THEN                             022710
    IFLGSD=0                                         022720
    CALL CLOGRS(GCSTR,FM,BM,RKFSS,MP,GCX,GCY,GCZ,BCY,ECZ,FC,YD,    022730
1 RM,OM,FT,BT,GT,CT,RT,MT,                      022740
1 IRY,IFLGCZ,WM1,WM2,WM3                          022750
1 ,PO,GM,WM4,WM5,WM6,WV1,WV2,WUU,WXX,XO,WXU,WM7,WM8) 022760
    IF (IO.EQ.0) THEN                              022770
      GO TO 2933                                    022780
    END IF                                           022790
    CALL FRMAUG(OT,RT,FT,BT,GCZ,MT,GCX,BCZ,FC,GCY,BCY,GT,XO,PO,FA,BA, 022800
1 GA,QA,QA,WM1,WM2,WM3,WM4,WM5,WM6,WM7,WM8,WMF,    022810
1 MAA,PA,PA,PA,GCZA,IRY,IFLGCZ,IFLGS0)           022820
    WRITE(KOUT,*) ' '                               022830
    WRITE(KOUT,*) 'DO YOU WISH TO CALCULATE THE EIGENVALUES OF THE CLOS 022840
1 ED-LOOP F MATRIX? Y OR N'                       022850
    READ(KIN,12)MSG                                022860
    IF (MSG.EQ.'Y') THEN                            022870
      NSAV1=NDIM                                    022880
      NDIM=NDIM2                                     022890
      NDIM1=NDIM2+1                                 022900
      WRITE(KOUT,*) 'THE EIGENVALUES OF THE CLOSED-LOOP F MATRIX ARE...' 022910
      CALL MEIGN(FA,WV3,WV4,IRFA,WPA)               022920
      NDIM=NSAV1                                    022930
      NDIM1=NDIM+1                                  022940
      END IF                                         022950
      CALL HYINTG(WMA,WMB,WMC,WME,GA,QA,FA,BA,IRFA,ICGA,IRY,IRQA) 022960
      ELSE                                           022970
C SOME SAMPLED DATA CONTROLLER IS WANTED          022980
  IFLGSD=1                                           022990
  WRITE(KOUT,*) 'DO YOU WISH TO MERELY DISCRETIZE THE CONTINUOUS TIME 023000
1 CONTROLLER, Y OR N'                             023010
  READ(KIN,12,END=2933)MSG                         023020
  IF (MSG.EQ.'Y') THEN                             023030
    WRITE(KOUT,*) '*****NOTE THAT WHEN ENTERING THE TIME INCREMENT 023040
1 IN THE CONTINUOUS TIME CONTROLLER SET UP, REMEMBER IT WILL BECO 023050
1 ME THE SAMPLE TIME FOR THE DISCRETIZED CONTROLLER*****' 023060
    CALL CLOGRS(GCSTR,FM,BM,RKFSS,MP,GCX,GCY,GCZ,BCY,BCZ,FC,YD,    023070
1 RM,OM,FT,BT,GT,CT,RT,MT,IRY,IFLGCZ,WM1,WM2,WM3,PO,GM,    023080
1 WM4,WM5,WM6,WV1,WV2,WUU,WXX,XO,WXU,WM7,WM8) 023090
    IF (IO.EQ.0) THEN                              023100
      GO TO 2933                                    023110
    END IF                                           023120
    CALL DISCRTZ(WM1,WM2,WM3,WM4,IRFM,DELTIM,FC,BCY,IRY,BCZ,IRMY,    023130
1 WM5,WM6,GCSTR,GT,OT,FT,BT,IRFT,ICGT,ICBT,IRHT,RKFSS,RT,GCX,    023140
1 ICBM)                                             023150
    WRITE(KOUT,*) ' '                               023160

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      WRITE(KOUT,*) 'DO YOU WANT TO CALCULATE THE EIGENVALUES OF THE STATE TRANSITION
      MATRIX FOR THE DISCRETIZED CONTROLLER? Y OR N?'
      READ(KIN,12)MSG
      IF (MSG.EQ.'Y') THEN
        WRITE(KOUT,*) 'THE EIGENVALUES OF THE STATE TRANSITION MATRIX FOR THE
        DISCRETIZED CONTROLLER ARE...'
        CALL MEIGN(WM2,WV1,WV2,IRFM,BCY)
        END IF
      C RKFSS, IN PRECEDING CALL STATEMENT ARE MERELY DUMMY WORK SPACES
      C GCSTR CONTAINS BCZD UPON RETURN FROM DSCRTZ
      CALL FRMAUG(WM6,RKFSS,WM5,GCSTR,GCZ,HT,GCX,WM4,WM2,GCY,WM3,WM1,XO,
      1 PJ,FA,BA,GA,CA,GUA,BCY,FC,WPA,WMB,WMC,WMD,WME,WPF,MXA,PXA,RA,
      1 PXVA,GCZA,IRY,IFLGZ,IFLGSD)
      ELSE
        IFLGZ=0
        CALL DLOGRS(GCX,GCY,GCZ,BCY,BCZ,WM3,FC,WM2,WM5,WM1,WM4,FM,BM,
      1 OM,GM,RM,WM,GT,QT,FT,BT,RT,MT,WXX,WUU,GCSTR,RKFSS,YD,
      1 IRY,IFLGZ,WXU,WM6,WM7,WM8,WM9,WM1,WV1,WV2)
        CALL FRMAUG(WM1,WP2,WM3,WM4,GCZ,HT,GCX,BCZ,FC,GCY,BCY,WM5,XO,
      1 PJ,FA,BA,GA,CA,GUA,RKFSS,GCSTR,WMA,WMB,WMC,WMD,WME,WMF,MXA,PXA,
      1 RA,PXVA,GCZA,IRY,IFLGZ,IFLGSD)
      C RKFSS,GCSTR ARE DUMMY WORK SPACE IN CALL TO FRMAUG
      END IF
      NSAV1=NDIM
      NSAV2=NDIM1
      NDIM1=NDIM2+1
      NDIM=NDIM2
      CALL EQUATE(WMA,FA,IRFA,IRFA)
      CALL EQUATE(WMC,BA,IRFA,IRY)
      CALL MAT1(GA,QA,IRFA,IRQA,IRQA,WME)
      CALL MAT4(WME,GA,IRFA,IRQA,IRFA,WMB)
      NDIM=NSAV1
      NDIM1=NSAV2
      C NOTE FC AND BCY IN CALL TO FRMAUG ARE DUMMY WORKSPACES
      END IF
      RETURN
2933 IO=0
      12 FC=MAT(41)
      END
* DECK XSU
      SUBROUTINE XSU(X,S,U,PHIJC,BJC,PHII,INTPII,PHIJ,BJ,
      1 TEMP,TEMP1,TEMP2,FM,BM,IRFM,ICM,WXX,WUU,WXU,PHIT)
      C THIS ROUTINE COMPUTES X'S AND U AT TIME T-SUB-I FOR USE IN THE
      C SAMPLED DATA CONTROLLER/DETERMINISTIC GAIN CALCULATION.
      C THIS ROUTINE APPROXIMATES THE INTEGRALS REQUIRED (SEE MAYBECK,
      C EQUATIONS 14.25) BY TREATING TIME VARYING ENTITIES IN THE
      C INTEGRANDS AS CONSTANTS OVER SOME SUBINTERVAL OF THE SAMPLE TIME
      C THAT IS CHOSEN BY THE USER
      REAL X (NDIM,NDIM), S(NDIM,NDIM), U(NDIM,NDIM),
      1PHIJC(NDIM,NDIM),BJC(NDIM,NDIM),PHII(NDIM,NDIM),
      1INTPII (NDIM,NDIM),PHIJ(NDIM,NDIM),BJ(NDIM,NDIM),
      1TEMP (NDIM,NDIM),TEMP1(NDIM,NDIM),TEMP2(NDIM,NDIM),
      1FM(NDIM,NDIM),BM (NDIM,NDIM),WXX(NDIM,NDIM),
      1WUU (NDIM,NDIM),WXU(NDIM,NDIM),PHIT(NDIM,NDIM)
      REAL COM1(1),COM2(1)
      CHARACTER MSG*6
      COMMON /PAIN1/NDIM,NDIM1,CCP1
      COMMON/PAIN2/COM2
      COMMON /INQU/ KIN,KOUT,KPLNCH
      COMMON /RNTIM/ RNTIME,DELTI
      COMMON /MAUNS/ MSG
      IZIT=0
      DO 702 IJK=1,1000
      C GIVE USER UP TO 1000 CHANCES TO CHOOSE DIFFERENT SUBINTERVAL LENGTH
      IF (IZIT.EQ.1) THEN
        WRITE(KOUT,12) ' '

```

```

WRITE(KOUT,*) 'DO YOU WISH TO RECOMPUTE X,S,U, BASED ON A DIFFERENT
1 SUBINTERVAL LENGTH, Y OR N?'
READ(KIN,*) MSG
11 FORMAT(21)
12 FORMAT(21,/)
IF (MSG.EQ.'N') THEN
RETURN
END IF
END IF
IZIT=1
WRITE(KOUT,*) 'ENTER THE NUMBER OF SUBINTERVALS TO USE IN THE APPROX
IMATIONS OF INTEGRALS NEEDED TO CALCULATE X, S, AND U (SUGGEST
1 OR MORE )>'
READ(KIN,*) INTVAL
C NOW INITIALIZE VARIABLES REQUIRED IN CALCULATIONS
C1=1.0
CALL IONT(IRFM,PHIJO,C1)
C1=0.0
CALL IONT(IRFM,X,C1)
CALL IONT(ICBM,U,C1)
DO 13 I=1,IRFM
DO 13 J=1,ICBM
BJO(I,J)=0
13 S(I,J)=0
C INITIALIZATION COMPLETE. NOW COMPUTE PHIJ, INTPHJ FOR SUBINTERVAL
C PHI,INTPHI ARE APPROXIMATED BY TAKING AVERAGE OF VALUES AT THE
C BEGINNING,END AND 8 POINTS IN THE MIDDLE OF EACH SUBINTERVAL
C THIS MEANS 9 SUB SUB INTERVAL POINTS TO BE CALCULATED. HOWEVER, 0
C ONLY 1 CALL TO INTEGRATE ROUTINES REQUIRED SINCE FM IS A CONSTANT
C MATRIX
DEL=1.0
SUBINT=DELIM/(4*INTVAL)
DO 23 JK=1,INTVAL
C COMPUTE PHIJ BJ AND THEN UPDATE PHIJC,BJO
C FOR EACH SUBINTERVAL
C1=1.0
DO 371 INTVL=1,4
DEL=DEL+SUBINT
CALL DSCRT(IRFM,FM,DEL,PHII,INTPII,5)
CALL MA001(IRFM,IRFM,PHIJC,PHII,TEMP,C1)
CALL EQUATE(PHIJO,TEMP,IRFM,IRFM)
C PHIJC=PHIJO+PHII
CALL MAT1(INTPII,8M,IRFM,IRFM,ICBM,TEMP1)
CALL MA001(IRFM,ICBM,BJO,TEMP1,TEMP,C1)
CALL EQUATE(BJO,TEMP,IRFM,ICBM)
C BJO=BJO+INTPII*8M
371 CONTINUE
C NOW CALCULATE BJ PHIJ
C1=0.2
CALL MSCALE(PHIJ,PHIJC,IRFM,IRFM,C1)
CALL MSCALE(BJ,BJO,IRFM,ICBM,C1)
C BJ, PHIJ NOW AVAILABLE FOR THIS SUBINTERVAL
C RESET PHIJO,BJO
CALL EQUATE(PHIJO,PHII,IRFM,IRFM)
CALL EQUATE(BJO,TEMP2,IRFM,ICBM)
C NOW UPDATE X,S,U FOR THIS SUBINTERVAL
C X=SUM OF (PHIJT*WXX*PHIJ*(DELIM/INTVAL)) FOR ALL JK
C S=SUM OF ((PHIJT*WXX*BJ+PHIJT*WU)*(DELIM/INTVAL))
CALL MAT4(PHIJ,WXX,IRFM,IRFM,IRFM,TEMP)
C TEMP= PHIJT * WXX
CALL MAT1(TEMP,PHIJ,IRFM,IRFM,IRFM,TEMP1)
CALL MAT1(TEMP,BJ,IRFM,IRFM,ICBM,TEMP2)
C1A=DELIM/INTVAL
CALL MA001(IRFM,IRFM,X,TEMP1,TEMP,C1A)
CALL EQUATE(X,TEMP,IRFM,IRFM)
C X IS NOW UPDATED FOR THIS SUB-SUB INTERVAL

```

CALL MAT4A(PHIJ,XXU,IFF,IFF,ICBM,TEMP1)	024590
C1=1.0	024591
CALL MADD1(IRFM,ICBM,TEMP2,TEMP1,TEMP,C1)	024592
CALL MADD1(IRFM,ICBM,S,TEMP,TEMP1,C1A)	024593
CALL EQUATE(S,TEMP1,IFF,ICBM)	024594
C S UPDATE FOR THIS SUB-SUB INTERVAL NOW COMPLETE	024595
C	024596
C U=SUM OF ((BJT*XX*BJ+WUU*BJT*XXU+XXUT*BJ)*(DELTIM/INTVAL)) FOR ALL JK	024597
C1=1.0	024598
CALL MAT3A(ICBM,IRFM,BJ,XX,TEMP)	024599
CALL MADD1(ICBM,ICBM,TEMP,WUU,TEMP1,C1)	024600
CALL MAT4A(BJ,XXU,ICBM,IRFM,ICBM,TEMP)	024601
CALL MADD1(ICBM,ICBM,TEMP1,TEMP,TEMP2,C1)	024602
CALL MAT4A(XXU,BJ,ICBM,IRFM,ICBM,TEMP)	024603
CALL MADD1(ICBM,ICBM,TEMP2,TEMP,TEMP1,C1)	024604
CALL MADD1(ICBM,ICBM,U,TEMP1,TEMP,C1A)	024605
CALL EQUATE(U,TEMP,ICBM,ICBM)	024606
C U UPDATE FOR THIS SUB-SUB INTERVAL NOW COMPLETE	024607
23 CONTINUE	024608
IO=5	024609
MSG='X(TI) IS'	024610
CALL MMATIO(X,IRFM,IRFM,IO,KIN,KOUT,NOIM,NOIM)	024611
MSG='U(TI) IS'	024612
CALL MMATIO(U,ICBM,ICBM,IO,KIN,KOUT,NOIM,NOIM)	024613
MSG='S(TI) IS'	024614
CALL MMATIO(S,IRFM,ICBM,IC,KIN,KOUT,NOIM,NOIM)	024615
782 CONTINUE	024616
RETURN	024617
END	024618
SUBROUTINE CNOISE(FM,GM,HN,OP,OM,IRFM,IRHM,ICBM,ICGM,WM1,WM2,WM3)	024619
DIMENSION FM(NDIM,NOIM),HM(NDIM,NOIM),BM(NDIM,NOIM),GM(NDIM,NOIM)	024620
1GM(NDIM,NOIM),COM1(1),WM1(NDIM,NOIM),WM2(NDIM,NOIM),WM3(NDIM,NOIM)	024621
COMMON/MAIN1/NOIM,NOIM1,COM1	024622
COMMON/INC1/KIN,KOUT,KPUNCH	024623
25 FORMAT(I1)	024624
WRITE(KOUT,*) '1ST OR 2ND ORDER SHAPING FILTER (1 OF 2) >'	024625
READ(KIN,25) ISIZE	024626
C CALCULATE THE DIMENSION OF THE AUGMENTED SYSTEM	024627
IRFS=IRFM+(ISIZE*ICBM)	024628
IF (ISIZE.NE.1) GO TO 10	024629
CALL FORDER(FM,BM,GM,OM,HN,IRFM,IRHM,ICBM,ICGM,IRFS,WM1)	024630
GO TO 20	024631
10 CALL SORDER(FM,BM,GM,OM,PP,IRFM,IRHM,ICBM,ICGM,IRFS,WM1,WM2,WM3)	024632
20 CONTINUE	024633
RETURN	024634
END	024635
SUBROUTINE FORDER(FM,BM,GM,OP,HP,IRFM,IRHM,ICBM,ICGM,IRFS,MU)	024636
DIMENSION FM(NDIM,NOIM),BM(NDIM,NOIM),GM(NDIM,NOIM),OM(NDIM,NOIM)	024637
1HM(NDIM,NOIM),COM1(1),COM2(1),MU(NDIM,NOIM)	024638
COMMON/MAIN1/NOIM,NOIM1,COM1	024639
COMMON/MAIN2/COM2	024640
COMMON/INC1/KIN,KOUT,KPUNCH	024641
COMMON/MAUN5/MSG	024642
REAL ENTRY,ENTRY1	024643
CHARACTER MSG*63	024644
C ZERO OUT BOTTOM PARTITIONS OF GM,BM,GM	024645
IFS=IRFM+1	024646
IGS=ICGM+1	024647
ICGS=ICGM+ICBM	024648
CALL ZPART(FM(IFS,1),ICBM,IRFS)	024649
CALL ZPART(BM(IFS,1),ICBM,ICBM)	024650
CALL ZPART(GM(IFS,1),ICBM,ICGS)	024651
CALL ZPART(HM(1,IFS),IRHM,ICBM)	024652
CALL ZPART(OM(IGS,1),ICBM,ICGS)	024653
CALL ZPART(MU(1,IGS),ICGM,ICBM)	024654
CALL ZPART(MU,ICBM,ICBM)	024655

C	DESIGN ONE SHAPING FILTER FOR EACH CONTROL CHANNEL	025150
	DO 60 IF=1,ICBM	025160
	WRITE(KOUT,*)'FILTER DESIGN',IF	025170
	WRITE(KOUT,*)'INPUT OPTIONS: (1) ENTER A AND 3,	025180
	1(2) ENTER Q, (3) GO TO NEXT FILTER OR EXIT IF DONE'	025190
	DO 90 IF=1,50	025200
27	FORMAT(I1)	025210
	WRITE(KOUT,*)'OPTION>'	025220
	READ(KIN,27)IOPT	025230
	GO TO (1,2,3)IOPT	025240
1	WRITE(KOUT,*)'ENTER A AND B>'	025250
	READ(KIN,*)ENTRY,ENTRY1	025260
	HU(IF,IF)=ENTRY-ENTRY1	025270
	FM((IRFM+IF), (IRFM+IF))=-1*ENTRY1	025280
	GO TO 90	025290
2	WRITE(KOUT,*)'ENTER Q>'	025300
	READ(KIN,*)ENTRY	025310
	QM((ICGM+IF), (ICGM+IF))=ENTRY	025320
	GO TO 90	025330
90	CONTINUE	025340
3	CONTINUE	025350
80	CONTINUE	025360
C	CALCULATE BM*HU AND PUT IN UPPER RIGHT PARTITION OF FM	025370
	CALL MAT1(BM,HU,IFFM,ICBM,ICBP,FM(1,IFS))	025380
C	INSERT BM INTO UPPER RIGHT PARTITION OF GM, DU=I	025390
	CALL EQUATE(GM(1,IGS),BP,IRFP,ICBP)	025400
C	PUT AN IDENTITY MATRIX IN LOWER RIGHT PARTITION OF GM	025410
	CALL IDNT(ICBM,GM(IGS,IGS),1.)	025420
C	CHANGE COLUMN DIMENSION OF GM AND ORDER OF SYSTEM	025430
	ICGM=IGS	025440
	IFFM=IRFS	025450
	IHOLD=IO	025460
	IO=5	025470
	MSG='QMAUG'	025480
	CALL MMATIO(QM,ICGM,ICGM,IC,KIN,KOUT,NDIM,NDIM)	025490
	IO=IHOLD	025500
	RETURN	025510
	END	025520
	SUBROUTINE SORDER(FM,BM,GM,QP,HM,IRFM,IFPM,ICBM,ICGM,IRFS,	025530
	IHU,FU,GU)	025540
	DIMENSION FM(NDIM,NDIM),BM(NDIM,NDIM),GM(NDIM,NDIM),	025550
	IQM(NDIM,NDIM),HM(NDIM,NDIM),PU(NDIM,NDIM),COM1(1),	025560
	ICOM2(1),FU(NDIM,NDIM),GU(NDIM,NDIM)	025570
	COMMON/MAIN1/NDIM,NDIM1,COM1	025580
	COMMON/MAIN2/COM2	025590
	COMMON/INGU/KIN,KOUT,KPUNCH	025600
	COMMON/MAUN5/MSG	025610
	REAL ENTRY,ENTRY1,BAC,BTC	025620
	IFS=IRFM+1	025630
	IGS=ICGM+1	025640
	ICGS=ICGM+ICBM	025650
	ICBM2=2*ICBM	025660
	CALL ZPART(FM(IFS,1),ICBM2,IRFS)	025670
	CALL ZPART(BM(IFS,1),ICBM2,ICBM)	025680
	CALL ZPART(QM(IGS,1),ICBM,ICGS)	025690
	CALL ZPART(QM(1,IGS),ICGM,ICBP)	025700
	CALL ZPART(GM(IFS,1),ICBM2,ICGS)	025710
	CALL ZPART(GM(1,IGS),IRFM,ICBP)	025720
	CALL ZPART(HM(1,IFS),IRHM,ICBM2)	025730
	CALL ZPART(FU,ICBM2,ICBM2)	025740
	CALL ZPART(IHU,ICBP,ICBM2)	025750
	CALL ZPART(GU,ICBM2,ICBM)	025760
C	DESIGN ONE SHAPING FILTER PER CONTROL CHANNEL	025770
	DO 80 IF=1,ICBM	025780
	WRITE(KOUT,*)'FILTER DESIGN',IF	025790
	WRITE(KOUT,*)'INPUT OPTIONS: (1) ENTER C, (2) ENTER D AND E,	025800

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	1(3) ENTER Q. (4) GO TO NEXT FILTER OR EXIT IF DONE	025810
	DO 90 IT=1,FC	025820
	WRITE(KOUT,*) 'OPTION>'	025830
26	FORMAT(I1)	025840
	READ(KIN,2)IOPT	025850
	GO TO (1,2,3,4)IOPT	025860
1	WRITE(KOUT,*) 'ENTER A>'	025870
	READ(KIN,*)ENTRY	025880
	HU(IF,(2*IF-1))=ENTRY	025890
	HU(IF,(2*IF))=1.	025900
	GO TO 90	025910
2	WRITE(KOUT,*) 'ENTER B AND C>'	025920
	READ(KIN,*)ENTRY,ENTRY1	025930
	BAC=-1*(ENTRY+ENTRY1)	025940
	BTC=-1*(ENTRY+ENTRY1)	025950
	FU((2*IF-1),(2*IF))=1.	025960
	FU((2*IF),(2*IF-1))=BTC	025970
	FU((2*IF),(2*IF))=BAC	025980
	GO TO 90	025990
3	WRITE(KOUT,*) 'ENTER Q>'	026000
	READ(KIN,*)ENTRY	026010
	QH((ICGM+IF),(ICGP+IF))=ENTRY	026020
90	CONTINUE	026030
4	CONTINUE	026040
	GU((IF*2),IF)=1.	026050
90	CONTINUE	026060
C	CALCULATE B=HU AND PUT IN UPPER RIGHT PARTITION OF FM	026070
	CALL MMUL(BM,HU,IRFM,ICBM,ICBM2,FM(1,IFS))	026080
C	INSERT FU INTO LOWER RIGHT PARTITION OF FM	026090
	CALL EQUATE(FM(IFS,IFS),FU,ICBM2,ICBM2)	026100
C	INSERT GU INTO LOWER RIGHT PARTITION OF GM	026110
	CALL EQUATE(GM(IFS,IGS),GU,ICBM2,ICBM)	026120
C	CHANGE COLUMN DIMENSION OF GM AND ORDER OF SYSTEM	026130
	ICGM=ICGS	026140
	IRFM=IRFS	026150
	RETURN	026160
	END	026170

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## APPENDIX C: Modifications and Additions to CGTPI

### C.1 Introduction

This appendix is intended to be used as a supplement to References 13 and 30. These sequences are user's guides to a computer-aided design program for designing controllers employing a Command Generator Tracker in the feedforward loop and a Proportional-plus-Integral regulator in the feedback loop (CGTPI). Reference 13 describes the original version of the program. Reference 30 contains a modified version that admits implicit model following for design of controller gains and that can be run in conjunction with a performance evaluation program (PFEVAL), described in the same reference. Modifications outlined in this chapter are made to the version in Reference 30.

As stated in Chapter III, only the PI regulator design portion of the program is used for this thesis. However, to execute PFEVAL, the user must first execute all three design paths of CGTPI. These include the PI regulator design, CGT design, and Kalman filter design. By proper choice of the CGT model matrices, the controlled degenerates into a simple PI regulator. These matrices are defined below:

$$A_m = [I] \quad (C-1)$$

$$B_m = [0] \quad (C-2)$$

$$C_m = [0] \quad (C-3)$$

$$D_m = [I] \quad (C-4)$$

Modifications to the program include an additional subroutine to apply the Doyle and Stein technique. This is bracketed and marked with



an "A" in the source code at the end of this appendix. Modifications to allow inputting colored-noise into the design model are bracketed and marked with an "M". Since the modifications made to CGTPI are not extensive, only the specific subroutines that contain changes are listed.

### C.2 Additions to CGTPI

One additional subroutine is added to CGTPI to allow modification of the dynamic driving noise covariance matrix,  $Q_d$ , via the Doyle and Stein technique as described in Section 3.4. Subroutine DAS is listed in lines 29730 through 29940. When executing the Kalman filter design option of CGTPI, this subroutine is called in line 18730. An input prompt will ask if the Doyle and Stein modification is desired. If so, then a prompt will ask for the scalar design parameter,  $q$ . The  $Q_d$  matrix is then modified as expressed in Equation (3-43) with the assumption that  $V = I$ . The modified  $Q_d$  matrix is also written to the LIST file.

### C.3 Modifications to CGTPI

The original version of CGTPI was not written with the intention of augmenting additional states with the design model. The modifications listed in the source code in this appendix will allow the technique of injecting time-correlated noise into the design model, but the method is awkward as will become apparent.

To apply the technique requires two executions of CGTPI. In the first execution, the unaugmented design model and truth model are entered to the program. Then, the PI regulator design, CGT design, and Kalman filter design paths are executed normally. All prompts to enter either filter gains or CGT gains should be answered with an "N". In the CGT

design path, the model matrices listed in Equations (D-1) through (D-4) can be entered if only a PI regulator is desired. It is useful to write the truth model to the SAVE file. An input prompt asks if it is desired to write the model to the SAVE file. If a "Y" is entered, the model matrices are stored by the program. This file is structured so that the design model, truth model, regulator gains, CGT gains, and command model can all be written to the file for later use. Copying the SAVE file to a DATA file allows the model and gain matrices to be read directly into the program, rather than be inputted from a terminal (Ref 13). The regulator and CGT gains are also written to the SAVE file. After following all three design paths, execution of the program is halted.

Prior to the second execution, the SAVE file is copied to a DATA file. Also, it is necessary to have the augmented matrices specified in Chapter IV available to be read into the program from the terminal. In the second execution, the augmented system is entered into the design model. The truth model can be read from the DATA file containing information from the previous execution. After entering the model, the CGT path is followed again. A prompt will ask if it is desired to read in the CGT gains. A "Y" is entered, the gains are read from the DATA file, and this path is exited. This modification is listed in lines 6600 to 6630 and 6910.

Next, the Kalman filter design path is followed with the augmented design model. After execution of this option, a prompt will inquire if it is desired to store performance analysis data to the SAVE file. A "Y" is entered, and another prompt will inquire if it is desired to

read controller gains from the DATA file. Again, a "Y" is entered. The modifications in lines 27230 through 27430 allow the program to read in PI controller gains from the previous execution and augment with columns by zeros so that the shaping filter states are not fed back through the controller. Performance evaluation data is stored and PFEVAL can thus be run normally as described in Reference 30.

The modified version also allows the option of reading in Kalman filter gains from a previous execution stored on the DATA file. These modifications are listed in lines 4230 to 4320.

Final modifications are given in lines 4470 through 4610 to reformat the PI controller gains and allow storage room for the augmented zeros.

```

2960= SUBROUTINE CGTXQ
2970= COMMON/MAIN1/NDIM,NDIM1,COM1(1)
2980= COMMON/MAIN2/COM2(1)
2990= COMMON/INOU/KIN,KOIT,KPUNCH
3000= COMMON/DESIGN/NUCOM,TSAMP,LFLRPI,LFLCGT,LFLKF,LTEVAL,LABORT
3010= COMMON/FILES/KSAVE,KDATA,KPLOT,KLIST,KTERM
3020= COMMON/SYSMTX/NUVM,SM(1)
3030= COMMON/ZMTX1/NUVM,ZM1(1)
3040= COMMON/ZMTX2/ZM2(1)
3050= COMMON/NDIMD/NND,NRD,NPD,NMD,NDD,NWD,NWDD,NPLD,NWPHWD,NNPR
3060= COMMON/LOCD/LAP,LGP,LPHI,LBD,LEX,LPHD,LQ,LQN,LQD,LC,LBY,LEY
3065= 1,LHP,LR
3070= COMMON/DSNMTX/NUVM,NODY,NOEY,DM(1)
3080= COMMON/NDIMC/NNC,NRC,NPC
3090= COMMON/LOCC/LPHC,LBDC,LCC,LBC
3100= COMMON/CMOMTX/NUVM,NEWCH,NOOC,CM(1)
3110= COMMON/NDIMT/NNT,NRT,NMT,NWT
3120= COMMON/LOCT/LPHT,LBOT,LQDT,LHT,LRT,LTDT,LTNT
3130= COMMON/TRUMTX/NUVM,TM(1)
3140= COMMON/LCNTRL/LPI1,LPT12,LPI22,LPHDL,LBDL
3150= COMMON/CONTROL/NUCTL,CTL(1)
3160= COMMON/LREGPI/LXDW,LUDW,LPHCL,LKX,LKZ
3170= COMMON/CREGPI/NVRPI,RPI(1)
3180= COMMON/LCGT/LA11,LA13,LA21,LA23,LA12,LA22,LKXA11,LKXA12,
3185= 1LKXA13
3190= COMMON/CCGT/NUCGT,CGT(1)
3200= COMMON/LKF/LEASD,LFLTRK,LFCOV
3210= COMMON/CKF/NUFLT,FLT(1)
3220= COMMON/AMC/AM(1)
3230= COMMON/BOG/BO(1)
3240= DIMENSION LD(15),ND(15)
3250= DATA NPLTZN/606/
3260= DATA IEQI,NO/-1,1HN/
3270= REWIND KLIST
3280= WRITE(KLIST,115) DATE(DUM),TIME(DUM)
3290= WRITE(KTERM,115) DATE(DUM),TIME(DUM)
3300= 115 FORMAT('1',27X,'* * * CGTPIF * * */14X,
3310= 1 'PROGRAM TO DESIGN A COMMAND GENERATOR TRACKER'*/8X,
3320= 2 'USING A REGULATOR WITH PROPORTIONAL PLUS INTEGRAL CONTROL
3325= 1'*/16X,
3330= 3 'AND A KALMAN FILTER FOR STATE ESTIMATION.'*/28X,
3340= 4 '* * * CGTPIF * * */11X,'DATE : ',A10//,11X,
3350= 5 'TIME : ',A10//)
3360= REWIND KSAVE
3370= REWIND KDATA
3380= WRITE(KSAVE,112) IEQI,NPLTZN
3390= DO 10 I=1,15
3400= 10 ND(I)=0
3410= DO 12 I=1,15
3420= 12 LD(I)=1
3430= LFLRPI=0
3440= LFLCGT=0
3450= LFLKF=0
3460= LTEVAL=0
3470= LABORT=0
3480= IPT=0

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3490=      ICGT=0
3500=      ITRU=0
3510=      IFLTR=0
3520=      ICODE=4
3530=      LFAVAL=0
3540=      LGCGT=0
3550=      NVCOM=MINO(NDIM,NVZM)
3560=      KOUT=KLIST
3570=      KPUNCH=KPILOT
3580=      IF(NVSM.GE.NPLTZM) GO TO 50
3590=      WRITE 101,NPLTZM
3600=      GO TO 1000
3610= 50   WRITE 102
3620=      READ*,TSAMP
3630=      IF(TSAMP.LE.0.) GO TO 50
3640=      WRITE(KLIST,103) TSAMP
3650= 103  FORMAT('OSAMPLE PERIOD IS ',F3.3,' SECONDS')
3660=      CALL SETUP(ND,LD,ICGT,ITRU,1)
3670=      IF(LABORT) 1000,100,1000
3680= 100  LABORT=0
3690=      WRITE 104
3700= 104  FORMAT('OCONTROLLER DESIGN (Y OR N) >')
3710=      READ 111,IANS
3720=      IF(IANS.EQ.NO) GO TO 500
3730=      LFLKF=0
3740=      CALL PINTX(IPI)
3750=      IF(LABORT) 1000,125,1000
3760= 125  WRITE 105
3770= 105  FORMAT('ODESIGN RFG/PI (Y OR N) >')
3780=      READ 111,IANS
3790=      IF(IANS.EQ.NO) GO TO 150
3800=      CALL SREGPI
3810=      IF(LABORT) 1000,200,1000
3820= 150  WRITE 106
3830= 106  FORMAT('ODESIGN CGT (Y OR N) >')
3840=      READ 111,IANS
3850=      IF(IANS.EQ.NO) GO TO 100
3860=      CALL SETUP(ND,LD,ICGT,ITRU,2)
3870=      WRITE 117
3880= 117  FORMAT('OTERMINATE CGT DESIGN PATH (Y OR N)? >')
3890=      READ 111,IANS
3900=      IF(IANS.NE.NO) GO TO 500
3910=      IF(ICGT) 155,100,155
3920= 155  IF(LABORT) 100,160,1000
3930= 160  CALL SCGT
3940=      IF(LABORT) 100,170,1000
3950= 170  IF(LFLCGT.LE.0) GO TO 125
3960= 200  LABORT=0
3970=      WRITE 107
3980= 107  FORMAT('OCONTROLLER EVALUATION WRT TRUTH MODEL (Y OR N) >')
3990=      READ 111,IANS
4000=      IF(IANS.EQ.NO) GO TO 250
4010=      CALL SETUP(ND,LD,ICGT,ITRU,3)
4020=      IF(LABORT) 200,260,1000
4030= 250  LTEVAL=0
4040= 260  CALL CEVAL
4050=      IF(LFLCGT.EQ.1) LGCGT=1
4060=      IF(LFAVAL.EQ.0.OR.LGCGT.EQ.0) GO TO 100
4070= 270  WRITE 600

```

```

4030= 400  FORMAT('WRITE PERFORMANCE EVALUATION DATA TO 'SAVE' FILE
4085=      1(Y OR N)
4090=      +>')
4100=      READ 111,IANS
4110=      IF(IANS.EQ.NO) GO TO 100
4120=      ICODE=ICODE+1
4130=      CALL PFDATA(ICODE,ND)
4140=      INUM=ICODE-4
4150=      WRITE 603,INUM
4160= 605  FORMAT('PERFORMANCE EVALUATION DATA, NO. 'I2,','WRITTEN TO
'SAVE
4170=      +' FILE')
4180=      GO TO 100
4190= 500  LABORT=0
4200=      WRITE 108
4210= 108  FORMAT('FILTER DESIGN (Y OR N) >')
4220=      READ 111,IANS
4230=      IF(IANS.EQ.NO) GO TO 900
4240=      WRITE 777
4250= 777  FORMAT('READ IN CGT GAINS (Y OR N) >')
4260=      READ 111,IANS
4270=      IF(IANS.EQ.NO) GO TO 505
M - 4280=      IF (LGCGT.NE.0) GO TO 505
4290=      IF (IFLTR.NE.0) GO TO 505
4300=      CALL READFS(SH,ND,4,IERR)
4310=      LGCGT=ND(2)
4320= 505  CALL FLTRK(IFLTR)
4330=      IF(IFLTR.EQ.0) GO TO 900
4340=      IF(LABORT) 1000,510,1000
4350= 510  CALL SETUP(ND,LD,ICGT,ITRU,3)
4360=      IF(LABORT) 500,525,1000
4370= 525  CALL FEVAL
4380= 530  IF(LABORT) 1000,540,1000
4390= 540  LFAVAL=1
4400=      IF(LGCGT.EQ.1) GO TO 270
4410=      GO TO 300
4420= 900  WRITE 109
4430= 109  FORMAT('END DESIGN RUNS (Y OR N) >')
4440=      READ 111,IANS
4450=      IF(IANS.EQ.NO) GO TO 100
4460=      IF(LFLRPI.EQ.0) GO TO 1000
M - 4470=      NPNTS=NRD*NNPR
4480=      ND(1)=NPNTS
4490=      ND(2)=LGCGT
4500=      ND(3)=LKX
4510=      ND(4)=LKZ
4520=      CALL FTMTX(RPI(LKX),SH,NPNTS,1)
4530=      ND(5)=NPNTS+1
4540=      IF(LGCGT.EQ.0) GO TO 910
4550=      ND(6)=LKXA11
4560=      ND(7)=LKXA12
4570=      ND(8)=LKXA13
4580=      NPNTS=NRD*(NNC+NRC+ND)
4590=      ND(1)=ND(1)+NPNTS
4600=      CALL FTMTX(CGT(LKXA11),SH(ND(5)),NPNTS,1)
M - 4610= 910  CALL WFTLED(4,ND(1),ND,SH)
4620=      WRITE 113
4630= 1000 CONTINUE
4640=      WRITE(KLIST,110)

```

```

4650=      REWIND KSAVE
4660=      REWIND KDATA
4670=      REWIND KLIST
4680=      WRITE 110
4690= 101  FORMAT(*0INSUFFICIENT MEMORY /SYSMTX/, NFED: ',I4)
4700= 102  FORMAT(*0ENTER SAMPLE PERIOD FOR DIGITAL CONTROLLER >*)
4710= 110  FORMAT(*0PROGRAM EXECUTION STOP*)
4720= 111  FORMAT(A3)
4730= 112  FORMAT(2I4)
4740= 113  FORMAT(6X,'REG/PI GAINS WRITTEN TO 'SAVE' FILE')
4750=      RETURN
4760=C END SUBROUTINE CGTXQ

```

```

6390= SUBROUTINE SCMD(ND,LD,ICGT)
6400= COMMON/DESIGN/NUCOM,TSAMP,LFLRPI,LFLCGT,LFLKF,LTEVAL,LABORT
6410= COMMON/FILES/KSAVE,KDATA,KPLOT,KLIST,KTERM
6420= COMMON/SYSMTX/NUSM,SM(1)
6430= COMMON/ZMTX1/NUZH,ZH1(1)
6440= COMMON/ZMTX2/ZM2(1)
6450= COMMON/NDIND/NND,NRD,NPD,NMD,NDD,NWD,NWDD,NPLD,NWPNWD,NNPR
6460= COMMON/NDIMC/NNC,NRC,NPC
6470= COMMON/CMDMTX/NVCM,NEWCM,NODC,CM(1)
6480= COMMON/LREGPI/LXDW,LUDW,LPHCI,LKX,LKZ
6490= COMMON/CREGPI/NVRPI,RPI(1)
6500= DIMENSION ND(1),LD(1)
6510= DATA NO/1HN/
6520= WRITE(KLIST,110)
6530= 110 FORMAT(////11X,5('* '), 'CGT DESIGN',5('* ')////)
6540= NEWCM=0
6550= IF(LFLRPI) 10,5,10
6560= S WRITE 102
6570= READ 111,IANS
6580= IF(IANS.EQ.NO) GO TO 8
6590= CALL READFS(SM,ND,4,IERR)
6600= NSIZE=ND(5)-1
6610= LKX=ND(3)
6620= LKZ=ND(4)
6630= CALL FTMTX(SM,RPI(LKX),NSIZE,1)
6640= IF(IERR.NE.0) RETURN
6650= CALL MATLST(RPI(LKX),NRD,NMD,'KX',KLIST)
6660= CALL MATLST(RPI(LKZ),NRD,NRD,'KZ',KLIST)
6670= LFLRPI=-1
6680= GO TO 10
6690= 8 IF(LFLCGT.GE.0) GO TO 9
6700= WRITE 103
6710= 103 FORMAT('OSYSTEM UNSTABLE - - OPEN-LOOP CGT NOT FEASIBLE')
6720= RETURN
6730= 9 LKX=1
6740= LKZ=1
6750= NSIZE=NRD*NND
6760= CALL ZPART(RPI(LKX),1,NSIZE,1)
6770= 10 IF(ICGT.EQ.0) GO TO 12
6780= WRITE 108
6790= 108 FORMAT(' MODIFY COMMAND MODEL (Y OR N) >')
6800= READ 111,IANS
6810= IF(IANS.EQ.NO) RETURN
6820= 12 CALL RSYS(SM,LD,ND,2,ICGT)
6830= IF(LABORT.NE.0) RETURN
6840= NEWCM=1
6850= CALL POLES(SM,NNC,2,ZM1,ZM2)
6860= IF(NPC.EQ.NPD) GO TO 15
6870= WRITE 104
6880= LABORT=-1
6890= RETURN
6900= 15 CALL OSCRTC(1.0,ZM1)

```



M - 6910= 102 FORMAT(' READ REG/PI GAINS FROM 'DATA' FILE (Y OR N) >')  
6920= 104 FORMAT('OCCOMMAND AND DESIGN MODEL OUTPUTS NOT EQUAL IN  
6930= 1NUMBER')  
6940= RETURN  
6950=C END SUBROUTINE SCMD

```

18250= SUBROUTINE FLTRK(IFLTR)
18260= COMMON/MAIN1/NDIM,NDIM1,COM1(1)
18270= COMMON/MAIN2/COM2(1)
18280= COMMON/DESIGN/NUCOM,TSAMP,LFLRPI,LFLCGT,LFLKF,LTEVAL,LABORT
18290= COMMON/FILES/KSAVE,KDATA,KPLOT,KLIST,KTERM
18300= COMMON/SYSMTX/HUSH,SH(1)
18310= COMMON/ZMTX1/NUZM,ZM1(1)
18320= COMMON/ZMTX2/ZM2(1)
18330= COMMON/NDIMD/NNH,NRQ,NPD,NMD,NDD,NWD,NWDD,NPLD,NWPNWD,NNPR
18340= COMMON/LOCD/LAP,LGP,LPHI,LBD,LEX,LPHD,LQ,LQN,LQD,LC,LDY,LEY
18345= 1,LHP,LR
18350= COMMON/DSNMTX/NUDM,NOHY,NOEY,DM(1)
18360= COMMON/LKF/LEADSN,LFLTRK,LFCOV
18370= COMMON/CKF/NUFLT,FLT(1)
18380= IF(NWPNWD.GT.0) GO TO 1
18390= WRITE(KTERM,108)
18400= 108 FORMAT('ONO DRIVING NOISES - - FILTER DESIGN ABORT')
18410= RETURN
18420= 1 IF(NMD.GT.0) GO TO 2
18430= WRITE(KTERM,109)
18440= 109 FORMAT('ONO MEASUREMENTS - - FILTER DESIGN ABORT')
18450= RETURN
18460= 2 WRITE(KLIST,110)
18470= 110 FORMAT('////11X,5(' * '), "FILTER DESIGN",5(' *')////)
18480= NSIZE=NPLD*(1+NPLD+NMD)
18490= IF(NSIZE.LE.NUFLT) GO TO 3
18500= WRITE 101,NSIZE
18510= 101 FORMAT('OINSUFFICIENT MEMORY /CKF/, NEED: ',I4)
18520= LABORT=NSIZE
18530= RETURN
18540= 3 NDIM=NPLD
18550= NDIM1=NDIM+1
18560= 5 IF(NWD.EQ.0) GO TO 12
18570= IF(IFLTR.LE.0) GO TO 6
18580= WRITE 105,NWD
18590= 105 FORMAT(' ENTER STATE NOISE STRENGTHS: ',I2)
18600= CALL RQWGS(DM(LQ),NWD,0)
18610= 6 CALL DVCTOR(NWD,DM(LQ),ZM1)
18620= CALL MATLST(ZM1,NWD,1,'Q',KTERM)
18630= 10 CALL MATLST(DM(LQ),NWD,NWD,'Q',KLIST)
18640= 12 IF(NWDD.EQ.0) GO TO 18
18650= IF(IFLTR.LE.0) GO TO 13
18660= WRITE 106,NWDD
18670= 106 FORMAT(' ENTER DISTURBANCE NOISE STRENGTHS: ',I2)
18680= CALL RQWGS(DM(LQN),NWDD,0)
18690= 13 CALL DVCTOR(NWDD,DM(LQN),ZM1)
18700= CALL MATLST(ZM1,NWDD,1,'QN',KTERM)
18710= 15 CALL MATLST(DM(LQN),NWDD,NWDD,'QN',KLIST)
18720= 18 CALL QDSCRT(DM(LQ),DM(LQN),ZM1,ZM2)
A - 18730= CALL DAS(DM(L3D),DM(LQD),ZM1,ZM2)
18740= IF(IFLTR.LE.0) GO TO 19
18750= WRITE 107,NMD
18760= 107 FORMAT(' ENTER MEASUREMENT NOISE STRENGTHS: ',I2)

```

```

18770=      CALL ROWGTS(DM(LR),NMD,0)
18780= 19    CALL DUCTOR(NMD,DM(LR),ZM1)
18790=      CALL MATLST(ZM1,NMD,1,"R",KTERM)
18800= 20    CALL MATLST(DM(LR),NMD,NMD,"R",KLIST)
18810= 25    CALL TERMTX(DM(LHP),SM,NMD,NDIM,2)
18820=      CALL TRANS2(NMD,NDIM,SM,ZM1)
18830=      LFCOV=LFLTRK+NDIM*NMD
18840=      CALL DUCTOR(NMD,DM(LR),FLT(LFCOV))
18850=      CALL KFLTR(NDIM,NMD,FLT,ZM1,DM(LD),FLT(LFCOV),ZM2,
18860= 1      FLT(LFLTRK),SM)
18870=      CALL TERMTX(SM,COM2,NDIM,NDIM,2)
18880=      IA=1
18890=      DO 30 I=1,NPLD
18900=      FLT(LFCOV-1+I)=SQRT(ZM2(IA))
18910= 30    IA=IA+NDIM
18920=      CALL MATLST(FLT(LFLTRK),NDIM,NMD,"KF",KLIST)
18930=      CALL MATLST(FLT(LFLTRK),NDIM,NMD,"KF",KTERM)
18940=      IFLTR=1
18950=      LFLKF=1
18960= 111   FORMAT(A3)
18970=      RETURN
18980=C END SUBROUTINE FLTRK

```

```

26970= SUBROUTINE PFDATA(ICODE,ND)
26980= COMMON/MAIN1/NDIM,NDIM1,COM1(1)
26990= COMMON/MAIN2/COM2(1)
27000= COMMON/INOU/KIN,KOUT,KPUNCH
27010= COMMON/DESIGN/NUCOM,TSAMP,LFLRPI,LFLCGT,LFLKFL,LFTEVAL,LABORT
27020= COMMON/FILES/KSAVE,KDATA,KPLOT,KLIST,KTERM
27030= COMMON/SYSTX/NUSM,SN(1)
27040= COMMON/ZMTX1/NUZM,ZM1(1)
27050= COMMON/ZMTX2/ZM2(1)
27060= COMMON/NDIMD/ND,NRD,NPD,NMD,NDD,NWD,NWDD,NPLD,NWPNWD,NNPR
27070= COMMON/LOCB/LAP,LGP,LPHI,LRO,LEX,LPHD,IQ,LQN,LQD,LC,LBY,LEY
27075= 1,LHP,LR
27080= COMMON/DSNMTX/NUDM,NODY,NOEY,OM(1)
27090= COMMON/NDIMC/NNC,NRC,NPC
27100= COMMON/LGCC/LPHC,LBDC,LCC,LDC
27110= COMMON/CDMTX/NUCM,NEWCM,NODC,CM(1)
27120= COMMON/NDINT/NNT,NRT,NMT,NWT
27130= COMMON/LOCT/LPHT,LBET,LQDT,LHT,LRT,LTDT,LTNT
27140= COMMON/TRUMTX/NUTM,TH(1)
27150= COMMON/LCCTRL/LPI11,LPI12,LPI21,LPI22,LPHDL,LBDL
27160= COMMON/CONTROL/NUCTL,CTL(1)
27170= COMMON/LREGPI/LXDW,LUDW,LPHCL,LKX,LKZ
27180= COMMON/CREGPI/NVRPI,RPI(1)
27190= COMMON/LCGT/LA11,LA13,LA21,LA23,LA12,LA22,LKXA11,LKXA12,
27195= 1LKXA13
27200= COMMON/CCGT/NUCGT,CGT(1)
27210= COMMON/LKF/LEADSN,LFLTRK,LFCOV
27220= COMMON/CKF/NUFLT,FLT(1)
27230= DIMENSION ND(1)
27240= DATA NO/1HN/
27250= WRITE 101
27260=101 FORMAT('OREAD IN CONTROLLER GAINS? (Y OR N) >')
27270= READ 103,IANS
27280=103 FORMAT(A3)
27290= IF (IANS.EQ.NO) GO TO 102
27300= CALL READFS(SM,ND,4,IERR)
27310= LKX=ND(3)
27320= LKZ=ND(4)
27330= LKXA11=ND(6)
27340= LKXA12=ND(7)
27350= LKXA13=ND(8)
27360= NNDP=(LKZ-LKX+1)/NRD
27370= CALL FTHTX(SM,RPI(LKX),NRD,NNDP)
27380= LL=LKX+NRD*NNDP
27390= LE=NND-NNDP
27400= CALL ZPART(RPI(LL),NRD,LE,NRD)
27410= LKZ=LL+NRD*LE
27420= LL=LL-LKX+1
27430= CALL FTHTX(SM(LL),RPI(LKZ),NRD,NPD)
27440= IL=ND(1)-ND(5)+1
27450= CALL FTHTX(SM(ND(5)),CGT(LKXA11),IL,1)
27460=102 ND(1)=NND
27470= ND(2)=NRD

```

```

27480=      ND(3)=NPD
27490=      ND(4)=NMD
27500=      ND(5)=NDD
27510=      ND(6)=NPLD
27520=      ND(7)=NNC
27530=      ND(8)=NRC
27540=      ND(9)=NPC
27550=      ND(10)=NNT
27560=      ND(11)=NRT
27570=      ND(12)=NMT
27580=      ND(13)=NODY
27590=      ND(14)=NOEY
27600=      NUZMS=NUZM
27610=C
27620=      CALL FTMTX(DM(LPHI),SM,NND,NND)
27630=      LL=NND*NND+1
27640=      CALL FTMTX(DM(LBD),SM(LL),NND,NRD)
27650=      LL=NND*NRD+LL
27660=C
27670=      IF(NDD.EQ.0) GO TO 100
27680=      CALL FTMTX(DM(LEX),SM(LL),NND,NDD)
27690=      LL=NND*NND+LL
27700=      CALL FTMTX(DM(LPHD),SM(LL),NDD,NDD)
27710=      LL=NDD*NRD+LL
27720=      IF(NODY.EQ.1) GO TO 90
27730=      CALL FTMTX(DM(LDY),SM(LL),NPD,NRD)
27740=      LL=NPD*NRD+LL
27750= 90    IF(NOEY.EQ.1) GO TO 95
27760=      CALL FTMTX(DM(LEY),SM(LL),NPD,NDD)
27770=      LL=NPD*NDD+LL
27780= 95    CALL FTMTX(DM(LHP),SM(LL),NMD,NPLD)
27790=      LL=NMD*NPLD+LL
27800=      GO TO 200
27810= 100  CONTINUE
27820=C
27830=      IF(NODY.EQ.1) GO TO 105
27840=      CALL FTMTX(DM(LDY),SM(LL),NPD,NRD)
27850=      LL=NPD*NRD+LL
27860= 105  CALL FTMTX(DM(LHP),SM(LL),NMD,NND)
27870=      LL=NMD*NND+LL
27880= 200  CALL FTMTX(DM(LC),SM(LL),NPD,NND)
27890=      LL=NPD*NND+LL
27900=      CALL FTMTX(CM(LPHC),SM(LL),NNC,NNC)
27910=      LL=NNC*NNC+LL
27920=      CALL FTMTX(CM(LBDC),SM(LL),NNC,NRC)
27930=      LL=NNC*NRC+LL
27940=      CALL FTMTX(CM(LCC),SM(LL),NPC,NNC)
27950=      LL=NPC*NNC+LL
27960=      CALL FTMTX(TM(LPHT),SM(LL),NNT,NNT)
27970=      LL=NNT*NNT+LL
27980=      CALL FTMTX(TM(LBDT),SM(LL),NNT,NRT)
27990=      LL=NNT*NNT+LL
28000=      CALL FTMTX(TM(LQDT),SM(LL),NNT,NNT)
28010=      LL=NNT*NNT+LL
28020=      CALL FTMTX(TM(LHT),SM(LL),NMT,NNT)
28030=      LL=NMT*NNT+LL

```

```

28040= CALL FTMTX(TM(LRT),SM(LL),NMT,NMT)
28050= LL=NMT*NMT+LL
28060= CALL FTMTX(RPI(LKX),SM(LL),NRD,NND)
28070= LL=NRD*NND+LL
28080= CALL FTMTX(RPI(LKZ),SM(LL),NRD,NPD)
28090= LL=NRD*NPD+LL
28100= CALL FTMTX(CGT(LKXA11),SM(LL),NRC,NNC)
28110= LL=NRC*NNC+LL
28120=C
28130= IF(NND.EQ.0) GO TO 300
28140= CALL FTMTX(CGT(LKXA13),SM(LL),NRD,NDD)
28150= LL=NRD*NDD+LL
28160= CALL FTMTX(FLT(LFLTRK),SM(LL),NPLD,NMD)
28170= LL=NPLD*NMD+LL
28180= CALL FTMTX(TM(LTDT),SM(LL),NND,NNT)
28190= LL=NND*NNT+LL
28200= CALL FTMTX(TM(LTNT),SM(LL),NDD,NNT)
28210= LL=NDU*NNF+LL
28220= GO TO 310
28230= 300 CONTINUE
28240=C
28250= CALL FTMTX(FLT(LFLTRK),SM(LL),NND,NMD)
28260= LL=NND*NMD+LL
28270= CALL FTMTX(TM(LTDT),SM(LL),NND,NNT)
28280= LL=NND*NNT+LL
28290= 310 SM(LL)=TSAMP
28300= ND(15)=LL
28310= CALL WFILED(ICODE,LL,ND,SM)
28320= NUZH=NUZMS
28330= RETURN
28340=C END SUBROUTINE PFDATA

```

```

SUBROUTINE DAS(BD,QD,BDP,QP)                                029660
COMMON/MAIN1/NDIM                                           029670
COMMON/NDIMD/NND,NRD,NPD,NMD,NDD,NUD,NMDD,NPLD,NMFD,NMNR    029700
COMMON/FILES/KSAVE,KDATA,KPLOT,KLIST,KTERM                 029710
DATA NO/1HN/                                                029720
WRITE 101                                                    029730
101 FORMAT('MODIFY Q BY DOYLE AND STEIN? (Y OR N) > ')    029740
READ 102,IANS                                               029750
102 FORMAT(A3)-----A-----029760
IF(IANS.EQ.NO) RETURN                                     029770
WRITE 103                                                    029780
103 FORMAT('ENTER Q=?')                                     029790
READ *,QDAS                                                 029800
QDAS=QDAS*QDAS                                              029810
CALL TFRNTX(BD,BDP,NND,NRD,2)                              029820
IF(NND.EQ.NPLD) GO TO 10                                    029830
L1=LA00R(NPLD,NND+1,1)                                     029840
10 CALL MAT2(NPLD,NRD,BDP,BDP,QP)                          029850
CALL MA001(NPLD,NPLD,QD,QP,QD,QDAS)                       029860
CALL MATLST(QD,NPLD,NPLD,QP,QP,KLIST)                     029870
RETURN                                                       029880
END                                                         029890

```

APPENDIX D: Additional AFTI/F-16  
Performance Data

D.1 Introduction

In Reference 6, it is stated that the Doyle and Stein technique for Kalman filter robustification is guaranteed to work only for a minimum-phase design model. That is, the model of the system to be controlled may not have transmission zeroes in the right-half s-plane.

The perturbation equations of motion described in Chapter V are linearized about a trim flight condition at an altitude of 10000 feet and a Mach number of 0.6. This yields a design model that is minimum phase. Initially, the trim condition was at the same altitude but with a Mach number of 0.8, to be consistent with other research done with the same model (Ref 12;27;29). At this design point, the eight-state controller model is non-minimum phase. The Doyle and Stein robustification was applied to this system with some unexpected results.

The results of a performance analysis of the eight-state continuous-time controller evaluated against a twelve-state truth model are presented in this chapter. First the performance was evaluated against a twelve-state truth model (10000 ft, M=0.6). Then, the response of the system at off-design condition is presented, with and without robustification.

D.2 Performance Evaluation of a Non-Minimum Phase Design Model

The performance of the non-minimum phase system at the design flight condition with a reduced-order design model is shown in Figure (D-1). The response of the system is stable with  $\theta$  converging to zero, and the standard deviation converging to a small finite value. The performance

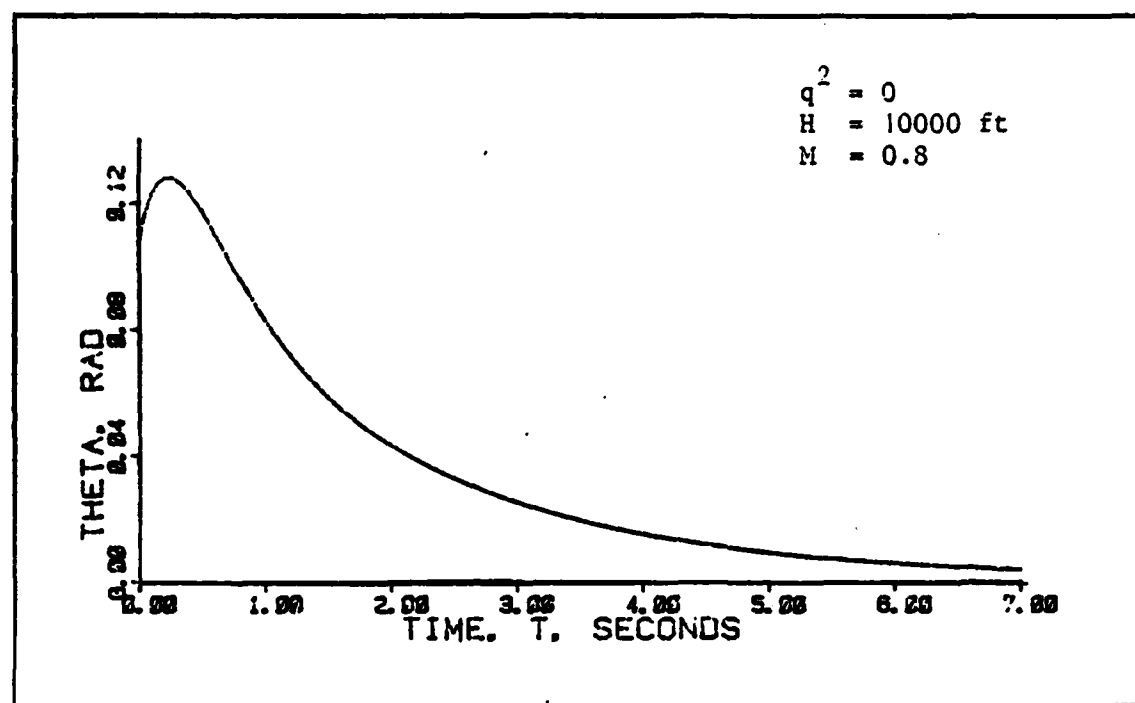


Figure D-1a: Unrobustified Non-Minimum Phase Mean Response of  $\Theta$

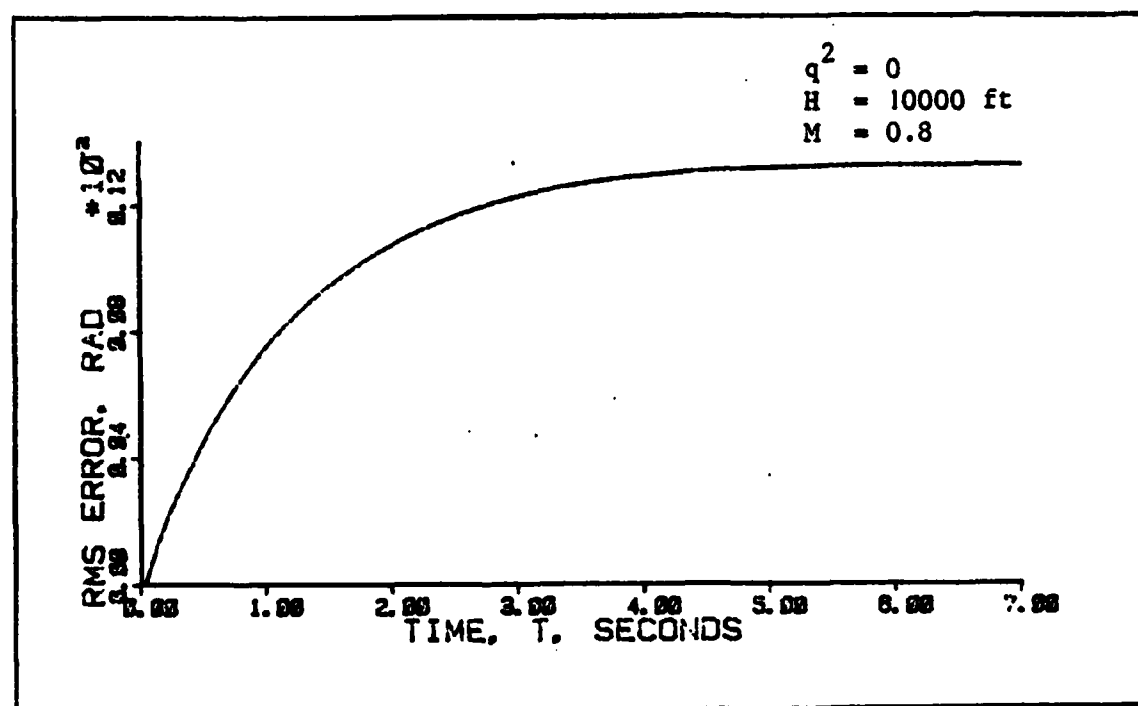


Figure D-1b: Standard Deviation of Theta for the Unrobustified Non-Minimum Phase System



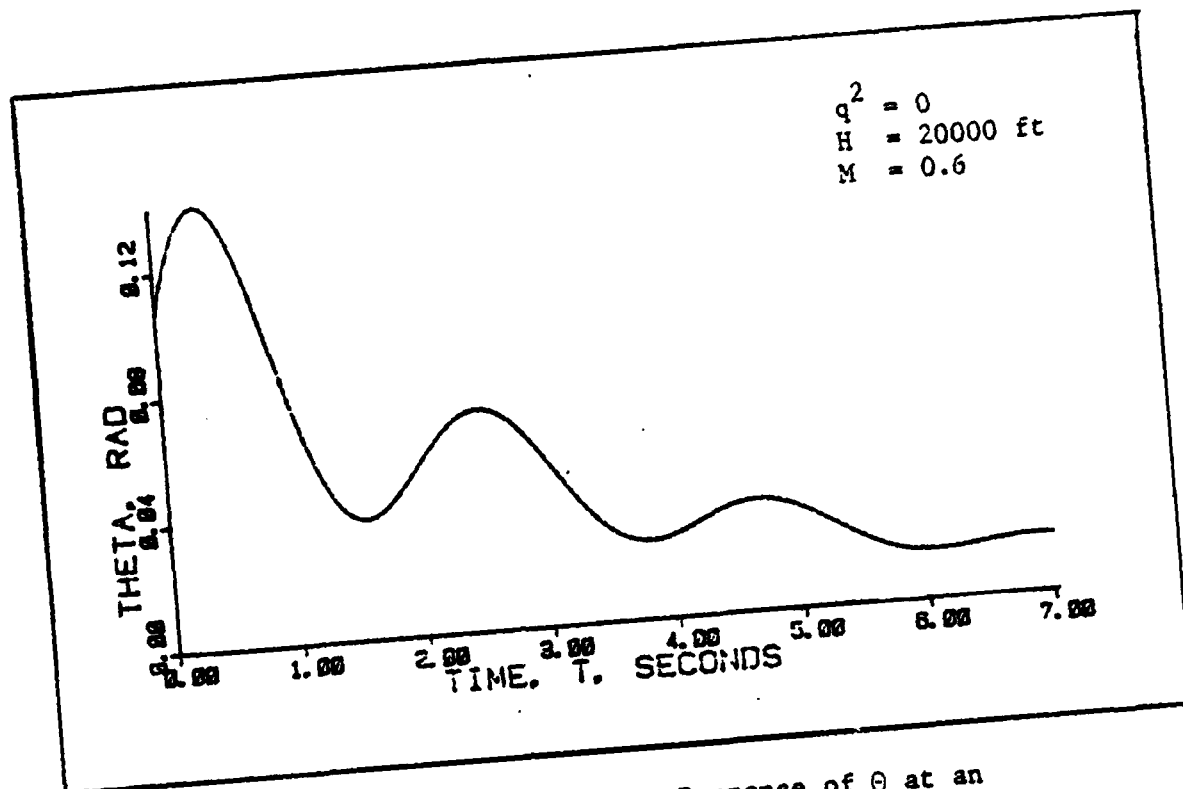


Figure D-2: Unrobustified Mean Response of  $\Theta$  at an Off-Design Condition

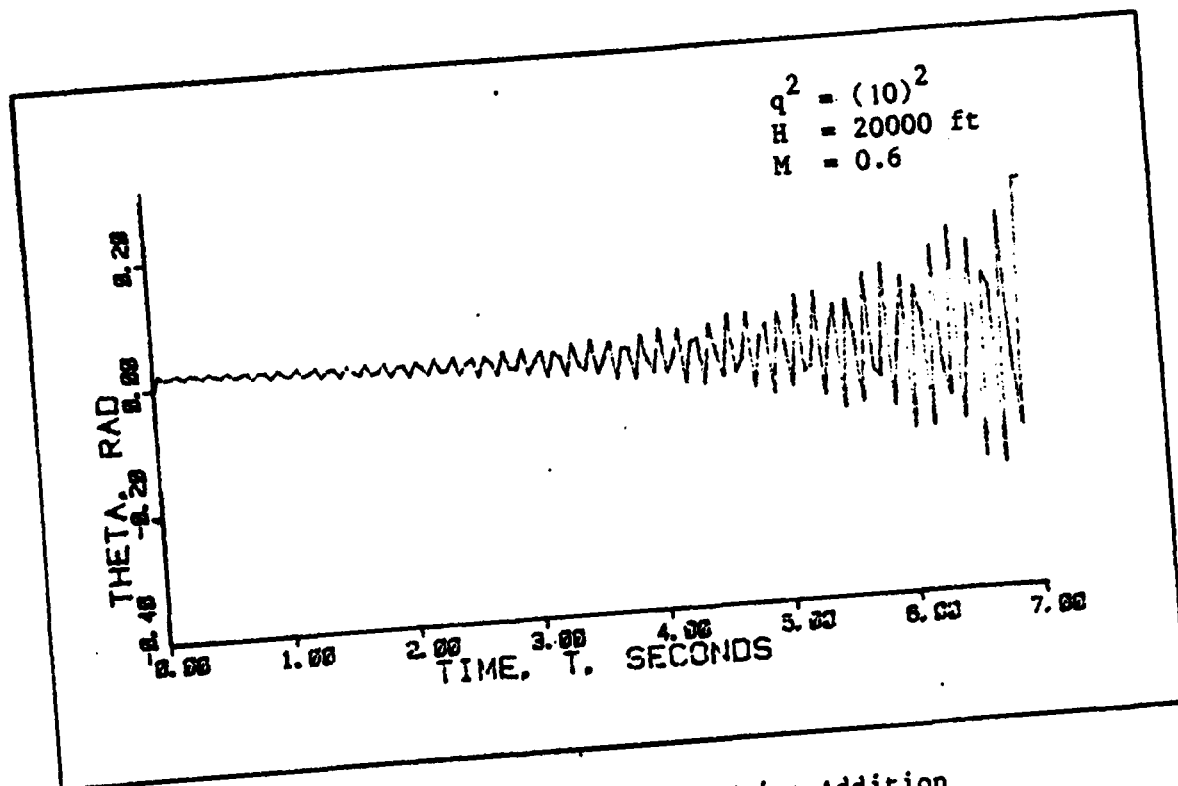


Figure D-3: Mean of Theta With White Noise Addition

of this controller is actually better than the one described in Chapters V and VI. The transient time and steady-state error are substantially better.

Figure (D-2) shows the mean of  $\Theta$  response when the flight condition is changed to an altitude of 20000 feet and a Mach number of 0.6. This response is still stable although slower and more oscillatory than the previous case. Figures (D-3) through (D-5) show the response of the same controller with progressively higher strengths of white noise added to the model at the control entry points. It is seen that the stability of the controlled system is lost with the application of the robustification technique. Examination of the eigenvalues of the closed-loop system matrix disclosed that, for any non-zero value of  $q$ , some of the eigenvalues are driven into the right-half  $s$ -plane.

If the response of the system at another off-design point (10000 feet,  $M = 0.6$ ) is examined (Figure D-6). It is seen that the mean of  $\Theta$  is diverging rapidly. However, as shown in Figure (D-7), if white noise of strength  $q^2 = 1000$  is added to the system model, the divergence is considerably slower. Figures (D-8) and (D-9) demonstrate that adding white noise with higher strength will stabilize the pitch attitude response. The mean of  $\Theta$  is actually converging to a steady-state value. Figures (D-10) through (D-12) show the same trend for the pitch rate,  $\dot{q}$ .

The figures in this appendix show that the trends observed in Chapter VI do not apply if the model used for controller and Kalman filter design is non-minimum phase. A case was shown where the stability of the system at an off-design flight condition was recovered by applying

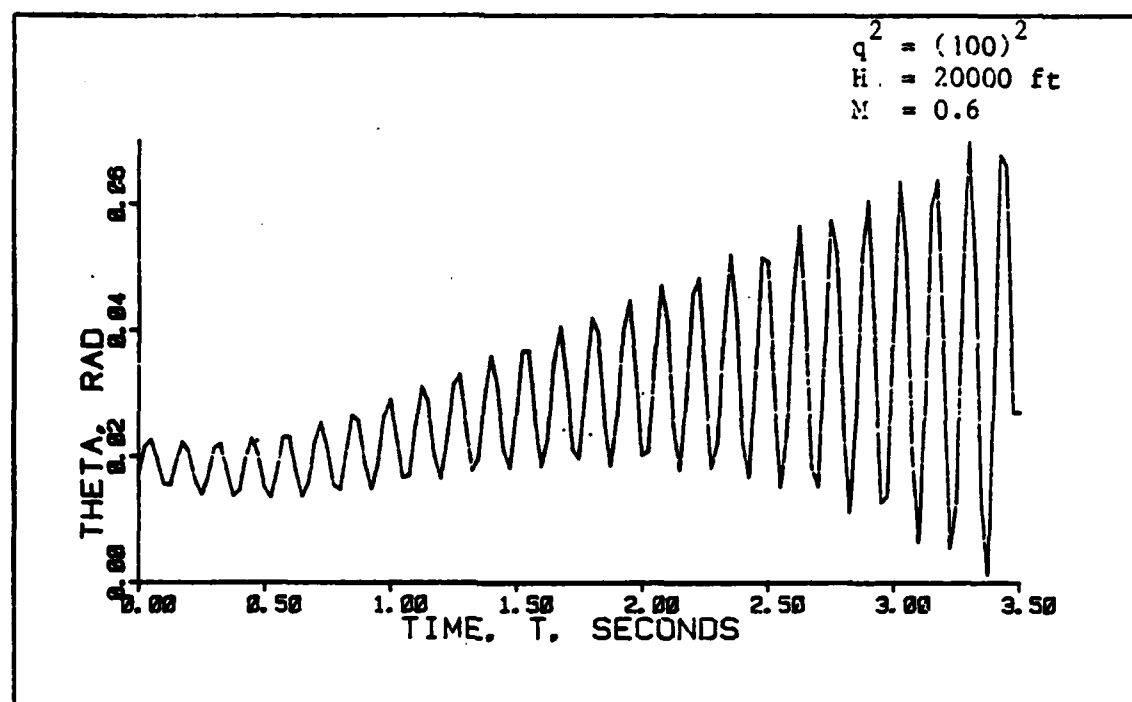


Figure D-4: Mean of Theta With White Noise Addition

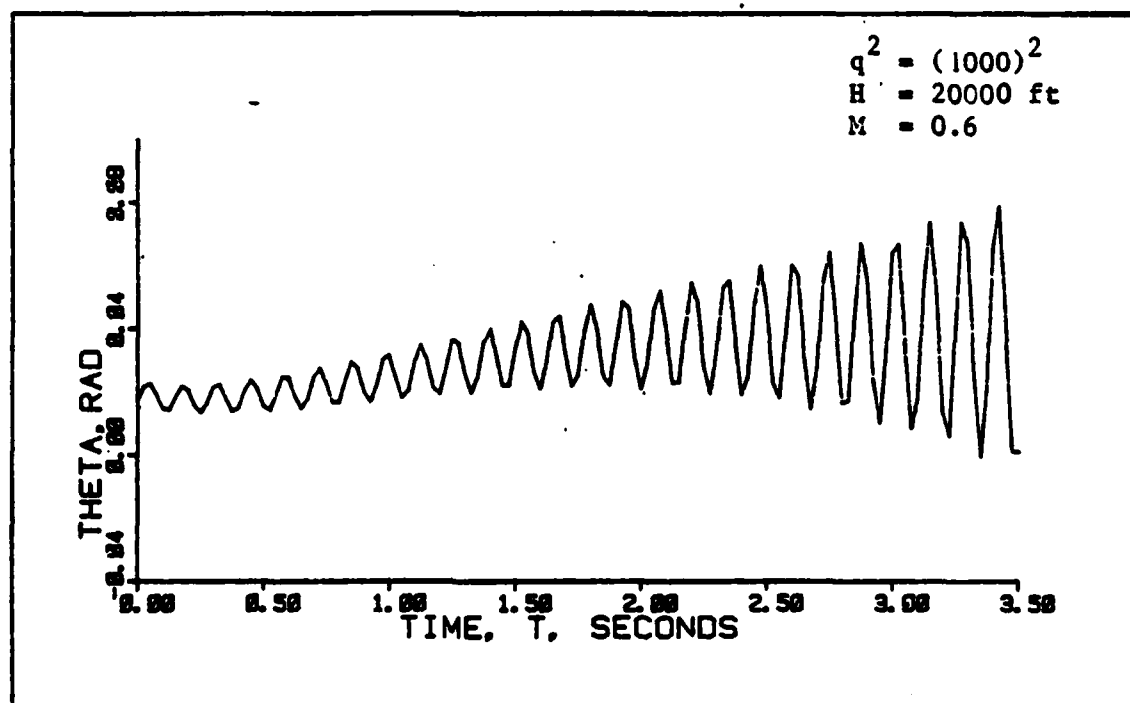


Figure D-5: Mean of Theta With White Noise Addition

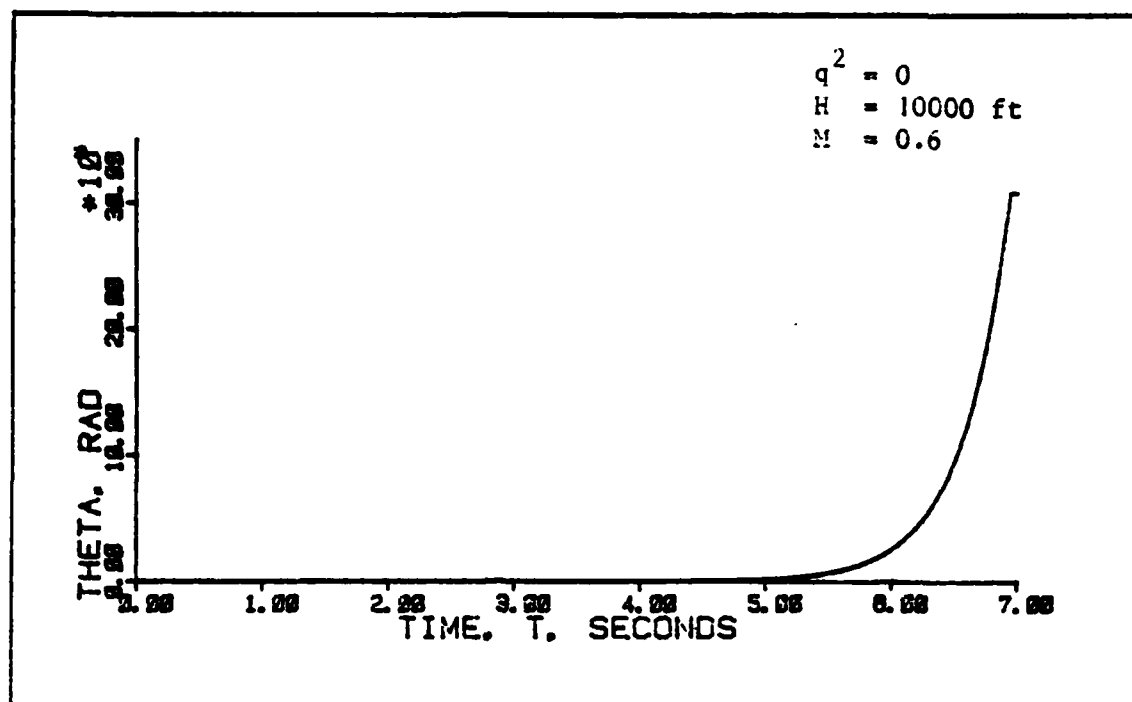


Figure D-6: Unrobustified Mean Response of  $\Theta$  at an Off-Design Condition

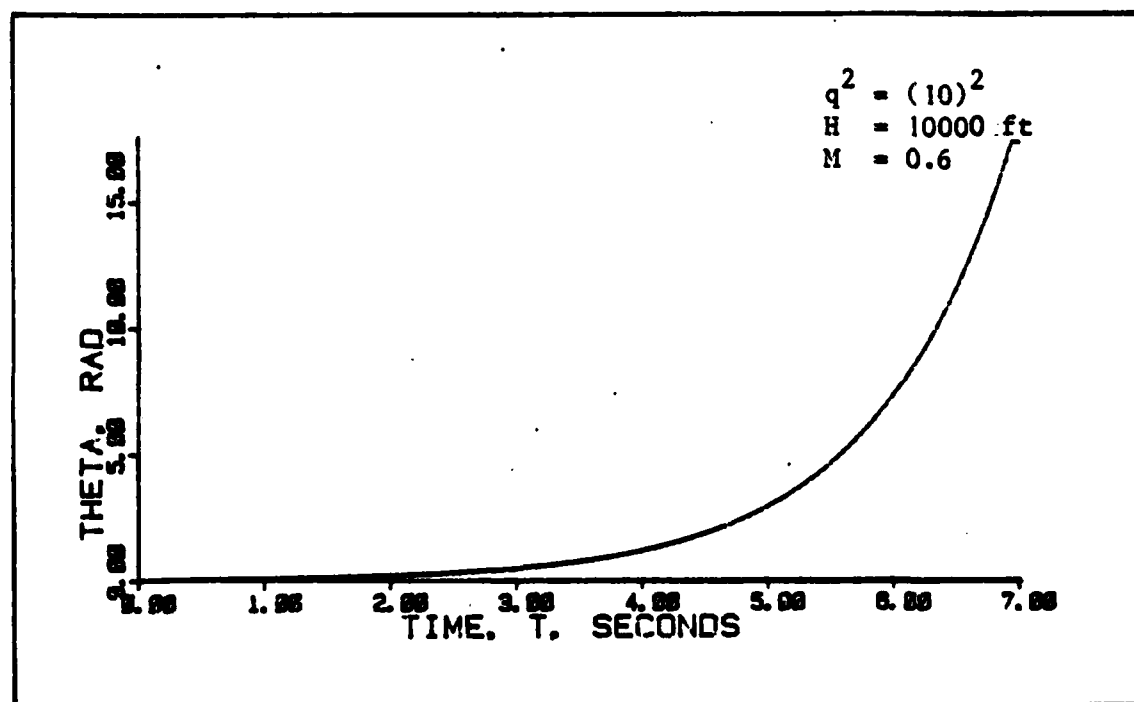


Figure D-7: Mean of Theta With White Noise Addition

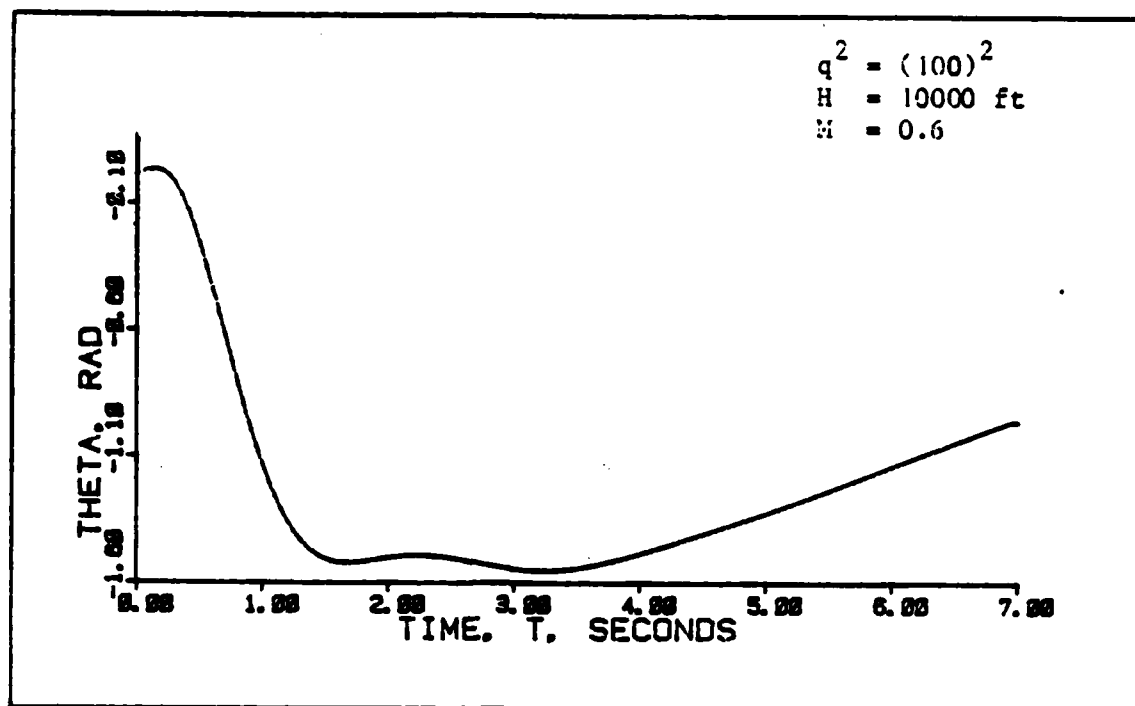


Figure D-8: Mean of Theta With White Noise Addition

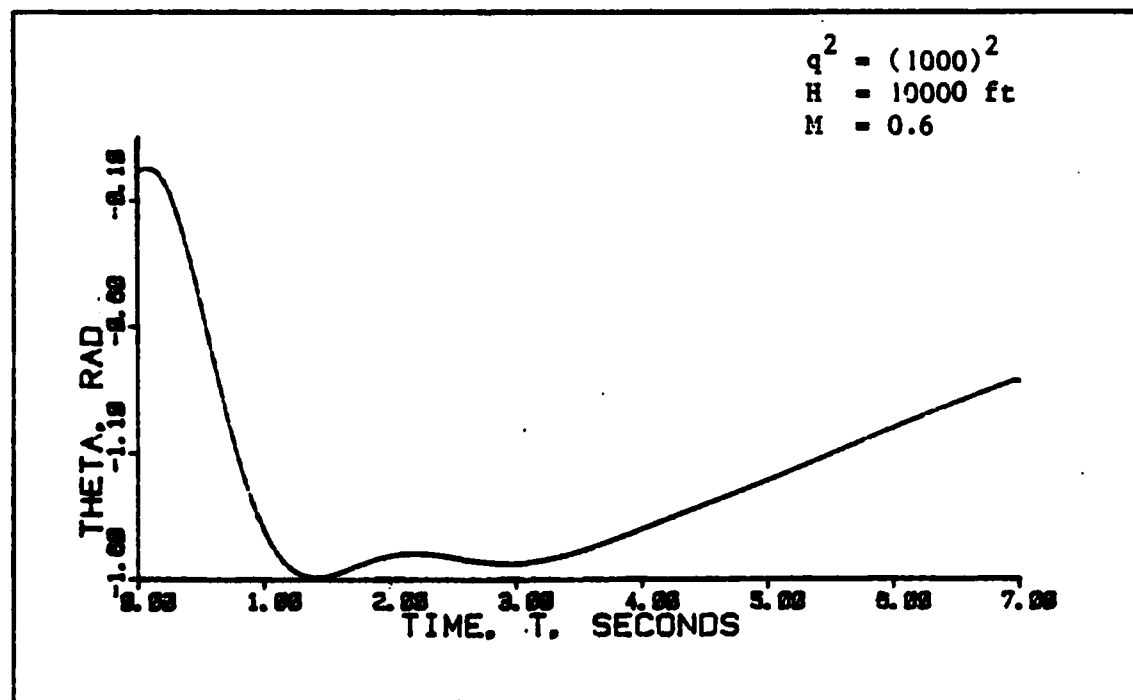


Figure D-9: Mean of Theta With White Noise Addition

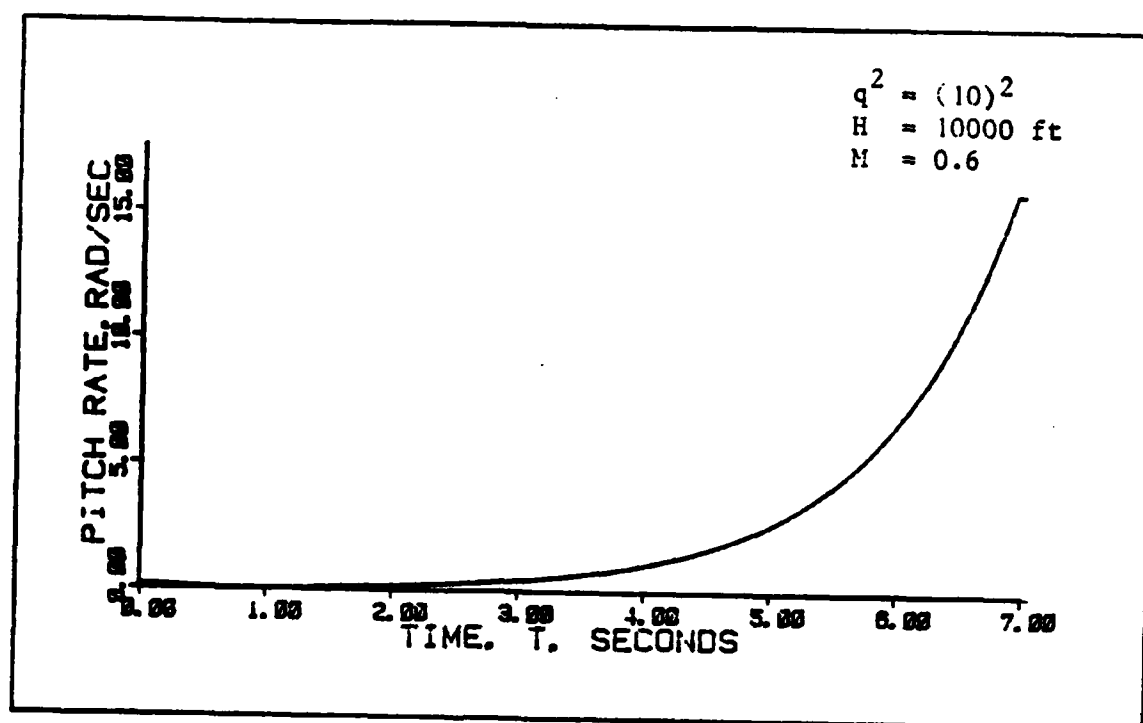


Figure D-10: Mean of  $q$  With White Noise Addition

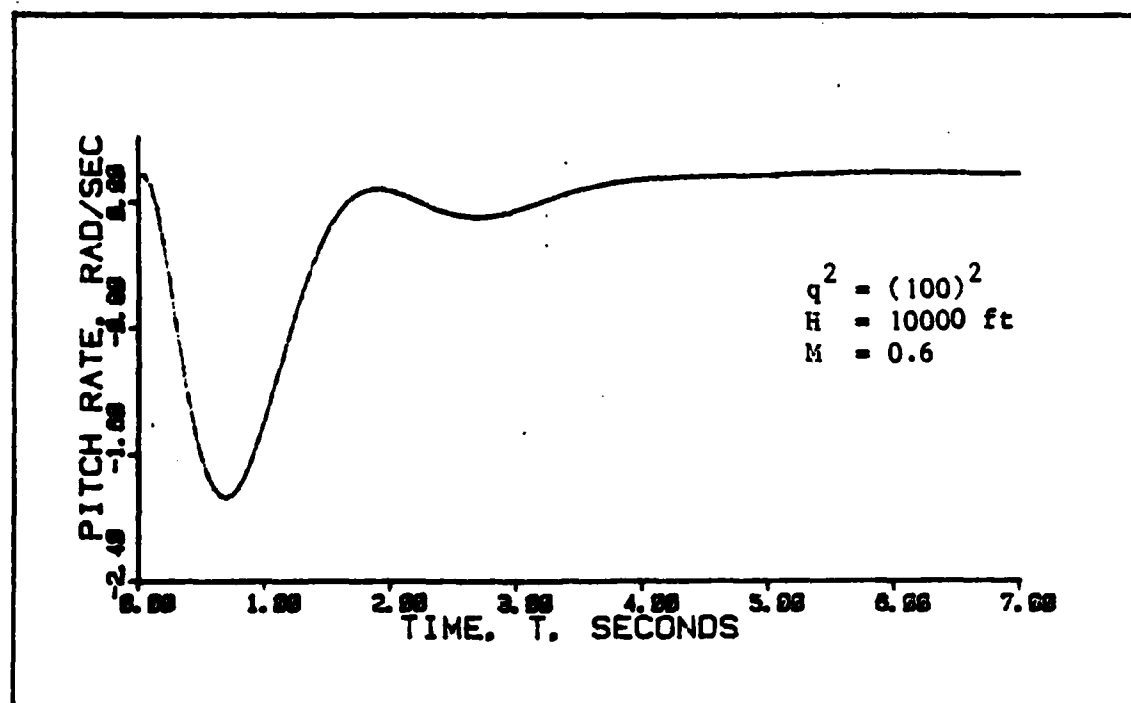


Figure D-11: Mean of  $q$  With White Noise Addition

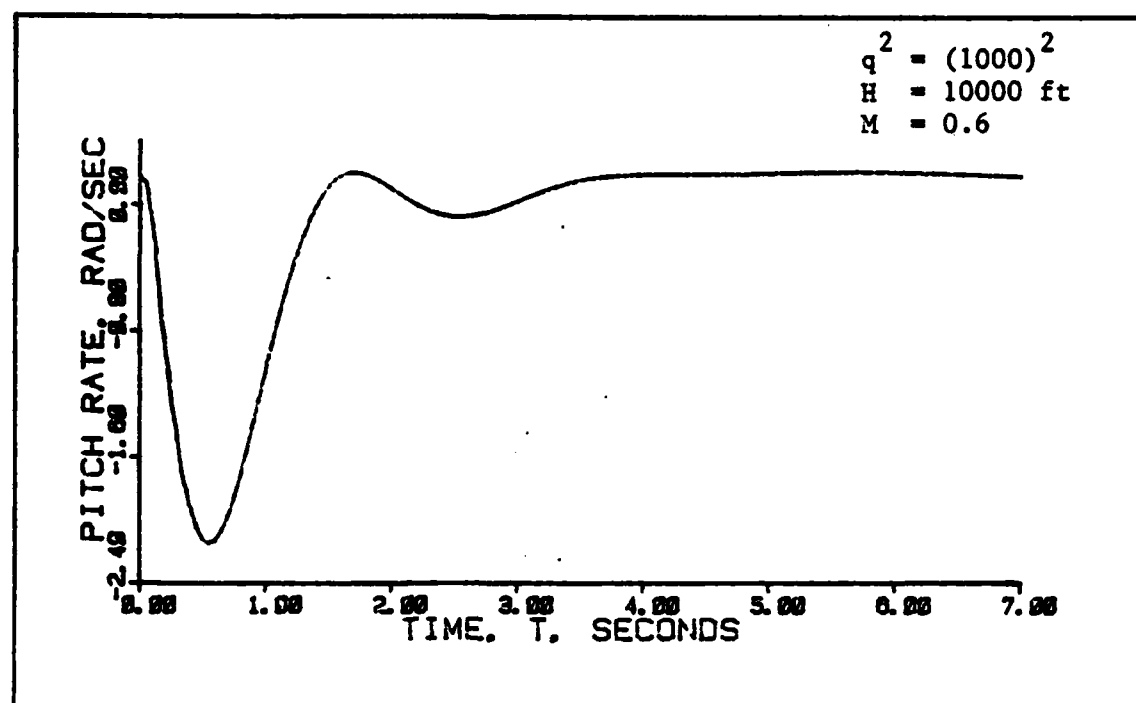


Figure D-12: Mean of  $q$  With White Noise Addition

the robustification technique. However, the noise addition destabilized the response of the system at another flight condition that was initially stable.

Reference 28 deals with the robustification technique of adding time-correlated noise to a system model at the control entry points. It states that this method is not constrained to minimum phase models as in the Doyle and Stein technique. This claim was not examined in this thesis, however, the models used would form a good basis for future research in this area.



Vita

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<p>This study examines the concept of robustifying a controlled system against differences which may exist between the real world system and the low-order design model upon which the controller design is based. The types of controllers considered are based upon the Linear system model, Quadratic cost, and Gaussian (LQG) noise process methodology of optimal control theory. It is assumed that full-state feedback is not available and a Kalman filter is employed to provide state estimates to the controller. Both continuous-time and sampled data controllers are considered.</p> <p>Two robustification techniques are considered. The first is the method of injecting zero mean white Gaussian noise into the design model at the point of entry of the control input during the process of tuning the Kalman filter. The second method is an extension of the first, where the white noise is replaced by a time-correlated noise. This allows the primary strength of the noise to be concentrated only in the frequency range where robustification is desired. Comparing the results of applying the two methods allows a designer</p>			
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to make a trade-off between the amount of desired robustification and the performance degradation at the design conditions which occurs when the techniques are applied.

Both methods are found to improve substantially the robustness properties of the controllers considered. For the specific flight control problem considered in this thesis, the technique of injecting white input noise into the design model produced the desired degree of robustification without prohibitively degrading performance at the design conditions.

The second method, though effective, did not yield substantial enough performance benefits over the first to warrant use in actual implementation.

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