

COMPUTER IDENTIFICATION AND CONTROL
OF A HEAT EXCHANGER

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ABSTRACT

The heat exchanger under study is an experimental model of the shell and tube heat exchanger. The self-tuning regulator algorithm of Åström and Wittenmark [17] is implemented on a digital computer, with the object of controlling the water temperature by manipulating the steam flow rate. The algorithm consists of an on-line least-squares parameter identification of a linear model, combined with a minimum variance control tuned according to the parameter estimates of the model.

Instrumentation and characteristics of the steady-state, dynamics, and noise of the heat exchanger are described. A model of the heat exchanger is obtained from a step response test for the purpose of comparison with the model obtained from the least-squares identification algorithm.

The algorithm does not determine the order of the plant model. This must be decided a priori. Other a priori decisions include the choice of a suitable sampling interval, an estimate of the noise variance, and the starting value of the model parameters. The results of experimental runs with variations in all of these are presented. Convergence difficulties in the parameter identification are encountered in some cases.

RESUME

L'échangeur de chaleur sous consideration est un modèle expérimental du type vapeur-eau. L'algorithme de contrôle de Åström et Wittenmark [17] est appliqué à un petit ordinateur. L'algorithme consiste en un estimateur des moindres carrés d'un modèle lineaire avec un régulateur à variance minimum calculé sur le modèle estimé. L'objet est de contrôler la température de l'eau par la manipulation du débit de vapeur.

Les caractéristiques en régimes permanent et dynamique sont décrites, ainsi que celles du bruit. Un modèle de l'échangeur est obtenu à partir de la réponse à l'échelon pour fin de comparaison avec les résultats de l'algorithme.

L'algorithme ne determine pas l'ordre du modèle, qu'il faut décider a priori. D'autres décisions a priori sont: le choix d'un intervalle d'échantillonnage, la variance du bruit, et les valeurs initiales des paramètres du modèle. Les résultats d'expériences avec variations dans chacun de ces quantités sont présentés. On a observé des difficultés de convergence de l'algorithme en quelques cas.

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CHAPTER 1

INTRODUCTION

1.1 Survey of Identification

Adaptive control is the subject of a great deal of current research. As yet, no general framework has been developed for the design of the adaptive controller. At present, the adaptive control system usually consists of two sub-systems: an identifier that determines system parameters, and a controller whose gains depend on the identified parameters.

Several problems arise from the use of this identifier/controller structure. One difficulty is that in order that the identification be accurate and complete, the control signal must excite all modes of the system. Furthermore, an identifier that works in open-loop or with a particular controller in closed-loop may not work with some other controller; with some controllers the identifier may not converge or may produce biased parameter estimates. Still another complaint of the present state of adaptive control, as pointed out by Tse and Athans [1], is that the adaptive controller should use not only the instantaneous estimates of the parameters but also the uncertainty

connected with the estimates. Because of the lack of a framework, the approach to these problems is not unified, as is evidenced by the widely differing methods of identification.

The experimental work presented in this paper is connected with one particular identifier/controller combination. Before describing the nature of the experiment, a brief survey of the methods of system identification follows. The emphasis of the review is on the philosophy behind the various methods. The purpose is to put in perspective the particular identification/control scheme chosen here, hence also the experiment, within the adaptive control field.

There are a number of identification methods that have been developed. Among the most popular in the North American literature are the Bayesian, the Correlation methods, the Extended Kalman Filter, the Stochastic Approximation methods, and the Least Squares method.

Bayesian: In Bayesian estimation a probabilistic approach is taken. The solution of the estimation problem consists of the probability density of the unknown. From this density an optimal estimate is made by taking the mean (minimum variance estimate), mode (maximum likelihood estimate), etc., depending on the loss function chosen.

Rather than producing an empirical probability density, the Bayesian estimator fine-tunes an assumed form of the density function. That is, a model of the unknown distribution is assumed, as well as an a priori distribution of the unknown model parameters. Via Bayes' rule, the Bayesian estimator modifies the distribution of the unknown parameters on the basis of observed data.

A feeling for Bayesian estimation can be obtained by looking at Bayes' rule. Suppose a quantity x is observed and a model of the distribution of x is assumed. The model contains an unknown parameter θ . The desired distribution is $f(x|\theta)$ with θ known. In order to get θ , the problem is turned around via Bayes' rule to get the distribution of θ based on observed values of x :

$$f(\theta|x) = \frac{f(x|\theta)f(\theta)}{\int_{-\infty}^{\infty} f(x|\theta)f(\theta)d\theta}$$

$f(x|\theta)$, whose form is assumed, is called the likelihood function. $f(\theta)$ is the assumed a priori distribution of θ , whereas $f(\theta|x)$ is the a posteriori distribution. A recursive computation is possible by substituting the a posteriori distribution for the a priori distribution after every observation of x .

The obvious difficulty with the method is the problem of choosing the likelihood function and the a priori

distribution (of unknown θ). Convergence requires that the likelihood function be a reasonable approximation of the true distribution of x . Furthermore, the computation is lengthy unless the likelihood function and the a priori distribution are chosen so that the a priori and a posteriori distributions are of the same form.

General presentations of Bayesian estimation can be found in [2] and [3]. In [4] Lainiotis derives a Bayesian recursive algorithm where the unknown distribution is modelled as an expansion of a set of orthogonal functions. The algorithm determines the coefficients of the expansion. The choice of the orthogonal basis is left as an open problem. In [5] Lainiotis presents an estimator that consists of a bank of Kalman filters matched to every possible value of a quantized set of the unknown θ . These filters produce the conditional estimates $f(x|\theta)$, and their outputs are weighted by the a posteriori probabilities. Computer simulations of the estimator were successful. Obviously, however, accuracy requires fine quantization. To meet this problem, Sengbush and Lainiotis [6] have proposed and tested a method that simplifies the parameter quantization.

Correlation methods: The idea behind the correlation methods is to extract the system parameters

from correlations of the output. The autocorrelation of the output is observed. Then equations relating the autocorrelation to the system parameters are solved for the unknown parameters. A more efficient method is obtained by using the autocorrelation of the innovation (one-step-prediction-error) sequence rather than that of the output directly. (The innovation is the difference between the actual output and the output predicted from the previous sample interval according to the estimated system.) The reason for the increased efficiency is that the innovation sequence is less correlated than the output. In fact, as the estimation converges, the innovation becomes more and more white. For both output and innovation correlation methods, the estimates are asymptotically unbiased, and higher efficiency is obtained by using higher order correlations.

In [7] Mehra derives two recursive algorithms, one based on output correlation, the other on innovation correlation. Both algorithms determine the covariance matrices of the process and measurement noises of a system. In [8, 9] Mehra presents an algorithm that identifies a system in canonical form. He also proposes a recursive method for determining the order of the system when it is not known. The method is based on testing

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the innovation sequence for whiteness.

Extended Kalman Filter: An optimal estimate of the state is given by the Kalman filter, whose design requires exact knowledge of the system parameters and noise statistics. The basis of the Extended Kalman Filter is to adjoin the unknown parameter vector to the state vector and design a Kalman filter for a linearized version of the augmented system, thereby obtaining estimates of both the state and parameters. The nonlinear augmented system is linearized by retaining only the first order terms of a Taylor series expansion about the augmented state vector.

Difficulties of the method lie in the requirement of exact noise statistics, and problems in programming and bias error analysis. Treatment of the Extended Kalman Filter (including filter equations) can be found in Jazwinski [3] and Saridis [10]. Convergent computer simulations are reported in [10].

Stochastic Approximation methods: The stochastic approximation methods essentially are hill-climbing search techniques suitably modified to operate in a stochastic environment. The identification problem is posed as a minimization of a loss function V of the parameters β . The solution is obtained through an

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iterative scheme of the form . .

$$\beta_{k+1} = \beta_k + (\text{gain})_{k+1} * (\text{error correction})_k \quad (1.1)$$

In a deterministic problem, the error correction term may be a negative gradient, and the gain term may be a constant (gradient method), a direction of steepest descent (steepest descent method), an inverse Hessian (Newton's method), etc. The significant modification found in the stochastic approximation is that account is taken of degradation of calculations arising from noisy data. The major modification lies in the gain factor of Equation 1.1. The gain may be a scalar or a matrix, the latter case having faster convergence at the expense of additional computation. The gain may be a prescribed sequence, such as the series $1/k$ in the scalar case; on the other hand, the algorithm may possess a learning property that maintains a best direction of search.

The main advantage of the stochastic approximation algorithms is their great simplicity and ease of implementation. Hence, although the method is not particularly noted for speed of convergence, it is a popular method for on-line identification. It should be noted, however, that in certain cases, the parameter estimates may be biased unless additional bias correction terms appear in Equation 1.1.

A good survey on stochastic approximation.

methods can be found in a paper by Saridis [11]. Specific open-loop algorithms are presented by Saridis and Stein in [12, 13]. These algorithms are extended by Saridis and Lobbia [14] to the closed-loop case; that is, a controller using estimates of the parameters is involved. In this case, unbiased estimates are produced in spite of the additional bias problems caused by feedback-induced correlations. Successful computer simulation results are presented for both the open-loop and closed-loop algorithms.

Least Squares method: The least squares method is the result of the solution of a regression-type problem. The problem is to fit a linear regression to an input-output history, the regression being, for instance, a system model in canonical form:

$$\begin{aligned} y(k) + a_1 y(k-1) + \dots + a_n y(k-n) \\ = b_1 u(k-1) + \dots + b_n u(k-n) + e(k) \end{aligned}$$

The $e(k)$ can be interpreted as noise or simply as residuals (equation errors). As in the stochastic approximation method, the basis of the least squares approach is a minimization; specifically, the minimization of the sum of the squares of the equation errors. The problem formulation and solution, as given in Åström and Eykhoff [15], are briefly reviewed in Chapter 3 of this paper. It is sufficient to say here that a recursion formula for the

solution is obtainable and it takes the form of Equation 1.1, and thus appears similar to a stochastic approximation algorithm. The distinction is that the error-correction factor is the one-step-prediction-error, a scalar; and the gain factor is a "learning" vector sequence. If the equation errors are correlated, the parameter estimates will be biased. Several ways of dealing with this problem have been developed. For example, in the method of generalized least squares [15, 16], uncorrelated residuals are obtained from a filter obtained from an autoregression fit of the residuals of an ordinary least squares fit.

Many of the references mentioned thus far provide simulation test results. Direct comparisons of the various on-line identification methods, however, are relatively rare and quite recent. Two such papers are by Saridis [10] and by Isermann et al [16]. The results in [16] lead the authors to the interesting suggestion that all the successful on-line methods that use the same a priori knowledge of the process have about the same performance over a long time period. This leaves the main advantages of the methods to depend on the process; for example, on any special use of a priori knowledge that can be made, on the type of input signal involved, on programming and computational expense, etc.

1.2 Statement of the Problem

The experiment presented in this paper tests the particular identifier/controller combination as presented by Åström and Wittenmark [17]. Specifically, this combination consists of an ordinary recursive least squares identification [15] with a minimum variance control computed according to the identified parameters. The result may be viewed as a self-tuning controller. It should be mentioned that although there are quite a number of on-line open-loop identification algorithms, successful closed-loop algorithms are relatively rare. The self-tuning control algorithm in [17] has been tested by extensive simulation studies by the authors of [17], and the algorithm is easily implemented on a mini-computer.

It has been mentioned that a successful open-loop identification algorithm will not necessarily work in closed-loop. This is the case here, where, although open-loop least squares convergence is guaranteed by theory, simulations indicate that the closed-loop self-tuning algorithm does not always converge. Moreover, if the input and noise are correlated, which is the case in closed-loop, the least squares parameter estimates will be biased. Furthermore, for this particular algorithm, it is not always possible to identify all the parameters. This problem, which will be explained in Chapter 3, can be circumvented,

however, by fixing one parameter.

A very interesting theoretical result is that if the algorithm converges, it will lead to the optimal minimum variance control even if the parameter estimates are biased [17]. Thus, the bias problem is neatly eliminated here. If the order of the model is too low, however, the algorithm will not lead to a minimum variance control if it converges.

On the whole, the simulation studies show that the algorithm does converge in many cases. Very often, the difference between convergence or divergence depends simply on the initial guess of the parameter values or on the value of the fixed parameter.

In the relatively small number of experimental industrial applications, the algorithm is not without success. In [18] is reported a successful application on a moisture content control loop in a paper mill in Sweden. Models of both third and fourth order were used. In this particular experiment, the algorithm exhibited good start-up and steady-state control, no convergence difficulties, and moderate insensitivity to model order and initial parameter values. A second successful industrial application of the algorithm is reported in [19]. In this case, the plant was an ore crusher.

It is the object of the experiment to further test the self-tuning control algorithm of [17] by applying it to another real-life plant, namely a steam-water heat exchanger. A quantitative performance evaluation is not undertaken. Instead, the questions asked are very basic: Does the algorithm converge? Can conditions be found that predictably affect convergence? The latter question has not yet been satisfactorily answered either by theory or by experiment.

CHAPTER 2

The Heat Exchanger System

2.1 Introduction

The heat exchanger is a device commonly used to transfer heat from one medium to another. The heat exchanger under study is an experimental model of the shell and tube type. Steam and water are the two media involved in the heat transfer. Within this heat exchanger the water flows through a bundle of seven 1" copper tubes 62" long. This set of tubes is enclosed within a shell of 6" steel pipe and the shell is surrounded by 1" thick insulation. Steam is allowed to flow into the shell and surround the water-carrying copper tubes. As the steam condenses, its latent heat of vaporization is transferred to the water through the copper tube walls. (See [20] for a theoretical study of this heat exchanger.)

In the experiment described in this paper the controlled variable is the temperature of the water flowing out of the heat exchanger (to be called output water temperature). The controlling variable is the steam flow rate. Disturbances exist in the form of water flow rate variations, temperature measurement noise, and temperature set-point changes. The object is to link the steam flow rate to the output water temperature by a sampled-data computer feedback

system.

The feedback link is an algorithm that implements the following control law:

$$U(k) = A_1 U(k-1) + A_2 U(k-2) + A_3 U(k-3) + A_4 U(k-4) \\ + B_0 y(k) + B_1 y(k-1) + B_2 y(k-2) + B_3 y(k-3) + B_4 y(k-4)$$

where $U(k)$ is the control signal (steam flow rate), and $y(k)$ is the output error signal (output water temperature minus set-point). The choice of coefficients A_i and B_i depends on the plant (heat exchanger) dynamics and the desired system performance. In this experiment, the parameters of a linear model of the plant are identified on-line by digital computer.

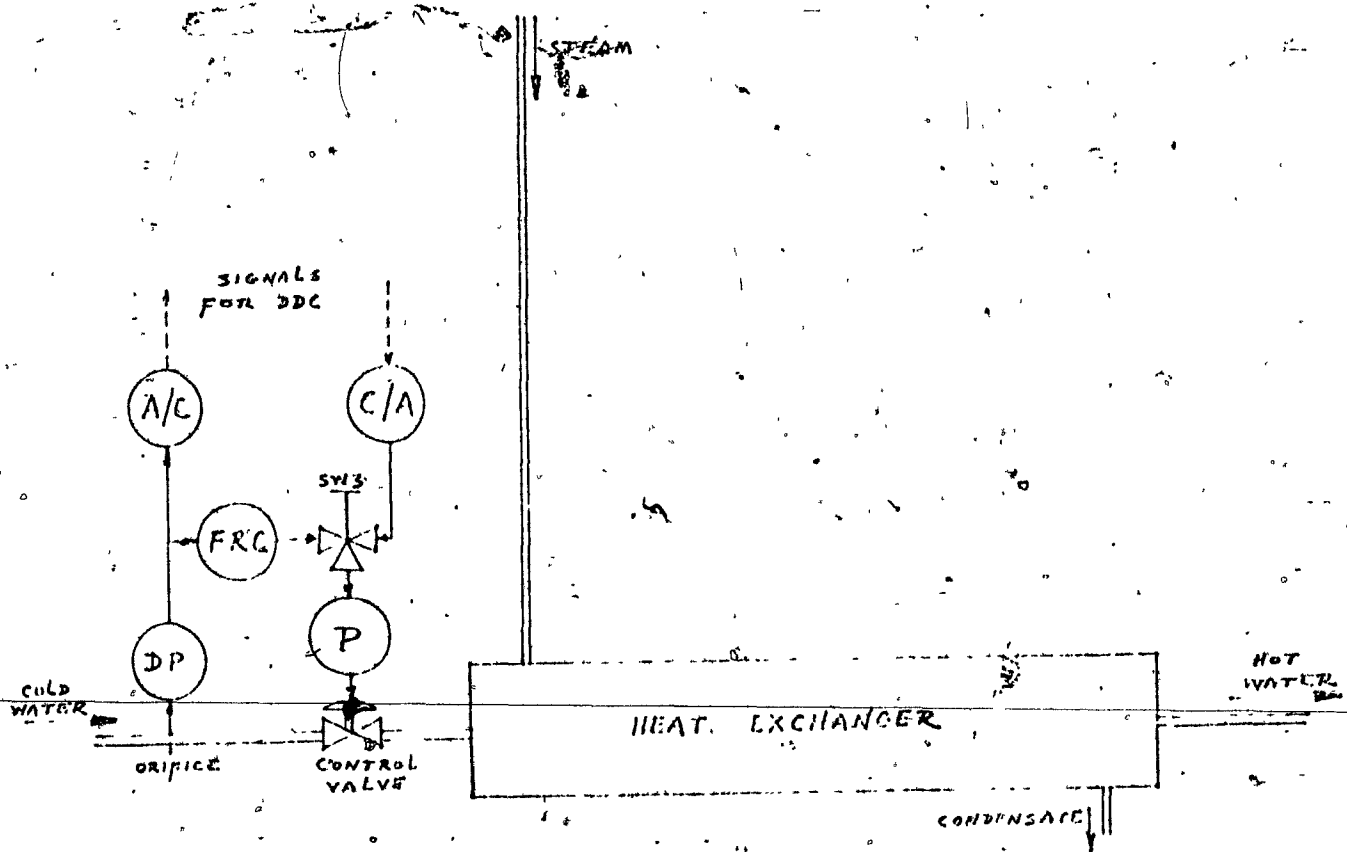
These parameters are used to compute the coefficients A_i and B_i of the control law according to a particular control policy. Specifically, the goal here is to minimize the variance of the output error. Since the identified model parameters are used to set the coefficients of the control law, the result is a self-tuning controller.

2.2 Water Flow Control

Figure 2.1 shows a block diagram of the water flow control system. The flow measurement takes place at the orifice, which essentially is a constriction in the piping. This constriction causes a pressure drop across the orifice. The pressure difference is proportional to the square of the

FIGURE 2.1

WATER FLOW CONTROL SYSTEM



FRC = Flow Recorder-Controller
 DP = Differential Pressure Cell
 P = Positioner
 A/C = Air-to-Current Converter
 C/A = Current-to-Air Converter

flow. This pressure difference signal is applied to a pneumatic differential pressure cell (dp cell), which converts the 0 to 100% differential water pressure to a 3 to 15 psi air pressure signal. This is the input to the pneumatic PID flow controller (FRC). The controller output drives a pneumatic positioner which positions the control valve stem with the aid of mechanical feedback.

The system is equipped with instrumentation that allows direct digital control (DDC) of the water flow. An air-to-current converter, whose output is bridged by a suitable resistor, provides the water flow measurement (see Figure 2.1). The computer control signal is applied to a current-to-air converter which then drives the positioner. The setting of switch-valve SW3-determines whether the positioner and control valve are driven by the pneumatic controller or by DDC.

The pneumatic controller did not produce good water flow control. A persistent limit cycle existed, even at minimum gain settings of the controller and positioner. The magnitude of the limit cycle was smallest at minimum controller gain with the positioner disabled and bypassed, but it was still perceptible. There appeared to be excessive gain in the loop, possibly due to an oversized control valve. Since good water flow regulation is not important to this

experiment, the loop was opened by operating the controller in manual mode.

2.3 Water Temperature Control

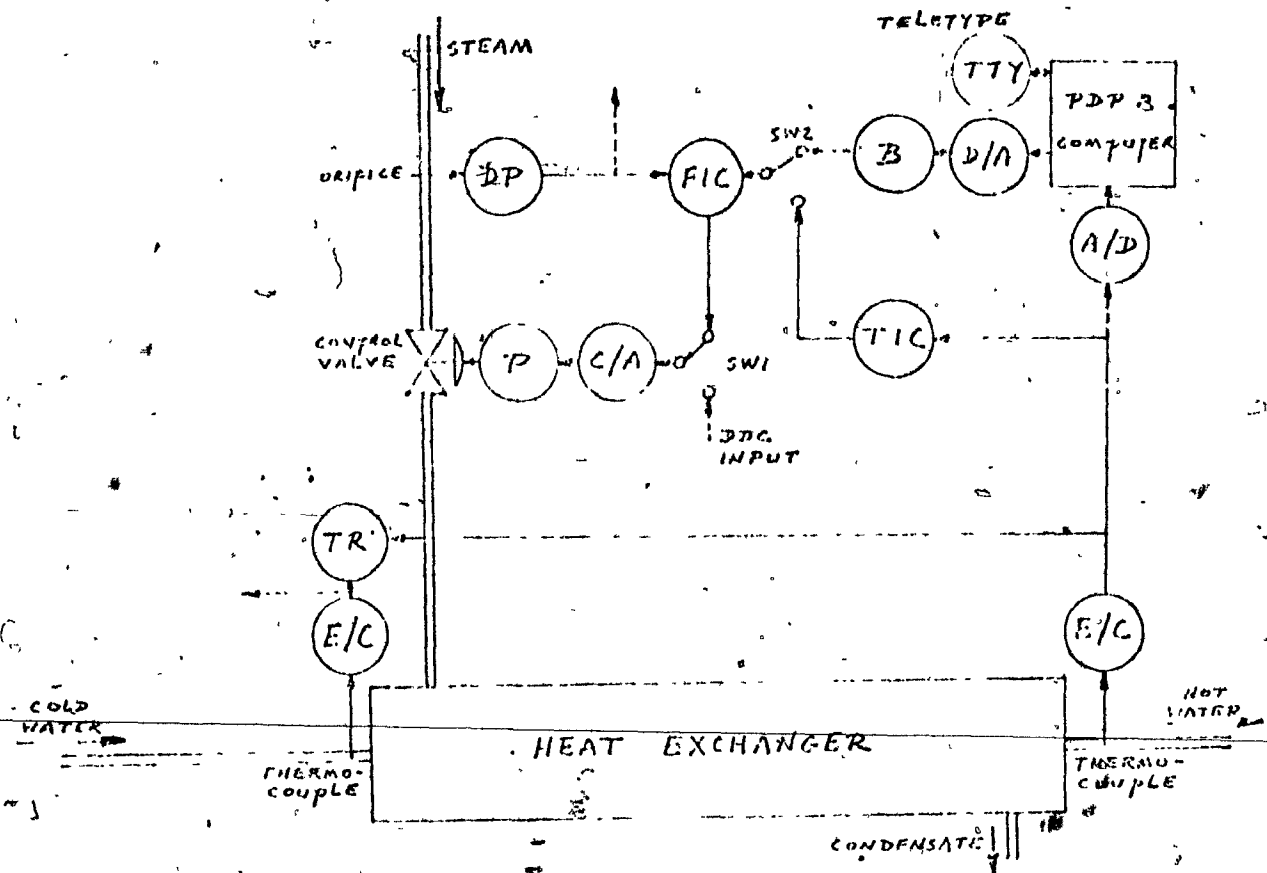
Figure 2.2 shows a block diagram of the water temperature control system. It consists of two loops in cascade. The inner loop is a steam flow control loop. The outer one is a temperature control loop, which determines the set-point of the inner flow loop.

In the inner loop the flow measurement is effected by means of an orifice and an electronic dp cell whose output is a 10 to 50 ma current. This is the input to the electronic PI flow controller (FIC). The controller output feeds a current-to-air converter which in turn drives a positioner and the steam control valve. The flow controller set-point signal is obtained from switch SW2.

At SW2 the outer temperature loop feeds the inner flow loop via the set-point signal. The temperature is measured by a copper-constantan thermocouple. After being converted into a 10 to 50 ma current, the temperature signal feeds two parallel blocks: an electronic PID temperature controller (TIC) and a PDP-8 computer system. Switch SW2 determines which of these temperature control blocks defines the set-point for the steam flow controller. Throughout this

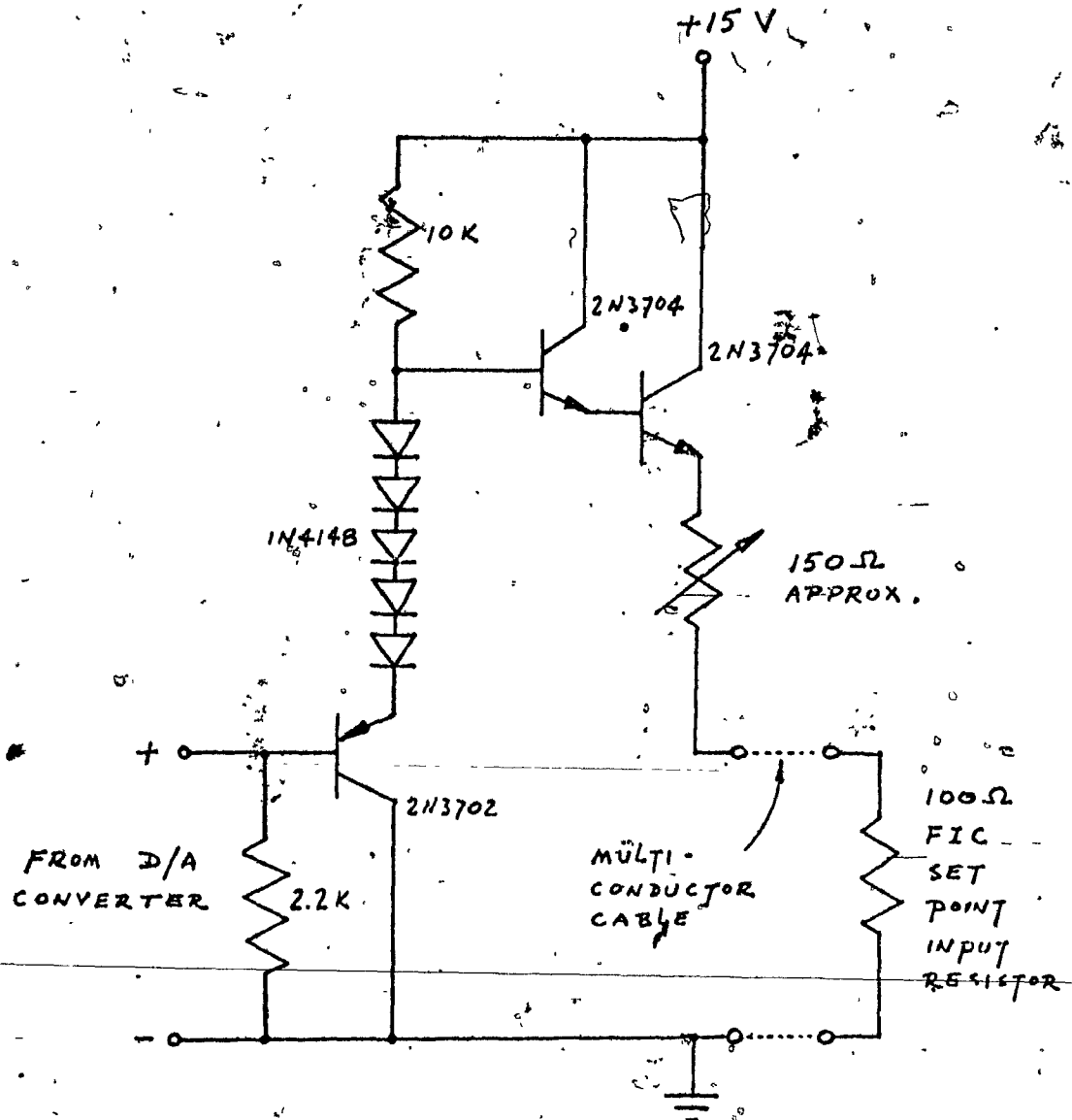
FIGURE 2.2.

WATER TEMPERATURE CONTROL SYSTEM



FIC = Flow Indicator-Controller
 TIC = Temperature Indicator-Controller
 P = Positioner
 DP = Differential Pressure Cell
 C/A = Current-to-Air Converter
 E/C = Emf-to-Current Converter
 TR = Temperature Recorder (Dual)
 A/D = Analog-to-Digital Converter
 D/A = Digital-to-Analog Converter
 B = Buffer

FIGURE 2.3



BUFFER CIRCUIT DIAGRAM

experiment, SW₂ was in the position shown in Figure 2.2; that is, the computer assumed the role of temperature controller.

In the case of computer control, the 10 to 50 ma current signals are not directly appropriate. At the input, the temperature measurement current from the emf-to-current converter is allowed to flow through a 200 Ω resistor. This provides a 2 to 10 volt input to the A/D converter. At the output, the 0 to 10 volt signal from the D/A converter goes through a buffer to provide the 10 to 50 ma flow controller set-point. A diagram of the buffer circuit is shown in Figure 2.3.

2.4 The Control Program

The computer used in controlling the heat exchanger is a Digital Equipment Corporation PDP-8/f. This is a general purpose mini-computer having a memory of 4096 12-bit words and a cycle time of 1.5 μ s. Peripherals include a teletype, a high speed paper tape reader/punch, an eight-channel analog-to-digital (A/D) converter, and a four-channel digital-to-analog (D/A) converter. Both the A/D and D/A converters operate on -10 to +10 volts and have conversion times of 80 μ s and 10 μ s respectively.

The computer program consists of two parts. One

part is the control program. Its function is to communicate with the operator/user and implement the control law. This program can be used alone if the user himself wishes to tune the controller; that is, specify the coefficients of the control law. The second part of the program, the identification program, was designed to overlay the control program and automatically specify the coefficients according to the plant identification. The resulting controller thus tunes itself. The identification program will be discussed in Chapter 3. The main features of the control program are described in the following paragraphs.

The control program can handle two inputs U_1 and U_2 , and two outputs y_1 and y_2 . The y_1 and y_2 are actually output error signals of the form $y_i = z_i - Y_i$, where z_i are set-points defined by the operator, and Y_i are the actual measured outputs. In the case of this experiment, U_1 drives the buffer (B in Figure 2.2) and y_1 is the temperature voltage across the 200Ω resistor at the A/D input.

The complete general control law has the following form:

$$\begin{aligned}
 U_1(k) = & A_1 U_1(k-1) + A_2 U_1(k-2) + A_3 U_1(k-3) + A_4 U_1(k-4) \\
 & + B_0 y_1(k) + B_1 y_1(k-1) + B_2 y_1(k-2) + B_3 y_1(k-3) + B_4 y_1(k-4) \\
 & + C_1 U_2(k-1) + C_2 U_2(k-2) + C_3 U_2(k-3) + C_4 U_2(k-4) \\
 & + D_0 y_2(k) + D_1 y_2(k-1) + D_2 y_2(k-2) + D_3 y_2(k-3) + D_4 y_2(k-4)
 \end{aligned}$$

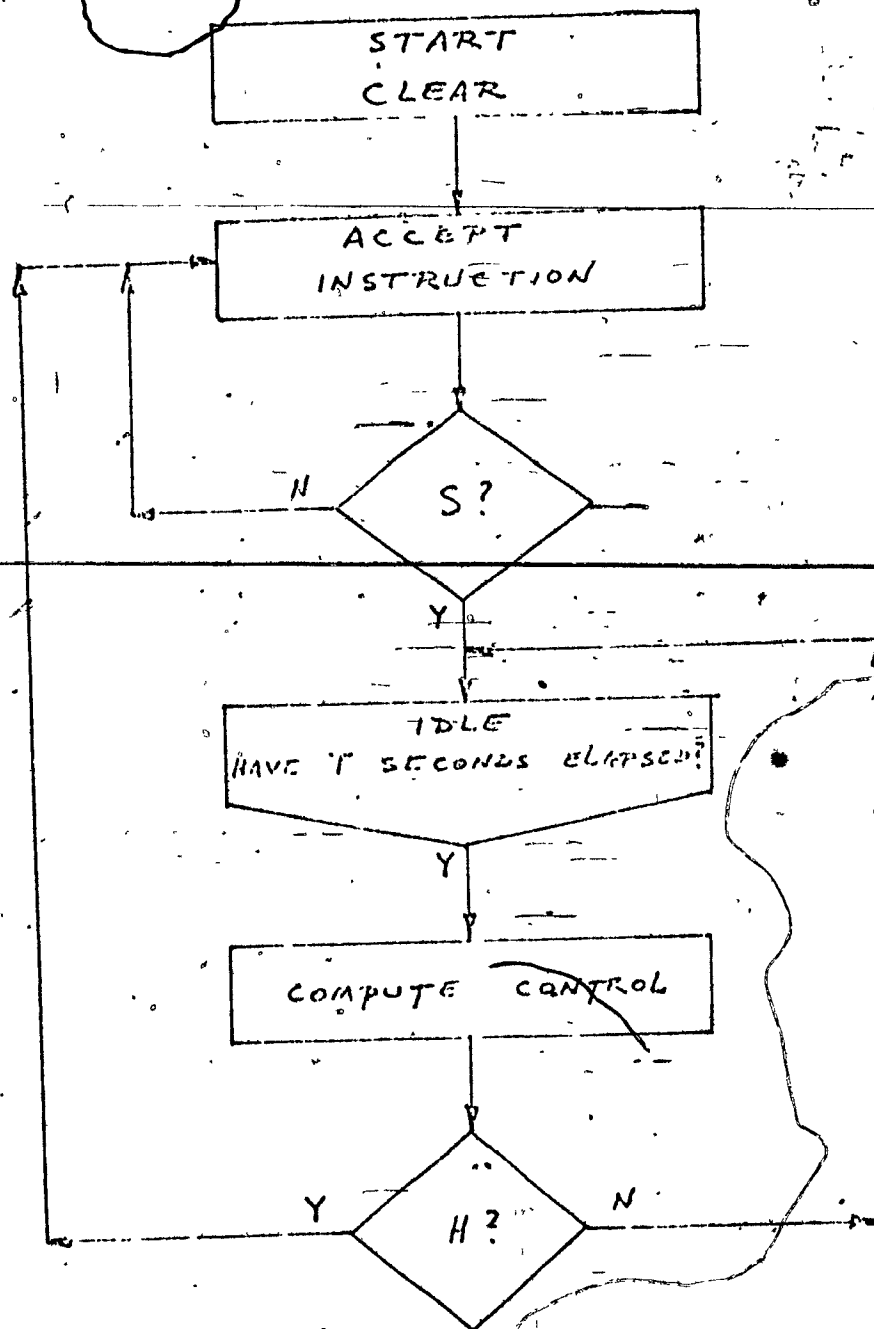
with a similar equation for U_2 . By appropriate specification of coefficients, the program can be made to control a multivariable process or two independent single-input-single-output processes. The restriction in the latter, however, is that the sampling rates must be identical.

Figure 2.4 shows a simplified flow-chart of the control program. In the instruction mode, an interpreter accepts instructions from the operator via teletype. The operator may specify the coefficients A_i , B_i , C_i , D_i for both U_1 and U_2 . He may also specify the set-points Z_1 and Z_2 , the sample interval T in seconds, and whether he wants the subsequent control history to be printed on teletype or punched on paper tape or both. If the operator types a format error, the program ignores the instruction, types a question mark, and awaits the next instruction.

When the operator types the letter S (for Start), the program enters the control mode and implements the control law that the operator just defined. In order to make changes, the operator can type the letter H (for Halt), whereupon the program returns to instruction mode at the end of the current control interval. The operator may then type in the corrected instruction(s) and again type S to continue control.

Since the PDP-8 is not equipped with a real-time clock, the A/D converter with interrupts is used in measuring

FIGURE 2.4



CONTROL PROGRAM FLOWCHART

the sample intervals. During the control mode of the program, A/D conversions are continuously being initiated and counted. When the count reaches a number related to T (specified by the operator), the count is reset to zero and a control computation begins. The control computation is done during the A/D conversion and is interrupted at the end of each conversion. Since the conversion time is relatively short, a large overhead is incurred in A/D interrupt servicing. The resulting computation time is about 200 ms per sample interval.

The computation is done in floating-point using a floating-point software package developed by Digital Equipment Corporation [21,23]. The floating-point number consists of a 12-bit two's complement exponent and a 24-bit two's complement mantissa with the radix point understood to precede the mantissa. The number is stored in three consecutive memory locations. Floating-point computation using this software package is slow but greatly facilitates handling a large range of numbers.

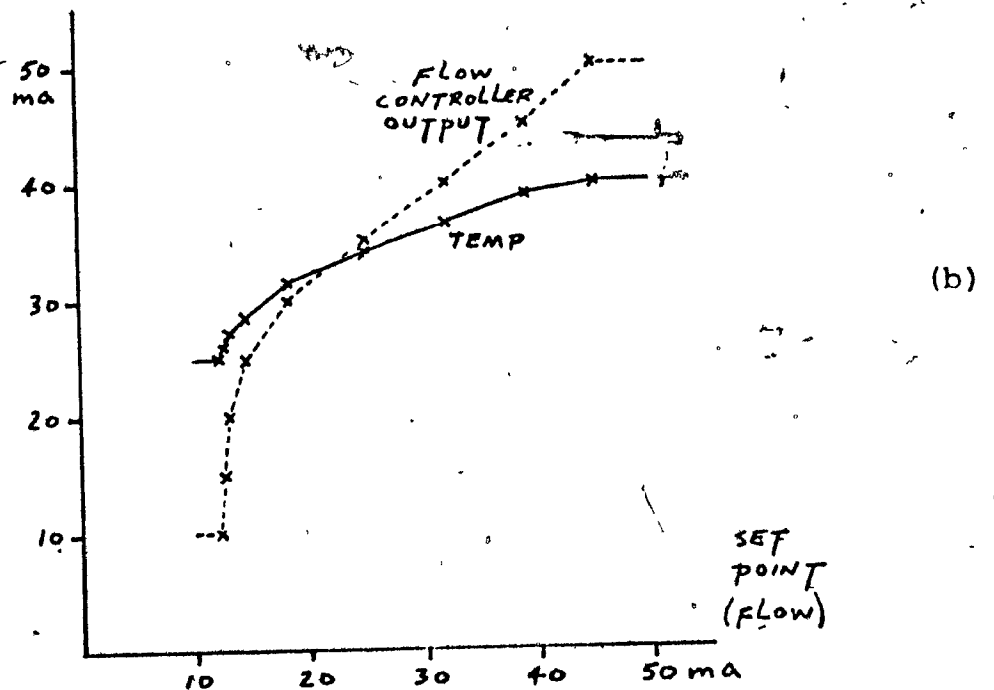
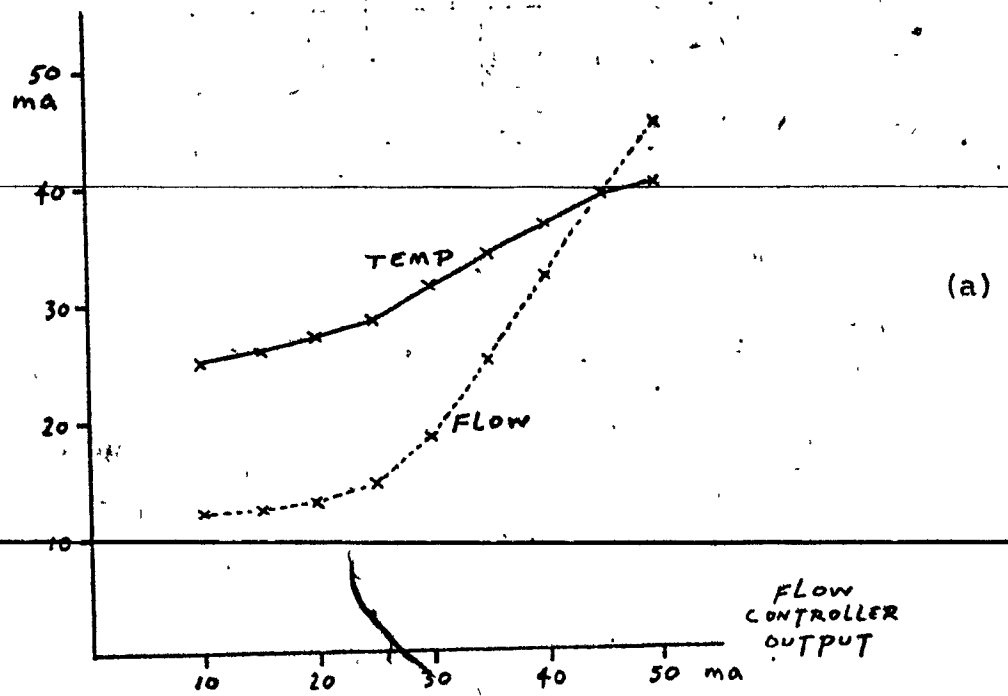
2.5 Open Loop Characteristics of the Heat Exchanger

In this section is presented some open loop data for the heat exchanger system. The data consists of the steady-state, the dynamics, and the noise characteristics.

In Figure 2.5a is shown the steady-state curve for both flow control loop and temperature control loop opened. This means simply that the steam flow controller is operated in manual mode. The solid line is the output water temperature current from the emf-to-current converter. The dotted line is the steam flow measurement from the electronic dp cell. The abscissa is the output of the steam flow controller. The flow measurement is not linear with the flow controller output due to the fact that the flow measurement is actually a differential pressure signal which is proportional to the square of the flow, and also due to control valve non-linearity.

Figure 2.5b is the steady-state curve for the flow control loop closed, i.e. the flow controller is in automatic mode, responding to the set-point signal. This in fact is the plant as far as the temperature controller is concerned. The non-linearity of the flow measurement has two significant effects on the closed flow control loop. One is that the output water temperature becomes non-linear, while it was roughly linear in the open loop. A second effect is that although the loop is stable at the lower flow rates, it becomes unstable at higher flow rates due to increased gain in the flow measurement. The resulting limit cycle disappears only at minimum controller gain, which renders the controller rather ineffective over much of the lower flow rate region. A satisfactory control is obtained by disabling and bypassing

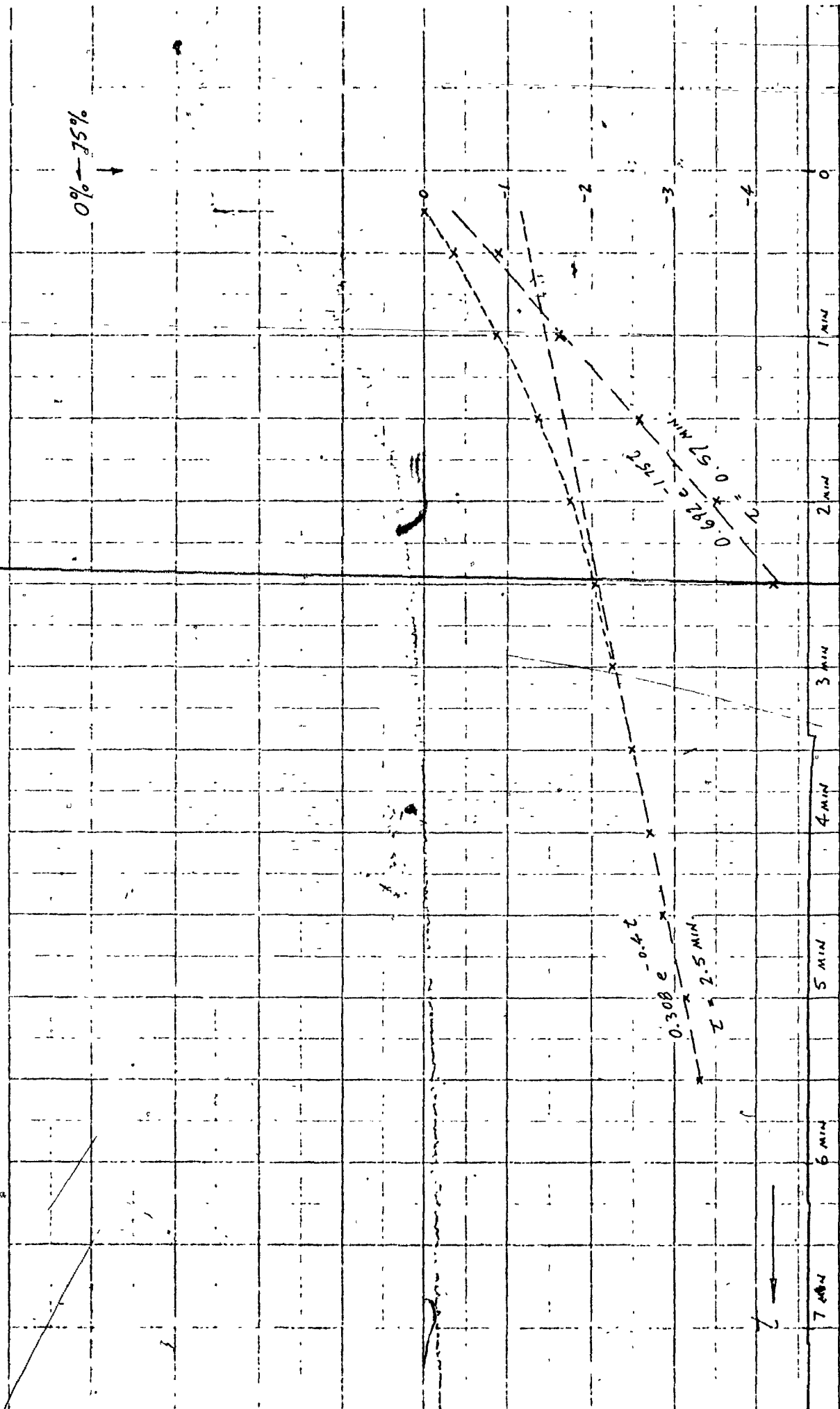
FIGURE 2.5



OPEN LOOP STEADY STATE CURVES

the positioner. The controller was then tuned to give an underdamped response with a settling time of about 1 second. A better solution, of course, would be to insert a square-root extractor following the dp cell. This would make all the curves in Figure 2.5 roughly linear and would allow the positioner to be in the loop. The value of the positioner lies in faster valve action and in the reduction of errors due to friction within the valve. The first advantage is not so important here since the 1 second flow response obtained without the positioner is quite short compared with the temperature response.

A step response analysis was done in order to get an estimate of the dynamics of the plant with both flow loop and temperature loop open. The step response is shown in Figure 2.6 with a graphical analysis immediately below it. The step was a downward step such that the flow controller output changed from 75% to 0%. Roughly, the response looks like two lags with a .25 minute delay. Normalizing the response and ignoring the delay, the response can be modelled as $c(t) = 1 - k_1 e^{-at} - k_2 e^{-bt}$, where a and b are both positive. Assuming that b is significantly greater than a , $c(t)$ approaches $1 - k_1 e^{-at}$ as t grows large. From the equality $\ln(1 - c(t)) = \ln(k_1 e^{-at}) = \ln(k_1) - at$ which follows, if $\ln(1 - c(t))$ is plotted in time, then $\ln(k_1)$ and $-a$ are readily obtained as the intercept and slope respectively of



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OPEN LOOP STEP RESPONSE

FIGURE 2.6

the $t = \infty$ asymptote. Thus, from the upper left asymptote in Figure 2.6 is obtained $k_1 = .308$ and $a = .4$. Similarly, by plotting $\ln(1 - c(t) - .308e^{-.4t}) = \ln(k_2) - bt$, k_2 and b are determined to be .692 and 1.75 respectively. Thus the response is $c(t) = 1 - .308e^{-.4t} - .692e^{-1.75t}$. The corresponding time constants are 2.5 minutes and .57 minutes. From this result, a discrete-time model of the plant will be developed in Chapter 3.

Noise presented two difficulties in this step analysis. One is accuracy, which is especially important in obtaining the $t = \infty$ asymptote. It was for this reason that a downward step change was chosen rather than upward, since it was observed that the noise was greater at higher temperatures. Another problem was that in getting a large enough signal/noise ratio for graphical analysis, it was necessary to use a large step change which would cause a non-linear response if the flow loop were closed. The fact that the temperature is roughly linear over the entire range in open loop is the reason for performing the step-response analysis with the flow loop open. This is not exactly the plant that the computer temperature controller sees, except in the case of small changes in a mid-temperature range. There, the temperature/output gain is about equal to the temperature/set-point gain, these being the open loop and closed loop gains respectively. Furthermore the relatively fast dynamics of the

flow controller are practically negligible as far as the heat exchanger model is concerned. Therefore the model obtained from the step-response test is valid as a plant model from the computer's point of view in the mid-temperature range.

An upward step change of the same magnitude (0% to 75% output) produces a noisier but very similarly shaped response, thus substantiating the temperature linearity when the flow loop is open. Near maximum steam flow, the noise becomes severe, steam begins to blow through the steam trap, and the heat exchanger dynamics appear to be different.

Because of this and because of the strong non-linearity at the low temperature region, the computer identification and control experiment was carried out in the mid-temperature range.

A simple noise analysis was done in the form of an autocorrelation of the mid-temperature steady-state. The noise was evident on a thermometer placed in the outlet water stream, hence it was not entirely measurement noise. Although the water flow recorder registered a steady flow, the flow out of the heat exchanger was not smooth and seemed to be one source of the noise.

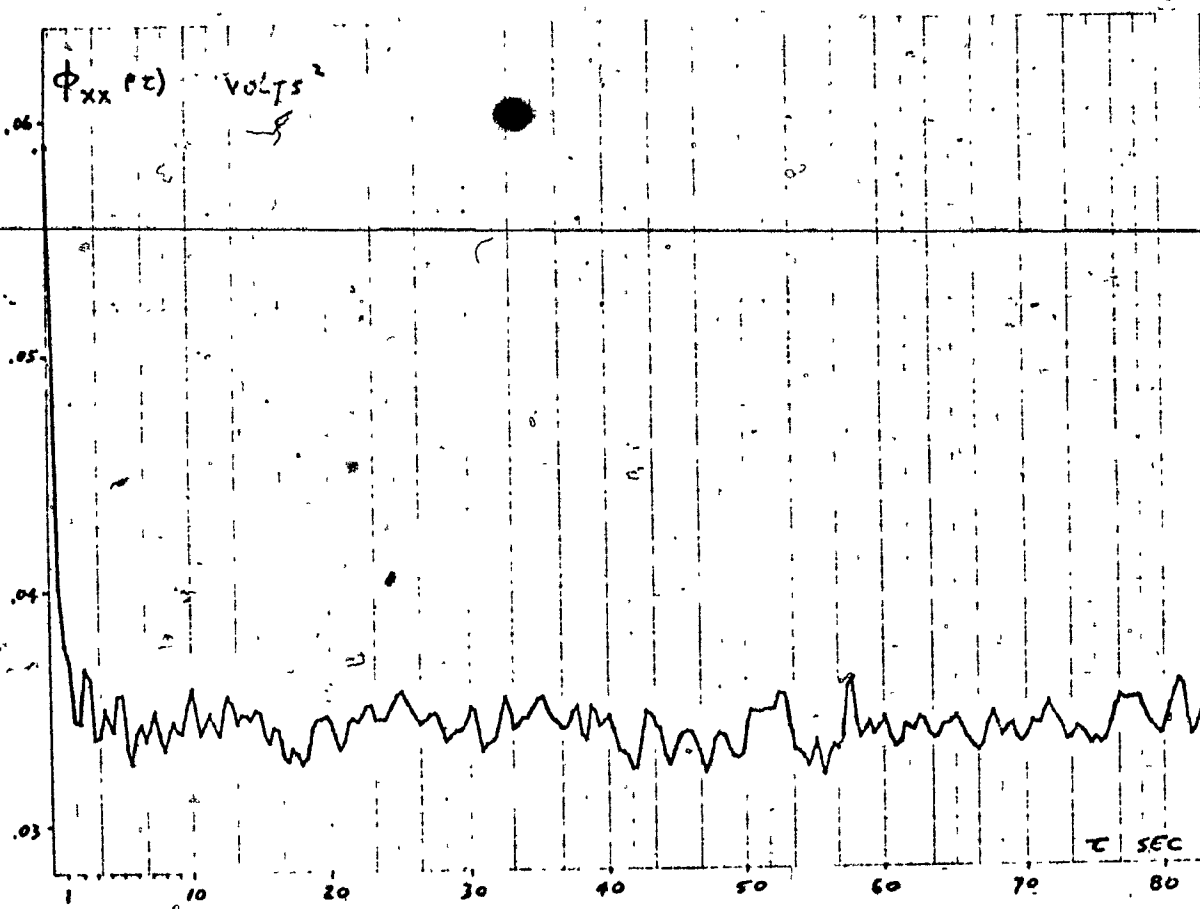
The autocorrelation was performed using the PDP-8 computer on a record of 2400 samples obtained at a sampling rate of 3 samples/second. Most of the dc value of the steady-state temperature signal was removed by subtraction in order

to minimize round-off errors in the autocorrelation calculation. The resulting autocorrelation function is shown in Figure 2.7. It is evident that to the heat exchanger system, whose dominant time constant is .57 minutes, the noise is essentially white. Assuming that the noise has zero mean, the noise variance is obtained as the difference between $\phi_{xx}(0)$ and $\phi_{xx}(\infty)$. Taking $\phi_{xx}(\infty)$ to be the mid-point of the garbage at the end of the autocorrelation curve of Figure 2.7:

$$\phi_{xx}(0) - \phi_{xx}(\infty) \approx .025 \text{ volt}^2$$

The corresponding rms value is about .16 volt.

FIGURE 2.7



AUTOCORRELATION OF THE STEADY STATE TEMPERATURE

CHAPTER 3

The Identification Program

3.1 Introduction

The identification program is an implementation of a self-tuning control algorithm presented in a paper by K.J. Åström and B. Wittenmark [17]. The algorithm consists of a least-squares parameter identification of the plant, and the determination of a minimum variance control law from the identified parameters. The least-squares identification algorithm is presented at length in [15].

The plant model has the following form:

$$\begin{aligned} y(k) + a_1 y(k-1) + \dots + a_n y(k-n) \\ = b_1 u(k-1) + \dots + b_n u(k-n) + e(k) \end{aligned} \quad (3.1)$$

where $u(k)$ is the control variable, $y(k)$ is the controlled variable (output) and $e(k)$ is a noise variable. The minimum variance performance criterion is $V = E[y^2(k)]$ where E is the expectation operator and V is to be minimized. The optimal control law for this performance criterion is

$$u(k) = \frac{1}{b_1} [a_1 y(k) + \dots + a_n y(k-n+1) - b_2 u(k-1) - \dots - b_n u(k-n+1)].$$

Note that the poles of this controller's transfer function exactly cancel the zeros of the plant. If the model parameters a_i and b_i were known, the control law coefficients could immediately be entered to the control program as follows:

$$A_1 = -\frac{b_2}{b_1}, \quad A_2 = -\frac{b_3}{b_1}, \quad \dots, \quad B_0 = \frac{a_1}{b_1}, \quad B_1 = \frac{a_2}{b_1}, \quad \dots$$

3.2 Least-Squares Parameter Estimation

In order to obtain the parameters a_i and b_i , the problem is posed as an optimization. The object is to minimize the function $V = \sum_{k=n}^{N+n} e^2(k)$ where the $e(k)$ are interpreted here as equation errors in the plant model

(Equation 3.1). Hence the name least-squares estimation.

$$\text{Let } y = \begin{bmatrix} y(n+1) \\ \vdots \\ y(n+N) \end{bmatrix}, \quad e = \begin{bmatrix} e(n+1) \\ \vdots \\ e(n+N) \end{bmatrix}$$

$$\phi = \begin{bmatrix} -y(n) & \dots & -y(1) & u(n) & \dots & u(1) \\ \vdots & & \vdots & \vdots & & \vdots \\ -y(n+N-1) & \dots & -y(N) & u(n+N-1) & \dots & u(N) \end{bmatrix}$$

$$\beta^T = [a_1 \ a_2 \ \dots \ a_n \ b_1 \ b_2 \ \dots \ b_n]$$

where y , e , and ϕ are simply a history of N consecutive sample intervals. Then $e = y - \phi\beta$ and V is the magnitude squared of vector e . The minimum of V is obtained by setting

$$\nabla V = 0, \text{ and the solution, if } \phi^T \phi \text{ is not singular, is}$$

$$\beta = [\phi^T \phi]^{-1} \phi^T y \quad (\text{see [15]}) \quad (3.2)$$

As $N \rightarrow \infty$ the covariance of β is given by $\sigma^2 [\phi^T \phi]^{-1}$ where σ^2 is the variance of $e(k)$.

The determination of β from Equation 3.2 is a one-shot computation on an input-output history of N samples. This does not lead to a suitable computer program because excessive computation would be required. In [15] a

recursion formula derived from Equation 3.2 is given.

Using this formula, the current parameter estimate β is updated at each sample interval. The recursion formula is the following:

$$\beta(N+1) = \beta(N) + \Gamma(N)[y(N+1) - \phi(N+1)\beta(N)] \quad (3.3)$$

where $\Gamma(N) = P(N)\phi^T(N+1)[\alpha + \phi(N+1)P(N)\phi^T(N+1)]^{-1}$.

$$P(N+1) = [I - \Gamma(N)\phi(N+1)]P(N)$$

$$\phi(N+1) = [-y(N) \quad -y(N-1) \quad \dots \quad -y(N-n+1) \quad u(N) \quad u(N-1) \quad \dots \quad u(N-n+1)]$$

$$P(N_0) = \alpha[\phi^T(N_0)\phi(N_0)]^{-1}$$

The vector Γ and the square matrix P are both of order $2n$.

The matrix P is the covariance of β if the $e(k)$ are independent with variance equal to α . If the estimate were perfect and

there were no disturbance, then $\phi(N+1)\beta(N)$ would be the value

of $y(N+1)$. Thus Equation 3.3 means that at each sample

interval β is updated by a quantity which is essentially the

product of a time varying gain $\Gamma(N)$ and the difference between

the actual output and the expected output. In order to start,

the recursion, β and P can be initialized using a priori

knowledge, gained possibly from a one-shot least-squares first

estimate. Alternatively $P(0)$ can be initialized simply to a

constant times the identity matrix, which is the course taken

in this experiment.

In brief then, at each control interval the algorithm updates the estimate of the parameters in the model:

$$y(k) + a_1 y(k-1) + \dots + a_n y(k-n) = b_1 u(k-1) + \dots + b_n u(k-n)$$

Then a minimum variance control law is defined from the parameters:

$$u(k) = \frac{1}{b_1} [a_1 y(k) + \dots + a_n y(k-n+1) - b_2 u(k-1) - \dots - b_n u(k-n+1)]$$

3.3 Convergence Properties of the Closed Loop

It can be shown that, if the plant is stable, and the $e(k)$ are independent with zero mean, and the input $u(k)$ is independent of $e(k)$ and persistently exciting of order n , then the least-squares estimate of β converges to the correct value as $N \rightarrow \infty$ [15]. It is to be noted that this convergence is that of an open loop identification, that is the input is generated by an external source and is uncorrelated with $e(k)$. No general convergence theorems have yet been proved for the closed loop, when the input is generated by feedback.

In this experiment, the $e(k)$ are not independent. To illustrate, consider a first order noise-free system: $x(k+1) + ax(k) = bu(k)$. Let the output be a measurement of $x(k)$ corrupted by additive white noise $y(k) = x(k) + n(k)$ (which is essentially the case in this experiment). Then the system seen by the identification algorithm would be

$$y(k+1) + ay(k) = bu(k) + n(k) + an(k-1). \quad [15]$$

The noise here is not white, but rather a filtered white noise where the filter is the plant. Thus the $e(k)$ in the plant model (Equation 3.1) is not an uncorrelated noise

variable. The parameter estimates may still converge, but the estimates will be biased. The estimate $\hat{\beta}$ of β will be biased by an amount given by

$$E(\hat{\beta} - \beta) = [E(\Phi^T \Phi)]^{-1} E(\Phi^T e) \quad [15]$$

where E is the expectation operator. A theorem presented in [17] states that if the parameter estimates converge in closed loop, the resulting controller converges to the minimum variance control. This result holds whether the noise is correlated or not. Thus even if the parameter estimates are biased because of correlated noise, the result will still be a minimum variance controller.

Another theorem in the same paper states that if ~~the parameter estimates converge~~, certain correlations of the input and output are zero. Specifically, if $e(k)$ is a noise function of order m , then

$$E[y(t+\tau)y(t)] = 0 \quad \tau = m+1, \dots, m+n$$

$$E[y(t+\tau)u(t)] = 0 \quad \tau = m+1, \dots, m+n$$

A second convergence difficulty is that since the input is generated by feedback, the input is correlated with the noise. Consider again the first order system $y(k+1) + ay(k) = bu(k)$. The minimum variance control is $u(k) = \frac{a}{b}y(k)$. If the estimates of the parameters a and b are in error by a common multiplicative factor, the resulting control is unchanged and the minimum variance performance criterion $V = E[y^2(k)]$ is unchanged. Furthermore, it can be

shown that the least-squares objective function $V = \sum_{k=n}^{N+n} e^2(k)$ will be identical for all a and b with the same ratio a/b [17]. Thus, in general, when the input is generated by feedback, the identification algorithm will determine the parameters only to within a common multiplicative factor. The resulting minimum variance control law will be unique, however, since the multiplicative factor cancels upon division by b_1 .

In the simulation studies done by Åström and Wittenmark, the value of b_1 was fixed. This serves to prevent the algorithm from making large excursions along the linear manifold on which the parameter estimates lie. Åström and Wittenmark report in their paper that the choice of b_1 is not crucial in obtaining convergence of the algorithm. In the experimental work done by the author, the value of b_1 was not fixed.

To summarize, because of correlated noise the parameter estimates will be biased but will still lead to a minimum variance control. Furthermore, because of feedback, the biased estimates may be out by a multiplicative factor and convergence is not guaranteed. If the parameter estimates converge, however, a unique minimum variance control law can be expected.

3.4 The Identification Program

The identification program was written to overlay the control program so that at each sample interval the parameter estimates are updated and the minimum variance control law is determined according to the algorithm just presented. After this tuning step, the control program computes the control variable. The parameter estimates are punched on paper tape at every sample interval, if so requested by the operator.

The identification program performs a fourth order identification and control. Orders greater than four cannot be handled by this program because of insufficient memory in the PDP-8. Additional overlays were written to generate lower order identification and control by changing the order of the matrix and vectors involved.

The computation time for fourth order is about four seconds. Since there are eight parameters, the vectors and the square matrix P are of order 8. Moreover, the computation is hampered by large overheads in the continuous use of the A/D for sample interval measurement, and in the use of software for floating-point arithmetic.

At the start, before asking the operator for instructions, the program sets all coefficients, variables and vectors to zero except for β . The parameter h_1 is

initialized to unity while the rest are zero. (The starting value of b_1 can be changed however via the switch register on the PDP-8 console.) The value of b_1 must not be zero since it appears as a divisor in the determination of the minimum variance control law. The matrix P is initialized to b_1 times the identity matrix. The value of α is set (via switch register) to the noise variance obtained from the autocorrelation study discussed in §2.5.

In §2.5 the estimates of two time constants have been obtained as 2.5 minutes and .57 minutes, where the shorter time constant dominates. In order to obtain reasonable values for the parameter estimates, the sample interval T was chosen to be .7 minutes (42 seconds). Reasonable values are such that there is a good pole-zero spread on the z -plane. If T were relatively small the plant would appear essentially as an integrator and the poles would tend towards +1 on the z -plane. In the extreme, the order of the system would decrease as seen by the identifier and the excess poles and zeros may end up anywhere. If T were very large, the plant would look like a delay and the poles would tend towards the origin and again the order of the system would decrease as seen by the identifier.

The identification program has one additional task, that of integrating the control signal $u(k)$. The reason for this is that in a noise-free steady-state, $y_1(k)$ (output

temperature) is a constant, $y_1(k)$ (output error) is zero, and $u_1(k)$ (the control signal) is zero. The steam flow set-point signal $U_1(k)$ however must be a non-zero constant. Thus $u_1(k)$ is not equal to $U_1(k)$ but rather is an additive correction term for $U_1(k)$. Hence $u_1(k)$ is integrated to obtain $U_1(k)$. The resulting overall flow-chart and system diagram are shown in Figures 3.1 and 3.2. In Figure 3.2, z_1 is the temperature set-point defined by teletype. v_1 is the output temperature and y_1 is the output error. The u_1 signal is the control variable which is integrated to produce U_1 , which is the steam flow set-point.

3.5 Discrete-Time Plant Model

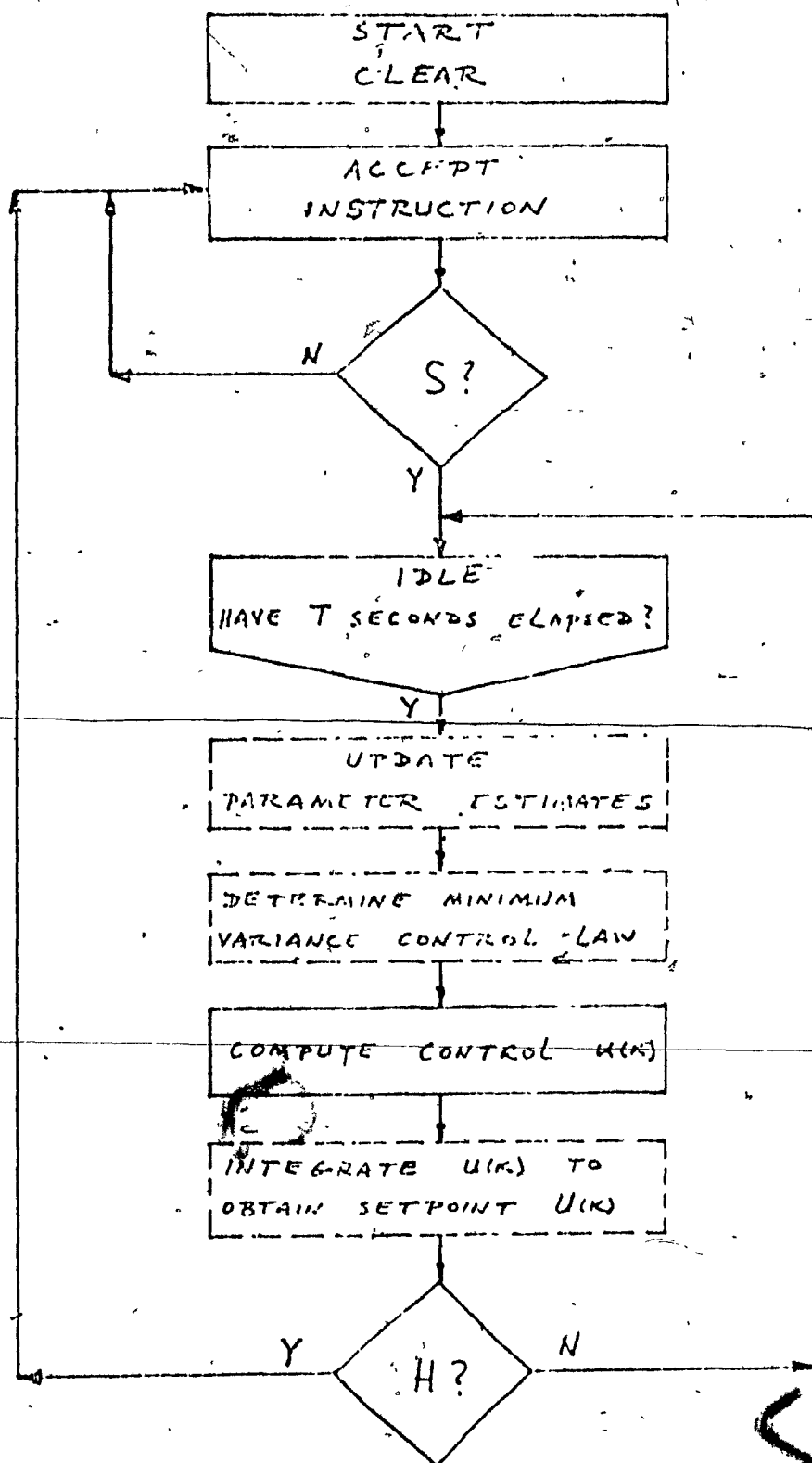
A discrete-time model of the plant was determined from the step response analysis for purposes of comparison with the least-squares identification. As far as the identification algorithm is concerned, the digital integrator in the identification program is part of the plant and is included in the model developed here.

From §2.5 the normalized step response of the heat exchanger (ignoring the delay) is

$$c(t) = 1 - .308e^{-.4t} - .692e^{-1.75t},$$

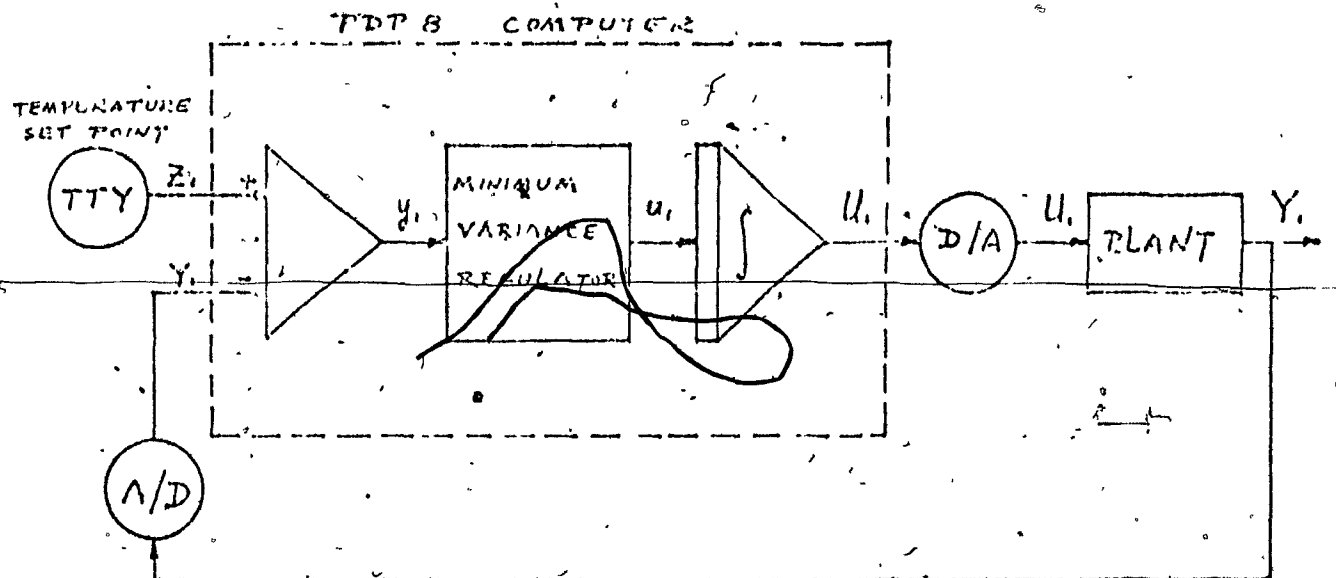
where the unit of time is the minute. For T_x , .7 minute

FIGURE 3.1



IDENTIFICATION AND CONTROL SYSTEM FLOWCHART

FIGURE 3.2



SELF-TUNING CONTROL SYSTEM
BLOCK DIAGRAM

the z-transform of $c(t)$ is

$$C(z) = \frac{z}{z-1} - \frac{.308z}{z-.756} - \frac{.692z}{z-.294}$$

$$= \frac{.563z(z-.695)}{(z-1)(z-.756)(z-.294)}$$

Since $C(z)$ is a step response, the transfer function is given by $C(z) \left(\frac{z}{z-1}\right)^{-1}$ where $\frac{z}{z-1}$ is the z-transform of the

integration operator. However, because of the digital integrator in series with the heat exchanger, the complete plant transfer function for unity gain is

$$C(z) \left(\frac{z}{z-1}\right)^{-1} \left(\frac{z}{z-1}\right) = C(z)$$

The gain depends on the operating region. From the discussion in §2.5, the analysis above is valid only in a mid-temperature region, where the current gain (from Figure 2.5) is approximately .325. That is, a 1 ma change in set-point produces a .325 ma change in temperature. Since the computer operates on voltages through the A/D and D/A, it is the voltage gain that is required. A .325 ma temperature change corresponds to a .065 volt change. A 1 ma set-point change requires a .250 volt change at the buffer input. Thus the voltage gain is .26 and the overall transfer function is $H(z) = .26C(z)$. When multiplied out, this gives

$$H(z) = \frac{.146z^2 - .102z}{z^3 - 2.05z^2 + 1.272z - .222}$$

In canonical form then, the model is

$$\begin{aligned} y(k) &= 2.05y(k-1) + 1.272y(k-2) - .222y(k-3) \\ &= .146u(k-1) - .102u(k-2) \end{aligned}$$

This is a third order system. The delay (including computation time) is not large relative to $T = 42$ seconds and has only a minor effect on the order of the zero at the origin. Hence it is ignored for this model.

3.6 Program Test

The complete program (identification and control) was tested on a plant consisting of a simple lag set up on a TR-48 analog computer. Nothing more was recorded than the fact that the fourth order identification and control produced fast convergence and good control. Following this test, the experiment proceeded to the heat exchanger. As a result of some convergence difficulties on the heat exchanger, however, further testing of the program using the analog computer was subsequently performed. The results of these tests are presented in Chapter 4.

CHAPTER 4

Experimental Results

4.1 Preliminary

The aim of the experiments was to determine how the closed loop identification and control performs under various conditions. The conditions that were varied are the order of the model, the sample interval, the starting value of parameter b_1 , and the value of α (the noise variance estimate). Only the first condition, the order of the model, was exhaustively varied, the other conditions being changed generally only to try to obtain convergence in "difficult" cases. The principal output of each experiment consists of answers to some yes-or-no questions: Do the parameter estimates converge? Is the closed loop system stable?

Convergence of the parameter estimates requires that the control $u(k)$ be persistently exciting. If the system is allowed to remain in a steady-state where $y(k)$ and $u(k)$ are small or zero, not much identification takes place. Hence the system was kept moving by means of set-point changes introduced by the operator via teletype.

A reduction of noise is not expected from the controller. If an accurate model were obtained from the identification, the minimum variance controller would make

$y(k) = e(k)$ in steady-state. Hence the variance of y will not be less than the noise variance.

Ideally, the plant should be identical in all the experiments. Throughout the experiments, the analog instrumentation settings were constant, and the water flow rate was a constant except for small fluctuations. However, the noise variance was not a constant. On days of cold weather, the additional heating load of the university degraded the quality of the steam available to the heat exchanger. The consequence is reduced noise and lower achievable temperatures. The dynamics did not change significantly. Thus a cold weather run is essentially a low-noise run.

Before proceeding to the experiments proper, some preliminary responses are presented here. Figures 4.1 and 4.2 show open loop step responses for $T = 42$ and $T = 14$ seconds. The upper curves are the output temperature voltage Y_1 . The lower curves are the steam flow set-point signal U_1 . The purpose of these records is to serve as a comparison with closed loop responses. In closed loop the noise will not be reduced, but the general response shape will be affected.

Another test was to obtain the response for a minimum variance control based on the step response model obtained in §3.5. The control program alone is required

since the parameters are given. Since the model has the form

$$y(k) + a_1 y(k-1) + a_2 y(k-2) + a_3 y(k-3) = b_1 u(k-1) + b_2 u(k-2),$$

the minimum variance control law is

$$u(k) = \frac{a_1}{b_1} y(k) + \frac{a_2}{b_1} y(k-1) + \frac{a_3}{b_1} y(k-2) - \frac{b_2}{b_1} u(k-1)$$

The steam flow set-point signal U_1 is the integral of the control variable: $U_1(k) = U_1(k-1) + u_1(k)$. Therefore by simple algebra,

$$\begin{aligned} U_1(k) = & \frac{a_1}{b_1} y_1(k) + \frac{a_2}{b_1} y_1(k-1) + \frac{a_3}{b_1} y_1(k-2) \\ & + \left(\frac{b_1 - b_2}{b_1} \right) U_1(k-1) + \left(\frac{b_2}{b_1} \right) U_1(k-2) \end{aligned}$$

Thus in the control law

$$B_0 = \frac{a_1}{b_1} = -14.0$$

$$B_1 = \frac{a_2}{b_1} = 8.72$$

$$B_2 = \frac{a_3}{b_1} = -1.52$$

$$A_1 = \frac{b_1 - b_2}{b_1} = 1.69$$

$$A_2 = \frac{b_2}{b_1} = -0.69$$

The resulting control is shown in Figure 4.3. Again, the upper curve is the output temperature y_1 . The heavy low-

frequency rectangular wave superimposed on the output is the set-point defined by the operator. The lower curve is U_1 .

It is evident that the control is not good. It appears that the gain is excessive. Assuming that b_1 and b_2 have been underestimated, their values were increased so as to produce a control law with lower gain. Figure 4.4 shows the result for gain reductions by factors of 10 and 2. The control became much less wild but the response slowed considerably at $1/10$ gain. In all cases, there were oscillation tendencies, with frequency and settling time decreasing with decreasing gain.

4.2 Fourth Order Model

The control algorithm using the fourth order model was successful in the sense that the parameter estimates did converge and the closed loop system was stable. Control histories (response and parameter convergence) for five particular runs are shown in Figures 4.5 to 4.9. Figure 4.10 illustrates the pole-zero locations of the transfer functions of the identified plant and of the corresponding minimum variance controller (the poles and zeros of the plant being on the left and those of the controller on the right). The diagram in Figure 4.10a corresponds to the step response model, obtained in §3.5.

Runs #1 and #2 are each typical of several runs. That is, several runs (using different values of α) more or less correctly identified the dominant higher frequency pole as in run #1 (see Figure 4.10b), while several other runs placed the pole near the lower frequency as in run #2 (see Figure 4.10c). In both cases, the pole at +1 (due to the digital integrator) was correctly identified. Also in both cases, a complex pole-pair almost cancels a complex zero-pair. This suggests that the algorithm sees the plant essentially as a simple lag in series with an integrator, and that a lower order identification may prove more successful. A further comparison of runs #1 and #2 is obtained by using only the control program and typing in the minimum variance control law based on the identified parameters. The responses shown in Figures 4.11 and 4.12 indicate that run #1, in which the estimated model is closer to the expected model, did in fact produce a better controller.

Run #3 is a low-noise run. The resulting model is similar to that of run #1 but the complex pole-zero pairs are further apart (see Figure 4.10d). Run #4 is also a low-noise run but with α set to a relatively high value of .64 volt².

Figure 4.10f is the pole-zero pattern for the step response model for $T = 14$ seconds. Note that because of the smaller sampling interval, the poles have moved closer to +1. Run #5 is for $T = 14$ seconds. Figure 4.10g indicates that the

two significant poles have been identified, but a zero appears outside the unit circle. Thus the identified plant is distinctly non-minimum phase. The transfer function of the corresponding controller is also non-minimum phase since the controller placed a pole directly on the non-minimum phase zero.

In all cases, the parameter estimates converged. The problem of the estimates being in error by an unknown multiplicative factor does not appear to be present, since meaningful pole-zero patterns are obtained from the estimated parameters directly. It is to be noted that in general, the parameter estimates are not to be accepted with confidence for the reasons discussed in §3.3.

4.3 Third Order Model

The third order model produced less success than came from the fourth order model. The system was not generally stable and the parameter estimates did not converge. Here, the problem of non-identifiability due to feedback became apparent. The parameter estimates made very large excursions which were not always on a linear path. The response became more sensitive to the value of α and very sensitive to the starting value of b_1 . For some values of b_1 an unstable system was produced. In these cases, incorrect parameter estimates led the system to saturate at minimum or

maximum temperature, whereupon no further identification took place.

Responses are shown in Figures 4.13 to 4.17.

Figure 4.18 illustrates the pole-zero patterns of the corresponding minimum variance controllers for stable intervals near the end of each run. The parameter estimates are definitely out by a large factor, hence no plant model and corresponding pole-zero pattern is obtained.

An attempt was made to obtain a PID controller by using the model

$$y(k) + a_1 y(k-1) + a_2 y(k-2) + a_3 y(k-3) = b_1 u(k-1) \quad (4.1)$$

The minimum variance controller is

$$u(k) = \frac{a_1}{b_1} y(k) + \frac{a_2}{b_1} y(k-1) + \frac{a_3}{b_1} y(k-2)$$

The steam flow set-point signal is the integral of $u(k)$:

$$U(k) = U(k-1) + \frac{a_1}{b_1} y(k) + \frac{a_2}{b_1} y(k-1) + \frac{a_3}{b_1} y(k-2)$$

The z-transform of the transfer function is

$$\frac{U(z)}{y(z)} = - \left(\frac{a_2}{b_1} + \frac{a_3}{b_1} \right) - \left(\frac{a_3}{b_1} \right) \frac{1}{z} + \left(\frac{a_1}{b_1} + \frac{a_2}{b_1} + \frac{a_3}{b_1} \right) \frac{z}{z-1} \quad (4.2)$$

where the three terms are proportional, derivative, and integral terms respectively.

The performance of the PID controller proved to be even worse and was even more sensitive to the starting value of b_1 . In run #9 (Figure 4.16) note that after 46 sample

intervals, incorrect parameter estimates produced a positive feedback and then a very low gain negative feedback, which then brought the system to the set-point very slowly. One explanation for the poor performance of the PID controller is that the model of Equation 4.1 places both the zeros at the origin. Hence the algorithm is not allowed to identify the zero that nearly cancels the non-dominant low frequency pole whose time constant is 2.5 minutes.

4.4 Second Order Model

The second order model also had convergence difficulties. As run #11 (Figure 4.19) shows, although the system generally followed the set-point, it went through periods of high and low gain, positive and negative feedback. Parameter a_1 refused to converge, while a_2 , b_1 , and b_2 either converged or were no longer identified towards the end of the run.

A PI controller was obtained by using the model of Equation 4.1 with $a_3 = 0$. Equation 4.2 then becomes

$$\frac{U(z)}{y(z)} = - \frac{a_2}{b_1} + \left(\frac{a_1}{b_1} + \frac{a_2}{b_1} \right) \frac{z}{z-1}$$

where the terms are proportional and integral terms. As in the third order case, the zero of the plant model transfer function is then fixed at the origin, which is all right here

since only two poles will be obtained. If the identification succeeds, these two poles would be the pole at +1 and the higher frequency pole corresponding to the dominant .57 minute time constant. Thus the pole-zero near-cancellation corresponding to the non-dominant pole (2.5 minutes time constant) need not even appear in the identified model, and it may be expected that this model would work. It did not work, however, as Figures 4.20 to 4.26 indicate. Parameter convergence was poor, and there were many cases of very low gain and/or positive feedback (see run #15, Figure 4.24).

The salient problem here is a recurrent situation of control signal saturation followed by a period of very low gain (see Figures 4.24 and 4.25). The control saturation is a consequence of excessive gain or positive feedback. If the system recovers, it emerges with a very low gain. The low gain is a result of an over-estimation of b_1 or an under-estimation of the other parameters. This is an over-reaction on the part of the identification algorithm to a previous under(over)-estimation which had led to the control saturation. The saturation/low-gain-recovery phenomenon appears to be a limit cycle of an unstable algorithm (for this model). The reason the phenomenon can be recurrent is that little is learned about the plant during the saturation and the low gain recovery. Since the $u(k)$ are small during this period, their effect is largely buried by noise. The

parameter identification during this time is then incorrect or very slow.

4.5 First Order Model

A first order model leads to an integral controller.

Setting $a_2 = a_3 = 0$, Equation 4.2 becomes
$$\frac{U(z)}{Y(z)} = \frac{a_1}{b_1} \frac{z}{z-1}$$

which is the transfer function of an integrator.

The system, using the first order model, was surprisingly more stable than that using the second and third order models. Run #18 (Figure 4.28) was a high-noise run. The parameters converged but yielded a rather high gain. Run #19 (Figure 4.29) was a low-noise run. The parameters also converged and the control was as good as any of the best in these experiments. For $T = 14$ seconds, however, convergence and stability were difficult to obtain.

4.6 Tests on a Simulated Plant

In view of the convergence difficulties apparent in the second and third order identifications, further testing was performed using the analog computer. The heat exchanger was simulated on the analog computer according to the step response model obtained in §2.5. The noise was not included in the simulation, since it

was hoped that something may be learned from the deterministic case that may have been previously hidden in the stochastic case. A primary purpose of these tests was to verify that the programming was error-free.

The behaviour of the program on the noise-free simulated system was about the same as that for the noisy real-life system: the first and fourth order identifications converged, and the second and third order identifications did not. The third order case had a strong tendency to saturate and therefore terminate the identification at a very early stage (even before any set-point change was made). The second order case was more stable but still produced poor control and divergent parameters. It was observed, however, that the divergent tendencies in the second order case occurred immediately at the set-point changes. That is, at each set-point change, the parameter values changed drastically but recovered at the next step to an only slightly different set of values. Eventually, however, recovery became less complete and the parameters began to diverge.

This behaviour suggests that the programming is not at fault, and that the difficulty may be eliminated by increasing the value of the noise parameter α . This proved to be the case. It was found that smooth consistent

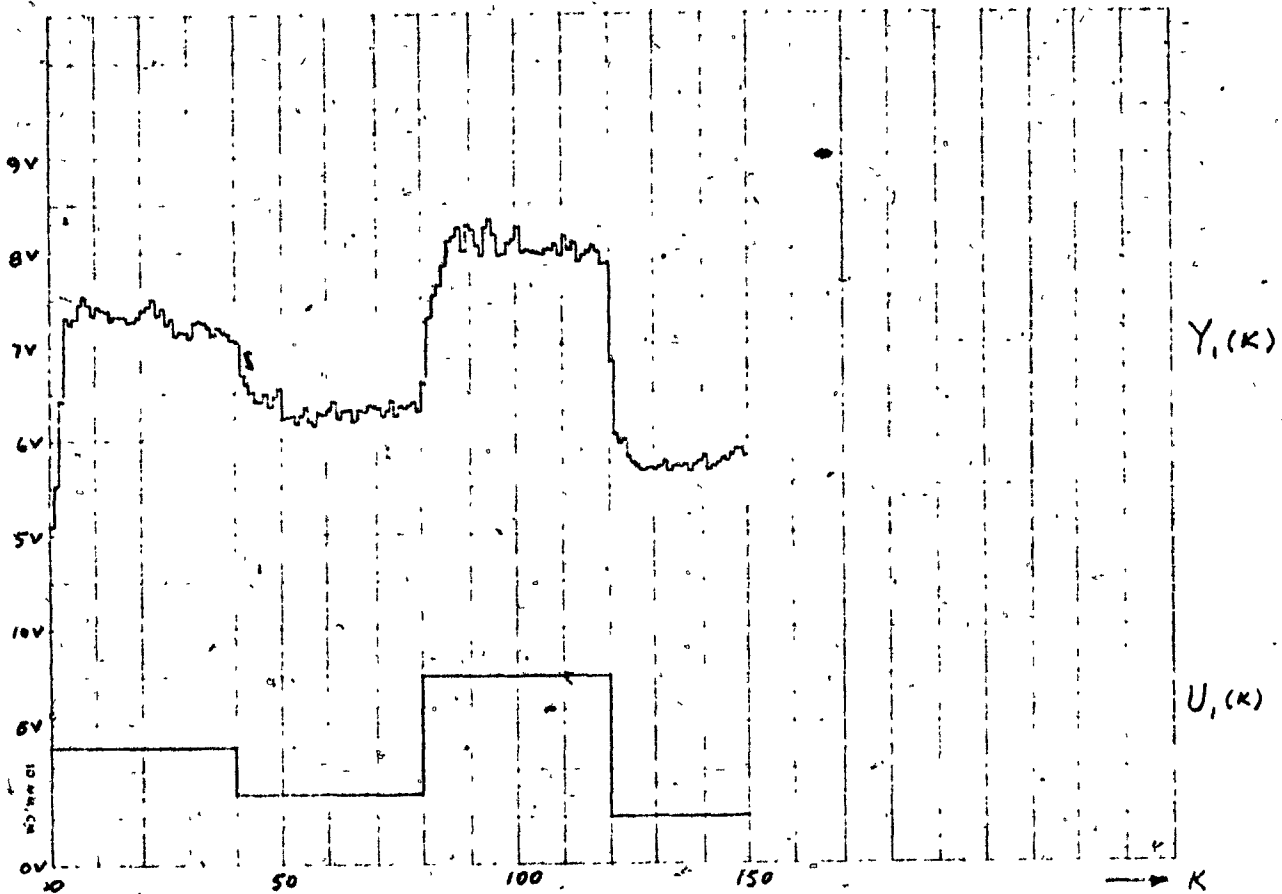
convergence could not be obtained in the second order case for α much less than about 6.5 volt^2 . Similarly, for the third order case, α had to be not less than about 13 volt^2 . Interestingly, in the third order case, the identification degraded much more rapidly with a reduction of α than in the second order case. In fact there appeared to be a rather sharp border between convergence and divergence, located approximately at $\alpha = 13 \text{ volt}^2$.

Decreasing α from this point leads to divergent identification. Increasing α , on the other hand, serves to slow down the identification. This can be expected when noting that α appears in the denominator of the expression for the gain factor Γ in Equation 3.3.

Models of first and fourth order produced no convergence difficulties. Consistent convergence in these cases was obtained for α as low as $2.5 \times 10^{-5} \text{ volt}^2$. In the fourth order case, this value of α led to convergence on the actual noisy heat exchanger as well as the noiseless simulation. Note that the actual heat exchanger had a noise variance of $.025 \text{ volt}^2$.

Results are shown in Figures 4.31 to 4.35. The success of these tests validates both the programming and the algorithm. However, it also raises the question of the significance of α . This question is taken up in the next chapter.

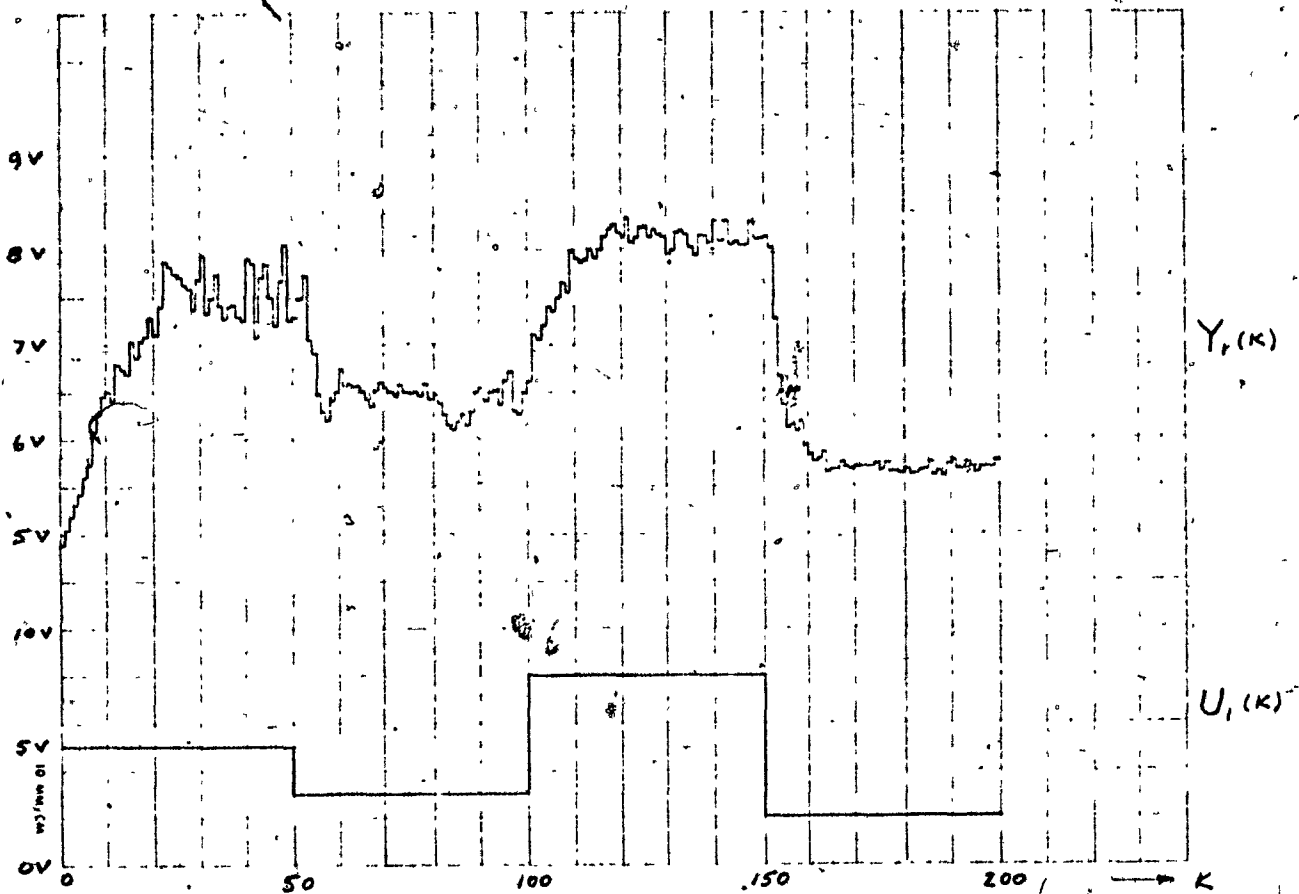
FIGURE 4.1



OPEN LOOP RESPONSE

 $T = 42$ seconds

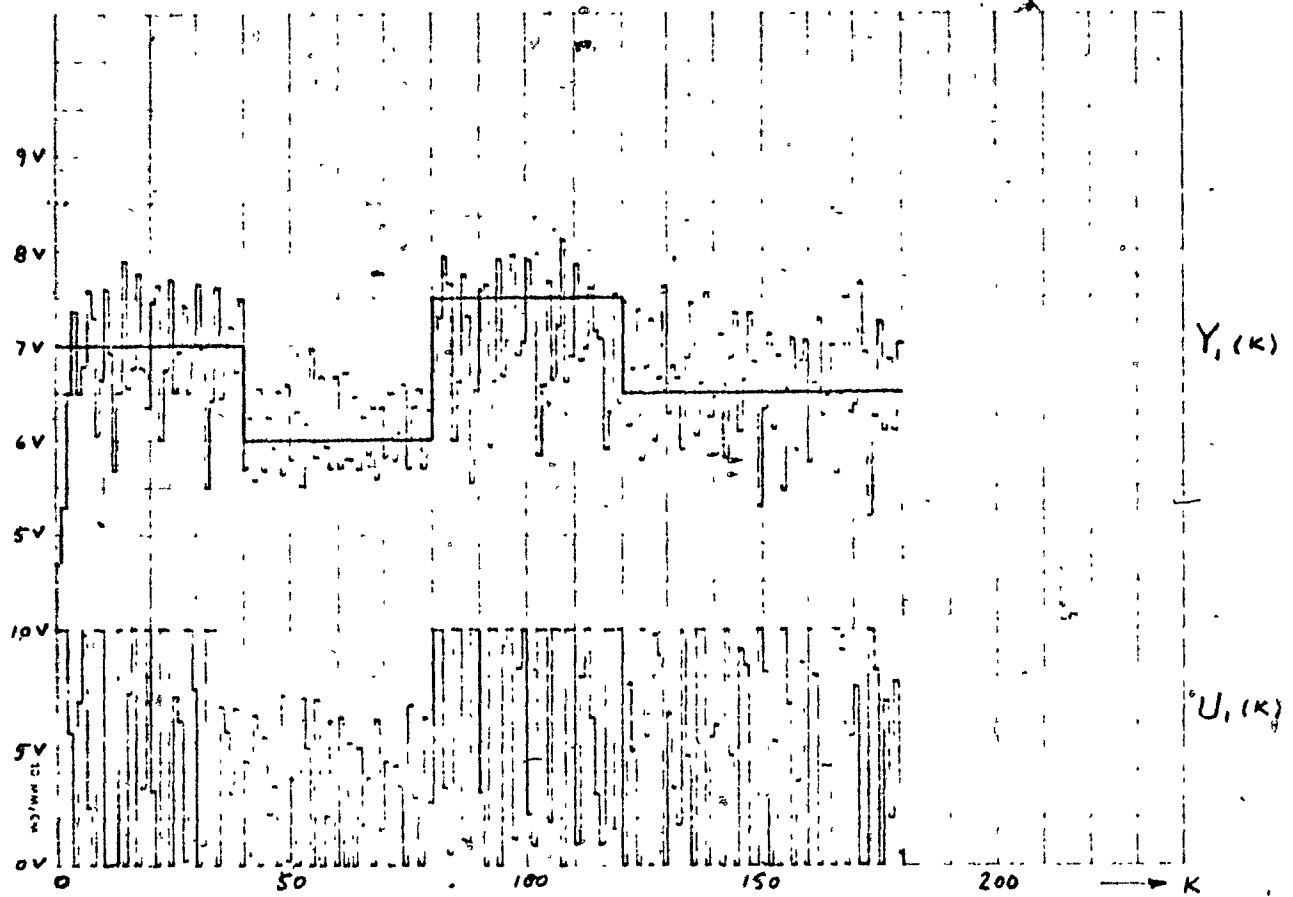
FIGURE 4.2



OPEN LOOP RESPONSE

 $T = 14$ seconds

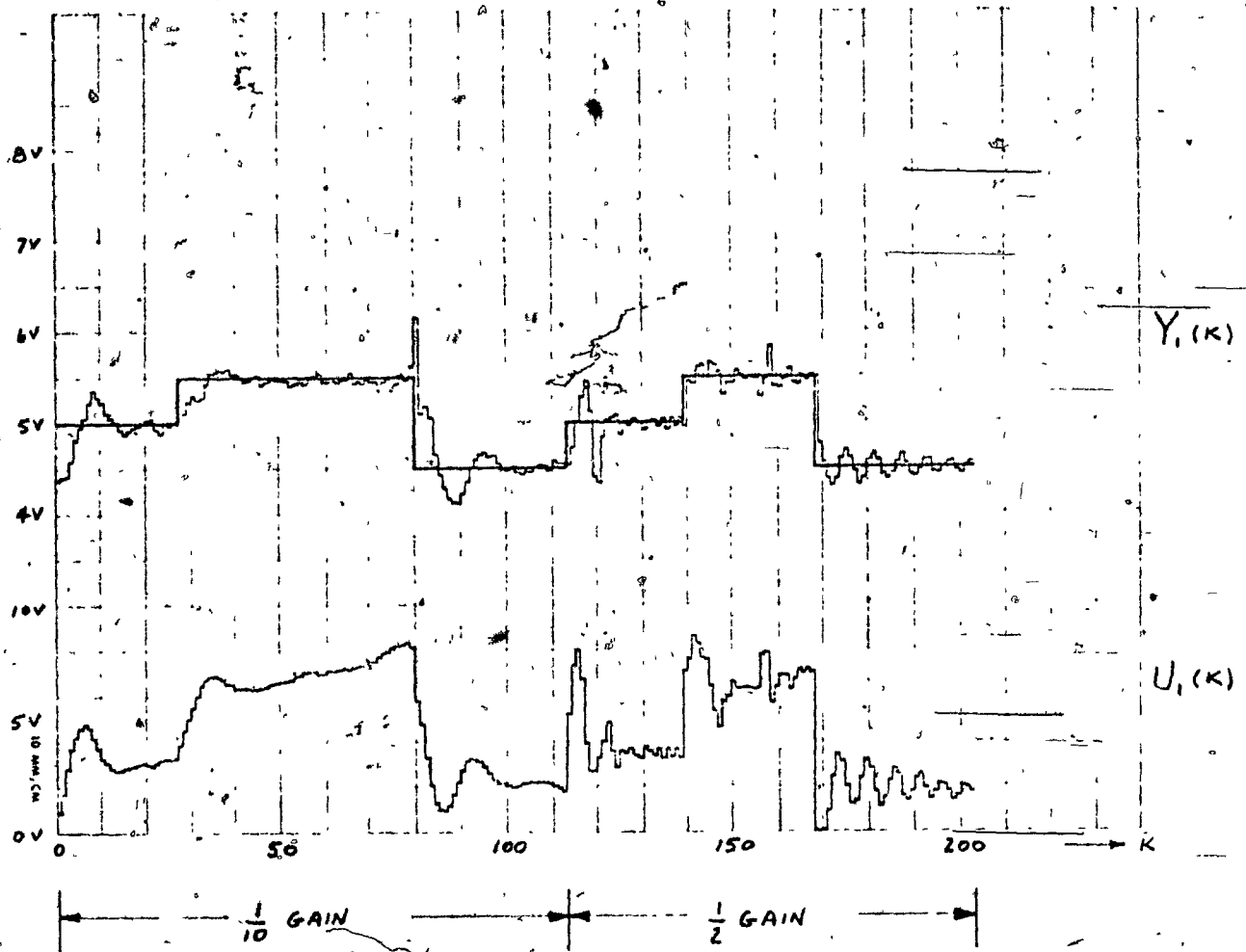
FIGURE 4.3



MINIMUM VARIANCE CONTROL
BASED ON STEP RESPONSE MODEL

$T = 42$ seconds

FIGURE 4.4

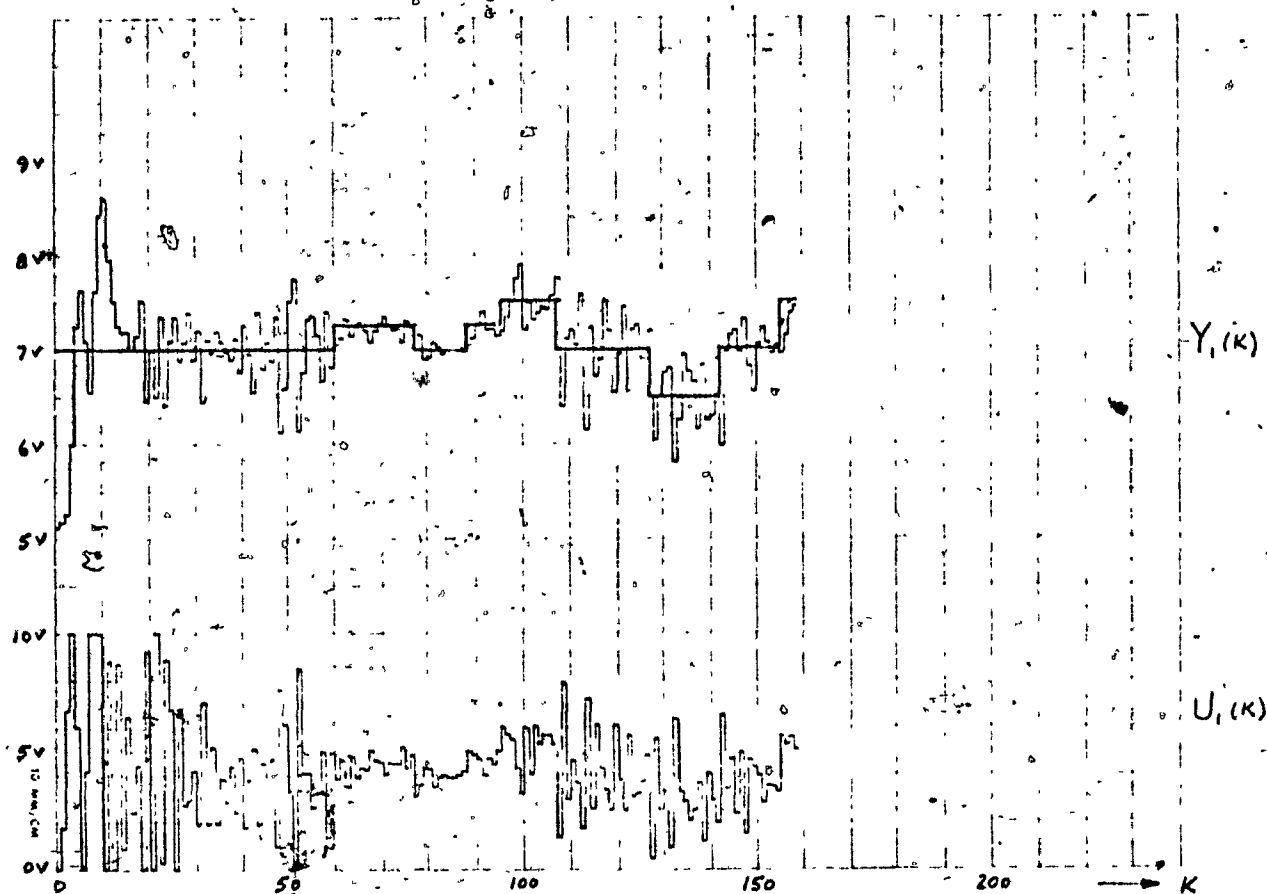


MINIMUM VARIANCE CONTROL

BASED ON STEP RESPONSE MODEL

 $T = 42$ seconds

FIGURE 4.5a



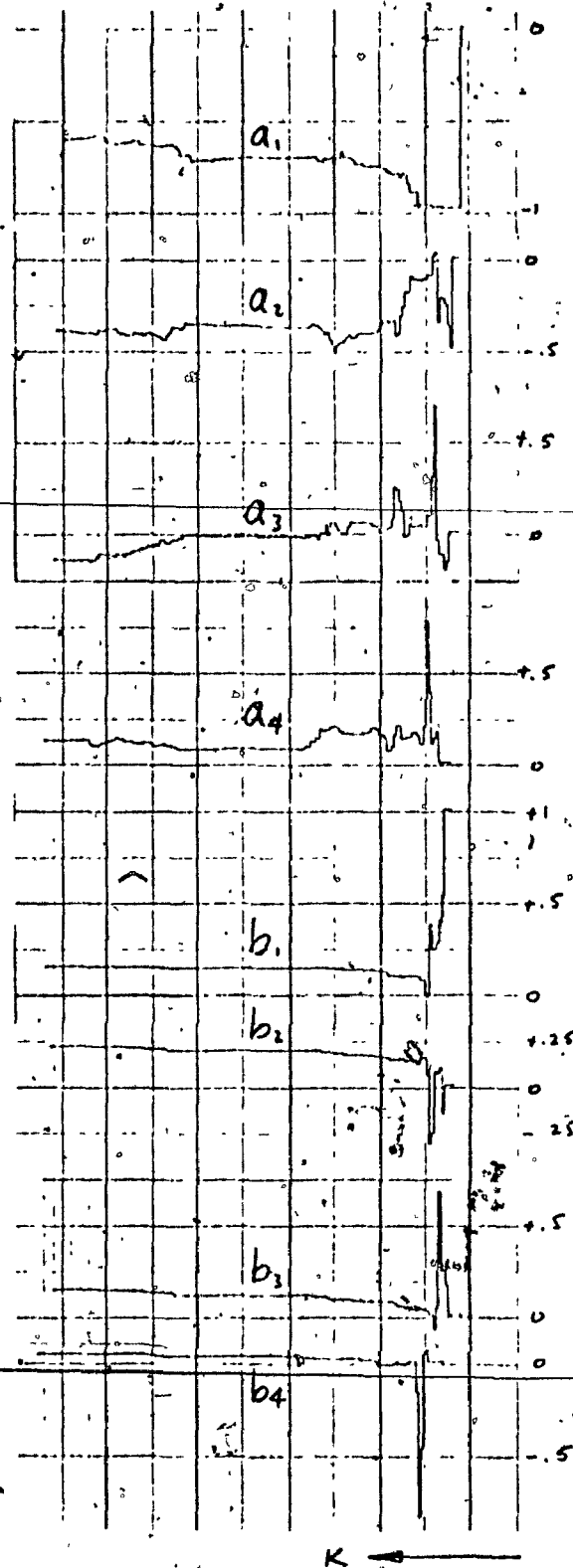
SELF-TUNING CONTROL

FOURTH ORDER

PUN # 1

 $T = 42$ seconds $b_1(0) = 1$

FIGURE 4.5b

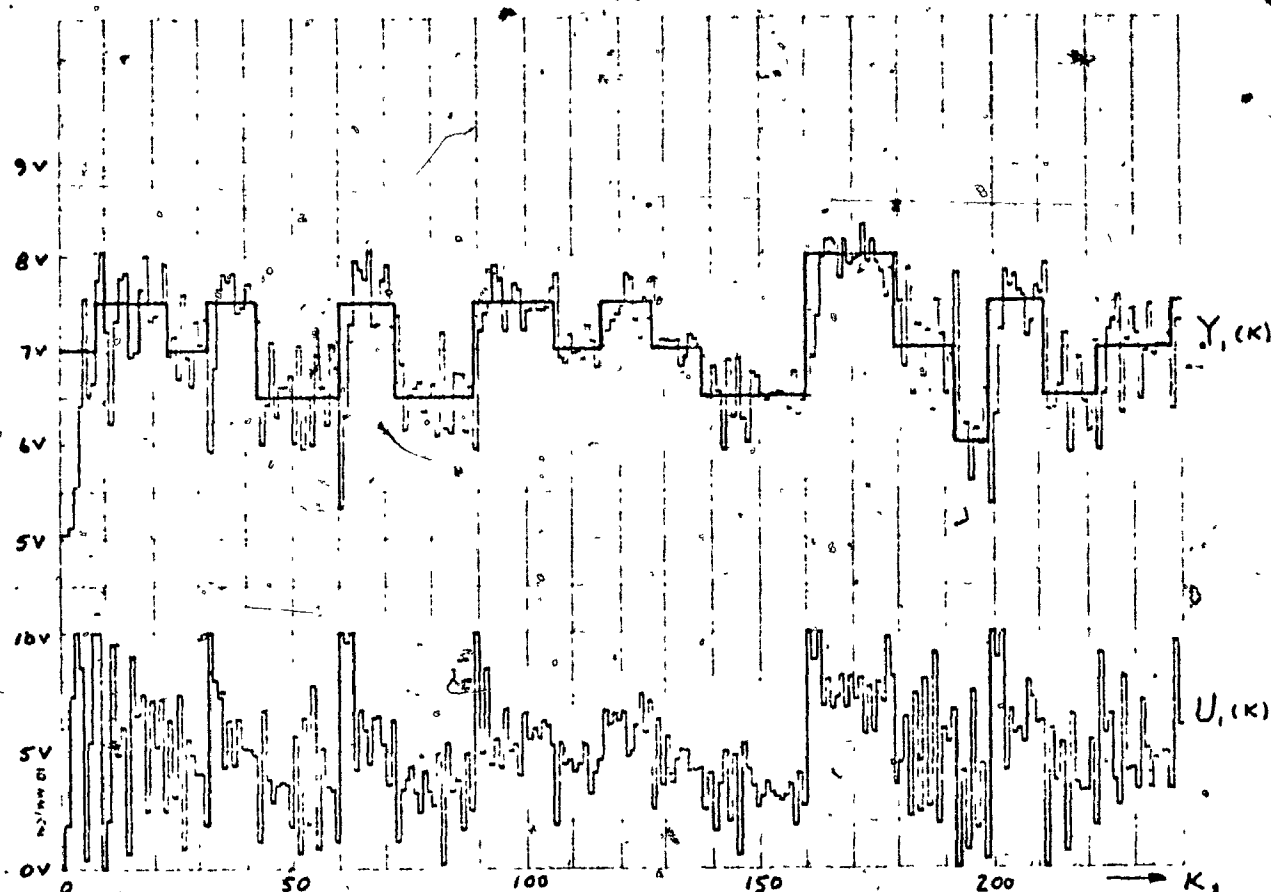


PARAMETER ESTIMATES

FOURTH ORDER

RUN # 1

FIGURE 4.6a



SELF-TUNING CONTROL

FOURTH ORDER

RUN # 2

 $T = 42$ seconds $b_1(0) = 1$

FIGURE 4.6a (Continued)

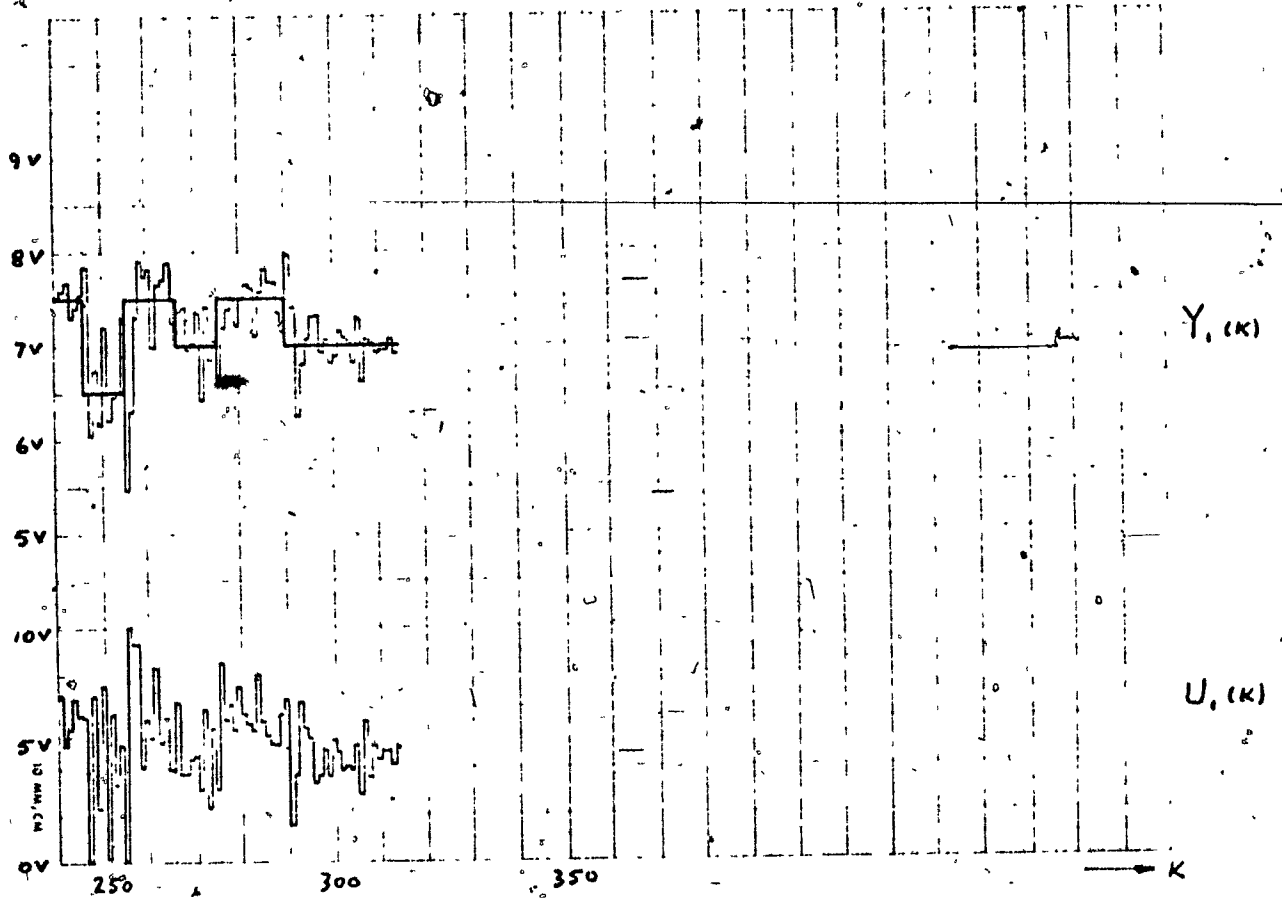
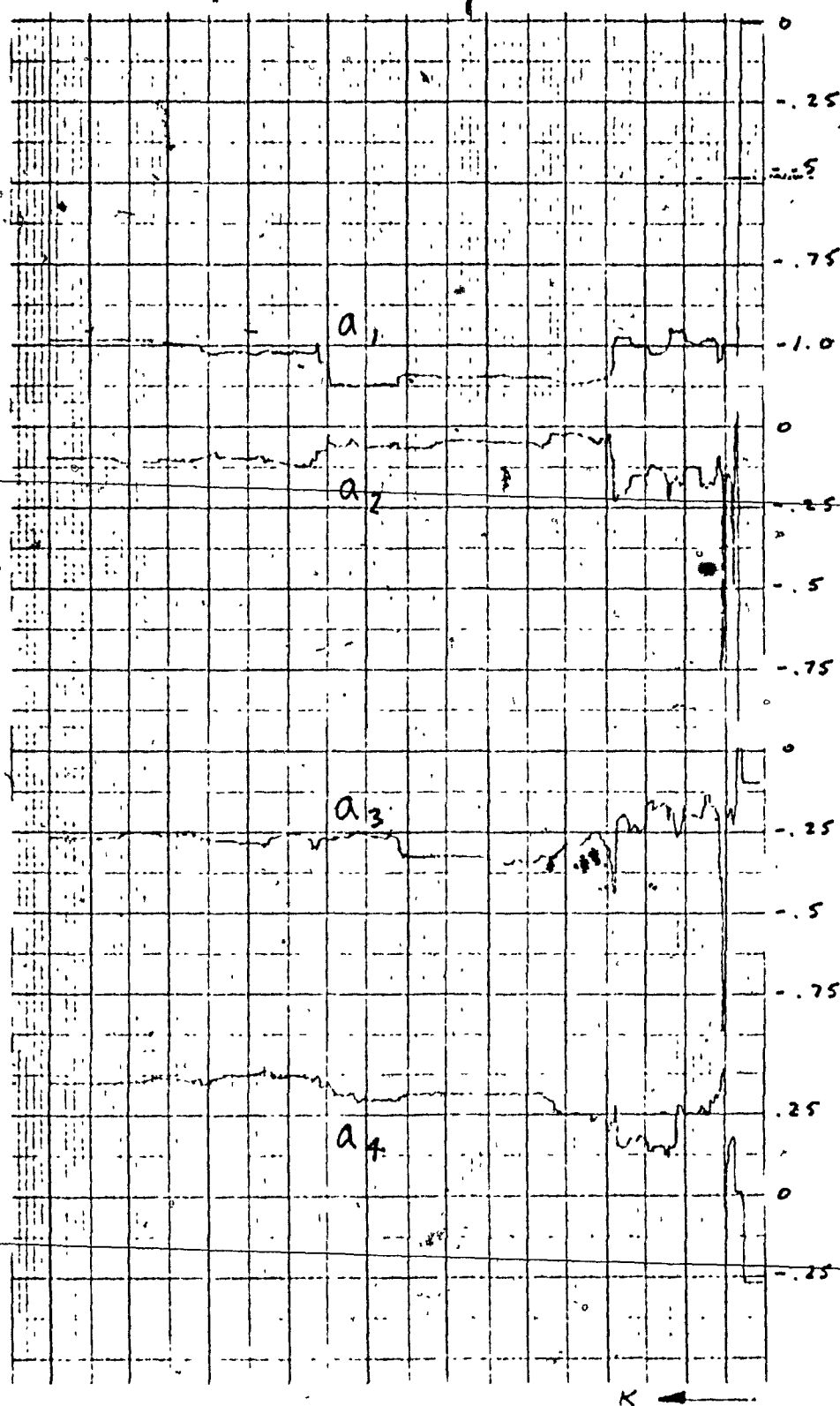


FIGURE 4.6b.



PARAMETER ESTIMATES

FOURTH ORDER

RUN # 2

FIGURE 4.61 (Continued)

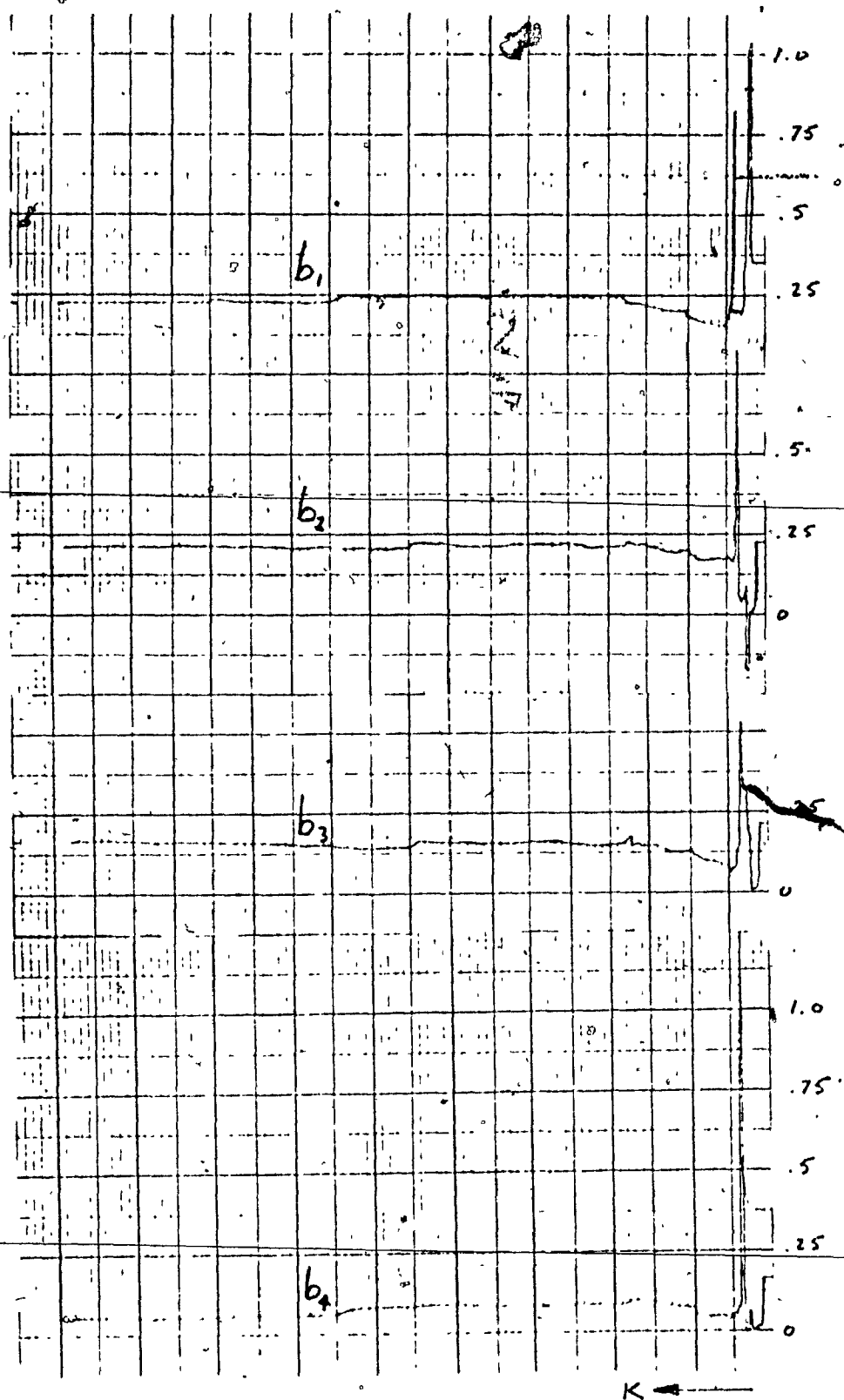
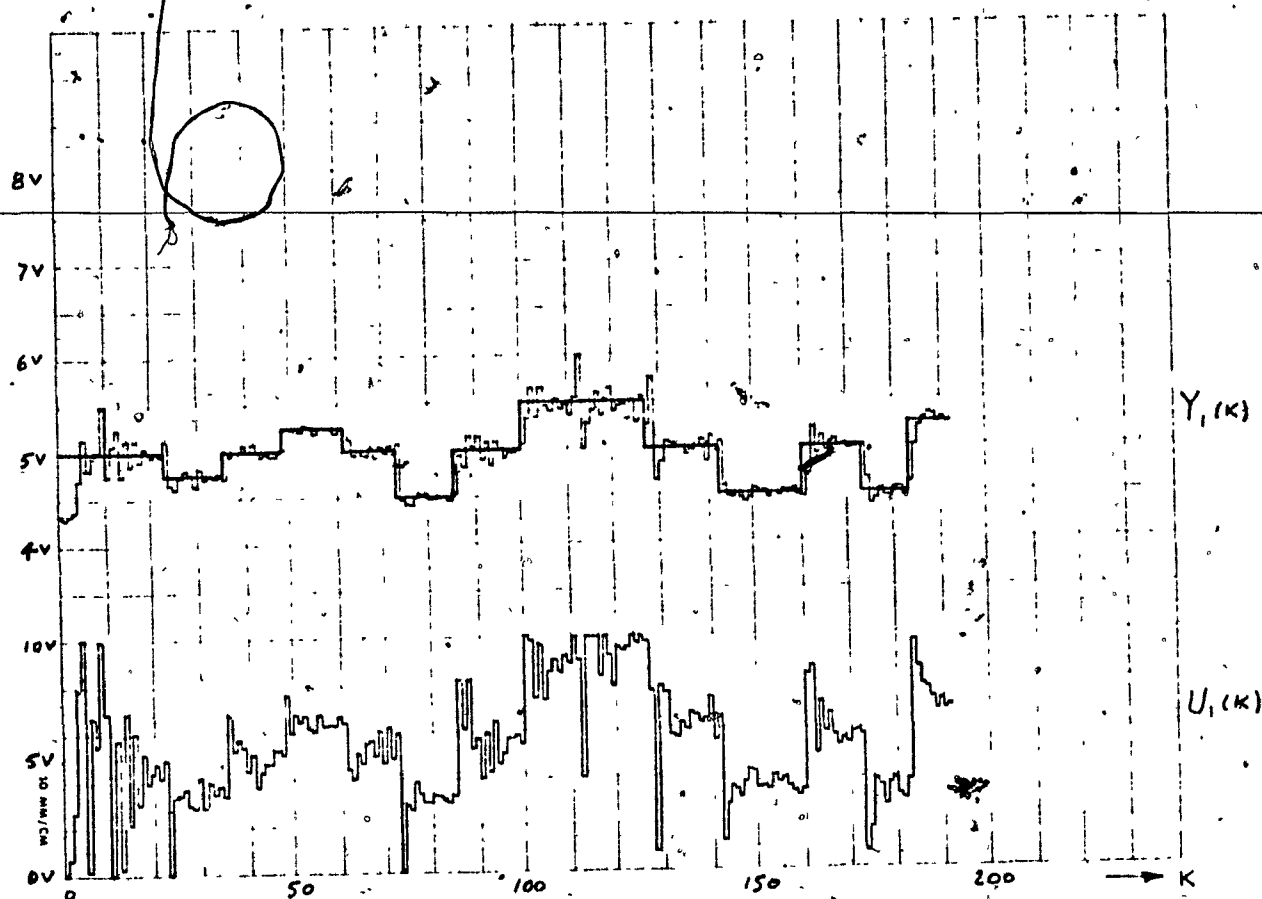


FIGURE 4.7a



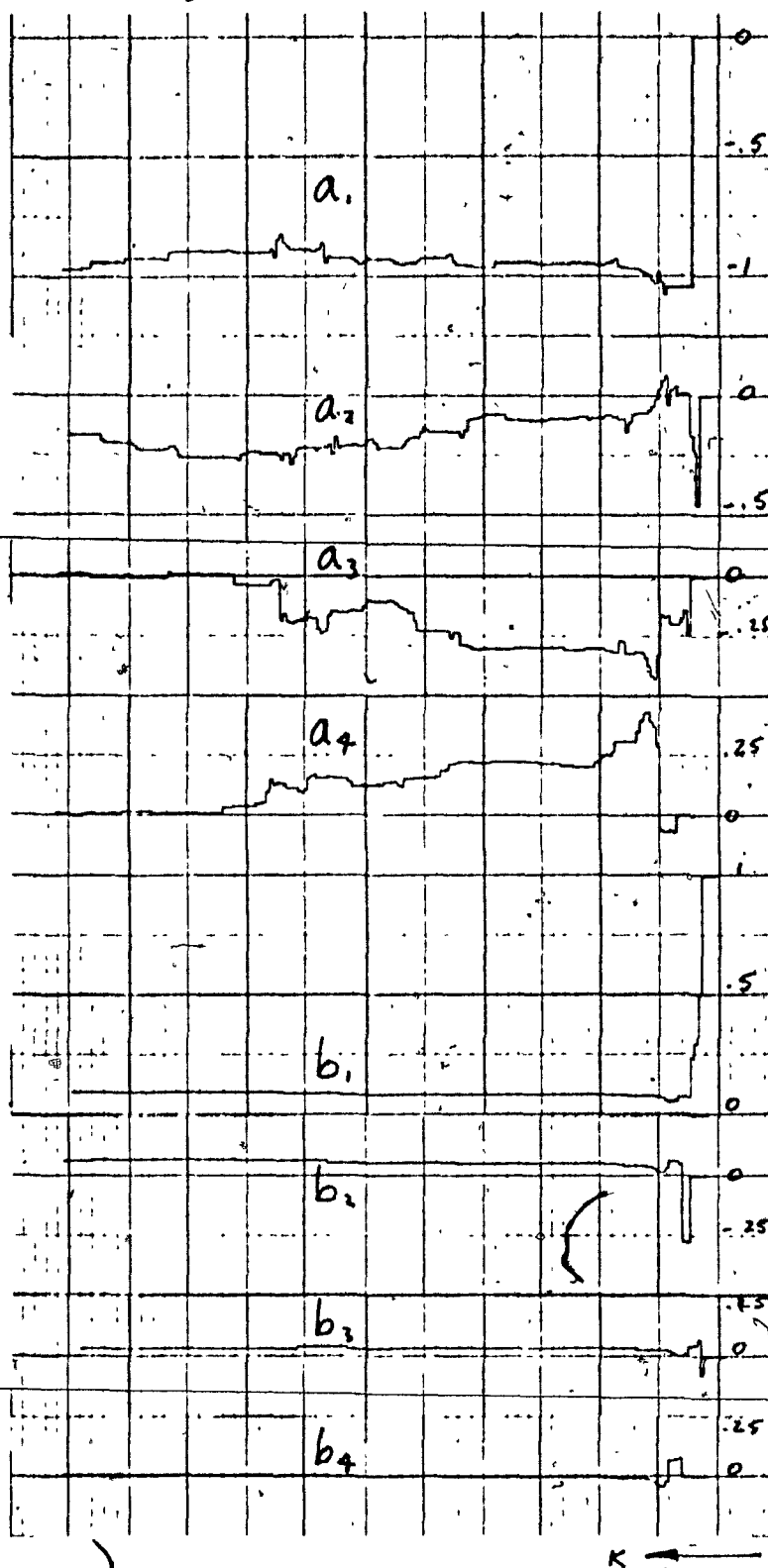
SELF-TUNING CONTROL

FOURTH ORDER

RUN # 3

 $T = 42 \text{ seconds}$ $b_1(0) = 1$

FIGURE 4.7b

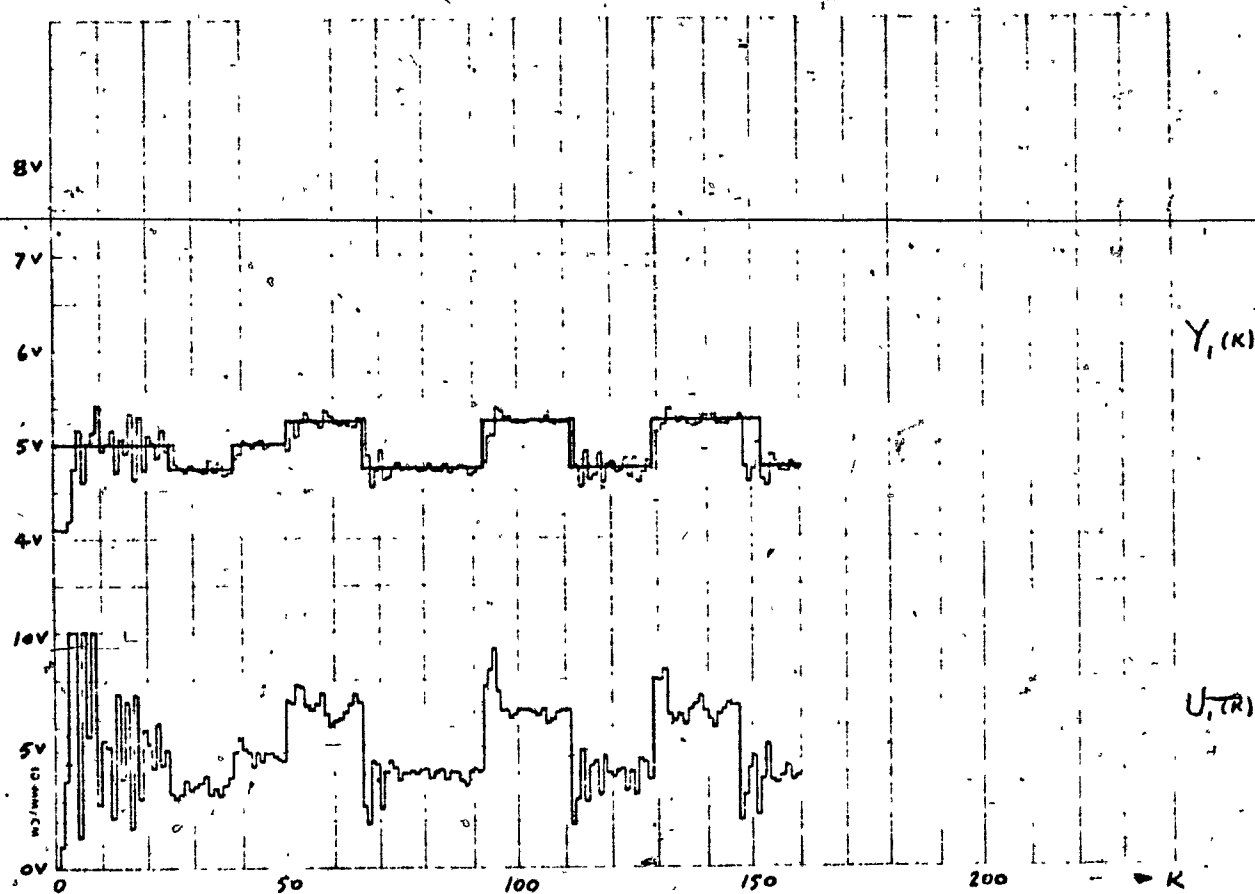


PARAMETER ESTIMATES

FOURTH ORDER

RUN # 3

FIGURE 4.8a



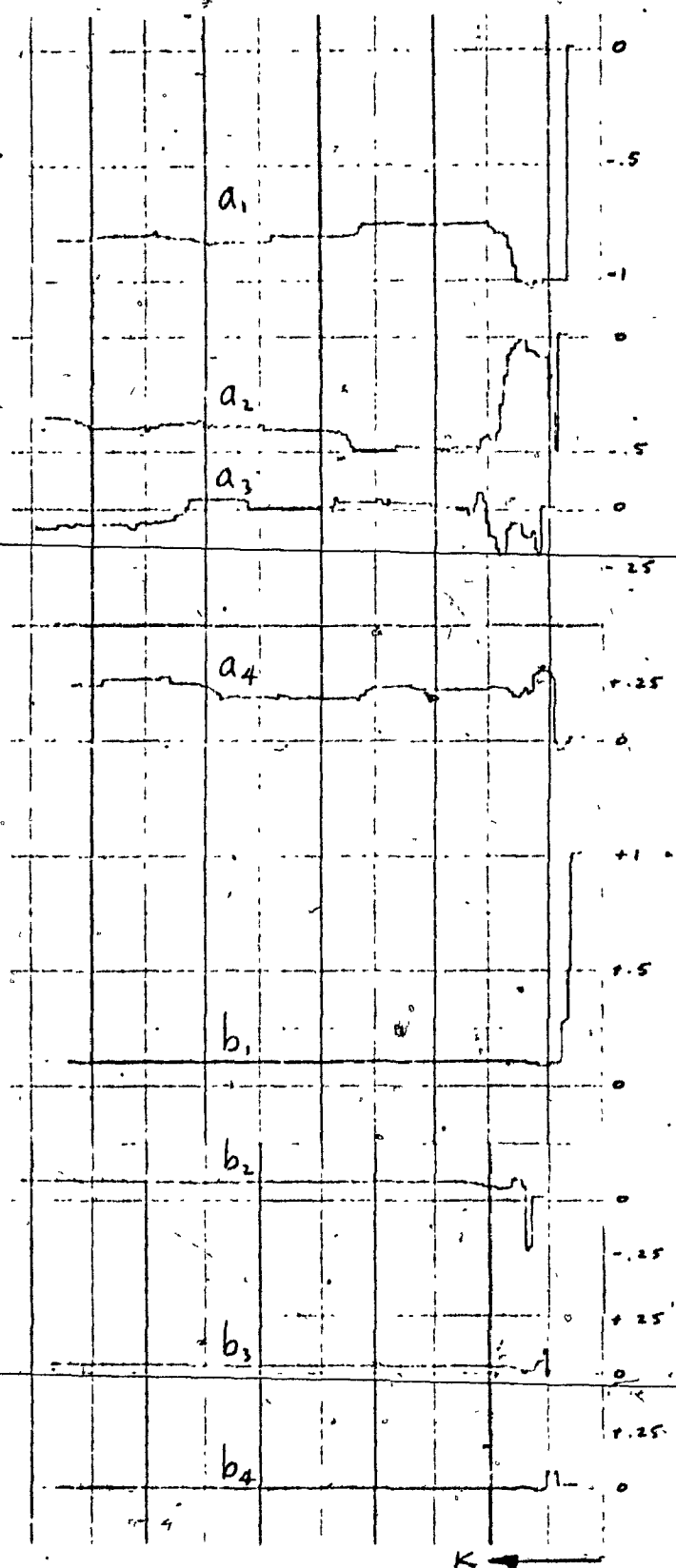
SELF-TUNING CONTROL

FOURTH ORDER

RUN # 4

 $T = 42$ seconds $b_1(0) = 1$ $\alpha = .64 \text{ volts}^2$

FIGURE 4.8b

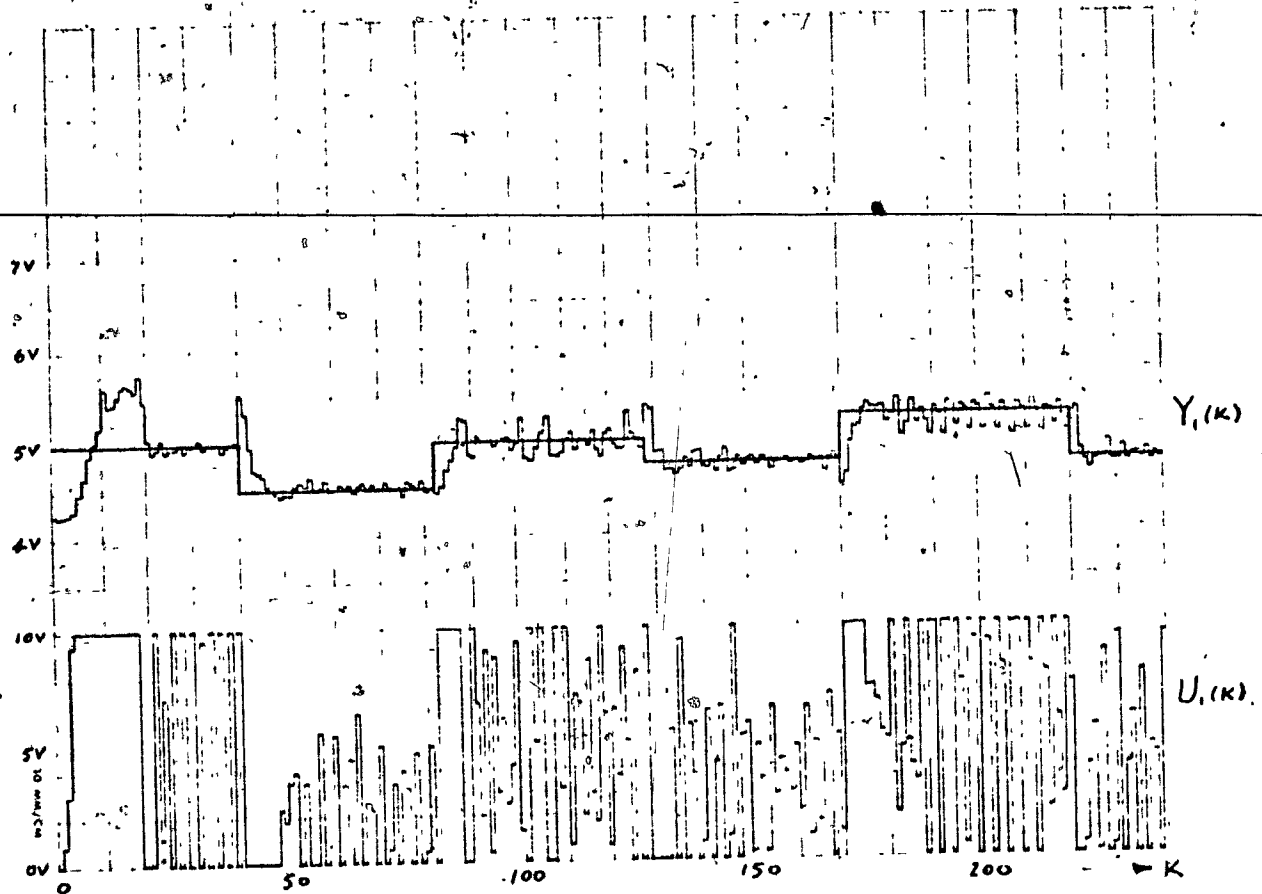


PARAMETER ESTIMATES

FOURTH ORDER

RUN # 4

FIGURE 4.9a



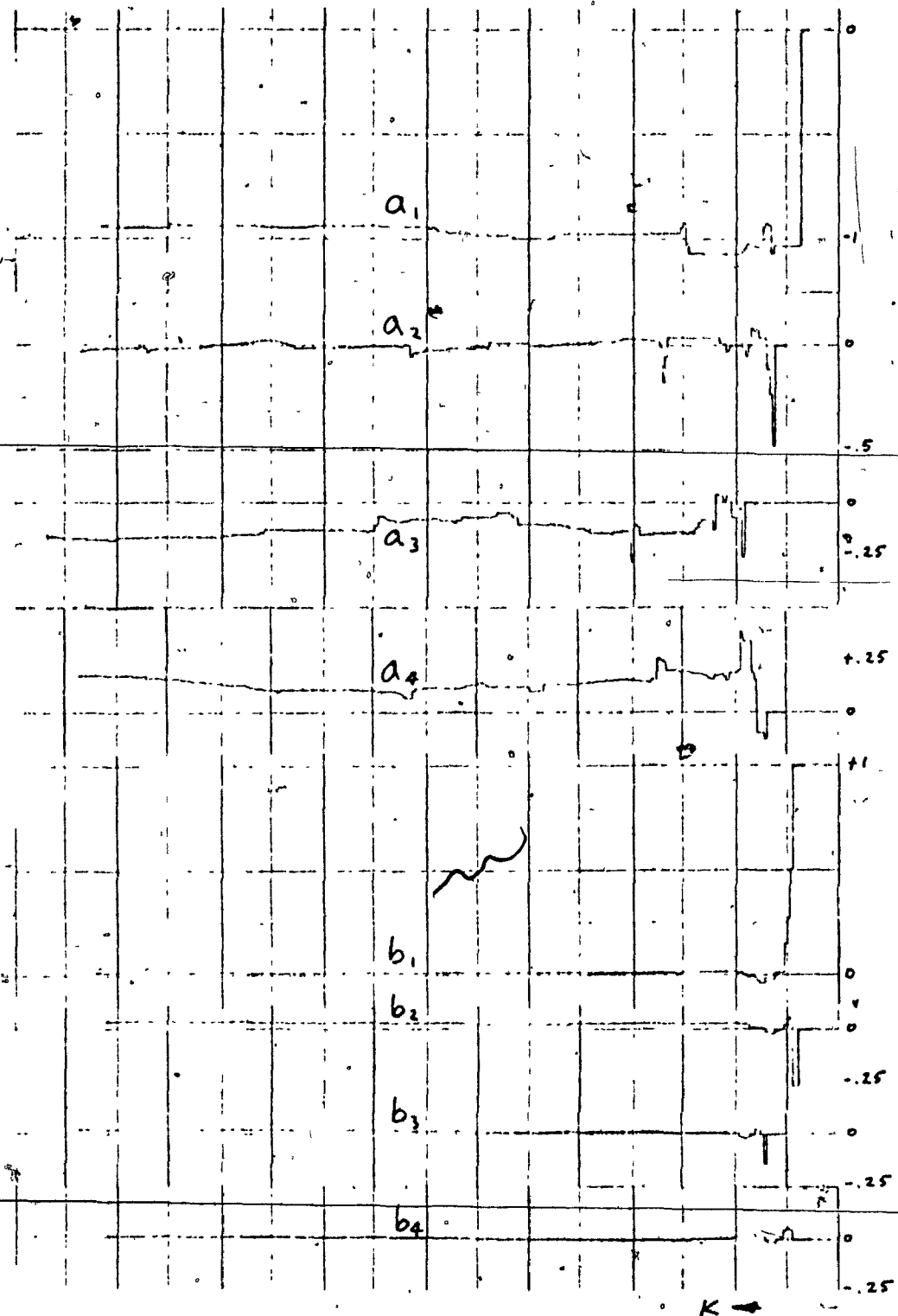
SELF-TUNING CONTROL ,

FOURTH ORDER

RUN # 5

 $T = 14$ seconds $b_1(0) = 1$

FIGURE 4.9b

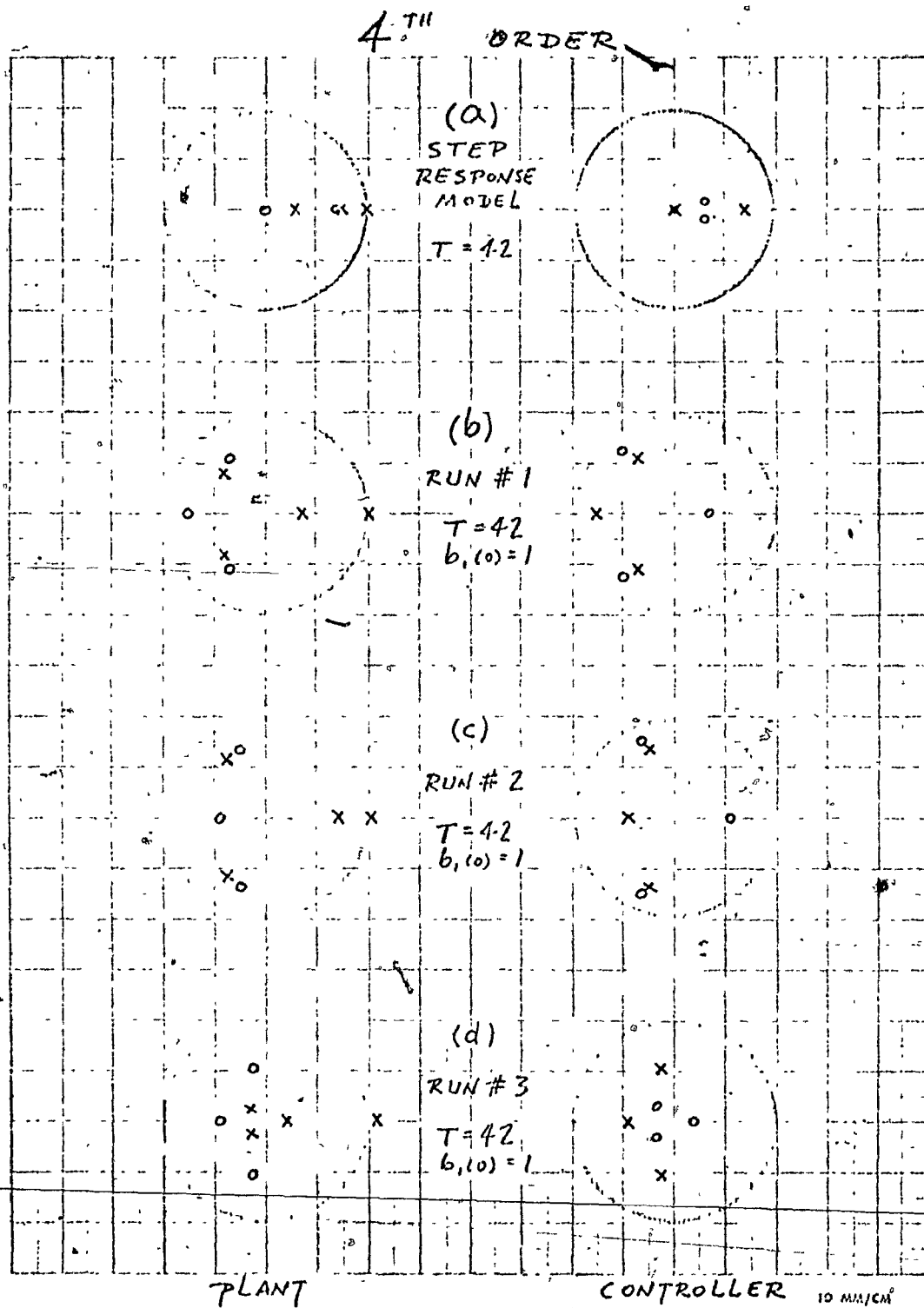


PARAMETER ESTIMATES

FOURTH ORDER

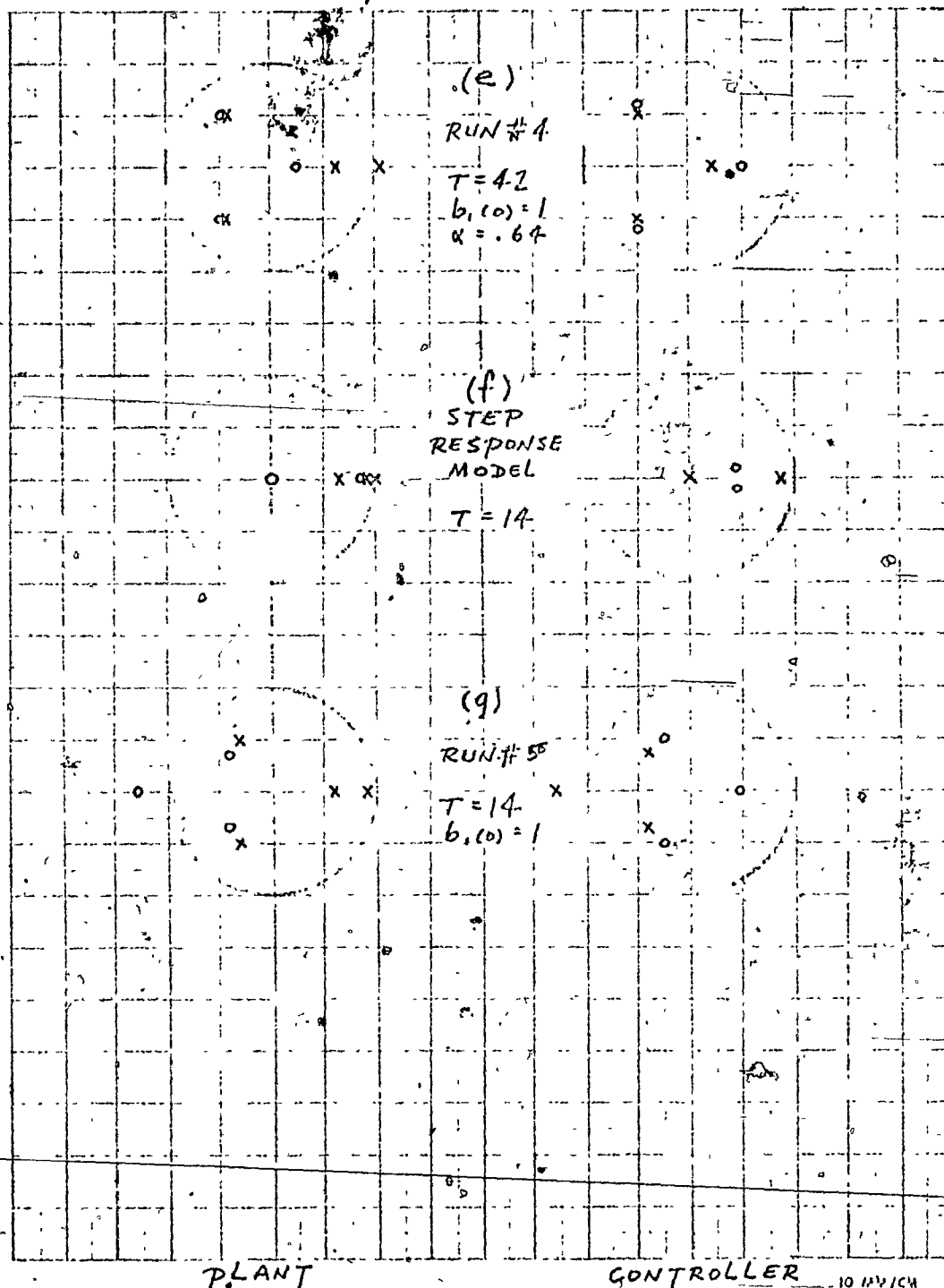
RUN # 5

FIGURE 4.10



POLE-ZERO PATTERN OF IDENTIFIED PLANT
AND OF CORRESPONDING MINIMUM VARIANCE CONTROLLER

FIGURE 4.10 (Continued)

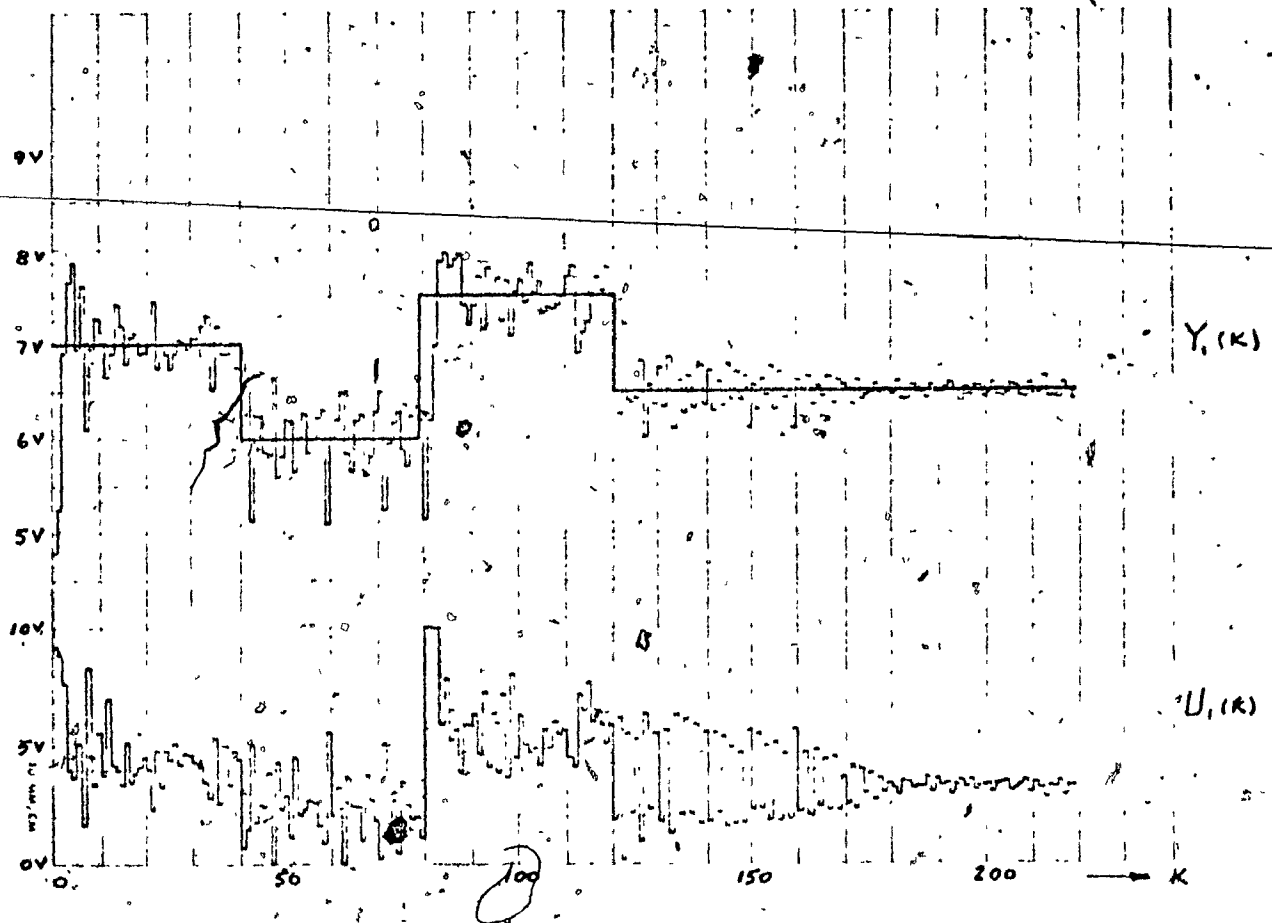
4TH ORDER

PLANT

CONTROLLER

10 MV/CM

FIGURE 4.11



MINIMUM VARIANCE CONTROL BASED ON RUN #1

 $T = 42$ seconds

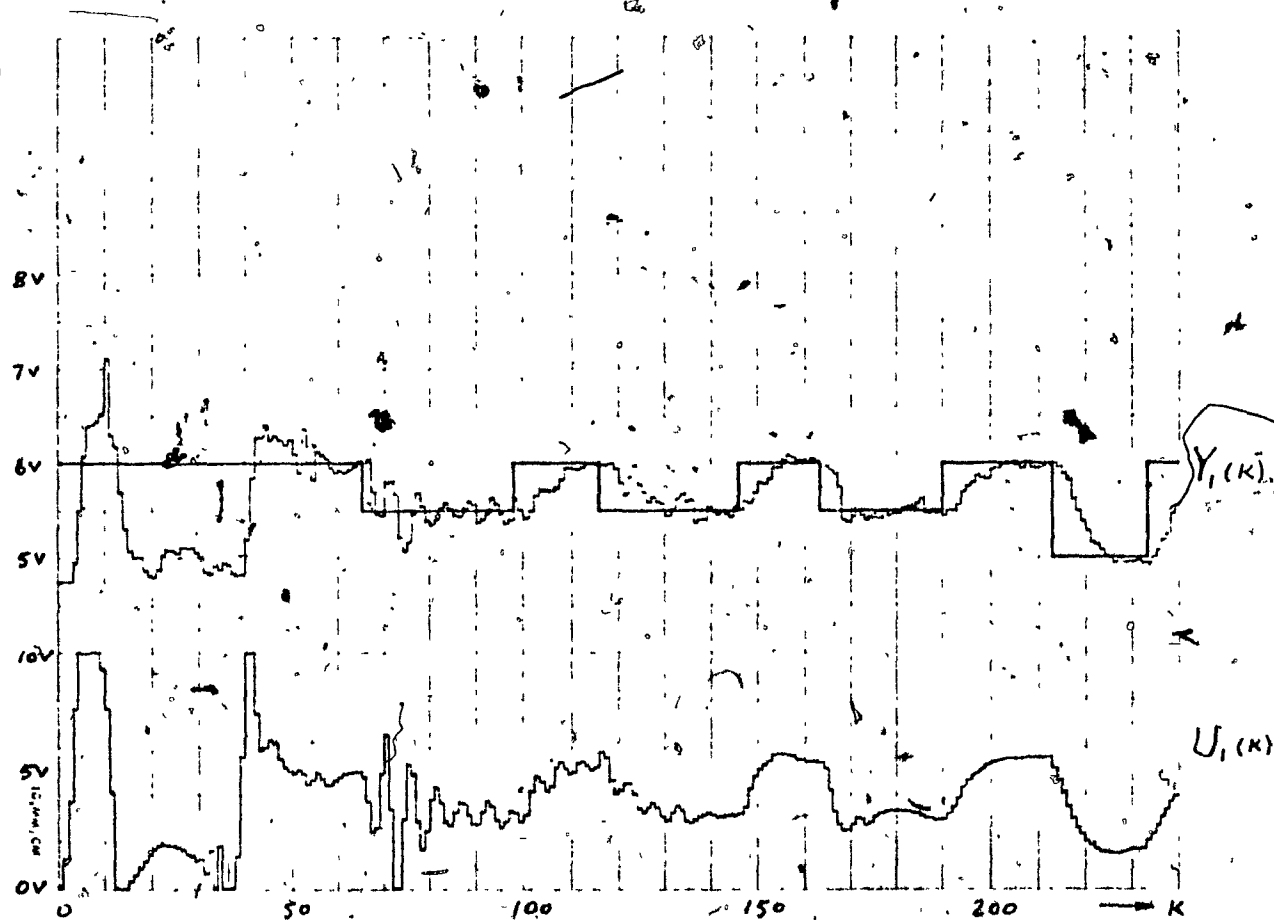
FIGURE 4.12



MINIMUM VARIANCE CONTROL BASED ON RUN # 2

$T = 42$ seconds

FIGURE 4.13a

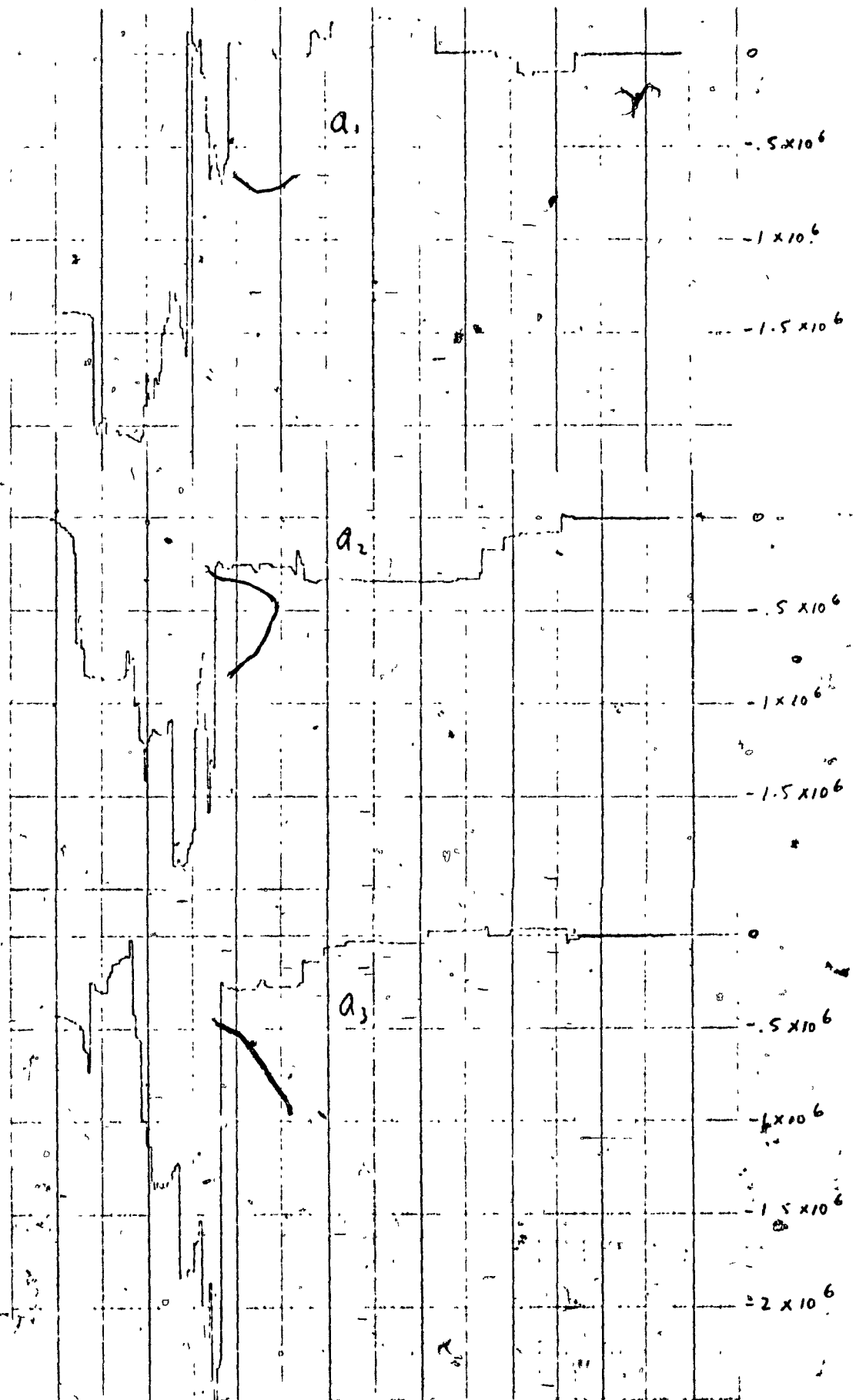


SELF-TUNING CONTROL

THIRD ORDER

RUN # 6

 $T = 42 \text{ seconds}$ $b_1(0) = 1$



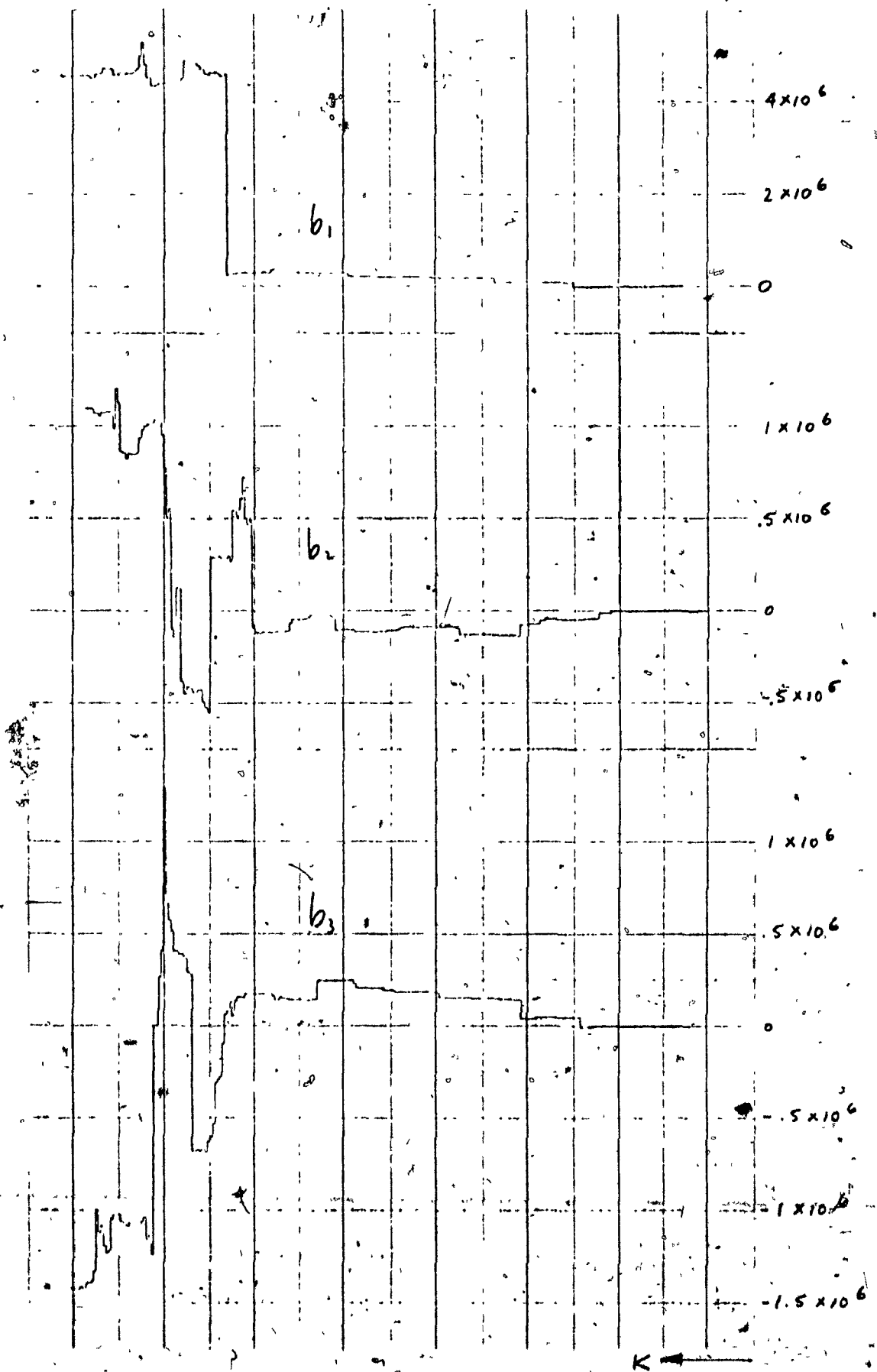
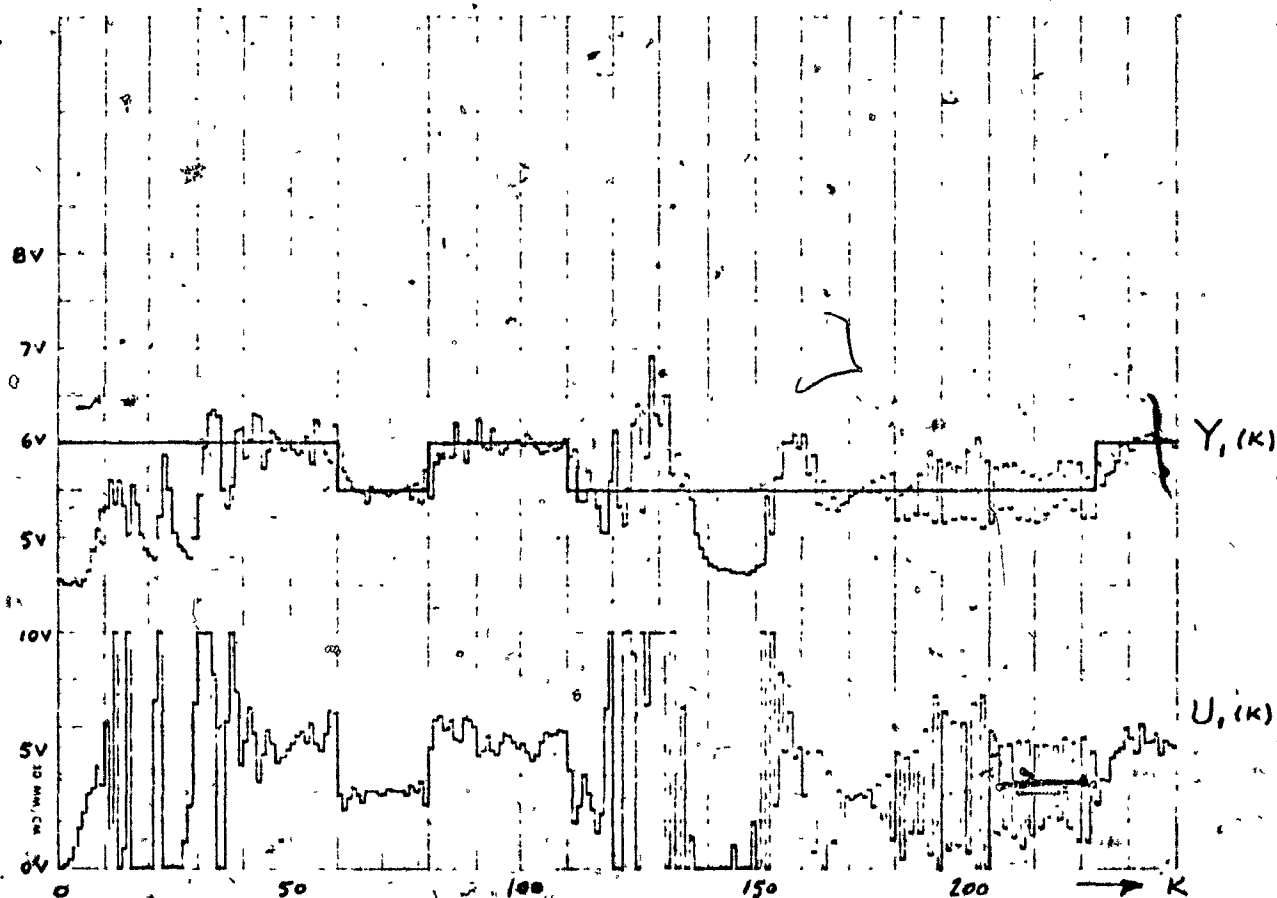


FIGURE 4.14a



SELF-TUNING CONTROL

THIRD ORDER

RUN

 $T = 42$ seconds $h_1(0) = 10$

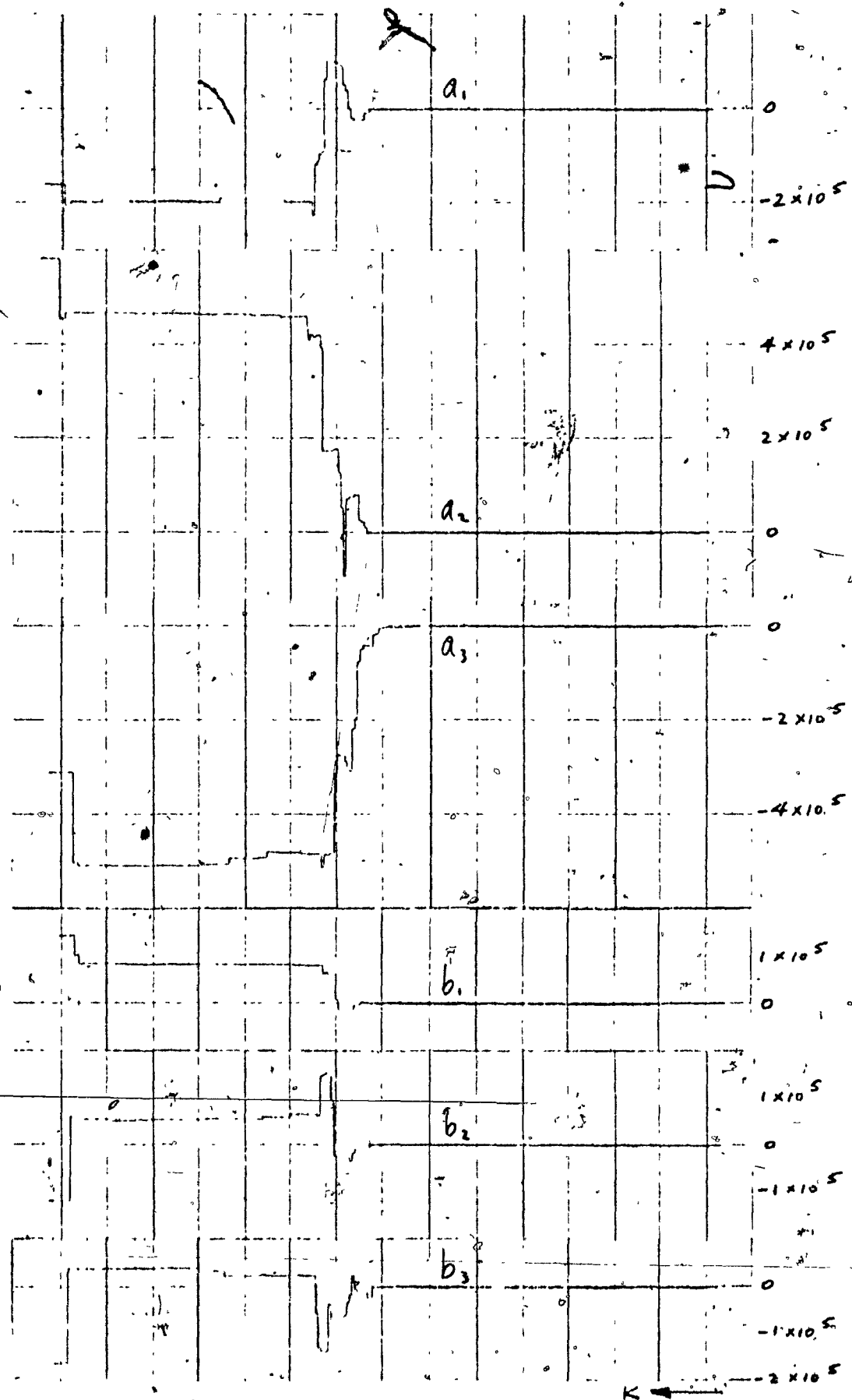
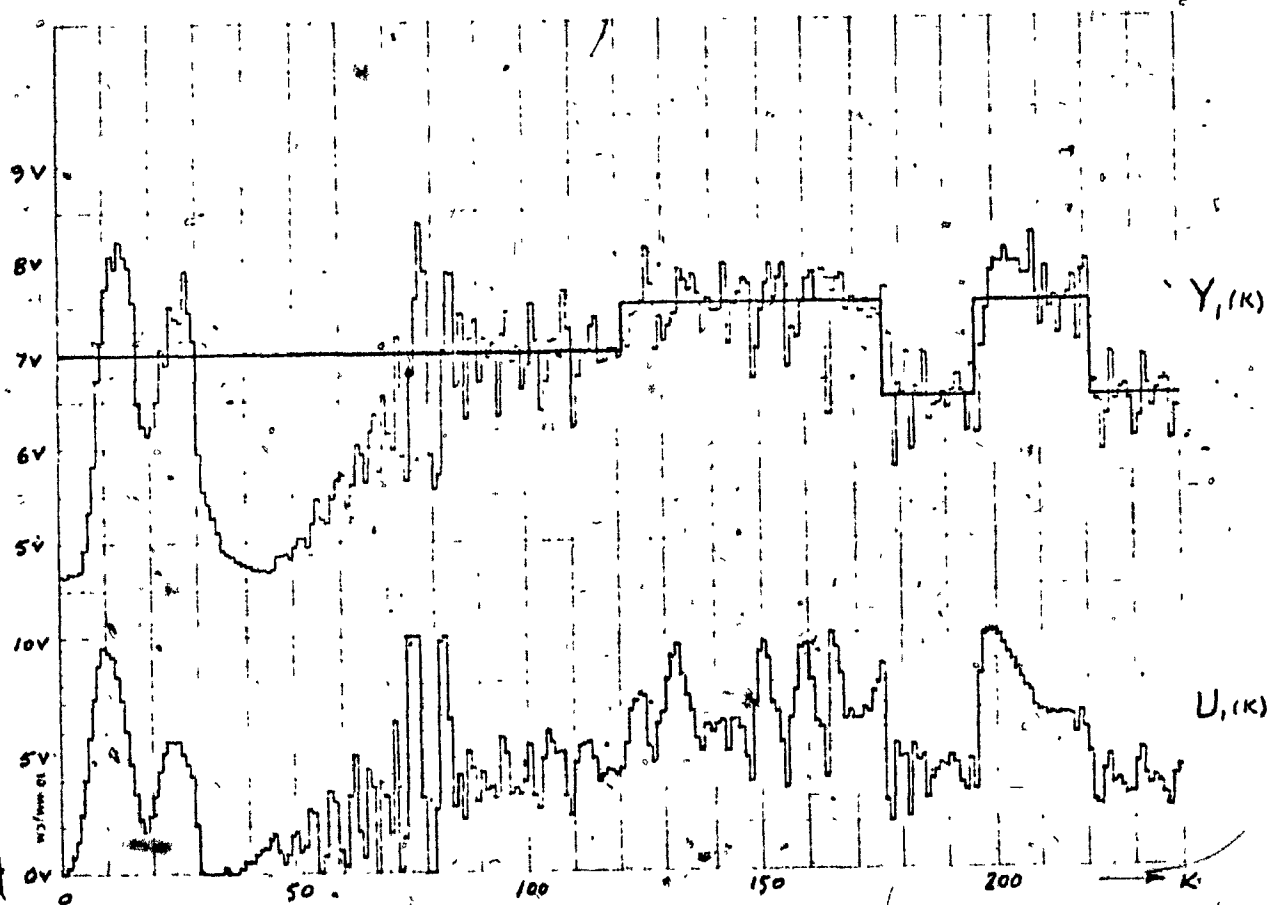


FIGURE 4.15a



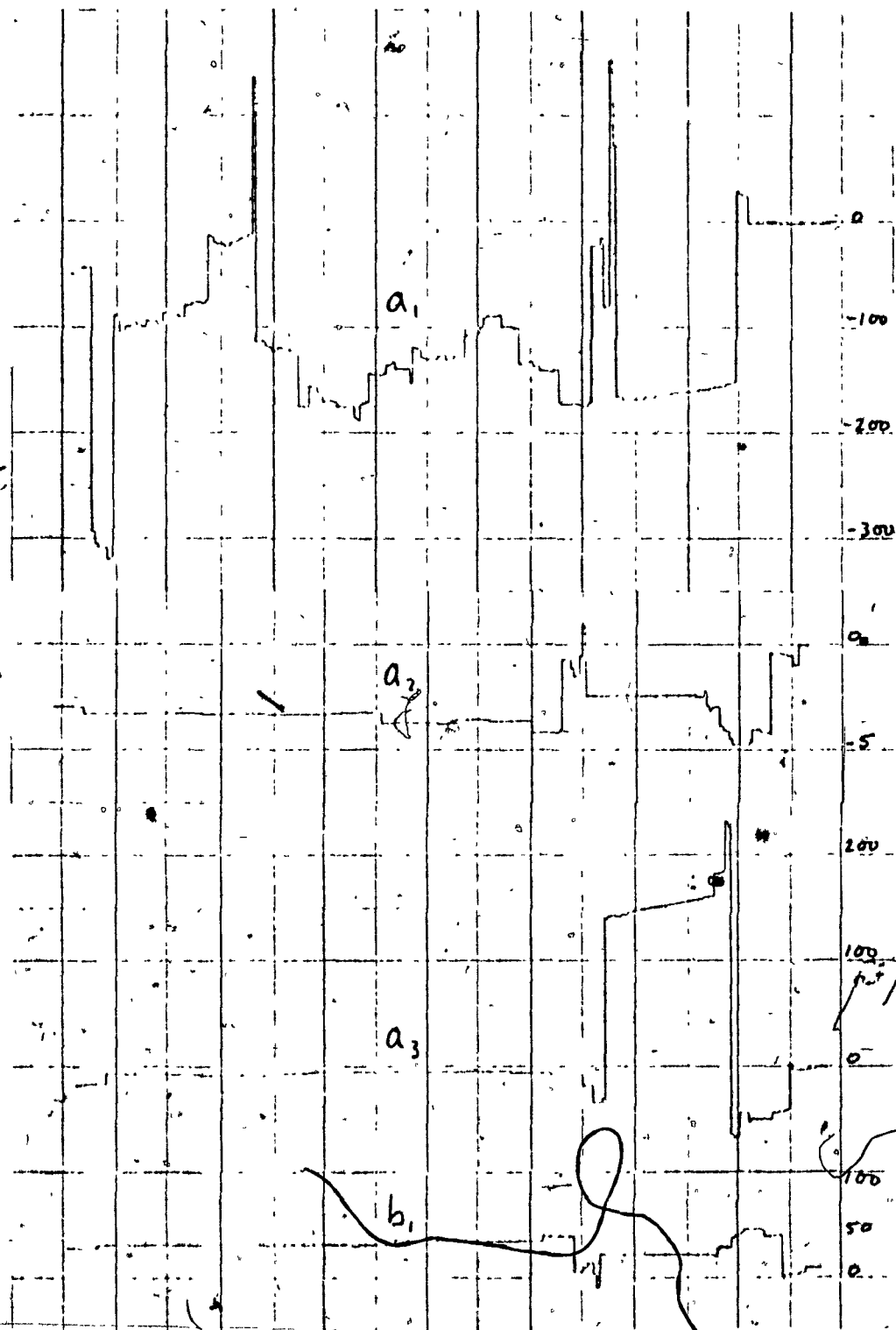
SELF-TUNING CONTROL

THIRD ORDER (PID)

RUN # 8

 $T = 42 \text{ seconds}$ $h_1(0) = 10$

FIGURE 4.15b

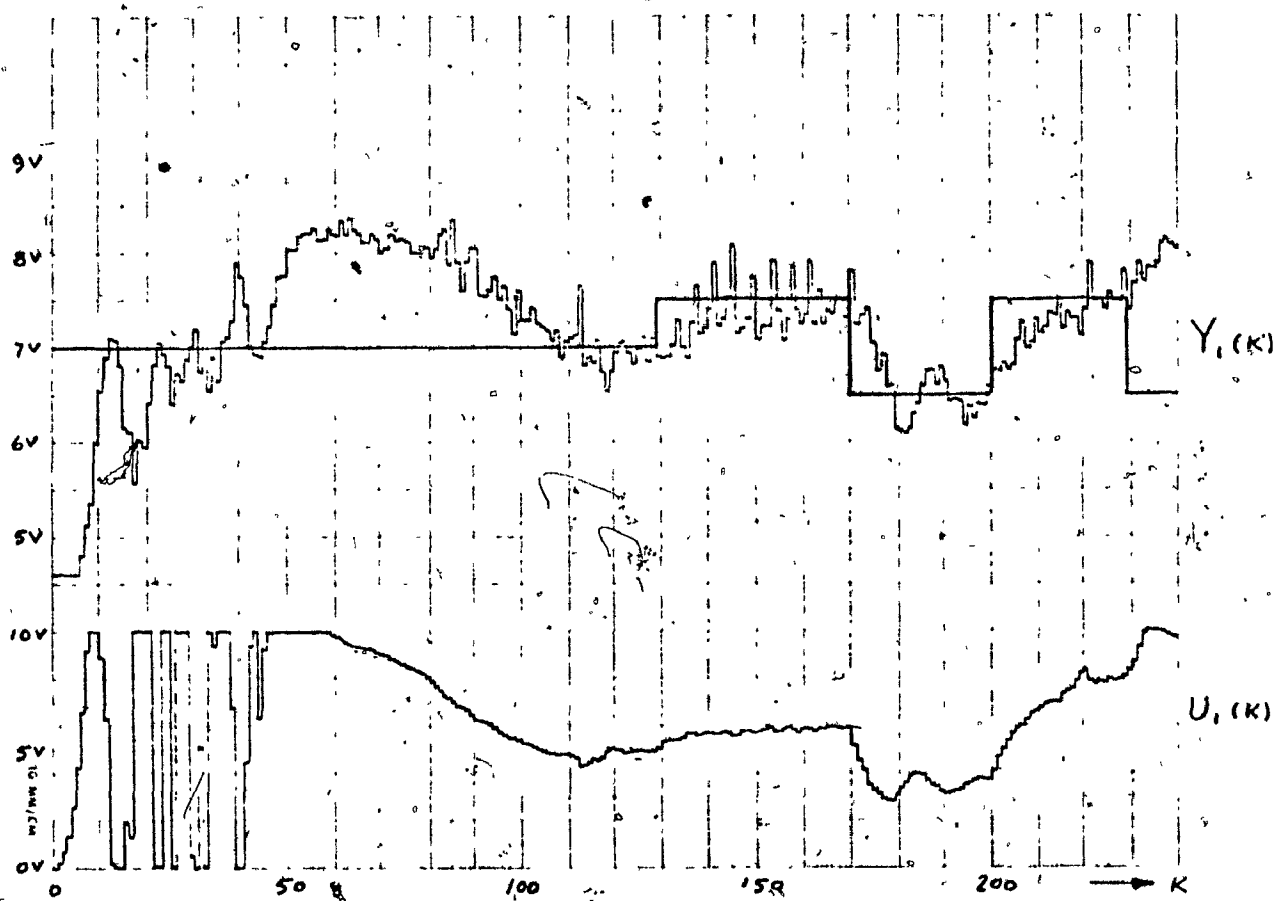


PARAMETER ESTIMATES

THIRD ORDER (PID)

RUN # 8

FIGURE 4.16a

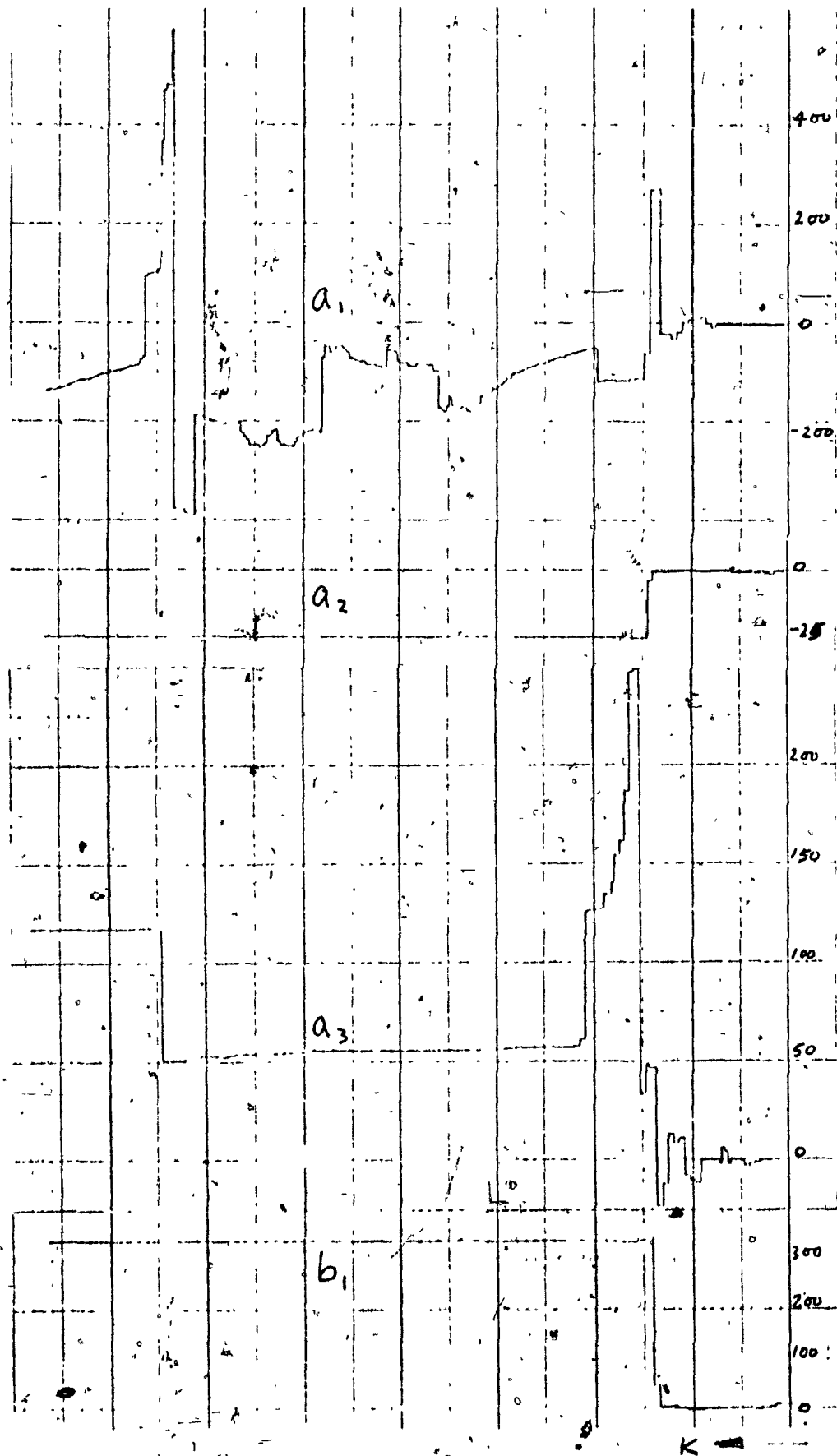


SELF-TUNING CONTROL

THIRD ORDER (PID)

RUN # 9

 $T = 14$ seconds $b_1(0) = 10$

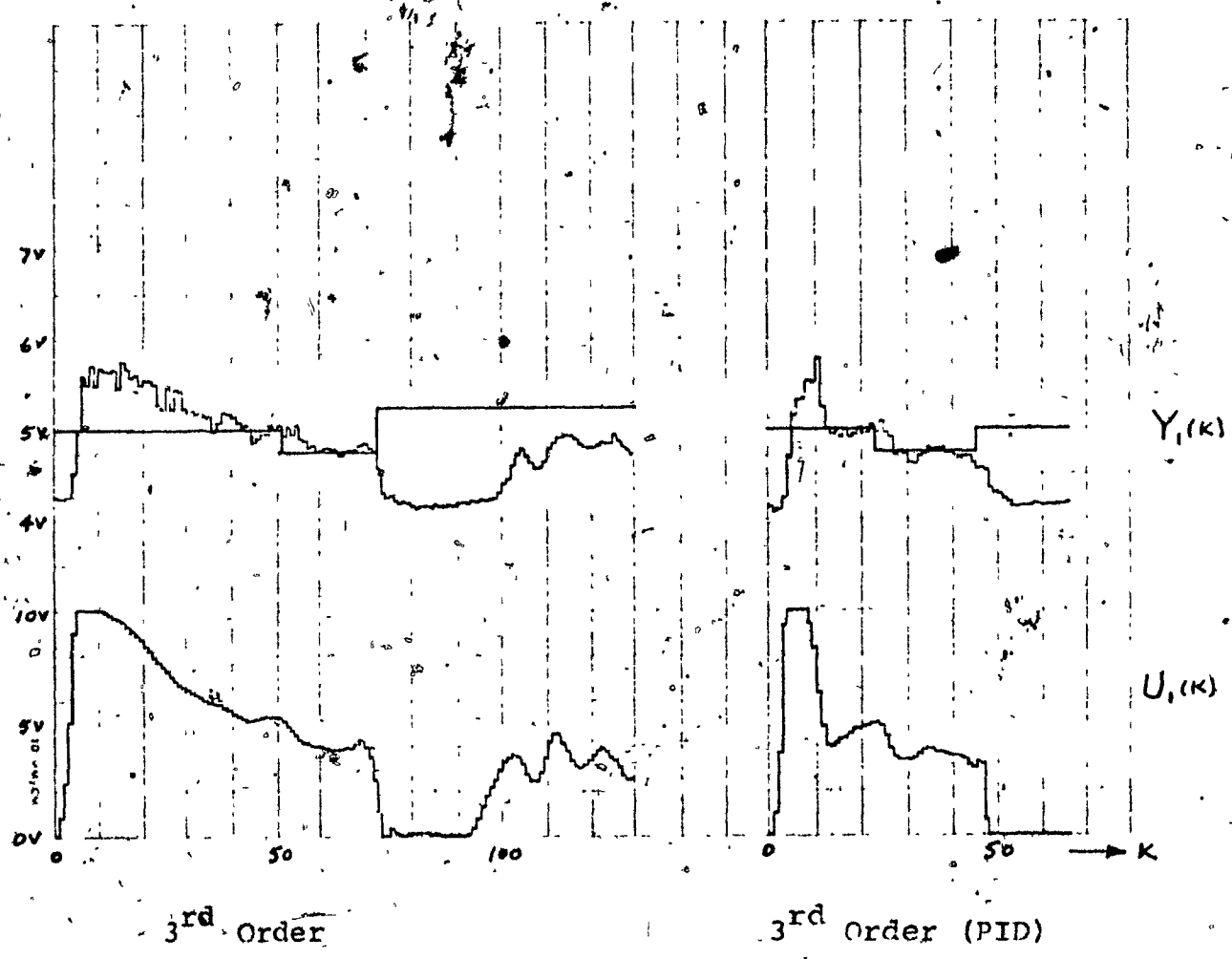


THIRD ORDER (PID)

PARAMETER ESTIMATES

— RUN # 9

FIGURE 4.17a



SELF-TUNING CONTROL

THIRD ORDER

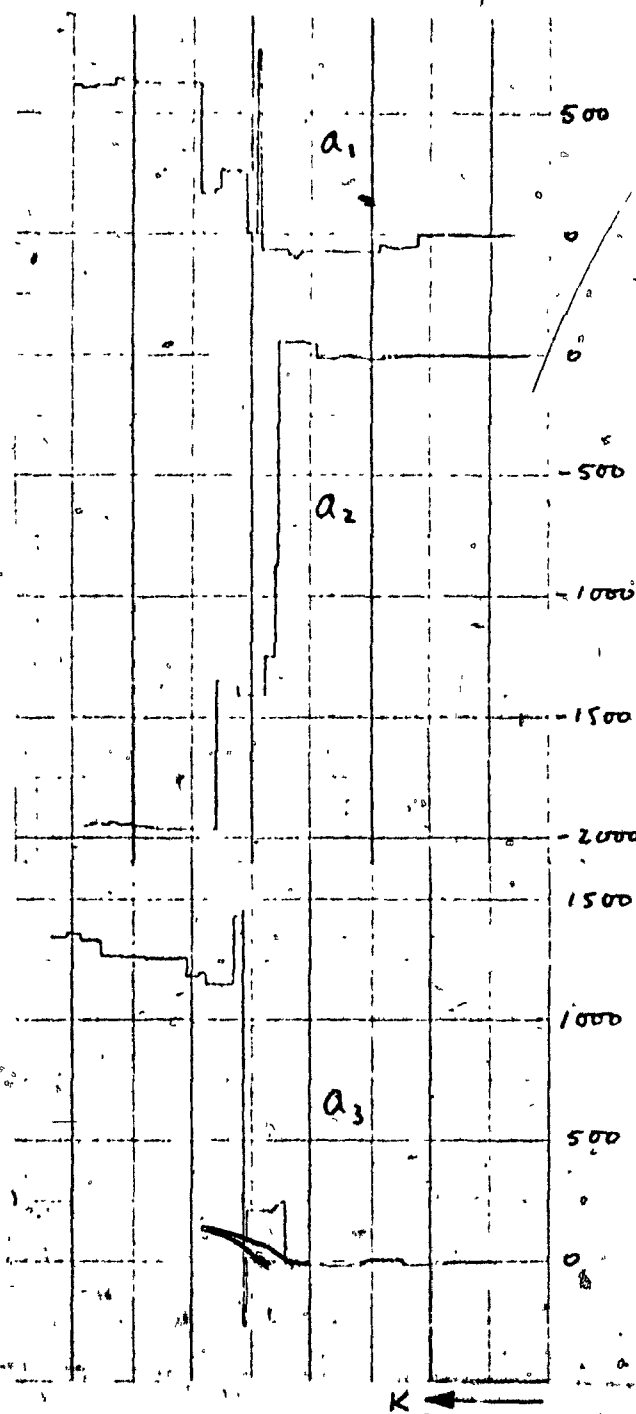
RUN # 10

$T = 42$ seconds

$b_1(0) = 1$

$a = .64 \text{ volts}^2$

FIGURE 4.17b



PARAMETER ESTIMATES

THIRD ORDER

RUN # 10

FIGURE 4.17b (Continued)

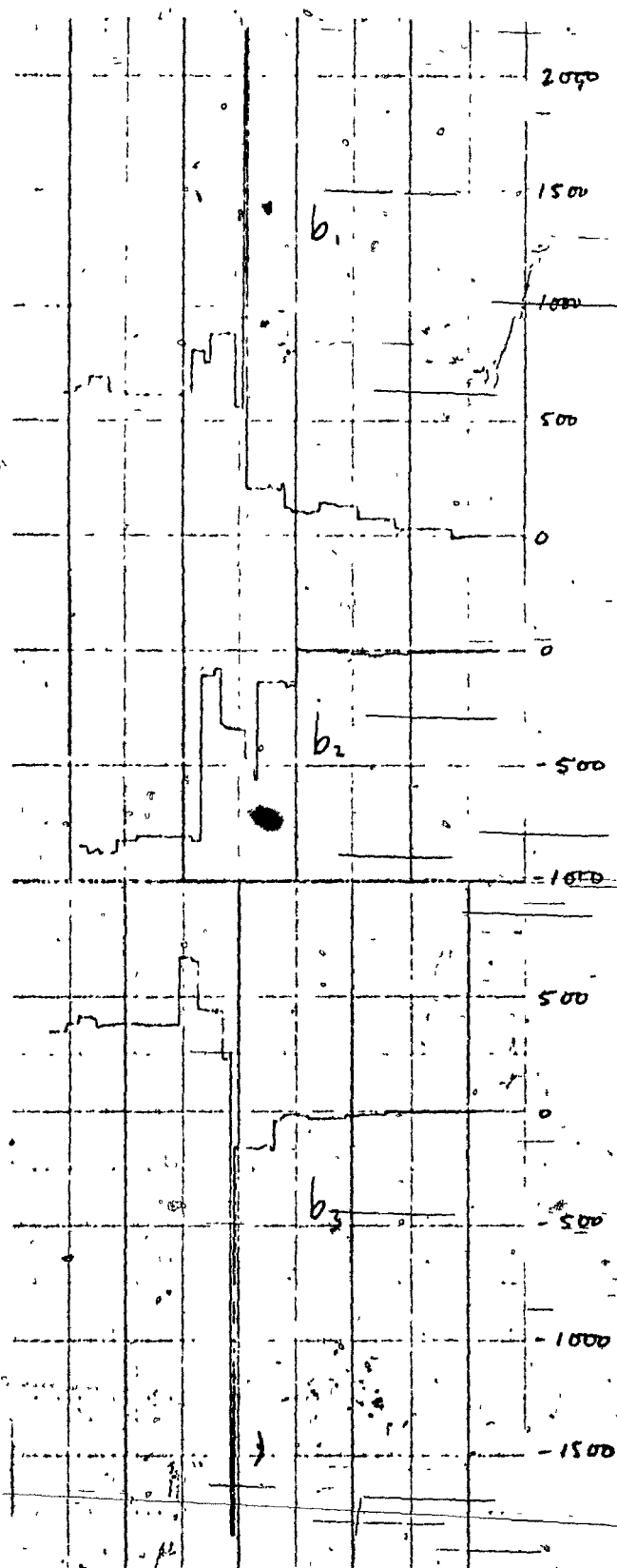
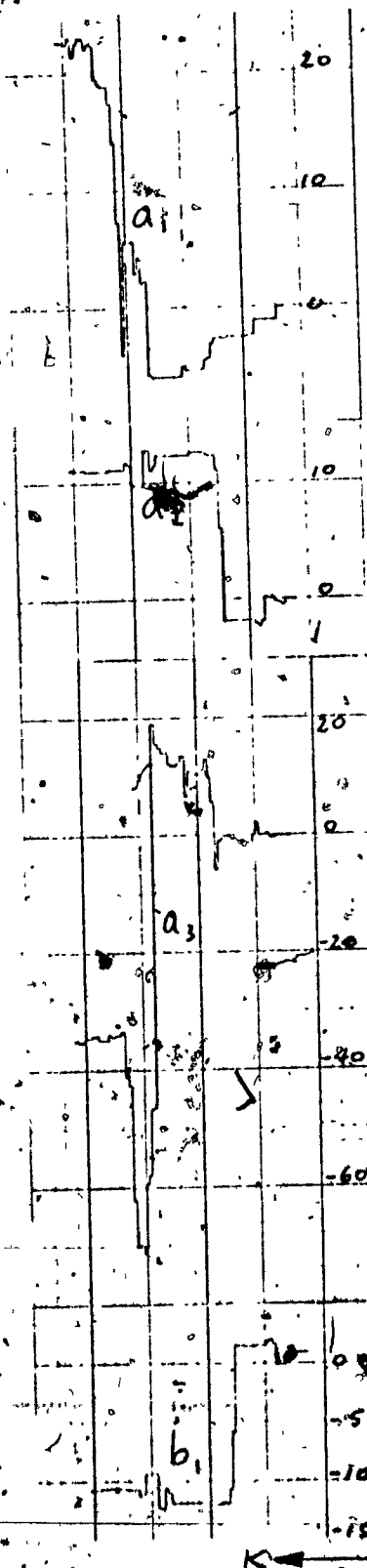


FIGURE 4.17c

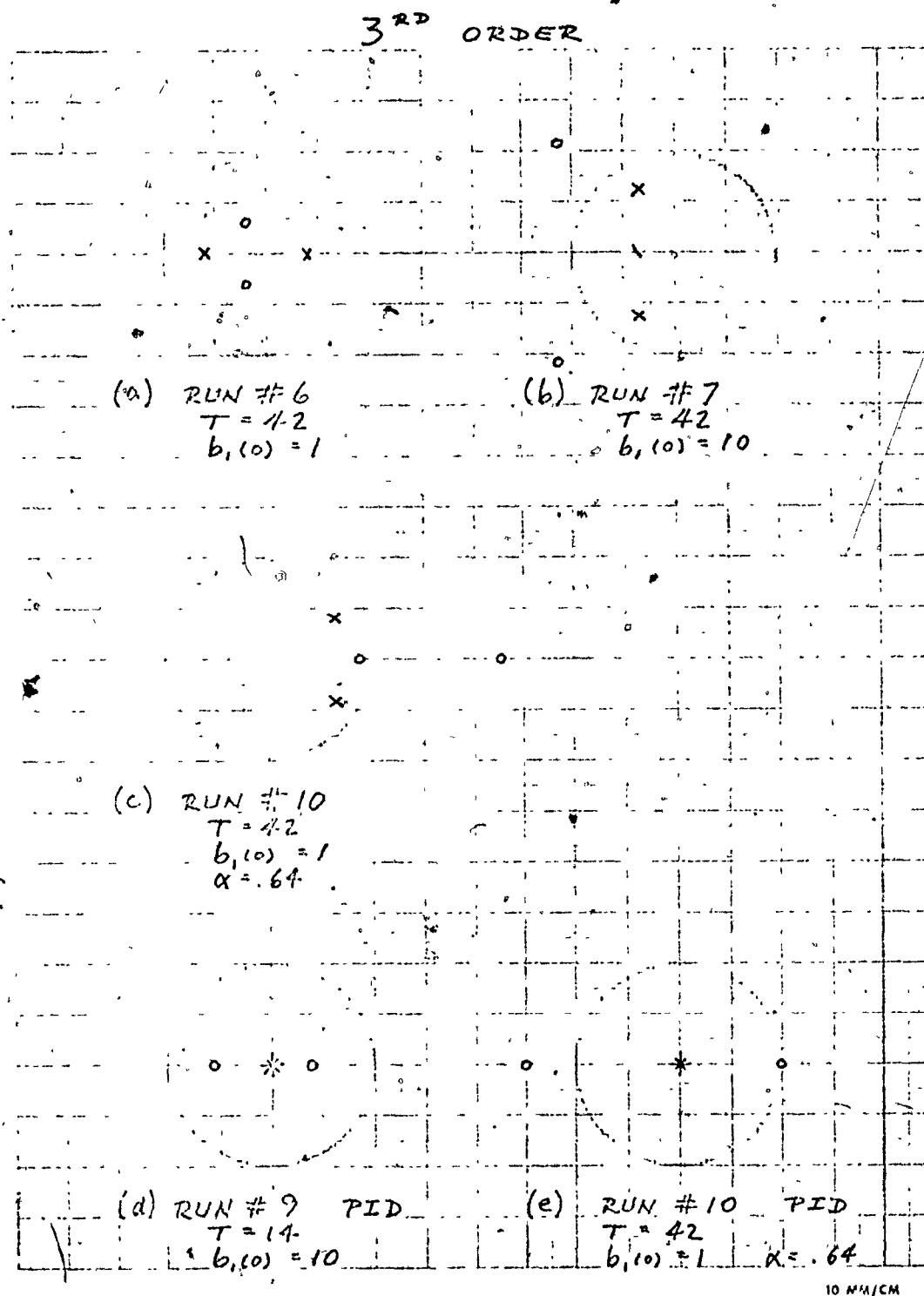


PARAMETER ESTIMATES

THIRD ORDER (PID)

RUN # 10

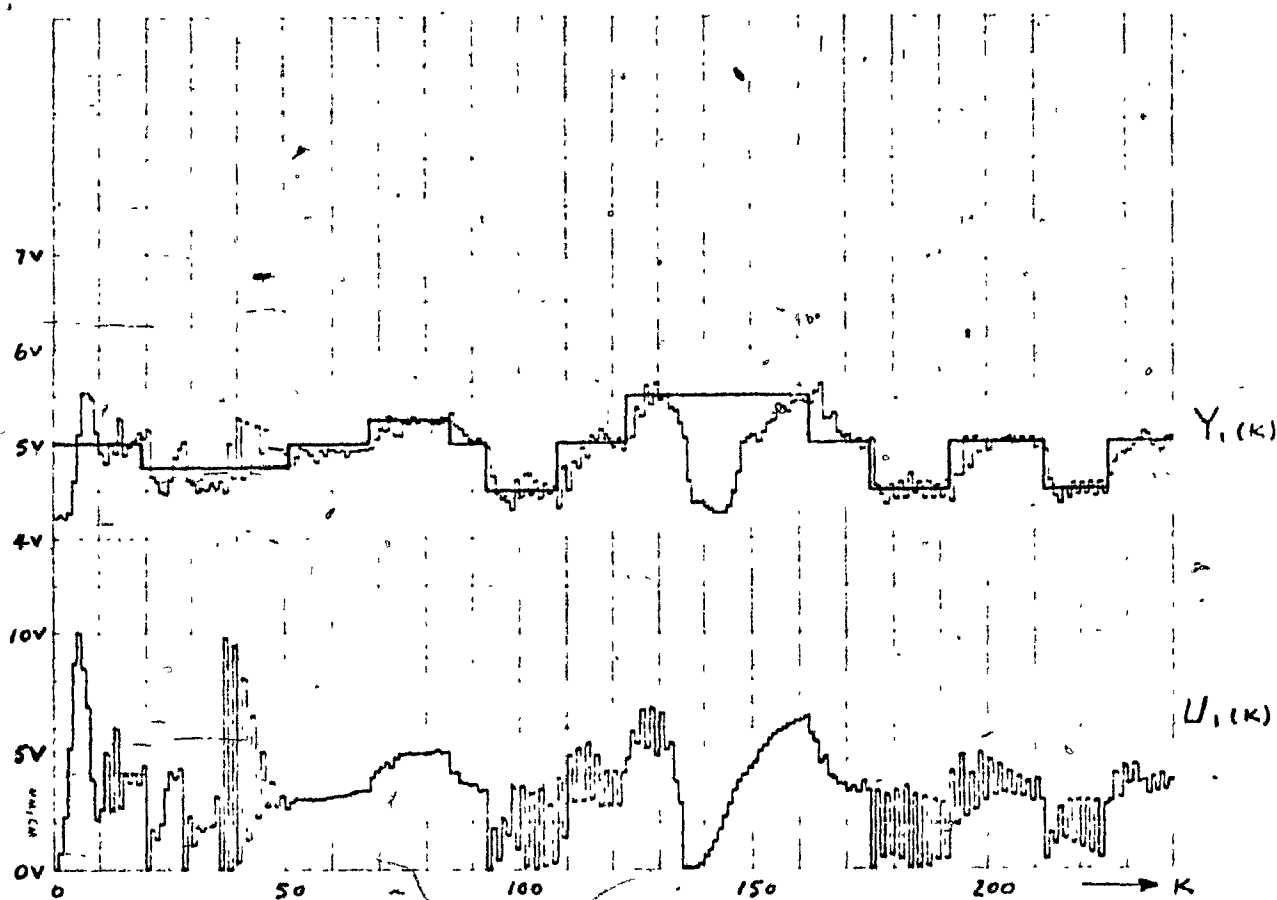
FIGURE 4.18



POLE-ZERO PATTERN OF MINIMUM VARIANCE CONTROLLER

OBTAINED FROM IDENTIFICATION ALGORITHM

FIGURE 4.19a

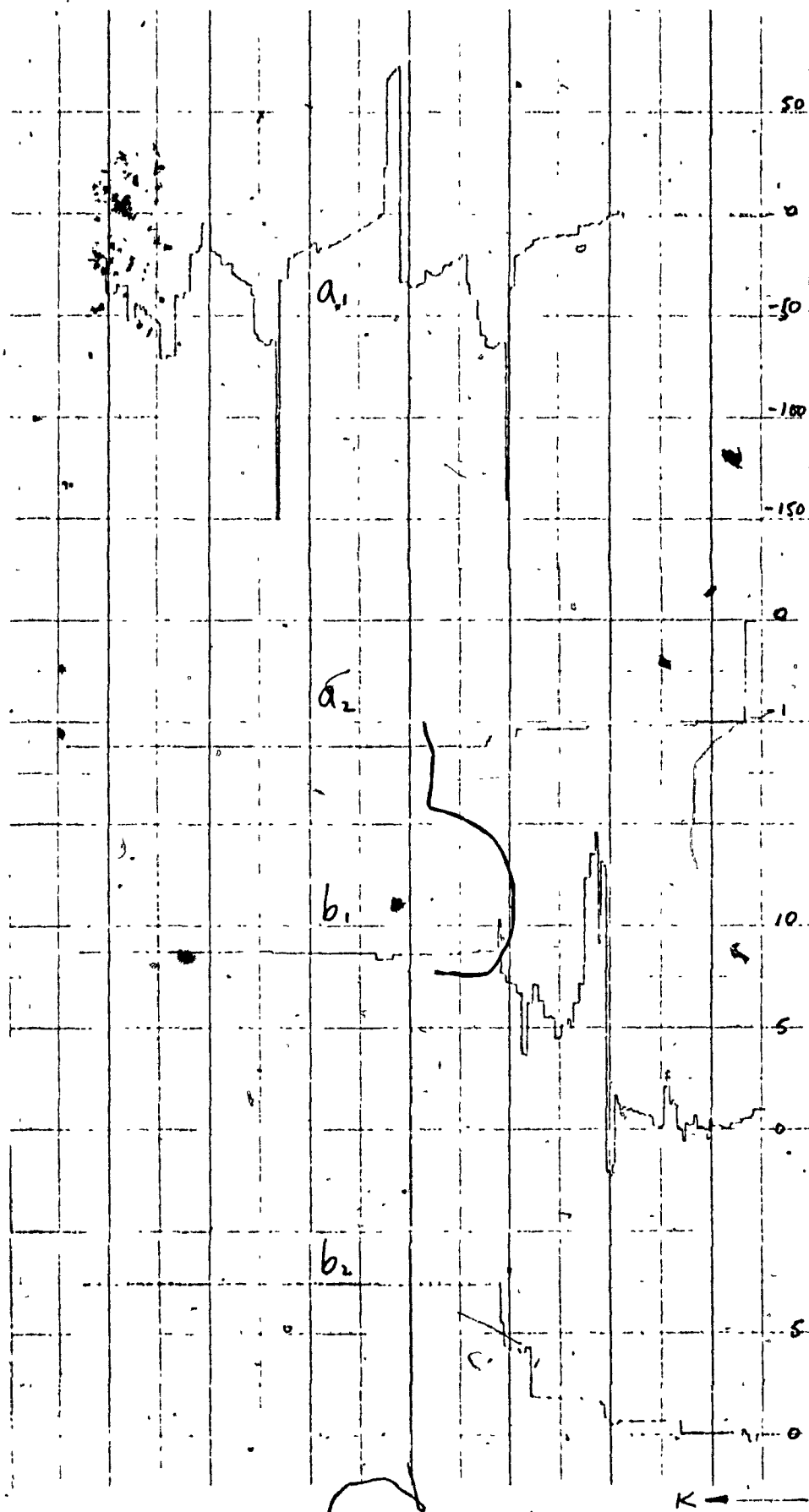


SELF-TUNING CONTROL

SECOND ORDER

RUN # 11

 $T = 42$ seconds $b_1(0) = 1$

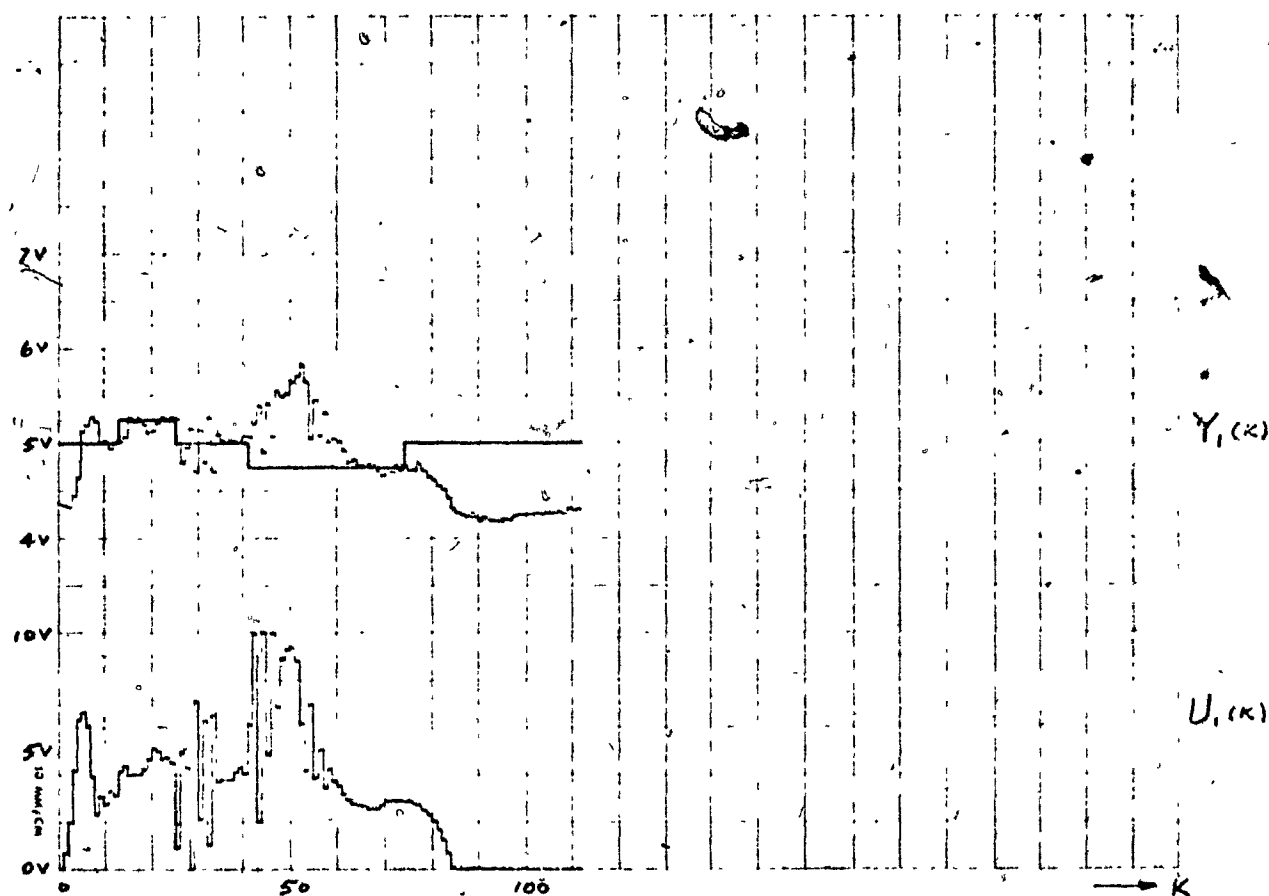


SECOND ORDER

PARAMETER ESTIMATES

RUN # 11

FIGURE 4.20a



SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 12

 $T = 42$ seconds $b_1(0) = 1$

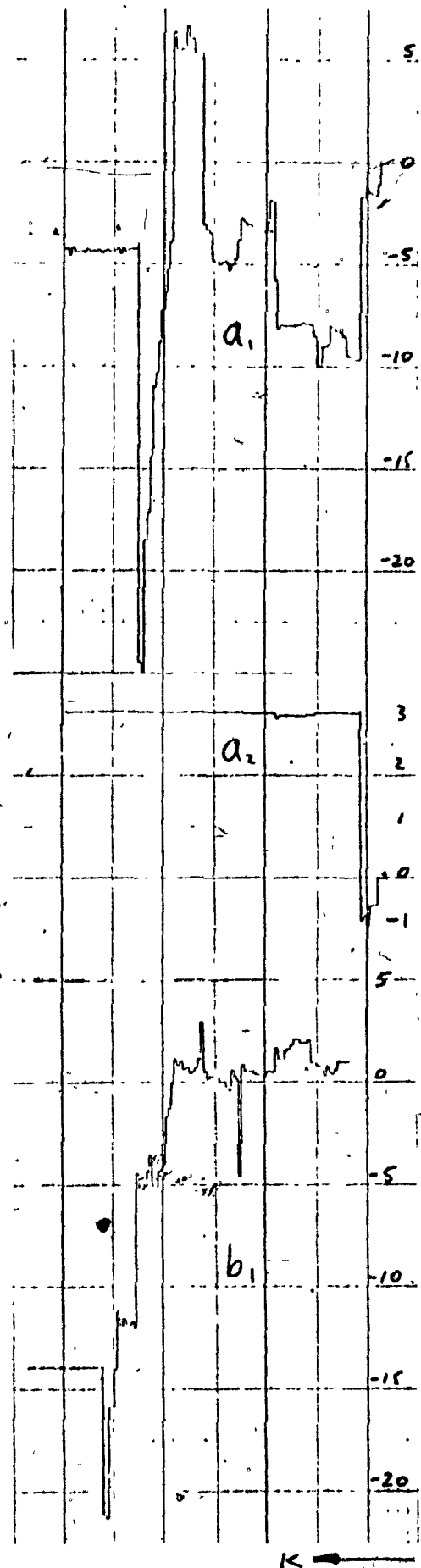
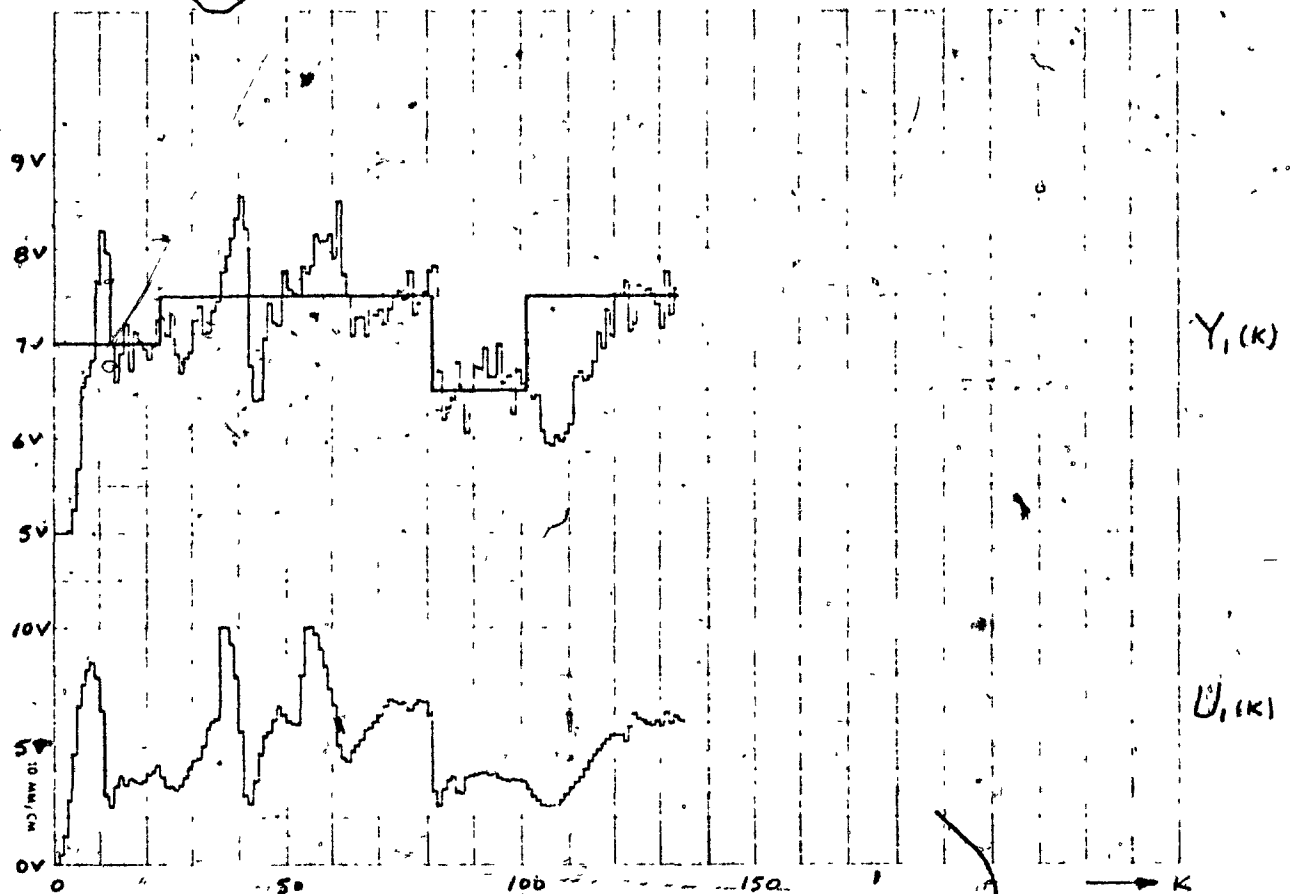


FIGURE 4.21a



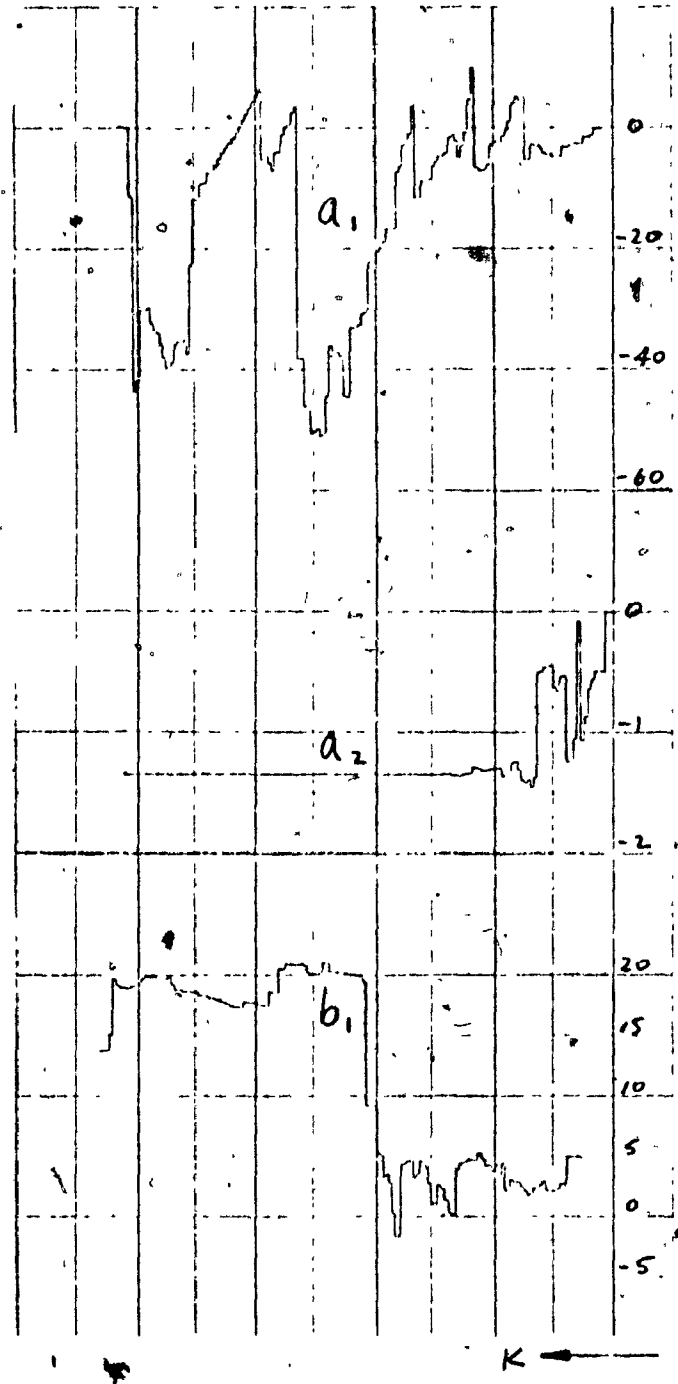
SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 13

 $T = 42 \text{ seconds}$
 $b_1(0) = 5$

FIGURE 4.21b

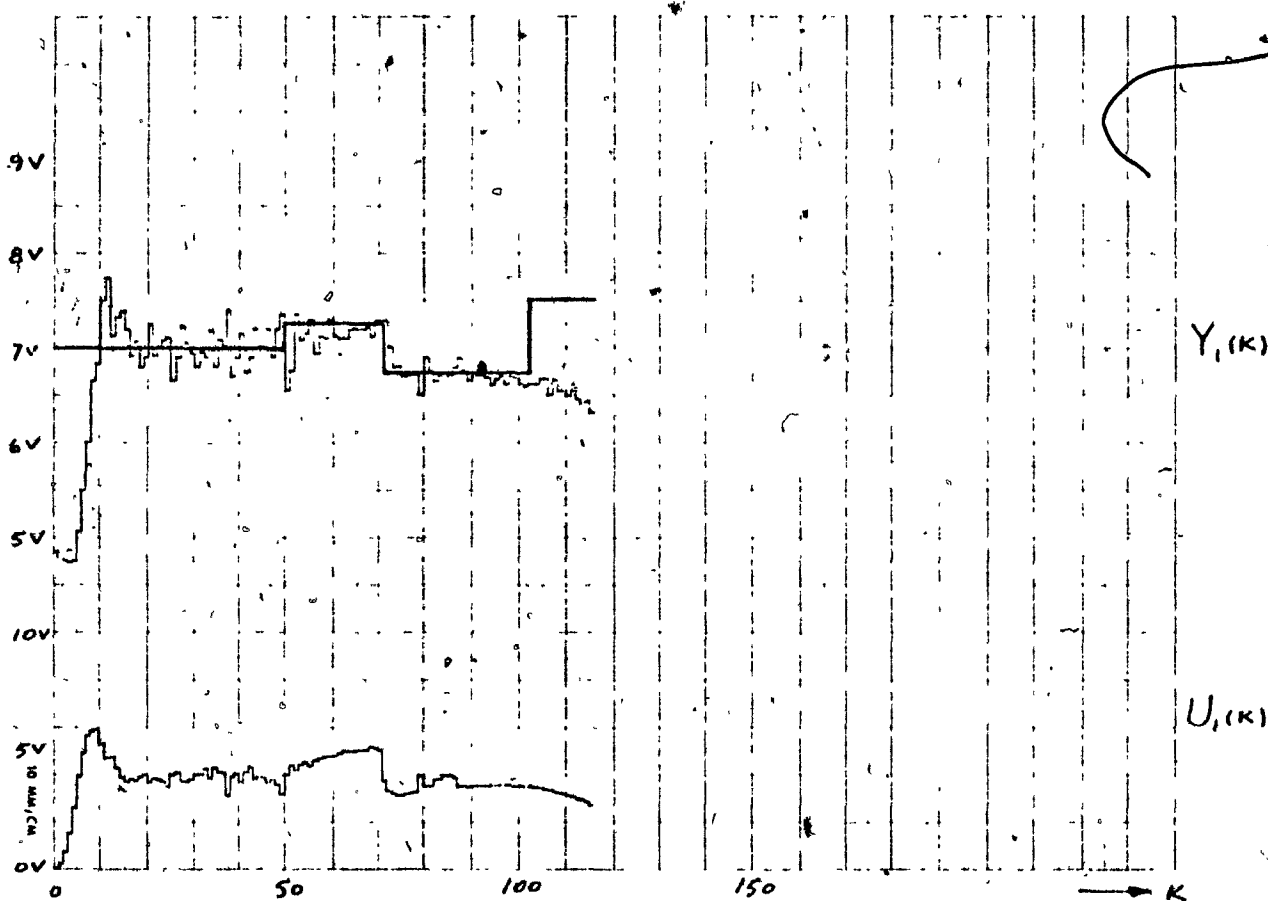


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 13

FIGURE 4.22a



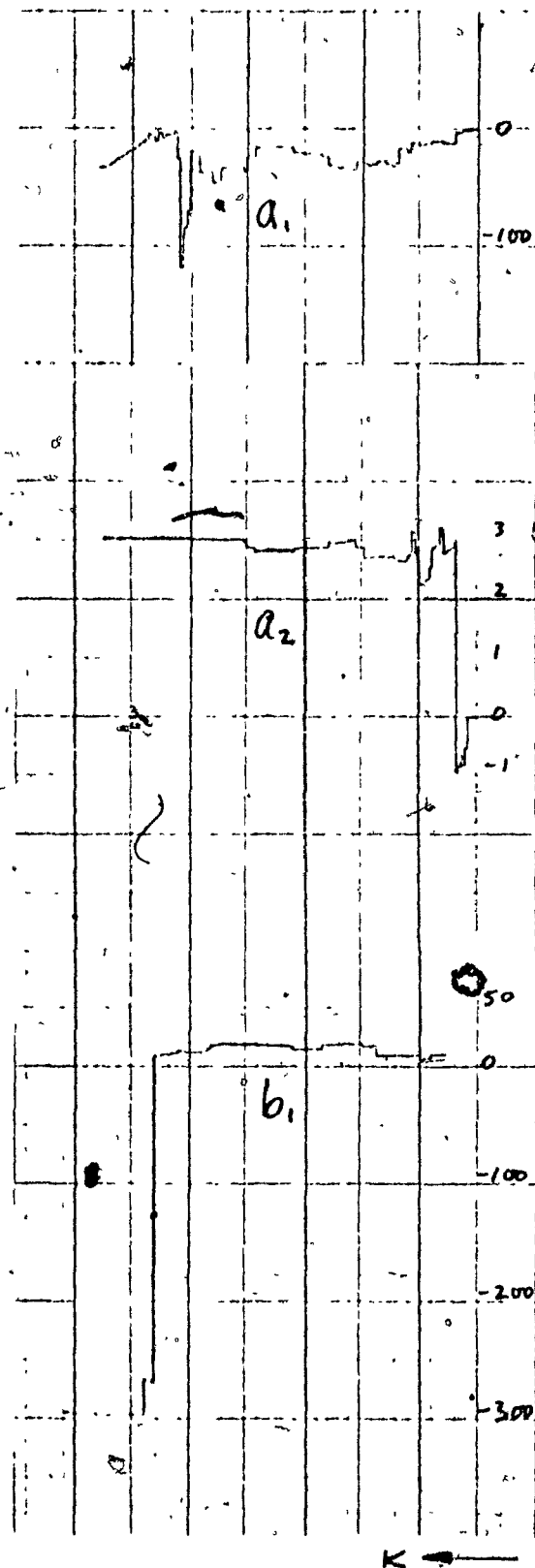
SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 14

 $T = 42 \text{ seconds}$ $b_1(0) = 10$

FIGURE 4.22b

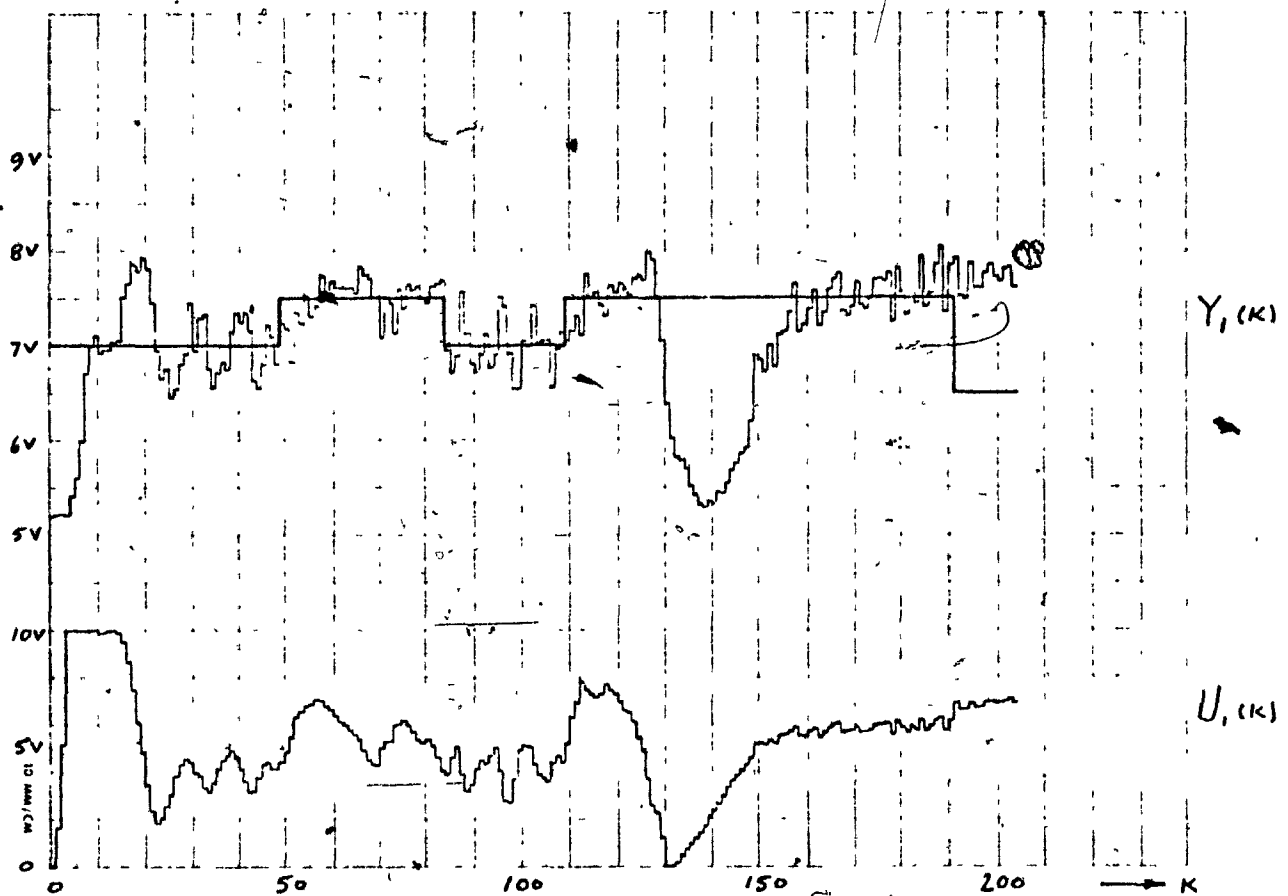


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 14

FIGURE 4.23a



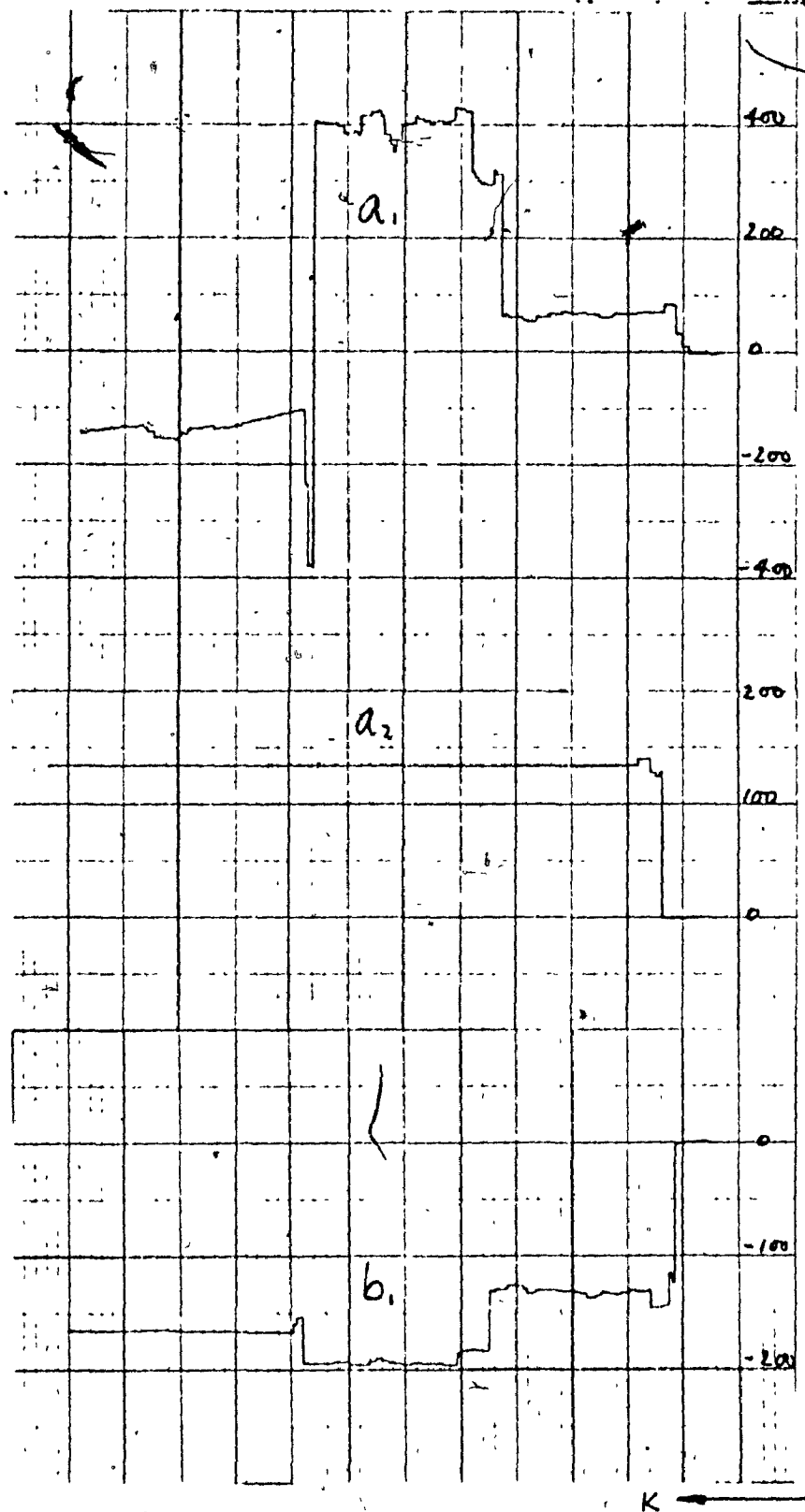
SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 15

 $T = 14 \text{ seconds}$ $b_1(0) = 1$

FIGURE 4.23b

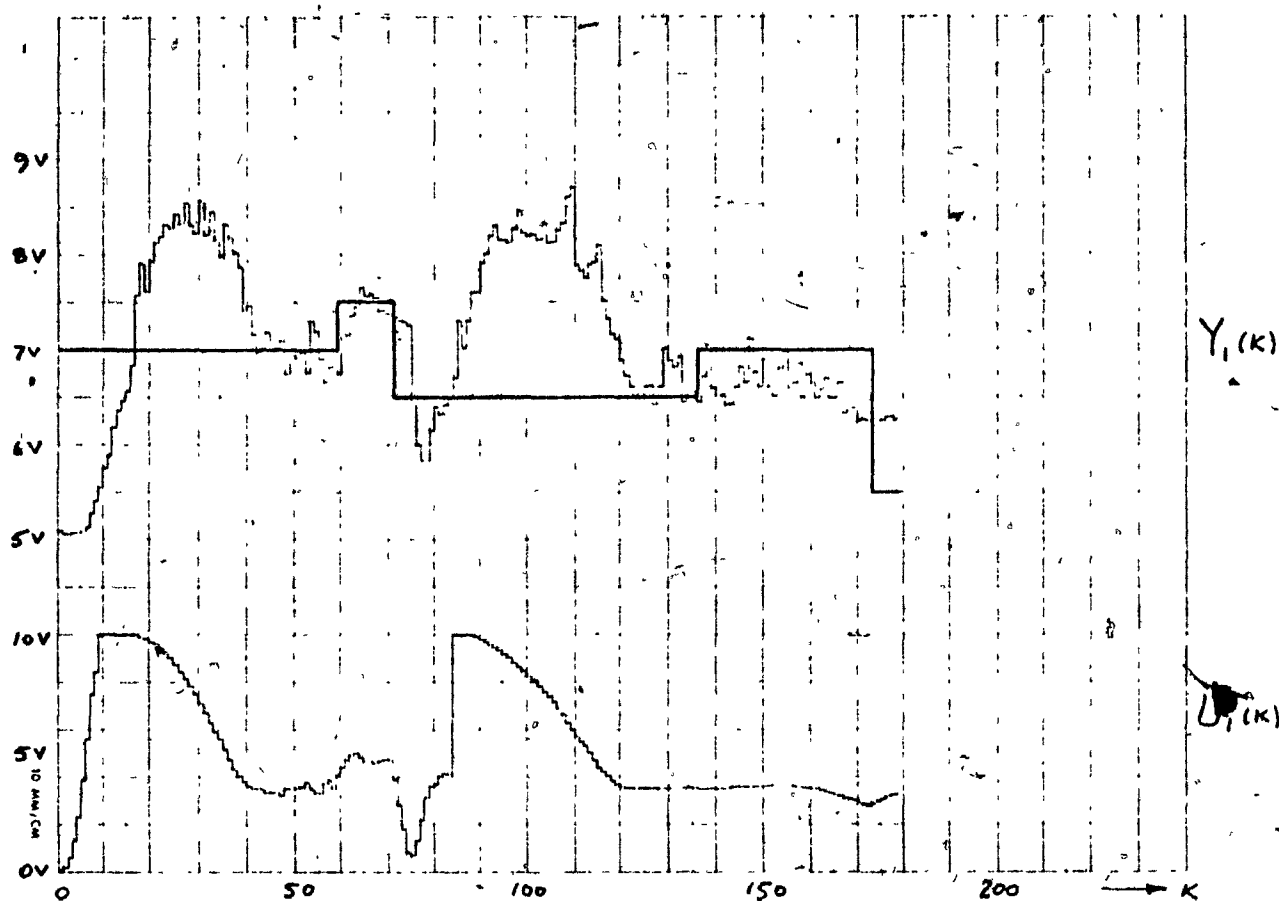


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 15

FIGURE 4.24a



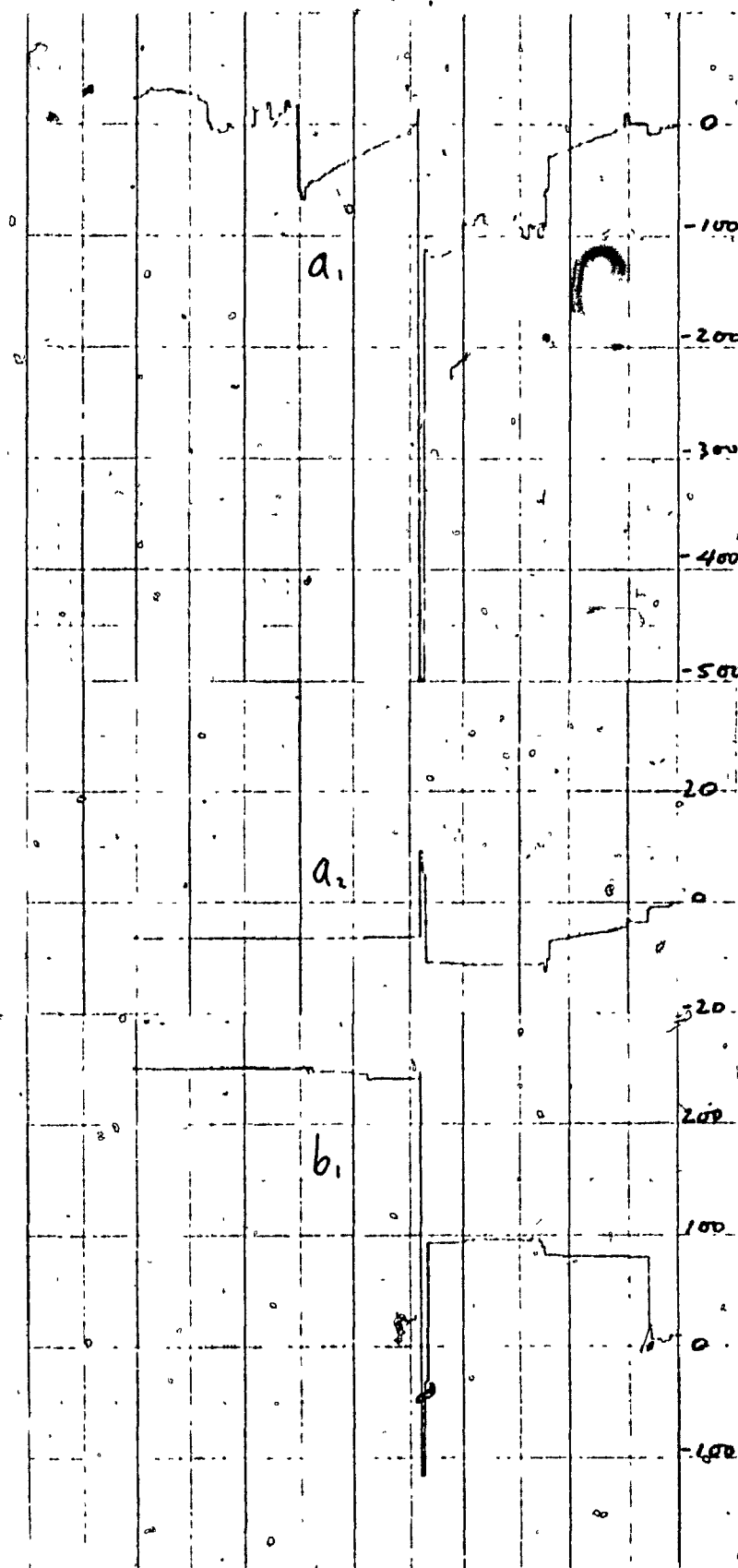
SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 16

 $T = 14$ seconds $b_1(0) = 10^{-4}$

FIGURE 4.24b

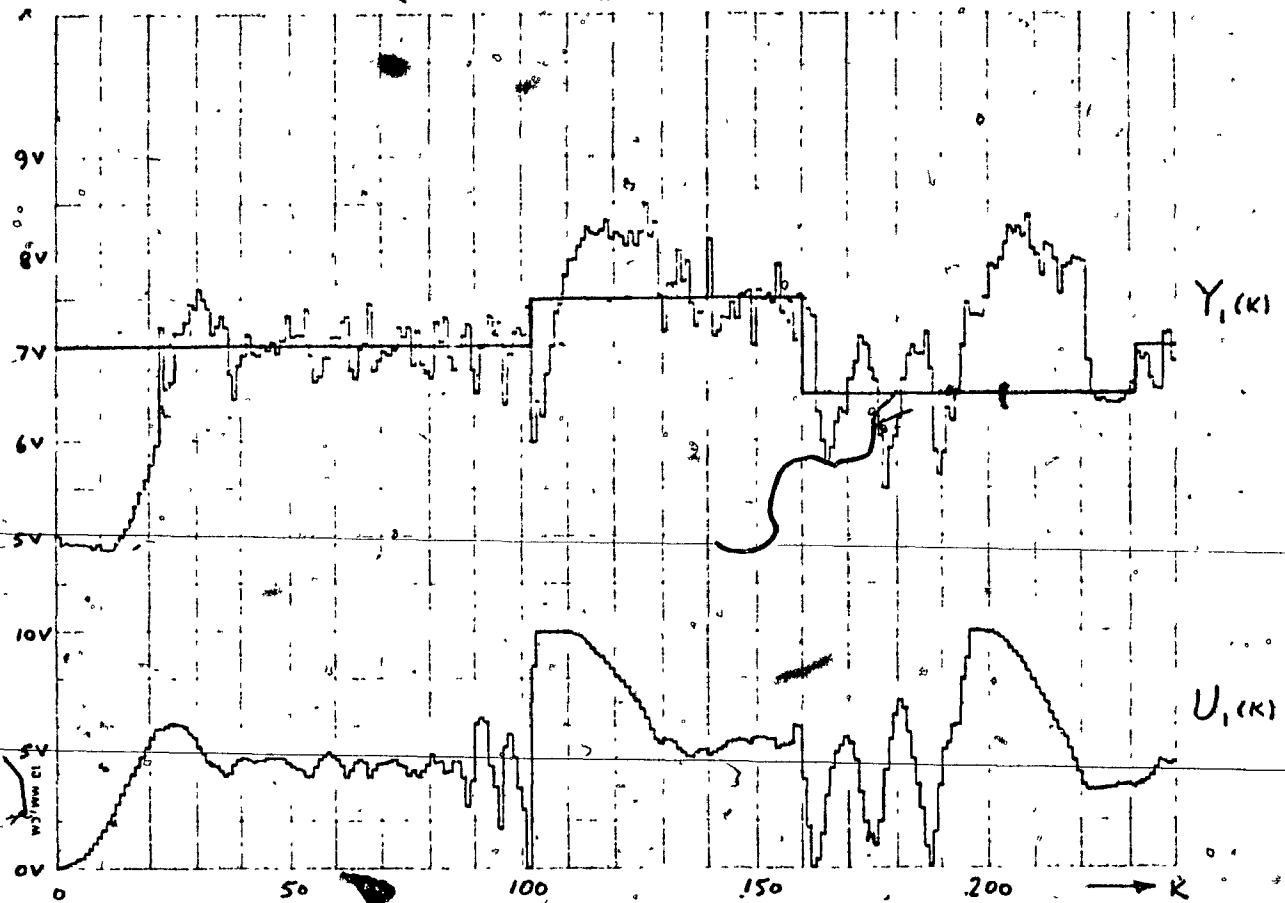


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 16

FIGURE 4.25a



SELF-TUNING CONTROL

SECOND ORDER (PI)

RUN # 17

 $T = 14$ seconds $b_1(0) = 64$

FIGURE 4.25a (Continued)

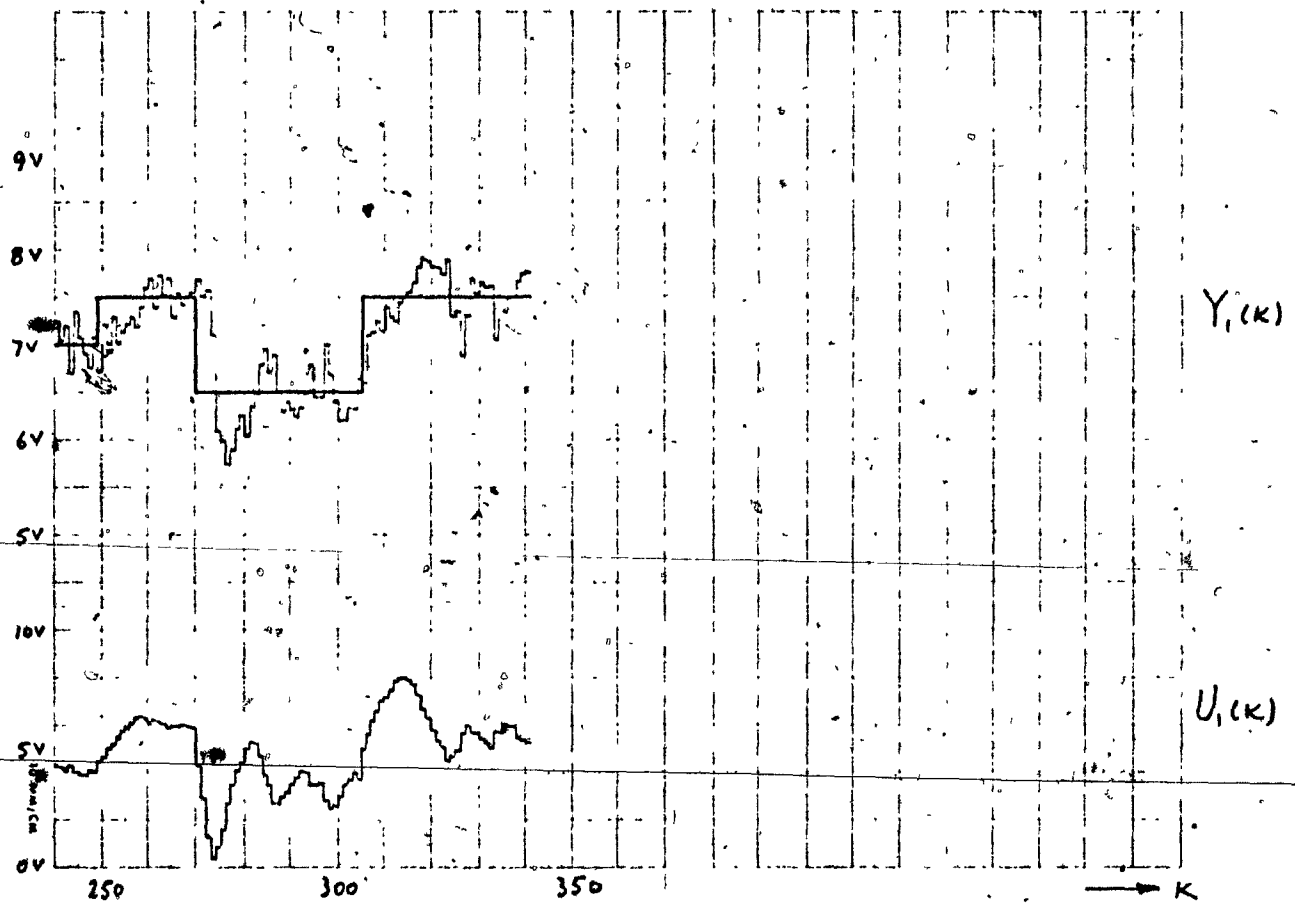
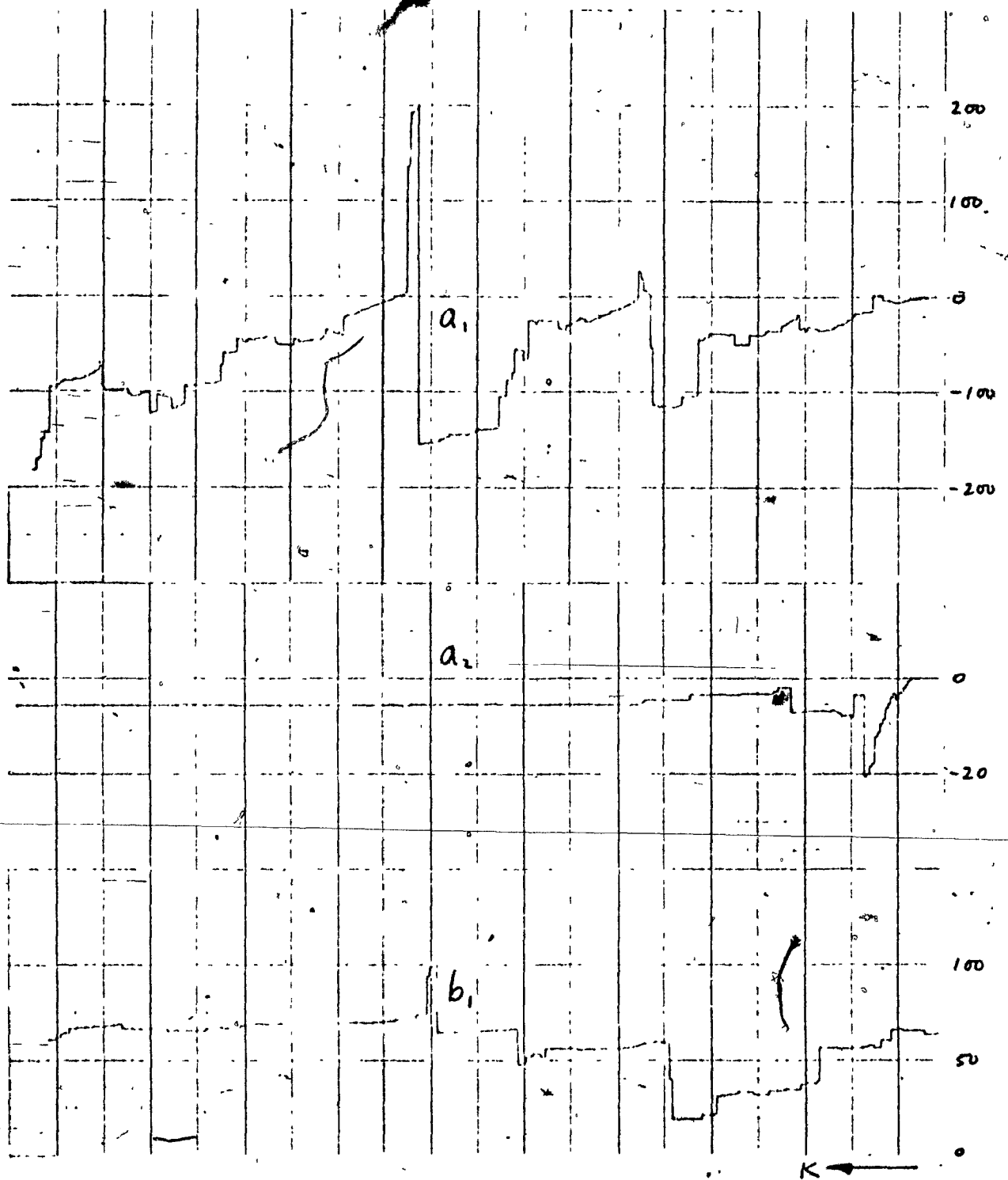


FIGURE 4.25b

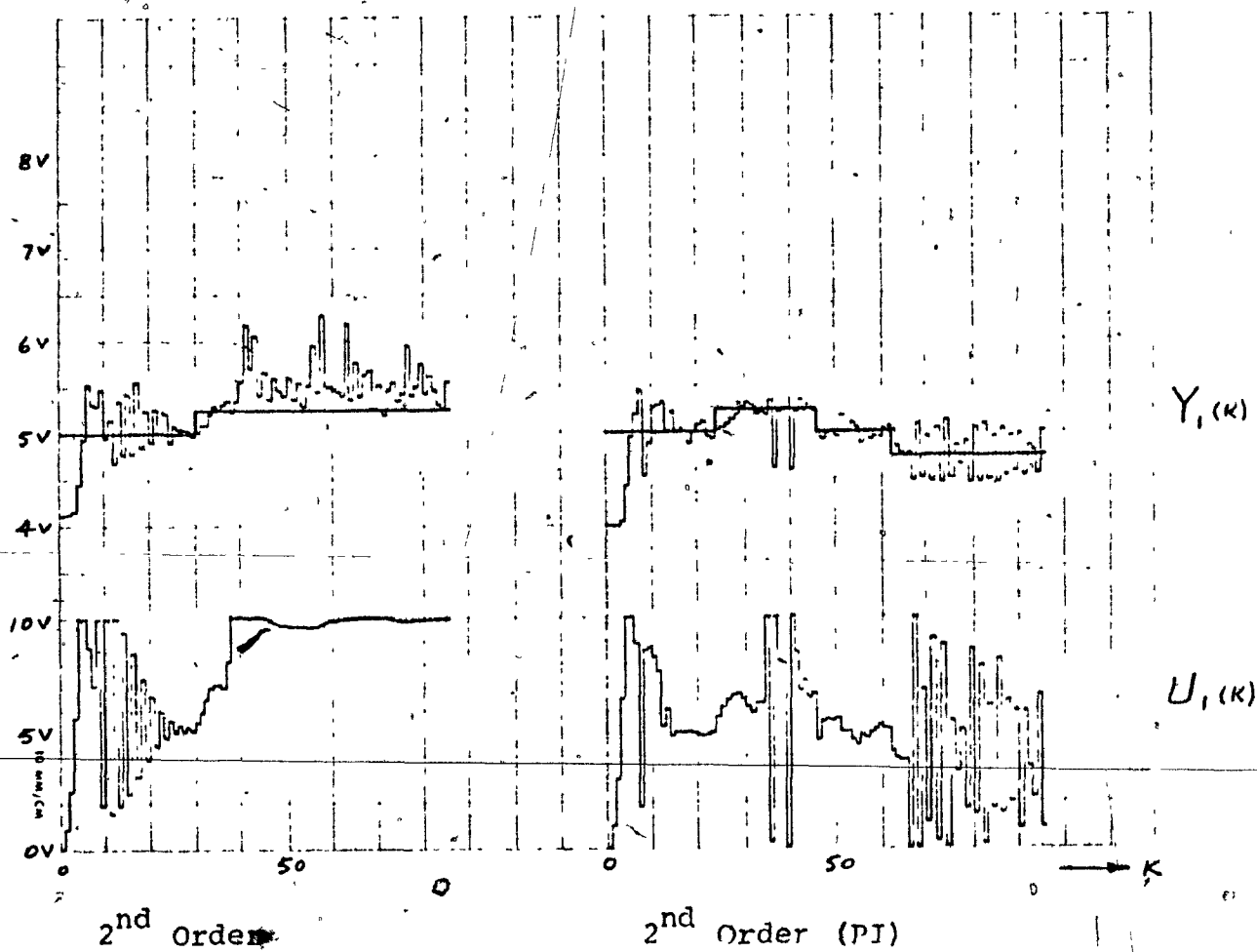


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 17

FIGURE 4.26a



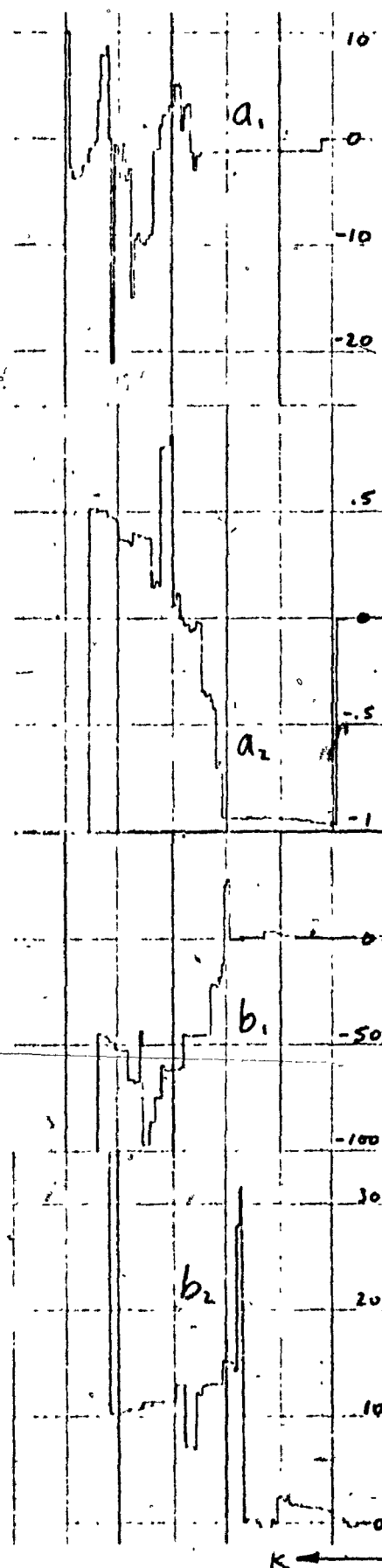
SELF-TUNING CONTROL

SECOND ORDER

RUN # 18

 $T = 42$ seconds $b_1(0) = .1$ $\alpha = .64$ volts²

FIGURE 4.26b

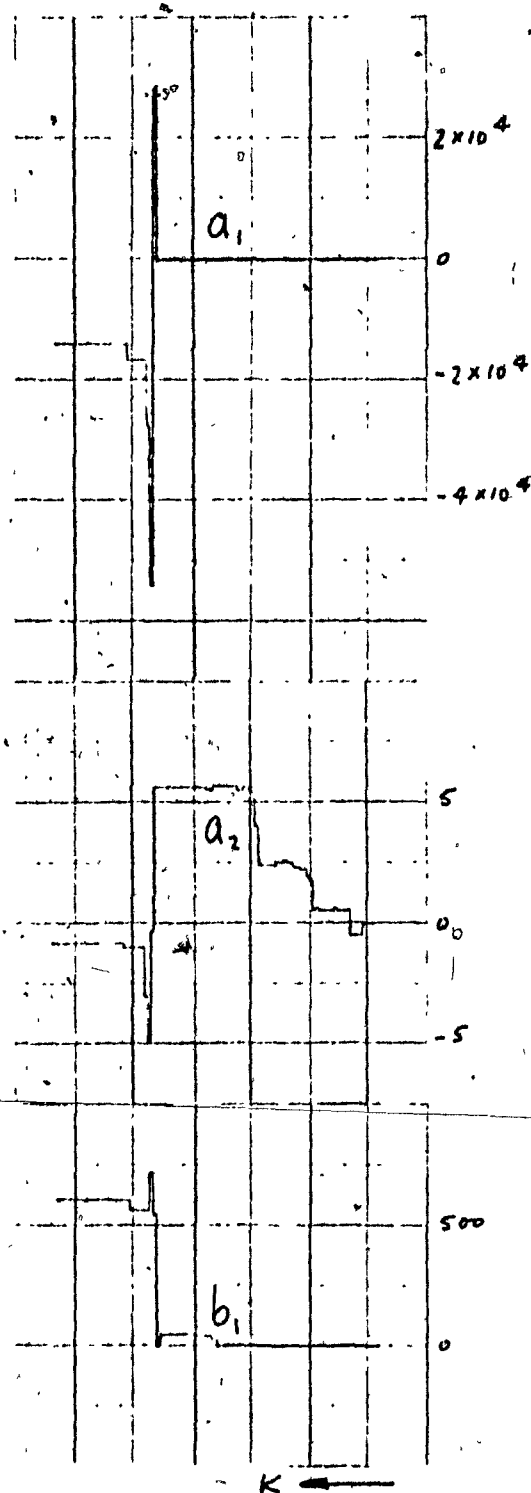


SECOND ORDER

PARAMETER ESTIMATES

RUN # 18

FIGURE 4.26c

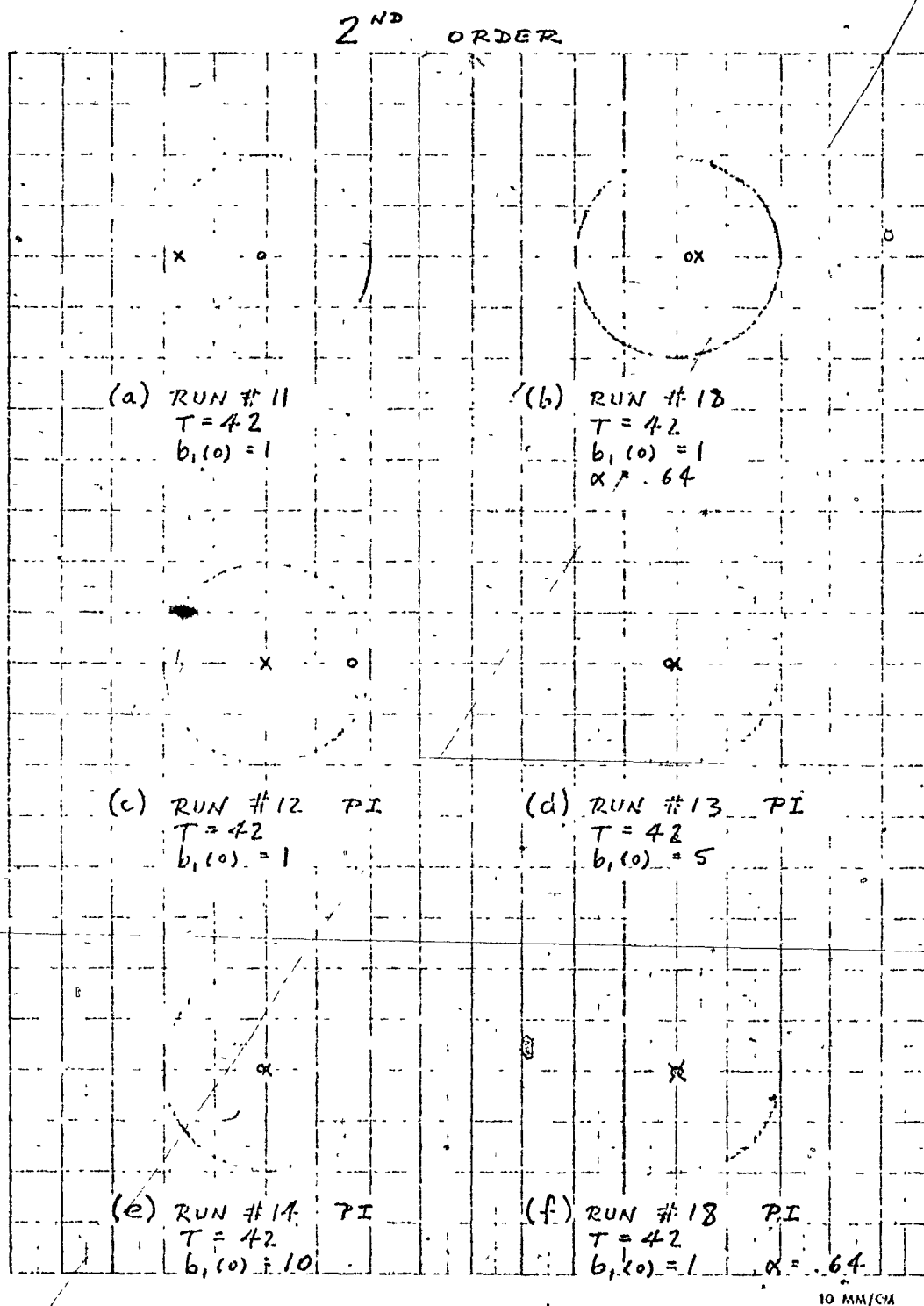


PARAMETER ESTIMATES

SECOND ORDER (PI)

RUN # 18

FIGURE 4.27



POLE-ZERO PATTERN OF MINIMUM VARIANCE CONTROLLER
 OBTAINED FROM IDENTIFICATION ALGORITHM

FIGURE 4.27 (Continued)

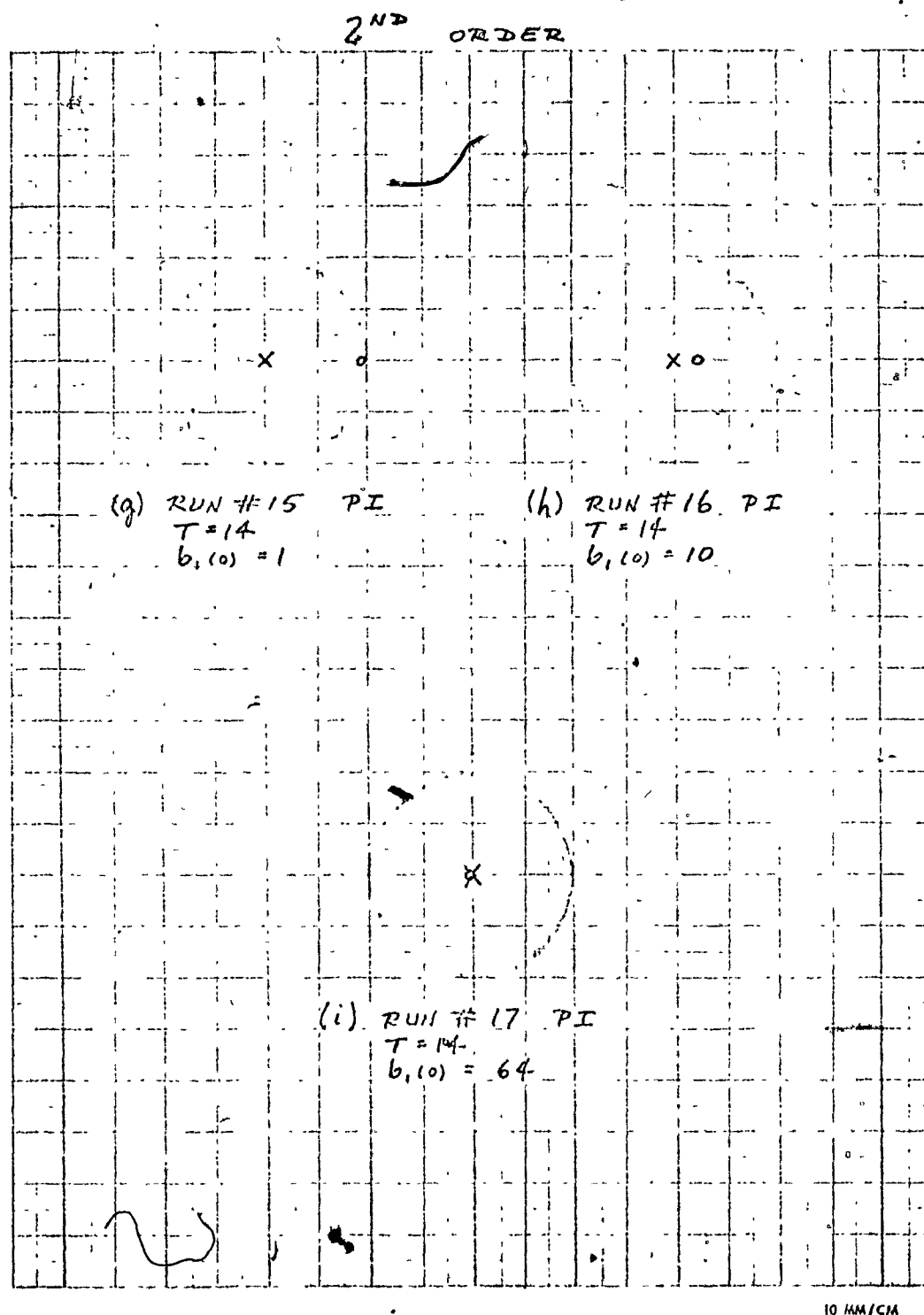
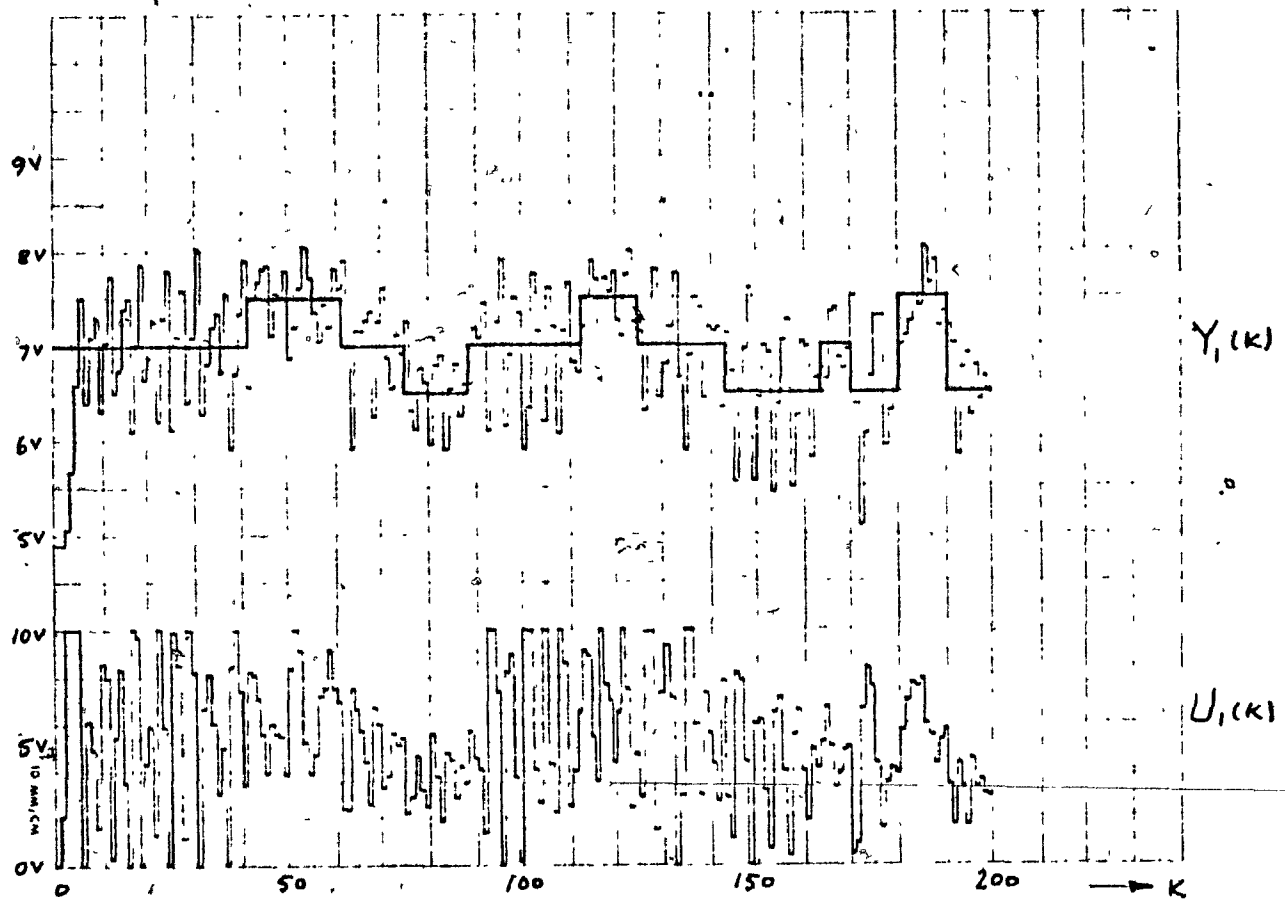


FIGURE 4.28a



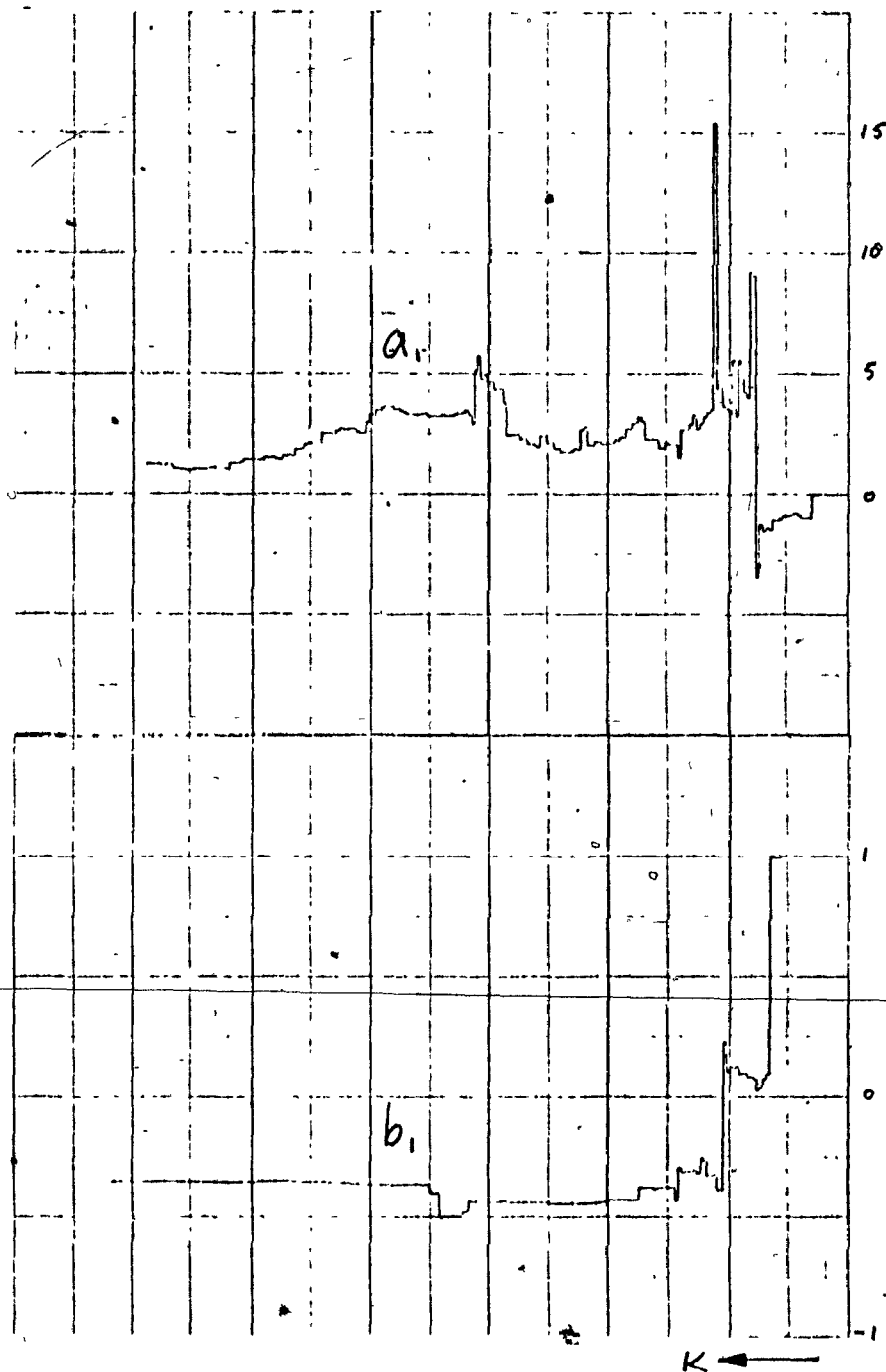
SELF-TUNING CONTROLLER

FIRST ORDER

RUN # 19

 $T = 42 \text{ seconds}$ $b_1(0) = 1$

FIGURE 4.28b

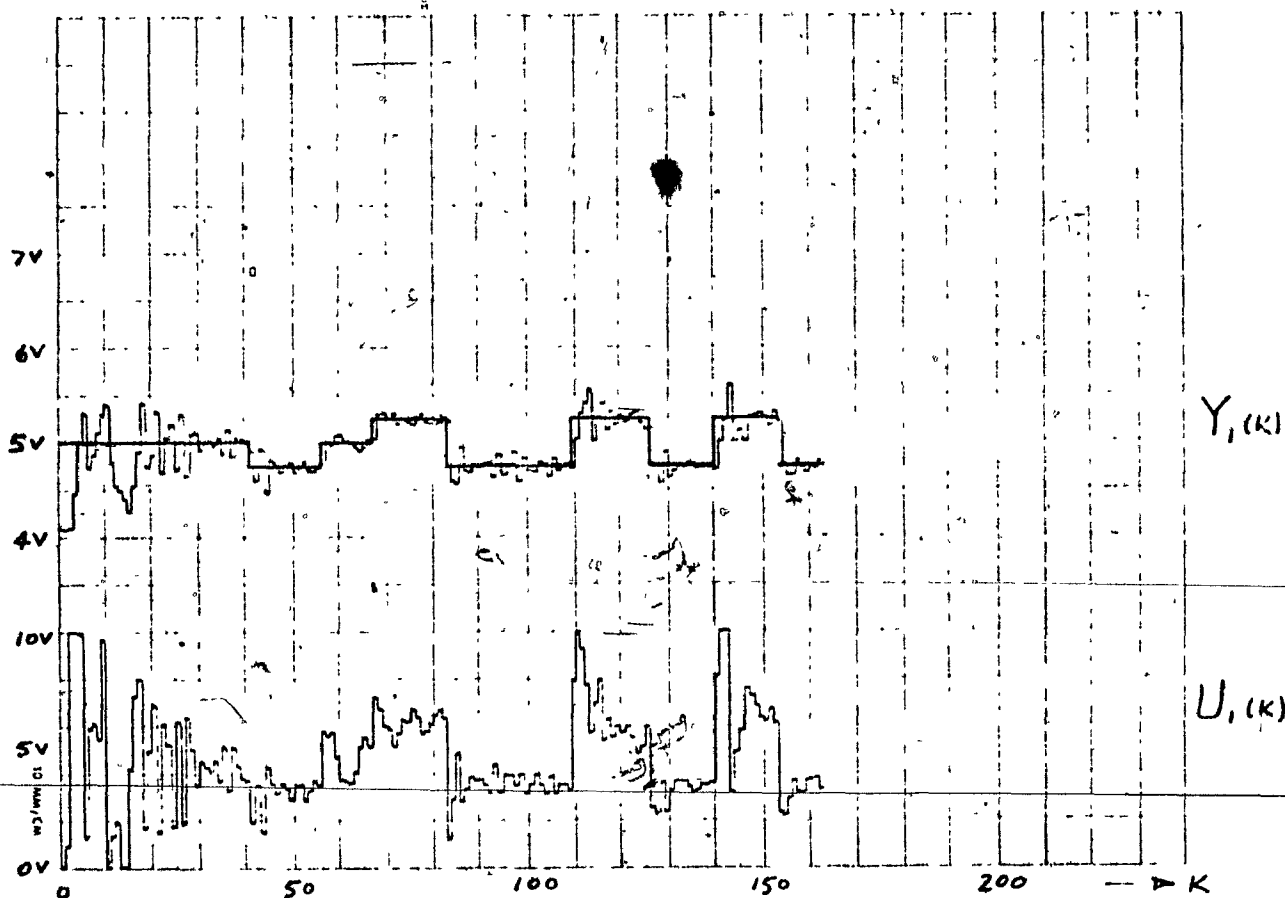


PARAMETER ESTIMATES

FIRST ORDER

RUN # 19

FIGURE 4.29a



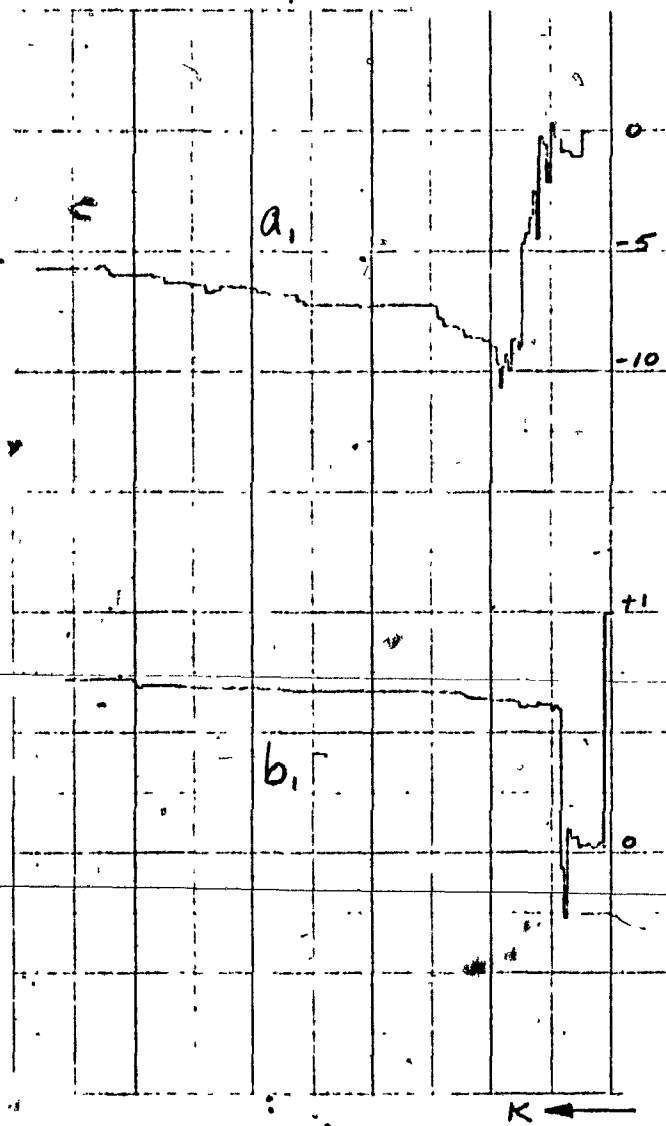
SELF-TUNING CONTROL

FIRST ORDER

RUN # 20

 $T = 42$ seconds $b_1(0) = 1$ $\alpha = .64 \text{ volts}^2$

FIGURE 4.29b

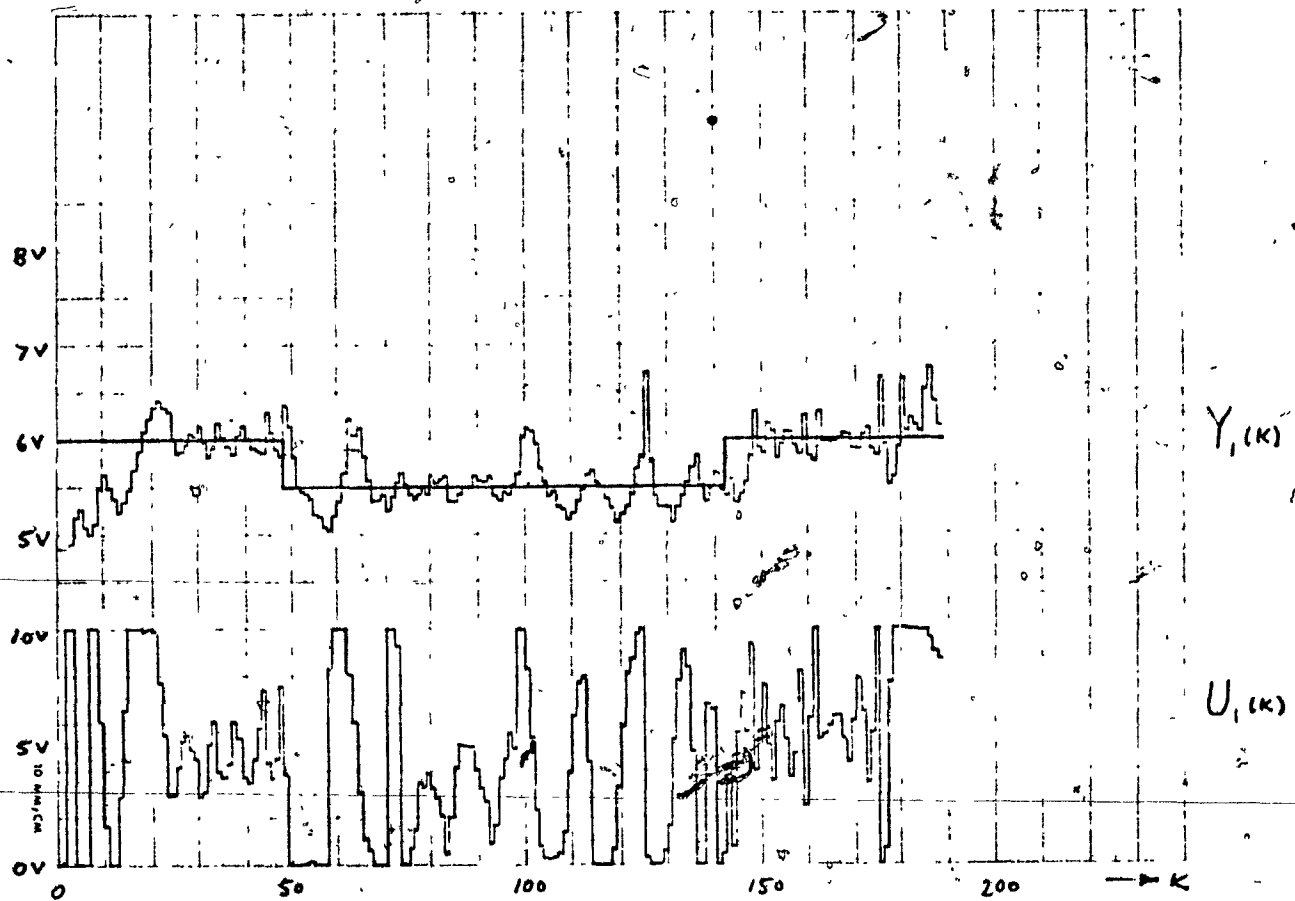


PARAMETER ESTIMATES

FIRST ORDER

RUN # 20

FIGURE 4.30a



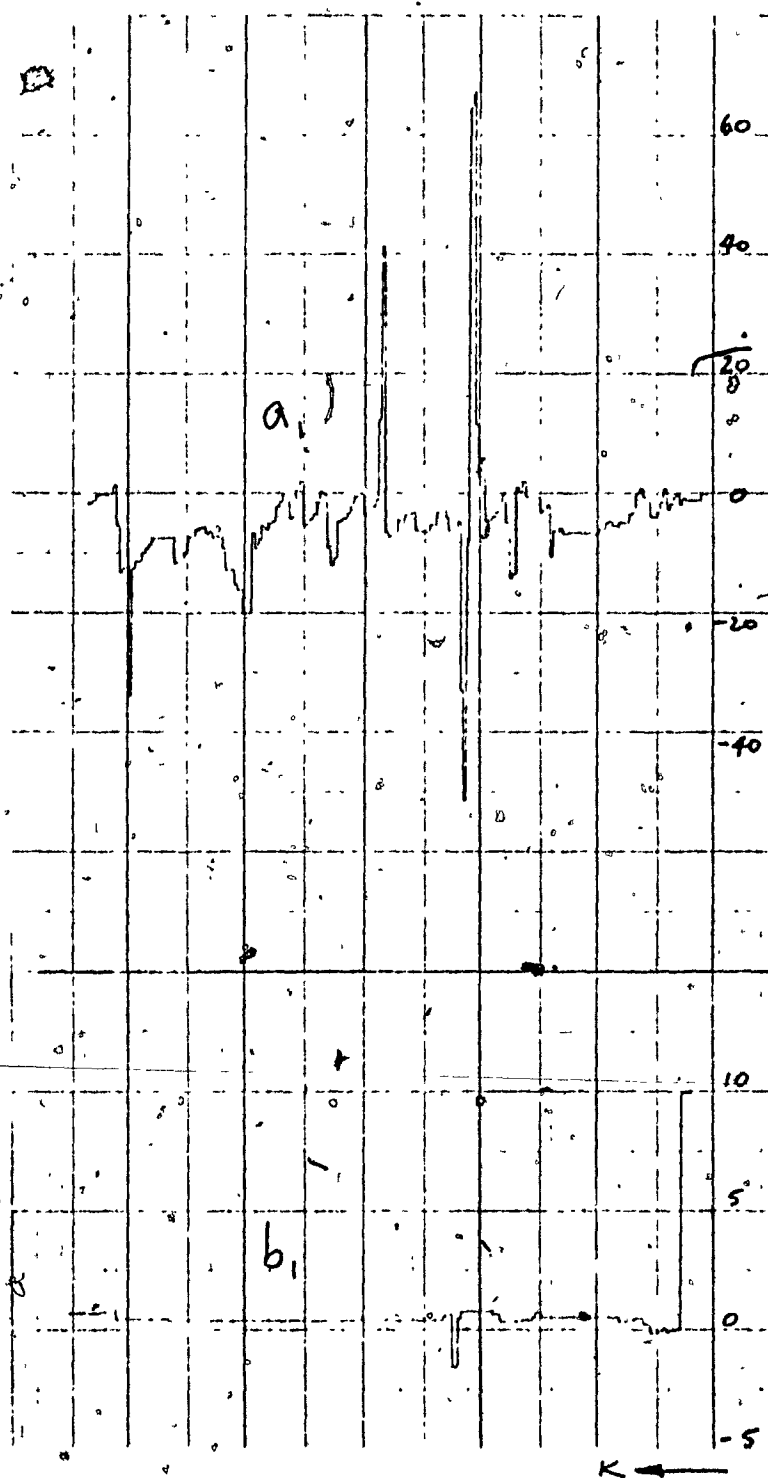
SELF-TUNING CONTROL

FIRST ORDER

RUN # 21

 $T = 14$ seconds $b_1(0) = 10$

FIGURE 4.30b

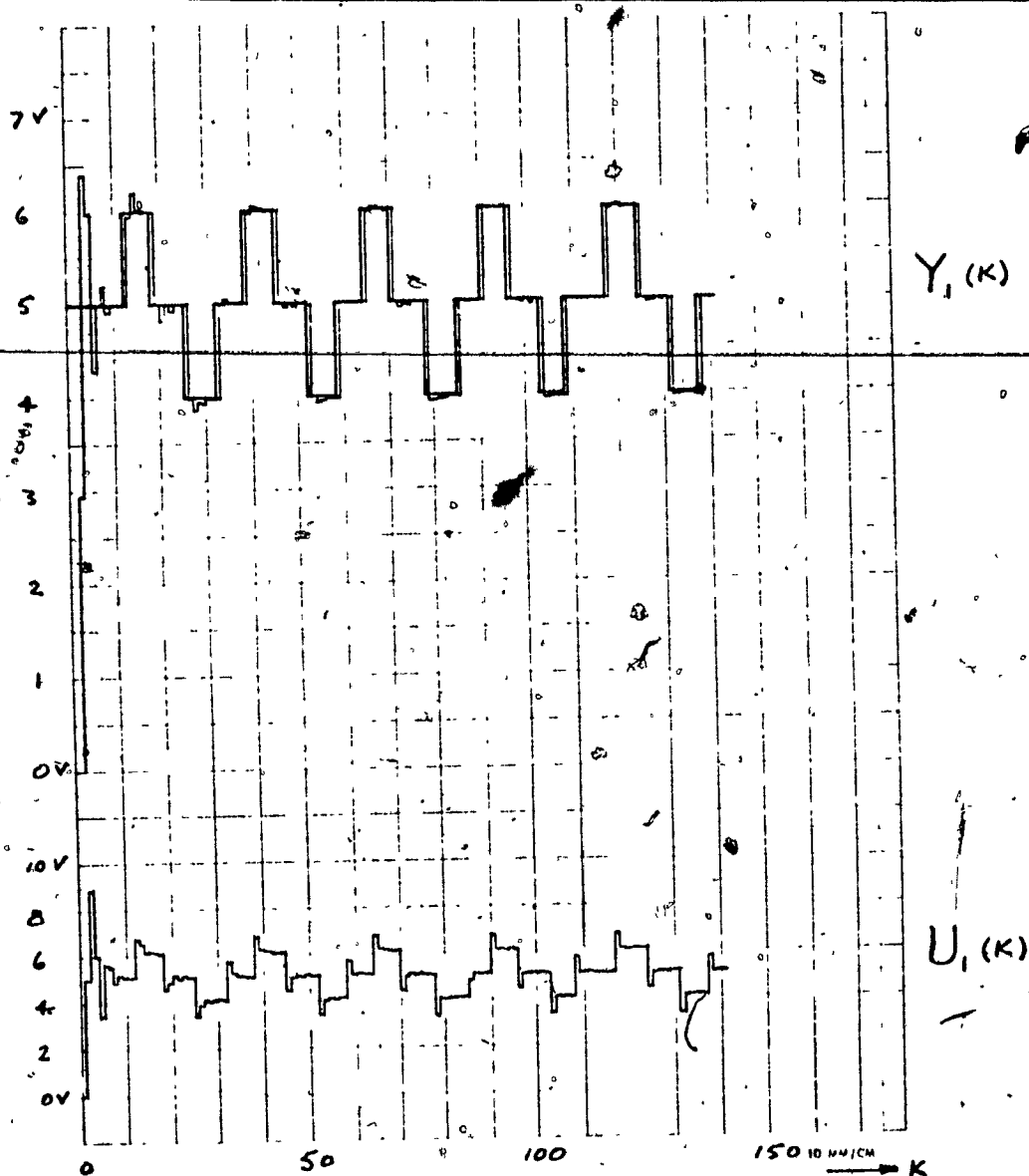


PARAMETER ESTIMATES

FIRST ORDER

RUN # 21

FIGURE 4.31



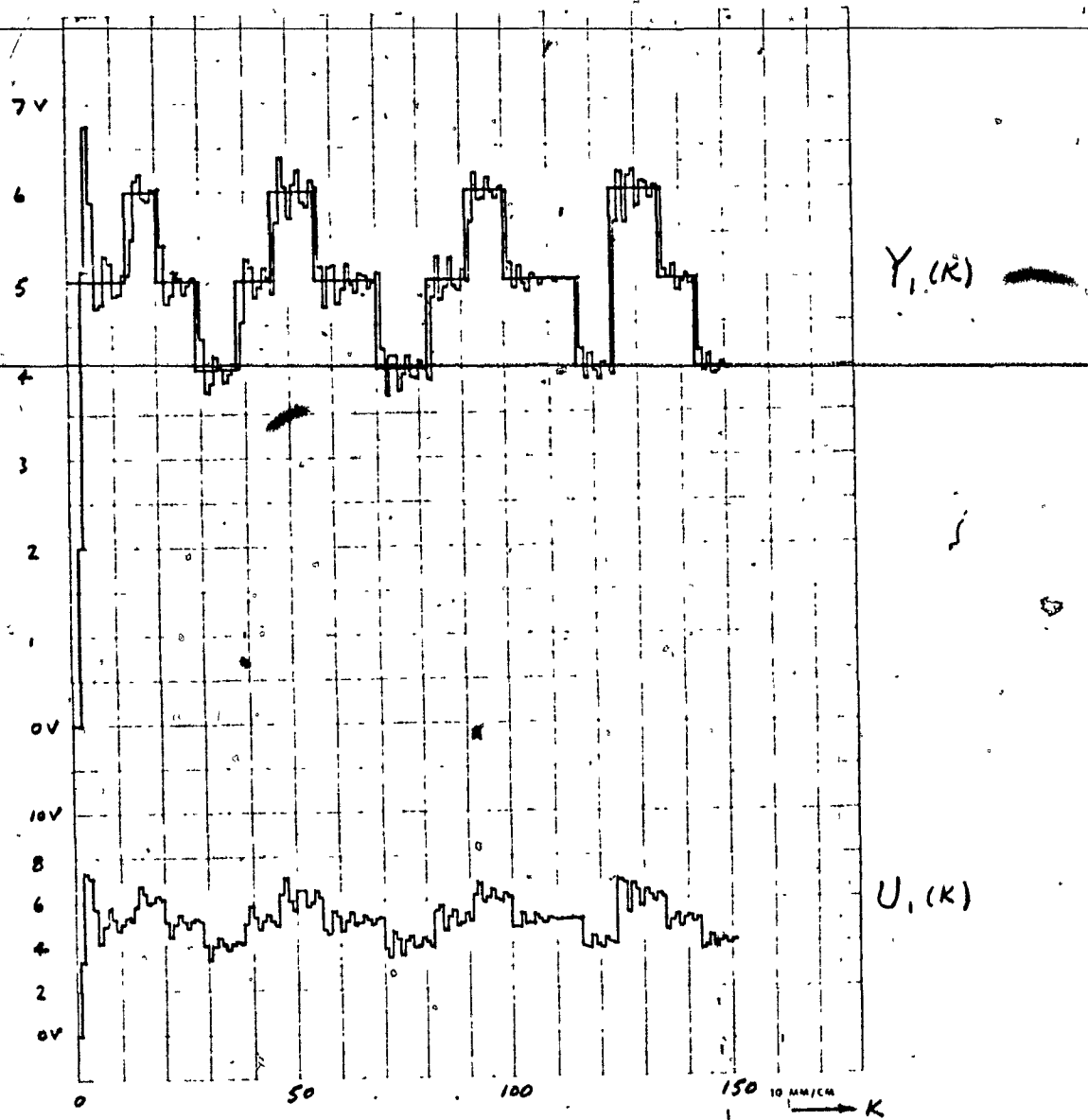
SELF-TUNING CONTROL

FOURTH ORDER

SIMULATED PLANT

 $T = 42 \text{ seconds}$ $b_1(0) = 1$ $\alpha = 2.5 \times 10^{-5} \text{ volt}^2$

FIGURE 4.32



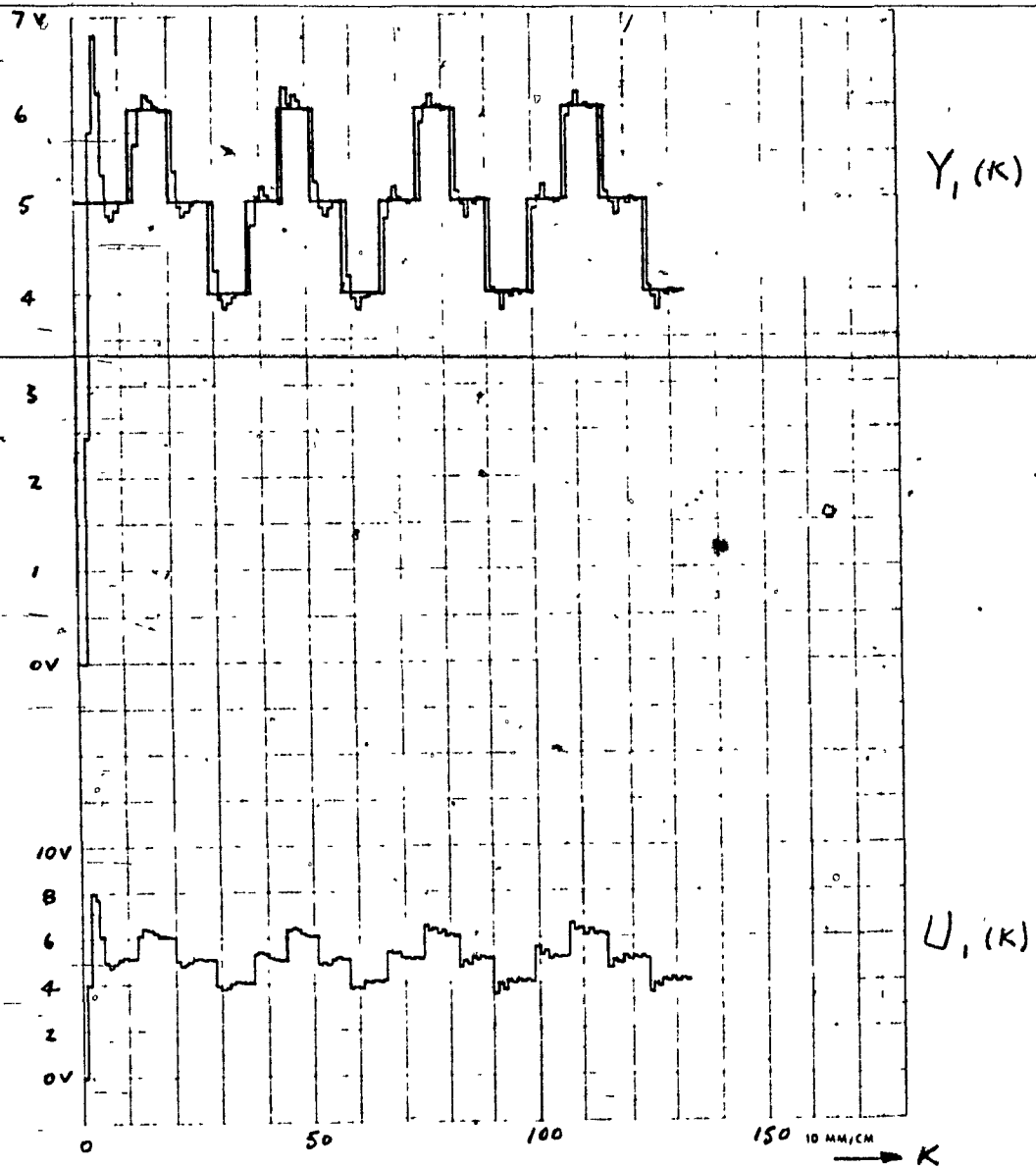
SELF-TUNING CONTROL

THIRD ORDER

SIMULATED PLANT

 $T = 42 \text{ seconds}$ $b_1(0) = 1$ $\alpha = 13.1 \text{ volt}^2$

FIGURE 4.33



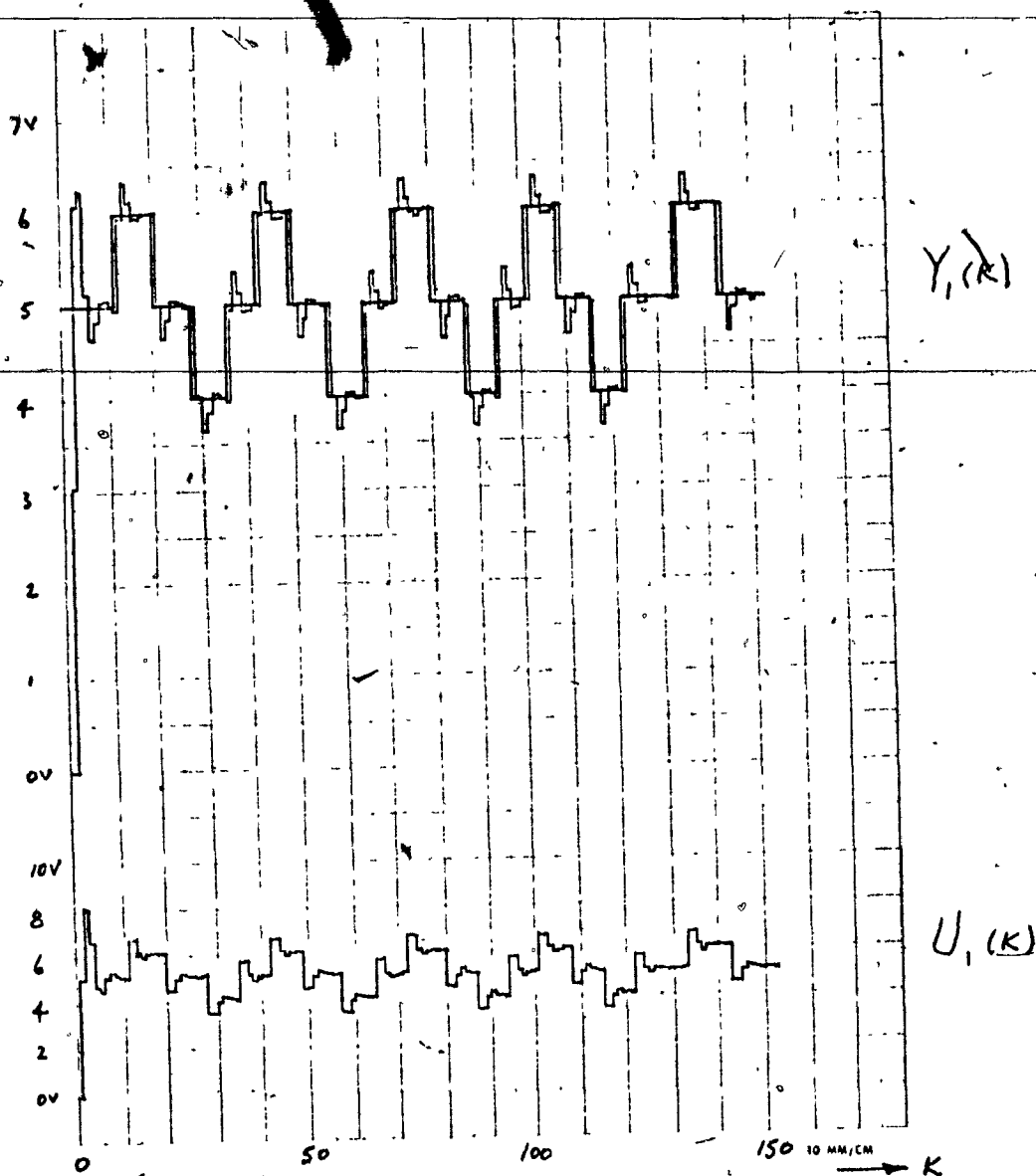
SELF-TUNING CONTROL

SECOND ORDER

SIMULATED PLANT

 $T = 42 \text{ seconds}$ $b_1(0) = 1$ $\alpha = 6.55 \text{ volt}^2$

FIGURE 4.34



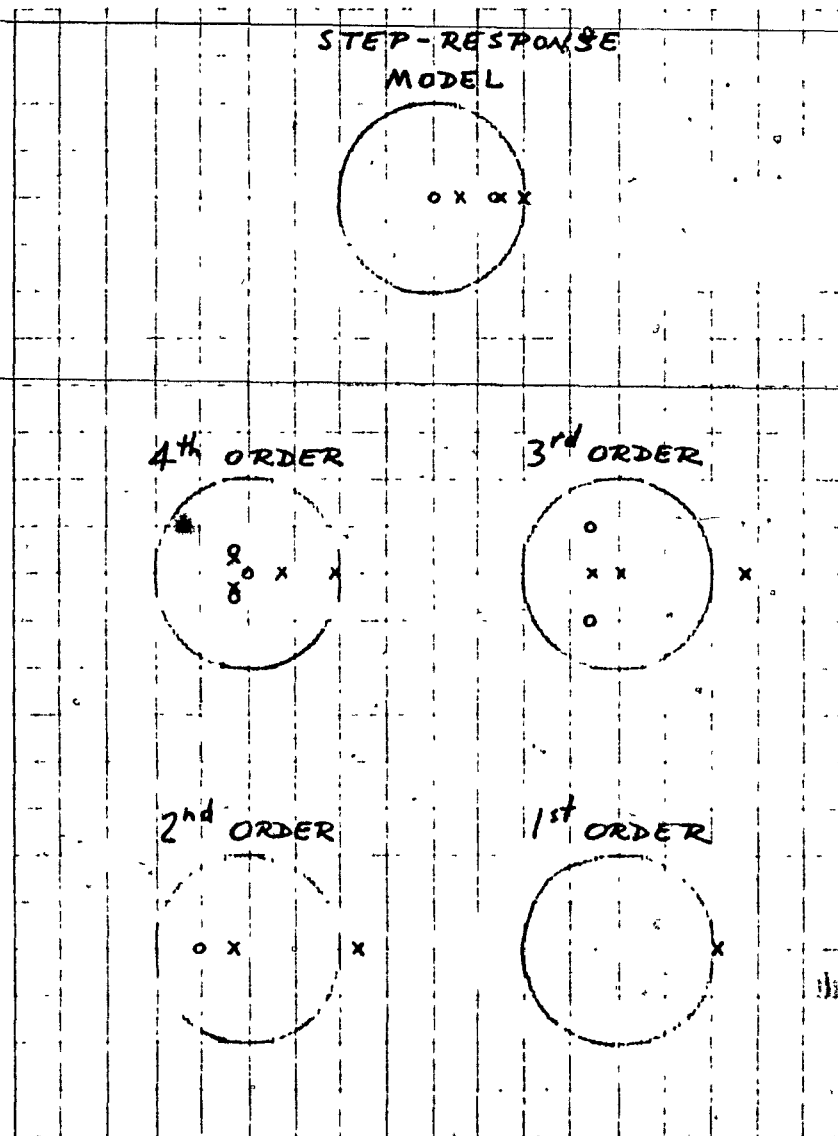
SELF-TUNING CONTROL

FIRST ORDER

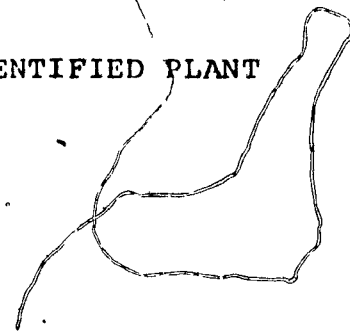
SIMULATED PLANT

 $T = 42 \text{ seconds}$ $b_1(0) = 1$ $\alpha = 2.5 \times 10^{-5} \text{ volt}^2$

FIGURE 4.35



POLE-ZERO PATTERN OF IDENTIFIED PLANT



CHAPTER 5

CONCLUSION

5.1 Summary

The heat exchanger was viewed as a single-input-single-output system. The input was a steam-flow set-point signal and the output was the outlet water temperature. The control algorithm used was a combination of a recursive least-squares parameter estimation using a linear model and a minimum variance regulator tuned on-line according to the estimated parameters. In a noise-free steady-state the input and output of the minimum variance regulator are zero. Thus the input to the regulator is required to be a temperature error signal, and the regulator output must be integrated to obtain a set-point signal for the steam flow controller. Hence the plant is augmented within the computer program by an integrator following the output of the regulator. To the identification algorithm, this integrator is part of the plant.

Convergence of the parameter identification cannot be guaranteed because of correlated noise (model equation errors) and because of feedback. Correlated noise leads to biased parameter estimates. The consequence of feedback is non-identifiability beyond specification of a linear manifold on which the parameter estimates lie. Nevertheless, if the parameter estimates converge, a unique minimum variance

controller can be expected.

For comparison purposes a model of the plant was obtained from a step response test. The model was valid only in a mid-temperature region. The heat exchanger appeared to be a combination of two lags with a minor delay. The model was augmented to include the digital integrator. A sampling interval of .7 minutes was chosen for most of the experiments so that all the parameters of the correct model would be significant. If not all the model parameters were significant, then the order of the model would be incorrect. An autocorrelation of the steady-state temperature was performed in order to obtain a noise variance estimate required for the identification algorithm.

Using a fourth order model, the least-squares minimum variance control algorithm produced a stable system and converging parameter estimates. The value of the noise variance parameter α did not appear to affect the results. In general, the pole at $+1$ due to the integrator and the pole corresponding to the dominant time constant of the heat exchanger were approximately correctly identified. Also produced were a pair of complex poles that nearly cancelled a pair of complex zeros. This suggested that a second order model would be more appropriate. (Experiments with second order models did not substantiate this hypothesis however.) In the fourth order experiments, the problem of non-

identifiability due to feedback did not make itself apparent.

Experiments with lower-order models met with less success however. The parameter estimates made large excursions, not always along a linear manifold. Stability became sensitive to the value of α and to the starting value of b_1 . Models leading to PI and PID controllers were even less successful. In the case of the PID controller (third order model) three poles of the plant are to be identified, while two zeros are fixed at the origin. The problem is that only one zero should actually be at the origin and the second cannot be properly adjusted. The first order model produced a simple integral controller. This turned out to have better stability and convergence properties than obtained from second and third order models.

Further tests on an analog computer simulation of the heat exchanger (with no noise) showed that the parameter α affects the algorithm much more than was previously apparent. The first and fourth order identifications converged for a wide range of α , including extremely small values. The second and third order identifications also converged, but only for relatively large values of α .

Overall, the experiment is a successful

application of the self-tuning control algorithm.

The question remains, however, of why α was important in the second and third order identifications, and unimportant in the first and fourth order identifications.

5.2 Conclusion

The experiment has been a successful application of the least-squares minimum-variance self-tuning control algorithm. On the whole, the result of the experiment is regarded as a further recommendation of this algorithm for use in industry. It has shown itself to be a robust algorithm, especially if the model order is high enough. In all cases, the algorithm used no a priori knowledge of the parameters. Also, parameter b_1 was not fixed. This opened the possibility of non-identifiability. Yet, for proper choices of α , the algorithm converged smoothly for all the model orders from first to fourth. Convergence was not slow either, with moderately good control being obtained within 20 or 30 steps.

A question remains, however, on the surprising effect α has on the algorithm. Parameter α is a noise variance parameter. Considering its location in the recursive formula of Equation 3.3, it can be viewed as a measure of the algorithm's immunity to noise. Giving a

large value to α makes the algorithm more immune to noise, but slows down the convergence. Reducing α speeds up convergence (by increasing gain Γ) but thereby makes the algorithm more susceptible to noise. It could be expected that too small a value for α would lead to diverging identification. Presumably, the best compromise is to set α equal to the actual plant noise variance. The question, then, is why does the fourth order identification converge in the stochastic environment of the heat exchanger with α three orders of magnitude less than

the actual measured noise; this while the second and third order identifications did not converge in the deterministic environment of the analog simulation until α corresponded to an rms noise value of about one third of the entire output range.

Since no such peculiarity is reported in the experiences of [18, 19], it is possible that it is a result here of the lack of the stabilizing effect of fixing the value of b_1 . In any case it is an interesting question whose answer would probably be very instructive about the general behaviour of the algorithm. It is suggested that further study of the algorithm, either experimental or theoretical, in a deterministic environment may illuminate the nature of the effect of α . The

reason for a deterministic study is that the least-squares identification scheme is essentially the solution of a deterministic optimization problem. If there is anything interesting about α of fundamental nature to be learned, a stochastic environment would tend to mask it.

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APPENDIX I

System Wiring Diagrams

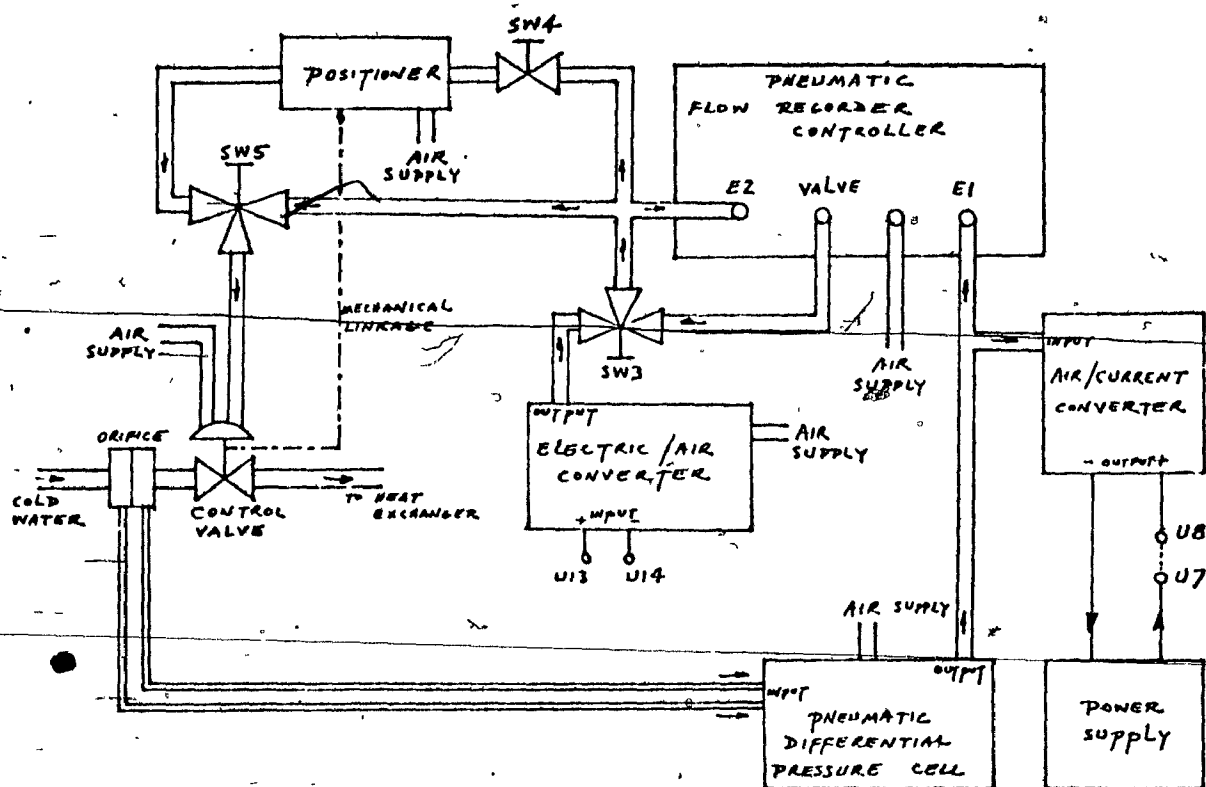
The illustrations in this Appendix are intended to provide more detail of the water flow control and water temperature control systems. The main purpose of this is to record the wiring of the heat-exchanger/computer-station link for purposes of future experimentation.

Figure Ia shows the water flow control system. The arrows indicate the direction of fluid and current flow. Terminals U13, U14 and U7, U8 are the input/output terminals for direct digital control (DDC). These terminals are located on the upper panel of terminals at the heat exchanger. By using the table in Figure Ic, the corresponding terminals at the computer station can be found.

Figure Ib shows the water temperature control system. The cold and hot water temperature signals are available at terminals U1, U2 and U3, U4 respectively. The steam flow signal appears at U11, U12. These signals are 10 to 50 ma currents and hence the terminal pairs must close the circuit. They may be shorted if the signal is not being used, or they may be bridged at the computer station by a resistor to produce a voltage signal. Terminals U15, U16 provide a direct digital control signal which can be engaged by setting switch SW1.

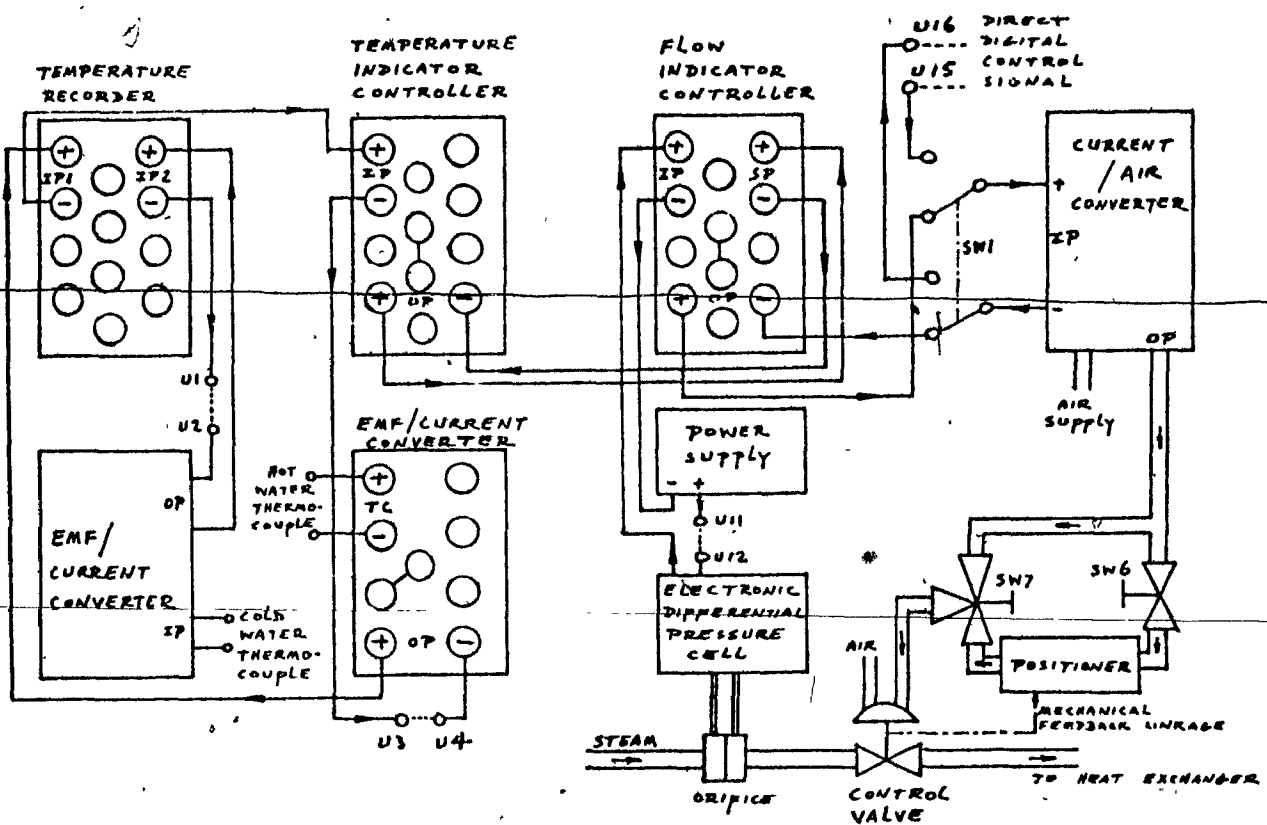
Switch SW2 is not shown here because it does not presently exist as a piece of hardware. The system is shown with the analog temperature controller feeding the set-point signal to the SP terminals of the steam flow controller. To effect the switch to computer set-point control, the temperature controller output is disconnected from the flow controller set-point and terminated by a 100 Ω resistor. The set-point terminals of the flow controller are then connected to terminals U5, U6 which carry the required set-point signal from the computer station.

FIGURE 1a



WATER FLOW CONTROL SYSTEM

FIGURE 1b



WATER TEMPERATURE CONTROL SYSTEM

FIGURE I c

TABLE OF HEAT-EXCHANGER/COMPUTER-STATION TERMINAL CONNECTION

ROOM 502
McCONNELL ENG. BLDG.

HEAT EXCHANGER	PANEL #	TERMINAL #	WRAPPER COLOUR	WIRE COLOUR	SIGNAL	PANEL #	TERMINAL #
U		1	GREEN	BLACK	+ COLD WATER	D	3
		2		WHITE	- TEMPERATURE MEASUREMENT		4
U		3	GREEN	YELLOW	+ HOT WATER	D	5
		4		BLACK	- TEMPERATURE MEASUREMENT		6
U		5	RED	RED	+ TEMPERATURE CONTROLLER	D	7
		6		BLACK	- SET-POINT COMMAND		8
U		7	RED	BLUE	+ WATER FLOW	D	9
		8		BLACK	- MEASUREMENT		10
U		9	BLUE	YELLOW	NOT USED	D	11
		10		RED			12
U		11	BLUE	RED	+ STEAM FLOW	C	1
		12		GREEN	- MEASUREMENT		2
U		13	BLUE	WHITE	+ WATER FLOW COMMAND	C	3
		14		RED	- (DIRECT DIGITAL CONTROL)		4
U		15	BLUE	BLUE	- STEAM FLOW COMMAND	C	5
		16		RED	+ (DDC AT SW1)		6
L		1	BLUE	GREEN	NOT USED	C	7
		2		BLACK			8
L		3	BLUE	ORANGE	NOT USED	C	9
		4		BLACK			10
L		5	BLUE	BROWN	NOT USED	C	11
		6		BLACK			12

APPENDIX II

Control Program Listing
(In Assembler Language [21,22])

/CONTROL PROGRAM

ADCL=6530
ADLN=6531
ADLE=6536
ADST=6532
ADSH=6534
ADSD=6533
DASH=6442
DALA=6443
DALD=6444
DAST=6446
DASI=6441

		*1
0001	5402	JMP I 2
0002	0200	SENV
		*5/
0005	7400	7400
0006	7200	7200
0007	5000	5000
		*11
0011	0012	H13, 13
0012	0001	H1024, 1
0013	2530	2530
0014	4460	4460
		*63
0003	1000	TTY, TTY
0004	1000	READ, READ
0005	1010	PRINT, PRINT
0006	1015	LF, LF
0007	1000	CR, CR
0070	1027	P, P
0071	1061	O, O
0072	1071	T, T
0073	1073	S, S
0074	0043	SET, SET
0075	0700	CLEAR, CLEAR
0076	1400	CLOCK, CLOCK
0077	1000	

0100	1644	XAD,	AD
0101	2127	XDA,	DA
0102	0554	XU1,	554
0103	0697	XU2,	697
0104	0570	XY1,	570
0105	0623	XY2,	623
0106	0600	XCOEF,	430
0107	7777	XX1,	-1
0110	7775	XX3,	-3
0111	0000	COUNT,	0
0112	0000	HUM,	0
0113	0013	U1,	13
0114	0000		0
0115	0000		0
0116	0013	U2,	13
0117	0000		0
0120	0000		0
0121	0013	V1,	13
0122	0000		0
0123	0000		0
0124	0013	V2,	13
0125	0000		0
0126	0000		0
0127	0013	Z1,	13
0130	0000		0
0131	0000		0
0132	0013	Z2,	13
0133	0000		0
0134	0000		0
0135	0000	E1,	0
0136	0000	E2,	0
0137	0000	EXIT,	EXIT
0140	0000	XDAO,	DAO
0141	0000	XERR,	ERROR
0142	0047	XFIN,	FIN
0143	0000	HALT,	0
0144	7433	LSTRT,	-345
0145	0000	LCOUNT,	0
0146	0000	IRSTRT,	0
0147	0000	RCOUNT,	0
0150	0003	W3,	3
			/
			/INTERRUPT SERVICING
			/
			W200
0200	0000	SERV,	DCA AC
0201	7204		GLK
0202	0000		DCA LINK
0203	0031		MSF
0204	7410		SNP
0205	5700		JMP I MINE
0206	0000		ATCH
0207	7410		SNP
0208	5707		JMP I MINE

/KEYBOARD

/A/D

0211	0041		TST
0212	7410		SNP
0213	5730		JMP I MITP /PRINTER
0214	0021		RST
0215	7410		SNP
0216	5731		JMP I MIPP /PUNCH
0217	6007		CAF
0200	7300	EXIT,	CLA CLL /RETURN FROM INTERRUPT SERVICING
0201	1306		TAD LINK
0202	7010		RAR
0203	1325		TAD AC
0204	6001		ION
0205	5400		JMP I 0
0206	7300	ERROR,	CLA CLL
0207	1045		TAD 45
0208	7510		SPA
0231	5236		JMP.+5
0232	7300		CLA CLL
0203	7040		CMA
0234	7010		RAR
0205	5705		JMP I 71
0206	7000		CLA CLL
0207	7000		CML
0240	7010		RAR
0241	7001		LAC
0242	5705		JMP I 71
0243	4404	SUB,	JMS I HEAD
0244	1700		TAD I HEAD
0245	7510		SPA
0246	5471		JMP I NO
0247	0111		DCA COUNT
0250	1111		TAD COUNT
0251	1107		TAD HHI
0252	7540		SMA SZA
0253	5471		JMP I NO
0254	7000		CLA CLL
0255	1111		TAD COUNT
0256	7004		TAL
0257	1111		TAD COUNT
0200	1334		TAD ADZ
0261	3111		DCA COUNT
0262	4477		JMS I XEP
0263	4405		JMS I 5
0264	4407		JMS I 7
0265	3306		TRM K00407
0266	0000		FEXT
0267	4542		JMS I XFIN
0270	3511		DCA I COUNT
0271	5463		JMP I PTTY
0272	0000	HEAD,	
0273	7300		CLA CLL
0274	1040		TAD HEAD
0275	0000		DCA I
0276	7000		CLA

0077	1410		TAD I 10
0300	7004		RAL
0301	7430		SZL
0302	5670		JMP I HEADER
0303	7510		RAR
0304	7440		SZA
0305	5314		JMP.+7
0306	1341		TAD KSP
0307	4405		JMS I NFPINT
0308	4465		JMS I NFPINT
0309	4465		JMS I NFPINT
0310	4465		JMS I NFPINT
0311	4465		JMS I NFPINT
0312	4465		JMS I NFPINT
0313	4465		JMS I NFPINT
0314	4465		JMS I NFPINT
0315	5276		JMP HEADER+4
0316	0000	CLDA,	C
0317	3724		DCA I XDACH
0320	4501		JMS I XDA
0321	2724		ISC I XDACH
0322	4501		JMS I XDA
0323	5710		JMP I CLDA
0324	0145	XDACH,	DACH
0325	0000	AC,	C,
0326	0000	LINK,	C,
0327	1000	LINK,	IAD
0328	1633	ITP,	ITP
0331	1660	ITP,	IPP
0332	1003	LINK,	LINK
0333	1162	LINK,	LINK
0334	0120	ADZ,	Z1+1
0335	2075	71,	DONE+1
0336	0010	X204P7,	C
0337	0145		0145
0340	4600		4600
0341	0040	KSP,	240
0342	0340	HEAD,	
0343	0240		040
0344	0240		040
0345	0332		332
0346	0261		061
0347	0000		0
0350	0331		331
0351	0261		061
0352	0000		0
0353	0305		305
0354	0261		061
0355	0000		0
0356	0325		325
0357	0261		061
0360	0000		0
0361	0332		332
0362	0262		062
0363	0000		0
0364	0001		001

0365	0000	262
0366	0000	262
0367	0305	305
0370	0262	262
0371	0000	0
0372	0305	305
0373	0000	262
0374	0215	215
0375	0212	212
0376	4000	4000
/		
/SYSTEM INITIALIZATION		
/		
*700		
0700	0000	CLEAR,
0701	7000	0
0702	7000	NOP
0703	1300	CLA
0704	3111	TAD IN242
0705	1100	DCA COUNT
0706	1100	TAD XCOEF
0707	3010	TAD IN1
0708	3010	DCA 10
0709	3010	DCA 1 10
0710	2111	ISF COUNT
0711	5310	JMP.-2
0712	7000	CLA
0713	1300	TAD XSSF
0714	3010	DCA 10
0715	7000	CLA
0716	1410	TAD I 10
0717	4400	JMS I XPRINT
0718	7000	SEA
0719	5310	JMP.-4
0720	4700	JMS I XCLDA
0721	5700	JMP I CLEAR
0722	0210	CLDA
0723	7500	IN242, -242
0724	0700	XSSF,
0725	0215	215
0726	0212	212
0727	0212	212
0728	0244	244
0729	0300	300
0730	0300	300
0731	0317	317
0732	0303	303
0733	0305	305
0734	0303	303
0735	0303	303
0736	0317	317
0737	0310	310
0738	0300	300

0757 0300
0751 0317
0752 0314
0753 0215
0754 0212
0755 0212
0756 0000

302
317
314
215
212
212
0

/START OF PROGRAM

1000 0000
1001 0032
1002 0046
1003 4475
1004 4467
1005 4466

*1000

IOF

KCC

TL5

JMS I NCLEAR

JMS I PCF

JMS I VLF

/ACCEPT INSTRUCTIONS FROM OPERATOR

1006 7000 TTY

CLA

TAD CAP

JMS I MPINT

CLA

JMS I MREAD

TAD MM132

SNA

1015 5474

JMS I MSET

/DEFINE SET POINTS Z1,Z2

1016 1354

TAD K5

1017 7450

SNA

1020 5326

JMP U

/DEFINE CONTROL LAW FOR U1 OR U2

1021 7001

IAC

1022 7450

SNA

1023 5472

JMP I MT

/DEFINE SAMPLE INTERVAL T

1024 7001

IAC

1025 7450

SNA

1026 5473

JMP I MS

/START COMMAND

1027 1150

TAD K3

1030 7450

SNA

1031 5470

JMP I MP

/REQUEST OUTPUT

1032 1352

TAD MM12

/ON PRINTER/PUNCH

1033 7450

SNA

1034 5272

JMP D

/DEFINE COEFFICIENT D

1035 7001

IAC

1036 7450

SNA

1037 5056

JMP C

/DEFINE COEFFICIENT C

1038 7001

IAC

1039 7450

SNA

1042 5260

JMP D

/DEFINE COEFFICIENT B

1043 7001

IAC

1044 7450

SNA

1045 5047

JMP A

/DEFINE COEFFICIENT A

1046 5471

JMS I MC

1047 1000

TAD MC

1048 0000

BCA MC

1049 1000

TAD MC

1050	3303		DCA 11
1053	1361		TAD INST2
1054	3277		DCA 12
1055	5376		JMP CONT
1056	1356	C,	TAD UA
1057	1363		TAD K27
1060	3357		DCA UAD
1061	5251		JMP A+2
1062	1356	3,	TAD UA
1063	1364		TAD K12
1064	3257		DCA UAD
1065	1365		TAD INST3
1066	3375		DCA 11
1067	1367		TAD INST4
1070	3277		DCA 12
1071	5276		JMP CONT
1072	1330	3,	TAD UA
1073	1371		TAD K39
1074	3357		DCA UAD
1075	5065		JMP B+3
1076	4404	CONT,	JMS I MREAD
1077		12,	C
1100	7510		SNA
1101	7410		SHP
1102	5471		JMP I MO
1103	3372		DCA ADDR
1104	1372		TAD ADDR
1105		11,	C
1106	7537		SNA SNA
1107	7410		SHP
1110	5471		JMP I MO
1111	4477		JMS I MEO
1112	1372		TAD ADDR
1113	7100		CLL
1114	7004		FAL
1115	1372		TAD ADDR
1116	1357		TAD UAD
1117	3372		DCA ADDR
1120	0000		HCC
1121	4405		JMS I 5
1122	4407		JMS I 7
1123	6772		FEXT I ADDR
1124			FEXT
1125	5000		JMP TTY
1126	1100	U,	TAD XCORP
1127	0000		DCA UA
1130	4404		JMS I MREAD
1131	1300		TAD K161
1132	7500		SNA
1133	7410		SHP
1134	5471		JMP I MO
1135	7400		SNA
1136	5000		JMP +0
1137	1107		TAD K11

1140	7027		SZA
1141	5071		JMP I MC
1142	1373		TAD RACC
1143	3356		DCA UA
1144	4464		JMS I READ
1145	1374		TAD MM15
1146	7440		SZA
1147	5071		JMP I MC
1150	4460		JMS I NLF
1151	5200		JMP TTY
1152	7014	MM14,	14
1153	7040	MM130,	-132
1154	7025	MS,	5
1155	7070	CA7,	076
1156	7070	CA,	40
1157	7070	CA0,	0
1160	1117	INST1,	TAD MM3
1161	1360	INST2,	TAD MM61
1162	7717	MM61,	-61
1163	7723	MM7,	33
1164	7717	MM2,	100
1165	1000	INST3,	TAD MM4
1166	7774	MM4,	-4
1167	1270	INST4,	TAD MM09
1170	7770	MM09,	-09
1171	7770	MM0,	07
1172	7770	MM00,	0
1173	7770	MM60,	466
1174	7770	MM15,	-15
1175	7770	MM15,	41200
1200	7000	READ,	0
1201	7270		CLA
1202	6031		HST
1203	5202		JMP.-1
1204	6030		HED
1205	4210		JMS PRINT
1206	7307		AND MASK1
1207	5670		JMP I READ
1210	7000	PRINT,	0
1211	6041		TSP
1212	5211		JMP.-1
1213	6046		TLE
1214	5010		JMP I PRINT
1215	7000	LF,	0
1216	7200		CLA
1217	1347		TAD MM12
1220	4210		JMS PRINT
1221	5015		JMP I LF
1222	7000	CP,	0
1223	7000		CLA
1224	1341		TAD MM15
1225	4210		JMS PRINT
1226	5000		JMP I CP
1227	4200		JMP TO

1230	4255	JMS READ
1231	1335	TAD HCC
1232	7510	SPA
1233	5261	JMP 0
1234	3111	DCA COUNT
1235	1111	TAD COUNT
1236	1110	TAD KM3
1237	7540	SMA SZA
1240	5261	JMP 0
1241	4255	JMS READ
1242	1336	TAD H15
1243	7440	SZA
1244	5261	JMP 0
1245	4215	JMS LF
1246	7350	CLA CLL
1247	3733	DCA P MORT1
1250	3734	DCA I MORT2
1251	1111	TAD COUNT
1252	7010	PAR
1253	7430	SZL
1254	0733	ISZ I MORT1
1255	7010	PAR
1256	7430	SZL
1257	0734	ISZ I MORT2
1260	5463	JMP I MTTY
1261	4222	JMS CP
1262	4215	JMS LF
1263	7200	CLA
1264	1342	TAD K277
1265	4210	JMS PRINT
1266	4222	JMS CP
1267	4215	JMS LF
1270	5463	JMP I MTTY
1271	4325	JMS EQ
1272	6032	HCC
1273	4405	JMS I 5
1274	4407	JMS I 7 0
1275	3345	FIFTY TWENTY
1276	5500	TEXT
1277	4542	JMS I MFIX
1300	7041	CIA
1301	3146	DCA HSTRT
1302	5463	JMP I MTTY
1303	4222	JMS CP
1304	4215	JMS LF
1305	7200	CLA
1306	1734	TAD I MORT2
1307	7440	SZA
1310	0744	JMS I MHEAD
1311	7200	CLA
1312	1146	TAD HSTRT
1313	3147	DCA HCOUNT
1314	1140	TAD LSTRT
1315	3143	DCA LCOUNT

1316	3112		DCA HUN
1317	3143		DCA HALT
1320	6642		TCF
1321	6530		ADCL
1322	4500		JMS I XAD
1323	6661		ION
1324	5324		JMP.
1325	6600	EQ.	S
1326	4200		JMS READ
1327	1343		TAD IN75
1328	7440		SZA
1331	5261		JMP C
1332	5725		JMP I EQ
1333	2151	NOPT1,	OPT1
1334	2152	NOPT2,	OPT2
1335	7720	NOC,	-6C
1336	7763	N15,	-15
1337	6177	NASH1,	177
1340	6212	N212,	212
1341	6215	N215,	215
1342	6277	N277,	277
1343	7700	NK75,	-75
1344	6270	NK75,	HEADER
1345	6115	TWENTY,	S
1346	2400		PASC
1347	6600		/
			/CONTROL ALGORITHM
			/IMPLEMENTATION OF CONTROL LAW
			/
			B1400
1400	7330	CLOCK,	NOP
1401	7330		CLA CLL
1402	4407		JMS I 7
1403	5127		FEET Z1
1404	7330		FNOR
1405	6504		FEUT I 1X1
1406	5121		FEET Y1
1407	7330		FNOR
1410	6504		FEUT I 1Y1
1411	6504		FEUT I 1Y1
1412	6600		FEET
1413	7330		CLA CLL
1414	1144		TAD 44
1415	1177		TAD IN71
1416	3744		DCA 44
1417	4540		JMS I 1X1
1420	3135		DCA 71
1421	4407		JMS I 7
1422	5127		FEET Z2
1423	7330		FNOR
1424	6504		FEUT I 1X2
1425	5121		FEET Y2
1426	7330		FNOR

1427	0005	FCST 1 NMC
1428	0505	IPUT 1 NMC
1429	0000	FENT
1430	7300	CLA CLL
1433	1044	TAD 44
1434	1107	TAD KMI
1435	3044	DCA 44
1436	4542	JMS 1 NFIX
1437	0136	DCA E2
1440	7200	CLA
1441	1106	TAD XCOEF
1442	3360	DCA COEF
1443	1102	TAD KUI
1444	3361	DCA VAR
1445	4327	JMS CTRL
1446	1366	TAD TEN
1447	3114	DCA U1+1
1450	1102	TAD KUI
1451	3361	DCA VAR
1452	4327	JMS CTRL
1453	1366	TAD TEN
1454	0117	DCA U2+1
1455	7000	NOP
1456	4542	JMS 1 NTAO
1457	1000	TAD KMI
1460	7101	IAC
1461	0111	DCA COUNT
1462	1102	TAD KUI
1463	1364	TAD K60
1464	1110	TAD KMI
1465	3361	DCA VAR
1466	1361	TAD VAR
1467	1110	TAD KMI
1470	3363	DCA VARN3
1471	4007	JMS 1 7
1472	5765	FCST 1 VARN3
1473	0701	FENT 1 VAR
1474	0000	FENT
1475	7200	CLA
1476	1001	TAD VAR
1477	2111	ISZ COUNT
1500	5064	JMP UP
1501	7000	CLA
1502	4007	JMS 1 7
1503	5113	FCST U1
1504	7000	FENT
1505	0000	FENT
1506	5116	FGET U2
1507	7000	FNOR
1510	0000	FENT
1511	0000	FENT
1512	7000	NOP
1513	7000	CLA
1514	1100	TAD HALT

1315	7030		CHA
1516	3316		JMP...
1517	7200		CLA
1523	0002		IOF
1521	0000		TLC
1520	1363		TAD NH
1523	4465		JMS I XPRINT
1524	4467		JMS I XCR
1525	4406		JMS I XLF
1526	5463		JMP I XTTY
1527	0000	CTEL	G
1530	7200		CLA
1531	3306		DCA TEM
1532	3307		DCA TEM+1
1533	3373		DCA TEM+2
1534	1362		TAD NH18
1535	3111		DCA COUNT
1536	4407		JMS I 7
1537	5760		FCET I COEF
1540	3761		FLPY I VAR
1541	1366		FADD TEM
1542	0000		FUT TEM
1543	0000		TEXT
1544	7200		CLA
1545	1360		TAD COEF
1546	1150		TAD H3
1547	3360		DCA COEF
1550	1031		TAD VAP
1551	1150		TAD H3
1552	3361		DCA VAR
1553	2111		ISE COUNT
1554	5336		JMP CTCL+7
1555	4542		JMS I XFIX
1556	0000		DCA TEM
1557	5727		JMP I CTRL
1560	0000	COEF	G
1561	0000	VAP	G
1562	7736	NH13	-22
1563	0010	NH	300
1564	0000	N60	66
1565	0000	VARH3	G
1566	0000	TEM	G
1567	0000		G
1570	0000		G

/ INTERRUPT SERVICE ROUTINES

/

81600

1600	6530	IAD	ADCL	/A/D INTERRUPT
1601	7300		CLA CLL	
1602	0532		ADPB	
1603	7004		TAL	
1604	7500		SFA	
1605	5000		JMP.+3	

1006	0273		AND MASK2
1007	7041		CLA
1010	3274		DCA ADREG
1011	1112		TAD NUK
1012	7450		SNA
1013	5221		JMP.+6
1014	7230		CLA
1015	1274		TAD ADREG
1016	3276		DCA BY2
1017	3112		DCA NUK
1020	5224		JMP.+4
1021	1274		TAD ADREG
1022	3275		DCA BY1
1023	2112		ISZ NUK
1024	2145		ISZ LCOUNT
1025	5242		JMP NOCLK
1026	1144		TAD LSTRT
1027	3145		DCA LCOUNT
1030	2147		ISZ NCOUNT
1031	5242		JMP NOCLK
1032	1146		TAD RSTRT
1033	3147		DCA NCOUNT
1034	1772		TAD NOCLOCK
1035	3000		DCA E
1036	1275		TAD BY1
1037	0102		DCA Y1+1
1040	1276		TAD BY2
1041	3125		DCA Y2+1
1042	4244	NOCLK	JMS AD
1043	5537		JMP I XEXIT
1044	0000	AD,	CL
1045	7300		CLA CLL
1046	1112		TAD NUK
1047	6531		ADEN
1050	1112		TAD NUK
1051	1277		TAD ENABLE
1052	6536		ADLE
1053	6532		ADST
1054	5044		JMP I AD
1055	0142	ITP,	TCF /PRINTER INTERRUPT
1056	2500		ISZ C
1057	3527		JMP I XEXIT
1060	0122	IPP,	PCF /PUNCH INTERRUPT
1061	2500		ISZ C
1062	5537		JMP I XEXIT
1063	0000	INT,	KCF /KEYBOARD INTERRUPT
1064	7200		CLA /DETECT HALT COMMAND
1065	6034		KFS
1066	6300		AND MSK1
1067	1301		TAD IN
1070	7450		SNA
1071	2143		ISZ HALT
1072	5537		JMP I XEXIT
1073	0777	HALT,	0777

1674	6666	ADDER,	3
1675	6666	BY1,	6
1676	6666	BY2,	6
1677	1659	ENABLE,	1666
1756	6177	MSH1,	177
1751	7675	INH,	-163

/OUTPUT ROUTINE

/

*2000

2557	6666	DAO,	6
2551	7355		CLA CLL
2552	3344		DCA DACH
2553	1114		TAD U1+1
2554	4327		JMS DA
2555	2344		ISE DACH
2556	1417		TAD U2+1
2557	4327		JMS DA
2558	1136		TAD Z1+1
2559	4241		JMS HALF
2560	4276		JMS OUT
2561	1136		TAD V1+1
2562	4241		JMS HALF
2563	4276		JMS OUT
2564	1136		TAD V1
2565	4276		JMS OUT
2566	1114		TAD U1+1
2567	4241		JMS HALF
2568	4276		JMS OUT
2569	1136		TAD E0+1
2570	4241		JMS HALF
2571	4276		JMS OUT
2572	1136		TAD E2
2573	4276		JMS OUT
2574	1117		TAD U2+1
2575	4241		JMS HALF
2576	4276		JMS OUT
2577	1136		TAD E2
2578	4276		JMS OUT
2579	1117		TAD U2+1
2580	4241		JMS HALF
2581	4276		JMS OUT
2582	4753		JMS I RET
2583	7675		NOT
2584	5666		JMS I DAO
2585	6666	HALT,	
2586	7166		CLL
2587	5517		SPA
2588	7666		CLL
2589	7616		RAR
2590	5641		JMS I HALF
2591	6666	FIN,	6
2592	7255		CLA
2593	1644		TAD 44
2594	7647		SPA SNA
2595	6666		J T.*2

/FLOATING TO FIXED POINT CONVERSION

2054	7000		CLA
2055	5047		JMP I FIN
2056	1047		TAD MM13
2057	7450		SNA
2060	5274		JMP DONE
2061	7500		SNA
2062	5541		JMP I MERR
2063	3544		DCA 44
2064	7100		CLL
2065	1045		TAD 45
2066	7510		SPA
2067	7000		CHL
2070	7010		RAR
2071	3045		DCA 45
2072	2044		ISS 44
2073	5204		JMP 47
2074	1045	DONE,	TAD 45
2075	3047		JMP I FIN
2076	5000	OUT,	
2077	3045		DCA BUFFER
2100	1051		TAD OPT1
2101	7440		SEA
2102	4007		JMP PUNCH
2103	1050		TAD OPT2
2104	7440		SZA
2105	0704		JMS I MPRINT
2106	5070		JMP I OUT
2107	0000	PUNCH,	0 /PAPER TAPE OUTPUT
2110	7000		CLA
2111	1045		TAD BUFFER
2112	7002		DSV
2113	0046		AND MSK77
2114	1050		TAD BITS
2115	6006		PLS
2116	5010		JMP.
2117	7000		CLA
2120	1045		TAD BUFFER
2121	0046		AND MSK77
2122	1050		TAD BITS
2123	6006		PLS
2124	5004		JMP.
2125	7000		CLA
2126	5007		JMP I PUNCH
2127	0000	DA,	0 /DATA OUTPUT
2130	0440		DATS
2131	7510		SPA
2132	1107		TAD MM1
2133	0440		DALD
2134	7000		CLA-CLL
2135	1044		TAD PACH
2136	0440		DALA
2137	0440		DAST
2138	0441		DASH
2139	5000		JMP.-1

2142	7200	CLA	
2143	5707	JMP I DA	
2144	0000	DACH,	0
2145	0000	BUFFER,	0
2146	0077	MEM77,	77
2147	7765	MEM13,	-13
2150	0000	DITE,	000
2151	0000	OPT1,	0
2152	0000	OPT2,	0
2153	2304	MPET,	DET
2154	2200	MPRINT,	PRNT
			*2200
2200	0000	PRNT,	0 /PRINTER OUTPUT
2201	7200	CLA	
2202	1734	TAD I MSUF	
2203	3045	DCA 45	
2204	1011	TAD M13	
2205	3044	DCA 44	
2206	3046	DCA 46	
2207	4407	JMS I 7	
2210	7000	THOR	
2211	0010	TDIN M1004	
2212	0000	PRNT	
2213	4542	JMS I MFIN	
2214	3734	DCA I MSUF	
2215	1734	TAD I MSUF	
2216	7500	SMA	
2217	5024	JMP.+5	
2220	7001	CIA	
2221	3734	DCA I MSUF	
2222	1347	TAD MINUS	
2223	7410	SNP	
2224	7200	CLA	
2225	1340	TAD M240	
2226	6046	TLS	
2227	5227	JMP.	
2230	7300	CLA CLL	
2231	1335	TAD M1000	
2232	4251	JMS DIGIT	
2233	1036	TAD M100	
2234	4251	JMS DIGIT	
2235	4276	JMS PERIOD	
2236	1337	TAD M10	
2237	4251	JMS DIGIT	
2240	7001	IAC	
2241	4251	JMS DIGIT	
2242	1346	TAD M240	
2243	6046	TLS	
2244	5244	JMP.	
2245	6046	TLS	
2246	5240	JMP.	
2247	7200	CLA	
2250	5600	JMP I PRNT	
2251	0000	DIGIT,	0

2252	3344	ECA DEC
2253	1344	TAD DEC
2254	7041	CLA
2255	3345	DCA HDEC
2256	1127	TAD HHI
2257	3111	DCA COUNT
2260	1734	TAD I XBUF
2261	2111	ISZ COUNT
2262	7000	NOP
2263	1343	TAD HDEC
2264	7000	SNA
2265	5261	JMP.-4
2266	1344	TAD DEC
2267	3734	DCA I XBUF
2270	1111	TAD COUNT
2271	1343	TAD H260
2272	6046	TLS
2273	5273	JMP.
2274	7200	CLA
2275	5051	JMP I DIGIT
2276	0000	PERIOD, C
2277	1342	TAD HPRD
2300	6046	TLS
2301	5301	JMP.
2302	7200	CLA
2303	5676	JMP I PERIOD
2304	0000	RET, C
2305	7200	CLA
2306	1733	TAD I YOPT1
2307	7450	SNA
2310	5314	JMP.+4
2311	7200	CLA
2312	6026	PLS
2313	5313	JMP.
2314	7200	CLA
2315	1732	TAD I YOPT2
2316	7450	SNA
2317	5704	JMP I RET
2300	7200	CLA
2321	1340	TAD HGR
2322	6046	TLS
2323	5323	JMP.
2324	7200	CLA
2305	1341	TAD HLF
2300	6046	TLS
2307	5027	JMP.
2330	7200	CLA
2301	5704	JMP I RET
2332	2152	YOPT2, OPT2
2333	2101	YOPT1, OPT1
2304	2145	XBUF, BUFFER
2335	1750	H1000, 1750
2300	0144	H100, 144
2337	0110	H10, 10

2341	0015	H07,	015
2341	0010	HL7,	010
2342	0050	H07D,	050
2343	0060	H200,	060
2344	0000	000,	0
2345	0000	H07C,	0
2346	0240	H240,	240
2347	0015	H100S,	15

APPENDIX III

Identification Program Listing

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/IDENTIFICATION PROGRAM
/
*20
0020 1710 XCLEAR, CLEAR
0021 2400 XIDENT, IDENT
0022 2730 XPCMB, PCMB
0023 0000 INDIR, 0
0024 3027 DCA TEM
0025 1427 TAD I TEM
0026 5423 JMP I INDIR
0027 0100 TEM, 0
*102
0102 0554 UIM1, 554
*107
0107 7777 KM1, -1
*111
0111 0000 COUNT, 0
*113
0113 0013 U1, 13
0114 0000 0
0115 0000 0
*121
0121 0013 Y1, 13
0122 0000 0
0123 0000 0
*127
0127 0013 Z1, 13
0130 0000 0
0131 0000 0
*150
0150 0000 K3, 3
0151 3400 VAC, 3400
0152 3430 MAC, 3430
0153 3730 MREG, 3730
0154 4230 P, 4230
0155 4530 PHI, 4530
0156 4560 BETA, 4560
0157 4610 GAMMA, 4610
0158 4640 FREQ, 4640

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0161	4643	VREG,	4643
0162	0013	SUM1,	13
0163	0000		0
0164	0000		0
0165	0000	ARG1,	0
0166	0000	ARG2,	0
0167	0000	ADI,	0
0170	0000	COUNT,	0
0171	7770	K18,	-10
0172	0573	YIM1,	573
0173	2151	OPT1,	2151
0174	2107	PUNCH,	2107
0175	2145	BUFFER,	2145
0176	7750	K124,	-30
0177	3777	K3777,	3777

			*701
0701	4420	JMS I XCLEAR	
			*1400
1400	4421	JMS I XIDENT	
			*1450
1450	4652	JMS I XSAT	
1451	7410	SKP	
1452	2050	XSAT,	SAT
1453	7000	VOP	
1454	7000	POP	

/MATRIX AND VECTOR INITIALIZATION

			*1710
1710	0000	CLEAR,	0
1711	7300	CLA CLL	
1712	3160	DCA SUM1+1	
1713	3164	DCA SUM1+2	
1714	1151	TAD VAC	
1715	1117	TAD K11	
1716	3010	DCA 10	
1717	1350	TAD K1273	
1720	3111	DCA COUNT	
1721	3410	DCA I 10	
1722	2111	ISZ COUNT	
1723	5321	JMP.-2	
1724	1171	TAD K18	
1725	3111	DCA COUNT	
1726	1154	TAD P	
1727	3165	DCA-ARG1	
1730	4407	RO,	JMS I 7
1731	5352	FBET KSTRT	
1732	6565	FPUT I ARG1	
1733	0000	FEXT	
1734	7320	CLA CLL	
1735	1165	TAD ARG1	
1736	1355	TAD T27	
1737	3165	DCA ARG1	
1740	2111	ISZ COUNT	

1741	5330		JMP R0
1742	4437		JIS I 7
1743	5352		FGET KSTRT
1744	6756		FPUT I XBETA5
1745	0000		FEXT
1746	7300		CLA CLL
1747	5710		JMP I CLEAR
1750	6505	Y1273,	-1273
1751	2661	YADDI,	ADDI
1752	2001	KSTRT,	I
1753	2000		2000
1754	0000		0
1755	0033	T27,	33
1756	4574	XBETA5,	4574
			*2003
2003	1163		TAD SUM1+1
			*2020
2020	1163		TAD SUM1+1
			*2037
2037	4402		JIS I KPC12
			/
			/CONTROL VARIABLE SATURATION HANDLER
			/
			*2350
2350	0000	SAT,	0
2351	7300		CLA CLL
2352	1114		TAD U1+1
2353	7510		SPA
2354	5370		JMP NEGU
2355	1163		TAD SUM1+1
2356	7500		S1A
2357	5366		JMP NEGU-2
2360	7300		CLA CLL
2361	1163		TAD SUM1+1
2362	7041		CIA
2363	1177		TAD X3777
2364	3114		DCA U1+1
2365	1177		TAD X3777
2366	3163		DCA SUM1+1
2367	5750		JMP I SAT
2370	1163	NEGU,	TAD SUM1+1
2371	7570		S1A
2372	5366		JMP NEGU-2
2373	7300		CLA CLL
2374	1163		TAD SUM1+1
2375	7041		CIA
2376	3114		DCA U1+1
2377	5366		JMP NEGU-2

/ IDENTIFICATION ALGORITHM

/ UPDATE BETA

/

*2400

2400	0000	IDENT,	0	ENTRY FROM CONTROL PROGRAM
2401	4761		JMS I XVXV	
2402	0157		GAMMA	
2403	0155		PHI	
2404	4762		JMS I XADDI	
2405	0152		VAC	
2406	4763		JMS I XMXM	
2407	0152		VAC	
2408	0154		P	
2411	4317		JMS MTXFR	
2412	0153		VREG	
2413	0154		P	
2414	4757		JMS I XGPHI	
2415	4764		JMS I XMXV	
2416	0154		P	
2417	0155		PHI	
2420	4765		JMS I XVDTV	
2421	0155		PHI	
2422	0151		VAC	
2423	4407		JMS I 7	
2424	5560		FGET I FREG	
2425	1372		FADD ALPHA	
2426	6560		FPUT I FREG	
2427	4550		FDIV I FREG	
2430	4560		FDIV I FREG	
2431	6560		FPUT I FREG	
2432	0000		FEXT	
2433	4766		JMS I XVXS	
2434	0155		PHI	
2435	0160		FREG	
2436	4301		JMS VTXFR	
2437	0151		VAC	
2440	0161		VREG	
2441	4764		JMS I XMXV	
2442	0154		P	
2443	0161		VREG	
2444	4301		JMS VTXFR	
2445	0151		VAC	
2446	0157		GAMMA	
2447	4760		JMS I XVJEG	
2450	0155		PHI	
2451	4765		JMS I XVDTV	
2452	0155		PHI	
2453	0156		BETA	
2454	4407		JMS I 7	
2455	5127		FGET 21	
2456	7000		FVOR	
2457	6375		FPUT REG	
2460	5121		FGET Y1	

2461	7000	FMOT	
2462	2375	FSUB REG	
2463	1560	FADD I FREG	
2464	6560	FPUT I FREG	
2465	0000	FEXT	
2466	4766	JMS I XVXS	
2467	0157	GAMMA	
2470	0160	FREG	
2471	4767	JMS I XVPLSV	
2472	0156	BETA	
2473	0151	VAC	
2474	4301	JMS VTXFR	
2475	0151	VAC	
2476	0156	BETA	
2477	4770	JMS I XPBETA	/COMPUTE MIN VARIANCE CONTROL
2500	5600	JMP I IDENT	/RETURN TO CONTROL PROGRAM

/INPUT-OUTPUT ROUTINES

/MATRIX ARITHMETIC ROUTINES

/

2501	0000	VTXFR,	0	/VECTOR TRANSFER ROUTINE
2502	7320	CLA	CLL	
2503	1701	TAD	I VTXFR	
2504	4023	JIS	INDIR	
2505	3165	DCA	ARG1	
2506	2301	LSZ	VTXFR	
2507	1701	TAD	I VTXFR	
2510	4023	JIS	INDIR	
2511	3166	DCA	ARG2	
2512	2301	LSZ	VTXFR	
2513	1171	TAD	X15	
2514	3101	DCA	COUNT	
2515	4335	JMS	TXFR	
2516	5701	JIP	I VTXFR	
2517	0000	MTXFR,	0	/MATRIX TRANSFER ROUTINE
2520	7320	CLA	CLL	
2521	1717	TAD	I MTXFR	
2522	4023	JIS	INDIR	
2523	3165	DCA	ARG1	
2524	2317	LSZ	MTXFR	
2525	1717	TAD	I MTXFR	
2526	4023	JIS	INDIR	
2527	3166	DCA	ARG2	
2530	2317	LSZ	MTXFR	
2531	1371	TAD	X164	
2532	3111	DCA	COUNT	
2533	4335	JMS	TXFR	
2534	5717	JMP	I MTXFR	
2535	0000	TXFR,	0	
2536	4407	JMS	I 7	
2537	5565	FEXT	I ARG1	
2540	5566	FPUT	I ARG2	
2541	0000	FEXT		
2542	4346	JMS	ARG	

2543	2111		ISZ COUNT
2544	5336		JMP TXF+1
2545	5735		JMP I TXFR
2546	0000	ARG,	0
2547	7300		CLA CLL
2552	1165		TAD ARG1
2551	1150		TAD K3
2552	3165		DCA ARG1
2553	1166		TAD ARG2
2554	1150		TAD K3
2555	3166		DCA ARG2
2556	5746		JMP I ARG
2557	2635	XGPHI,	GPHI
2560	2705	VNEG,	VNEG
2561	3027	VXV,	VXV
2562	2661	XADDI,	ADDI
2563	3200	XIXI,	IXI
2564	3237	XIXV,	IXV
2565	3117	XVDTV,	VDTV
2566	3000	XVXS,	VXS
2567	3265	XVPLSV,	VPLSV
2574	2600	XPRETA,	PRETA
2571	7700	X154,	-100
2572	0001	ALPHA,	1
2573	2000		2000
2574	0000		0
2575	0000	REG,	0
2576	0000		0
2577	0000		0
			/
			/COMPUTE MINIMUM VARIANCE CONTROL LAW
			/
			*2600
2600	0000	PRETA,	0
2601	4400		JIS I 7
2602	5556		FGET I BETA1
2603	4755		FDIV I BETA5
2604	6764		FPUT I B0
2605	5752		FGET I BETA2
2606	4755		FDIV I BETA5
2607	6765		FPUT I B1
2610	5753		FGET I BETA3
2611	4755		FDIV I BETA5
2612	6766		FPUT I B2
2613	5754		FGET I BETA4
2614	4755		FDIV I BETA5
2615	6767		FPUT I B3
2616	5774		FGET I XZERO
2617	2756		FSUB I BETA6
2620	4755		FDIV I BETA5
2621	6761		FPUT I A1
2622	5774		FGET I XZERO
2623	2757		FSUB I BETA7
2624	4755		FDIV I BETA5

2625	6762		FPUT I A2
2626	5774		FGET I XZERO
2627	2760		FSUB I BETAS
2630	4755		FDIV I BETAS
2631	6763		FPUT I A3
2632	0000		FEXT
2633	7300		CLA CLL
2634	5600		JMP I PBETA
2635	0000	GPPI,	0 /GET VECTOR PHI
2636	4770		JMS I XVTXFR
2637	0172		YIMI
2640	0155		PHI
2641	4305		JMS VNEG
2642	0155		PHI
2643	1102		TAD UIMI
2644	3165		DCA ARG1
2645	1371		TAD PHIS
2646	3165		DCA ARG2
2647	1372		TAD X14
2650	3111		DCA COUNT
2651	4407	R12,	JMS I 7
2652	5565		FGET I ARG1
2653	6566		FPUT I ARG2
2654	0000		FEXT
2655	4773		JMS I XARG
2656	2111		ISZ COUNT
2657	5251		JMP R12
2660	5635		JMP I GPPI
2661	0000	ADDI,	0 /ADD IDENTITY MATRIX
2662	7300		CLA CLL
2663	1661		TAD I ADDI
2664	4023		JMS INDIR
2665	3165		DCA ARG1
2666	2261		ISZ ADDI
2667	1171		TAD X18
2670	3111		DCA COUNT
2671	4407	R9,	JMS I 7
2672	5565		FGET I ARG1
2673	1775		FADD I XONE
2674	6565		FPUT I ARG1
2675	0000		FEXT
2676	7200		CLA
2677	4165		TAD ARG1
2700	1376		TAD X27
2701	3165		DCA ARG1
2702	2111		ISZ COUNT
2703	5271		JMP R9
2704	5661		JMP I ADDI
2705	0000	VNEG,	0 /MULTIPLY VECTOR BY -1
2706	7300		CLA CLL
2707	1705		TAD I VNEG
2710	4023		JMS INDIR
2711	3165		DCA ARG1
2712	2205		ISZ VNEG

2713	1171		TAD K18
2714	3111		DCA COUNT
2715	3044	R3,	DCA 44
2716	3045		DCA 45
2717	3046		DCA 46
2722	4407		JMS I 7
2721	2565		FEND I APGI
2722	6565		FEND I APGI
2723	0000		FEXT
2724	4773		JMS I KAPE
2725	2111		ISZ COUNT
2726	5315		JIP 23
2727	5705		JIP I VIEG
2730	0300	PC12,	0 /PUNCH PARAMETER ESTIMATES ON PAPER TAPE
2731	1573		STAD I OPT1
2732	7450		SJA
2733	5730		JIP I PCMB
2734	7200		CLA
2735	1176		TAD K124
2736	3111		DCA COUNT
2737	1156		TAD BETA
2740	1107		TAD K11
2741	3010		DCA 10
2742	1410		TAD I 10
2743	3575		DCA I BUFFER
2744	4574		JMS I PUNCH
2745	2111		ISZ COUNT
2746	5342		JIP - 4
2747	6026		PLS
2750	5350		JIP.
2751	5730		JIP I PCMB
2752	4563	BETA2,	4563
2753	4566	BETA3,	4566
2754	4571	BETA4,	4571
2755	4574	BETA5,	4574
2756	4577	BETA6,	4577
2757	4602	BETA7,	4602
2760	4605	BETA8,	4605
2761	0400	A1,	400
2762	0403	A2,	403
2763	0406	A3,	406
2764	0414	B3,	414
2765	0417	B1,	417
2766	0422	B2,	422
2767	0425	B3,	425
2770	2501	XVTXFR,	VTXFR
2771	4544	PHI5,	4544
2772	7774	K14,	-4
2773	2546	KAPG,	APG
2774	3166	XZERO,	ZERO
2775	3163	XOIE,	OIE
2776	0033	X27,	33

3000	0000	VXS,	*3000	0	/MULTIPLY VECTOR BY SCALAR
3001	7300		CLA	CLL	
3002	1600		TAD	I VXS	
3003	4023		JMS	INDIR	
3004	3165		DCA	ARG1	
3005	2200		ISZ	VXS	
3006	1600		TAD	I VXS	
3007	4023		JMS	INDIR	
3010	3167		DCA	ADI	
3011	2200		ISZ	VXS	
3012	1151		TAD	VAC	
3013	3166		DCA	ARG2	
3014	1171		TAD	K18	
3015	3111		DCA	COUNT	
3016	4407	R6,	JMS	I 7	
3017	5565		RGET	I ARG1	
3020	3567		FMPY	I ADI	
3021	6566		FPUT	I ARG2	
3022	0000		FEXT		
3023	4753		JMS	I YARG	
3024	2111		ISZ	COUNT	
3025	5216		JMP	R6	
3026	5600		JMP	I VXS	
3027	0000	VXV,	0	/VECTOR CROSS PRODUCT	
3030	7300		CLA	CLL	
3031	1627		TAD	I VXV	
3032	4023		JMS	INDIR	
3033	3167		DCA	ADI	
3034	2227		ISZ	VXV	
3035	1171		TAD	K18	
3036	3170		DCA	COUNT	
3037	1152		TAD	VAC	
3040	3166		DCA	ARG2	
3041	1627	R3,	TAD	I VXV	
3042	4023		JMS	INDIR	
3043	3165		DCA	ARG1	
3044	1171		TAD	K18	
3045	3111		DCA	COUNT	
3046	4407	R7,	JMS	I 7	
3047	5567		RGET	I ADI	
3050	3565		FMPY	I ARG1	
3051	6566		FPUT	I ARG2	
3052	0000		FEXT		
3053	4753		JMS	I YARG	
3054	2111		ISZ	COUNT	
3055	5246		JMP	R7	
3056	1167		TAD	ADI	
3057	1150		TAD	K3	
3060	3167		DCA	ADI	
3061	2170		ISZ	COUNT	
3062	5241		JMP	R8	
3063	2227		ISZ	VXV	
3064	5627		JMP	I VXV	

3065	0000	VPLSV,	0	/ADD TWO VECTORS
3066	7300			CLA CLL
3067	1665			TAD I VPLSV
3070	4023			JMS INDIR
3071	3165			DCA ARG1
3072	2265			ISZ VPLSV
3073	1665			TAD I VPLSV
3074	4023			JMS INDIR
3075	3166			DCA ARG2
3076	2265			ISZ VPLSV
3077	1174			TAD K18
3100	3111			DCA COUNT
3101	1151			TAD VAC
3102	3167			DCA ADI
3103	4407	R5,		JMS I 7
3104	5565			FGET I ARG1
3105	1566			FADD I ARG2
3106	6567			FPUT I ADI
3107	0000			FEXT
3110	4753			JMS I YARG
3111	1167			TAD ADI
3112	1150			TAD K3
3113	3167			DCA ADI
3114	2111			ISZ COUNT
3115	5303			JMP R5
3116	5665			JMP I VPLSV
3117	0000	VDTV,	0	/VECTOR DOT PRODUCT
3120	7300			CLA CLL
3121	1717			TAD I VDTV
3122	4023			JMS INDIR
3123	3165			DCA ARG1
3124	2317			ISZ VDTV
3125	1717			TAD I VDTV
3126	4023			JMS INDIR
3127	3166			DCA ARG2
3130	1171			TAD K18
3131	3111			DCA COUNT
3132	3044			DCA 44
3133	3045			DCA 45
3134	3046			DCA 46
3135	4407			JMS I 7
3136	6560			FPUT I FREG
3137	0000			FEXT
3140	4407	R4,		JMS I 7
3141	5565			FGET I ARG1
3142	3566			FGET I ARG2
3143	1560			FADD I FREG
3144	6560			FPUT I FREG
3145	0000			FEXT
3146	4753			JMS I YARG
3147	2111			ISZ COUNT
3150	5303			JMP R4
3151	2317			ISZ VDTV
3152	5717			JMP I VDTV

3153	2546	YADG,	ARG
3154	0000	THREE,	0
3155	3362		DCA TEMP
3156	1362		TAD TEMP
3157	7104		CLL RAL
3160	1362		TAD TEMP
3161	5754		JMP I THREE
3162	0000	TEMP,	0
3163	0001	ONE,	1
3164	2000		2000
3165	0000		0
3166	0000	ZERO,	0
3167	0000		0
3170	0000		0

*3200

3200	0000	MX1,	0	MATRIX MULTIPLICATION
3201	7300		CLA CLL	
3202	1600		TAD I MX1	
3203	4023		JMS INDIR	
3204	3374		DCA ARG5	

3205	2200		ISZ MX1
3206	1600		TAD I MX1
3207	4023		JMS INDIR
3210	3371		DCA ARG6
3211	2200		ISZ MX1
3212	1171		TAD K18
3213	3360		DCA KCOUNT
3214	1153		TAD ARG6
3215	3372		DCA ARG7
3216	4276		JMS MTRANS
3217	3371		ARG6
3220	4237	R2,	JMS MXV
3221	3370		ARG5
3222	3371		ARG6
3223	4761		JMS I YUTYFR
3224	1151		VAC
3225	3372		ARG7
3226	1371		TAD ARG6
3227	1362		TAD K24
3230	3371		DCA ARG6
3231	1372		TAD ARG7
3232	1362		TAD K24
3233	3372		DCA ARG7
3234	2360		ISZ KCOUNT
3235	5220		JMP R2
3236	5600		JMP I MX1

3237	0000	MXV,	0	MATRIX-VECTOR MULTIPLICATION
3240	7300		CLA CLL	
3241	1637		TAD I MXV	
3242	4023		JMS INDIR	
3243	3366		DCA ARG3	
3244	2237		ISZ MXV	
3245	1637		TAD I MXV	
3246	4023		JMS INDIR	

3247	3367		DCA ARG4
3250	1151		TAD VAC
3251	3167		DCA ADI
3252	2237		ISZ MKV
3253	1171		TAD K18
3254	3170		DCA KOUNT
3255	4764	RI,	JMS I YVDTV
3256	3366		ARG3
3257	3367		ARG4
3260	4407		JMS I 7
3261	5560		FGET I FREG
3262	6567		FPUT I ADI
3263	0000		FEXT
3264	7200		CLA
3265	1366		TAD ARG3
3266	1362		TAD K24
3267	3366		DCA ARG3
3270	1167		TAD ADI
3271	1150		TAD K3
3272	3167		DCA ADI
3273	2170		ISZ KOUNT
3274	5255		JMP RI
3275	5637		JMP I MKV
3276	0000	MTRANS, 0	MATRIX TRANSPOSITION
3277	7300		CLA CLL
3300	1676		TAD I MTRANS
3301	4023		JMS IUDIR
3302	3167		DCA ADI
3303	2276		ISZ MTRANS
3304	1171		TAD K18
3305	3111		DCA COUNT
3306	3356		DCA II
3307	3357	RII,	DCA JJ
3310	1171		TAD K18
3311	3170		DCA KOUNT
3312	1356	RIQ,	TAD II
3313	7041		CIA
3314	1357		TAD JJ
3315	7550		SPA SNA
3316	5346		JIP MS
3317	7300		CLA CLL
3320	1356		TAD II
3321	7006		RTL
3322	7004		RAL
3323	1357		TAD JJ
3324	4765		JMS I X3
3325	1167		TAD ADI
3326	3165		DCA ARG1
3327	1357		TAD JJ
3330	7006		RTL
3331	7004		RAL
3332	1356		TAD II
3333	4765		JMS I X3
3334	1167		TAD ADI

3335	3166		DCA ARG2
3336	4407		JMS I 7
3337	5565		FGET I ARG1
3340	6560		FPUT I FREG
3341	5566		FGET I ARG2
3342	6565		FPUT I ARG1
3343	5560		FGET I FREG
3344	6566		FPUT I ARG2
3345	0000		FEXT
3346	7200	IS,	CLA
3347	2357		ISZ JJ
3350	2170		ISZ COUNT
3351	5312		JMP R10
3352	2356		ISZ II
3353	2111		ISZ COUNT
3354	5307		JMP R11
3355	0000		JMP I MTRANS
3356	0000	II,	0
3357	0000	JJ,	0
3360	0000	KCOUNT,	0
3361	2501	MTXFR,	VTXFR
3362	0230	X20,	30
3363	2517	YMTXFR,	MTXFR
3364	3117	YVDTV,	VDTV
3365	3154	X3,	THREE
3366	0000	ARG3,	0
3367	0000	ARG4,	0
3370	0000	ARG5,	0
3371	0000	ARG6,	0
3372	0000	ARG7,	0