

MONTE-CARLO EVALUATION OF DIGITAL FILTERS  
FOR FIRE CONTROL SYSTEMS

Michael Gordon Ketron

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## Monterey, California



# THESIS

MONTE-CARLO EVALUATION OF DIGITAL FILTERS  
FOR FIRE CONTROL SYSTEMS

by

Michael Gordon Ketron

June 1974

Thesis Advisor:

D. E. Kirk

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|--|-----------------------|--|
| 1. REPORT NUMBER   | 2. GOVT ACCESSION NO. | 3. RECIPIENT'S CATALOG NUMBER                                      |
| 4. TITLE (and Subtitle)<br><br>MONTE-CARLO EVALUATION OF DIGITAL<br>FILTERS FOR FIRE CONTROL SYSTEMS   |                       | 5. TYPE OF REPORT & PERIOD COVERED<br>Master's Thesis<br>June 1974 |
|  |                       | 6. PERFORMING ORG. REPORT NUMBER                                   |
| 7. AUTHOR(s)<br><br>Michael Gordon Ketron  |                       | 8. CONTRACT OR GRANT NUMBER(s)                                     |
| 9. PERFORMING ORGANIZATION NAME AND ADDRESS<br><br>Naval Postgraduate School<br>Monterey, California 93940   |                       | 10. PROGRAM ELEMENT, PROJECT, TASK<br>AREA & WORK UNIT NUMBERS     |
| 11. CONTROLLING OFFICE NAME AND ADDRESS<br><br>Naval Postgraduate School<br>Monterey, California 93940   |                       | 12. REPORT DATE<br>June 1974                                       |
|  |                       | 13. NUMBER OF PAGES  |
| 14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office)<br><br>Naval Postgraduate School<br>Monterey, California   |                       | 15. SECURITY CLASS. (of this report)<br><br>Unclassified           |
|  |                       | 15a. DECLASSIFICATION/DOWNGRADING<br>SCHEDULE                      |
| 16. DISTRIBUTION STATEMENT (of this Report)<br><br>Approved for public release; distribution unlimited.  |                       |  |
| 17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report)   |                       |  |
| 18. SUPPLEMENTARY NOTES  |                       |  |
| 19. KEY WORDS (Continue on reverse side if necessary and identify by block number)   |                       |  |
| 20. ABSTRACT (Continue on reverse side if necessary and identify by block number)<br><br>Techniques are discussed for Monte-Carlo evaluation of several filters in a digital fire control system performing against an aircraft trajectory developed to simulate a diving attack. The programs developed are designed to allow for future modeling work in a three-dimensional theater of operations with expedient selection of the |                       |  |



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Monte-Carlo Evaluation of Digital Filters  
for Fire Control Systems

by

Michael Gordon Ketron  
Lieutenant (junior grade), United States Navy  
B.S., University of Tennessee, 1971

Submitted in partial fulfillment of the  
requirements for the degree of

MASTER OF SCIENCE IN ELECTRICAL ENGINEERING

from the  
NAVAL POSTGRADUATE SCHOOL  
June 1974



## ABSTRACT

Techniques are discussed for Monte-Carlo evaluation of several filters in a digital fire control system performing against an aircraft trajectory developed to simulate a diving attack. The programs developed are designed to allow for future modeling work in a three-dimensional theater of operations with expedient selection of the options that are discussed, and a comprehensive user's guide is provided. Comparison is made between a model of the filter currently used in the MK-86 Digital Fire Control System and some promising combinations of the several options available.



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## I. INTRODUCTION

With the increasing cost and complexity of modern gunfire control systems there exists a correspondingly increasing need for thorough evaluation of simulated performance with accurate models before major commitment is made.

Specifically, the MK-86 Digital Gunfire Control System has come under such scrutiny and this thesis is one in a series designed to define, evaluate and offer alternative approaches to the existing system. While the original motivation for the study in this thesis was to explore alternative techniques to those used in the existing system, the resulting derivations and computer programs lend themselves to other system modeling situations within certain restrictions. The development of this thesis is intended to allow future modeling work to easily take advantage of the results contained herein.

The pattern followed in a typical cycle of the gunfire control system is the acquisition of data by a radar, analysis of the measurements in a filtering section, and the generation of gun orders in a third general section. The entire model should study all of these segments, but in order to simplify the problem each segment may be studied separately. This thesis is designed for studying the filtering section and assumes that input data is available from a





typical radar and that the results would be used by another section to generate the gun orders.

In order to evaluate possible filtering methods for analyzing the radar data it was necessary to generate realistic, simulated radar measurements for the model. Therefore, a hypothetical track was developed with the capacity for variation to provide the necessary input data. In addition, the various special features that are presented in this thesis were organized in the computer simulation for straightforward selection and implementation thus offering maximum flexibility for evaluation of the various filtering methods. This designed-in choice of options was intended, also, to lend the general program to other applicable situations.



## II. PROBLEM DESCRIPTION

### A. KALMAN FILTER THEORY

Sequential estimation is characterized by the serial recursive processing of observations taken in time sequence. The result of every processing cycle is the current best estimate of the vector being estimated. This estimate therefore embodies all observation data up to and including the current observation. As a new observation is made, the current estimate is updated to reflect this most recent data. In such an estimation scheme the calculations are identical in nature from cycle to cycle so they are ideally suited for implementation on a digital computer.

The Kalman filter [1] is a recursive filter of the type applicable to a digital fire control system in which discrete observations are available from the radar. The filter offers the capability of not only generating estimates of the observed system's states but, also, of predicting future system (or plant) states.

The linear discrete system for which the Kalman filter is designed is characterized by the state and output equations

$$\underline{x}(k+1) = \underline{\phi} \underline{x}(k) + \underline{\Delta} \underline{u}(k) + \underline{\Gamma} \underline{w}(k) \quad (1)$$

$$\underline{z}(k) = \underline{H} \underline{x}(k) + \underline{v}(k) \quad (2)$$

where

$\underline{x}(k)$  is the n-dimensional state vector at time  $t = kT$ ,

$\underline{u}(k)$  is the p-dimensional deterministic input vector at time  $t = kT$ ,



$\underline{z}(k)$  is the m-dimensional vector of measurements or observations taken at time  $t = kT$ ,

$\underline{w}(k)$  and  $\underline{v}(k)$  are q-dimensional and m-dimensional noise processes, respectively, at time  $t = kT$ ,

$\underline{\phi}$  is the  $n \times n$  state transition matrix--assumed to be known,

$\underline{H}$  is the  $m \times n$  observation matrix--assumed to be known,

$\underline{\Delta}$  and  $\underline{\Gamma}$  are  $n \times p$  and  $n \times q$  matrices, respectively, which relate the deterministic and non-deterministic forcing terms to the state vector --they are assumed to be known,

$T$  is the time between measurements, and  $k$  is a non-negative integer.

The noise statistics are summarized below

$$E[\underline{v}(k) \underline{v}^T(j)] = \underline{R}(k) \underline{\delta}(k, j) \quad (3)$$

$$\underline{\Gamma} E[\underline{w}(k) \underline{w}^T(j)] \underline{\Gamma}^T = \underline{Q}(k) \underline{\delta}(k, j) \quad (4)$$

$$E[\underline{v}(k) \underline{w}^T(j)] = 0 \text{ for all } (k, j) \quad (5)$$

$$\underline{\delta}(k, j) = \begin{cases} 0 & k \neq j \\ 1 & k = j \end{cases} \quad (6)$$

It is assumed that the initial state is a random variable with known mean and covariance

$$E[\underline{x}(0)] = \underline{\hat{x}}_0, \quad E\{[\underline{x}(0) - \underline{\hat{x}}_0][\underline{x}(0) - \underline{\hat{x}}_0]^T\} = \underline{P}_0 \quad (7)$$

In addition, it is assumed that the measurement noise and initial state are uncorrelated

$$E[\underline{x}(0) \underline{v}^T(k)] = 0 \text{ for all } k \quad (8)$$

and that the forcing noise and initial state are uncorrelated

$$E[\underline{x}(0) \underline{w}^T(k)] = 0 \text{ for all } k \quad (9)$$



The Kalman filter equations are summarized below

$$\underline{G}(k) = \underline{P}(k/k-1) \underline{H}^T [\underline{H} \underline{P}(k/k-1) \underline{H}^T + \underline{R}(k)]^{-1} \quad (10)$$

$$\underline{P}(k/k) = (\underline{I} - \underline{G}(k) \underline{H}) \underline{P}(k/k-1) \quad (11)$$

$$\underline{P}(k+1/k) = \underline{\phi} \underline{P}(k/k) \underline{\phi}^T + \underline{Q}(k) \quad (12)$$

$$\hat{\underline{x}}(k/k) = \hat{\underline{x}}(k/k-1) + \underline{G}(k) [\underline{z}(k) - \underline{H} \hat{\underline{x}}(k/k-1)] \quad (13)$$

$$\hat{\underline{x}}(k+1/k) = \underline{\phi} \hat{\underline{x}}(k/k) + \underline{\Delta} \underline{u}(k) \quad (14)$$

where the notation  $(k/k-1)$  is defined as a condition at time  $t = kT$  given information up to and including time  $t = (k-1)T$ .

The matrices in these equations are

- $\underline{G}(k)$  -  $n \times m$  gain matrix
- $\underline{R}(k)$  -  $m \times m$  covariance matrix of measurement error
- $\underline{P}(k/k)$  -  $n \times n$  covariance matrix of estimation error
- $\underline{I}$  -  $n \times n$  identity matrix
- $\underline{P}(k+1/k)$  -  $n \times n$  prediction covariance matrix
- $\underline{Q}(k)$  -  $n \times n$  covariance matrix of state excitation defined as  $\underline{\Gamma} E[\underline{w}(k) \underline{w}^T(k)] \underline{\Gamma}^T$
- $\hat{\underline{x}}(k/k)$  -  $n \times 1$  optimal (minimum variance) estimates of  $\underline{x}(k)$
- $\underline{z}(k)$  -  $m \times 1$  observation vector
- $\hat{\underline{x}}(k+1/k)$  -  $n \times 1$  optimal predicted value of  $\underline{x}(k)$ .

A block diagram of the discrete plant and Kalman filter is shown in Figure 1.

The Kalman filter takes advantage of all previous state measurements,  $\underline{z}$ , along with their respective error estimates,





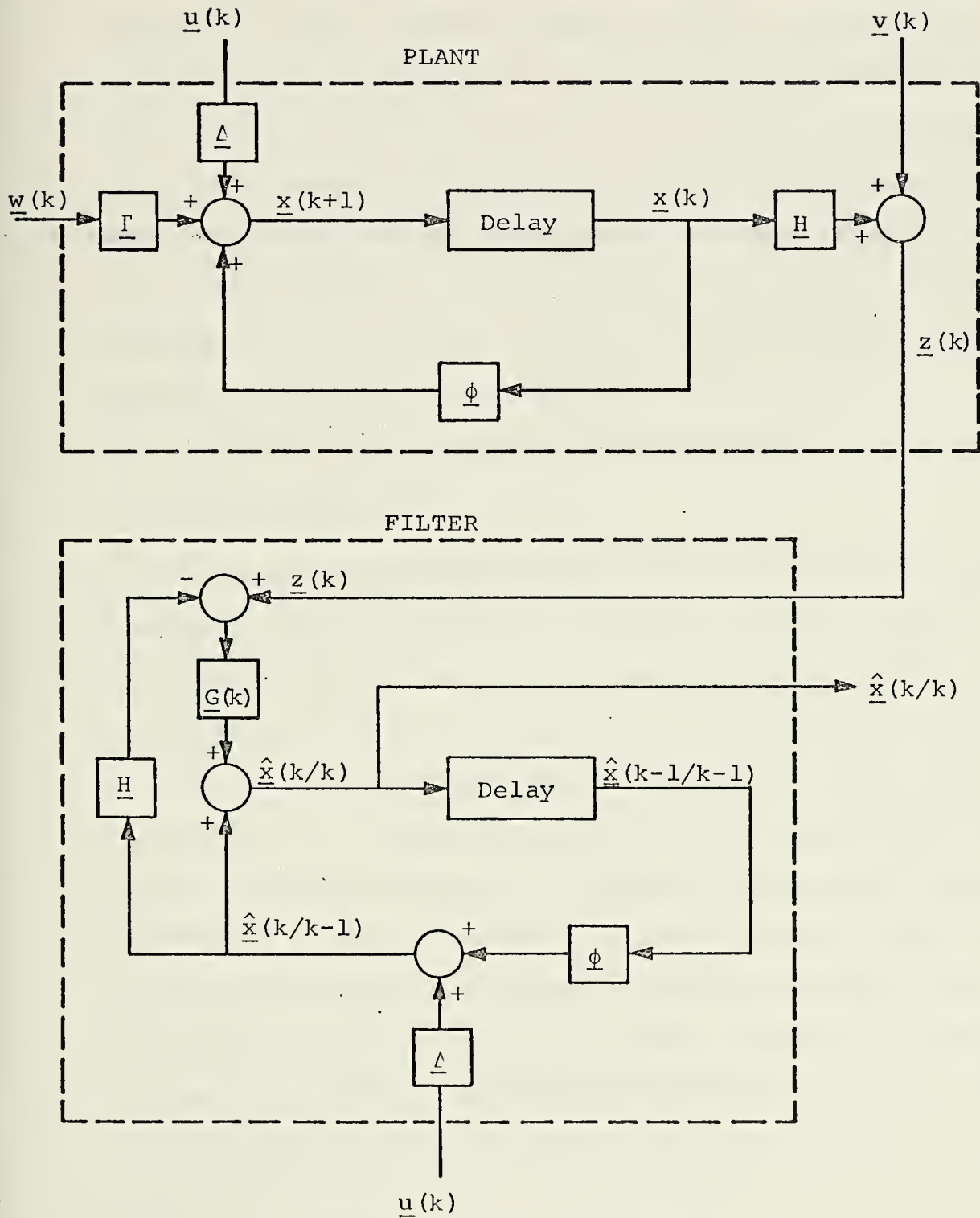


Figure 1. Block Diagram of the Discrete Plant and Kalman Filter.



and predicts ahead what the system states should be based on the state transition matrix,  $\underline{\phi}$ , and any known deterministic forcing  $\underline{u}$ . When a new measurement becomes available for current state estimation the filter takes the predicted state vector from the previous iteration,  $\hat{\underline{x}}(k/k-1)$ , and corrects it by some amount depending on the difference between the predicted vector and the measurement vector. Normally the amount of correction is a fraction (less than one) of the difference and is defined by the gain matrix,  $\underline{G}(k)$ , which has been calculated using equations (10) - (12) so that the state estimates yield minimum variance estimates.

#### B. SYSTEM MODELS

In order to apply the Kalman filter to a given situation a model for the plant must be specified. In the case of a fire control system the plant is defined as the target. The motion of the target is characterized by quantities such as position, velocity, acceleration, acceleration rate, etc.

For most applications the plant can be approximated as either a constant-velocity or constant-acceleration target. In terms of a three-dimensional tracking problem with filtering in an orthogonal coordinate system the resultant state vector for the constant-velocity model includes six states --three for position definition and three for velocity. The constant-acceleration model also includes three acceleration states.

In this thesis, for example, the filtering is done in an XYZ Cartesian coordinate system, shown in Figure 2, using



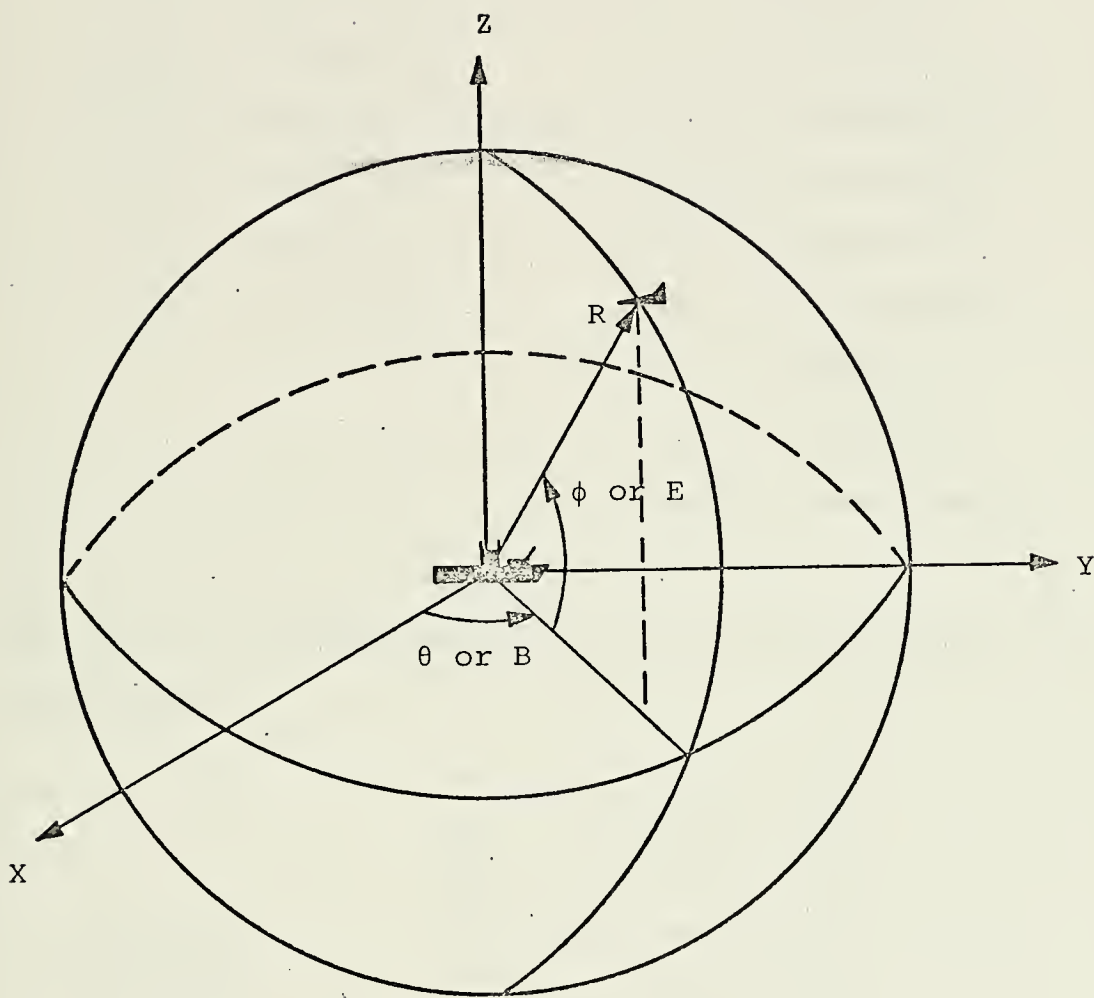


Figure 2. Coordinate System.



both constant-velocity and constant-acceleration models with the resultant state vectors

Constant-velocity Model

$x_1$  x position  
 $x_2$  x velocity  
 $x_3$  y position  
 $x_4$  y velocity  
 $x_5$  z position  
 $x_6$  z velocity

Constant-Acceleration Target

$x_1$  x position  
 $x_2$  x velocity  
 $x_3$  x acceleration  
 $x_4$  y position  
 $x_5$  y velocity  
 $x_6$  y acceleration  
 $x_7$  z position  
 $x_8$  z velocity  
 $x_9$  z acceleration

The state transition matrices,  $\phi$ , for the two models are shown below

Constant-Velocity  $\phi =$  
$$\begin{bmatrix} 1 & T & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & T & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & T \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \quad (17)$$





$$\begin{array}{l}
 \text{Constant} \\
 \text{Acceleration}
 \end{array}
 \underline{\phi} =
 \begin{bmatrix}
 1 & T & \frac{T^2}{2} & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1 & T & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 1 & T & \frac{T^2}{2} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 1 & T & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 1 & T & \frac{T^2}{2} \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & T \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1
 \end{bmatrix}
 \quad (18)$$

For the models used in defining the plant in this thesis the deterministic and non-deterministic forcing terms,  $\underline{\Delta u}(k)$  and  $\underline{\Gamma w}(k)$ , are assumed to be zero.

### C. MONTE-CARLO SIMULATION

The basic premise upon which computer simulation is justified is that the model used for study accurately portrays the phenomenon as it occurs in practice. In order to assure that this requirement is satisfied repeated independent experiments yielding classical random-sample statistics are necessary. The requirement as it applies to actual simulation exercises implies that those parameters in the model that are variable require alteration in a random manner dictated by the expected randomness in practice.

Monte-Carlo evaluation is the process of satisfying the need for randomness in some of the model parameters and is especially applicable with computer simulation where the



necessary calculating capacity and speed is available. Further discussion of Monte-Carlo techniques and applications may be found in [2].

The following discussion presents the necessary information for deciding how many Monte Carlo runs, or iterations, are required to satisfy some desired statistical confidence in the simulation. The source of theoretical information against which empirical sampling of available random number generators are compared is [3].

The use of available random number generators, such as GAUSS or NORMAL for normally distributed random numbers, is the most straightforward method for obtaining the desired randomness to introduce into the simulation. Two factors that must be considered in evaluating the performance of a random number generator are (1) the statistics of the empirical samples compared with the desired statistics and (2) the confidence for which the results of (1) are guaranteed.

For the type of random numbers needed with the Monte-Carlo simulation of the filtering process in a fire control system the merit of acceptance is based on the actual mean and standard deviation (or variance) of the normally distributed random numbers compared to the desired, or specified, values of these quantities. Considering the mean of the distribution first, several variables can be defined that are used in calculating a desired number of samples to achieve an acceptable mean. These include



- d - a margin of acceptable error of the sample mean expressed as a percentage of the desired standard deviation, such as 0.1 (10% of the standard deviation),
- $\alpha$  - the risk involved in the result, also signifying (100 -  $\alpha$ ) per cent confidence in the result,
- $\beta$  - defined as  $(1 - \alpha/2)$ .

Entering Figure 3 with the value of  $\beta$  the variable  $t$  is defined. The number of samples necessary to give a sample mean within  $d$  of the desired mean with a known risk of  $\alpha$  is given by

$$n = \frac{t^2 \sigma^2}{d^2} \quad (19)$$

It can be seen that if  $d$  is expressed as a percentage of  $\sigma$ , the actual value of the standard deviation is not required to solve for the number of samples needed.

To determine the number of samples necessary for desired statistics of the standard deviation refer to Figure 4. Here again, the margin of acceptable error is expressed as a percentage of the standard deviation, and this percentage is the parameter  $P$  scaled along the abscissa. The ordinate is labeled "degrees of freedom" and represents the number of samples needed to achieve the various confidences plotted as bars across the graph.

Working through an example shows how the curves are used. Suppose it is desired to execute enough Monte-Carlo runs to assure 90% confidence in sample statistics having an actual mean and an actual standard deviation within 10% of its desired value. Considering the mean first:



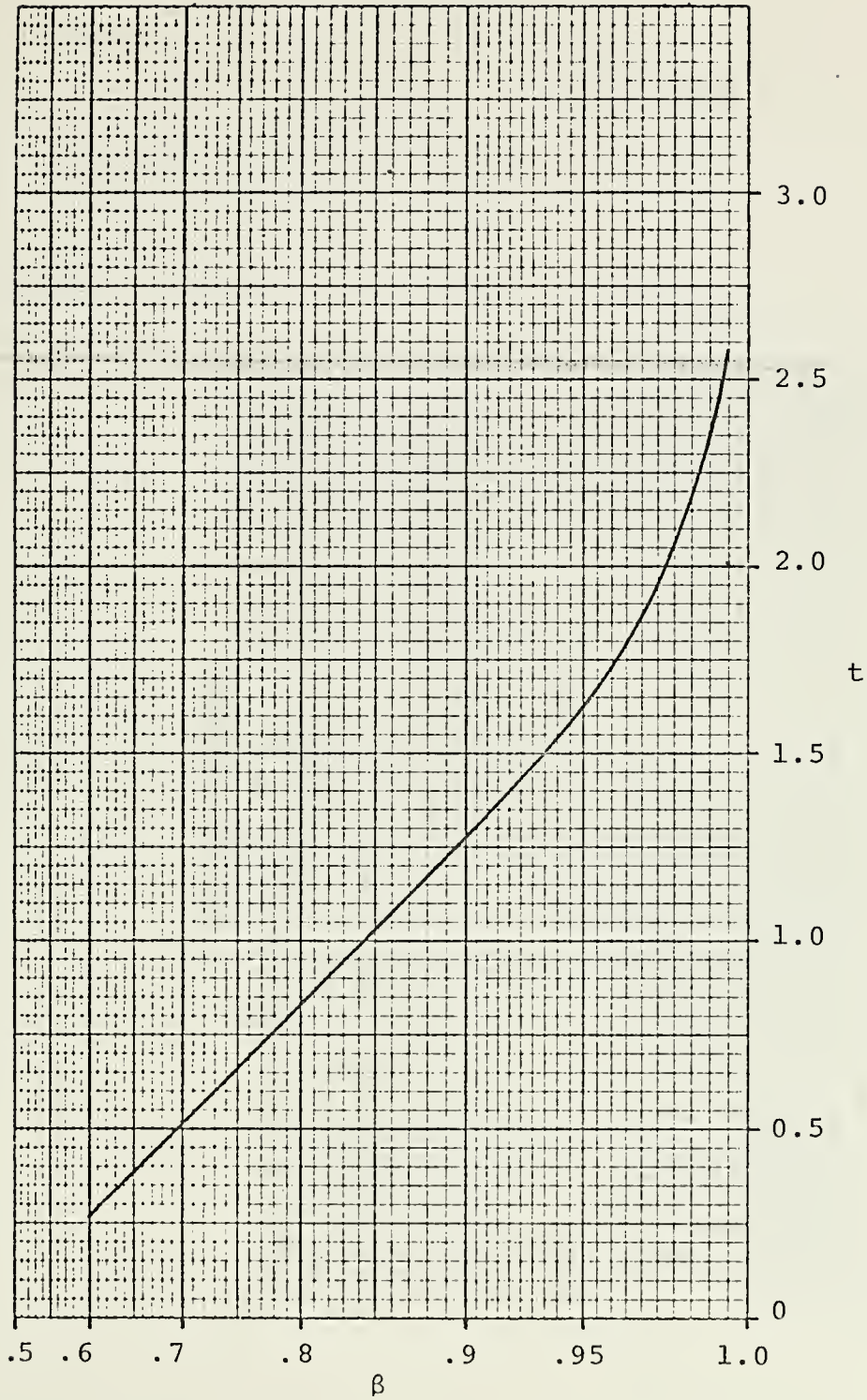


Figure 3.  $t - \beta$  Diagram.





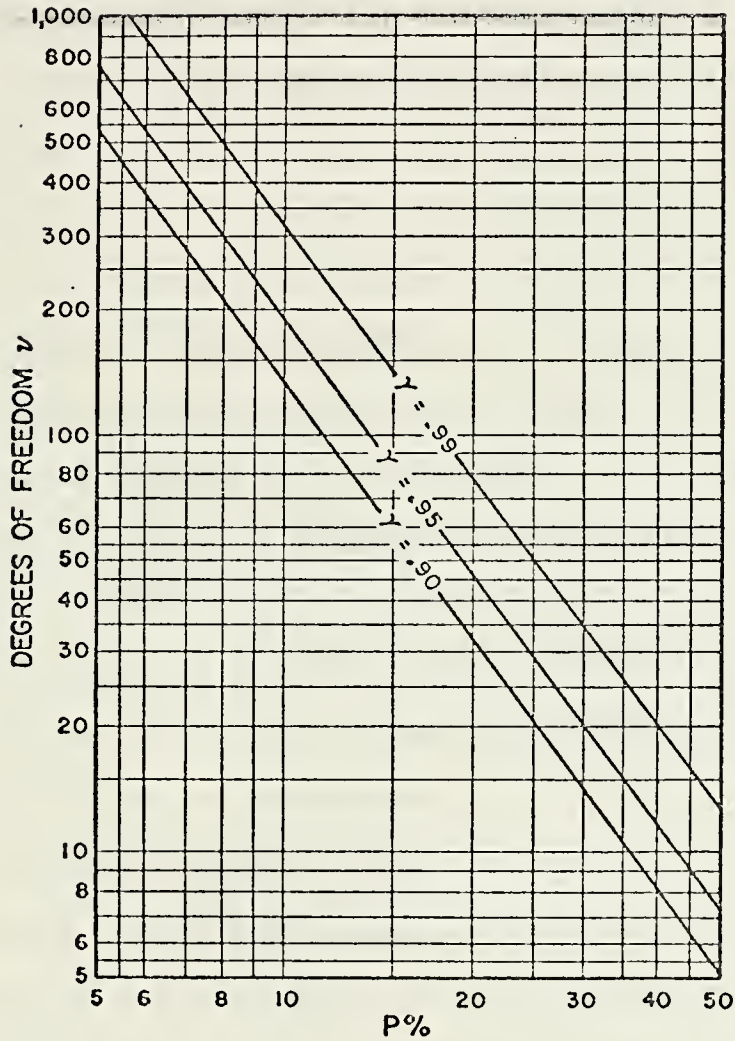


Figure 4. Number of Samples Required to Estimate the Standard Deviation within P Percent of Its True Value with Confidence Coefficient  $\gamma$ .



$$\beta = 1 - \frac{\alpha}{2} = 1 - \frac{.1}{2} = .95 . \quad (20)$$

From Figure 3 the value of  $t$  is found to be 1.645, and

$$n = \frac{(1.645)^2 \sigma^2}{(.1)^2 \sigma^2} = 270 . \quad (21)$$

Referring to Figure 4 to find the number of samples required for the desired standard deviation results gives  $n = 140$ .

Therefore, in order to satisfy all restrictions 270 samples would be required.

Verification of these theoretical results involved sampling two subroutines available in the NPS Computer Facility Sublibrary, GAUSS and NORMAL. The approach used was to sample each number generator for 1000 ensembles of several numbers of samples in each ensemble. The number 1000 was chosen to obtain steady-state values in the statistics determined for the runs of length 50, 75, 100, 150, 200, 500, 700, 1000, and 5000 samples each. A summary of the results is included as Table I,<sup>1</sup> and the FORTRAN program for the study may be found in Appendix A.

It may be noted in Table I that the results for the normally distributed random number generator GAUSS are incomplete, and this is a consequence of the substantially greater execution time required than for NORMAL. Whereas NORMAL used

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<sup>1</sup>Theoretical results are limited in accuracy to one's ability to read Figure 3. Also, percent of standard deviations within criteria of desired standard deviations are not shown for theoretical calculations because (1) Figure 4 is not complete and (2) the limiting factor in satisfactory M-C simulation is the satisfaction of requirements for the mean, not for the standard deviation.



TABLE I  
COMPARISON OF EMPIRICAL AND THEORETICAL  
NORMALLY DISTRIBUTED RANDOM NUMBERS

| Number of Samples | Percent Criteria | Theoretical<br>$\bar{M}^a$ | NORMAL      |        | GAUSS       |        |
|-------------------|------------------|----------------------------|-------------|--------|-------------|--------|
|                   |                  |                            | $\bar{M}^a$ | $SD^b$ | $\bar{M}^a$ | $SD^b$ |
| 50                | 5                | 28.0                       | 27.0        | 32.7   | 27.4        | 37.2   |
|                   | 10               | 52.4                       | 51.9        | 63.7   | 52.4        | 64.8   |
|                   | 15               | 71.2                       | 69.0        | 82.8   | 72.3        | 83.9   |
|                   | 20               | 84.2                       | 83.9        | 94.3   | 85.7        | 94.5   |
|                   | 25               | 92.4                       | 92.9        | 98.1   | 94.2        | 98.4   |
| 75                | 5                | 33.6                       | 33.7        | 42.2   | 34.7        | 44.2   |
|                   | 10               | 62.0                       | 60.4        | 75.1   | 60.6        | 76.6   |
|                   | 15               | 80.8                       | 80.2        | 82.5   | 81.3        | 93.2   |
|                   | 20               | 91.6                       | 91.9        | 98.5   | 92.0        | 99.0   |
|                   | 25               | 96.8                       | 97.0        | 99.8   | 97.6        | 99.9   |
| 100               | 5                | 38.6                       | 39.0        | 49.8   | 37.3        | 50.7   |
|                   | 10               | 68.0                       | 67.9        | 83.2   | 66.9        | 83.3   |
|                   | 15               | 86.8                       | 86.3        | 96.0   | 87.1        | 97.2   |
|                   | 20               | 95.4                       | 95.1        | 99.2   | 95.7        | 99.6   |
|                   | 25               | 98.8                       | 98.3        | 100.0  | 99.1        | 100.0  |
| 150               | 5                | 47.0                       | 45.2        | 60.0   | 45.3        | 60.8   |
|                   | 10               | 78.0                       | 78.3        | 89.6   | 78.2        | 92.1   |
|                   | 15               | 93.4                       | 91.9        | 99.0   | 94.2        | 99.3   |
|                   | 20               | 98.4                       | 98.0        | 99.9   | 99.0        | 100.0  |
|                   | 25               | 99.8                       | 99.7        | 100.0  | 99.9        | 100.0  |
| 200               | 5                | 52.0                       | 52.7        | 66.8   | 52.6        | 68.7   |
|                   | 10               | 84.2                       | 82.2        | 95.3   | 83.6        | 95.2   |
|                   | 15               | 96.4                       | 96.0        | 99.8   | 96.3        | 99.9   |
|                   | 20               | 99.6                       | 99.4        | 100.0  | 99.1        | 100.0  |
|                   | 25               | 100.0                      | 99.9        | 100.0  | 99.9        | 100.0  |
| 500               | 5                | 73.4                       | 75.3        | 87.0   | 72.6        | 88.9   |
|                   | 10               | 97.0                       | 96.7        | 99.9   | 97.0        | 99.9   |
|                   | 15               | 100.0                      | 99.9        | 100.0  | 99.9        | 100.0  |
|                   | 20               | 100.0                      | 100.0       | 100.0  | 100.0       | 100.0  |
| 700               | 5                | 81.4                       | 81.0        | 93.5   | 81.5        | 93.7   |
|                   | 10               | 99.2                       | 99.6        | 100.0  | 98.6        | 100.0  |
|                   | 15               | 100.0                      | 100.0       | 100.0  | 100.0       | 100.0  |
| 1000              | 5                | 89.0                       | 90.4        | 96.2   | -           | -      |
|                   | 10               | 100.0                      | 99.9        | 100.0  | -           | -      |
|                   | 15               | 100.0                      | 100.0       | 100.0  | -           | -      |
| 5000              | 5                | 100.0                      | 100.0       | 100.0  | -           | -      |

Note (a) - Percent of Means within percent criteria of desired standard deviation

Note (b) - Percent of Standard Deviations within same percent criteria as (a)



approximately 25 minutes of execution time to carry out the entire analysis, GAUSS required 90 minutes to evaluate up through 700 sample ensemble runs. For this reason NORMAL is recommended when normally distributed random numbers are desired.





### III. DEVELOPMENT OF MONTE-CARLO SIMULATION PROGRAM (MCSP)

#### A. TRACK MODEL

In order to provide realistic data for testing filter performance against an airborne target it was necessary to develop a model based on a conventional dive bomb approach by a modern aircraft against a seaborne target. From knowledgeable sources the parameters of such an approach were obtained and, in accordance with the 4-Hz sampling rate of the radar, used to develop a simulation including such features as a constant-acceleration dive, a pull-out following ordinance release, and evasive maneuvers while withdrawing in a gradual climb. The individual segments of the 233-point track will be discussed as well as the methods one may use to develop other tracks.

While in actual operation a radar measures range, bearing, and elevation (RBE), and perhaps rates as well, convenience dictated development of the track in an XYZ coordinate system centered at the ship, or radar,<sup>2</sup> as shown in Figure 2. Such parameters as ship's speed, bearing, roll, pitch, etc. were not considered in the simulation to avoid model complexity. These quantities, while offering potential sources of error to any overall tracking system, were avoided to concentrate on the filter performance and not the various correction factors applied to input measurement data.

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<sup>2</sup>The radar and the gun are assumed to be at the center of the ship, or origin of coordinate system, and at sea level.



Satisfying the requirement for randomness in the track data points, the variable factors mentioned earlier for Monte-Carlo simulation, involved, when using one basic track, the addition of randomly distributed Gaussian measurement noise to every point. An alternative approach might have been to run the simulation with many independent tracks randomly distributed about the expected track, but this would have caused much additional complexity and increased execution time in the simulation.

In order to realistically generate noisy measurements the track's XYZ data points, after being read by the Monte-Carlo Simulation Program (MCSP), are converted to RBE, the desired Gaussian measurement noise with zero mean and appropriate standard deviation is added to each data point, conversion back into the same XYZ coordinate system takes place, and the data is available as simulated coordinate-converted output from the radar ready for analysis.

The original track is layed out with the aircraft approaching from the +X direction with initial radar contact initiated at a range of 7200 yards and altitude of 4333 yards. Its initial velocity is approximately 250 knots, and in the scenario of the attack has just climbed to the given altitude from a low or medium level approach in preparation for the dive phase of the attack. A summary of the track segments follows:



| Track Points | Segment Description | G Turn | Comments                    |
|--------------|---------------------|--------|-----------------------------|
| 1 - 33       | Initial Dive        | 1      | Assumes 40° Dive Angle      |
| 34 - 113     | Accelerating Dive   | -      | 139.16 yd/sec-249.75 yd/sec |
| 114 - 137    | Pullout             | 4      | To altitude of 958 yards    |
| 138 - 141    | Starboard Turn      | 2      |                             |
| 142 - 145    | Port Turn           | 2      |                             |
| 146 - 149    | Port Turn           | 2      |                             |
| 150 - 157    | Port Turn           | 3.5    | Begin 10° Climb -           |
| 158 - 165    | Starboard Turn      | 3.5    | Velocity of 249.76          |
| 166 - 173    | Port Turn           | 3.0    | yd/sec is maintained        |
| 174 - 181    | Starboard Turn      | 3.0    |                             |
| 182 - 193    | Port Turn           | 3.0    |                             |
| 194 - 205    | Starboard Turn      | 2.5    |                             |
| 206 - 221    | Port Turn           | 2.5    |                             |
| 222 - 233    | Starboard Turn      | 2.0    |                             |

Note that the port turn at points 142 - 145 is followed by another similar turn from points 146 - 149. The need for the additional segment rather than creating just one turn from points 142 - 149 will become clear later in the discussion.

The geometric relationships used to translate the desired track into mathematical terms include both basic, intuitive approaches and some requiring direct application to a given target orientation. These general techniques will now be discussed; the implementation of the particular track segments in a FORTRAN program may be found in Appendix B.

The following sub-sections portray the various possible maneuvers and track orientations used in developing the track for the MCSP.

A.1 The initial conditions include the starting point anywhere in the coordinate system with the velocity vector parallel to an axis. Specified parameters for a turn can include angle of turn, time of execution, tangential velocity,



constant radial acceleration, and final translation with respect to starting point. (Examples are: Points 1 - 33, 138-141, 146 - 149)

The geometrical arrangement is shown in Figure 5, and the following pertinent definitions and equations apply:  $a$  is the radial acceleration,  $vel$  is the tangential velocity, and  $T$  is the time of maneuver; thus

$$a = \frac{32.2}{3} \text{ yds/sec}^2 \cdot G \text{ of turn} = vel^2/R, \quad (22)$$

$$C = (4h(2R-h))^{1/2}, \quad S = R\theta = 2R\sin^{-1}C/2R, \\ T = S/2 \cdot vel. \quad (23)$$

Using the pertinent equations certain parameters of the turn can be specified while assuring that the other parameters are reasonable. This results in values for  $R$ ,  $\theta$ , and the number of track points in the segment and provides incrementally growing  $\Delta\theta$ 's. Then each new track point is given by

$$X_n = X_o \pm R \sin \Delta\theta \quad (24)$$

$$Y_n = Y_o \pm (R - R \cos \Delta\theta). \quad (25)$$

Note: The optional signs above are dependent upon the orientation of the maneuver on the axes.

A.2 The initial conditions include the starting point anywhere with the velocity vector at some angle,  $\tau$ , with respect to an axis. The turn is in the plane of two of the axes with certain parameters known, such as those mentioned in the previous sub-section. Four types of turn are possible in this scenario:





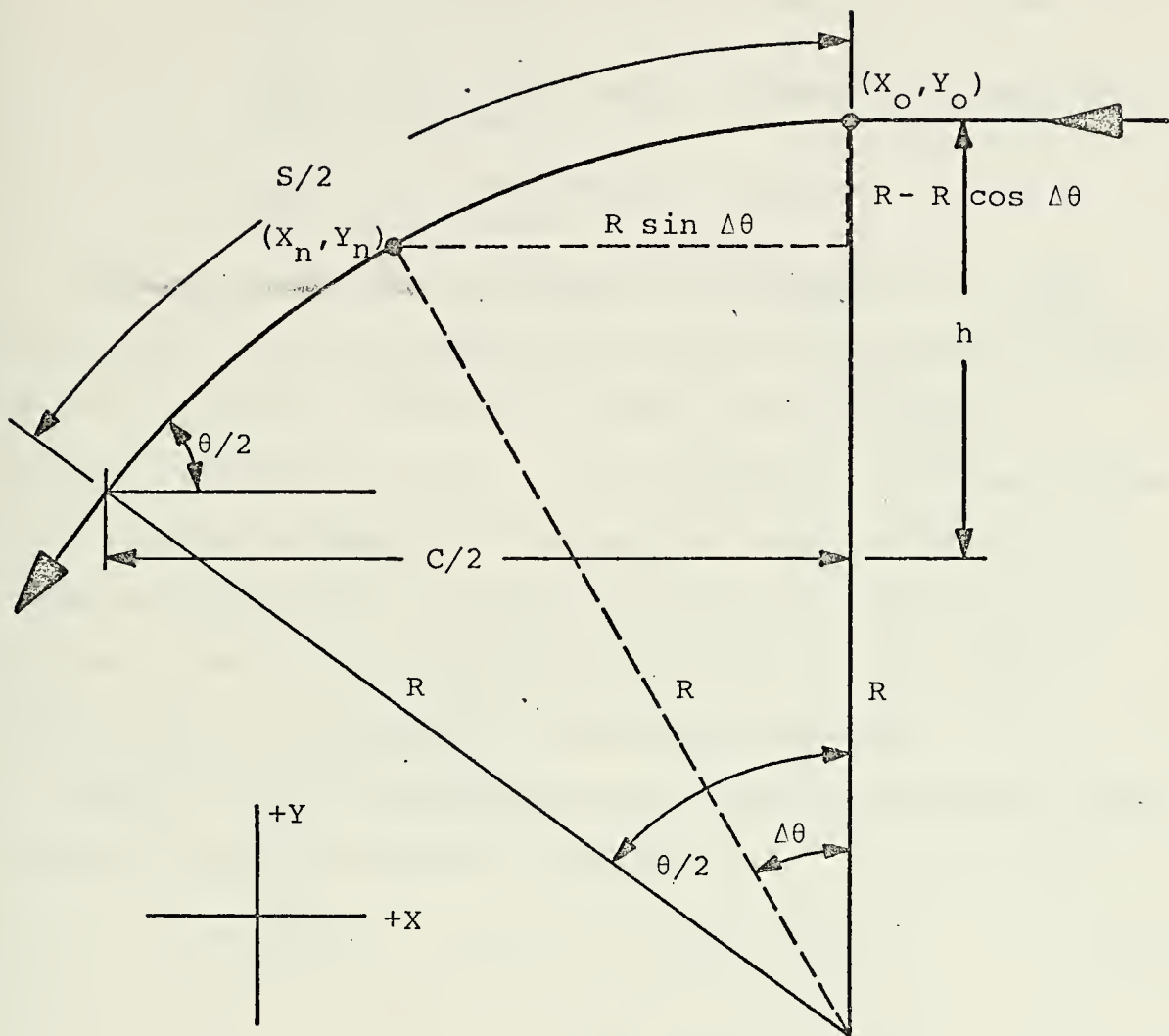


Figure 5. Geometry for Turn with Velocity Vector Initially Parallel to Axis.



- a - Initial velocity vector  $\tau$  degrees to right of axis; right turn,
- b - Initial velocity vector  $\tau$  degrees to right of axis; left turn,
- c - Initial velocity vector  $\tau$  degrees to left of axis; right turn,
- d - Initial velocity vector  $\tau$  degrees to left of axis; left turn.

The same pertinent equations as presented in A.1 are valid here, and the geometrical layouts are shown in Figures 6 and 7. Figure 6 applies to cases a and d above, and Figure 7 to cases b and c. For comparison all cases will be based on the assumption of an initial velocity vector at  $\tau$  degrees with respect to the X axis and the turn in the XY plane. However, the procedure is directly applicable to other axes and planes with similar orientation.

The following development for obtaining the desired quantities  $\Delta\phi$  and  $C'$  applies to Figure 6.

$$\alpha = \frac{180-\Delta\theta}{2}, \lambda = \beta-\alpha, \beta = 180 - (\tau+\Delta\theta) \quad (26)$$

$$\therefore \lambda = 180 - \tau - \Delta\theta - 90 + \frac{\Delta\theta}{2} = 90 - \tau - \frac{\Delta\theta}{2} \quad (27)$$

and 
$$\Delta\phi = 90 - \lambda = \tau + \frac{\Delta\theta}{2} \quad (28)$$

Also, it can be seen that  $C' = 2 R \sin (\Delta\theta/2)$ .

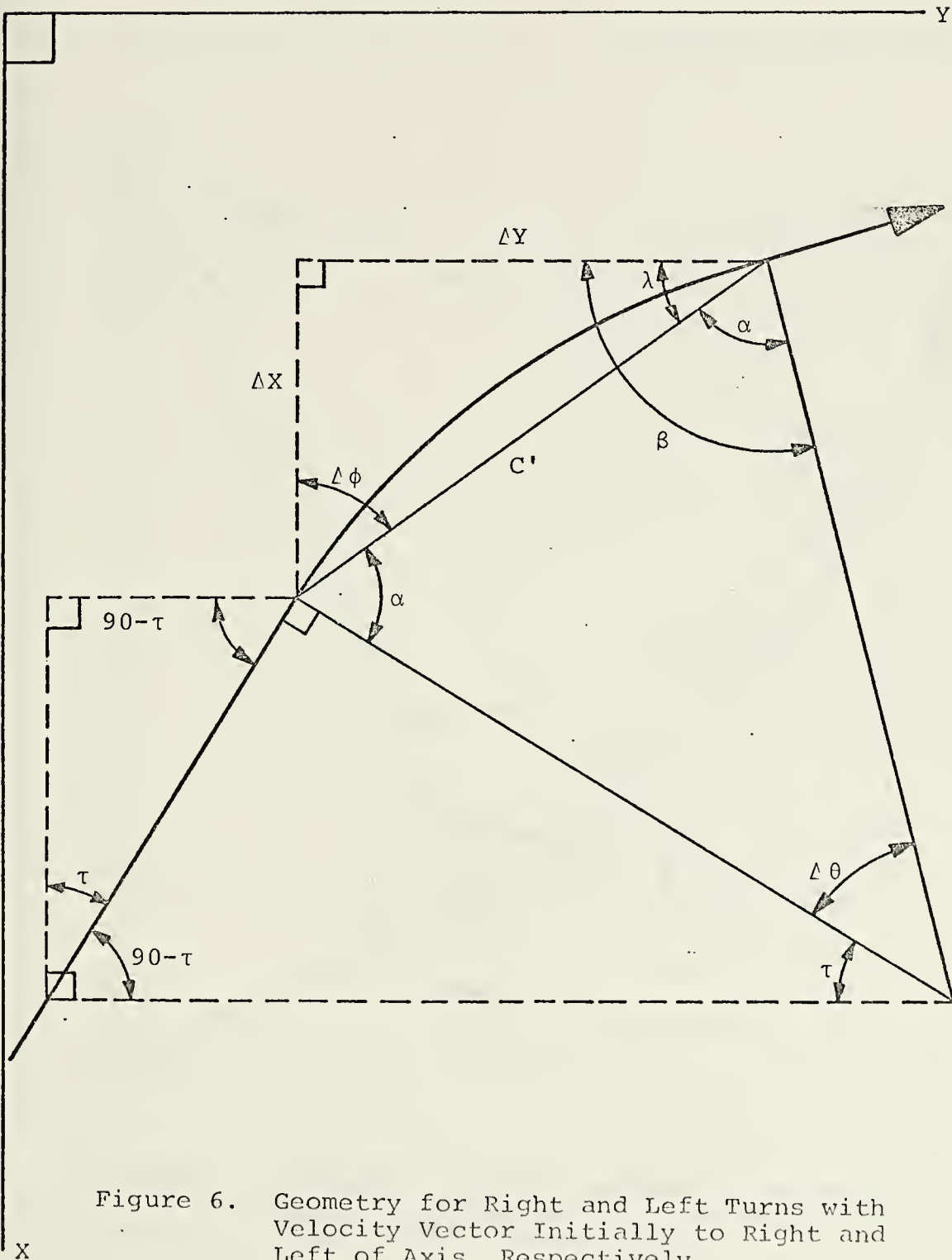
Similarly, the following development applies to Figure 7.

$$\alpha = \frac{180-\Delta\theta}{2}, \beta = 90 + (90-\tau) = 90 - \frac{\Delta\theta}{2} \quad (29)$$

$$\Delta\phi = \beta - \alpha = 180 - \tau - 90 + \frac{\Delta\theta}{2} = 90 - \tau + \frac{\Delta\theta}{2} \quad (30)$$

Again, it can be seen that  $C' = 2 R \sin (\Delta\theta/2)$ .







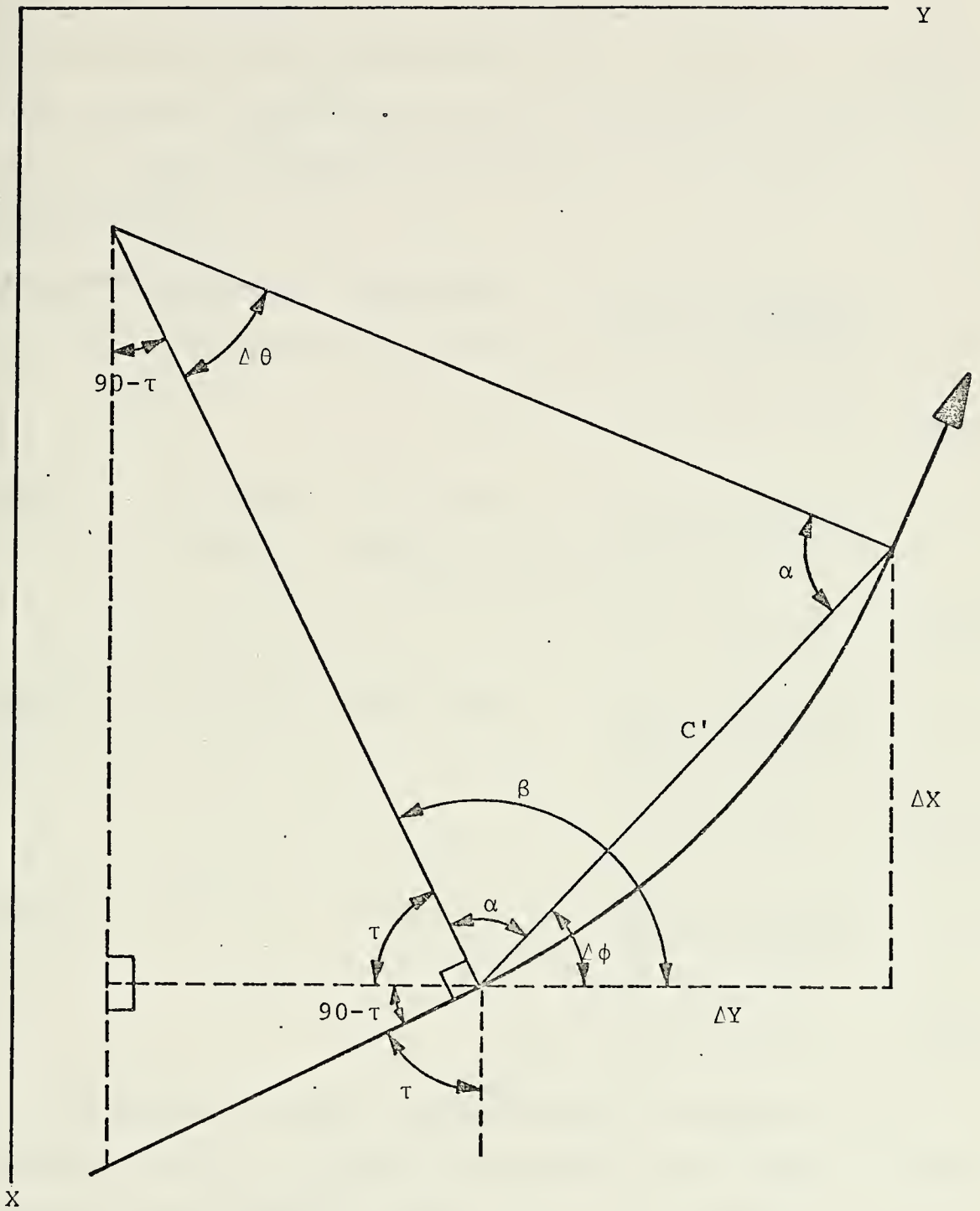


Figure 7. Geometry for Right and Left Turns with Velocity Vector Initially to Left and Right of Axis, Respectively.









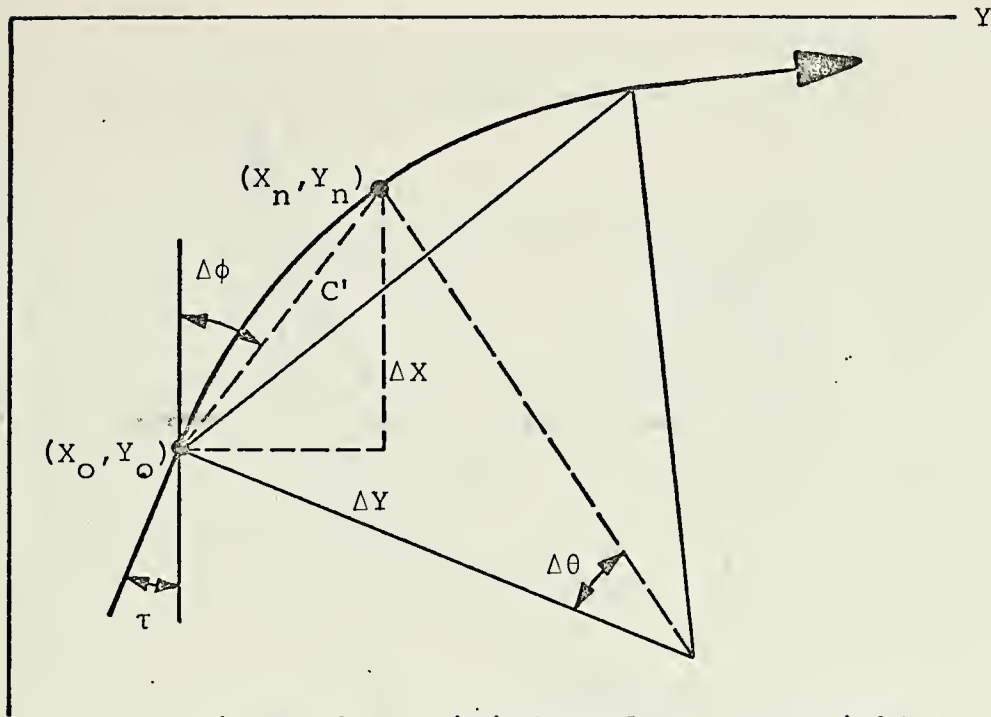


Figure 8. Initial Angle,  $\tau$ , to Right of Axis - Right Turn.

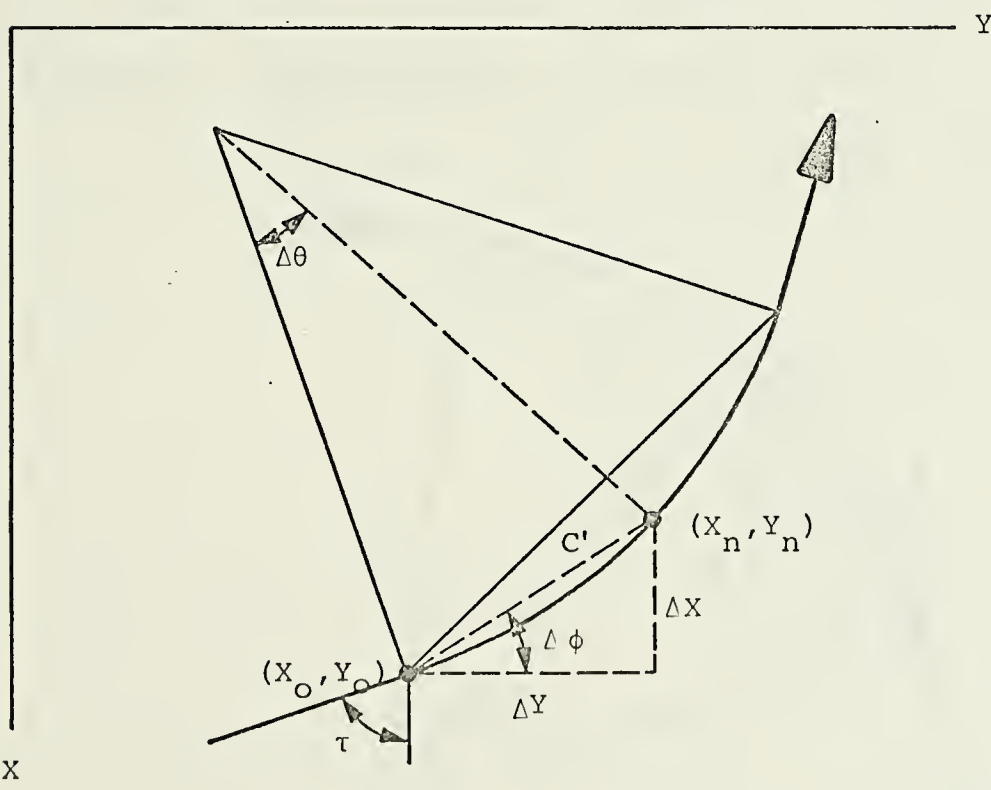


Figure 9. Initial Angle,  $\tau$ , to Right of Axis - Left Turn.



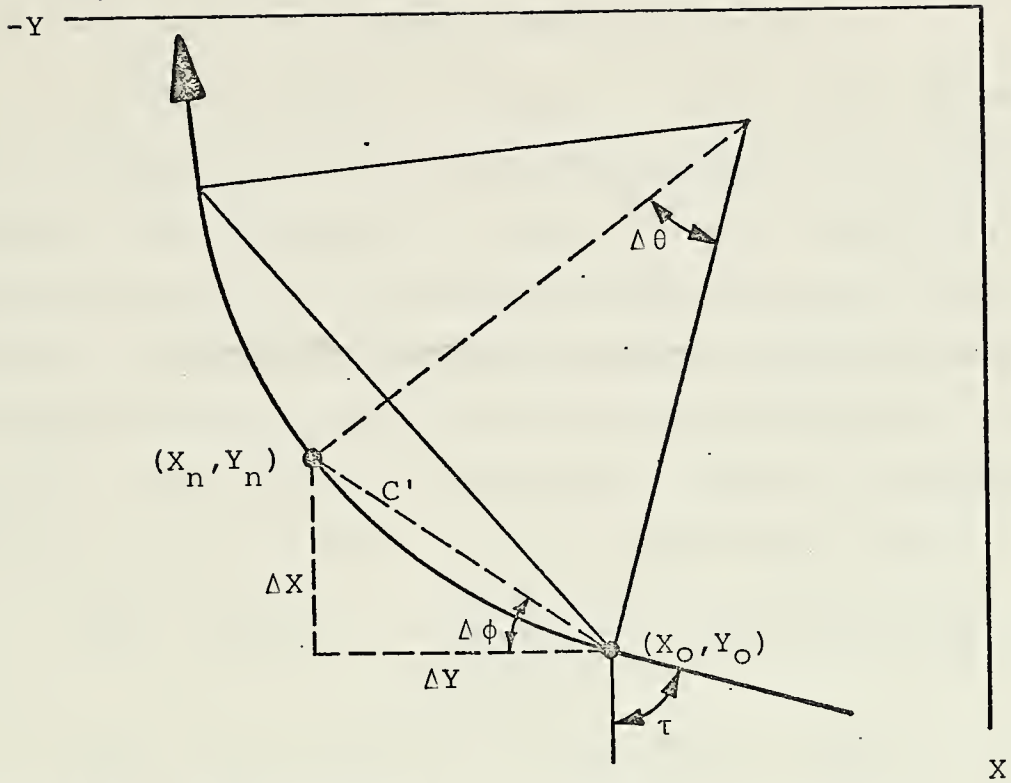


Figure 10. Initial Angle,  $\tau$ , to Left of Axis - Right Turn.

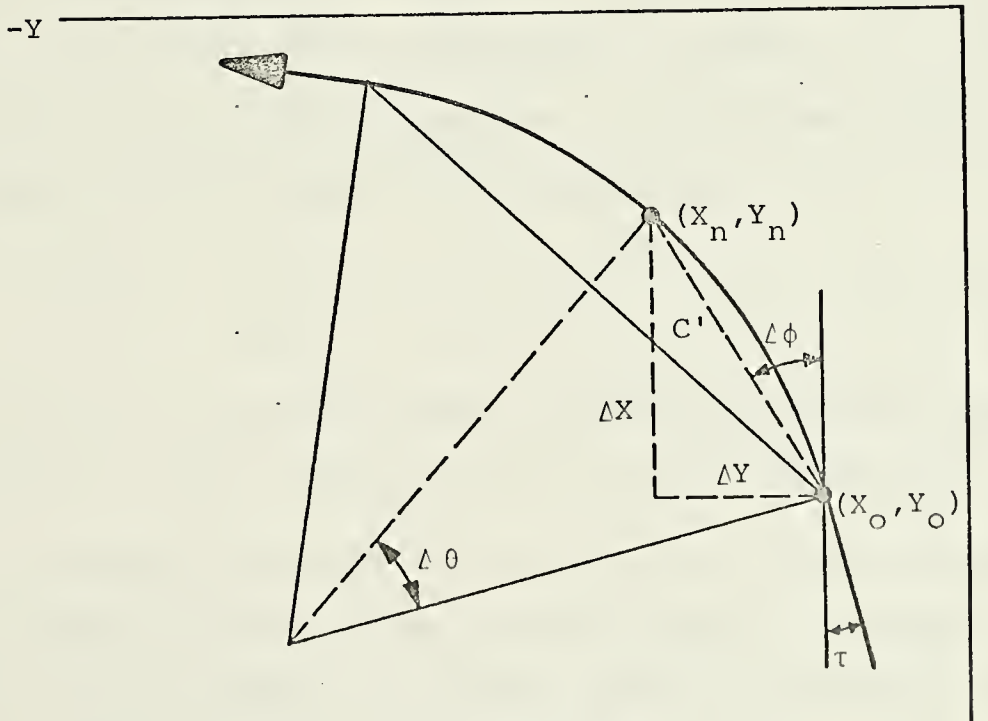


Figure 11. Initial Angle,  $\tau$ , to Left of Axis - Left Turn.



coordinates may begin to change in directions opposite to that discussed in the scenario described above.

A.3 Initial conditions include a starting point anywhere in the coordinate system with constant-acceleration motion in a straight line. (Example: Points 34 - 113) Specified parameters may include initial and final velocity, time of execution, acceleration (assumed constant), angle of segment with respect to some plane, and final translation from starting point. The geometric arrangement is shown in Figure 12, and the pertinent kinematic relationships follow below

$$\text{Final Velocity} = \frac{2 \cdot S}{\left( \begin{array}{c} \text{Time of} \\ \text{Execution} \end{array} \right)} - \text{Initial Velocity}, \quad (35)$$

$$\text{Final Vel.} = \text{Init. Vel.} + \text{Acceleration} \cdot \text{Time}, \quad (36)$$

$$Z \text{ Translation} = S \cdot \sin \alpha, \quad (37)$$

$$X \text{ Translation} = S \cdot \cos \alpha, \quad (38)$$

$$S = \frac{1}{2} (\text{Init. Vel.} + \text{Final Vel.}) \cdot \text{Time}. \quad (39)$$

Using these equations to define a track segment, the incremental displacements can be defined as:

$$\Delta X = \Delta S \cdot \cos \alpha, \quad (40)$$

$$\Delta Z = \Delta S \cdot \sin \alpha. \quad (41)$$

Then each point along the track is defined by applying these incremental changes relative to the previous track point.

A.4 The last situation to be discussed is that applying to the evasive retreat of the aircraft from the seaborne target. In this scenario a gradual climb is initiated after passing over the ship, and the climb angle of  $\beta$  degrees





requires an additional perspective into the aircraft's motion. Using the techniques discussed previously in subsection A.2 the aircraft's track was defined in the XY plane of the coordinate system. If, however, the incremental Z motion is such as to give a  $\beta$  climb angle the apparent effect in the plane of the aircraft is to cause an increasing radially accelerating turn while maintaining the same constant-acceleration turn in the XY plane. It can be seen in Figure 13 that in the plane of the aircraft the radius of curvature decreases from  $R_1$  to  $R_2$ ; therefore, the radial acceleration increases as it is inversely related to the radius in a constant velocity model by  $R = \text{Vel}^2 / \text{Acceleration}$ . This apparent complication is justifiable in the context of actual operation in that an aircraft might, in fact, execute a turn with an increasing amount of radial acceleration. This causes no problems in the simulation as long as the angle,  $\tau$ , between the appropriate axis and the aircraft's velocity vector is continuously updated in order to correctly initiate the succeeding maneuver.

Two plots of the aircraft motion throughout the entire maneuver are included as Figures 14 and 15. Figure 14 depicts the XZ plane, and Figure 15 shows a top view of the XY plane.

## B. TRACK OPTIONS

The groundwork for the track development having been discussed, next follows a description of the options available in using this or any other track in the MCSP. The discussion



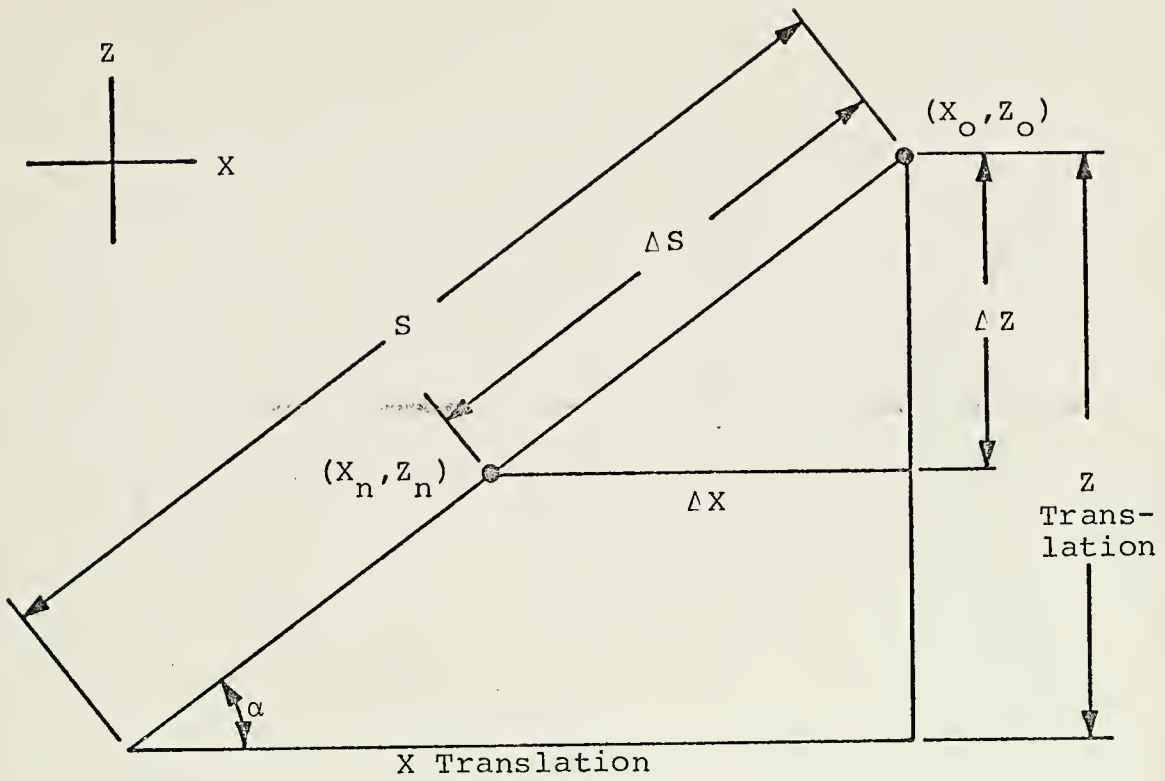


Figure 12. Constant Acceleration Motion in a Straight Line.

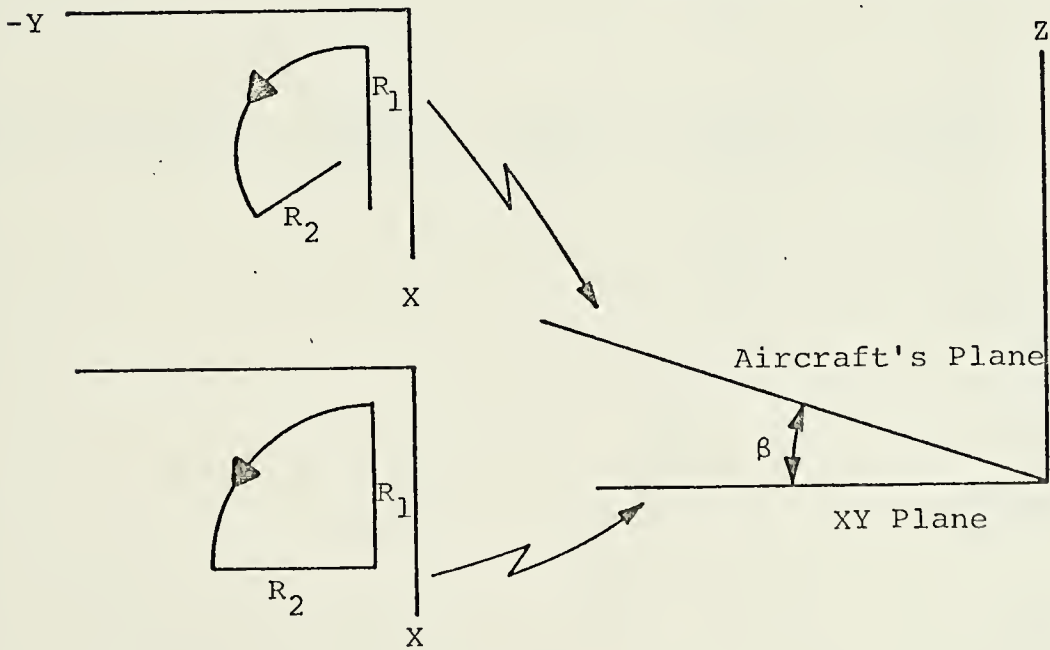
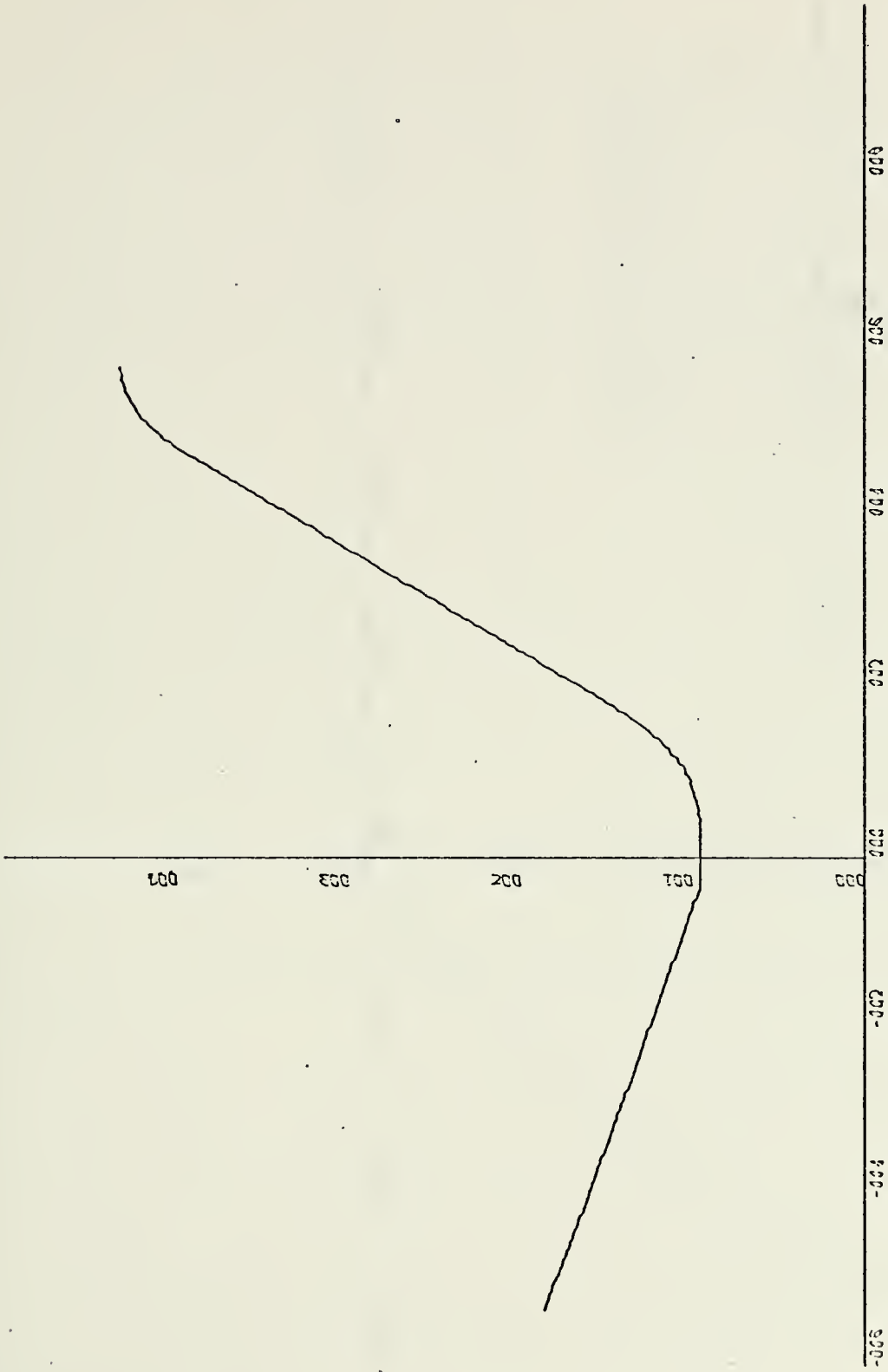


Figure 13. Constant Radial Acceleration Turn with Climb Angle of  $\beta$ .

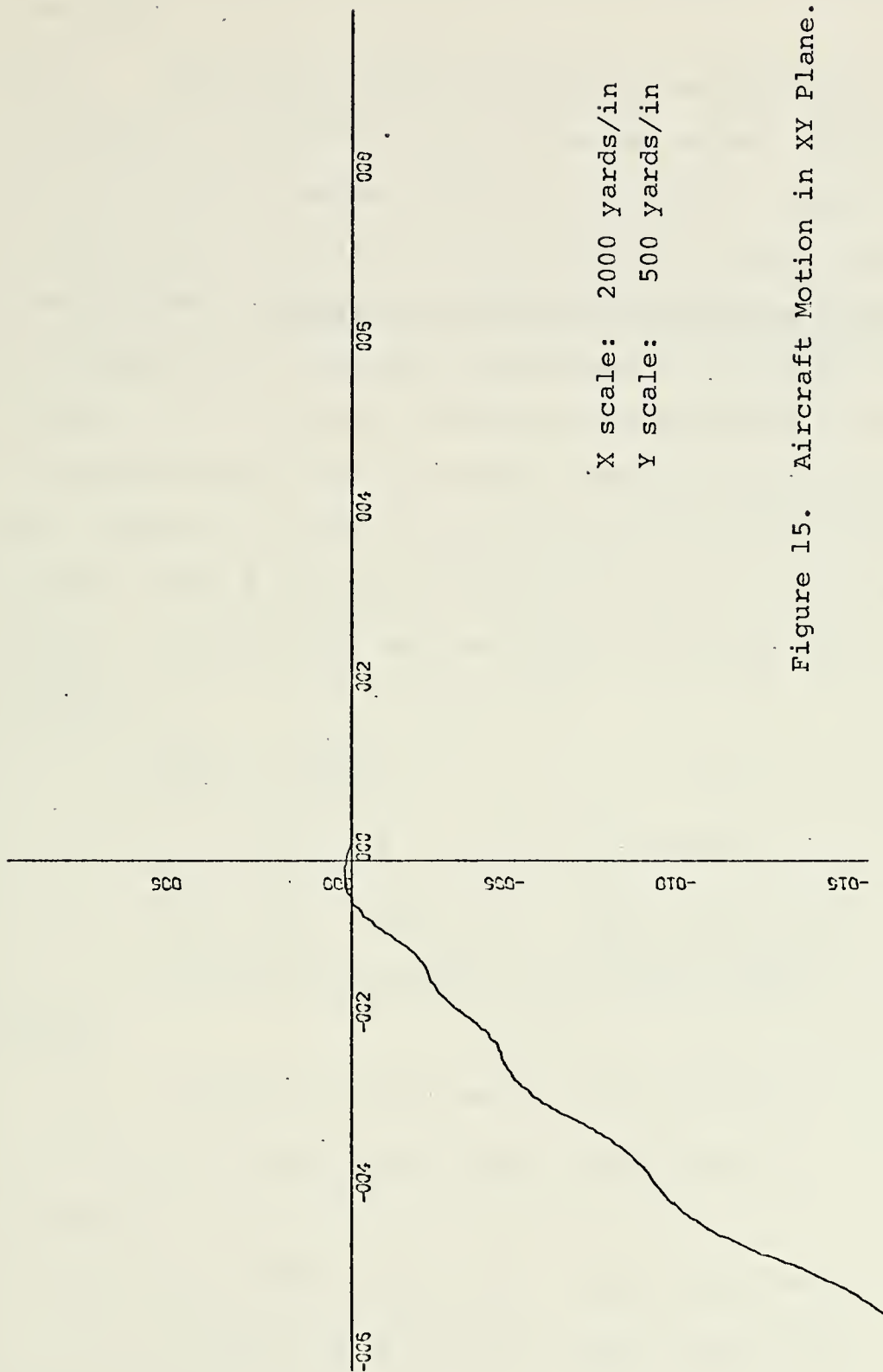




X scale: 2000 yards/in  
 Z scale: 1000 yards/in

Figure 14. Aircraft Motion in XZ Plane.





X scale: 2000 yards/in  
 Y scale: 500 yards/in

Figure 15. Aircraft Motion in XY Plane.





of implementing the desired options follows in a later section.

B.1 The first option available is the selection of direction of approach of the airborne target relative to the coordinate system centered at the ship. This rotational capability is achieved by inputting the original track in XYZ coordinates, converting into RBE, and adding a desired radial increment to the bearing coordinate of every point shown in Figure 16. If track velocities are also input to the program (an option that is covered later), it can be seen that the component in the Z direction is not affected while those in the X and Y directions are. Following is a derivation to handle this problem, and the vector presentation is shown in Figure 17.

$$|\text{VEL}'| = |\text{VEL}| \quad (42)$$

$$\dot{X} = -\text{VEL} \cdot \cos \beta \quad \dot{Y} = \text{VEL} \cdot \sin \beta \quad (43)$$

$$\dot{X}' = -\text{VEL}' \cdot \cos (\beta - \Delta) = -\text{VEL} \cdot \cos (\beta - \Delta) \quad (44)$$

$$\dot{Y}' = \text{VEL}' \cdot \sin (\beta - \Delta) = \text{VEL} \cdot \sin (\beta - \Delta) \quad (45)$$

Using

$$\cos (\beta - \Delta) = \cos \beta \cos \Delta + \sin \beta \sin \Delta \quad (46)$$

$$\sin (\beta - \Delta) = \sin \beta \cos \Delta - \cos \beta \sin \Delta \quad (47)$$

We have

$$\dot{X}' = -\text{VEL} \cdot \cos \beta \cos \Delta - \text{VEL} \cdot \sin \beta \sin \Delta \quad (48)$$

$$\dot{Y}' = \text{VEL} \cdot \sin \beta \cos \Delta - \text{VEL} \cdot \cos \beta \sin \Delta \quad (49)$$

And finally,



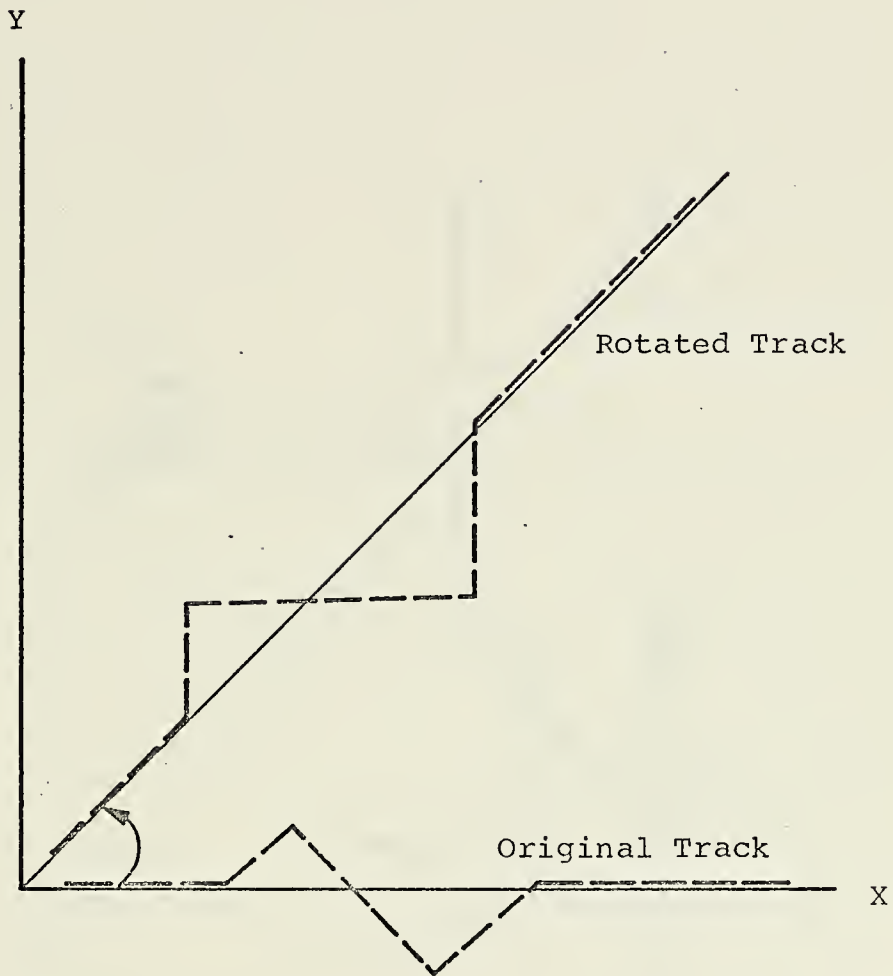


Figure 16. Rotation of Track.



Y

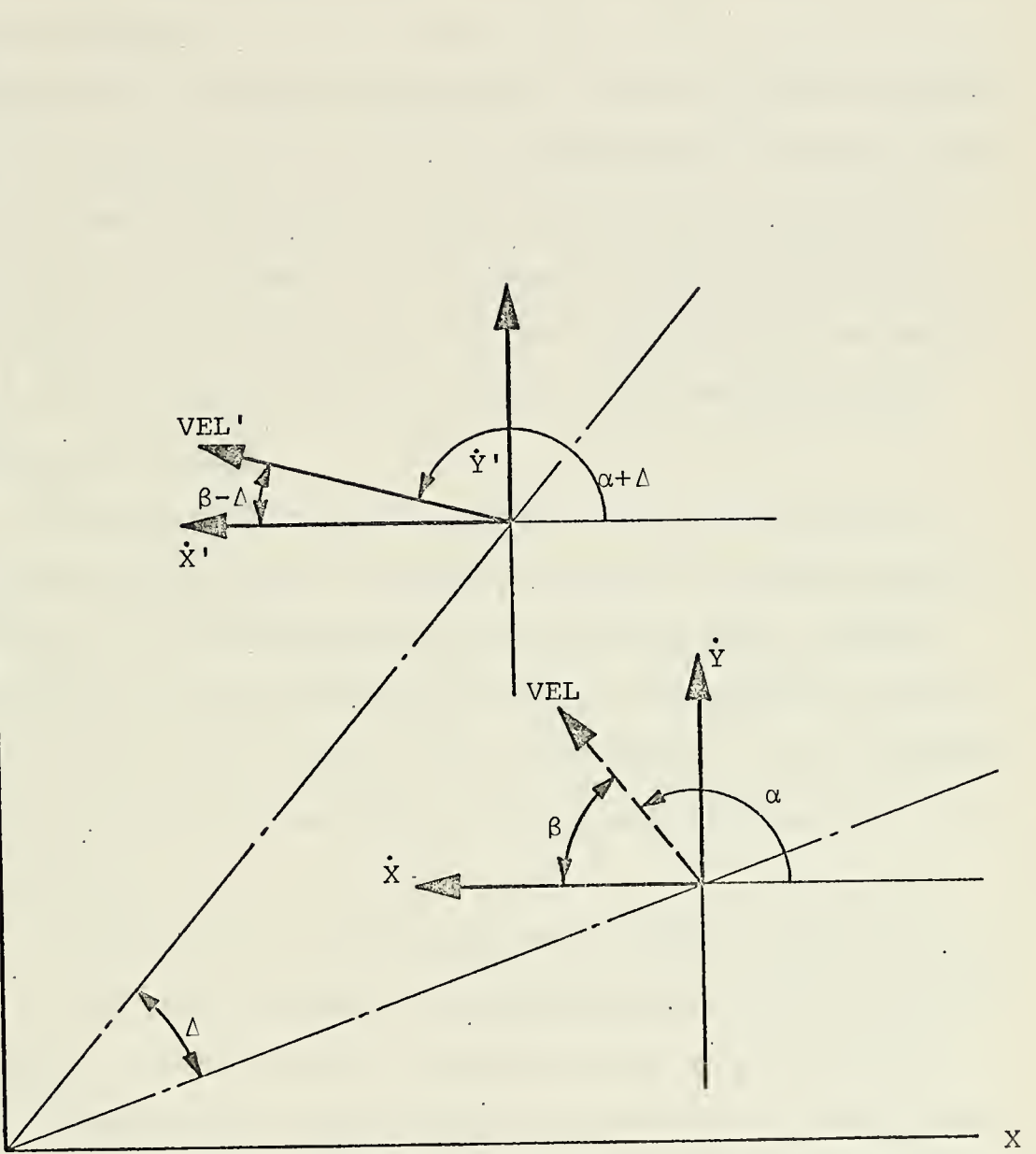


Figure 17. Velocity Vectors for Track Rotation.



$$\dot{X}' = \dot{X} \cos \Delta - \dot{Y} \sin \Delta \quad (50)$$

$$\dot{Y}' = \dot{Y} \cos \Delta + \dot{X} \sin \Delta \quad (51)$$

where  $\dot{X}$  and  $\dot{Y}$  are the original track velocities and  $\Delta$  is the angle of rotation.

The MCSP is arranged to provide a choice of the following angles of approach:  $0^\circ$ ,  $45^\circ$ ,  $90^\circ$ , and  $202.5^\circ$ . If other than these choices is desired the MCSP can be easily altered by referring to the segment of the program several statements after label 13 in the listing of the MCSP included as Appendix C. The variable DEL can be set to the desired radial increment in radians.

B.2 The second option available is the translation of all of the track points by desired amounts for each coordinate axis. The implementation is discussed later, but the method simply involves adding the desired increments to all of the original track points' XYZ coordinates. The velocity components at each track point are not affected by the translation.

B.3 A third option is the choice of adding measurement noise to the track points. If requested, the noise is added to the RBE coordinates and conversion back into noisy XYZ occurs. This noise, as now exists in the MCSP is mean zero, and the user selects the desired standard deviations in range, bearing, and elevation (units are yards, radians, radians).

In conclusion, the procedure for implementation of a track and possible options requires that the XYZ data points be generated offline and read into the MCSP where various





input parameters control the rotation, translation, and noise options applied.

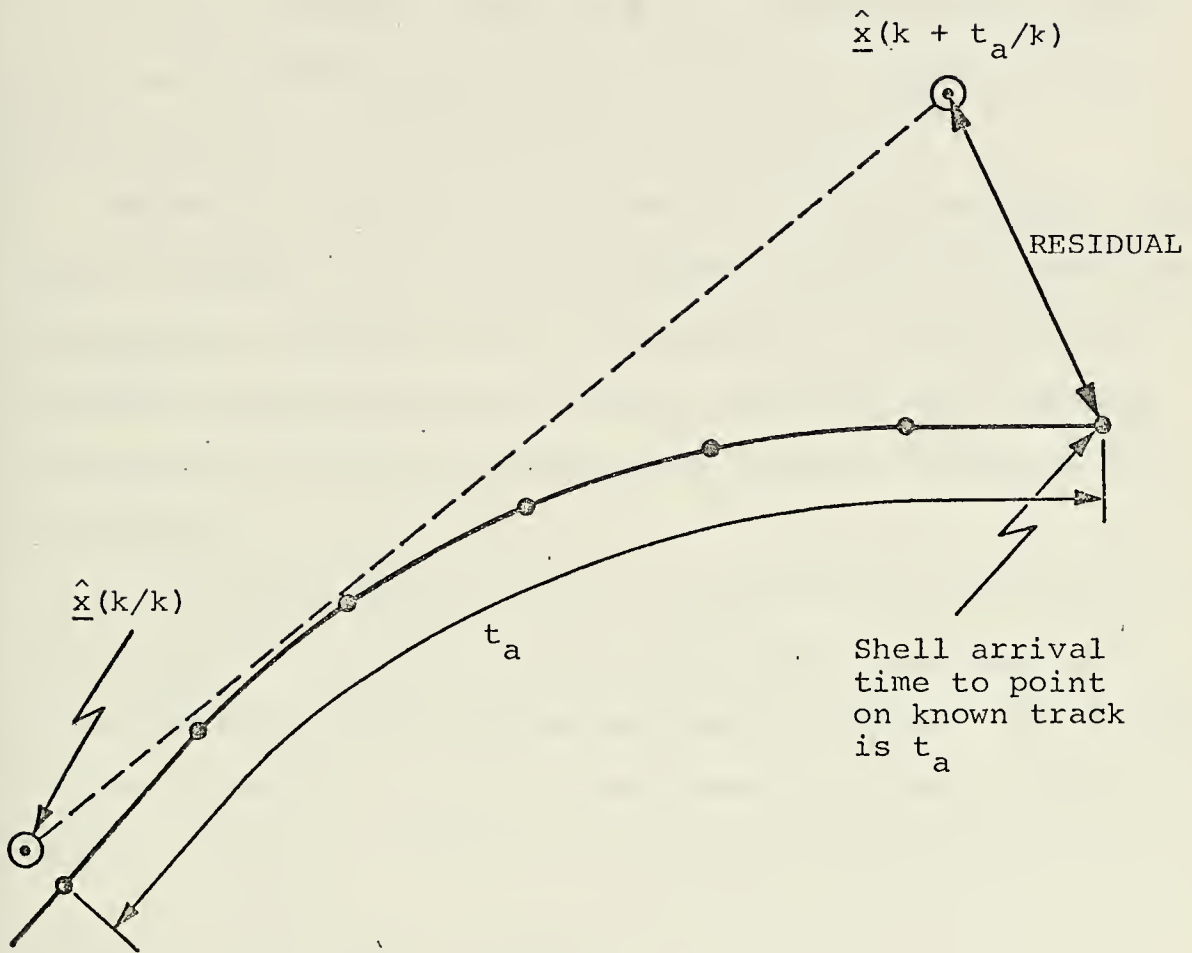
### C. SHELL-AT-TARGET ACCURACY

A feature which is an option in the MCSP provides approximate shell-at-target miss distances for additional filter evaluation. This feature is based on the firing table for a 5-inch 54-caliber gun and provides the approximate miss distances, or residuals, in X, Y, and Z directions and, also, the absolute miss distance with time of shell flight. Sections of the MCSP that are specifically responsible for these calculations are Subroutines ACCUR and STUDY.

The principles involved in obtaining the residuals will now be described. The first step required for implementation of the routine requires that the input XYZ track data be provided to program GUNFIRE, included as Appendix D, for offline calculation of the shell flight times to each point using interpolation in the 5-inch 54 firing table, included also in Appendix D. Program GUNFIRE is set up to provide these times for data points out to a range of 7500 yards and having an elevation of from 15 to 90 degrees above the horizon.

The known information includes the XYZ data points, the data point spacing, the shell times to each data point, and the  $\phi$ , or state transition matrix, and the objective, as depicted in Figure 18, is to step back in time from a track point for which the shell arrival time,  $t_a$ , is known to a filter estimated point calculated an amount  $t_a$  earlier.





● - True track points

⊙ - Filter outputs

Figure 18. Shell-at-target Accuracy Diagram.



Knowing this estimated point  $\hat{\underline{x}}(k/k)$  and the  $\underline{\phi}$  matrix the filter's prediction of target location an amount  $t_a$  later is included in the vector defined by  $\underline{\phi}^{(\frac{t_a}{T})} \hat{\underline{x}}(k/k)$  where  $T$  is the time between samples. The residual is then defined to be equal to the difference between the position states in the vector  $\hat{\underline{x}}(k+t_a/k)$  and the actual track point.

Two major problems arise, however, in this approach. The first is that if  $t_a$  is not an integer multiple of track time increments, a filter estimate of states is not available. Secondly, implementation of this method generally requires raising the  $\underline{\phi}$  matrix to a fractional power--a situation to be avoided.

A reasonable approximation to the desired results above is to step back in time to the first filter estimate prior to the desired time of filter estimation vector. This amounts to some fraction,  $\alpha$ , of the time between samples, and the required vector of estimates is given, approximately, by

$$\hat{\underline{x}}(k+\alpha/k) = \underline{\phi}^{(\frac{\alpha}{T})} \hat{\underline{x}}(k/k). \quad \text{Hence,}$$

$$\hat{\underline{x}}(k+\alpha+t_a/k+\alpha) = \underline{\phi}^{(\frac{t_a}{T})} \hat{\underline{x}}(k+\alpha/k+\alpha) \quad (52)$$

$$\approx \underline{\phi}^{(\frac{t_a}{T})} \underline{\phi}^{(\frac{\alpha}{T})} \hat{\underline{x}}(k/k) \quad (53)$$

$$\approx \underline{\phi}^{(\frac{t_a+\alpha}{T})} \hat{\underline{x}}(k/k) \quad (54)$$

but  $\frac{t_a+\alpha}{T}$  is an integer, thus solving the problem of raising  $\underline{\phi}$  to a fractional power.



EXAMPLE: Given that the time for a shell to reach data point 100 is 9.7 time periods (the same as assuming a 1-Hz sampling rate) finds the filter's predicted position.

SOLUTION: Because  $\hat{\underline{x}}(90.3/90.3)$  is not available, approximate it is  $\underline{\phi}^{0.3} \hat{\underline{x}}(90/90)$ . Then the filter's prediction vector at time 100 is given by

$$\begin{aligned}\hat{\underline{x}}(100/90) &= \underline{\phi}^{0.3} \underline{\phi}^{9.7} \hat{\underline{x}}(90/90) \\ &= \underline{\phi}^{10} \hat{\underline{x}}(90/90) .\end{aligned}\tag{55}$$

It should be noted that when the shell flight times are generated by GUNFIRE these are applicable only to the original track or a track rotated by the MCSP. If translation of a track is to be performed by the MCSP the shell-at-target miss distances generated will be erroneous, and this option should not be used.

#### D. ONLINE CALCULATION OF COVARIANCE MATRIX OF POSITION MEASUREMENT NOISE

The covariance matrix of measurement noise,  $\underline{R}$ , is a measure of the noise associated with the measurements and is used in calculating the gains.  $\underline{R}$  is defined as  $E[\underline{n}(k) \underline{n}^T(k)]$  where  $\underline{n}(k)$  is the additive noise vector at time  $k$ .

In problems where the measurements are linearly related to the filter states,  $\underline{R}$  can be approximated as a constant matrix for the entire track, an option available in the MCSP.

However, when the observations presented to the filter are range, bearing, and elevation (with noise assumed mutually independent) and the state equations are in rectangular coordinates, the required coordinate conversion generates a coupled, non-independent relationship among the noise terms





in the filtering coordinate system. This section presents a derivation to define the resultant  $\underline{R}$  matrix for the given coordinate conversion.

The radar measures range,  $r$ , bearing,  $\theta$ , and elevation,  $\phi$ , and these measurements are made available to the estimator at sample times with the parameter  $k$  as an index. The measurements also include independent noise terms  $n_1(k)$ ,  $n_2(k)$ , and  $n_3(k)$  that are assumed to be characterized by random, white distributions with mean zero and known variances. This implies that

$$E[n_i(k)]_{i=1,2,3} = 0 \quad \text{for all } k \quad (56)$$

$$E[n_1(k)n_1(j)] \quad \left\{ \begin{array}{l} 0 \quad j \neq k \\ \sigma_R^2 \quad j = k \end{array} \right. \quad (57)$$

$$E[n_2(k)n_2(j)] \quad \left\{ \begin{array}{l} 0 \quad j \neq k \\ \sigma_B^2 \quad j = k \end{array} \right. \quad (58)$$

$$E[n_3(k)n_3(j)] \quad \left\{ \begin{array}{l} 0 \quad j \neq k \\ \sigma_E^2 \quad j = k \end{array} \right. \quad (59)$$

$$\left. \begin{array}{l} E[n_1(k)n_2(j)] \\ E[n_1(k)n_3(j)] \\ E[n_2(k)n_3(j)] \end{array} \right\} \begin{array}{l} 0 \\ 0 \\ 0 \end{array} \quad \text{for all } k, j \quad (60)$$

The terms  $\sigma_R^2$ ,  $\sigma_B^2$ , and  $\sigma_E^2$  must be known. The coordinate system is the same as shown in Figure 2.

The rectangular coordinates of the target are

$$X(k) = r(k) \cos \theta(k) \cos \phi(k) \quad (61)$$

$$Y(k) = r(k) \sin \theta(k) \cos \phi(k) \quad (62)$$

$$Z(k) = r(k) \sin \phi(k) \quad (63)$$



However, the output of the radar is

$$S_1(k) = r(k) + n_1(k) \quad (64)$$

$$S_2(k) = \theta(k) + n_2(k) \quad (65)$$

$$S_3(k) = \phi(k) + n_3(k) \quad (66)$$

and what the estimator receives is given by the noisy XYZ values denoted by  $Z_1(k)$ ,  $Z_2(k)$ ,  $Z_3(k)$  and defined as follows

$$Z_1(k) = S_1(k) \cos S_2(k) \cos S_3(k) \quad (67)$$

$$Z_2(k) = S_1(k) \sin S_2(k) \cos S_3(k) \quad (68)$$

$$Z_3(k) = S_1(k) \sin S_3(k) \quad (69)$$

If equations (64) - (66) are substituted into equations (67) - (69), trigonometric identities used with subsequent expansion, and it is assumed that the bearing measurement noise,  $n_2(k)$ , and elevation measurement noise,  $n_3(k)$ , are small so that  $\cos n_2(k) \approx 1$ ,  $\sin n_2(k) \approx n_2(k)$ ,  $\cos n_3(k) \approx 1$ , and  $\sin n_3(k) \approx n_3(k)$ ,<sup>3</sup> we have

$$Z_1(k) = X(k) + v_1(k) \quad (70)$$

$$Z_2(k) = Y(k) + v_2(k) \quad (71)$$

$$Z_3(k) = Z(k) + v_3(k) \quad (72)$$

where  $v_1(k)$ ,  $v_2(k)$ , and  $v_3(k)$  represent additive noise and are given by

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<sup>3</sup>In most radar tracking systems the standard deviations in bearing and elevation are less than .005 radians, or .286 degrees. The  $3\sigma$  point, which includes 99% of this is, therefore, approximately 0.86 degrees. For comparison,  
 $\sin 0.86^\circ = .0153$  vs. .015 radians  
 $\cos 0.86^\circ = .9999$  vs. 1.



$$\begin{aligned}
v_1(k) = & - r(k) n_3(k) \cos \theta(k) \sin \phi(k) \\
& - r(k) n_2(k) \sin \theta(k) \cos \phi(k) \\
& + r(k) n_2(k) n_3(k) \sin \theta(k) \sin \phi(k) \\
& + n_1(k) \cos \theta(k) \cos \phi(k) \\
& - n_1(k) n_3(k) \cos \theta(k) \sin \phi(k) \\
& - n_1(k) n_2(k) \sin \theta(k) \cos \phi(k) \\
& + n_1(k) n_2(k) n_3(k) \sin \theta(k) \sin \phi(k) \quad (73)
\end{aligned}$$

$$\begin{aligned}
v_2(k) = & - r(k) n_3(k) \sin \theta(k) \sin \phi(k) \\
& + r(k) n_2(k) \cos \theta(k) \cos \phi(k) \\
& - r(k) n_2(k) n_3(k) \cos \theta(k) \sin \phi(k) \\
& + n_1(k) \sin \theta(k) \cos \phi(k) \\
& - n_1(k) n_3(k) \sin \theta(k) \sin \phi(k) \\
& + n_1(k) n_2(k) \cos \theta(k) \cos \phi(k) \\
& - n_1(k) n_2(k) n_3(k) \cos \theta(k) \sin \phi(k) \quad (74)
\end{aligned}$$

$$\begin{aligned}
v_3(k) = & r(k) n_3(k) \cos \phi(k) + n_1(k) \sin \phi(k) \\
& + n_1(k) n_3(k) \cos \phi(k) \quad (75)
\end{aligned}$$

Using the assumed mutual independence of the noise processes and the state variables which implies  $E[n_i n_j] = E[n_i] \cdot E[n_j]$ ,  $i \neq j$ , gives

$$E[v_1(k)] = E[v_2(k)] = E[v_3(k)] = 0, \text{ for all } k, \text{ because}$$

$$E[n_1(k)] = E[n_2(k)] = E[n_3(k)] = 0.$$



Then, if it is assumed that  $r^2(k) \gg \sigma_R^2$ , the following terms of the R matrix can be defined. In every case

$$E[v_m(k) v_n(j)] = 0 \text{ for } j \neq k \text{ and } m, n = 1, 2, 3 \quad (76)$$

$$\begin{aligned} E[v_1(k) v_1(j)] &= r^2(k) \sigma_E^2 \cos^2 \theta(k) \sin^2 \phi(k) \\ &\quad + r^2(k) \sigma_B^2 \sin^2 \theta(k) \cos^2 \phi(k) \\ &\quad + r^2(k) \sigma_B^2 \sigma_E^2 \sin^2 \theta(k) \sin^2 \phi(k) \\ &\quad + \sigma_R^2 \cos^2 \theta(k) \cos^2 \phi(k) \end{aligned} \quad (77)$$

$$\begin{aligned} E[v_1(k) v_2(j)] &= E[v_2(k) v_1(j)] = \\ &\quad r^2(k) \cos \theta(k) \sin \theta(k) \sin^2 \phi(k) \\ &\quad \cdot [\sigma_E^2 - \sigma_B^2 \sigma_E^2] \\ &\quad + \sin \theta(k) \cos \theta(k) \cos^2 \phi(k) [-r^2(k) \sigma_B^2 + \sigma_R^2] \end{aligned} \quad (78)$$

$$\begin{aligned} E[v_1(k) v_3(j)] &= E[v_3(k) v_1(j)] = \\ &\quad \cos \phi(k) \sin \phi(k) \cos \theta(k) [\sigma_R^2 - r^2(k) \sigma_E^2] \end{aligned} \quad (79)$$

$$\begin{aligned} E[v_2(k) v_2(j)] &= r^2(k) \sigma_E^2 \sin^2 \theta(k) \sin^2 \phi(k) \\ &\quad + r^2(k) \sigma_B^2 \cos^2 \theta(k) \cos^2 \phi(k) \\ &\quad + r^2(k) \sigma_B^2 \sigma_E^2 \cos^2 \theta(k) \sin^2 \phi(k) \\ &\quad + \sigma_R^2 \sin^2 \theta(k) \cos^2 \phi(k) \end{aligned} \quad (80)$$

$$\begin{aligned} E[v_2(k) v_3(j)] &= E[v_3(k) v_2(j)] = \\ &\quad \sin \theta(k) \sin \phi(k) \cos \phi(k) [\sigma_R^2 - r^2(k) \sigma_E^2] \end{aligned} \quad (81)$$





$$\begin{aligned}
E[v_3(k) v_3(j)] &= r^2(k) \sigma_E^2 \cos^2 \phi(k) \\
&+ \sigma_R^2 \sin^2 \phi(k)
\end{aligned}
\tag{82}$$

A matrix summary of  $\underline{R}$  for the three-dimensional coordinate system is

$$R = \begin{bmatrix} E[v_1(k) v_1(j)] & E[v_1(k) v_2(j)] & E[v_1(k) v_3(j)] \\ E[v_2(k) v_1(j)] & E[v_2(k) v_2(j)] & E[v_2(k) v_3(j)] \\ E[v_3(k) v_1(j)] & E[v_3(k) v_2(j)] & E[v_3(k) v_3(j)] \end{bmatrix}
\tag{83}$$

This covariance matrix is included in the MCSP by the subroutine RNOISE and functions in the following manner. The current filter estimate is used to predict ahead one time increment to define the expected position states. These predicted XYZ values are taken by routine RNOISE, converted to RBE, and, along with the standard deviations,  $\sigma_R$ ,  $\sigma_B$ , and  $\sigma_E$  input as data, which, when squared, give the necessary variances, are used to define the covariance matrix of measurement noise required for calculating the gains for the next filter estimate. That is,  $\hat{\underline{x}}(k + 1/k)$  is used to obtain  $\underline{R}$  for computing the gains needed to determine  $\hat{\underline{x}}(k + 1/k + 1)$ .

Again, this approach is an option which can be used with the Monte-Carlo simulation.

#### E. COVARIANCE MATRIX OF STATE EXCITATION

A filter designed to exactly follow a constant-velocity target will lag an accelerating target unless some technique



is incorporated to allow the filter to respond to the changes with increased gains to more heavily weight the latest observations. Increasing these gains is equivalent to increasing the bandwidth of the filter thus insuring that it is able to respond to deviations from the expected target performance.

The compromise made, however, is that additional measurement noise is able to enter into filter estimation and prediction through the increased bandwidth thus degrading the ability of the filter to reject measurement noise. There is, therefore, a balance which must be met between too small a covariance matrix of state excitation,  $\underline{Q}$ , and a resultant lagging filter and too large a  $\underline{Q}$  with subsequent erratic filter behavior.

The MCSP can accommodate the use of a  $\underline{Q}$  matrix in two ways--by using a constant input matrix of the anticipated covariance of state excitation in the calculation of all gains or by generating, online,  $\underline{Q}$  matrices through a procedure that is adaptive to the residuals detected between predicted and estimated filter states. The alternative to these methods, as the MCSP currently exists, is to use no  $\underline{Q}$  and input zeroes in its place.

The first of these methods requires some prior analysis of the target motion before the Monte-Carlo simulation is exercised, and with a modeled track, such as that designed for simulation with the MCSP, the variances can be evaluated using the track data.



The most direct method for obtaining terms to define a constant  $\underline{Q}$  matrix is to take successive differences of the position data to approximate velocities, accelerations, and acceleration rates. Having these arrays of accelerations and acceleration rates in each dimension of the coordinate system, the mean and variance of each array can be calculated with the usual statistical procedures. The resultant quantities provide the necessary variances for determining  $\underline{Q}$  for either constant-velocity or constant-acceleration models if the means produced are, in fact, close to zero (the assumed mean of random forcing). The implementation of this approach is included at the end of the track generating program in Appendix B, and the output of that program is also there showing the variances generated by the above technique.

While these variances are defined from actual track data it must be remembered that the resulting constant  $\underline{Q}$  matrix will inevitably be sub-optimal since the accelerations or acceleration rates do not, in most cases, occur uniformly over the track; for some points  $\underline{Q}$  will be "too small," thus not compensating for target motion and for other points  $\underline{Q}$  will be "too large" allowing for measurement noise degradation of filter performance.

The use of a well-chosen constant  $\underline{Q}$  matrix does, however, offer two distinct advantages. The first is that, if well chosen, it cannot help but improve the performance of a filter not dynamically adequate to follow a maneuvering target and, second, it offers simplicity and subsequent reduced



computation time and core area requirements in comparison to an adaptive technique.

The  $\underline{Q}$  matrix, when implemented as a constant matrix, is defined by

$$\underline{Q} = \underline{\Gamma} E[\underline{w} \underline{w}^T] \underline{\Gamma}^T \quad (84)$$

where  $E[\underline{w} \underline{w}^T]$  is the covariance matrix of state excitation, and  $\underline{\Gamma}$  is necessary to relate changes in velocity or acceleration for the constant-velocity or constant-acceleration filters, respectively, through successive integrations, to the appropriate states. For the XYZ coordinate system, when the random forcing is assumed to be de-coupled,

$$E[\underline{w} \underline{w}^T] = \begin{bmatrix} \sigma_X^2 & 0 & 0 \\ 0 & \sigma_Y^2 & 0 \\ 0 & 0 & \sigma_Z^2 \end{bmatrix}$$

The matrix  $\underline{\Gamma}$  for a de-coupled constant-velocity filter in three dimensions is given by

$$\underline{\Gamma}_V = \begin{bmatrix} \frac{T^2}{2} & 0 & 0 \\ T & 0 & 0 \\ 0 & \frac{T^2}{2} & 0 \\ 0 & T & 0 \\ 0 & 0 & \frac{T^2}{2} \\ 0 & 0 & T \end{bmatrix}$$





where  $T$  is the time between measurements. The constant-acceleration  $\underline{\Gamma}$  matrix for a de-coupled target motion model is

$$\underline{\Gamma}_A = \begin{bmatrix} T^3/6 & 0 & 0 \\ T^2/2 & 0 & 0 \\ T & 0 & 0 \\ 0 & T^3/6 & 0 \\ 0 & T^2/2 & 0 \\ 0 & T & 0 \\ 0 & 0 & T^3/6 \\ 0 & 0 & T^2/2 \\ 0 & 0 & T \end{bmatrix}$$

In the MCSP the matrices  $\underline{\Gamma}$  and  $E[\underline{w} \underline{w}^T]$  are input separately, and the necessary matrix products are formed to define the  $\underline{Q}$  matrix.

The second of the two methods available is an adaptive technique described in (4). The  $\underline{Q}$  matrix, at any point along the track, is defined to be

$$\underline{Q}(k) = K_1 [\hat{\underline{x}}(k/k) - \hat{\underline{x}}(k/k-1)] [\hat{\underline{x}}(k/k) - \hat{\underline{x}}(k/k-1)]^T - K_2 \cdot \underline{Q}(k-1) \quad (88)$$

where  $\hat{\underline{x}}(k/k)$  is the filter's state estimation vector at time  $k$  and  $\hat{\underline{x}}(k/k-1)$  is the state prediction vector determined at time  $k-1$ . It is this residual between the predicted states at time  $k$  using knowledge through time  $k-1$  and the estimated



states at time  $k$  which offers a measure of the state excitation acting to cause the discrepancy. The terms  $K_1$  and  $K_2$  are weighting constants used to limit the degree to which  $\underline{Q}$  can change from one point to the next. By increasing  $K_1$ ,  $\underline{Q}$  becomes much more responsive to residuals and, also, noisy estimates while an increase in  $K_2$  results in a type of damping on the  $\underline{Q}$  matrix.

This is, to say the least, a seat-of-the-pants approach, but, given a very complex modeling scenario, it offers many points in its favor. First, it is very simple and easy to implement with small additional computation time and storage required. Secondly, it does, in fact, vary the magnitude of  $\underline{Q}$  quickly, from point to point, rather than waiting for some threshold to be passed before alteration. Also, in a steady-state situation where the residual goes to zero, so does  $\underline{Q}$ , thus maintaining minimum bandwidth as desired.

It should be noted that the nature of gain calculations requires that the  $\underline{Q}$  matrix generated at point  $k$  is used to determine the gains for calculating the filter estimates at time  $k + 1$ , but given a reasonably high sampling rate the loss in performance should not be significant.

This adaptive method is easily requested, as will be shown in the section on using the MCSP, and the logic for its implementation appears at the end of subroutine UPDATE.

#### F. INITIALIZATION

The initialization of a Kalman filter can greatly affect the speed with which filter estimates and predictions improve.



Many techniques are available, and three options exist in the MCSP for user selection.

The first is a default option which is used if the other two methods are not requested. The filter initializing vector,  $\hat{\underline{x}}(0/-1)$ , has its position states defined as the first observation,  $\underline{z}$ , and the remaining states are initialized to zero. In conjunction with this choice it is advisable to initialize the prediction covariance matrix with very large values to assure large gains during the transient period.

The second possibility is to read in the initial conditions, an option that will be described in the next section. This is fine if reasonable initial conditions are known, but care must be taken not to detract from filter performance by choosing erroneous values offering worse initial conditions than the default case. Again, it pays to choose a large prediction covariance matrix if the gains are not precalculated.

The third option is to have the MCSP generate "synthetic" estimates by solving difference equations, online. With this method it takes two measurements to generate a velocity estimate and a third to generate an acceleration estimate if needed. After either two or three measurements, depending on whether a constant-velocity or constant-acceleration model is used, an initializing estimation covariance matrix is generated as follows in the MCSP:



Constant-Velocity Model

$$\underline{P}(1/1) = \begin{bmatrix} 1 & 1/T & 0 & 0 & 0 & 0 \\ 1/T & 2/T^2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1/T & 0 & 0 \\ 0 & 0 & 1/T & 2/T^2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1/T \\ 0 & 0 & 0 & 0 & 1/T & 2/T^2 \end{bmatrix} \quad (89)$$

Constant-Acceleration Model

$$\underline{P}(2/2) = \begin{bmatrix} 1 & 1/T & 1/T^2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/T & 2/T^2 & 3/T^3 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/T^2 & 3/T^3 & 6/T^4 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1/T & 1/T^2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1/T & 2/T^2 & 3/T^3 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1/T^2 & 3/T^3 & 6/T^4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1/T & 1/T^2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1/T & 2/T^2 & 3/T^3 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1/T^2 & 3/T^3 & 6/T^4 \end{bmatrix} \quad (90)$$

The appropriate estimation covariance matrix is then used to generate a prediction covariance matrix for calculating the next gain matrix. With this option it is not necessary to input an initial prediction covariance matrix. The logic for its implementation can be found in subroutine UPDATE of the MCSP included as Appendix C.





#### IV. DESCRIPTION OF MONTE-CARLO SIMULATION PROGRAM (MCSP)

##### A. APPLICATION

The Monte-Carlo Simulation Program is designed for use in comparing 6th and 9th-order filter evaluations of airborne tracking methods and incorporates user controlled selection of the options mentioned in the previous sections and several other features that are discussed in this section. A listing of the program in FORTRAN IV appears in Appendix C. Comments appear liberally throughout the listing to aid in following the flow of the program as well as for pertinent information related to variable definitions and input instructions. The following sub-sections elaborate on the procedures to follow in using the MCSP and describe the output provided.

##### B. INPUTS

The input cards are described in sequence as required by the MCSP with description of the different arrangements dictated by the options requested. Unless otherwise specified the data should be double precision, D, format and right justified in the fields designated because calculations in the MCSP are done in double precision to minimize truncation errors.

|         |                          | <u>CARD 1</u> |  |  |
|---------|--------------------------|---------------|--|--|
| COL(S)  | VARIABLE<br>(ARRAY) NAME | FORMAT        | DESCRIPTION  |  |
| 1 - 10  | LOOP                     | I 10          | Integer specifying the number of Monte-Carlo runs desired                                      |  |
| 11 - 20 | KVEL                     | I 10          | An integer "1" specifies that true velocity data for the track is available and is to be input |  |



| COL(S)  | VARIABLE<br>(ARRAY) NAME | FORMAT | DESCRIPTION  |
|---------|--------------------------|--------|--|
| 31 - 44 | LOOK                     | 14 I 1 | <p>A 14-element integer array for outputting plots on the high speed printer. A "1" in the specified column produces a plot as follows:</p> <p>31 - Mean miss distance when shell arrives (from shell-at-target calculations). Default option automatically prohibits this output if the gunfire option is not requested.</p> <p>32 - Mean filter estimation error in X position calculation</p> <p>33 - Same for Y coordinate</p> <p>34 - Same for Z coordinate</p> <p>35 - Variance of X position estimates</p> <p>36 - Same for Y</p> <p>37 - Same for Z</p> <p>38 - Mean filter estimation error in X velocity calculation</p> <p>39 - Same for Y</p> <p>40 - Same for Z</p> <p>41 - Variance of X velocity estimates</p> <p>42 - Same for Y</p> <p>43 - Same for Z</p> <p>44 - Estimated target tangential velocity in ft/sec</p> |
| 51 - 70 | TIME                     | D 20.0 | Time between measurements in secs.   |

CARD 2

| COL(S) | VARIABLE<br>(ARRAY) NAME | FORMAT | DESCRIPTION  |
|--------|--------------------------|--------|--|
| 1 - 5  | N                        | I 5    | Integer specifying the number of states, i.e. order of filter such as 6 or 9 |
| 6 - 10 | NN                       | I 5    | An integer specifying the number of track points or measurements             |



| COL(S)  | VARIABLE<br>(ARRAY) NAME | FORMAT  | DESCRIPTION  |
|---------|--------------------------|---------|--|
| 11 - 15 | OPT                      | I 5     | An integer "1" causes the program to generate synthetic measurements for filter initialization   |
| 16 - 20 | BEGIN                    | I 5     | An integer "1" signifies that the filter initialization vector $\hat{x}(0/-1)$ is to be input data   |
| 21 - 25 | IR                       | I 5     | An integer "0" causes the program to generate, online, $R$ matrices as discussed in III.D. An integer "1" signifies that a constant $R$ matrix is input data.  |
| 26 - 30 | ADAPTQ                   | I 5     | An integer "1" incorporates the adaptive $Q$ matrix calculations   |
| 31 - 35 | IG                       | I 5     | An integer "1" causes the MCSP to punch out the gain matrices for each point on the track with one row of a matrix per card and having two coded integers in the first 10 columns of each card. The first specifies the track point and the second gives the row of the gain matrix for that point on the track. The gains are punched in a D format in columns 11-30, 31-50, and 51-70 of each card. <u>NOTE: It is necessary to change the second control card of the source deck from // EXEC FORTCLG to // EXEC FORTCLGP when using this option.</u> |
| 36 - 40 | IGAIN                    | I 5     | This integer parameter controls the handling of gains in the MCSP. An integer "0" causes gains to be generated online for each MC run. An integer "1" generates gains on the first run with a noiseless track and uses these gains for all subsequent noisy tracks. An integer "2" causes gains to be read in by rows, 3 number per card. The data should be in D format in 20 column fields beginning in column 11.   |
| 41 - 80 | LABEL                    | .10 A 4 | A 10-element alphanumeric array for run identification that appears at beginning of output   |



CARD 3

| COL(S)  | VARIABLE<br>(ARRAY) NAME | FORMAT | DESCRIPTION  |
|---------|--------------------------|--------|--|
| 1 - 15  | RR                       | D 15.0 | Standard deviation of noise added to range data in yards. In MCSP this is added to .001. Range to provide a realistic figure |
| 16-30   | RB                       | D 15.0 | Standard deviation of noise added to bearing data in radians   |
| 31-45   | RE                       | D 15.0 | Standard deviation of noise added to elevation data in radians   |
| 46 - 60 | K1                       | D 15.0 | Weighting constant for adaptive <u>Q</u> option which multiplies previous <u>Q</u>   |
| 61 - 75 | K2                       | D 15.0 | Weighting constant for adaptive <u>Q</u> option which multiplies new residual matrix   |

CARD 4

| COL(S)  | VARIABLE<br>(ARRAY) NAME | FORMAT | DESCRIPTION  |
|---------|--------------------------|--------|--|
| 1 - 5   | IDIR                     | I 5    | An integer which specifies the rotation of the target track desired:<br>0 - no rotation<br>1 - 45° counterclockwise<br>2 - 90° counterclockwise<br>3 - 200.5° counterclockwise |
| 6 - 10  | INDEX                    | I 5    | An integer controlling the addition of measurement noise to the track<br>0 - no noise<br>1 - add noise   |
| 11 - 15 | IGUN                     | I 5    | An integer "1" causes MCSP to calculate and output shell-at-target residuals   |
| 16 - 35 | TRANSX                   | D 20.0 | Desired translation of all X-axis data points in yards   |
| 36 - 55 | TRANSY                   | D 20.0 | Same for Y axis  |
| 56 - 75 | TRANSZ                   | D 20.0 | Same for Z axis  |





Card 5 begins the input section for matrices, and, therefore, the description given is for a typical card of the matrix with an indication of the number of cards that should be in the matrix. This section also depends upon the options that have been chosen above, and the possible combinations are discussed below.

1 - If gains are to be read in (IGAIN = 2) they are done so here under the format specified under the IGAIN = 2 description. There should be  $NN \cdot N$  cards. (Next go to #1 below to read  $\phi$  matrix.)

The following optional matrices apply when gains are calculated by the MCSP.

2 - If the  $\underline{R}$  matrix is to be computed by the MCSP and synthetic measurements for initialization are not requested (OPT  $\neq$  1) the following sequence applies.

(a) Read initial prediction covariance matrix,  $\underline{P}(0/-1)$ , an  $N \times N$  matrix. Data is read with an 8D10.5 format, thus there are  $N$  cards for 6th-order and  $2 \cdot N$  cards for 9th-order with the 9th element of each matrix row on a card by itself in the first 10 columns following the card with the first 8 elements punched.

(b) Read initial matrix of random forcing covariances,  $E[\underline{w} \underline{w}^T]$ , a (3 X 3) matrix with the same input format as (a).

(c) Read the (3 X N)  $\underline{F}$  matrix with the same format as (a).

3 - If  $\underline{R}$  is to be a constant matrix, the instructions in 2 above apply with the addition of reading the (3 X 3)  $\underline{R}$  matrix between (a) and (b). The format is the same as 2 (a) above.

4 - If gains are calculated online and synthetic measurements are desired for filter initialization (OPT = 1) it is not necessary to provide a prediction covariance matrix. Therefore, in this case either the matrix of expected forcing covariances or a covariance matrix of measurement noise should appear first in this section of optional matrices.

These optional matrices having been arranged correctly, the next two matrices follow for all four possibilities.



- 1 - The  $N \times N$  plant dynamics matrix,  $\underline{\phi}$ , is read. Again, the format follows that described in case 2 (a) above.
- 2 - The  $3 \times N$  (for a three-dimensional problem) measurement matrix,  $\underline{H}$ , is read next with the same format as for  $\underline{\phi}$ . There should be either six or 12 cards depending on the order of the filter.

The next major group of cards pertains to the track data. The format for all track data is to read the XYZ components for a given point with a 3D20.0 specification for fields beginning in column nine. If true velocities are available and to be read they should be punched according to the same format, and the velocity data card for a given track point should follow immediately after the corresponding position data card and prior to the next point's position data card. There should be either NN or  $2 \cdot NN$  cards in this segment depending on whether velocity data is available or not.

Following the track data, if IGUN = 1 implying that the shell-at-target calculations are desired, should appear the shell flight times punched one per card in the first 20 columns. There should be NN of these cards, and the format statement controlling their input is D20.0.

The last two cards possibly needed are for the initial conditions for the state vector  $\underline{\hat{x}}(0/-1)$ , called for when BEGIN = 1. The same format is followed as for the track data with the position data on the first card and velocity data on the second. In the case of a 9th-order filter the acceleration initial conditions are automatically set to zero.

This concludes the discussion of input data organization, and a complete sample data set is included in an example in Appendix E.



### C. OUTPUT

Most of the output provided by the MCSP is self-explanatory, but a summary is included here with some additional insight into what the output actually means and a few peculiarities that the user should be aware of. A complete sample output is included with the example in Appendix E.

The first general output segment includes specification of some of the input parameters as well as the initial condition matrices. The matrices printed depend on what is relevant to the run requested. For example, if gains are read in, no prediction covariance matrix or random forcing matrix is provided. Also, if the adaptive  $\underline{Q}$  method is used the actual initial state excitation matrix is printed instead of the output for constant- $\underline{Q}$  filter runs which includes the variances of random forcing that are provided as input data.

The next output segment, if requested, is the performance table showing the shell-at-target residuals. Because it takes several seconds for a shell to reach the target after filtering has been initiated, the output shows no residuals until this flight time has been satisfied. In the case of the track designed for the MCSP with no track translation, this amounts to 36 time increments. Also printed is a figure of merit for the given filtering scheme which is the average shell-at-target miss distance for those track points where it is applicable.



Following the gunfire figure of merit is a filter performance factor representing the rms error in filter position estimates averaged over the entire track. If true velocity data is available, a similar performance factor is printed which represents the rms error in filter velocity estimates averaged over the entire track.

The succeeding output table displays actual filter estimation accuracy by showing the mean and variance of the error between the filter estimate and the true track data. If velocity data is not available, the output for the velocity statistics presents the mean and variance of the estimates themselves.

The last and optional set of output is the plots requested on the first data card. These are produced by the routine PLOTP with one plot per page and headings already provided for the three-dimensional XYZ coordinate system.





## V. SUMMARY OF REPRESENTATIVE SIMULATION RUNS

### A. DESCRIPTION OF RUNS AND NUMERICAL RESULTS

This section presents the results of the simulations using some of the various combinations of options available in the MCSP. Rather than include the extensive outputs of the simulations only relative figures of merit for the shell-at-target accuracy calculations and filter performance factors are listed in Table II to provide some insight into the effects of possible combinations of selected options. In addition, results are included that were obtained using the MK-86 filter simulation developed in [5].

Descriptions of the various runs follow that include mention of the unique options used for each run. Unless otherwise specified the controls and initial conditions used in all runs include:

- (a) 100 Monte-Carlo runs,
- (b)  $\underline{\phi}$ ,  $\underline{\Gamma}$  and  $\underline{H}$  are the same as described previously for 6th and 9th-order filters,
- (c)  $\underline{P}(k/k-1)$  is "large" (1.D05 for all diagonal terms),
- (d) Covariances of random forcing input matrix = 0,
- (e) Standard deviations of Gaussian noise added to track are  $\sigma_R = 5 + .001 \times \text{Range}$ ,  $\sigma_B = .002$  radians,  $\sigma_E = .002$  radians,
- (f) No rotation or translation of track data,
- (g) Initialization of filter by letting the first observation define the position states in  $\hat{\underline{x}}(0/-1)$ .



Reference made to "Constant Q" in the descriptions signifies that the covariances of random forcing used for defining the constant Q matrix are those calculated by successive differences in the track generating program. It should be kept in mind that this process is based on the analysis of a noiseless track, and the use of the covariances thus obtained are inherently small for the noisy-track simulations with subsequent estimator error.

| <u>Run Number</u> | <u>Order of Filter</u> | <u>Special Features</u>  |
|-------------------|------------------------|--|
| 1                 | 6                      | 1 Monte-Carlo run<br>No additive noise on track  |
| 2                 | 6                      | IGAIN = 0 (gains calculated every run)   |
| 3                 | 6                      | IGAIN = 1 (gains calculated on first run with noiseless track)                               |
| 4                 | 6                      | IGAIN = 1<br>Constant <u>Q</u> (values listed in Appendix B)                                 |
| 5                 | 6                      | Same as 4<br>OPT = 1 (synthetic initialization)<br><u>P</u> (k/k-1) = 0                      |
| 6                 | 6                      | Same as 4<br>BEGIN = 1 ( $\hat{x}(0/-1)$ is set to 0)  |
| 7                 | 6                      | IGAIN = 0<br>Adaptive <u>Q</u> with $K_1 = .1, K_2 = .9$                                     |
| 8                 | 6                      | IGAIN = 0<br>Adaptive <u>Q</u> with $K_1 = .5, K_2 = .5$                                     |
| 9                 | 6                      | IGAIN = 0<br>Adaptive <u>Q</u> with $K_1 = .9, K_2 = .1$                                     |
| 10                | 6                      | IGAIN = 0<br>Adaptive <u>Q</u> with $K_1 = .1, K_2 = .9$<br>IDIR = 1 (45° rotation of track) |



| <u>Run Number</u> | <u>Order of Filter</u> | <u>Special Features</u>   |
|-------------------|------------------------|---|
| 11                | 6                      | IGAIN = 0<br>Adaptive $\underline{Q}$ with $K_1 = .5, K_2 = .5$<br>IDIR = 1                       |
| 12                | 6                      | IGAIN = 0<br>Adaptive $\underline{Q}$ with $K_1 = .9, K_2 = .1$<br>IDIR = 1                       |
| 13                | 9                      | 1 Monte-Carlo run<br>No additive noise on track   |
| 14                | 9                      | IGAIN = 1   |
| 15                | 9                      | IGAIN = 1<br>Constant $\underline{Q}$   |
| 16                | 9                      | IGAIN = 0<br>Adaptive $\underline{Q}$ with $K_1 = .1, K_2 = .9$                                   |
| 17                | 9                      | IGAIN = 0<br>Adaptive $\underline{Q}$ with $K_1 = .5, K_2 = .5$                                   |
| 18                | 9                      | IGAIN = 0<br>Adaptive $\underline{Q}$ with $K_1 = .9, K_2 = .1$                                   |
| 19                | 9                      | Same as 18<br>Only position and velocity estimates used for shell-at-target accuracy calculations |
| 20                | 6-9                    | Model of MK-86 estimator<br>No additive noise on track<br>1 Monte-Carlo run                       |

The second column of Table II lists shell-at-target figures of merit normalized to the results obtained with the MK-86 estimator. This is done to indicate relative performance against what must be considered a severe test of the filters studied. Due to compatibility problems between the MK-86 estimator and the MCSP the results for run 20 are based on a noiseless track with one Monte-Carlo simulation. This produces optimal statistics since noise on the track degrades filter performance, and the comparison between the MK-86



estimator and the other filters in the study is slanted in favor of the former.

Additional insight into the potential success of the fire control system estimators in generating accurate firing orders is reflected in Table III where summaries of the numbers of hits within several radial miss distances of the target are tabulated for the better filters. Again, these figures would undoubtedly improve given a less maneuvering target, but it is worthwhile to know the possible consequences of a threat as reflected in the modeled track. The numbers shown are out of a total of 197 shells fired in the simulation.

#### B. EVALUATION OF RESULTS

An analysis of the numerical results presented in Table II reveals several items of interest. In this section the influence of such factors as the method of determining and using gains, initialization, type of  $\underline{Q}$  matrix used, and direction of approach are empirically compared, and possible explanations for the trends shown are discussed.

The motivation for runs 1-3 and 13-14 was to check the reliability of the Monte-Carlo statistics and to compare the various techniques for determining and using gains. With noise of mean zero added to the track in runs 2, 3, and 14 the filter estimate errors should have been consistent with those generated in runs 1 and 13 (single run, no additive noise simulations). In addition, runs 2 and 3 were executed





TABLE II  
ESTIMATOR PERFORMANCE TABLE

| Run Number | Normalized RMS Shell-at-Target Miss Distance | RMS Mean Filter Position Estimation Error - yds | RMS Mean Filter Velocity Estimation Error - yds/sec |
|------------|--|---|---|
| 1          | 5.49   | 463.83  | 79.02   |
| 2          | 5.48   | 463.67  | 78.99   |
| 3          | 5.48   | 463.80  | 78.99   |
| 4          | 0.93   | 5.19  | 11.22   |
| 5          | 0.94   | 5.29  | 11.42   |
| 6          | 0.93   | 5.22  | 11.22   |
| 7          | 2.40   | 8.48  | 58.32   |
| 8          | 2.06   | 5.88  | 47.92   |
| 9          | 0.96   | 2.06  | 16.34   |
| 10         | 2.40   | 8.91  | 61.18   |
| 11         | 2.06   | 6.12  | 50.82   |
| 12         | 0.96   | 2.12  | 17.46   |
| 13         | 3.44   | 190.22  | 50.13   |
| 14         | 3.46   | 190.32  | 50.26   |
| 15         | 1.74   | 1.79  | 6.63  |
| 16         | 1.62   | 7.13  | 32.74   |
| 17         | 1.46   | 4.35  | 26.09   |
| 18         | 1.19   | 1.59  | 10.25   |
| 19         | 0.79   | 1.59  | 10.25   |
| 20         | 1.00   | 6.30  | 10.43   |



TABLE III  
NUMBER OF HITS WITHIN SEVERAL MISS  
DISTANCES FOR SELECTED FILTERS

| Run | Description   | 15 | 20 | 25 | 30 |
|-----|---|----|----|----|----|
| 4   | 6th order<br>Constant $\underline{Q}$                               | 3  | 6  | 10 | 16 |
| 9   | 6th order<br>Adaptive $\underline{Q}$<br>$K_1 = .9, \bar{K}_2 = .1$ | 0  | 0  | 0  | 0  |
| 15  | 9th order<br>Constant $\underline{Q}$                               | 35 | 41 | 44 | 52 |
| 18  | 9th order<br>Adaptive $\underline{Q}$<br>$k_1 = .9, \bar{K}_2 = .1$ | 18 | 30 | 40 | 43 |
| 19  | Same as 18<br>Accel. terms<br>not used                              | 0  | 6  | 13 | 20 |
| 20  | MK-86 Estimator   | 0  | 3  | 6  | 12 |



to test the reliability of using gains calculated on the first run having a noiseless track against the results of a simulation having gains calculated on every run. As can be seen in Table II both tests showed desired results--the statistics for 100 runs with noisy tracks gave the expected mean errors, and the gain calculations on the first run having a noiseless track provided reliable results if used for all other runs having noisy tracks. It should be noted that calculating and using gains in this manner is only possible when  $\underline{Q}$  is a constant matrix. Attempts to use IGAIN = 1 with adaptive  $\underline{Q}$  simulations resulted in major discrepancies because the noise on the track is a major factor influencing the magnitude of the  $\underline{Q}$  matrix from point to point, and the noiseless track did not provide similar residuals.

Runs 4-6 were designed to evaluate the various initialization methods, and the results show only minor variation.

A most interesting trend in the application of the adaptive  $\underline{Q}$  option is revealed by runs 7-12 and 16-18. It is apparent that for the track used it is beneficial to provide some damping in the form of a high  $K_1/K_2$  ratio. A plausible explanation is that additive noise can cause undesirable residual variation and subsequent erratic  $\underline{Q}$  matrices if allowed to too heavily weight the updated  $\underline{Q}$  matrix. Selecting the correct ratio shows obvious improvement in filter performance and provides better performance factors in Table II than those for the constant  $\underline{Q}$  runs with 6th and



9th order filters. However, the numbers of hits within the several radii shown in Table III do not show the same pattern, and this suggests that if the requirement to have a few shells within a small radius is more important than having a lot of shells within a larger radius, then a less responsive  $\underline{Q}$  matrix calculating technique is needed. Because in most airborne filtering situations the random forcing on the target is unknown, the successful implementation of an adaptive technique seems to be the only way to deal with all possible situations.

The rotation of the track in runs 10-12 was provided to check that option and confirm that the trends shown are not functions of the direction of approach of the target.

A final area of comparison is between the 6th and 9th order filters with similar options used. As expected the 9th order filter provided smaller mean filter estimation errors because of its ability to exactly follow a constantly-accelerating target. Specific runs for comparison are 4 and 15 for the constant  $\underline{Q}$  runs and 9 and 18 for the best of the adaptive  $\underline{Q}$  runs. As shown in Table II the two methods for determining  $\underline{Q}$  provide similar results for the 6th-order filter with the constant  $\underline{Q}$  method producing slightly smaller filter estimation errors. In the case of the 9th-order filter the adaptive  $\underline{Q}$  technique provides superior position estimates, but the constant  $\underline{Q}$  method is superior in generating velocity estimates.





A most noteworthy incompatibility is the contradiction shown for shell-at-target miss distances between 6th and 9th-order estimators. From Table II it is suggested that the 6th-order filters offer superior performance in minimizing the miss distances, but the actual superiority belongs to the 9th-order filters as evidenced by Table III. The discrepancy can be explained by the fact that the 6th-order filters consistently provide miss distances that are in the range from 50 to 150 yards but fail to reduce this substantially whereas the 9th-order filters do provide these minimum miss distances but are substantially worse at ranges greater than 3000 yards due to the effects of acceleration estimation errors. At great distances these acceleration estimation errors are important due to the larger flight times of the shells.

Run 19 was included as a special study to determine the shell-at-target accuracy for the filter in run 18 when only position and velocity estimates were used for prediction. Some internal changes were required in the MCSP to obtain these results that are intended to reinforce the comments made above.

The results for run 20 present the performance of the existing system against the scenario of the MCSP track. This is a combination 6th-9th order filter using only position and velocity estimates to generate predicted target location for determining gunfire orders.



## VI. CONCLUSIONS AND RECOMMENDATIONS

Judging from the results presented in the previous section the techniques and derivations presented in this thesis have been shown to be applicable to airborne tracking problems with possible results superior to those obtained using present estimators. The use of a Monte-Carlo simulator has been beneficial in insuring that the results obtained are statistically legitimate for the scenario studied. Also, the capacity to vary the implementation of the many options available has made it possible to determine those factors that offer the greatest potential for improved estimator performance.

While the results presented in this thesis are encouraging, the scope of the study must be reiterated in order to avoid drawing too broad conclusions regarding the applicability to aircraft tracking problems. The track developed and used was selected because of its common use today in typical aircraft tactics, but much additional analysis would be necessary to categorically accept the relative capacities of the several filters presented herein. It is the adherence to such an analytical approach that can insure that the filters acquired for modern gunfire control systems are capable of handling the threats they are expected to experience.



The results have shown that given a complex target approach, basically simple filters are capable of surprising results if reinforced by techniques designed specifically for the model scenario. The online generation of the covariance matrix of measurement noise and adaptive covariance matrices of random forcing are examples that contribute to minimizing filter estimation errors. These features are, of course, dependent upon the availability of computation capacity, but with the advent of smaller, faster, and less expensive computers the restrictions imposed are becoming more liberal all the time.

This thesis was intended to correlate some of the important concepts pertinent to airborne tracking filters into an easily used simulation program for current and future study. Some areas that appear to offer additional potential for filter improvement and simulation accuracy include continued study in the field of adaptive covariance matrices of random forcing, a feature seen to be most powerful in this thesis, further work in the area of computation time and storage requirements for implementing various filtering approaches, and refinement of the shell-at-target accuracy calculations. Furthermore, implementation of simulation-generated techniques into actual models for testing is mandatory to provide feedback for refining the simulators, thus saving investment in inferior systems due to acceptance of incorrect simulation results.



## APPENDIX A

### PROGRAM LISTING FOR MONTE-CARLO RUN REQUIREMENT EVALUATION

The program written for analyzing the existing generators of normally distributed random numbers is included in this appendix. It provides statistics for empirical mean and standard deviation within 5, 10, 15, 20, and 25 percent of the desired standard deviation for ensembles taken a variable number of times.

In the program the following variables can be altered to obtain additional statistics describing expected results in Monte-Carlo simulations

ITER - The number of samples in each ensemble

N - The number of ensembles determined by data cards using an I5 format in the first five columns.

The program is written to terminate when  $N = 5000$ , but this can be revised by referring to the statement identified by NORM0094.





```

C THIS PROGRAM ANALYZES THE AVAILABLE RANDCM NUMBER GENERATORS FOR
C NORMALLY DISTRIBUTED RANDOM NUMBERS
C
C REAL MEAN, NUM1, NUM2, NUM3, NUM4, NUM5, NUM6, NUM7, NUM8, NUM9, NUM10
C DIMENSION MEAN(10000), STDDEV(10000)
C
C ITER IS THE NUMBER OF TIMES THE REQUESTED NUMBER OF SAMPLES ARE
C TAKEN TO ASSURE STEADY-STATE CONDITIONS
C
C 10 ITER = 1000
C
C ITER IS NUMBER OF ITERATIONS OF 'N' SAMPLES EACH
C
C 25 READ(5,25) N
C    FORMAT(I5)
C
C XMEAN IS THE REQUESTED MEAN
C XSTD IS THE REQUESTED STANDARD DEVIATION
C
C XMEAN = 0.0
C XSTD = 2.0
C WRITE(6,100) N, ITER, XMEAN, XSTD
C 100 FORMAT(I11,10X,'EXPERIMENTAL NORMAL DISTRIBUTION',/,
C 1 11X,'PARAMETERS FOR ',I3,' SAMPLES TAKEN',I5,' TIMES',/,11X,
C 2 ' WITH GIVEN MEAN=',F4.0,' STD. DEV.=',F4.0, '//')
C
C THE FOLLOWING CARD IS NOT PRESENT WHEN 'GAUSS' IS USED
C
C CALL OVFLOW
C KERNEL = 2568317
C DO 1000 I=1, ITER
C   MEAN(I) = 0
C   STDDEV(I) = 0
C   DO 2000 J=1,N
C     WHEN GAUSS IS USED THE CALLING STATEMENT IS 'CALL GAUSS (KERNEL,
C     XSTD,XMEAN,TEMP)', AND THE FOLLOWING TWO CARDS ARE ELIMINATED
C
C     CALL NORMAL(KERNEL,TEMP,1)
C     TEMP = TEMP*XSTD
C     TERM1 = FLOAT(J-1)/FLOAT(J)
C     TERM2 = 1.0/FLOAT(J)
C     MEAN(I) = MEAN(I)*TERM1+TERM2*TEMP
C     STDDEV(I) = STDDEV(I)+(TEMP-MEAN(I))**2
C     STDDEV(I) = Sqrt(STDDEV(I)/(N-1))
C   2000 NUM1=0
C   1000 STDDEV(I) = 0
C   50 NUM2=0

```



NORM0048  
 NORM0049  
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 NORM0051  
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 NORM0054  
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 NORM0092  
 NORM0093  
 NORM0094  
 NORM0095

```

NUM3=0
NUM4=0
NUM5=0
NUM6=0
NUM7=0
NUM8=0
NUM9=0
NUM10=0
DC 300
I=1
ITER 5, 10, 15, 20, 25 PERCENT OF STD DEV RANGE OF MEAN
FIND RISK (MEAN(I)-XMEAN) NUM1=NUM1+1
DIFF=ABS(.05*XSTD)) NUM2 = NUM2+1
IF(DIFF.LE.(.15*XSTD)) NUM3=NUM3+1
IF(DIFF.LE.(.25*XSTD)) NUM4 = NUM4+1
IF(DIFF.LE.(.25*XSTD)) NUM5=NUM5+1
FIND PRECISIONS FOR STANDARD DEVIATION ERRORS OF
S, 10, 15, 20, 25 PERCENT
DIFF=ABS(STDDEV(I)-XSTD) NUM6=NUM6+1
IF(DIFF.LE.(.05*XSTD)) NUM7=NUM7+1
IF(DIFF.LE.(.15*XSTD)) NUM8=NUM8+1
IF(DIFF.LE.(.20*XSTD)) NUM9=NUM9+1
IF(DIFF.LE.(.25*XSTD)) NUM10=NUM10+1
CONTINUE
ITER=1
R15=NUM1/FILTER*100
R10=NUM2/FILTER*100
R15=NUM3/FILTER*100
R25=NUM4/FILTER*100
S15=NUM5/FILTER*100
S10=NUM6/FILTER*100
S20=NUM7/FILTER*100
S25=NUM8/FILTER*100
WRITE(6,400) R5, S5, R10, S10, R15, S15, R20, S20, R25, S25
FORMAT(IH0,5X,PERCENT,10X,PERCENT OF MEANS,11X,11X,12X,
1, WITHIN PERCENT SD,/,/,6X,CRITERIA OF,14X,CRITERIA OF,/,/,
27X,DESIRED SD,16X,/,6X,60(1-),/,/,
10X,15,17X,F7.3,/,
5X,10,17X,F7.3,/,
9X,15,17X,F7.3,/,
9X,20,17X,F7.3,/,
9X,25,17X,F7.3,/,
IF(N.NE.5000) GO TO 10
STOP
  
```



APPENDIX B  
TRACK GENERATING PROGRAM

The following program generates the track data used in the simulations conducted for comparison in this thesis. Both position and velocity data are produced for each of the 233 points on the track, and, in addition, the means and variances of accelerations and acceleration rates are calculated and averaged over the entire track to provide data used in the "Constant  $\underline{Q}$ " simulations.

The results of the latter calculations follow, and it should be remembered that the validity of using the variances in defining a constant  $\underline{Q}$  matrix is dependent on having enough sample points to insure accelerations and acceleration rates with the assumed mean of random forcing (zero).

|   |         |                                     |
|---|---------|-------------------------------------|
| Mean acceleration in X direction:             | -1.675  | yd/sec <sup>2</sup>                 |
| "                  Y      "      :            | -1.419  | "                                   |
| "                  Z      "      :            | 0.772   | "                                   |
| Variance of acceleration in X direction:      | 45.739  | (yd/sec <sup>2</sup> ) <sup>2</sup> |
| "                  Y      "      :            | 296.869 | "                                   |
| "                  Z      "      :            | 224.467 | "                                   |
| Mean acceleration rate in X direction:        | -0.131  | yd/sec <sup>3</sup>                 |
| "                  Y      "      :            | -0.351  | "                                   |
| "                  Z      "      :            | 0.207   | "                                   |
| Variance of acceleration rate in X direction: | 106.984 | (yd/sec <sup>3</sup> ) <sup>2</sup> |
| Variance of acceleration rate in Y direction: | 957.846 | "                                   |
| Variance of acceleration rate in Z direction: | 4249.27 | "                                   |



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 TRAC0003  
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 TRAC0035  
 TRAC0036  
 TRAC0037  
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 TRAC0039  
 TRAC0040  
 TRAC0041  
 TRAC0042  
 TRAC0043  
 TRAC0044  
 TRAC0045  
 TRAC0046  
 TRAC0047

THIS PROGRAM GENERATES THE TRACK AND VELOCITY DATA AS WELL AS  
 GENERATING THE VARIANCES USED FOR FINDING A CONSTANT,  
 OPTIMAL, Q MATRIX

```

IMPLICIT REAL*8 (A-H,L-Z)
DIMENSION X(233),Y(233),Z(233)
1 XVEL(233),YVEL(233),ZVEL(233)
2 VSTOR(233,3),ASTOR(233,3),ARSTOR(233,3)
DEGRAD = 57.2957795131D0
X(1) = 17250.D0
Y(1) = 0.D0
Z(1) = 1300.D0
XVEL(1) = -417.5D0
YVEL(1) = 0.D0
ZVEL(1) = 0.D0
THETA = 80.D0/DEGRAD
R = 6680.D0/THETA
ANGLE = 0.D0
DC 10 I=2,33
ANGLE = ANGLE+1.25D0/DEGRAD
ZTEMP = R*DCOS(ANGLE)
XVEL(I) = -417.5D0*DCOS(ANGLE)
YVEL(I) = 0.D0
ZVEL(I) = -417.5D0*DSIN(ANGLE)
X(I) = X(1)-XTEMP
Y(I) = 0.D0
Z(I) = Z(1)-(R-ZTEMP)

```

THIS IS END OF INITIAL DIVE PHASE - BEGIN ACCELERATING DIVE

```

SIN40 = DSIN(40.D0/DEGRAD)
COS40 = DCOS(40.D0/DEGRAD)
TIME = .25D0
DIST = 7500.D0/SIN40
VEL = (2.D0*DIST)/20.D0-417.5D0
CLDVEL = (VEL-417.5D0)/80.D0
DC 20 I=54,113
X(I) = X(I-1)-(OLDVEL*TIME+DVEL*TIME/2.D0)*COS40
Y(I) = Y(I-1)-(OLDVEL*TIME+DVEL*TIME/2.D0)*SIN40
Z(I) = Z(I-1)-(OLDVEL+DVEL
XVEL(I) = -CLDVEL*DCOS(40.D0/DEGRAD)
YVEL(I) = 0.D0
ZVEL(I) = -CLDVEL*DSIN(40.D0/DEGRAD)

```





```

C C THIS IS END OF ACCELERATING DIVE - BEGIN PULLOUT
R = 12.00*VEL/TTHETA
ANGLE = 0.00
ADEL = 40.00/24.00/DEGRAD
DC 30 I=114,137
ANGLE = ANGLE+ADEL
CPRIME = 2.00*R*DSIN(ANGLE/2.00)
PHI = 40.00/DEGRAD - ANGLE/2.00
XVEL(I) = -VEL*DCOS(40.00/DEGRAD-ANGLE)
YVEL(I) = 0.00
ZVEL(I) = -VEL*DSIN(40.00/DEGRAD-ANGLE)
X(I) = X(113)-CPRIME*DCOS(PHI)
Z(I) = Z(113)-CPRIME*DSIN(PHI)
Y(I) = Y(I-1)
30
C C THIS IS END OF PULLOUT - BEGIN EVASIVE MANEUVER TO RIGHT OF 5 DEG
TTHETA = 5.00/DEGRAD
R = VEL /TTHETA
ANGLE = 0.00
DC 40 I=138,141
ANGLE = ANGLE+1.2500/DEGRAD
XVEL(I) = -VEL*DCOS(ANGLE)
YVEL(I) = VEL*DSIN(ANGLE)
ZVEL(I) = 0.00
X(I) = X(137)-R*DSIN(ANGLE)
Y(I) = Y(137)+R*DCOS(ANGLE)
Z(I) = Z(137)
40
C C THIS IS END OF STARBOARD MANEUVER - REPEAT TO PORT
ANGLE = 0.00
DC 50 I=142,145
ANGLE = ANGLE+1.2500/DEGRAD
CPRIME = 2.00*R*DSIN(ANGLE/2.00)
PHI = 85.00/DEGRAD+ANGLE/2.00
XVEL(I) = -VEL*DCOS(5.00/DEGRAD-ANGLE)
YVEL(I) = VEL*DSIN(5.00/DEGRAD-ANGLE)
ZVEL(I) = 0.00
X(I) = X(141)-CPRIME*DSIN(PHI)
Y(I) = Y(141)+CPRIME*DCOS(PHI)
Z(I) = Z(141)
50
C C THIS IS END OF PORT MANEUVER - EXECUTE ANOTHER SIMILAR PORT MAN.
ANGLE = 0.00
DC 60 I=146,149

```

TRAC0048  
TRAC0049  
TRAC0050  
TRAC0051  
TRAC0052  
TRAC0053  
TRAC0054  
TRAC0055  
TRAC0056  
TRAC0057  
TRAC0058  
TRAC0059  
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TRAC0090  
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TRAC0093  
TRAC0094  
TRAC0095



```

ANGLE = ANGLE+1.2500/DEGRAD
XVEL(I) = -VEL*DCOS(ANGLE)
YVEL(I) = -VEL*DSIN(ANGLE)
ZVEL(I) = 0.00
X(I) = X(145)-R*DSIN(ANGLE)
Y(I) = Y(145)-{R-R*DCOS(ANGLE)}
Z(I) = Z(145)

```

60

CCC

THIS IS END OF PORT MANEUVER - PLANE NOW HAS PASSED OVER SHIP -  
EXECUTE 3.5 G PORT TURN FOR 2 SECS AND BEGIN 10 DEG CLIMB

```

SIN10 = DSIN(10.00/DEGRAD)
COS10 = DCOS(10.00/DEGRAD)
ANGLE = 0.00
GRAV = 32.200
TAU = 5.00/DEGRAD
R = VEL**2/(3.500*GRAV)
THETA = S/R
DELTH = THETA/8.00
DC70 I = 150.157 DELTH
ANGLE = DELTH
C PHI = DO*AR*DSIN(ANGLE/2.00)
XVEL(I) = -VEL*DSIN(ANGLE/2.00)
YVEL(I) = -VEL*DSIN(ANGLE/2.00)
ZVEL(I) = -VEL*DSIN(ANGLE/2.00)
X(I) = X(149)-C*DCOS(PHI)
Y(I) = Y(149)-C*DSIN(PHI)
Z(I) = Z(149)+C*SIN10
TAU = TAU+THETA

```

70

CCC

END OF TURN - BEGIN 3.5 G STARBOARD TURN FOR 2 SECS.

```

ANGLE = 0.00
DO 80 I = 158, 165
  DELTH = ANGLE+DELTH
  C PHI = 2.00*AR*DSIN(ANGLE/2.00)
  XVEL(I) = -VEL*DCOS(ANGLE/2.00)
  YVEL(I) = -VEL*DSIN(ANGLE/2.00)
  ZVEL(I) = -VEL*DSIN(ANGLE/2.00)
  X(I) = X(157)-C*DCOS(PHI)
  Y(I) = Y(157)-C*DSIN(PHI)
  Z(I) = Z(157)+C*SIN10
  TAU = TAU-THETA

```

80

CCC

END OF TURN - BEGIN 3.0 G PORT TURN FOR 2 SECS.

TRAC00096  
TRAC00097  
TRAC00098  
TRAC00099  
TRAC0100  
TRAC0101  
TRAC0102  
TRAC0103  
TRAC0104  
TRAC0105  
TRAC0106  
TRAC0107  
TRAC0108  
TRAC0109  
TRAC0110  
TRAC0111  
TRAC0112  
TRAC0113  
TRAC0114  
TRAC0115  
TRAC0116  
TRAC0117  
TRAC0118  
TRAC0119  
TRAC0120  
TRAC0121  
TRAC0122  
TRAC0123  
TRAC0124  
TRAC0125  
TRAC0126  
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 TRAC0151  
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 TRAC0183  
 TRAC0184  
 TRAC0185  
 TRAC0186  
 TRAC0187  
 TRAC0188  
 TRAC0189  
 TRAC0190  
 TRAC0191

```

C
  ANGLE = 0.00
  R = VEL**2/(3.00*GRAV)
  THETA = S/R
  DELTH I = 166,173
  DO 90 I = 166,173
  ANGLE = ANGLE+DELTH
  C = 2.00*R*DSIN(ANGLE/2.00)
  PHI = TAU+ANGLE/2.00*DCOS(TAU+ANGLE)
  XVEL(I) = -VEL*DSIN(TAU+ANGLE)
  YVEL(I) = -VEL*DSIN(TAU+ANGLE)
  ZVEL(I) = VEL*SIN10
  X(I) = X(165)-C*DCOS(PHI)
  Y(I) = Y(165)-C*DSIN(PHI)
  Z(I) = Z(165)+C*SIN10
  TAU = TAU+THETA
  90
  C
  C
  END OF TURN -- BEGIN 3.0 G STARBOARD TURN FOR 2 SECS.

```

```

  ANGLE = 0.00
  DO 100 I = 174,181
  DELTH I = 174,181
  ANGLE = ANGLE+DELTH
  C = 2.00*R*DSIN(ANGLE/2.00)
  PHI = 90.00/DEGRAD-TAU+ANGLE/2.00
  XVEL(I) = -VEL*DSIN(TAU-ANGLE)
  YVEL(I) = -VEL*DSIN(TAU-ANGLE)
  ZVEL(I) = VEL*SIN10
  X(I) = X(173)-C*DCOS(PHI)
  Y(I) = Y(173)-C*DSIN(PHI)
  Z(I) = Z(173)+C*SIN10
  TAU = TAU-THETA
  100
  C
  C
  END OF TURN -- BEGIN 3.0 G PORT TURN FOR 3 SECS.

```

```

  ANGLE = 0.00
  S = VEL**3.00
  THETA = S/R
  DELTH I = 182,193
  DO 110 I = 182,193
  ANGLE = ANGLE+DELTH
  C = 2.00*R*DSIN(ANGLE/2.00)
  PHI = TAU+ANGLE/2.00*DCOS(TAU+ANGLE)
  XVEL(I) = -VEL*DSIN(TAU+ANGLE)
  YVEL(I) = -VEL*DSIN(TAU+ANGLE)
  ZVEL(I) = VEL*SIN10
  X(I) = X(181)-C*DCOS(PHI)
  Y(I) = Y(181)-C*DSIN(PHI)
  Z(I) = Z(181)+C*SIN10
  110
  C
  C

```



TRAC00192  
 TRAC00193  
 TRAC00194  
 TRAC00195  
 TRAC00196  
 TRAC00197  
 TRAC00198  
 TRAC00199  
 TRAC00200  
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 TRAC00202  
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 TRAC00210  
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 TRAC00219  
 TRAC00220  
 TRAC00221  
 TRAC00222  
 TRAC00223  
 TRAC00224  
 TRAC00225  
 TRAC00226  
 TRAC00227  
 TRAC00228  
 TRAC00229  
 TRAC00230  
 TRAC00231  
 TRAC00232  
 TRAC00233  
 TRAC00234  
 TRAC00235  
 TRAC00236  
 TRAC00237  
 TRAC00238  
 TRAC00239

TAU = TAU+THETA  
 END OF TURN - BEGIN 2.5 G STARBOARD TURN FOR 3 SECS.

ANGLE = 0.00  
 S = VEL\*\*2/(2.500\*GRAV)  
 R = TETA  
 DELTH = TETA/12.00  
 DO I=194,205  
 ANGLE = ANGLE+DELTH  
 C = 2.00\*R\*DSIN(ANGLE/2.00)  
 PHI = 90.00/DEGRAD - TAU+ANGLE  
 XVEL(I) = -VEL\*COS(10\*DCOS(TAU-ANGLE))  
 YVEL(I) = -VEL\*SIN(10\*DCOS(TAU-ANGLE))  
 ZVEL(I) = VEL\*SIN(10\*DCOS(TAU-ANGLE))  
 X(I) = X(193)-C\*DCOS(PHI)  
 Y(I) = Y(193)+C\*SIN(10\*DCOS(TAU-ANGLE))  
 Z(I) = Z(193)+C\*SIN(10\*DCOS(TAU-ANGLE))  
 TAU = TAU-THETA

120  
 END OF TURN - BEGIN 2.5 G PORT TURN FOR 4 SECS.

ANGLE = 0.00  
 S = VEL\*4.00  
 R = TETA  
 DELTH = TETA/16.00  
 DO I=206,221  
 ANGLE = ANGLE+DELTH  
 C = 2.00\*R\*DSIN(ANGLE/2.00)  
 PHI = TAU+ANGLE  
 XVEL(I) = -VEL\*COS(10\*DCOS(TAU+ANGLE))  
 YVEL(I) = -VEL\*SIN(10\*DCOS(TAU+ANGLE))  
 ZVEL(I) = VEL\*SIN(10\*DCOS(TAU+ANGLE))  
 X(I) = X(205)-C\*DCOS(PHI)  
 Y(I) = Y(205)+C\*SIN(10\*DCOS(TAU+ANGLE))  
 Z(I) = Z(205)+C\*SIN(10\*DCOS(TAU+ANGLE))  
 TAU = TAU+THETA

130  
 END OF TURN - BEGIN 2.0 G STARBOARD TURN FOR 3 SECS.

ANGLE = 0.00  
 S = VEL\*3.00  
 R = TETA  
 DELTH = TETA/12.00  
 DO I=222,233  
 ANGLE = ANGLE+DELTH  
 C = 2.00\*R\*DSIN(ANGLE/2.00)





TRAC0240  
 TRAC0241  
 TRAC0242  
 TRAC0243  
 TRAC0244  
 TRAC0245  
 TRAC0246  
 TRAC0247  
 TRAC0248  
 TRAC0249  
 TRAC0250  
 TRAC0251  
 TRAC0252  
 TRAC0253  
 TRAC0254  
 TRAC0255  
 TRAC0256  
 TRAC0257  
 TRAC0258  
 TRAC0259  
 TRAC0260  
 TRAC0261  
 TRAC0262  
 TRAC0263  
 TRAC0264  
 TRAC0265  
 TRAC0266  
 TRAC0267  
 TRAC0268  
 TRAC0269  
 TRAC0270  
 TRAC0271  
 TRAC0272  
 TRAC0273  
 TRAC0274  
 TRAC0275  
 TRAC0276  
 TRAC0277  
 TRAC0278  
 TRAC0279  
 TRAC0280  
 TRAC0281  
 TRAC0282  
 TRAC0283  
 TRAC0284  
 TRAC0285  
 TRAC0286  
 TRAC0287

```

PHI = 90. DO/DEGRAD-TAU+ANGLE/2. DO
XVEL(I) = -VEL#COS10*DCOS(TAU-ANGLE)
YVEL(I) = -VEL#COS10*DSIN(TAU-ANGLE)
ZVEL(I) = VEL#SIN10
X(I) = X(221)-C#DCOS(PHI)
Y(I) = Y(221)-C#DCOS(PHI)
Z(I) = Z(221)+C#SIN10
DC 150 I=1,233

```

140

C  
 C  
 C  
 C

CHANGE POSITION AND VELOCITY DATA TO YARDS AND YARDS/SEC  
 INSTEAD OF FEET AND FEET/SEC

```

X(I) = X(I)/3. DO
Y(I) = Y(I)/3. DO
Z(I) = Z(I)/3. DO
VEL = DSQRT(XVEL(I)**2+YVEL(I)**2+ZVEL(I)**2)
XVEL(I) = XVEL(I)/3. DO
YVEL(I) = YVEL(I)/3. DO
ZVEL(I) = ZVEL(I)/3. DO

```

THE POSITION/VELOCITY VALUES ARE PUNCHED ON CARD PAIRS WITH  
 THE 3 POSITION VALUES ON THE FIRST CARD AND THE 3 VELOCITY  
 VALUES ON THE SECOND

C  
 C  
 C  
 C  
 C

```

WRITE(7,4778) I, X(I), Y(I), Z(I), I, XVEL(I), YVEL(I), ZVEL(I), VEL
WRITE(6,4777) I, X(I), Y(I), Z(I), I, XVEL(I), YVEL(I), ZVEL(I), VEL
FCRMAT(5X, I3, 3D20.10, /, 5X, I3, 3D20.10)
FCRMAT(5X, I3, 3D20.10, /, 5X, I3, 3D20.10, 10X, D20.6)

```

150  
 4778  
 4777

C  
 C  
 C  
 C  
 C

THE REMAINDER OF THE PROGRAM GENERATES VARIANCES OF ACCELERATION  
 AND ACCELERATION RATES BY TAKING SUCCESSIVE DIFFERENCES AND  
 ANALYZING OVER THE ENTIRE TRACK

```

DO 160 I=1,232
VSTOR(I,1) = (X(I+1)-X(I))/ .25D0
VSTOR(I,2) = (Y(I+1)-Y(I))/ .25D0
VSTOR(I,3) = (Z(I+1)-Z(I))/ .25D0
DC 170 I=1,231
ASTOR(I,1) = (VSTOR(I+1,1)-VSTOR(I,1))/ .25D0
ASTOR(I,2) = (VSTOR(I+1,2)-VSTOR(I,2))/ .25D0
ASTOR(I,3) = (VSTOR(I+1,3)-VSTOR(I,3))/ .25D0
DO 180 I=1,230
ARSTOR(I,1) = (ASTOR(I+1,1)-ASTOR(I,1))/ .25D0
ARSTOR(I,2) = (ASTOR(I+1,2)-ASTOR(I,2))/ .25D0
ARSTOR(I,3) = (ASTOR(I+1,3)-ASTOR(I,3))/ .25D0
AMEANY = 0. DO
AMEANZ = 0. DO

```

160  
 170  
 180



```

AVARX = 0.00
AVARY = 0.00
AVARZ = 0.00
DC 190 I=1,231
AMEANX+ASTOR(I,1)
AMEANY+ASTOR(I,2)
AMEANZ+ASTOR(I,3)
AVARX+ASTOR(I,1)**2
AVARY+ASTOR(I,2)**2
AVARZ+ASTOR(I,3)**2
AMEANX = AMEANX/231.00
AMEANY = AMEANX/231.00
AMEANZ = AMEANX/231.00
AVARX = AVARX/231.00-AMEANX**2
AVARY = AVARX/231.00-AMEANX**2
AVARZ = AVARZ/231.00-AMEANX**2
WRITE(6,2719) AMEANX,AMEANZ,AVARX,AVARY,AVARZ
FORMAT(IH1,9X,'ACC MEAN X=',D20.6,/, 'ACC MEAN Y=',D20.6,/,
1 10X,'ACC MEAN Z=',D20.6,/, 'VAR X=',D20.6,/, 'VAR Y=',D20.6,/,
3 ;/, '10X', 'VAR Z=',D20.6,/,/)
AMEANX = 0.00
AMEANY = 0.00
AMEANZ = 0.00
AVARX = 0.00
AVARY = 0.00
AVARZ = 0.00
DC 200 I=1,230
AMEANX+ARSTOR(I,1)
AMEANY+ARSTOR(I,2)
AMEANZ+ARSTOR(I,3)
AVARX+ARSTOR(I,1)**2
AVARY+ARSTOR(I,2)**2
AVARZ+ARSTOR(I,3)**2
AMEANX = AMEANX/230.00
AMEANY = AMEANX/230.00
AMEANZ = AMEANX/230.00
AVARX = AVARX/230.00-AMEANX**2
AVARY = AVARX/230.00-AMEANX**2
AVARZ = AVARX/230.00-AMEANX**2
WRITE(6,2720) AMEANX,AMEANZ,AVARX,AVARY,AVARZ
FORMAT(IH9,9X,'ACC RATE MEAN X=',D20.6,/, 'ACC RATE MEAN Y=',
1 D20.6,/, '10X', 'ACC RATE MEAN Z=',D20.6,/, 'VAR X=',D20.6,/,
2 'VAR Y=',D20.6,/, '10X', 'VAR Z=',D20.6,/)
3 SIGP
END
TRAC0288
TRAC0289
TRAC0290
TRAC0291
TRAC0292
TRAC0293
TRAC0294
TRAC0295
TRAC0296
TRAC0297
TRAC0298
TRAC0299
TRAC0300
TRAC0301
TRAC0302
TRAC0303
TRAC0304
TRAC0305
TRAC0306
TRAC0307
TRAC0308
TRAC0309
TRAC0310
TRAC0311
TRAC0312
TRAC0313
TRAC0314
TRAC0315
TRAC0316
TRAC0317
TRAC0318
TRAC0319
TRAC0320
TRAC0321
TRAC0322
TRAC0323
TRAC0324
TRAC0325
TRAC0326
TRAC0327
TRAC0328
TRAC0329
TRAC0330
TRAC0331
TRAC0332

```



## APPENDIX C

### LISTING OF MONTE-CARLO SIMULATION PROGRAM (MSCP)

A complete listing of the Monte-Carlo Simulation Program is included herein. For user convenience the statement numbers are in ascending order allowing for easy location of designated statements, and the cards are labeled in the right margin for reference and ordering if required.

Variable and array descriptions appear at the beginning of the program with instructions for the input parameters immediately following each READ statement. Format statements controlling headings on the output plots are included in the program statements MCSP0600-MSCP0694.



```

THIS IS THE MONTE-CARLO SIMULATICN PROGRM (MCSP)
AND CAN ANALYZE UP TO A 233 PCINT TRACK AS MANY AS
99999 TIMES TO OBTAIN TRUE MCNTE-CARLO STATISTICS
OF FILTER PERFORMANCE
1 IMPLICIT REAL*8(Q,H,R,G,P,A,B,C,D,S,I)
  REAL*8 II,X,Y,Z,XTRUE,YTRUE,ZTRUE,EST,RTRUE,BTRUE,ETRUE,RT,BT,ET,XM
  REAL*4 SEEABS,SEEDRD
  INTEGER*4 OPT,OPTNO,BEGIN,ADAPTQ
  DIMENSION PRED(9,9), LABEL(10), STATES(9,9), RADAR(9,9), XTRUE(233
1), YTRUE(233), ZTRUE(233), QP(9,9), PKKMIP(9,9), RP(9,9), A(9,9)
2B(9,9), GSTOR(9,3,233), TVELX(233), TVELY(233), TVELZ(233), RTRUE(
3233), BTRUE(233), ETRUE(233), SEEABS(233), LOCK(14),
+ STARTI(9), GAMMA(9,9), GAMMAT(9,9), HCLD(9,9),
CCMCON II(9,9), Q(9,9), R(9,9), G(9,9), PHI(9,9), PKK(9,9), PKKMI
1(9,9), RMEAN(233,6), RSD(6), RESID(233,4), AHEAD(233), EST(9,233), PHIPR
2M(9,9,40), PKSTOR(9,233)
  DESCRIPTION OF ARRAYS FOLLOWS:
  PRED -- FILTERS DESCRIPTION OF X(K+1/K)
  LABELS -- USER DESCRIPTION OF RUN
  RADAR -- FILTER ESTIMATES
  XTRUE -- MEASUREMENT INPUT 'X' VALUES
  YTRUE -- NOISELESS INPUT 'Y' VALUES
  ZTRUE -- NOISELESS INPUT 'Z' VALUES
  QP -- COVARIANCE MATRIX OF STATE EXCITATION
  PKKMIP -- PRIMARY MATRIX MEASURED IN CALCULATING POWER
  RP -- TEMPORARY MATRIX USED IN CALCULATING POWER
  A -- TEMPORARY MATRIX USED IN CALCULATING POWER
  B -- TEMPORARY MATRIX USED IN CALCULATING POWER
  GSTOR -- STORAGE
  TVELX -- VELOCITY
  TVELY -- VELOCITY
  TVELZ -- VELOCITY
  RTRUE -- RANGE CALCULATED FROM TRUE X,Y,Z
  BTRUE -- BEARING CALCULATED FROM TRUE X,Y,Z
  ETRUE -- ELEVATION CALCULATED FROM TRUE X,Y,Z
CALL QVFLOW
KERNL1=3790437
KERNL2=30117221
KERNL3=1993183
DEGRAD=57.295779513100
II -- IDENTITY MATRIX

```

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CC





MCSPO0048  
 MCSPO0049  
 MCSPO0050  
 MCSPO0051  
 MCSPO0052  
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 MCSPO0059  
 MCSPO0060  
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 MCSPO0092  
 MCSPO0093  
 MCSPO0094  
 MCSPO0095

```

    Q  - COVARIANCE MATRIX OF STATE EXCITATION (VARIABLE)
    GAMMA - GAMMA TO RELATE NON-DETERMINISTIC
    FORCING TO STATES
    H R - MEASUREMENT MATRIX MEASUREMENT ERROR MATRIX (VARIABLE)
    G - OPTIMUM GAIN, I.E. PLANT DYNAMICS P (K/K) (VARIABLE)
    PHI - ESTIMATION COVARIANCES, LALS IN EACH DIRECTION TOTAL MISS
    PKKMI - MEAN POSITION RESIDUALS IN EACH DIRECTION PLUS TRACK
    RMEAN - SUM OF POSITION RESIDUALS TO GET ESTIMATES
    RSD - POSITION SHELL FILTER ESTIMATES
    RESID - TIMES ARRAY FOR POWERS OF PHIMATRIX UP TO 40
    AHEAD - STORAGE FOR FILTER VARIANCES
    ESTPHIPRM - STORAGE FOR FILTER VARIANCES
    PKSTOR - STORAGE FOR FILTER VARIANCES
    DC 4 I=1,233
    DC 1 K=1,9
    PKSTOR(K,I)=0.00
    EST(2,J)=0.00
    RMEAN(I,J)=0.00
    DC 3 L=1,4
    RESID(I,L)=0.00
    DC 5 I=1,6
    RSD(I)=0.00
    READ (5,6) LOOP,KVEL,(LOOK(I),I=1,14),TIME
    FCRMAT (2I10,10X,14I1,6X,D20.0)
    LOOP IS THE NUMBER OF MONTE CARLO RUNS
    KVEL = 1 WILL SET UP THE PROGRAM FOR READING IN TRUE VELOCITIES
    IN ADDITION TO TRACK POSITIONS TO BE PRINTED
    ARRAY ,LOOK, DETERMINES PLOTS TO BE PRINTED
    IF INPUT ARRAY WILL BE PRINTED
    LOOK(1): MEAN DISTANCE WHEN SHELL ARRIVES
    LOOK(2): MEAN ERROR IN X POSITION
    LOOK(3): MEAN ERROR IN Y POSITION
    LOOK(4): MEAN ERROR IN Z POSITION
    LOOK(5): X POSITION VARIANCE
    LOOK(6): Y POSITION VARIANCE
    LOOK(7): Z POSITION VARIANCE
    LOOK(8): MEAN VELOCITY ERROR IN X DIRECTION
    LOOK(9): MEAN VELOCITY ERROR IN Y DIRECTION
    LOOK(10): MEAN VELOCITY ERROR IN Z DIRECTION
  
```

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 C CCCCCCCCCCCCCCCCCC



MCSP00096  
 MCSP00097  
 MCSP00098  
 MCSP00099  
 MCSP0100  
 MCSP0101  
 MCSP0102  
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 MCSP0143

```

LOOK(11): X VELOCITY VARIANCE
LOOK(12): Y VELOCITY VARIANCE
LOOK(13): Z VELOCITY VARIANCE
LOOK(14): TARGET SPEED IN FT/SEC
'TIME, IS THE TIME BETWEEN MEASUREMENTS

7 READ (5,7) N,NN,OPT,BEGIN,IR,ADAPTQ,IG,IGAIN,(LABEL(I),I=1,10)
FCRMAT (8I5,10A4)

```

```

N----- NUMBER OF STATES OR CRDR OF FILTER
NN----- NUMBER OF STEPS - I.E. MEASUREMENTS
CPT----- A , I, HERE WILL GENERATE SYNTHETIC MEASUREMENTS
BEGIN----- A , I, MEANS THAT THE STATE INITIAL CONDITION
MATRIX X(0/-1) IS TO BE READ IN AT THE END OF THE DATA
IR----- A , 0, WILL CAUSE THE PROGRAM TO GENERATE ONLINE
R MATRICES. A , 1, WILL READ IN THE CONSTANT R MATRIX
ADAPTQ---- A , 1, WILL CAUSE THE Q MATRIX TO BE CHANGED
ONLINE DEPENDENT ON THE RESIDUAL BETWEEN PREDICTED AND
ESTIMATED STATES.
IG----- A , 1, MEANS PUNCH OUT GAINS - NOTE: MUST ALSO CHANGE
IGAIN---- A , 0, EXEC FORTCLG ... TC // EXEC FORTCLGP ...
EACH MONTE CARLO RUN WILL CAUSE GAINS TO BE GENERATED ONLINE FOR
NOISELESS TRACK AND USE FOR ALL OTHER RUNS
SUBROUTINE 'MREADI'

```

```

8 READ (5,8) RR,RB,RE,K1,K2
FCRMAT (5D15.0)

RR = RANGE STANDARD DEVIATION
RB = BEARING STANDARD DEVIATION
PE = ELEVATION STANDARD DEVIATION
K1 AND K2 ARE WEIGHTING CONSTANTS WHICH DETERMINE THE EXTENT TO
WHICH THE EFFECT OF THE LATEST RESIDUAL WILL ENTER INTO THE
Q MATRIX. I.E. K1 X OLD Q + K2 X NEW Q

9 READ (5,9) IDIR,INDEX,IGUN,TRANSX,TRANSY,TRANZ
FCRMAT (3I5,3D20.0)

```

```

INDEX = 0 : NO NOISE ADDED ----- INDEX = 1 : NOISE ADDED
IDIR DETERMINES DIRECTION OF APPROACH
0 DEGS. IS RIGHT BEAM AND ANGLES INCREASE COUNTERCLOCKWISE
IDIR = 0 ANGLE = 0 DEGS.
IDIR = 1 ANGLE = 45 DEGS.
IDIR = 2 ANGLE = 90 DEGS.

```

C C C C C  
 C C C C C C C C C C C C C C C  
 C C C C C C C C C C C C C C C  
 C C C C C C C C C C C C C C C



```

CC C C C C C
IDIR = 3 ANGLE = 202.5 DEGS
IGUN=1 WILL CALCULATE SHELL AT TARGET RESIDUALS
IPANSX = TRANSLATION OF ALL X DATA POINTS
IPANSY = TRANSLATION OF ALL Y DATA POINTS
IPANSZ = TRANSLATION OF ALL Z DATA POINTS
M=N/3
IF (OPT.EQ.1.AND.N.EQ.9) OPTNO=3
IF (OPT.EQ.1.AND.N.EQ.6) OPTNO=2
NUM=N/M
N2=N/3+1
N3=N*2/3+1
N4=N2+1
N5=N3+1
DC 10 I=1,9
DC 10 J=1,9
DC 10 I,J)=0.DO
RACAR(I,J)=0.DO
STATES(I,J)=0.DO
PRED(I,J)=0.DO
CP(I,J)=0.DO
PHI(I,J)=0.DO
PKK(I,J)=0.DO
PKKMIP(I,J)=0.DO
G(I,J)=0.DO
H(I,J)=0.DO
HCLD(I,J) = 0.DO
GAMMA(I,J) = 0.DO
GAMMAT(I,J) = 0.DO
IF (IGAIN.EQ.0.OR.IGAIN.EQ.1) GO TO 12
DC 11 I=1,NN
IF GAINS ARE TO BE READ, THEY ARE DONE SO HERE - THERE WILL BE
NN QUANTITY OF (N X NUM) MATRICES
CALL MREAD1 (G,N,NUM)
DC 11 J=1,N
DC 11 K=1,NUM
11 GSTOR(J,K,I)=G(J,K)
GC TO 13
INITIAL CONDITIONS ON COVARIANCE OF ESTIMATION ERROR READ HERE
THIS WILL BE A (N X N) MATRIX
12 IF(OPT.NE.1) CALL MREAD(PKKMIP,N,N)
IF THE 'R' MATRIX IS NOT TIME VARYING, READ IT IN HERE
CC C C C C C

```

```

MCSP0144
MCSP0145
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MCSP0189
MCSP0190
MCSP0191

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CC      THIS WILL BE A (NUM X NUM) MATRIX
CC      IF(IR.NE.0) CALL MREAD(RP,NUM,NUM)
CC      READ INITIAL CONDITION VARIANCES
CC      OF RANDOM FORCING MATRIX
CC      THIS WILL BE A (NUM X NUM) MATRIX
CC      CALL MREAD (QP,NUM,NUM)
CC      READ GAMMA MATRIX - THIS WILL BE A (N X NUM) MATRIX
CC      CALL MREAD(GAMMA,N,NUM)
CC      READ PHI MATRIX HERE - THIS WILL BE A (N X N) MATRIX
CC      CALL MREAD (PHI,N,N)
CC      READ MEASUREMENT MATRIX HERE - THIS WILL BE A (NUM X N) MATRIX
CC      CALL MREAD (H,NUM,N)
CC      IF (IDIR.EQ.0) DEL=0.00
CC      IF (IDIR.EQ.1) DEL=45.00/DEGRAD
CC      IF (IDIR.EQ.2) DEL=90.00/DEGRAD
CC      IF (IDIR.EQ.3) DEL=202.500/DEGRAD
CC      SINDEL=DSIN(DEL)
CC      CCSDEL=DCOS(DEL)
CC      DC 15 I=1,NN
CC
CC      THE FOLLOWING SECTION READS THE TRUE (X,Y,Z) DATA AND
CC      CALCULATES TRUE (R,B,E)
CC
14      READ (5,14) XTRUE(I),YTRUE(I),ZTRUE(I)
      FORMAT (6X,3D20.0)
      XTRUE(I)=XTRUE(I)+TRANSX
      YTRUE(I)=YTRUE(I)+TRANSY
      ZTRUE(I)=ZTRUE(I)+TRANSZ
      XT=XTRUE(I)
      YT=YTRUE(I)
      ZT=ZTRUE(I)
      RT=CSQRT(XT**2+YT**2+ZT**2)
      BT=DATAN2(ZT,DSQRT(XT**2+YT**2))
      IF (YT.GE.0.00.AND.XT.NE.0.00) BT=DATAN2(YT,XT)+DEL
      IF (YT.GT.0.00.AND.XT.EQ.0.00) BT = PI/2.00 + DEL
      IF (YT.EQ.0.00.AND.XT.LT.0.00) BT = 180.00/DEGRAD + DEL
      IF (YT.LT.0.00.AND.XT.LT.0.00) BT=DATAN2(DABS(YT),DABS(XT))+DEL+180.00/DEGRAD

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MCSP0192
MCSP0193
MCSP0194
MCSP0195
MCSP0196
MCSP0197
MCSP0198
MCSP0199
MCSP0200
MCSP0201
MCSP0202
MCSP0203
MCSP0204
MCSP0205
MCSP0206
MCSP0207
MCSP0208
MCSP0209
MCSP0210
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MCSP0226
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MCSP0230
MCSP0231
MCSP0232
MCSP0233
MCSP0234
MCSP0235
MCSP0236
MCSP0237
MCSP0238

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IF (YT.LT.0.DO.AND.XT.EQ.0.DO) BT = 3.DO*PI/2.DO + DEL
IF (YT.LT.0.DO.AND.XT.GT.0.DO) BT = -DATAN2(DABS(YT),XT)+DEL+360.DO/
1 DEGRAD
RTTRUE(I)=RT
BTRUE(I)=BT
ETTRUE(I)=ET
XTRUE(I)=RT*DCOS(BT)*DCOS(ET)
YTRUE(I)=RT*DSIN(BT)*DCOS(ET)
ZTRUE(I)=RT*DSIN(ET)
WRITE(6,4444) I,XTRUE(I),YTRUE(I),ZTRUE(I),RT,BT,ET
1 C
TVELX(I)=0.DO
TVELY(I)=0.DO
TVELZ(I)=0.DO
2 C
3 C
TRUE VELOCITIES ARE READ IN IF AVAILASLE
4 C
IF (KVEL.EQ.1) READ (5,14) TVELX(I),TVELY(I),TVELZ(I)
STOREV = TVELX(I)
TVELX(I) = STOREV *COSDEL-TVELY(I)*SINDEL
TVELY(I) = TVELY(I)*COSDEL+ STOREV *SINDEL
AHEAD(I)=0.DO
15 CCONTINUE
15 C
TIMES FOR SHELL TO REACH TRACK ARE READ IN IF REQUESTED
16 C
IF (IGUN.NE.1) GO TO 18
DC 16 I=1,NN
17 READ (5,17) AHEAD(I)
17 FORMAT (D20.0)
16 C
DETERMINE THE FILTER INITIALIZATION TERMS
18 C
IF (BEGIN.NE.1) GO TO 20
18 READ (5,14) START(1),START(N2),START(N3),START(N4),START
18 IN5)
DC 19 I=1,9
19 PRED(I,1)=START(I)
GO TO 21
20 CALL NJISE (RTRUE(1),BTRUE(1),ETRUE(1),RR,RE,RE,KERNL1,KERNL2,KERN
19 L3,X,Y,Z,INDEX)
KERNL1=3790437
KERNL2=30117221
KERNL3=1993183
START(1)=X
START(N2)=Y
START(N3)=Z
PRED(1,1)=X
PRED(N2,1)=Y

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MCSP0239
MCSP0240
MCSP0241
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MCSP0248
MCSP0249
MCSP0250
MCSP0251
MCSP0252
MCSP0253
MCSP0254
MCSP0255
SUPP0255
MCSP0256
MCSP0257
MCSP0258
MCSP0259
MCSP0260
MCSP0261
MCSP0262
MCSP0263
MCSP0264
MCSP0265
MCSP0266
MCSP0267
MCSP0268
MCSP0269
MCSP0270
MCSP0271
MCSP0272
MCSP0273
MCSP0274
MCSP0275
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MCSP0277
MCSP0278
MCSP0279
MCSP0280
MCSP0281
MCSP0282
MCSP0283
MCSP0284
MCSP0285

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32 FCPMAT (IH0,/,/, ' COVARIANCES OF RANDOM FORCING FOR CALCULATION OF
   GC MATRI X',/,/)
33 GC TO 35
34 WRITE (6,34) K1,K2
35 FCPMAT (IH0,/,/, ' INITIAL COVARIANCES OF RANDOM FORCING MATRIX OF
   STATE EXCITATION FOR ADAPTIVE Q FILTER',/,/, ' WEIGHTING FACTOR
   FOR Q MATRIX UPDATE ARE - K1=,D20.6,/,45X,K2=,D20.6,/,/)
36 CALL MWRITE (QP,NUM,NUM)
   DEFINE Q MATRIX AS GAMMA X (COVARIANCES OF RANDOM FORCING,
   INPUT AS QP) X GAMMA TRANSPOSE
37 CALL TRANS (GAMMA,N,NUM,GAMMAT)
38 CALL PROD (GAMMA,QP,N,NUM,NUM,HOLD)
39 CALL PROD (HOLD,GAMMAT,N,NUM,N,QP)
   GP IS NOW THE INITIAL CONDITION COVARIANCE MATRIX
   OF STATE EXCITATION
39 IF (IR.EQ.0) GO TO 37
40 WRITE (6,36)
36 FCPMAT (IH0,/,/, ' CONSTANT R OR COVARIANCE OF MEASUREMENT ERROR MAT
   RIX',/,/)
   CALL MWRITE (RP,NUM,NUM)
40 GC TO 41
37 WRITE (6,38)
38 FCPMAT (IH0,/,/, ' R OR COVARIANCE OF MEAS. ERROR MATRIX GENERATED B
   Y PROGRAM',/,/)
41 GC TO 41
39 WRITE (6,40)
40 FCPMAT (IH0,/,/, ' GAINS ARE READ IN',/,/)
41 WRITE (6,42) RR,RE STAN DEV,RR=,1PD20.12,/,10X,
42 FCPMAT (IH0,9X,DEV,RB=,1PD20.12,/,10X,
   BEARING STAN DEV,RE=,1PD20.12)
43 IF (OPT.EQ.1) WRITE (6,43) OPTNC
44 FCPMAT (IH0,/,/, ' THE FIRST ,12, ESTIMATES OF POSITION WILL BE TH
   OBSERVED POSITION',/,/)
45 GC TO 44
46 FCPMAT (IH0,/,/, ' STATE INITIAL CONDITIONS ARE',/,/)
47 CALL MWRITE (PRED,N,1)
48 IF (IR.NE.0) GO TO 46
49 CALL RNOISE (PRED,RB,RR,RE,N)
50 DC 45 I=1,NUM
51 DC 45 J=1,NUM
52 PP(I,J)=R(I,J)
53 CCNTINUE

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MCSP03334
MCSP03335
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MCSP03376
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MCSP03379
MCSP03380
MCSP03381

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C C THE FOLLOWING LOOP IS THE MONTE CARLO SIMULATION
C C
C C DC 58 ITER=1,LOOP
C C DC 47 I=1,N
C C DC 47 J=1,N
C C G(I,J)=GP(I,J)
C C PCKM1(I,J)=PKKM1P(I,J)
C C R(I,J)=RP(I,J)
47 DC 48 I=1,N
48 PRED(I,1)=START(I)
C C THE FOLLOWING LOOP STEPS THROUGH THE TRACK POINTS
C C
C C DC 57 L=1,NN
C C IF DESIRED, NOISE IS ADDED TO THE TRACK POINTS
C C CALL NOISE (RTRUE(L),BTRUE(L),ETRUE(L),RR,RE,RE,KERNL1,
C C I KERNL2,KERNL3,X,Y,Z,INDEX)
C C IFLAG = 0
C C THE FOLLOWING CARD SETS THE OPTICN FOR USING INITIAL OBSERVATIONS
C C TO GENERATE PREDICTIONS INSTEAD OF FILTER ESTIMATES
C C IF (OPT.EQ.1.AND.L.LE.OPTND) IFLAG=1
C C THE FOLLOWING SECTION HANDLES THE VARIOUS GAIN OPTICNS
C C
C C IF (IGAIN.EQ.0) GO TO 52
C C IF (IGAIN.EQ.1.AND.ITER.NE.1) GO TO 50
C C IF (IGAIN.EQ.2) GO TO 50
C C CALL GAIN (N,NUM,ITER,L,IFLAG,KVEL)
C C DC 49 IZ=1,N
C C DC 49 IX=1,NUM
49 GSTOR(IZ,IX,L)=G(IZ,IX)
C C GO TO 53
C C DC 51 IZ=1,N
C C DC 51 IX=1,NUM
51 G(IZ,IX)=GSTOR(IZ,IX,L)
C C GO TO 53
52 CALL GAIN (N,NUM,ITER,L,IFLAG,KVEL)
C C FOR MOST SITUATIONS THE RADAR MEASUREMENTS ARE MADE NOISY
C C
C C 53 RADAR(1,1)=X
C C RADAR(2,1)=Y
C C RADAR(3,1)=Z
C C
MCSP0382
MCSP0383
MCSP0384
MCSP0385
MCSP0386
MCSP0387
MCSP0388
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MCSP0390
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MCSP0397
MCSP0398
MCSP0399
MCSP0400
MCSP0401
MCSP0402
MCSP0403
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MCSP0429

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C          HCWEVER, FOR I GAIN MONTE CARLO SIMULATION TRUE TRACK PCINTS ARE
C          USED - THIS IS ONLY DONE ON THE FIRST MONTE CARLO RUN
C          IF (IGAIN.NE.1.OR.ITER.NE.1) GO TO 54
C          RADAR(1,1)=XTRUE(L)
C          RADAR(2,1)=YTRUE(L)
C          RADAR(3,1)=ZTRUE(L)
54          CCNTINUE
C          'UPDATE' WILL GENERATE ESTIMATED AND PREDICTED FILTER STATES
C          CALL UPDATE (N,NUM,RADAR,STATES,PRED,IFLAG,L,TIME,IGAIN,ITER,K1,K2
1,ADAPTQ)
DC 55 I=1,9
C          THE FOLLOWING MATRIX SAVES THE ESTIMATED STATES FOR USE IN
C          CALCULATING MISS DISTANCES WHEN THE SHELL REACHES THE TARGET
55          EST(I,L)=STATES(I,1)
C          THE FOLLOWING 'IF' CALLS FCR COVARIANCE CF MEASUREMENT NOISE ONLY
C          WHEN NECESSARY
C          IF((IGAIN.EQ.0.OR.(IGAIN.EQ.1.AND.ITER.EQ.1)).AND.IR.EQ.0)
1          CALL RNOISE(PRED,RB,RR,RE,N)
C          'ACCUR' IS CALLED TO CALCULATE SHELL AT TARGET MISS DISTANCE
C          WHEN SUFFICIENT TIME HAS PASSED FOR SHELL TO GET TO TARGET
C          IF (IGUN.NE.1) GO TO 56
C          TSPACE=1.DO/TIME
C          IF (FLOAT(L).GT.(TSPACE*AHEAD(L)+1.DO)) CALL ACCUR (XTRUE(L),YTRUE
1(L),ZTRUE(L),L,N,ITER)
C          'SCENAR' IS CALLED TO PROVIDE MONTE CARLO STATISTICS
56          CALL SCENAR (XTRUE(L),YTRUE(L),ZTRUE(L),STATES,N,L,ITER,IGAIN,TVEL
1X(L),TVELY(L),TVELZ(L),KVEL)
57          CCNTINUE
58          CCNTINUE
C          OUTPUT PERFORMANCE TABLE IF SHELL AT TARGET RESIDUALS CALCULATED
C          IF (IGUN.NE.1) GO TO 60
C          WRITE (6,59)
C          FCRRMAT (1H1,'POINT',7X,'TRUE',10X,'FILTER',8X,'RESIDUALS',8X,
59          'TIME',10X,'MEAN',/, 'NUMBER',6X,'TRACK',10X,'PRED',25X,
2          'OF FLI',9X,'MISS',/,13X,'POINTS',8X,'POINTS',24X,'-----',

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 MCSP0525

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3  7X,'DISTANCE',/,55X,'TRACK TIME',/,55X,'WHEN FIRED',/,IX,
4  88('7'),//)
    THE FOLLOWING SECTION DOES FINAL COMPUTATIONS ON STATISTICAL
    QUANTITIES AND OUTPUTS THEM
60 DC 62 I=1,NN
    RMEAN'S ARE CHANGED TO MEAN OF ESTIMATE AT EACH POINT FROM SUM
    OF ESTIMATES FOR ALL MONTE CARLO RUNS
61 RMEAN(I,J)=RMEAN(I,J)/FLCAT(LLOOP)
    THE FOLLOWING 6 TERMS ARE USED TO CALCULATE FILTER PERFORMANCE
    FACTORS, AND REPRESENT SUMS OF MEAN ERRORS
    RSD(1)=RSD(1)+DABS(RMEAN(I,1)-XTRUE(I))
    RSD(2)=RSD(2)+DABS(RMEAN(I,2)-YTRUE(I))
    RSD(3)=RSD(3)+DABS(RMEAN(I,3)-ZTRUE(I))
    RSD(4) = RSD(4) + DABS(RMEAN(I,4)-TWELX(I))
    RSD(5) = RSD(5) + DABS(RMEAN(I,5)-TWELY(I))
    RSD(6) = RSD(6) + DABS(RMEAN(I,6)-TWELZ(I))
62 CONTINUE
    IF (IGUN.NE.1) GO TO 66
    IFIRE = 0
    I=1,NN
    RESID(I,1)=RESID(I,1)/FLOAT(LLOOP)
    RESID(I,2)=RESID(I,2)/FLOAT(LLOOP)
    RESID(I,3)=RESID(I,3)/FLCAT(LLOOP)
    RESID(I,4)=DSQRT(RESID(I,1)**2+RESID(I,2)**2+RESID(I,3)**2)
    IF INSUFFICIENT TIME HAD PASSED TO CALCULATE MISS DISTANCE
    IF THEN DO NOT OUTPUT STATISTICS
    IF (RESID(I,4).EQ.0.00) GO TO 63
    IFIRE = IFIRE + 1
    CALL STUDY (I,XTRUE(I),YTRUE(I),ZTRUE(I),RESID(I,1),RESID(I,2),RES
1  ID(I,3),AHEAD(I),RESID(I,4))
63 CONTINUE
    GUNSUM = 0.00
    DC 64 I=1,NN
    GUNSUM = GUNSUM + RESID(I,4)
    GMISS = GUNSUM/FLCAT(IFIRE)
    WRITE (6,65) GMISS
65 FORMAT('H0,///,IGX,'AVERAGE SHELL AT TARGET MISS DISTANCE =',
1  D20.8,' YARDS.')
66 PERFX=RSD(1)/FLOAT(NN)
  
```



```

PERFY=RSD(2)/FLOAT(NN)
PERFZ=RSD(3)/FLOAT(NN)
PERFXV = RSD(4)/FLOAT(NN)
PERFYV = RSD(5)/FLOAT(NN)
PERFZV = RSD(6)/FLOAT(NN)
CC
CC
      CALCULATE FILTER PERFORMANCE FACTOR
PERF=DSQRT(PERFX**2+PERFY**2+PERFZ**2)
WRITE (6,67) PERF
67  FORMAT (1H0, '//', 10X, 'PERFORMANCE FACTOR CF FILTER, AVERAGE FILTER
      POSITION ESTIMATE ERROR = ', D20.8, ' YARDS',)
      1  PERFV = DSQRT(PERFXV**2+PERFYV**2+PERFZV**2)
      IF(KVEL.EQ.1) WRITE(6,120) PERFV
      WRITE (6,68)
68  FORMAT (1H1, 'POINT NO', 6X, 'POSITION', 12X, 'VELOCITY',
      1, 11X, 'VELOCITY VARIANCES',
      2, /, 15X, 'ESTIMATE', 12X, '/ERRORS', 11X,
      3, 'AROUND ESTIMATE', /, 16X, 'AROUND TRUTH', /,
      4  88(' ',), //)
      DO 73 I=1,NN
      DO 69 J=1,9
69  PKSTOR(J,I)=PKSTOR(J,I)/FLOAT(LCOP)
      RMEAN'S NOW BECOME THE EXPECTED VALUE OF ESTIMATICN ERROR
RMEAN(I,1)=RMEAN(I,1)-XTRUE(I)
RMEAN(I,2)=RMEAN(I,2)-YTRUE(I)
RMEAN(I,3)=RMEAN(I,3)-ZTRUE(I)
      PKSTOR'S BECOME VARIANCES DEFINED AS EXPECTED VALUE OF SQUARE
      OF ESTIMATION ERROR MINUS EXPECTED VALUE OF ESTIMATION ERROR**2
      E(X**2) - E(X)**2
CC
CC
      PKSTOR(1,I)=PKSTOR(1,I)-RMEAN(I,1)**2
      PKSTOR(N2,I)=PKSTOR(N2,I)-RMEAN(I,2)**2
      PKSTOR(N3,I)=PKSTOR(N3,I)-RMEAN(I,3)**2
      IF (KVEL.NE.1) GO TO 71
      RMEAN(I,4)=RMEAN(I,4)-TVELX(I)
      RMEAN(I,5)=RMEAN(I,5)-TVELZ(I)
      RMEAN(I,6)=RMEAN(I,6)-TVELZ(I)
      PKSTOR(2,I)=PKSTOR(2,I)-RMEAN(I,4)**2
      PKSTOR(N4,I)=PKSTOR(N4,I)-RMEAN(I,5)**2
      PKSTOR(N5,I)=PKSTOR(N5,I)-RMEAN(I,6)**2
      WRITE (6,70) I, RMEAN(I,1), PKSTOR(1,I), RMEAN(I,4), PKSTOR(2,I), RMEAN(NCSP0571
      1(I,2), PKSTOR(N2,I), RMEAN(NCSP0572
      2, RMEAN(I,6), PKSTOR(N5,I),
      70  FORMAT (1H0, I5, 4X, 'X:', D15.6, 4X, 'X:', D15.6, 9X, 'X:', D15.6, /,

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MCSP0573

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1 10X,'Y:',D15.6,4X,'Y:',D15.6,D15.6,9X,'Y:',D15.6,/,
2 10X,'Z:',D15.6,4X,'Z:',D15.6,D15.6,9X,'Z:',D15.6,/,
GC TO 73

```

IF TRUE VELOCITIES ARE NOT AVAILABLE, THEN CALCULATE MEAN AND VARIANCE OF ESTIMATED VELOCITY, NOT ERROR OF ESTIMATED VELOCITY

```

71 AVEL=DSQRT(RMEAN(I,4)**2+RMEAN(I,5)**2+RMEAN(I,6)**2)
PKSTOR(2,I)=PKSTOR(2,I)-RMEAN(I,4)**2
PKSTOR(N4,I)=PKSTOR(N4,I)-RMEAN(I,5)**2
PKSTOR(N5,I)=PKSTOR(N5,I)-RMEAN(I,6)**2
WRITE(6,72) I,RMEAN(I,1),PKSTOR(1,I),PKSTOR(2,I),RMEAN(I,2),PKSTOR
1 R(N2,I),AVEL,15,4X,'X:',D15.6,4X,'X:',D15.6,24X,'X:',D15.6,/,
72 1 10X,'Y:',D15.6,4X,'Y:',D15.6,D15.6,9X,'Y:',D15.6,/,
2 10X,'Z:',D15.6,4X,'Z:',D15.6,D15.6,9X,'Z:',D15.6,/,
73 CONTINUE

```

FOLLOWING IS THE SEGMENT FOR OUTPUTING GRAPHS

```

DC 74 I=1,NN
74 SEEABS(I)=I
IF (LOOK(I),EQ.0.OR.IGUN.NE.1) GO TO 77
WRITE(6,75) I,10X,'KETRON MK-86 THESIS',/,11X,
FOR MAT (I,75) I,10X,'KETRON MK-86 THESIS',/,11X,
1 'MEAN MISS DISTANCE WHEN SHELL ARRIVES IN YDS.',/,,
DC 76 I=1,NN
76 SEEORD(I)=RESID(I,4) SEEORD,NN,0)
CALL PLOTTP (SEEABS,SEEORD,NN,0)
IF (LOOK(2),EQ.0) GO TO 80
WRITE(6,78) I,10X,'KETRON MK-86 THESIS',/,11X,
1 'MEAN ERROR IN X POSITION',/,,
DC 79 I=1,NN
79 SEEORD(I)=RMEAN(I,1)
CALL PLOTTP (SEEABS,SEEORD,NN,0)
IF (LOOK(3),EQ.0) GO TO 83
WRITE(6,81) I,10X,'KETRON MK-86 THESIS',/,11X,
1 'MEAN ERROR IN Y POSITION',/,,
DC 82 I=1,NN
82 SEEORD(I)=RMEAN(I,2)
CALL PLOTTP (SEEABS,SEEORD,NN,0)
IF (LOOK(4),EQ.0) GO TO 86
WRITE(6,84)

```

MCSP05774  
MCSP05775  
MCSP05776  
MCSP05777  
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MCSP0580  
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MCSP0584  
MCSP0585  
MCSP0586  
MCSP0587  
MCSP0588  
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MCSP0591  
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MCSP0600  
MCSP0601  
MCSP0602  
MCSP0603  
MCSP0604  
MCSP0605  
MCSP0606  
MCSP0607  
MCSP0608  
MCSP0609  
MCSP0610  
MCSP0611  
MCSP0612  
MCSP0613  
MCSP0614  
MCSP0615  
MCSP0616  
MCSP0617  
MCSP0618  
MCSP0619  
MCSP0620  
MCSP0621

CCCCC

CCCCC





```

84 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'MEAN ERROR IN Z POSITION',//)
85 I=1,NN
86 SEECD(I)=RMEAN(I,3)
87 CALL PLOTP (SEEABS,SEECD,NN,0)
88 IF (LOOK(5).EQ.0) GO TO 89
89 WRITE (6,87)
90 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'X POSITION VARIANCE',//)
91 I=1,NN
92 SEECD(I)=PKSTOR(I,I)
93 CALL PLOTP (SEEABS,SEECD,NN,0)
94 IF (LOOK(6).EQ.0) GO TO 92
95 WRITE (6,90)
96 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'Y POSITION VARIANCE',//)
97 I=1,NN
98 SEECD(I)=PKSTOR(N2,I)
99 CALL PLOTP (SEEABS,SEECD,NN,0)
100 IF (LOOK(7).EQ.0) GO TO 95
101 WRITE (6,93)
102 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'Z POSITION VARIANCE',//)
103 I=1,NN
104 SEECD(I)=PKSTOR(N3,I)
105 CALL PLOTP (SEEABS,SEECD,NN,0)
106 IF (LOOK(8).EQ.0) GO TO 98
107 WRITE (6,96)
108 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'MEAN VELOCITY ERROR IN X DIRECTION',//)
109 I=1,NN
110 SEECD(I)=RMEAN(I,4)
111 CALL PLOTP (SEEABS,SEECD,NN,0)
112 IF (LOOK(9).EQ.0) GO TO 101
113 WRITE (6,99)
114 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'MEAN VELOCITY ERROR IN Y DIRECTION',//)
115 I=1,NN
116 SEECD(I)=RMEAN(I,5)
117 CALL PLOTP (SEEABS,SEECD,NN,0)
118 IF (LOOK(10).EQ.0) GO TO 104
119 WRITE (6,102)
120 FORMAT (IHL,IOX,'KETRON MK-86 THESIS',/,11X,
1 'MEAN VELOCITY ERROR IN Z DIRECTION',//)
121 I=1,NN
122 SEECD(I)=RMEAN(I,6)
123 CALL PLOTP (SEEABS,SEECD,NN,0)
124 IF (LOOK(11).EQ.0) GO TO 107

```

```

MCSP0622
MCSP0623
MCSP0624
MCSP0625
MCSP0626
MCSP0627
MCSP0628
MCSP0629
MCSP0630
MCSP0631
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MCSP0633
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MCSP0635
MCSP0636
MCSP0637
MCSP0638
MCSP0639
MCSP0640
MCSP0641
MCSP0642
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MCSP0645
MCSP0646
MCSP0647
MCSP0648
MCSP0649
MCSP0650
MCSP0651
MCSP0652
MCSP0653
MCSP0654
MCSP0655
MCSP0656
MCSP0657
MCSP0658
MCSP0659
MCSP0660
MCSP0661
MCSP0662
MCSP0663
MCSP0664
MCSP0665
MCSP0666
MCSP0667
MCSP0668
MCSP0669

```



```

WRITE (6,105)
FORMAT (IHI,ILOX,'KETRON MK-86 THESIS',/,11X,
1 'X VELOCITY VARIANCE',//)
DC 106 I=1,NN
106 SEORD(I)=PKSTOR(2,I)
CALL PLOTP (SEEABS,SEORD,NN,0)
107 IF (LOOK(12),EQ.0) GO TO 110
WRITE (6,108)
108 FORMAT (IHI,ILOX,'KETRON MK-86 THESIS',/,11X,
1 'Y VELOCITY VARIANCE',//)
DC 109 I=1,NN
109 SEORD(I)=PKSTOR(N4,I)
CALL PLOTP (SEEABS,SEORD,NN,0)
110 IF (LOOK(13),EQ.0) GO TO 113
WRITE (6,111)
111 FORMAT (IHI,ILOX,'KETRON MK-86 THESIS',/,11X,
1 'Z VELOCITY VARIANCE',//)
DC 112 I=1,NN
112 SEORD(I)=PKSTOR(N5,I)
CALL PLOTP (SEEABS,SEORD,NN,0)
113 IF (LOOK(14),EQ.0) GO TO 116
WRITE (6,114)
114 FORMAT (IHI,ILOX,'KETRON MK-86 THESIS',/,11X,
1 'TARGET SPEED IN FT/SEC',//)
DC 115 I=1,NN
115 SEORD(I)=DSQRT((RMEAN(I,4)+TVELX(I))**2+(RMEAN(I,5)+TVELY(I))**2+
1 (RMEAN(I,6)+TVELZ(I))**2)*3.D0
CALL PLOTP (SEEABS,SEORD,NN,0)
116 CONTINUE
C
C THE FOLLOWING SEGMENT PUNCHES GAIN MATRICES IF REQUESTED
C
IF (IG.NE.1) GO TO 119
DC 117 I=1,NN
DC 117 J=1,NN
WRITE (7,118) I,J,(GSTOR(J,J,I),JJ=1,3)
117 FORMAT (3X,2I4,3D20.8)
118 STOP
119 FCPMAT(IH0,/,40X,'AVERAGE FILTER VELOCITY ESTIMATE ERROR = ',
1 D20.8,' YARDS/SEC.')
121 FCRMAT(IH0,/,', GAMMA MATRIX',//)
END

```

MCSP0670  
MCSP0671  
MCSP0672  
MCSP0673  
MCSP0674  
MCSP0675  
MCSP0676  
MCSP0677  
MCSP0678  
MCSP0679  
MCSP0680  
MCSP0681  
MCSP0682  
MCSP0683  
MCSP0684  
MCSP0685  
MCSP0686  
MCSP0687  
MCSP0688  
MCSP0689  
MCSP0690  
MCSP0691  
MCSP0692  
MCSP0693  
MCSP0694  
MCSP0695  
MCSP0696  
MCSP0697  
MCSP0698  
MCSP0699  
MCSP0700  
MCSP0701  
MCSP0702  
MCSP0703  
MCSP0704  
MCSP0705  
MCSP0706  
MCSP0707  
MCSP0708  
MCSP0709  
MCSP0710  
MCSP0711

SUBROUTINE NOISE (RT,BT,ET,R1,RB,RE,KERNL1,KERNL2,KERNL3,X,Y,Z,
MCSP0712



```

C C C
1 INDEX)
THIS SUBROUTINE TAKES THE INPUT R,B,E DATA, ADDS NOISE, AND THEN
C CONVERTS INTO NOISY X,Y,Z.
C
IMPLICIT REAL*8(A-H,L-Z)
REAL*4 RERR,BERR,EERR
RERROR = 0
BERROR = 0
EERROR = 0
IF(INDEX.EQ.0) GO TO 1
C
C SET RANGE STANDARD DEVIATION AT INPUT VALUE + .1 PER CENT RANGE
C
RK=PI+.001D0*RT
CALL NORMAL (KERNL1,RERR,1)
RERROR=RR*DBLE(RERR)
CALL NORMAL (KERNL2,BERR,1)
BERROR=RB*DBLE(BERR)
CALL NORMAL (KERNL3,EERR,1)
EERROR=RE*DBLE(EERR)
1 K=RT+RERROR
E=ET+BERROR
C
C CONVERT INTO NOISY X,Y,Z DATA
C
X=R*DCOS(B)*DCOS(E)
Y=R*DSIN(B)*DCOS(E)
Z=R*DSIN(E)
RETURN
END
C C C

```

```

SUBROUTINE UPDATE (N,NUM,RADAR,STATES,PRED,IFLAG,L,TIME,IGAIN,ITER
1,K1,K2,ADAPTQ)
C
C THIS ROUTINE COMPUTES THE FILTER ESTIMATE X(K/K)
C AND FILTER PREDICTION X(K+1/K)
C IT ALSO GENERATES THE SYNTHETIC ESTIMATES IF REQUESTED, AND
C THE ADAPTIVE Q MATRIX IS UPDATED IF DESIRED
C
IMPLICIT REAL*8(P,Q,H,G,R,I,S)
REAL*8 TEMPI,TEMP2,TEMP3,II,AHEAD,EST,K1,K2
INTEGER*4 ADAPTQ
DIMENSION RADAR(9,9), STATES(9,9), PRED(9,9), TEMPI(9,9), TEMP2(9,MCSP07113
MCSP07114
MCSP07115
MCSP07116
MCSP07117
MCSP07118
MCSP07119
MCSP07120
MCSP07121
MCSP07122
MCSP07123
MCSP07124
MCSP07125
MCSP07126
MCSP07127
MCSP07128
MCSP07129
MCSP07130
MCSP07131
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MCSP07145
MCSP07146
MCSP07147
MCSP07148
MCSP07149
MCSP07150
MCSP07151
MCSP07152
MCSP07153
MCSP07154
MCSP07155

```



```

19), TEMP3(9,9), QOLD(9,9), QRESID(9,9), SPD IFF(9,9), SPTR(9,9)
CCMCMCN II(9,9), Q(9,9), H(9,9), R(9,9), G(9,9), PHI(9,9), PKK(9,9), PHIPK
1(S,9), RMEAN(233,6), RSD(6), RESID(233,4), AHEAD(233), EST(9,233),
2M(9,9,40), PKSTOR(9,233)
DC I I=1,9
DC J J=1,9
TEMP1(I,J)=0.00
TEMP2(I,J)=0.00
TEMP3(I,J)=0.00
1 IF(IFLAG.EQ.1) GO TO 2
CC
CC X(K/K) = X(K/K-1) + G(K)(Z(K)-HX(K/K-1))
CALL PRCD (H,PRED,NUM,N,1,TEMP1)
CALL SUB (RADAR,TEMP1,NUM,1,TEMP3)
CALL PRCD (G,TEMP3,N,NUM,1,TEMP2)
CALL ADD (PRED,TEMP2,N,1,STATES)
CC
CC IFLAG = 1 SIGNIFIES THAT THE ESTIMATED POSITIONS ARE TO BE THE
CC MEASURED POSITIONS - THIS OPTION IS ONLY AVAILABLE FOR
CC THE CASES WHERE GAINS ARE CALCULATED BY THE PROGRAM
2 IF(IFLAG.NE.1.OR.IGAIN.EQ.2) GO TO 5
CALL TRANS (PHI,N,N,TEMP2)
CC
CC THE FOLLOWING SEGMENT GENERATES SYNTHETIC FILTER ESTIMATES AND
CC COVARIANCES OF ESTIMATION ERROR IF REQUESTED BY 'OPT' = 1
N2=N/3+1
N3=N*2/3+1
N4=N+1
N5=N3+1
STATES(1,1)=RADAR(1,1)
STATES(N2,1)=RADAR(2,1)
STATES(N3,1)=RADAR(3,1)
IF (L5.EQ.1) GO TO 5
STATES(N4,1)=(RADAR(1,1)-EST(1,L-1))/TIME
STATES(N5,1)=(RADAR(2,1)-EST(N2,L-1))/TIME
STATES(N5,1)=(RADAR(3,1)-EST(N3,L-1))/TIME
PKK(1,1)=1.00
PKK(N2,N2)=1.00
PKK(N3,N3)=1.00
PKK(1,2)=1.00/TIME
PKK(N2,N4)=PKK(1,2)
PKK(N3,N5)=PKK(1,2)
PKK(2,1)=1.00/TIME
PKK(N4,N2)=PKK(2,1)
PKK(N5,N3)=PKK(2,1)

```

```

MCSP0759
MCSP0758
MCSP0757
MCSP0756
MCSP0760
MCSP0761
MCSP0762
MCSP0763
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MCSP0767
MCSP0768
MCSP0769
MCSP0770
MCSP0771
MCSP0772
MCSP0773
MCSP0774
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MCSP0776
MCSP0777
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MCSP0789
MCSP0790
MCSP0791
MCSP0792
MCSP0793
MCSP0794
MCSP0795
MCSP0796
MCSP0797
MCSP0798
MCSP0799
MCSP0800
MCSP0801
MCSP0802
MCSP0803

```





MCSP0804  
 MCSP0805  
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 MCSP0808  
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 MCSP0841  
 MCSP0842  
 MCSP0843  
 MCSP0844  
 MCSP0845  
 MCSP0846  
 MCSP0847  
 MCSP0848  
 MCSP0849  
 MCSP0850  
 MCSP0851

```

    C C C C C
    PKK(2,2)=2.DO/TIME**2
    PKK(N4,N4)=PKK(2,2)
    PKK(N5,N5)=PKK(2,2)
    IF FILTER IS 2ND ORDER OR 3 SAMPLES HAVE NOT BEEN TAKEN TO
    CALCULATE FINAL 3RD ORDER ESTIMATES AND COVARIANCE TERMS,
    SKIP THE FOLLOWING SECTION
    IF (N.EQ.6) GO TO 3
    IF (N.EQ.9.AND.LT.3) GO TO 5
    N6=N4+1
    N7=N5+1
    STATES(N6,1)=(RADAR(1,1)-2.DO*EST(1,L-1)+EST(1,L-2))/TIME**2
    STATES(N7,1)=(RADAR(2,1)-2.DO*EST(N2,L-1)+EST(N2,L-2))/TIME**2
    STATES(N3,1)=(RADAR(3,1)-2.DO*EST(N3,L-1)+EST(N3,L-2))/TIME**2
    PKK(1,3)=1.DO/TIME**2
    PKK(N2,N6)=PKK(1,3)
    PKK(N3,N7)=PKK(1,3)
    PKK(2,3)=3.DO/TIME**3
    PKK(N4,N6)=PKK(2,3)
    PKK(N5,N7)=PKK(2,3)
    PKK(3,1)=1.DO/TIME**2
    PKK(N6,N2)=PKK(3,1)
    PKK(N7,N3)=PKK(3,1)
    PKK(3,2)=3.DO/TIME**3
    PKK(N6,N4)=PKK(3,2)
    PKK(N7,N5)=PKK(3,2)
    PKK(3,3)=6.DO/TIME**4
    PKK(N6,N6)=PKK(3,3)
    PKK(N7,N7)=PKK(3,3)
    DC 4 I=1,N
    DC 4 J=1,N
    PKKM1(I,J)=0.DO
    CALL PRCD (PHI,PKK,N,N,TEMP1)
    CALL PRCD (TEMP1,TEMP2,N,N,N,TEMP3)
    CALL ADD(TEMP3,Q,N,N,PKKM1)
    CCNTINUE
    THE FOLLOWING SEGMENT IS USED TO UPDATE THE ADAPTIVE Q MATRIX
    IF (ADAPTQ.NE.1.OR.IGAIN.EQ.2.OR.(IGAIN.EQ.1.AND.ITER.NE.1)) GO TO
    1 8
    DC 6 I=1,N
    DC 6 J=1,N
    SPDIF(I,J)=STATES(I,J)-PRED(I,J)
    CALL TRANS (SPDIFF,N,N,SPTR)
    CALL PROD (SPDIFF,SPTR,N,N,N,QRESID)
    DC 7 I=1,N
  
```



MCSP0852  
 MCSP0853  
 MCSP0854  
 MCSP0855  
 MCSP0856  
 MCSP0857  
 MCSP0858  
 MCSP0859  
 MCSP0860  
 MCSP0861

```

DC 7 J=1,N
QOLD(I,J)=Q(I,J)
Q(I,J) = KI*QOLD(I,J) + K2*QRESID(I,J)
7 CC CONTINUE
      X(K+1/K) = PHI * X(K/K)
8 CALL PROC (PHI, STATES,N,N,1,PRED)
  RETURN
  END
  
```

CC

MCSP0862  
 MCSP0863  
 MCSP0864  
 MCSP0865  
 MCSP0866  
 MCSP0867  
 MCSP0868  
 MCSP0869  
 MCSP0870  
 MCSP0871  
 MCSP0872  
 MCSP0873  
 MCSP0874  
 MCSP0875  
 MCSP0876  
 MCSP0877  
 MCSP0878  
 MCSP0879  
 MCSP0880  
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 MCSP0882  
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 MCSP0884  
 MCSP0885  
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 MCSP0887  
 MCSP0888  
 MCSP0889  
 MCSP0890  
 MCSP0891  
 MCSP0892  
 MCSP0893  
 MCSP0894

```

SUBROUTINE SCENAR (X,Y,Z, STATES,N,L,ITER,IGAIN,TVELX,TVELY,TVELZ,
1VEL)
  THIS SUBROUTINE COMPUTES THE MEANS OF THE FILTER ESTIMATES
  FOR 'ITER' SAMPLES FOR USE IN CALCULATING THE FILTER
  PERFORMANCE FACTOR AND ALSO THE VARIANCES WHEN GAINS ARE READ
  IMPLICIT REAL*8(P,Q,H,G,R,T,S)
  REAL*8 X, Y,Z,XCAL,YCAL,ZCAL,TERM1,TERM2,II,AHEAD,EST
  DIMENSION II(9,9),Q(9,9),H(9,9),R(9,9),G(9,9),PHI(9,9),PKKM1
  1(9,9),RMEAN(233,6),RSD(6),RESID(233,4),AHEAD(233),EST(9,233),PHIPR
  2N(9,9,40),PKSTOR(9,233)
  N2=N/3+1
  N3=N*2/3+1
  N4=N2+1
  N5=N3+1
  XCAL=STATES(1,1)
  YCAL=STATES(N2,1)
  ZCAL=STATES(N3,1)
  RMEAN'S ARE SUM OF FILTER ESTIMATES IN THIS ROUTINE - AFTER FULL
  MONTE CARLO RUN HAS BEEN CONCLUDED THE MEANS ARE CALCULATED BY
  DIVIDING THESE SUMS BY THE NUMBER OF MONTE CARLO RUNS IN MAIN
  RMEAN(L,1)=RMEAN(L,1)+XCAL
  RMEAN(L,2)=RMEAN(L,2)+YCAL
  RMEAN(L,3)=RMEAN(L,3)+ZCAL
  RMEAN(L,4)=RMEAN(L,4)+STATES(2,1)
  RMEAN(L,5)=RMEAN(L,5)+STATES(N4,1)
  RMEAN(L,6)=RMEAN(L,6)+STATES(N5,1)
  
```

CCCC

```

  RMEAN(L,1)=RMEAN(L,1)+XCAL
  RMEAN(L,2)=RMEAN(L,2)+YCAL
  RMEAN(L,3)=RMEAN(L,3)+ZCAL
  RMEAN(L,4)=RMEAN(L,4)+STATES(2,1)
  RMEAN(L,5)=RMEAN(L,5)+STATES(N4,1)
  RMEAN(L,6)=RMEAN(L,6)+STATES(N5,1)
  
```

CCCC

```

  SIMILARLY, PKSTOR'S ARE THE SUMS OF SQUARES OF ESTIMATION ERROR
  
```

CC













```

DIMENSION A(9,9), B(9,9), C(9,9)
DO 1 I=1,N
DO 1 J=1,M
1 C(I,J)=A(I,J)+B(I,J)
RETURN
END

```

```

MCSP0984
MCSP0985
MCSP0986
MCSP0987
MCSP0988
MCSP0989

```

```

SUBROUTINE SUB (A,B,N,M,C)
THIS SUBROUTINE SUBTRACTS THE NXM MATRIX B FROM THE NXM MATRIX
A AND STORES THE RESULT IN C
C
C
REAL*8 A,B,C
DIMENSION A(9,9), B(9,9), C(9,9)
DO 1 I=1,N
DO 1 J=1,M
1 C(I,J)=A(I,J)-B(I,J)
RETURN
END

```

```

MCSP0990
MCSP0991
MCSP0992
MCSP0993
MCSP0994
MCSP0995
MCSP0996
MCSP0997
MCSP0998
MCSP0999
MCSP1000
MCSP1001

```

```

SUBROUTINE PROD (A,B,N,M,L,C)
THIS SUBROUTINE COMPUTES THE MATRIX PRODUCT AB AND STORES THE
RESULT IN C
A = NXM B = MXL C = NXL
C
C
REAL*8 A,B,C
DIMENSION A(9,9), B(9,9), C(9,9)
DO 1 I=1,N
DO 1 J=1,L
1 C(I,J)=0.DO
DO 2 I=1,N
DO 2 J=1,L
DO 2 K=1,M
2 C(I,J)=C(I,J)+A(I,K)*B(K,J)
RETURN
END

```

```

MCSP1002
MCSP1003
MCSP1004
MCSP1005
MCSP1006
MCSP1007
MCSP1008
MCSP1009
MCSP1010
MCSP1011
MCSP1012
MCSP1013
MCSP1014
MCSP1015
MCSP1016
MCSP1017
MCSP1018

```



MCSP1019  
 MCSP1020  
 MCSP1021  
 MCSP1022  
 MCSP1023  
 MCSP1024  
 MCSP1025  
 MCSP1026  
 MCSP1027  
 MCSP1028  
 MCSP1029  
 MCSP1030  
 MCSP1031

```

SUBROUTINE TRANS (A,N,M,C)
  THIS SUBROUTINE FORMS THE MATRIX TRANSPOSE OF A STORING THE
  RESULT IN C
  A = NXM
  C = MXN
  REAL*8 A,C
  DIMENSION A(9,9), C(9,9)
  DO 1 I=1,N
  DO 1 J=1,M
  C(J,I)=A(I,J)
  1 RETURN
  END
  
```

MCSP1032  
 MCSP1033  
 MCSP1034  
 MCSP1035  
 MCSP1036  
 MCSP1037  
 MCSP1038  
 MCSP1039  
 MCSP1040  
 MCSP1041  
 MCSP1042  
 MCSP1043  
 MCSP1044

```

SUBROUTINE MREAD (A,N,M)
  THIS SUBROUTINE READS AN NXM MATRIX A ACCORDING TO THE FORMAT
  8D10.5. THE ENTRIES IN THE FIRST ROW OF A ARE READ FIRST, THEN
  THE ENTRIES IN THE SECOND ROW, ETC.
  REAL*8 A
  DIMENSION A(9,9)
  DO 1 I=1,N
  READ (5,2) (A(I,J),J=1,M)
  1 FORMAT (8D10.5)
  RETURN
  END
  
```

MCSP1045  
 MCSP1046  
 MCSP1047  
 TMCSP1048  
 MCSP1049  
 MCSP1050  
 MCSP1051  
 MCSP1052  
 MCSP1053  
 MCSP1054  
 MCSP1055  
 MCSP1056

```

SUBROUTINE MREAD1 (A,N,M)
  THIS SUBROUTINE READS AN NXM MATRIX A ACCORDING TO THE FORMAT
  10X,3D20.8). THE ENTRIES IN THE FIRST ROW OF A ARE READ FIRST,
  THE ENTRIES IN THE SECOND ROW, ETC.
  REAL*8 A
  DIMENSION A(9,9)
  DO 1 I=1,N
  READ (5,2) (A(I,J),J=1,M)
  1 FORMAT (10X,3D20.8)
  RETURN
  END
  
```



END

MCSP1057

```

C
C
SUBROUTINE MWRITE (A,N,M)
  THIS SUBROUTINE WRITES THE ENTRIES OF THE NXM MATRIX A
  REAL*8 A
  DIMENSION A(9,9)
  DO 1 I=1,N
  DO 2 J=1,M
    WRITE (6,2) (I,J,A(I,J),J=1,M)
  1 FORMAT (2(3X,1(I,1,1,1))= ',1PD19.12)
  2
  RETURN
  END

```

MCSP1058  
 MCSP1059  
 MCSP1060  
 MCSP1061  
 MCSP1062  
 MCSP1063  
 MCSP1064  
 MCSP1065  
 MCSP1066  
 MCSP1067  
 MCSP1068

```

C
C
SUBROUTINE GAUSS3 (N,EPSI,A,X,KER,K)
  THIS SUBROUTINE INVERTS AN N X N MATRIX THAT HAS BEEN
  DIMENSIONED K X K IN THE CALLING PROGRAM
  REAL*8 A,X,Y,D
  DIMENSION A(I), X(1), L(9), M(9), Y(9,9)
  DO 1 I=1,N
  DO 2 J=1,N
    IND=(I-1)*K+J
    Y(I,J)=A(IND)
  1 KER=1
  N2=2*N
  CALL ARRAY (2,N,N,9,9,Y,Y)
  CALL DMINV (Y,N,D,L,M)
  CALL ARRAY (1,N,N,9,9,Y,Y)
  IF (D.EQ.0) KER=2
  DO 2 I=1,N
  DO 2 J=1,N
    IND=(I-1)*K+J
    X(IND)=Y(I,J)
  2
  RETURN
  END

```

MCSP1069  
 MCSP1070  
 MCSP1071  
 MCSP1072  
 MCSP1073  
 MCSP1074  
 MCSP1075  
 MCSP1076  
 MCSP1077  
 MCSP1078  
 MCSP1079  
 MCSP1080  
 MCSP1081  
 MCSP1082  
 MCSP1083  
 MCSP1084  
 MCSP1085  
 MCSP1086  
 MCSP1087  
 MCSP1088  
 MCSP1089  
 MCSP1090  
 MCSP1091



```

SUBROUTINE ARRAY (MODE,I,J,N,M,S,D)
DIMENSION S(1), D(1)
REAL*8 S,D
NI=N-I
IF (MODE-1) 1,1,3
1 IJ=I*J+1
  NM=N*J+1
  DC 2 K=1,J
  NM=NM-NI
  DC 2 L=1,I
  IJ=IJ-1
  NM=NM-1
2 D(NM)=S(IJ)
3 IJ=0
  NM=0
  DC 5 K=1,J
  DC 4 L=1,I
  IJ=IJ+1
  NM=NM+1
4 S(IJ)=D(NM)
5 NM=NM+NI
6 RETURN
END

```

```

MCSP11092
MCSP11093
MCSP11094
MCSP11095
MCSP11096
MCSP11097
MCSP11098
MCSP11099
MCSP11100
MCSP11101
MCSP11102
MCSP11103
MCSP11104
MCSP11105
MCSP11106
MCSP11107
MCSP11108
MCSP11109
MCSP11110
MCSP11111
MCSP11112
MCSP11113
MCSP11114
MCSP11115

```

```

SUBROUTINE RNOISE (XM,RB1,RRO,REL,N)

```

```

THIS SUBROUTINE CALCULATES THE COVARIANCE CF MEASUREMENT NOISE
MATRIX, R, FOR ONLINE CALCULATION IF REQUESTED

```

```

RB1,RR1,REL ARE INPUT AS STANDARD DEVIATIONS

```

```

IMPLICIT REAL*8(P,Q,H,G,R,T,B)
REAL*8 I,X,DUMMY,RAN,BEAR,ELEV,SINB,SINE,CCSB,COSE,SSQB,SSQE,CSQB
1,CSQE,RSQ,XM,AHEAD,EST
COMMON I(9,9),Q(9,9),H(9,9),G(9,9),PHI(9,9),PKK(9,9),PKKM1
1(9,9),KMEAN(233,6),RSD(6),RESID(233,4),AHEAD(233),EST(9,233),PHIPR
2M(9,9,40),PKSTOR(9,233)
DIMENSION X(9), XM(9,9)
DC 1 I=1,9
1 X(I)=XM(I,1)
PI=3.14159265358979323846
N2=N/3+1

```

```

MCSP11116
MCSP11117
MCSP11118
MCSP11119
MCSP11120
MCSP11121
MCSP11122
MCSP11123
MCSP11124
MCSP11125
MCSP11126
MCSP11127
MCSP11128
MCSP11129
MCSP11130
MCSP11131
MCSP11132
MCSP11133
MCSP11134

```

```

CCCCC

```





```

N3=N*2/3+1
RAN=DSQRT(X(1)**2+X(N2)**2+X(N3)**2)
RR1=RR0+.001D0*GAN
IF(X(1).EQ.0.0) GO TO 2
IF(X(N2).EQ.0.0D0.AND.X(1).GT.0.0D0) BEAR=0.0D0
IF(X(N2).EQ.0.0D0.AND.X(1).LT.0.0D0) BEAR=PI
IF(X(N2).NE.0.0D0) BEAR=DATAN(X(N2)/X(1))
GO TO 3
2 IF(X(N2).GT.0.0) BEAR = PI/2.0
IF(X(N2).LT.0.0) BEAR=3.*PI/2.
3 DUMMY=X(N3)/RAN
GO TO 5
4 BEAR=0.0D0
DUMMY=1.0D0
5 IF(DUMMY.NE.1.0) ELEV=PI/2.0
IF(DUMMY.NE.1.0) ELEV=DATAN(DUMMY/DSQRT(1.-DUMMY**2))
SINB=DSIN(BEAR)
COSB=DCOS(BEAR)
SSQB=SINB**SINE
CSQB=COSB**SINE
RSQB=COSB**RE
RRE=RSQB**RAN**RBI
RRI=RRE**RBI
RE=RE1**RRI
R(1,1)=RSQB**RE*CSQB**SSQE+RSQ**R*CSQB**SSQB**SSQE+RR*CSQB**SSQE
R(1,2)=RSQ**SINB**SSQE*(RE-R*RE)+SINB**CSQB**SSQE*(RR-RSQ**R)
R(2,1)=RSQ**RE**SSQB**SSQE+RSQ**R*CSQB**SSQB**RE*CSQB**SSQE+RR*SSQB**SSQE
R(2,2)=RSQ**SINE**COSE*(RR-RSQ**RE)
R(2,3)=R(2,2)
R(1,3)=COSE**SINE**COSB*(RR-RSQ**RE)
R(3,1)=R(1,3)
R(3,3)=RSQ**CSQB**RE+SSQE**RR
RETURN
END

```

MCSP11135  
MCSP11136  
MCSP11137  
MCSP11138  
MCSP11139  
MCSP11140  
MCSP11141  
MCSP11142  
MCSP11143  
MCSP11144  
MCSP11145  
MCSP11146  
MCSP11147  
MCSP11148  
MCSP11149  
MCSP11150  
MCSP11151  
MCSP11152  
MCSP11153  
MCSP11154  
MCSP11155  
MCSP11156  
MCSP11157  
MCSP11158  
MCSP11159  
MCSP11160  
MCSP11161  
MCSP11162  
MCSP11163  
MCSP11164  
MCSP11165  
MCSP11166  
MCSP11167  
MCSP11168  
MCSP11169  
MCSP11170  
MCSP11171  
MCSP11172  
MCSP11173  
MCSP11174  
MCSP11175  
MCSP11176

MCSP11177

SUBROUTINE ACCUR (X,Y,Z,L,N,ITER)



MCSPI1178  
 MCSPI1179  
 MCSPI1180  
 MCSPI1181  
 MCSPI1182  
 MCSPI1183  
 MCSPI1184  
 MCSPI1185  
 MCSPI1186  
 MCSPI1187  
 MCSPI1188  
 MCSPI1189  
 MCSPI1190  
 MCSPI1191  
 MCSPI1192  
 MCSPI1193  
 MCSPI1194  
 MCSPI1195  
 MCSPI1196  
 MCSPI1197  
 MCSPI1198  
 MCSPI1199  
 MCSPI1200  
 MCSPI1201  
 MCSPI1202  
 MCSPI1203  
 MCSPI1204  
 MCSPI1205  
 MCSPI1206  
 MCSPI1207  
 MCSPI1208

THIS SUBROUTINE CALCULATES THE SHELL AT TARGET RESIDUALS IF  
 REQUESTED

```

IMPLICIT REAL*8(Q,H,R,G,P,A,B,T,S,X,Y,Z,E)
REAL*8 DIFF,II
DIMENSION A(9,9), Q(9,9), H(9,9), R(9,9), G(9,9), PHI(9,9), PKK(9,9), PKKMI(9,9),
CMMCN(233,6), RSD(6), RESID(233,4), AHEAD(233), EST(9,233), PHIPRM(9,233)
1(9,9), RMEAN(233,6), RSD(6), RESID(233,4), AHEAD(233), EST(9,233), PHIPRM(9,233)
2M(9,9,40), PKSTOR(9,233)
DIFF=FLOAT(L)-4.00*AHEAD(L)
LESS=DIFF
DO 2 I=1,9
  DO 1 J=2,9
    A(I,J)=EST(I,LESS)
  1 AHEAD=AHEAD(L)*4.00+1.00
DO 3 I=1,9
  DO 3 J=1,9
    TEMP(I,J)=PHIPRM(I,J,IAHEAD)
  CALL PRGD (TEMP,A,N,N,1,TRY)
  N2=N/3+1
  N3=N*2/3+1
  XCAL=TRY(1,1)
  YCAL=TRY(N2,1)
  ZCAL=TRY(N3,1)
  RESID(L,1)=RESID(L,1)+(XCAL-X)
  RESID(L,2)=RESID(L,2)+(YCAL-Y)
  RESID(L,3)=RESID(L,3)+(ZCAL-Z)
RETURN
END
  
```

C  
 C  
 C

MCSPI1209  
 MCSPI1210  
 MCSPI1211  
 MCSPI1212  
 MCSPI1213  
 MCSPI1214  
 MCSPI1215  
 MCSPI1216  
 MCSPI1217  
 MCSPI1218  
 MCSPI1219  
 MCSPI1220

SUBROUTINE STUDY ( I,X,Y,Z,RESIDX,RESIDY,RESIDZ,AHEAD,RMISS)  
 THIS ROUTINE OUTPUTS THE SHELL AT TARGET RESIDUAL  
 PERFORMANCE TABLE IF REQUESTED

```

REAL*8 X,Y,Z,RESIDX,RESIDY,RESIDZ,CALX,CALY,CALZ,AHEAD,TIME,RMISS
CALX=X+RESIDX
CALY=Y+RESIDY
CALZ=Z+RESIDZ
TIME=FLOAT(I)-AHEAD*4.00
WRITE (6,1) I,X,CALX,RESIDX,AHEAD
WRITE (6,2) Y,CALY,RESIDY,RMISS
  
```

C  
 C  
 C



```

1 WRITE (6,3) Z,CALZ,RESIDZ,TIME
2 FFORMAT (1H0,I4,I,X=,3D15.6,D14.5)
3 FFORMAT (1H,5X,I,Y=,3D15.6,-----
  FFORMAT (1H,5X,I,Z=,3D15.6,D14.5,/,
  ,D15.6)
  RETURN
  END

```

```

MCSPI1221
MCSPI1222
MCSPI1223
MCSPI1224
MCSPI1225
MCSPI1226

```



APPENDIX D

LISTING OF SHELL FLIGHT TIME INTERPOLATION PROGRAM

The program that interpolates in the 5 in - 54 gunfire table (included after the program listing) to generate the shell-at-target flight times is included in this appendix. In its present form it can provide times for tracks operating in a region bounded by

|                         |               |
|-------------------------|---------------|
| minimum range           | - 500 yards   |
| maximum range           | -7500 yards   |
| minimum elevation angle | - 15 degrees  |
| minimum elevation angle | - 90 degrees. |

If track data points fall outside of these limits the program can be easily expanded by enlarging the arrays and revising the linear interpolation equations.





```

CCCCC
THIS PROGRAM GENERATES THE SHELL FLIGHT TIMES TO POINTS ON A
TRACK BY INTERPOLATING IN A 5 IN 54 GUNFIRE TABLE.
BOTH THE INTERPOLATION ARRAY AND THE TRACK DATA ARE INPUT,
AND THE FLIGHT TIMES ARE PUNCHED.

REAL*8 X,Y,Z,R,E,ARRAY,TIME,RSTEP,ESTEP,RSTEP1,ESTEP1,T1,T2,FRAC
DIMENSION ARRAY(15,16)

CC   READ THE INTERPOLATION ARRAY BY ROWS
100  READ(5,100) ((ARRAY(I,J),J=2,16),I=1,15)
      FORMAT(8D10.0,/,7D10.0)

CCCC  THE FOLLOWING READ STATEMENT IS A SUPPLEMENT TO READ
      THE FIRST COLUMN
200  READ(5,100) (ARRAY(I,1),I=1,15)
      WRITE(6,200) ((ARRAY(I,J),J=1,8),I=1,15)
      FCORMAT(1H1,10X,INTERPOLATION ARRAY,/,/(2X,8D16.4))
300  WRITE(6,300) ((ARRAY(I,J),J=9,16),I=1,15)
      FCORMAT(1H0,/,/(2X,8D16.4))
      WRITE(6,50)
50   FCORMAT(1H1,14X,'X',14X,'Y',14X,'Z',14X,'R',14X,'E',12X,
1    'TIME',10X,'CL TIME',9X,'STEPS',/)
      DO 1000 I=1,233

CCCC  READ THE INPUT TRACK DATA
150  READ(5,150) X,Y,Z
      FCORMAT(8X,3D20.0)
      R=DSQRT(X**2+Y**2+Z**2)
      E=DATAN2(Z,DSQRT(X**2+Y**2))
      ESTEP=500.00
      RSTEP=15.00
      INDR=1
      INDE=1
6    IF(R.LT.RSTEP) GO TO 7
      INDR=INDR+1
      GO TO 6
7    RSTEP1= RSTEP-500.00
8    IF(E.LT.ESTEP) GO TO 10
      ESTEP=ESTEP+5.00
      INDE=INDE+1
      GO TO 8
10   ESTEP1 = ESTEP-5.00

```



```

FRAC = (R-(INDR-1)*500.D0)/500.D0
T1 = ARRAY(INDR-1, INDE-1)+FRAC*(ARRAY(INDR, INDE-1)-ARRAY(INDR-1,
1 INDE-1))
T2 = ARRAY(INDR-1, INDE)+FRAC*(ARRAY(INDR, INDE)-ARRAY(INDR-1, INDE))

C THE FLIGHT TIME IS CALCULATED IN SECS
C
TIME = T1+(E-ESTEPI )/5.D0*(T2-T1)
WRITE(6,51) I,X,Y,Z,R,E,TIME
FORMAT(IH0,I3,6D15.6,5X,'-----',10X,'-----')
CCONTINUE
51
1000 STOP
END

```



5 INCH 54 CALIBER GUN I.V. 2500 F.S.

TIME OF FLIGHT IN SECONDS

POSITION ANGLE IN DEGREES

| SLANT RANGE YARDS | 0     | 5     | 10    | 15    | 20    | 25    | 30    | 35    | 40    | 45    | 50    | 55    | 60    | 65    | 70    | 75    | 80    | 85    | 90    |
|-------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| 500               | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  | 0.61  |
| 1000              | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.24  | 1.25  |
| 1500              | 1.88  | 1.88  | 1.89  | 1.89  | 1.89  | 1.89  | 1.89  | 1.89  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  | 1.90  |
| 2000              | 2.55  | 2.55  | 2.56  | 2.56  | 2.56  | 2.56  | 2.57  | 2.57  | 2.57  | 2.57  | 2.58  | 2.58  | 2.58  | 2.58  | 2.58  | 2.58  | 2.58  | 2.58  | 2.58  |
| 2500              | 3.24  | 3.24  | 3.25  | 3.25  | 3.25  | 3.26  | 3.26  | 3.27  | 3.27  | 3.27  | 3.28  | 3.28  | 3.28  | 3.29  | 3.29  | 3.29  | 3.29  | 3.29  | 3.29  |
| 3000              | 3.95  | 3.95  | 3.96  | 3.97  | 3.97  | 3.98  | 3.98  | 3.99  | 4.00  | 4.00  | 4.01  | 4.01  | 4.01  | 4.02  | 4.02  | 4.02  | 4.02  | 4.02  | 4.02  |
| 3500              | 4.68  | 4.69  | 4.70  | 4.71  | 4.71  | 4.72  | 4.73  | 4.74  | 4.75  | 4.75  | 4.76  | 4.77  | 4.77  | 4.78  | 4.78  | 4.78  | 4.78  | 4.78  | 4.78  |
| 4000              | 5.44  | 5.45  | 5.46  | 5.47  | 5.48  | 5.49  | 5.50  | 5.51  | 5.52  | 5.53  | 5.54  | 5.55  | 5.55  | 5.56  | 5.56  | 5.57  | 5.57  | 5.57  | 5.57  |
| 4500              | 6.22  | 6.24  | 6.25  | 6.26  | 6.28  | 6.29  | 6.30  | 6.31  | 6.33  | 6.34  | 6.35  | 6.36  | 6.37  | 6.37  | 6.38  | 6.38  | 6.39  | 6.39  | 6.39  |
| 5000              | 7.04  | 7.05  | 7.07  | 7.08  | 7.10  | 7.11  | 7.13  | 7.14  | 7.16  | 7.17  | 7.19  | 7.20  | 7.21  | 7.22  | 7.23  | 7.23  | 7.24  | 7.24  | 7.24  |
| 5500              | 7.88  | 7.89  | 7.91  | 7.93  | 7.95  | 7.97  | 7.99  | 8.01  | 8.02  | 8.04  | 8.06  | 8.07  | 8.08  | 8.09  | 8.10  | 8.11  | 8.12  | 8.12  | 8.12  |
| 6000              | 8.75  | 8.77  | 8.79  | 8.81  | 8.83  | 8.85  | 8.87  | 8.90  | 8.92  | 8.94  | 8.96  | 8.98  | 8.99  | 9.01  | 9.02  | 9.03  | 9.03  | 9.04  | 9.04  |
| 6500              | 9.65  | 9.67  | 9.70  | 9.72  | 9.74  | 9.77  | 9.80  | 9.82  | 9.85  | 9.87  | 9.89  | 9.91  | 9.93  | 9.95  | 9.96  | 9.98  | 9.98  | 9.99  | 9.99  |
| 7000              | 10.59 | 10.61 | 10.64 | 10.66 | 10.69 | 10.72 | 10.75 | 10.78 | 10.81 | 10.84 | 10.87 | 10.89 | 10.91 | 10.93 | 10.95 | 10.96 | 10.97 | 10.98 | 10.98 |
| 7500              | 11.56 | 11.59 | 11.61 | 11.64 | 11.67 | 11.71 | 11.74 | 11.78 | 11.81 | 11.84 | 11.88 | 11.91 | 11.93 | 11.96 | 11.99 | 11.99 | 12.01 | 12.01 | 12.02 |
| 8000              | 12.57 | 12.60 | 12.63 | 12.66 | 12.69 | 12.73 | 12.77 | 12.81 | 12.85 | 12.89 | 12.93 | 12.96 | 13.00 | 13.02 | 13.05 | 13.07 | 13.08 | 13.09 | 13.09 |
| 8500              | 13.62 | 13.65 | 13.68 | 13.71 | 13.75 | 13.79 | 13.84 | 13.88 | 13.93 | 13.98 | 14.02 | 14.06 | 14.10 | 14.14 | 14.17 | 14.19 | 14.21 | 14.22 | 14.22 |
| 9000              | 14.71 | 14.74 | 14.77 | 14.81 | 14.85 | 14.90 | 14.95 | 15.00 | 15.06 | 15.11 | 15.16 | 15.21 | 15.26 | 15.30 | 15.34 | 15.36 | 15.38 | 15.40 | 15.40 |
| 9500              | 15.84 | 15.87 | 15.90 | 15.95 | 15.99 | 16.05 | 16.11 | 16.17 | 16.23 | 16.29 | 16.36 | 16.42 | 16.47 | 16.52 | 16.56 | 16.59 | 16.62 | 16.63 | 16.64 |
| 10000             | 17.02 | 17.05 | 17.08 | 17.13 | 17.18 | 17.24 | 17.31 | 17.38 | 17.46 | 17.53 | 17.60 | 17.67 | 17.74 | 17.80 | 17.85 | 17.89 | 17.92 | 17.94 | 17.94 |
| 10500             | 18.25 | 18.27 | 18.31 | 18.36 | 18.42 | 18.49 | 18.57 | 18.65 | 18.74 | 18.83 | 18.91 | 19.00 | 19.07 | 19.14 | 19.20 | 19.25 | 19.29 | 19.31 | 19.32 |
| 11000             | 19.53 | 19.55 | 19.59 | 19.64 | 19.71 | 19.79 | 19.88 | 19.98 | 20.08 | 20.18 | 20.29 | 20.39 | 20.48 | 20.57 | 20.64 | 20.70 | 20.74 | 20.77 | 20.78 |
| 11500             | 20.86 | 20.88 | 20.92 | 20.98 | 21.06 | 21.15 | 21.25 | 21.37 | 21.49 | 21.61 | 21.74 | 21.86 | 21.97 | 22.07 | 22.16 | 22.23 | 22.29 | 22.32 | 22.33 |
| 12000             | 22.23 | 22.26 | 22.30 | 22.37 | 22.46 | 22.57 | 22.69 | 22.83 | 22.97 | 23.12 | 23.27 | 23.41 | 23.55 | 23.67 | 23.78 | 23.87 | 23.93 | 23.97 | 23.99 |
| 12500             | 23.66 | 23.69 | 23.74 | 23.82 | 23.92 | 24.05 | 24.20 | 24.36 | 24.53 | 24.71 | 24.89 | 25.06 | 25.22 | 25.37 | 25.50 | 25.61 | 25.69 | 25.74 | 25.75 |
| 13000             | 25.13 | 25.16 | 25.23 | 25.33 | 25.45 | 25.60 | 25.77 | 25.97 | 26.17 | 26.38 | 26.60 | 26.81 | 27.01 | 27.19 | 27.35 | 27.48 | 27.58 | 27.64 | 27.67 |
| 13500             | 26.62 | 26.68 | 26.77 | 26.88 | 27.03 | 27.21 | 27.42 | 27.65 | 27.89 | 28.15 | 28.41 | 28.67 | 28.92 | 29.15 | 29.35 | 29.52 | 29.65 | 29.73 | 29.75 |
| 14000             | 28.15 | 28.23 | 28.34 | 28.49 | 28.67 | 28.89 | 29.14 | 29.41 | 29.71 | 30.03 | 30.35 | 30.67 | 30.99 | 31.29 | 31.55 | 31.77 | 31.94 | 32.05 | 32.08 |
| 14500             | 29.71 | 29.82 | 29.96 | 30.15 | 30.37 | 30.63 | 30.93 | 31.27 | 31.64 | 32.03 | 32.44 | 32.86 | 33.27 | 33.67 | 34.02 | 34.32 | 34.56 | 34.70 | 34.75 |
| 15000             | 31.30 | 31.44 | 31.63 | 31.85 | 32.13 | 32.45 | 32.82 | 33.24 | 33.70 | 34.20 | 34.73 | 35.29 | 35.85 | 36.39 | 36.89 | 37.33 | 37.68 | 37.90 | 37.97 |
| 15500             | 32.92 | 33.11 | 33.33 | 33.62 | 33.95 | 34.35 | 34.81 | 35.34 | 35.93 | 36.59 | 37.30 | 38.06 | 38.86 | 39.68 | 40.47 | 41.20 | 41.80 | 42.21 | 42.35 |
| 16000             | 34.58 | 34.81 | 35.09 | 35.44 | 35.86 | 36.35 | 36.93 | 37.60 | 38.38 | 39.27 | 40.27 | 41.41 | 42.70 | 44.17 | 45.91 | 48.26 |       |       |       |
| 16500             | 36.27 | 36.56 | 36.91 | 37.33 | 37.85 | 38.47 | 39.21 | 40.10 | 41.15 | 42.42 | 43.98 | 46.01 | 49.31 |       |       |       |       |       |       |
| 17000             | 38.01 | 38.36 | 38.79 | 39.31 | 39.95 | 40.74 | 41.70 | 42.90 | 44.43 | 46.47 | 49.76 |       |       |       |       |       |       |       |       |
| 17500             | 39.78 | 40.21 | 40.73 | 41.38 | 42.18 | 43.20 | 44.50 | 46.22 | 48.73 |       |       |       |       |       |       |       |       |       |       |
| 18000             | 41.61 | 42.12 | 42.76 | 43.56 | 44.58 | 45.93 | 47.77 |       |       |       |       |       |       |       |       |       |       |       |       |
| 18500             | 43.48 | 44.11 | 44.88 | 45.80 | 47.21 | 49.07 |       |       |       |       |       |       |       |       |       |       |       |       |       |
| 19000             | 45.42 | 46.17 | 47.13 | 48.39 |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |
| 19500             | 47.44 | 48.34 | 49.52 |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |
| 20000             | 49.53 | 48.67 |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |



## APPENDIX E

### SAMPLE SIMULATION

A sample problem is included to display a representative input data set and the output that can be expected from the MCSP. A description of the desired simulation follows:

Number of Monte-Carlo runs - 100,

Track has 233 points with 4-Hz, sampling rate with approximate velocity data available,

Direction of approach - 45°,

Order of filter - 6,

Initialization by synthetic measurements is desired,

Adaptive  $\underline{Q}$  with  $K_1 = .9$ ,  $K_2 = .1$ ,

Noise statistics are  $\sigma_R = 5.0 + .001 \times \text{Range}$

$\sigma_B = \sigma_E = .002$  radians,

Shell flight times are available, and shell-at-target accuracy calculations are desired,

$\underline{\phi}$ ,  $\underline{\Gamma}$ , and  $\underline{H}$  are standard 6th order matrices,

Desired output includes all available plots.

An abridged listing of the input data appears next followed by representative portions of the output produced.









```

10 0.54388831170 04 0.0 0.43026909950 04
10 -0.130+9261820 03 0.0 -0.27150069810 02
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **

```

TRACK DATA CONTINUES IN SAME FCRMAT

```

0.51702916290 04 -0.14898673760 04 0.183223115360 04
-0.22767851600 03 -0.0930786678660 02 0.43371113810 02
-0.52283387310 04 -0.15128732770 04 0.18430978380 04
-0.22962578050 03 -0.088165457970 02 0.43371113810 02
-0.52868667220 04 -0.15346267080 04 0.185338716490 04
-0.23146703370 03 -0.063211533920 02 0.43371113810 02
-0.53458485810 04 -0.155511176290 04 0.186646317260 02
-0.23320142540 03 -0.073219193550 02 0.43371113810 02
0.11920540 02
0.111370800 02
0.110813100 02
0.110247610 02
0.109574520 02
0.1089093590 02
0.108506220 02
0.107914330 02
0.107357050 02
0.106753370 02
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **

```

SHELL FLIGHT TIMES CONTINUE IN SAME FCRMAT

```

0.816023150 01
0.820923710 01
0.837852630 01
0.846804230 01
0.859773040 01
** ** ** **
** ** ** **
** ** ** **
** ** ** **
** ** ** **

```

END OF DATA SET FOR THIS EXAMPLE



ADAPTIVE - THESIS SAMPLE  
ORDER OF FILTER= 0

NUMBER OF MONTE CARLO PUNS= 100  
CCOE FOR DIRECTION OF APPROACH= 1  
TRANSLATION OF X DATA PCINIS= 0.0  
Y DATA PCINIS= 0.0  
Z DATA PCINIS= 0.0

NCISE CODE= 1

PHI MATRIX

|        |                    |     |        |                    |     |
|--------|--------------------|-----|--------|--------------------|-----|
| (1,1)= | 1.0000000000000000 | 0.0 | (1,2)= | 2.5000000000000000 | -01 |
| (1,3)= | 0.0                | 0.0 | (1,4)= | 0.0                | 0.0 |
| (1,5)= | 0.0                | 0.0 | (1,6)= | 1.0000000000000000 | 00  |
| (2,1)= | 0.0                | 0.0 | (2,2)= | 0.0                | 0.0 |
| (2,3)= | 0.0                | 0.0 | (2,4)= | 0.0                | 0.0 |
| (2,5)= | 0.0                | 0.0 | (2,6)= | 0.0                | 0.0 |
| (3,1)= | 1.0000000000000000 | 0.0 | (3,2)= | 0.0                | 0.0 |
| (3,3)= | 0.0                | 0.0 | (3,4)= | 2.5000000000000000 | -01 |
| (3,5)= | 0.0                | 0.0 | (3,6)= | 0.0                | 0.0 |
| (4,1)= | 0.0                | 0.0 | (4,2)= | 0.0                | 0.0 |
| (4,3)= | 0.0                | 0.0 | (4,4)= | 1.0000000000000000 | 00  |
| (4,5)= | 0.0                | 0.0 | (4,6)= | 0.0                | 0.0 |
| (5,1)= | 0.0                | 0.0 | (5,2)= | 0.0                | 0.0 |
| (5,3)= | 1.0000000000000000 | 0.0 | (5,4)= | 2.5000000000000000 | -01 |
| (5,5)= | 0.0                | 0.0 | (5,6)= | 0.0                | 0.0 |
| (6,1)= | 0.0                | 0.0 | (6,2)= | 0.0                | 0.0 |
| (6,3)= | 0.0                | 0.0 | (6,4)= | 0.0                | 0.0 |
| (6,5)= | 0.0                | 0.0 | (6,6)= | 1.0000000000000000 | 00  |

H CR MEASUREMENT MATRIX

|        |                    |     |        |     |     |
|--------|--------------------|-----|--------|-----|-----|
| (1,1)= | 1.0000000000000000 | 0.0 | (1,2)= | 0.0 | 0.0 |
| (1,3)= | 0.0                | 0.0 | (1,4)= | 0.0 | 0.0 |
| (1,5)= | 0.0                | 0.0 | (1,6)= | 0.0 | 0.0 |
| (2,1)= | 0.0                | 0.0 | (2,2)= | 0.0 | 0.0 |
| (2,3)= | 1.0000000000000000 | 0.0 | (2,4)= | 0.0 | 0.0 |
| (2,5)= | 0.0                | 0.0 | (2,6)= | 0.0 | 0.0 |
| (3,1)= | 0.0                | 0.0 | (3,2)= | 0.0 | 0.0 |
| (3,3)= | 0.0                | 0.0 | (3,4)= | 0.0 | 0.0 |
| (3,5)= | 1.0000000000000000 | 0.0 | (3,6)= | 0.0 | 0.0 |

GAMMA MATRIX

|        |     |        |     |
|--------|-----|--------|-----|
| (1,1)= | 0.0 | (1,2)= | 0.0 |
| (1,3)= | 0.0 | (2,2)= | 0.0 |
| (2,1)= | 0.0 | (3,2)= | 0.0 |
| (2,3)= | 0.0 | (4,2)= | 0.0 |
| (3,1)= | 0.0 | (5,2)= | 0.0 |
| (3,3)= | 0.0 | (6,2)= | 0.0 |
| (4,1)= | 0.0 |        |     |
| (4,3)= | 0.0 |        |     |
| (5,1)= | 0.0 |        |     |
| (5,3)= | 0.0 |        |     |
| (6,1)= | 0.0 |        |     |
| (6,3)= | 0.0 |        |     |



P(K/K-1) OR PREDICTION COVARIANCE MATRIX

```

(1,1) = 0.0
(1,2) = 0.0
(1,3) = 0.0
(1,4) = 0.0
(1,5) = 0.0
(2,1) = 0.0
(2,2) = 0.0
(2,3) = 0.0
(2,4) = 0.0
(2,5) = 0.0
(3,1) = 0.0
(3,2) = 0.0
(3,3) = 0.0
(3,4) = 0.0
(3,5) = 0.0
(4,1) = 0.0
(4,2) = 0.0
(4,3) = 0.0
(4,4) = 0.0
(4,5) = 0.0
(5,1) = 0.0
(5,2) = 0.0
(5,3) = 0.0
(5,4) = 0.0
(5,5) = 0.0
(6,1) = 0.0
(6,2) = 0.0
(6,3) = 0.0
(6,4) = 0.0
(6,5) = 0.0
    
```

INITIAL COVARIANCES OF RANDOM FORCING MATRIX OF STATE EXCITATION FOR ADAPTIVE Q FILTER

WEIGHTING FACTORS FOR C MATRIX UPDATE ARE - K1= 0.9000000 00  
 K2= 0.1000000 00

```

(1,1) = 0.0
(1,2) = 0.0
(1,3) = 0.0
(2,1) = 0.0
(2,2) = 0.0
(3,1) = 0.0
(3,2) = 0.0
(3,3) = 0.0
    
```

R CR COVARIANCE OF MEAS. ERROR MATRIX GENERATED BY PROGRAM

RANGE STAN OE,RR= 5.0000000000000 00  
 BEARING STAN OE,RB= 2.0000000000000-03  
 ELEVATION STAN OE,RE= 2.0000000000000-03

THE FIRST 2 ESTIMATES OF POSITION WILL BE THE OBSERVED POSITION

STATE INITIAL CONDITIONS ARE

```

(1,1) = 4.0727172264900 03
(2,1) = 0.0
(3,1) = 4.0632459995820 03
(4,1) = 0.0
(5,1) = 4.3510848373670 03
(6,1) = 0.0
    
```





| POINT NUMBER | TRUE TRACK POINTS | FILTER TRACK POINTS | RESIDUALS     | TIME OF FLT TRACK TIME WHEN FIRED | MEAN MISS DISTANCE |
|--------------|-------------------|---------------------|---------------|-----------------------------------|--------------------|
|              |                   |                     |               |                                   |                    |
| 37           | X = 0.3264150 04  | 0.3262890 04        | -0.1257670 01 | 0.896300 01                       | 0.1515640 04       |
|              | Y = 0.3264150 04  | 0.2903960 04        | -0.1250190 04 | 0.114820 01                       |                    |
|              | Z = 0.3869000 04  | 0.4711080 04        | 0.8420770 03  |                                   |                    |
| 38           | X = 0.3244460 04  | 0.3204730 04        | -0.3972560 02 | 0.889610 01                       | 0.4169030 03       |
|              | Y = 0.3244460 04  | 0.370230 04         | -0.7420330 02 | 0.241540 01                       |                    |
|              | Z = 0.3845640 04  | 0.4255960 04        | 0.4063180 03  |                                   |                    |
| 39           | X = 0.3224580 04  | 0.3056660 04        | -0.1681180 03 | 0.883010 01                       | 0.6507390 03       |
|              | Y = 0.3224580 04  | 0.2926990 04        | -0.2976910 03 | 0.367960 01                       |                    |
|              | Z = 0.3822060 04  | 0.4308930 04        | 0.4868760 03  |                                   |                    |
| 40           | X = 0.3204520 04  | 0.3059470 04        | -0.1450480 03 | 0.876340 01                       | 0.1149160 04       |
|              | Y = 0.3204520 04  | 0.2331720 04        | -0.8727990 03 | 0.494630 01                       |                    |
|              | Z = 0.3798250 04  | 0.4351560 04        | 0.7333150 03  |                                   |                    |
| 41           | X = 0.3184270 04  | 0.2952780 04        | -0.2414850 03 | 0.869610 01                       | 0.1343100 04       |
|              | Y = 0.3184270 04  | 0.2125590 04        | -0.7034980 04 | 0.621540 01                       |                    |
|              | Z = 0.3774220 04  | 0.4399870 04        | 0.8244780 03  |                                   |                    |
| 42           | X = 0.3183630 04  | 0.2837710 04        | -0.3261210 03 | 0.862820 01                       | 0.1355600 04       |
|              | Y = 0.3183630 04  | 0.2223370 04        | -0.7620440 04 | 0.748700 01                       |                    |
|              | Z = 0.3743970 04  | 0.4353400 04        | 0.8054320 03  |                                   |                    |
| 43           | X = 0.3143210 04  | 0.2787870 04        | -0.3553350 03 | 0.855970 01                       | 0.1241530 04       |
|              | Y = 0.3143210 04  | 0.2119970 04        | -0.3232320 03 | 0.876120 01                       |                    |
|              | Z = 0.3725490 04  | 0.4475670 04        | 0.7501790 03  |                                   |                    |
| 44           | X = 0.3122400 04  | 0.2708350 04        | -0.4140490 03 | 0.849060 01                       | 0.1170760 04       |
|              | Y = 0.3122400 04  | 0.244660 04         | -0.8777060 03 | 0.100380 02                       |                    |
|              | Z = 0.3700800 04  | 0.4555680 04        | 0.6548860 03  |                                   |                    |

\*\*\*\*\* PERFORMANCE TABLE OUTOUT CONTINUES FOR REMAINING TRACK POINTS \*\*\*\*\*

AVERAGE SHELL AT TARGET MISS DISTANCE = 0.18277740 03 YARDS

PERFORMANCE FACTOR OF FILTER, AVERAGE FILTER POSITION ESTIMATE ERROR = 0.265126090 01 YARDS

AVERAGE FILTER VELOCITY ESTIMATE ERROR = 0.228808250 02 YARDS/SEC

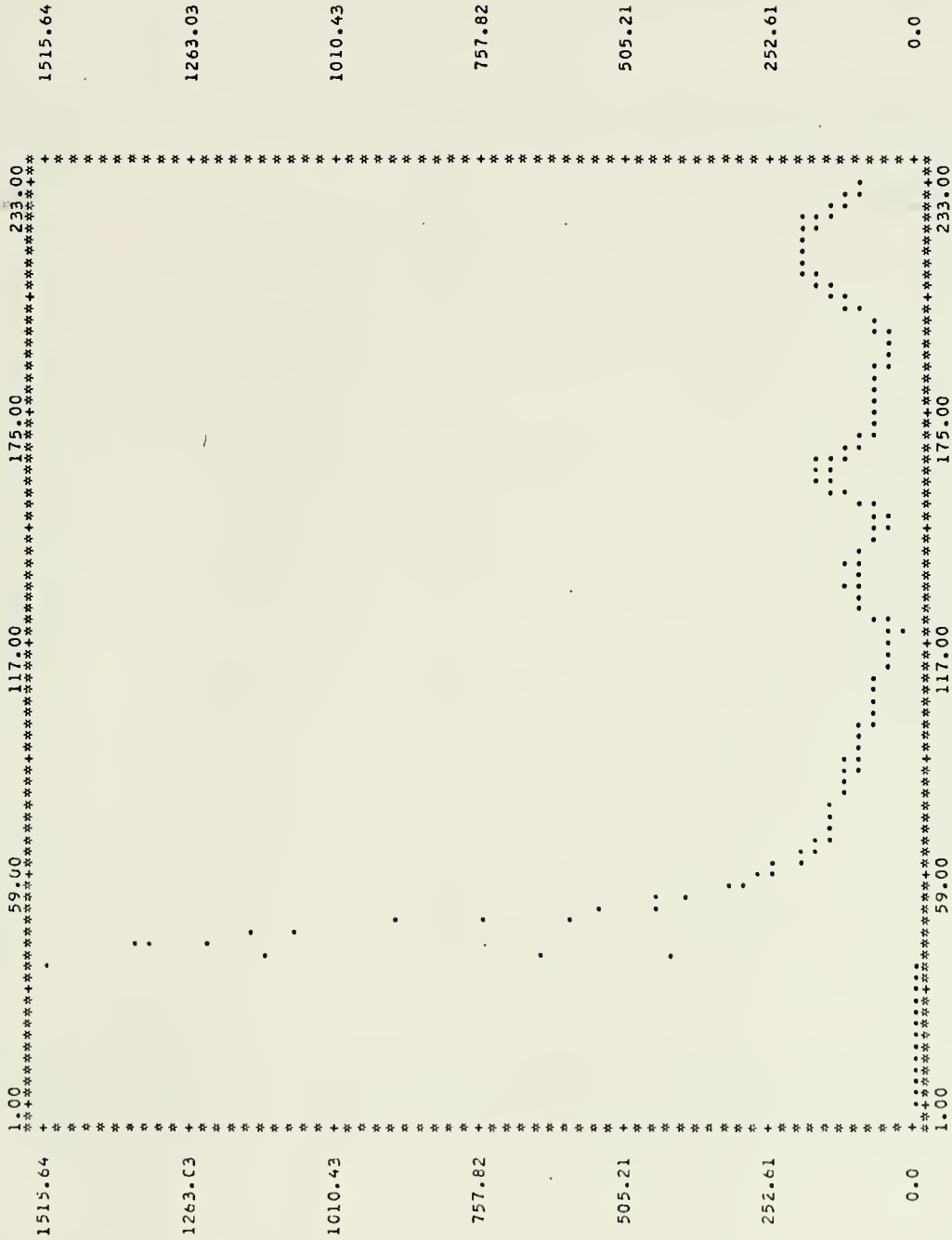


| POINT NO | POSITION ESTIMATE ERRORS                                | POSITION VARIANCES                                    | VELOCITY / ERRORS                               | VELOCITY VARIANCES AROUND ESTIMATE / AROUND TRUTH      |
|----------|---|---|---|--|
| 1        | X: -0.5869500 00<br>Y: 0.8877200 00<br>Z: 0.2667850 01  | X: 0.1264470 03<br>Y: 0.1589200 03<br>Z: 0.1409170 03 | 0.9251410 01<br>-0.1307940 03<br>0.4167560 02   | X: 0.1000710 03<br>Y: 0.5753950 03<br>Z: 0.4003770 02  |
| 2        | X: 0.7545570 00<br>Y: 0.1257540 01<br>Z: 0.8300160 00   | X: 0.1515110 03<br>Y: 0.1315970 03<br>Z: 0.1596330 03 | 0.5350410 01<br>0.1493670 01<br>-0.5833430 01   | X: 0.4147960 04<br>Y: 0.4641090 04<br>Z: 0.4518960 04  |
| 3        | X: 0.7057070 00<br>Y: 0.1470990 01<br>Z: -0.3008100 01  | X: 0.5628680 03<br>Y: 0.5842220 03<br>Z: 0.5516330 03 | -0.8456680 01<br>-0.3404860 02<br>0.3861960 01  | X: 0.3755090 04<br>Y: 0.8318880 04<br>Z: 0.5082250 04  |
| 4        | X: 0.1747570 01<br>Y: -0.9144870 01<br>Z: 0.1500260 00  | X: 0.5450270 03<br>Y: 0.2939770 03<br>Z: 0.7351480 03 | -0.5625930 01<br>-0.8527660 02<br>0.3149020 02  | X: 0.1923290 04<br>Y: 0.9721730 04<br>Z: 0.2291400 04  |
| 5        | X: 0.6144890 01<br>Y: -0.1184530 02<br>Z: 0.6170080 01  | X: 0.3272940 03<br>Y: 0.1837750 03<br>Z: 0.5795850 03 | -0.8143510 01<br>-0.1035920 03<br>0.4550990 02  | X: 0.1897520 04<br>Y: 0.1131200 05<br>Z: 0.2083000 04  |
| 6        | X: 0.7955340 01<br>Y: -0.1154870 02<br>Z: 0.9196870 01  | X: 0.1883330 03<br>Y: 0.2489670 03<br>Z: 0.5278870 03 | -0.1626740 02<br>-0.1058320 03<br>0.4551000 02  | X: 0.2371090 04<br>Y: 0.1567610 05<br>Z: 0.22319410 04 |
| 7        | X: 0.7645690 01<br>Y: -0.8484210 01<br>Z: 0.97714840 01 | X: 0.1438700 03<br>Y: 0.1250970 03<br>Z: 0.5943640 03 | -0.2684820 02<br>-0.1066440 03<br>0.4398590 02  | X: 0.1768330 04<br>Y: 0.1290470 05<br>Z: 0.1874850 04  |
| 8        | X: 0.6205060 01<br>Y: -0.4205540 01<br>Z: 0.8757070 01  | X: 0.1343870 03<br>Y: 0.1444630 03<br>Z: 0.5234990 03 | -0.2990000 02<br>-0.9361990 02<br>0.38555810 02 | X: 0.2034090 04<br>Y: 0.1340990 05<br>Z: 0.1934440 04  |
| 9        | X: 0.3108800 01<br>Y: -0.124540 01<br>Z: 0.7188760 01   | X: 0.1365750 03<br>Y: 0.1237080 03<br>Z: 0.1886110 03 | -0.3800480 02<br>-0.9702130 02<br>0.38995310 02 | X: 0.2036340 04<br>Y: 0.1147520 05<br>Z: 0.1995420 04  |
| 10       | X: 0.1329180 01<br>Y: -0.1445250 01<br>Z: 0.4922810 01  | X: 0.9378140 02<br>Y: 0.1117670 03<br>Z: 0.1326700 03 | -0.3746770 02<br>-0.9168930 02<br>0.32809510 02 | X: 0.1835920 04<br>Y: 0.9681160 04<br>Z: 0.1624800 04  |

\*\*\*\*\* FILTER ERROR TABLE CONTINUES FOR REMAINING TRACK POINTS \*\*\*\*\*



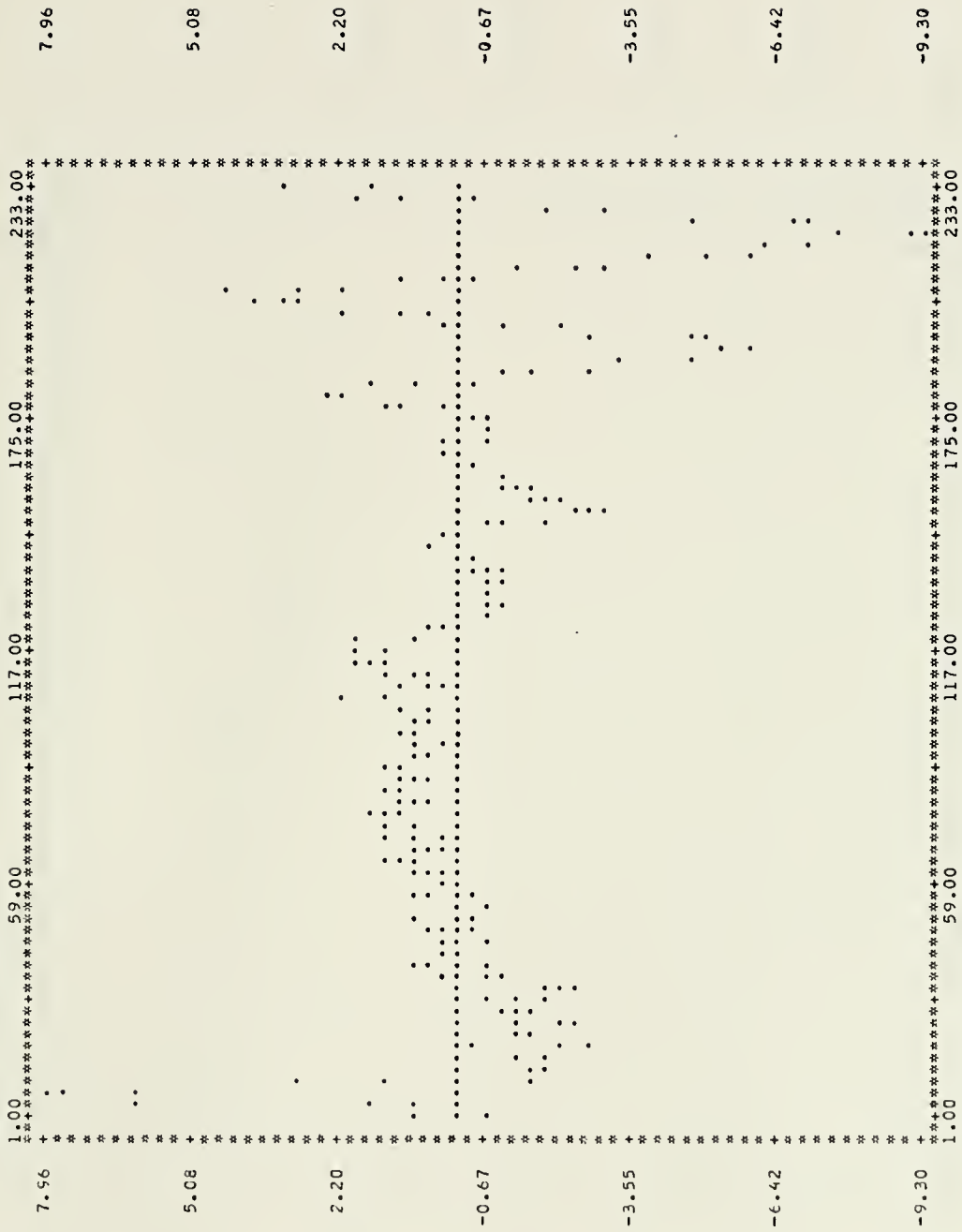
KETRON MK-86 THESIS  
 MEAN MISS DISTANCE WHEN SHELL ARRIVES IN YDS.



X-SCALE: "\*" = 0.290E 01 UNITS  
 Y-SCALE: "\*" = 0.253E 02 UNITS



NETRON MK-86 THESIS  
MEAN ERROR IN X POSITION



X-SCALE: "X"= 0.290E 01 UNITS  
Y-SCALE: "Y"= 0.288E 00 UNITS

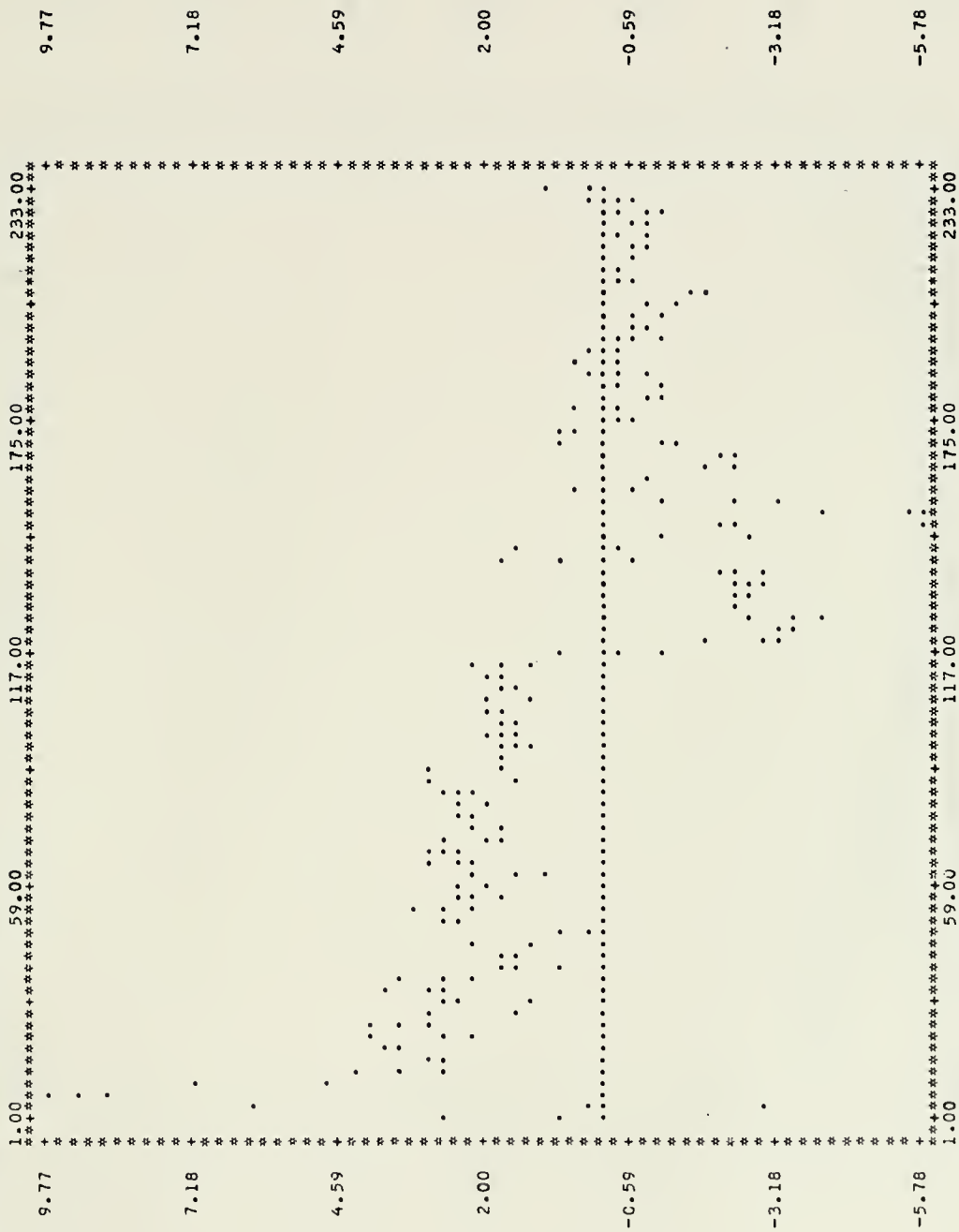








KETRON MK-86 THESIS  
 MEAN ERROR IN Z POSITION



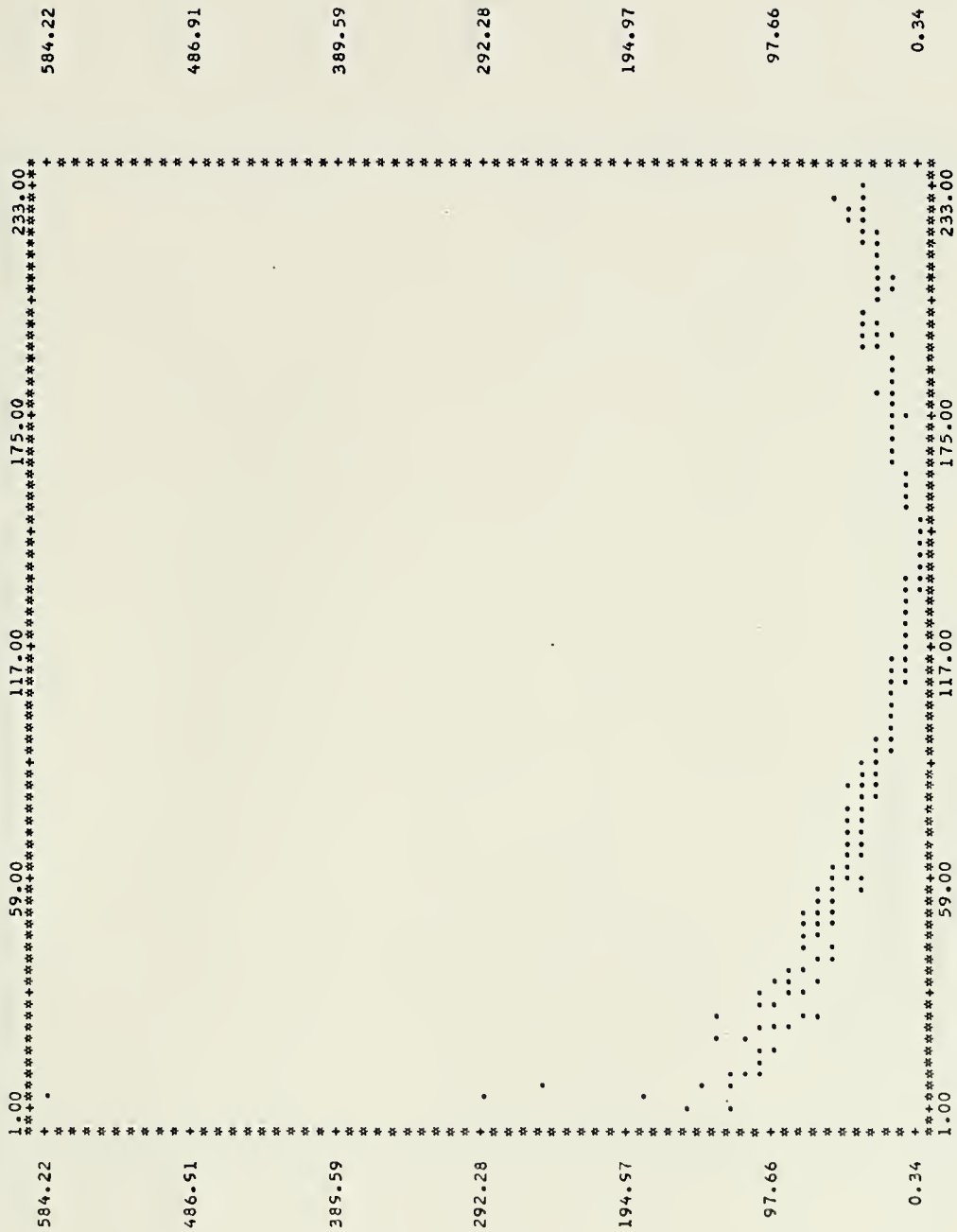
X-SCALE: "\*" = 0.290E 01 UNITS  
 Y-SCALE: "\*" = 0.259E 00 UNITS







KETRON MK-86 THESIS  
Y POSITION VARIANCE

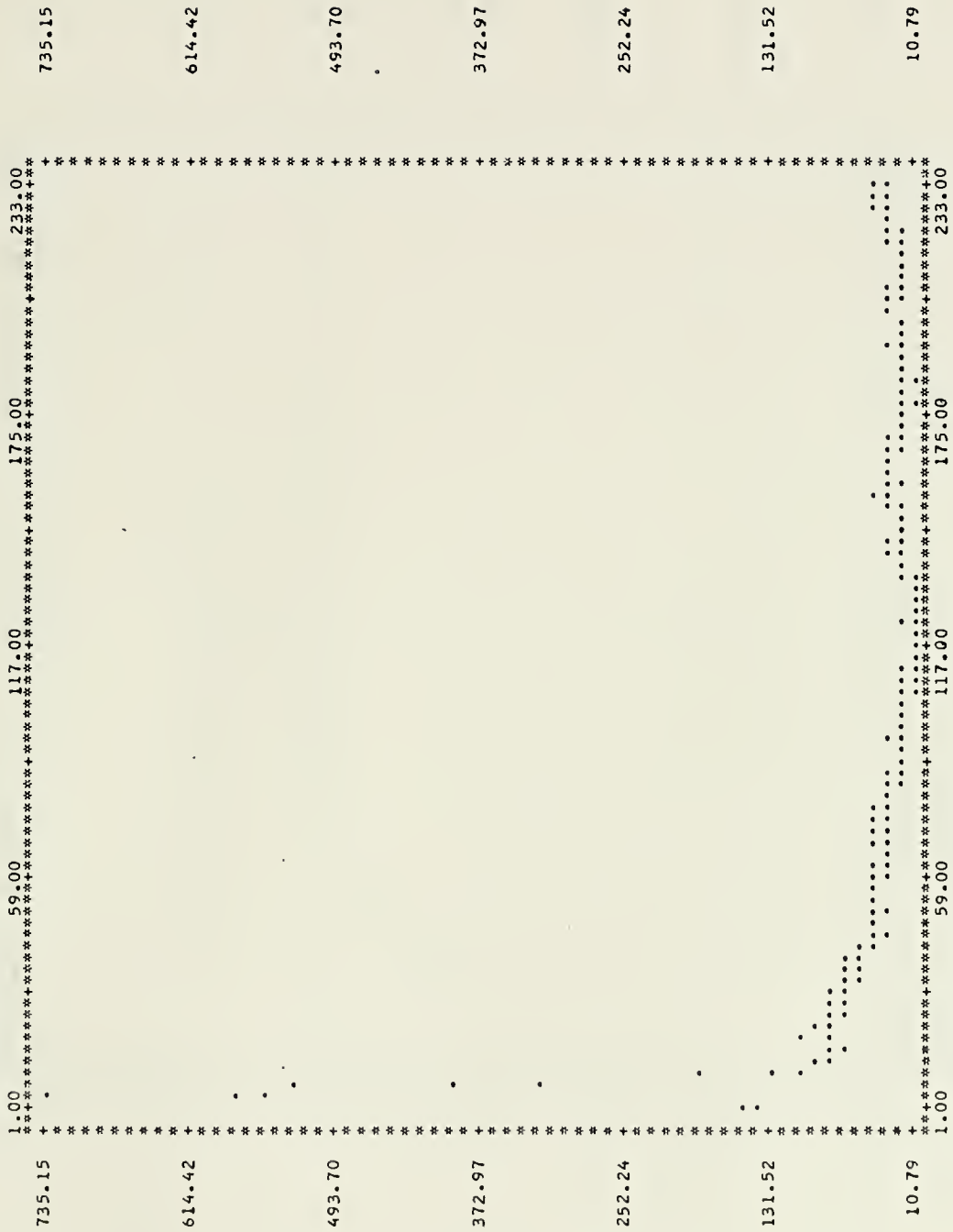


X-SCALE: " " = 0.290E 01 UNITS  
Y-SCALE: " " = 0.973E 01 UNITS





KETRON MK-86 THESIS  
Z POSITION VARIANCE



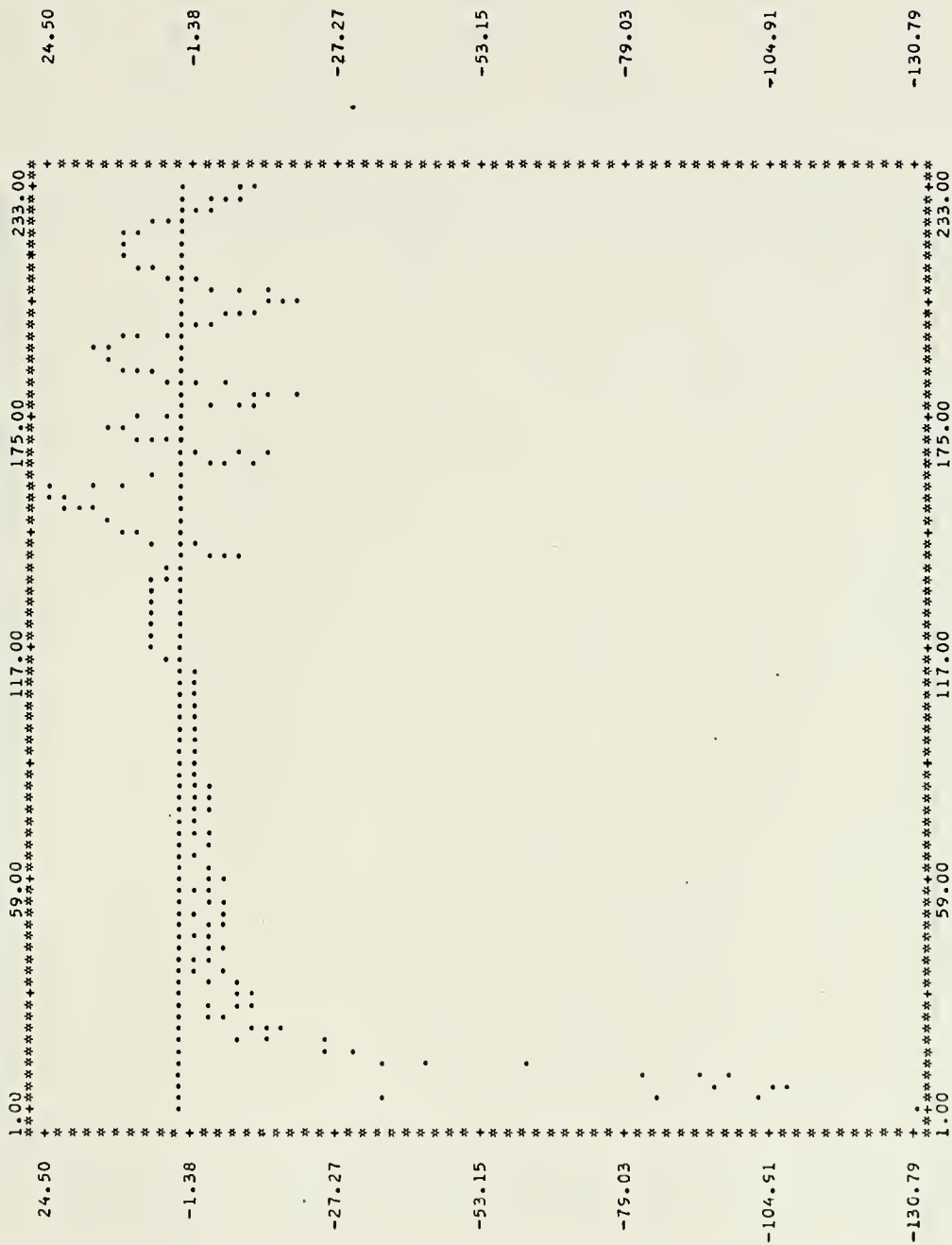
X-SCALE: "X"= 0.290E 01 UNITS  
Y-SCALE: "Y"= 0.121E 02 UNITS







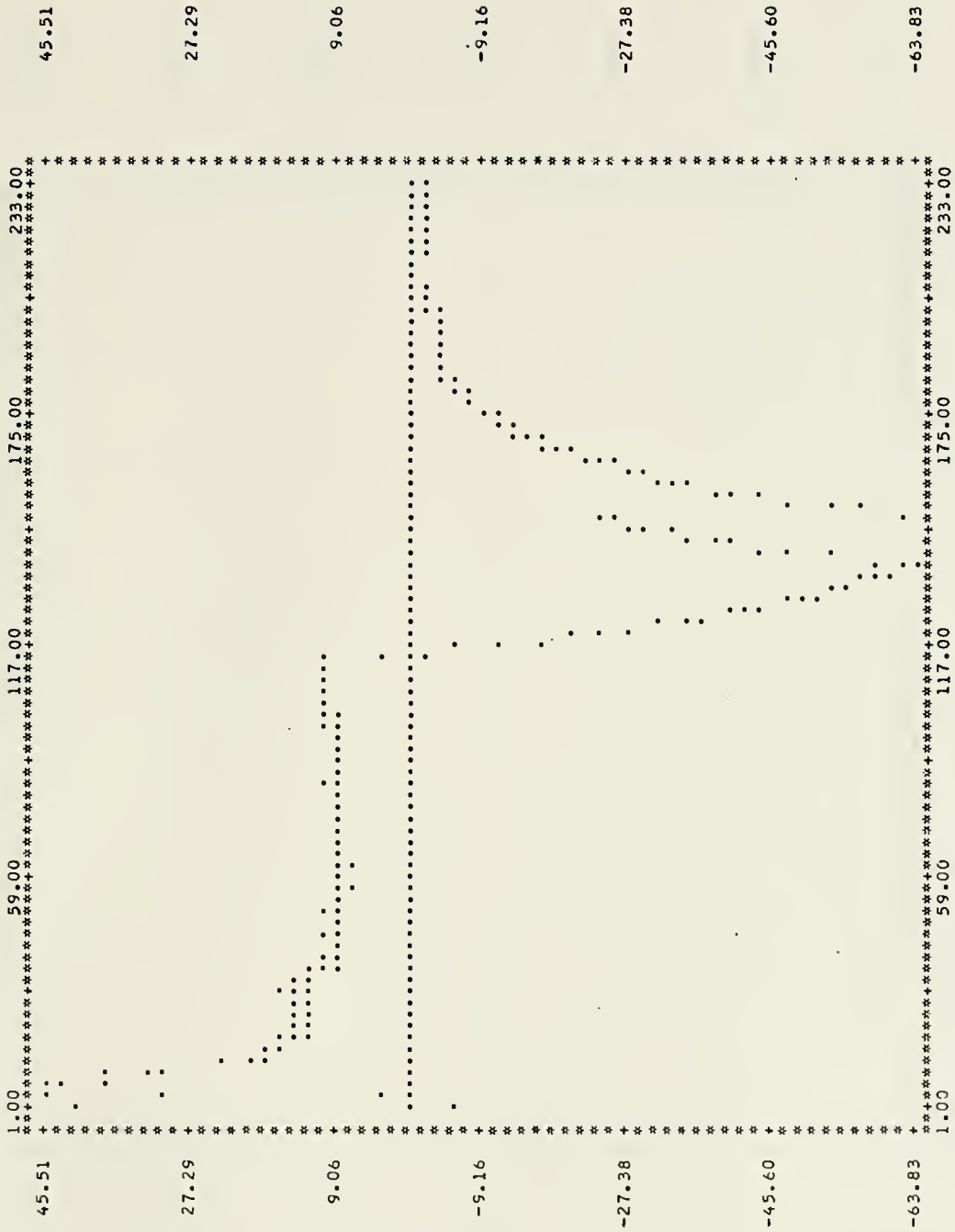
KETRON MK-86 THESIS  
MEAN VELOCITY ERROR IN Y DIRECTION



X-SCALE: "##"= 0.290E 01 UNITS  
Y-SCALE: "##"= 0.259E 01 UNITS



KETRON MK-86 THESIS  
MEAN VELOCITY ERROR IN Z DIRECTION



X-SCALE: " " = 0.290E 01 UNITS  
Y-SCALE: " " = 0.182E 01 UNITS





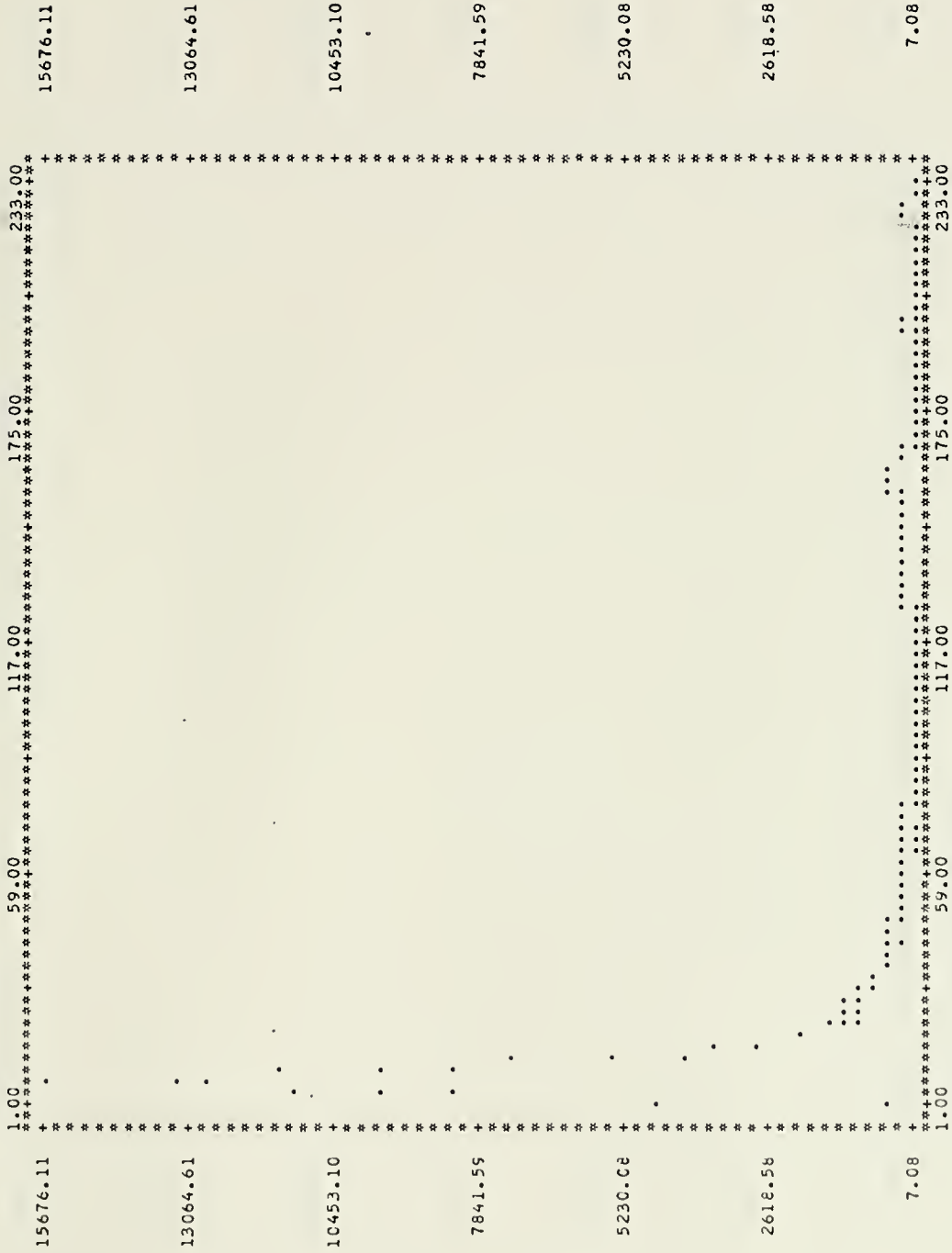
KETON MK-86 THESIS  
X VELOCITY VARIANCE



X-SCALE: "\*"= 0.290E 01 UNITS  
Y-SCALE: "\*"= 0.691E 02 UNITS



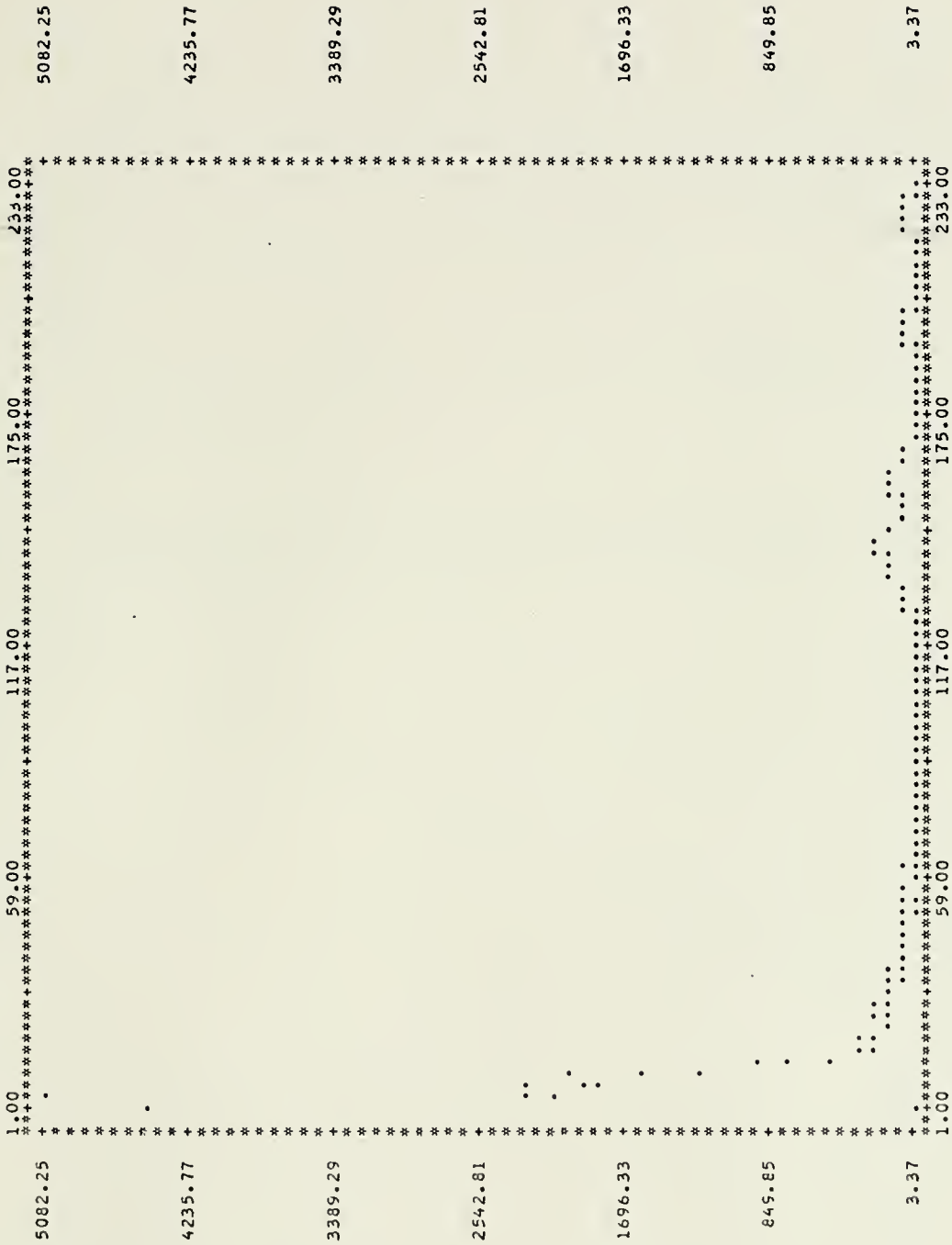
KETRON MK-86 THESIS  
Y VELOCITY VARIANCE



X-SCALE: "\*" = 0.290E 01 UNITS  
Y-SCALE: "\*" = 0.261E 03 UNITS



KETRON MK-86 THESIS  
Z VELOCITY VARIANCE

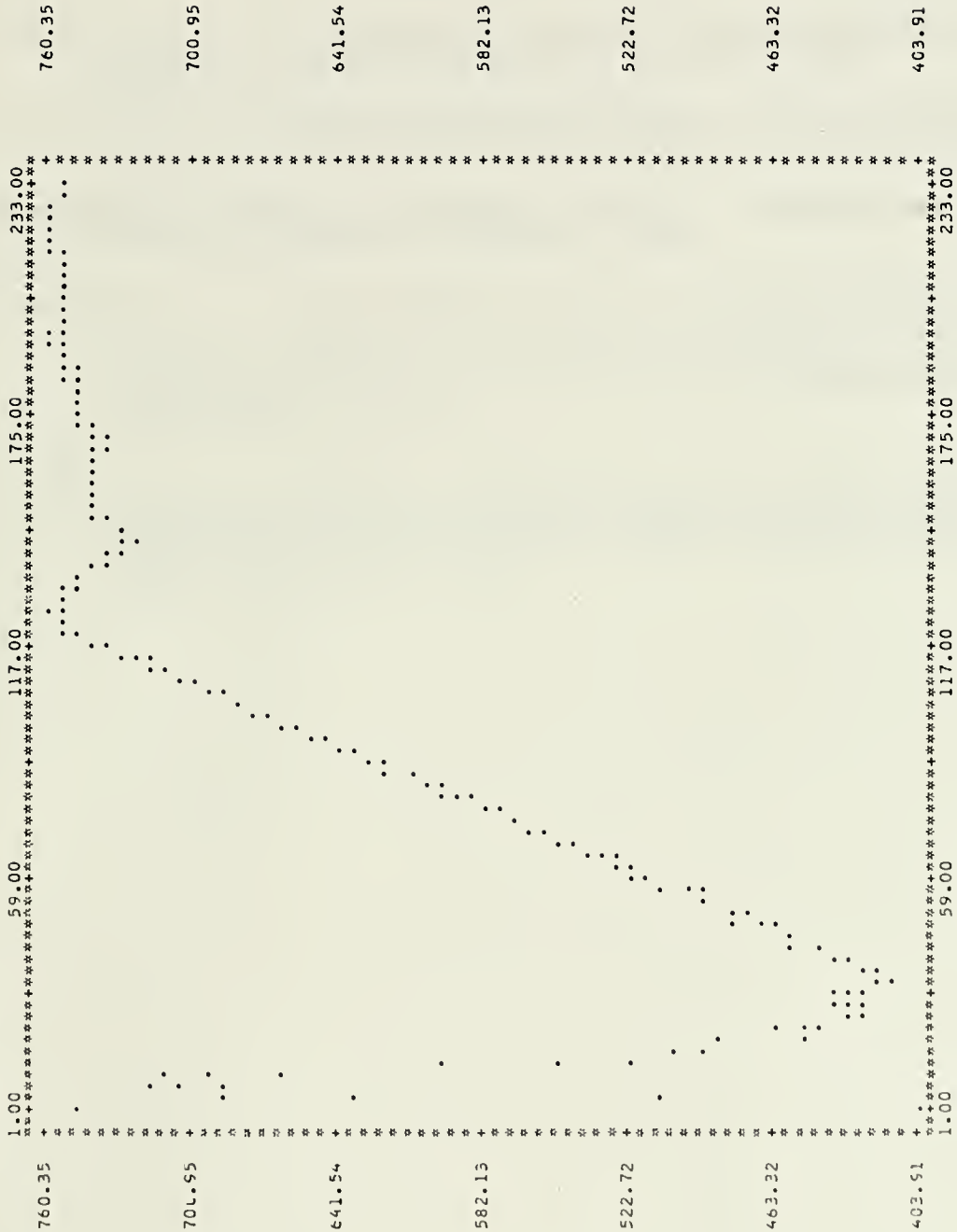


X-SCALE: "\*" = 0.290E 01 UNITS

Y-SCALE: "\*" = 0.646E 02 UNITS



KRTFON MK-86 THESES  
 TARGET SPEED IN FT/SEC



X-SCALE: "\*" = 0.290E 01 UNITS  
 Y-SCALE: "\*" = 0.554E 01 UNITS





## BIBLIOGRAPHY

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5. Will, T. J., A FORTRAN Simulation of the Target Data Filtering Section of the AA Tracking Loop in the MK-86 GFCS, Master's Thesis, Naval Postgraduate School, Monterey, 1973.



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94831 S  
7 JUN 76  
7 JUN 76

23472  
23646

152415  
Thesis  
K3953 Ketron  
c.1 Monte-Carlo evaluation  
of digital filters for  
fire control systems.

94831 S  
7 JUN 76  
7 JUN 76

23472  
23646

152415  
Thesis  
K3953 Ketron  
c.1 Monte-Carlo evaluation  
of digital filters for  
fire control systems.

thesK3953

Monte-Carlo evaluation of digital filter



3 2768 001 03222 0

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