NAVAL POSTGRADUATE SCHOOL Monterey, California



THESIS

KALMAN FILTERING TECHNIQUES
APPLIED TO
AIRBORNE DIRECTION-FINDING
AND EMITTER LOCATION

by

L. Laddie Coburn

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H. A. Titus

Thesis Advisor:
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Kalman Filtering Techniques
Applied to
Airborne Direction-Finding
and Emitter Location

by

L. Laddie Coburn Lieutenant, United States Navy B.S., United States Naval Academy, 1965

Submitted in partial fulfillment of the requirements for the degree of

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I. INTRODUCTION

One goal of an boome direction-finding (DF) is to obtain the location of an electromagnetic radiation source in latitude/longitude coordinates on the earth's surface. Over the years many emitter location techniques have been devised, and much research has been devoted to optimizing bearing angle measurements in angle-only direction-finding and pavigation systems. Much of the effort has been directed toward physical components such as the directional receiving antenna and the improvement of its characteristics. In airborne DF systems current typical bearing angle-of-arrival (AOA) accuracy is found to be about $\pm 2^{\circ}$ at microwave frequencies. Despite such accuracy a multiple bearing fix will yield considerable emitter position ambiguity at any significant range between the emitter and the airborne DF system.

To improve this situation a statistical optimal estimate of each received emitter bearing angle was computed using sequential estimation techniques, commonly called Kalman filtering. Additionally, an adaptive gating technique was utilized which selectively filtered out bearing lines which did not closely associate with other bearing lines already filtered and correlated to a distinct emitter location. Smoothing techniques were utilized to improve the unfiltered initial bearing angle, and a smoothed initial bearing angle and filtered final bearing

angle were used in a spherical earth triangulation solution to estimate the emitter position. The entire sorting, filtering, smoothing, gating, and triangulation procedure was organized within a software computer program capable of automatically estimating multiple emitter locations given noisy emitter bearing angle data and aircraft navigation data.

II. ANALYSIS OF THE PROBLEM

A. APPLICATION

The objective of this analytical study was to develop an optimal emitter location algorithm which would have application to military aircraft flights in an operational environment. The computer program developed could be utilized in an airborne digital computer system to give real-time analysis of emitter parameters in a multiple emitter environment, or it could be used in a ground-based computer system to give post-flight analysis of large quantities of compiled data.

Since all data sampling and processing was done at discrete time intervals, discrete sequential estimation techniques were utilized. The real facility in using sequential estimation techniques is that each subsequent computation is based only on the new observation and the last previously calculated estimate, and these recursive equation computations require a minimum of computer storage to filter a large quantity of data. This makes such an emitter location algorithm ideal for airborne computer application [1].

To allow generality of application, the software system was developed to process multiple emitter locations, to sort and filter large quantities of emitter data, to filter data at non-uniform sampling intervals, and to estimate emitter locations on any region of the earth's surface. It was initially assumed that no errors existed in the aircraft

navigation data; hence these position data together with two or more filtered emitter bearing angles could be used to compute the emitter location (see Figure 1). If a deterministic aircraft navigation error was known to exist, it could be compensated for in the emitter location algorithm. If a random error existed in the aircraft navigation data, it could be minimized by using a non-zero Q matrix to account for random excitation noise (see Section III, Part B, page 17) or by Kalman filtering the aircraft navigation data as well as the emitter bearing data. Kalman filtering techniques have also been developed to optimize navigation system accuracy [2].

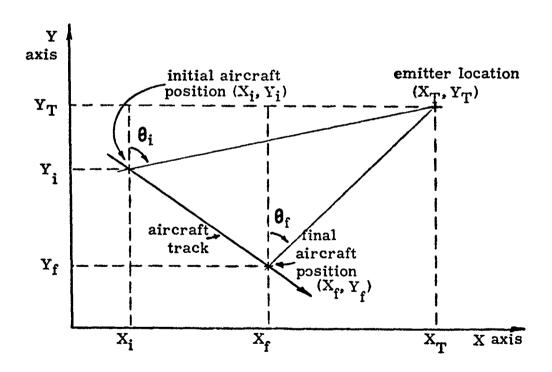


FIGURE 1
GEOMETRY OF EMITTER LOCATION

B. KALMAN FILTER UTILIZATION

A signal X, measured in the presence of random white Gaussian noise V, is filtered in an attempt to recover the original signal. The filtered estimate of the state is denoted by \widehat{X} . From Kalman filter theory [3] for discrete systems using first-order difference equations, the kth noisy observation Z(k) of the state is given by

where $\underline{H}(k)$ is the observation matrix and $\underline{V}(k)$ is the measurement noise. Since a discrete or sampled system is being considered, k denotes the kth time sample.

The state of a general, discrete system with no deterministic forcing function is given by

$$X(k+1) = \Phi(k+1,k)X(k) + \Gamma W(k)$$
 $k = 1, 2, ...$ (2)

where $\Phi(k, 1, k)$ is the state transition matrix, Γ is a distribution matrix related to the random forcing function, and $\Psi(k)$ is a random forcing function to account for random excitation noise. It was assumed that the noise sequences have zero mean and second-order statistics described by

$$\mathbf{E}\left[\underline{\mathbf{V}}(\mathbf{k})\underline{\mathbf{V}}(\mathbf{j})^{\mathrm{T}}\right] = \underline{\mathbf{R}}(\mathbf{k})\delta(\mathbf{k}\mathbf{j}) \tag{3}$$

$$\underline{\Gamma} \mathbf{E} \left[\underline{\mathbf{W}}(\mathbf{k}) \underline{\mathbf{W}}(\mathbf{j})^{\mathrm{T}} \right] \underline{\Gamma}^{\mathrm{T}} = \underline{\mathbf{Q}}(\mathbf{k}) \mathbf{S}(\mathbf{k}\mathbf{j})$$
(4)

and
$$\mathbf{E}\left[\underline{\mathbf{V}}(\mathbf{k})\underline{\mathbf{W}}(\mathbf{j})^{\mathbf{T}}\right] = \underline{\mathbf{0}}$$
 for all \mathbf{k}, \mathbf{j} . (5)

where
$$8(kj) = \begin{cases} 0 & k \neq j \\ & \text{is the Kronecker delta.} \end{cases}$$

The Kalman filter recursion equations [1] are summarized below where $\hat{X}(k/j)$ denotes the estimate of the state X(k) based upon j measurement observations $Z(1), Z(2), \ldots, Z(j)$.

$$\underline{\mathbf{G}}(\mathbf{k}) = \underline{\mathbf{P}}(\mathbf{k}/\mathbf{k}-1)\underline{\mathbf{H}}(\mathbf{k})^{\mathrm{T}} \left[\underline{\mathbf{H}}(\mathbf{k})\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k}-1)\underline{\mathbf{H}}(\mathbf{k})^{\mathrm{T}} + \underline{\mathbf{R}}(\mathbf{k})\right]^{-1}$$
(6)

$$\mathbf{P}(\mathbf{k}/\mathbf{k}) = \mathbf{P}(\mathbf{k}/\mathbf{k}-1) - \mathbf{G}(\mathbf{k})\mathbf{H}(\mathbf{k})\mathbf{P}(\mathbf{k}/\mathbf{k}-1)$$
 (7)

$$\underline{\mathbf{P}}(\mathbf{k}+1/\mathbf{k}) = \underline{\Phi}(\mathbf{k}+1,\mathbf{k})\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k})\underline{\Phi}(\mathbf{k}+1,\mathbf{k})^{\mathrm{T}} + \underline{\mathbf{Q}}(\mathbf{k})$$
(8)

$$\widehat{\underline{X}}(k/k) = \widehat{\underline{X}}(k/k-1) + \underline{G}(k) \left[\underline{Z}(k) - \underline{H}(k) \widehat{\underline{X}}(k/k-1) \right]$$
 (9)

$$\widehat{\underline{X}}(k/k-1) = \underline{\Phi}(k, k-1)\widehat{\underline{X}}(k-1/k-1)$$
(10)

The Kalman filter gains are represented by the matrix $\underline{G}(k)$, and $\underline{P}(k/j)$ represents the error covariance matrix of the various estimates.

For the problem being considered the states are bearing angle-of-arrival and bearing rate, and it is desired to filter the noisy observed bearing angles to obtain optimal estimates of these states. The system dynamics were approximated by a $1/s^2$ plant which results in a state transition matrix given by

$$\underline{\Phi}(k+1,k) = \begin{bmatrix} 1 & T(k+1) \\ 0 & 1 \end{bmatrix}.$$

Since only the bearing angle of the state was observed, R(k) and W(k) become scalar variance terms and the bracketed terms in equations (6) and (9) become scalars as well. This considerably simplified reducing these matrix equations to scalar form (see Appendix A); however, allowing a non-uniform sampling interval required that the

term T(k+1) remain a variable in each of the applicable recursion equations. This required that error covariance terms and Kalman filter gains be computed on line since they were a function of the sampling interval; therefore, these terms could not be precomputed and stored.

Several assumptions are implicit here. Modeling the system dynamics with a 1/s² transfer function gives a linear approximation that may not adequately represent the actual system equations, which are non-linear [4]. It may then become necessary to augment the dynamic model with an additional difference equation which includes the second derivative of the measured bearing angle, or extended Kalman filtering techniques may be employed to more accurately approach the non-linearity of the dynamic system. Extended Kalman filtering techniques are discussed in Section III, Part F, of this paper.

that the statistical nature of Kalman filtering techniques requires that the statistics of the measurement noise $\underline{V}(k)$ and the excitation noise $\underline{W}(k)$ be completely specified, or if unknown they must be assumed. If no deterministic errors are known to be coloring these noise sources, the assumption of white noise may be a good one although it may not exactly model the actual noise bearing error. Kalman filtering assumes that the noise is independent from one sampling interval to the next [5]. This may not be an accurate assumption since successive bearing angle observations from a single emitter might have similar error statistics. Any deterministic bias in aircraft antenna bearing angle-of-arrival measurement would further negate this assumption. This difficulty was reported by Reeves [6], who recommended that emitter bearing angle-of-arrival measurement noise be

considered to have both random and correlated components. This would require augmenting the state vector equation with an additional variable and further complicate the Kalman filtering process.

C. SMOOTHING TECHNIQUES

Since Kalman filtering techniques sequentially filter the second and subsequent bearing angle observations associated with a single emitter, the first bearing angle associated with a particular emitter remains unfiltered and no optimal estimate exists. Thus a technique of filtering backward in time or smoothing was used to improve the estimate of the initial bearing angle. The smoothing equations are similar to Kalman filter equations in that they are recursive and functions of similar statistical parameters, but they update or smooth previous data based on more recent data that has been optimally estimated.

A general technique proposed by Rauch [7] gives the smoothed estimate of the initial observation after k time intervals as

$$\widehat{\underline{X}}(1/k) = \underline{\widehat{X}}(1/k-1) + \underline{D}(1/k)\underline{G}(k) \left[\underline{Z}(k) - \underline{H}(k)\underline{\widehat{X}}(k/k-1)\right]$$
(11)

where
$$\underline{D}(1/k) = \underline{D}(1/k-1)\underline{P}(k-1/k-1)\underline{\Phi}(k,k-1)^{T}\underline{P}(k/k-1)^{-1}$$
 (12)

For a non-uniform sampling interval $\underline{D}(1/k)$ becomes a function of T(k) and must be solved sequentially before the smoothed initial estimate $\widehat{\underline{X}}(1/k)$ can be computed. For an analysis of these equations reduced to scalar form, see Appendix B. These scalar smoothing equations were included in the software computer program and yielded a smoothed estimate of the initial bearing angle and bearing rate.

III. COMPUTATIONAL PROCEDURE

A. INITIAL DATA SORT

environment with multiple emitters at various RF frequencies, emitter bearing angle-of-arrival errors, and a constant velocity moving aircraft. Aircraft navigation and emitter target data was sampled and recorded at discrete but time-varying intervals. Since the number of emitters was unknown and large quantitites of signal data were recorded at various frequencies, pulse repetition frequencies (PRF), and pulse widths (PW), it was necessary that the data be initially sorted by frequency and PRF to initially estimate the number of distinct emitter targets. This was accomplished with sufficiently small frequency and PRF gates to separate and associate emitter parameters with a distinct emitter target. Additionally, it was found that multiples of PRF may exist which associate with a single emitter, so a check for PRF multiples had to be included within the PRF sorting subroutine. One set of data processed by this program included PRF multiples given on page 59.

Multiple emitters of the same frequency and PRF could still result in an ambiguous solution; therefore, each bearing angle initially associated with a distinct emitter by the initial data sort was sequentially filtered and gated to determine if it correlated with this emitter.

Due to the large quantity of data to be processed, any bearing angle not

correlated to a distinct emitter after being filtered and gated was discarded from the data bank. Further investigation and greater computer storage capability could allow such discarded bearing angles to be reassociated with a different emitter; however, this process was not developed within this analysis.

B. KALMAN FILTER INITIALIZATION

Initialization of the Kalman filter equations requires a knowledge of the system dynamics, the statistical properties of the Kalman filter parameters, and some information about the initial state of the system. Since it was necessary to associate each sorted data sample (emitter bearing angle and associated emitter data) with a particular emitter, the notation JSET (I, J) was devised, where I is the emitter target number with which the data sample is associated and J is the jth sample of sequential data associated with that emitter. JSET(I, J) gives the sequential sample number for this data sample when listed together with all other data. See page 63 for a listing of initial JSET data. To initialize the Kalman filter equations, KI was used as the initial JSET value; i.e., JSET(I, 1) = KI for the ith emitter target.

To filter emitter bearing angles-of-arrival, the noisy observation of the state from equation (1) is given by

$$\mathbf{\theta}(\mathbf{k}) = \mathbf{H}(\mathbf{k}) \mathbf{\Theta}(\mathbf{k}) + \mathbf{V}(\mathbf{k}) \tag{13}$$

where
$$\underline{\boldsymbol{\theta}}(k) = \begin{bmatrix} \boldsymbol{\theta}(k) \\ \dot{\boldsymbol{\theta}}(k) \end{bmatrix}$$
 = vector of noisy measured observations,
$$\underline{\boldsymbol{\theta}}(k) = \begin{bmatrix} \boldsymbol{\theta}(k) \\ \dot{\boldsymbol{\theta}}(k) \end{bmatrix}$$
 = state vector of exact emitter bearing angle

and bearing rate, and $\underline{H}(k) = \begin{bmatrix} 1 & 0 \end{bmatrix}$ is the observation matrix for measurement of only noisy bearing angle θ . Since the bearing rate $\dot{\theta}(k)$ was not measured, it must be estimated to initialize the sequential estimation process. This bearing rate depends on the speed and heading of the aircraft with respect to the emitter, the relative bearing angle-of-arrival, and the range r to the emitter. Since the range to an unknown emitter is unknown, it must be estimated. For this program $\mathbf{r} = 150$ nautical miles was assumed and

$$\hat{\theta}(KI) = \frac{v}{r} \sin(BRNG) \times \frac{PIRAD}{3600}$$
 degrees/second (14)

where BRNG = relative bearing angle-of-arrival

PIRAD = 57.29578 degrees/radian

v = aircraft velocity in knots.

If an estimated value of range to the emitter is known and different from 150 nautical miles, then this new value should be used in equation (14) and substituted into the computer program.

Since an initial optimal estimate was not available to the filtering process described by equations (9) and (10), the first bearing angle optimal estimate THTD(KI) was assumed equal to the noisy measured bearing angle THETAD(KI), and the first bearing rate estimate TDTD(KI)

was assumed equivalent to $\hat{\boldsymbol{\theta}}(KI)$. The smoothing process was similarly initialized.

The initial uncertainty of emitter bearing angle and angle rate on filter initialization was accounted for in the initial values of the error covariance matrix P(1/0). Since stationary emitter locations were considered, the random forcing function W(k) of the dynamic state equation (2) could be set to zero if exact knowledge of the observer's aircraft position were known. However, since the aircraft position may have both random and deterministic errors associated with its position measurement and because of the simplified Φ matrix, a non-zero Q matrix was utilized. For W(k) having a zero mean and $E[W(k)^2] = 1.0$ degrees, from equation (4)

$$\underline{Q}(k) = \begin{bmatrix} \frac{T^{2}(k)}{2} \\ T(k) \end{bmatrix} \begin{bmatrix} 1.0 \end{bmatrix} \begin{bmatrix} \frac{T^{2}(k)}{2} & T(k) \end{bmatrix} = \begin{bmatrix} \frac{T^{4}(k)}{4} & \frac{T^{3}(k)}{2} \\ \frac{T^{3}(k)}{2} & T^{2}(k) \end{bmatrix}. \quad (15)$$

The value of R(k), the scalar variance of the measurement noise, was assumed to be constant and estimated according to the angle measurement accuracy of the DF system being considered. The initial \underline{D} matrix for the sequential smoothing process was initialized as the identity matrix (see Appendix B).

When utilizing this program for emitter location with different known measurement and covariance statistics, these initialization parameters must be changed in the computer program to ensure optimal solution locations. The only other term which must be modified by the

program user is the value of NUM, which is the total number of data samples to be sorted and filtered by this program. Since the data for each emitter must be filtered separately, the filter process occurs once for each distinct emitter target as determined by the initial data sort. Once an optimal estimate of emitter bearing angle and bearing rate have been initialized, the filter process is ready to commence.

C. KALMAN FILTER PROCESS

The Kalman filter equations utilized to filter the states, emitter bearing angle θ and bearing rate $\dot{\theta}$, are those noted on page 11 shown as equations (6) through (10) with \dot{X} replaced by $\dot{\theta}$. Since these are vector equations and tedious to handle within a computer program, their scalar counterparts were derived (see Appendix A) and utilized within the program. The optimal estimation equation is similar to equations (9) and (10) and is given by

$$\begin{bmatrix} \widehat{\boldsymbol{\theta}}(k/k) \\ \widehat{\boldsymbol{\theta}}(k/k) \end{bmatrix} = \begin{bmatrix} 1 & T(k) \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \widehat{\boldsymbol{\theta}}(k-1/k-1) \\ \widehat{\boldsymbol{\theta}}(k-1/k-1) \end{bmatrix} + \begin{bmatrix} G1(k) \\ G2(k) \end{bmatrix} E(k)$$
(16)

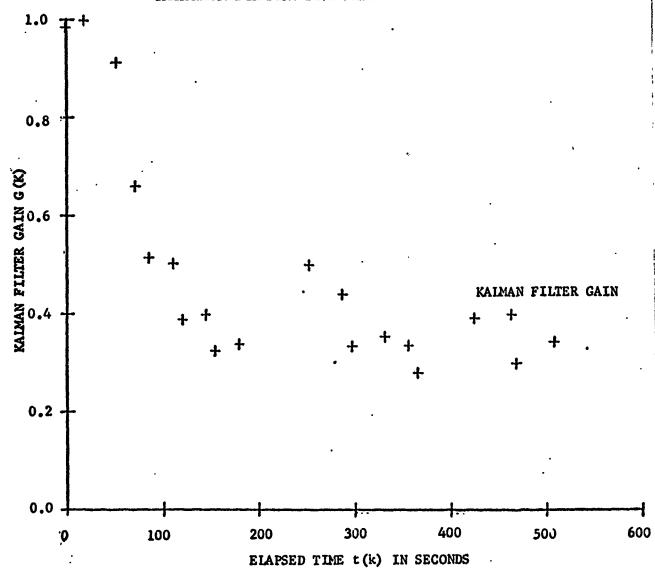
where
$$E(k) = \theta(k) - \hat{\theta}(k-1/k-1) - T(k)\hat{\theta}(k-1/k-1)$$
 (17)

results in a scalar term since only one of the state variables was measured.

Due to the non-uniform sampling interval, the Kalman filter gains G1(k) and G2(k) may vary and not approach a steady state value at a uniform rate as they would with a constant sampling interval. This is shown graphically in Figure 2, which depicts the transient Kalman

FIGURE 2

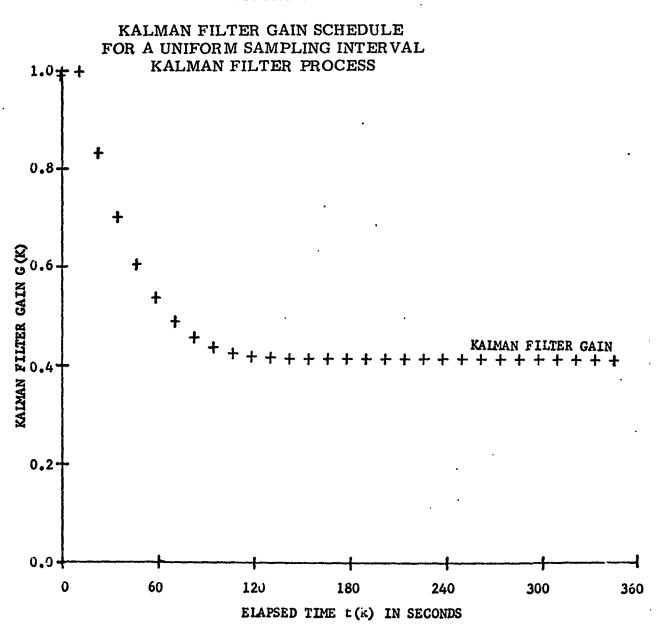
KALMAN FILTER GAIN SCHEDULE FOR A NON-UNIFORM SAMPLING INTERVAL KALMAN FILTER PROCESS



NOTES:

- 1. Points plotted above represent discrete values of Kalman filter gain G1(K) associated with filtering the bearing angle of the state vector.
- 2. Since the prediction covariance matrix (P(k+1/k)) given by equation (8) is a function of $\Phi(k+1,k)$, which contains the sampling interval T(k+1), it can be seen that the Kalman filter gain G(k) will also vary with T(k+1) as in equation (6). As the sampling interval increases, the gain will also increase.

FIGURE 3



NOTES:

- 1. Points plotted above represent discrete values of Kalman filter gain G1(K) associated with filtering the bearing angle of the state vector.
- 2. After ten sampling intervals the gain has nearly reached a steady state value of 0.417. This asymptote is determined by the non-zero value of the Q matrix.
- 3. If system random excitation noise (i.e., the \underline{Q} matrix) were assumed to be zero, then the Kalman filter gain would approach zero upon reaching steady state.

filter gain schedule of a non-uniform sampling interval. Figure 3 shows the Kalman filter gain schedule of a uniform, twelve-second sampling interval computer simulation, which was developed to allow an error analysis to be made on the Kalman filter program.

The adaptive gate was chosen to be a function of the prediction variance term P11(k/k-1), which is a variable function, and the measurement noise variance R(k), which is constant. The adaptive gate was set to selectively compare the filtered bearing angle with previously correlated bearing angles. If a filtered bearing does not fall within the adaptive gate, it is discarded and the sampling interval increases to the last previous correlated bearing. This increases the Kalman filter gain and opens the adaptive gate to a wider allowable value until a subsequent bearing angle is found which correlates to this emitter, then causing the adaptive gate to reduce in size. The adaptive gate was chosen to be

$$GATE(K) = C \cdot \sqrt{P11(k) + R(K)}$$
 (18)

where C is a constant scale factor. Since the prediction covariance P(k/k-1) may be defined as

$$\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k}-1) = \mathbf{E}\left\{\left[\underline{\boldsymbol{\theta}}(\mathbf{k}) - \underline{\widehat{\boldsymbol{\theta}}}(\mathbf{k}/\mathbf{k}-1)\right]\left[\underline{\boldsymbol{\theta}}(\mathbf{k}) - \underline{\widehat{\boldsymbol{\theta}}}(\mathbf{k}/\mathbf{k}-1)\right]^{\mathrm{T}}\right\}, (19)$$

which is a function of the mean square error between the true value of the bearing angle and its optimal estimate, it can be seen that as the estimate error increases the adaptive gate will increase and as the optimal estimate approaches the true value the gate size will decrease as is desired. The constant term R(k) prevents the gate from getting so narrow that no emitter bearing angles could correlate to a distinct emitter, and the variable term P11(k) allows the gate size to "adapt" to the transient filter errors as was noted above.

D. INITIAL BEARING ANGLE SMOOTHING PROCESS

In order to obtain a smoothed estimate of the first bearing angle associated with each emitter target, it was necessary to solve equation (11) sequentially within the Kalman filter process since the gain term was recomputed after each sampling interval. The smoothed initial bearing angle estimation equation used in the software program is similar to equation (11) and is given by

$$\begin{bmatrix}
\widehat{\boldsymbol{\theta}}(1/k) \\
\widehat{\boldsymbol{\theta}}(1/k)
\end{bmatrix} = \begin{bmatrix}
\widehat{\boldsymbol{\theta}}(1/k-1) \\
\widehat{\boldsymbol{\theta}}(1/k-1)
\end{bmatrix} + \begin{bmatrix}
D11(1/k) D12(1/k) \\
D21(1/k) D22(1/k)
\end{bmatrix} \begin{bmatrix}
G1(k) \\
G2(k)
\end{bmatrix} E(k), (20)$$

where E(k) is defined by equation (17) and the scalar terms in the \underline{D} matrix are as derived in Appendix B. To compute these terms, both $\underline{\Phi}(k,k-1)^{-1}$ and $\underline{P}(k/k-1)^{-1}$ had to be determined. The inverse of the state transition matrix is obviously given by

$$\underline{\underline{\Phi}}(k, k-1)^{-1} = \begin{bmatrix} 1 & -T(k) \\ 0 & 1 \end{bmatrix} . \tag{21}$$

Should the error covariance matrix become singular because of numerical inaccuracy in computation, no inverse will exist. In this case corrective action must be taken before attempting to compute the inverse. Several references have discussed this problem, suggested

solutions, and Schmidt [8] gives a survey of current techniques in this area. For this program, the IBM 360 subroutine MINV was utilized to compute the error covariance matrix inverse with good numerical accuracy.

E. EMITTER LOCATION ALGORITHMS

Several techniques were investigated to obtain a computationally simple yet accurate algorithm to calculate the emitter locations. Since more than two bearing lines may result in an ambiguous emitter location, it was most logical to use only the smoothed initial bearing angle $\widehat{\theta}_i$ and the filtered final bearing angle $\widehat{\theta}_f$ to compute the location. Any such location procedure would give most accurate results if the two bearings intersected at right angles; however, infrequent data collection, short-leg flight tracks, or long-range distances from the emitter generally prevent such choice of initial and final bearing angles.

1. Plane Triangulation Solution

The basic "flat earth" triangulation solution was presented by Kayton and Fried [9] and discussed in some detail in Refs. 6 and 10. Figure 1 depicts the geometry of the triangulation method of emitter location solution, and the solution equations are given by

$$\tan \widehat{\theta}_i = \frac{X_T - X_i}{Y_T - Y_i}$$
 (22)

and
$$\tan \hat{\theta}_f = \frac{X_T - X_f}{Y_T - Y_f}$$
 (23)

where the emitter location coordinates (X_T, Y_T) are unknown and must be determined.

If latitude (L)/longitude (λ) coordinates are substituted for the X/Y axes in the flat earth model, the following equations result:

$$\tan \hat{\theta}_{i} = \frac{(\lambda_{T} - \lambda_{i}) \cos L_{T}}{L_{T} - L_{i}}$$
(24)

$$\tan \hat{\theta}_{f} = \frac{(\lambda_{T} - \lambda_{f}) \cos L_{T}}{L_{T} - L_{f}} . \tag{25}$$

These equations gave a more accurate solution but were more complex to program since one of the unknowns (L_T) occurs both as an explicit function and as a cosine function. This difficulty was resolved by solving for an intermediate value of emitter latitude (TILA) based on an aircraft midlatitude cosine function (WAV) which was known and then by recomputing emitter latitude (TLA) based on the cosine function of the intermediate emitter latitude. The resulting equations are given by

$$WAV = \frac{L_i + L_f}{2}$$
 (26)

$$TILA = \frac{(\lambda_{f} - \lambda_{i})\cos(WAV) + L_{i}\tan\hat{\theta}_{i} - L_{f}\tan\hat{\theta}_{f}}{\tan\hat{\theta}_{i} - \tan\hat{\theta}_{f}}$$
(27)

$$TLA = \frac{(\lambda_{f} - \lambda_{i})\cos(TILA) + L_{i}\tan\widehat{\theta}_{i} - L_{f}\tan\widehat{\theta}_{f}}{\tan\widehat{\theta}_{i} - \tan\widehat{\theta}_{f}}$$
(28)

TLO =
$$\lambda_f - (L_T - L_f) \tan \hat{\theta}_f / \cos(TILA)$$
 (29)

where TLA = emitter target latitude L_T

TLO = emitter target longitude λ_T

L; = initial aircraft position latitude

 λ_i = initial aircraft position longitude

L_f = final aircraft position latitude

 λ_f = final aircraft position longitude.

These equations are derived in Appendix C and are equivalent to the "round world" model developed on pages 41-43 of Ref. 6.

2. Spherical Triangulation Solution

Two approaches to a spherical earth solution were taken.

The first involved the derivation of a spherical triangulation solution with two equations resulting from two spherical triangles as shown in Fig. 4. These equations are derived in Appendix D and are given by

$$\tan \hat{\theta}_{i} = \frac{\tan \left[(\lambda_{T} - \lambda_{i}) \cos L_{T} \right]}{\sin(L_{T} - L_{i})}$$
(30)

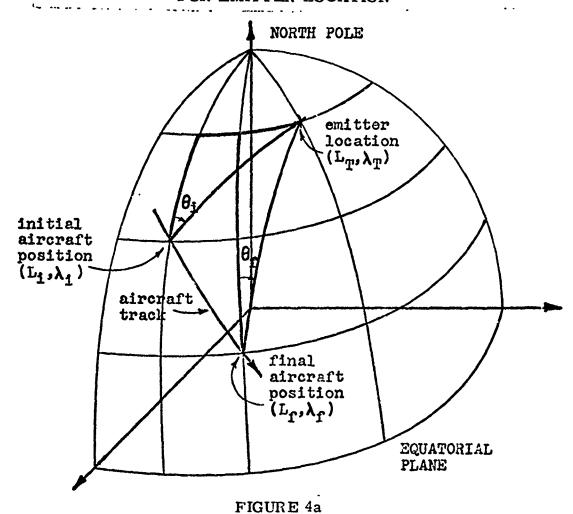
$$\tan \hat{\theta}_{f} = \frac{\tan \left[(\lambda_{T} - \lambda_{f}) \cos L_{T} \right]}{\sin(L_{T} - L_{f})} . \tag{31}$$

These equations represent exact solutions for emitter location on a spherical earth of constant radius. It is known that the earth is an oblate spheroid so small errors may be introduced at large distances from the emitter with such a spherical earth solution. However, according to Kayton and Fried [9], a constant radius of curvature can be used everywhere on the earth with an error of less than 0.2 percent of distance.

If the emitter location problem is concerned only with microwave frequency emitters, then relatively short distances from the

FIGURE 4

SPHERICAL EARTH TRIANGULATION SOLUTION FOR EMITTER LOCATION



SPHERICAL TRIANGULATION
GEOMETRY
(L_T, \lambda_i)

A
(L₁, \lambda_i)

FIGURE 4b

INITIAL SPHERICAL

TRIANGLE

(

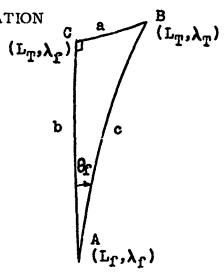


FIGURE 4c FINAL SPHERICAL TRIANGLE

aircraft to the emitter will exist since microwave electromagnetic (EM) energy propagation approximates a straight line path (line-of-sight) on the earth's surface. Reintjes and Coate [11] show that the maximum distance for direct EM wave propagation to a radar horizon is given by

$$d = \sqrt{2h_a} + \sqrt{2h_e}, \qquad (32)$$

where d = radar horizon in statute miles

ha = aircraft altitude in feet

h_e = emitter altitude in feet

assuming a 4/3 effective radius of the earth for standard EM energy refraction characteristics. If h_e is unknown or assumed to be zero, then aircraft altitudes (h_a) up to 80,000 feet should not yield radar horizons greater than 400 statute miles (350 nautical miles). At this distance latitude/longitude differences between the aircraft and emitter are small (less than 6°), and it can be seen that the emitter location algorithm developed from equations (24) and (25) is nearly equivalent to the exact spherical solution from equations (30) and (31) since the tangent and sine functions of small angles are approximately equal to the radian measure of the angles themselves. Thus for ranges up to 360 nautical miles from the emitter, the planar solution will be less than 0.5 percent in error from the spherical solution. For ranges of less than 150 nautical miles the error may be neglected and the plane triangulation solution may be considered sufficiently accurate for emitter location estimation.

If greater accuracy is desired, the spherical triangulation equations (30) and (31) may be solved by an iterative procedure as suggested in Appendix E. It was found to be tedious to attempt a closed-form solution.

3. Earth-Centered Coordinate System Solution

Another approach to the spherical earth solution was considered using tangent plane projections and an earth-centered coordinate system (x_1, x_2, x_3) , where the x_3 axis is the polar axis and the x_1 , x_2 plane is the equatorial plane of the earth. An observation plane was defined for each emitter-bearing line, which passed through the center of the earth and inscribed a great circle arc on the earth's surface between the aircraft and emitter positions. The procedure for obtaining the emitter location required solving the equations of the observation planes, two at a time, with a constraint equation that the emitter be located on the earth's surface. Assuming a constant earth radius and letting x_3 equal a constant, the emitter position may be computed from the equations of the plane (x_1', x_2', c) where x_3 = constant as shown below.

$$a_{11}x_1' + a_{12}x_2' = -a_{13}c$$
 (33)

$$a_{21}x_1' + a_{22}x_2' = -a_{23}c$$
 (34)

These equations were then transformed back to the original coordinate set (x_1, x_2, x_3) by the equation

$$x_i = Ax_i$$
 $i = 1, 2, 3.$ (35)

where
$$A = Re/(x_1^2 + x_2^2 + c^2)^{\frac{1}{2}}$$
 (36)

and Re = average radius of the earth.

With proper coordinate rotation, it is possible to have the projection plane $(x_3 = c)$ tangent to the earth's surface and centered at the aircraft observation position.

F. EXTENDED KALMAN FILTERING

It may be desirable to directly filter and update the estimates of emitter position (L_T, λ_T) as described by Reeves [6]. If only bearing angle-of-arrival information is available as observable data, however, a non-linear transformation must be utilized to transform observable bearing angles into filtered position estimates.

This was accomplished by extended Kalman filtering techniques utilizing the concepts of small perturbation theory and a Taylor series expansion about an initial point. Equation (1) then becomes

$$\underline{Z}(k) = N[\underline{X}(k)] + \underline{V}(k)$$
 (37)

where Z(k) represents the observable bearing angles, X(k) is the emitter position vector, and N represents the non-linear transformation given by

$$\widehat{\boldsymbol{\theta}} = \tan^{-1} \left(\frac{(\lambda_{\mathrm{T}} - \lambda) \cos L_{\mathrm{T}}}{L_{\mathrm{T}} - L} \right). \tag{38}$$

If the position error of the state vector X(k) is given by

$$\widetilde{X}(k) = X(k) - \widehat{X}(k), \qquad (39)$$

then the true position $\underline{X}(k)$ may be expanded about the most recent optimal estimate $\widehat{\underline{X}}(k)$ in a Taylor series expansion with higher order.

terms neglected as shown below:

$$N\left[X(k)\right] = N\left[\widehat{X}(k)\right] + \underline{M}(k)\underline{\widetilde{X}} + \dots \qquad (40)$$

where
$$\underline{\underline{M}}(k) = \frac{\partial \underline{N}}{\partial \underline{\underline{X}}} \Big|_{\hat{X}} = \left[\frac{\partial \hat{\theta}}{\partial \lambda_{\mathrm{T}}} - \frac{\partial \hat{\theta}}{\partial L_{\mathrm{T}}} \right]$$
 (41)

for the emitter location algorithm given by equations (24) and (25).

The Kalman filter recursion equations (6) through (8) may then be rewritten to account for this non-linear observation matrix and are given by

$$\underline{G}(k) = \underline{P}(k/k-1)\underline{M}(k)^{T} \left[\underline{M}(k)\underline{P}(k/k-1)\underline{M}(k)^{T} + \underline{R}(k)\right]^{-1}$$
(42)

$$\mathbf{P}(\mathbf{k}/\mathbf{k}) = \mathbf{P}(\mathbf{k}/\mathbf{k}-1) - \mathbf{G}(\mathbf{k})\mathbf{M}(\mathbf{k})\mathbf{P}(\mathbf{k}/\mathbf{k}-1)$$
 (43)

$$\underline{\mathbf{P}}(\mathbf{k}+1/\mathbf{k}) = \underline{\underline{\Phi}}(\mathbf{k}+1,\mathbf{k})\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k})\underline{\underline{\Phi}}(\mathbf{k}+1,\mathbf{k})^{\mathrm{T}} + \underline{\mathbf{Q}}(\mathbf{k}) . \tag{44}$$

Since the non-linear filter is processing estimates of a stationary emitter position, the Φ matrix becomes the identity matrix and the equations are considerably simplified and reduced to scalar form (see Appendix F).

The greatest difficulty in this approach was found in initialization of the extended Kalman filter process. Errors or inaccuracies in initialization may cause the recursive computations to diverge, and it was therefore decided to commence extended Kalman filter processing only after linear filtering and smoothing of the bearing angle observations had been completed and an initial optimal estimate of emitter position obtained. This would allow the extended Kalman filter process to be used as a final correction procedure to further improve the final optimal estimates of emitter position.

IV. PRESENTATION OF RESULTS

A. MONTE CARLO SIMULATION ANALYSIS

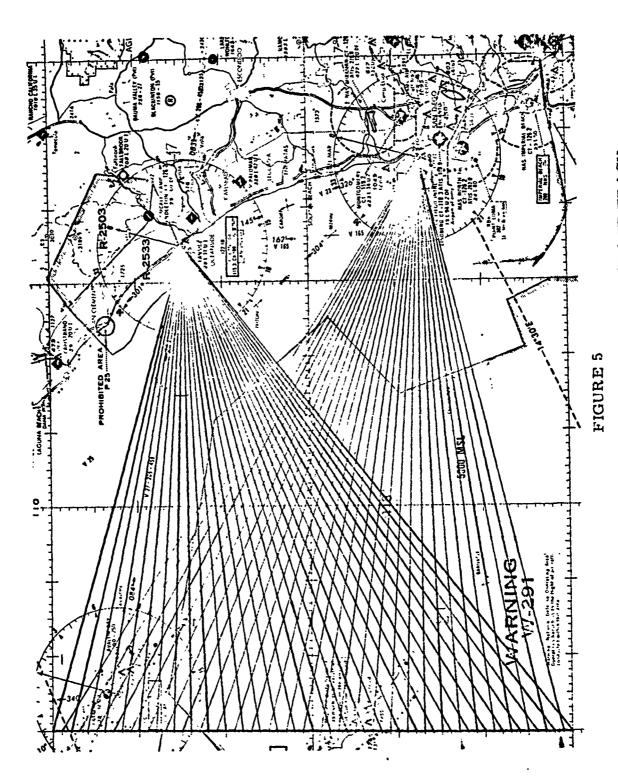
Since this Kalman filter program was developed to process actual aircraft flight data, a Monte Carlo simulation was utilized to accomplish program validity checks and error analysis studies. Figure 5 depicts the geometry of the simulation with an aircraft flying due south at 600 knots off the coast of southern California. Two VORTAC stations were chosen as emitters with accurately known positions given as

EMITTER #1	Oceanside VORTAC Channel 100 Frequency 1197 mHz	33.24055° N -117.41694° W
	- •	

EMITTER #2 San Diego VORTAC 32.78222°N
Channel 125 -117.22444°W.
Frequency 1212 mHz

Bearing angle-of-arrival information was assumed to arrive every six seconds alternately from each target giving a twelve-second uniform sampling interval for simplicity. Known true bearing angles were computed numerically in the absence of measurement noise and allowed position location within twenty feet of the known locations given above.

In the presence of computer-generated random measurement noise of one degree variance, a large number of simulation runs were executed to obtain a statistically valid error analysis. The results of this analysis are given in Appendix G and summarized below for the two simulation emitters. A range error was computed from each



PLOT OF MONTE CARLO SIMULATION AIRCRAFT TRACK AND EMITTER BEARING ANGLES-OF-ARRIVAL

position estimation latitude and longitude error, and all errors were statistically averaged for the total number of runs.

MONTE CARLO SIMULATION ERROR ANALYSIS

average error in nautical miles

emitter	average latitude error	average longitude error	average range error	closest point of approach	normalized estimated position range error
1	0.44852	0.23376	0.53692	54.5	0.985%
2	0.39961	0.23108	0.50758	64.7	0.785%

These results show that the basic plane triangulation solution emitter location algorithm is capable of estimating position within one percent of range at emitter distances of less than 100 nautical miles.

B. AIRCRAFT FLIGHT DATA ANALYSIS

The Kalman filter program was also utilized to process and filter aircraft flight data of the form discussed in Section III, Part A, of this paper. This data processing tested maximum utility of the computer program since certain data samples would not correlate with predicted emitter locations and had to be selectively gated out. Figures 6 through 10 show graphs of noisy unfiltered bearing angle-of-arrival information initially associated with a distinct emitter target. The dotted lines show unfiltered bearing angles and the solid lines show the smoothed initial and filtered final bearing angle. It can be seen that Figures 7, 9, and 10 have erroneous bearing angles which must be excluded from the

Kalman filtering process by the adaptive gating technique. This is shown to occur in the results of the computer output on page 64. Also comparing the final JSET data (page 65) with the initial JSET data (page 63) shows which data samples were not correlated to a distinct emitter target and were discarded.

The emitter location estimation accuracy obtained from the results of processing and filtering flight data was not as good as that obtained from the Monte Carlo simulation; however, it was estimated that the standard deviation of bearing angle measurement noise was approximately 12° and that some aircraft navigation error may have existed as well. Therefore, typical emitter location estimation within 3 percent on an average range from aircraft to emitter target of 100 nautical miles was considered reasonable.

FIGURE 6

PLOT OF NOISY MEASURED BEARING ANGLES
ASSOCIATED WITH EMITTER TARGET

NTAR = 1

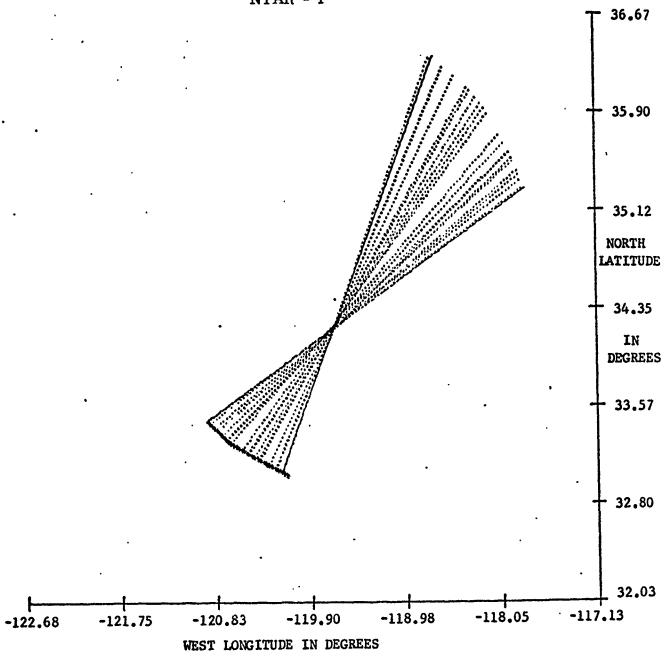


FIGURE 7

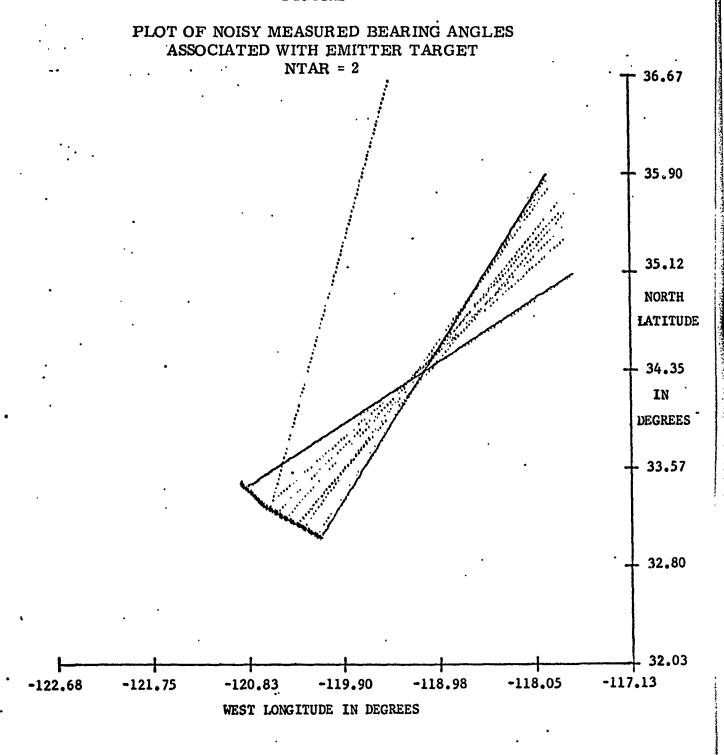


FIGURE 8

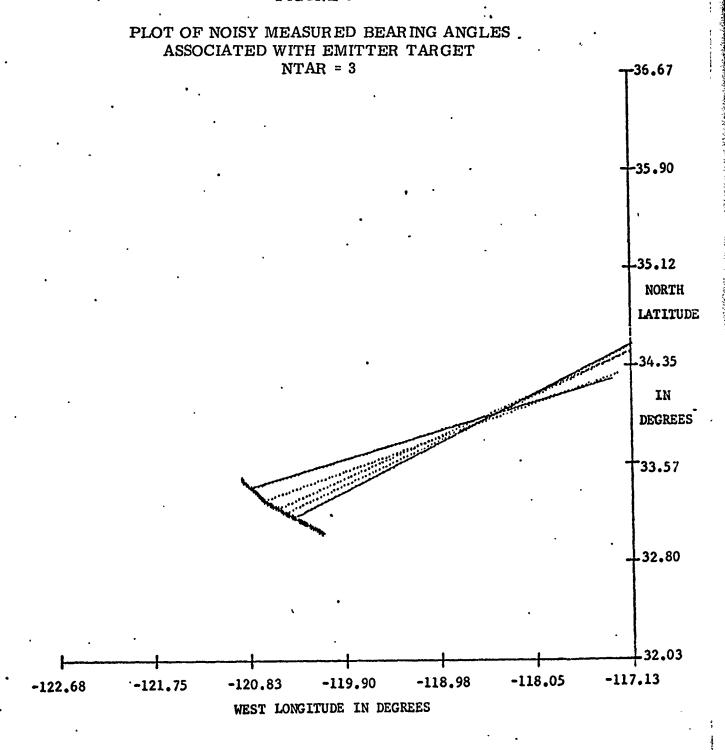
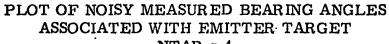


FIGURE 9



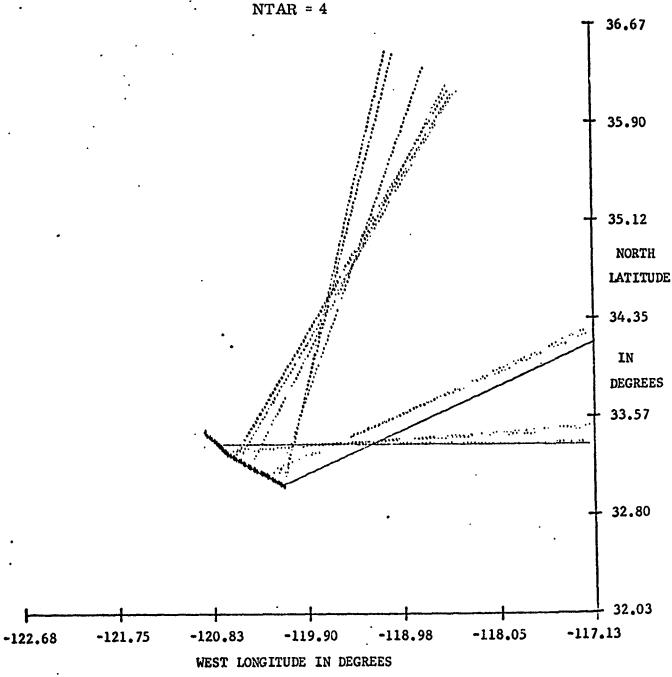
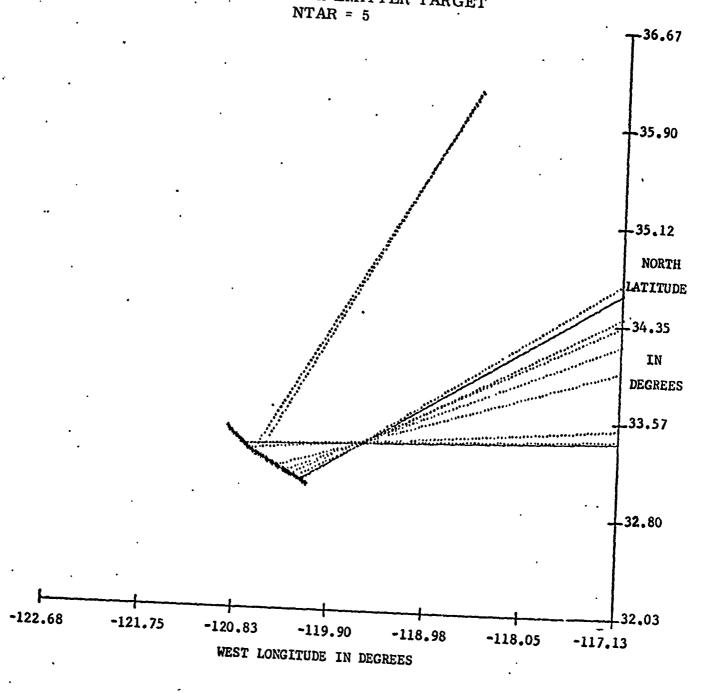


FIGURE 10

PLOT OF NOISY MEASURED BEARING ANGLES ASSOCIATED WITH EMITTER TARGET NTAR = 5



v. CONCLUSIONS

It was found that Kalman filtering techniques have useful application in the field of airborne direction-finding and emitter location procedures. Many airborne directional antenna receiving systems have $\pm 10^{9}$ typical bearing accuracies which yield poor results in conventional plotting and triangulation procedures for emitter position fixing. The Kalman filtering process, in effect, statistically filters out random signal error or measurement noise, resulting in an optimal estimate of the true signal.

By computer processing techniques these emitter data may be sorted and sequentially filtered to yield optimal estimates of bearing angle-of-arrival information. Computerized solutions, as suggested within this paper, may then be utilized to compute numerical emitter locations in latitude/longitude coordinates to a high degree of accuracy. For limited-storage airborne computer applications the Kalman filtering process and plane triangulation solution should be readily adaptable to airborne implementation to yield real-time cockpit display outputs with typical estimated emitter position error within three percent of the exact value.

For more precise resolution of emitter location, the spherical earth triangulation solution may be considered; or alternately, the extended Kalman filtering techniques may be utilized to improve or

update plane triangulation solution initial estimates of emitter location.

These procedures are somewhat more involved, however, and would require greater computer memory and processing time.

APPENDIX A

DERIVATION OF SCALAR KALMAN FILTER RECURSION EQUATIONS

The Kalman filter recursion equations from page 11, equations (6), (7), and (8), are listed below:

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

$$P(k/k) = P(k/k-1) - G(k)H(k)P(k/k-1)$$
 (2A)

$$\underline{\mathbf{P}}(\mathbf{k}+1/\mathbf{k}) = \underline{\boldsymbol{\Phi}}(\mathbf{k}+1,\mathbf{k})\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k})\underline{\boldsymbol{\Phi}}(\mathbf{k}+1,\mathbf{k})^{\mathrm{T}} + \underline{\mathbf{Q}}(\mathbf{k})$$
(3A)

where in this application $\Phi(k+1,k) = \begin{bmatrix} 1 & T(k+1) \\ 0 & 1 \end{bmatrix}$,

$$\underline{\mathbf{H}}(\mathbf{k}) = \begin{bmatrix} 1 & 0 \end{bmatrix}$$
,

the \underline{P} and \underline{Q} matrices may be considered to have scalar components given by

$$\begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} Q11 & Q12 \\ Q21 & Q22 \end{bmatrix} \quad \text{respectively,}$$

and the \underline{G} matrix is a 2x1 matrix of the form $\begin{bmatrix} G1 \\ G2 \end{bmatrix}$.

Writing equation (1A) in matrix notation yields

$$\begin{bmatrix} G1 \\ G2 \end{bmatrix} = \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + R \begin{bmatrix} -1 \\ 0 \end{bmatrix} (4A)$$

and for the observation matrix $\underline{H}(k) = \begin{bmatrix} 1 & 0 \end{bmatrix}$, the inverse term becomes a scalar allowing the gain terms to be computed directly, as is shown below:

$$\begin{bmatrix} G1 \\ G2 \end{bmatrix} = \begin{bmatrix} P11 \\ P21 \end{bmatrix} \begin{bmatrix} P11 + R \end{bmatrix}^{-1}$$
 (5A)

which results in the scalar gain equations

G1(k) =
$$\frac{P11(k/k-1)}{P11(k/k-1) + R(k)}$$
 (6A)

$$G2(k) = \frac{P21(k/k-1)}{P11(k/k-1) + R(k)}$$
 (7A)

From equation (19) which defines the \underline{P} matrix, it can be seen that since $\hat{\theta}$ and $\hat{\theta}$ are statistically independent, where

$$\tilde{\theta} = \Theta - \hat{\theta}$$
,

then P12 = P21, and the \underline{P} matrix is symmetric. From equation (15) it can be seen that the \underline{Q} matrix is also symmetric. Equation (7A) then becomes

$$G2(k) = \frac{P12(k/k-1)}{P11(k/k-1) + R(k)} . \tag{8A}$$

To solve for the prediction covariance scalar terms, equation (2A) was substituted into (3A) giving

$$\begin{bmatrix} P11 & P12 \\ P12 & P22 \end{bmatrix} = \begin{bmatrix} 1 & T \\ 0 & 1 \end{bmatrix} \begin{bmatrix} P11 & P12 \\ P12 & P22 \end{bmatrix} - \begin{bmatrix} G1 \\ G2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ P12 & P22 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ P12 & P22 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ T & 1 \end{bmatrix} + \begin{bmatrix} Q11 & Q12 \\ Q12 & Q22 \end{bmatrix}$$

$$k+1/k \qquad k/k-1 \qquad k/k-1$$

which upon simplifying and writing in scalar form yields equations as noted on page 84 of the Kali: in filter program.

APPENDIX B

DERIVATION OF SCALAR SMOOTHING EQUATIONS

The general smoothing equations (11) and (12) are given on page 13 and are repeated below:

$$\frac{\Lambda}{X}(1/k) = \frac{\Lambda}{X}(1/k-1) + D(1/k)C(k)E(k)$$
 (1B)

where
$$\underline{D}(1/k) = \underline{D}(1/k-1)\underline{P}(k-1/k-1)\underline{\Phi}(k, k-1)^{T}\underline{P}(k/k-1)^{-1}$$
 (2B)

and
$$\underline{E}(k) = \underline{Z}(k) - \underline{H}(k)\underline{X}(k/k-1)$$
. (3B)

In order to solve for a smoothed value of the state X(1/k), it is first necessary to solve equation (2B) sequentially since the state transition matrix Φ is a function of the variable sampling interval as given by

$$\underline{\Phi}(k,k-1)^{T} = \begin{bmatrix} 1 & 0 \\ T(k) & 1 \end{bmatrix} . \tag{4B}$$

It is also necessary to obtain the inverse of the error covariance matrix P in order to solve equation (2B). On the IBM 360, the matrix inverse routine MINV was used with good numerical accuracy, and the results were programmed as follows:

$$\underline{P}^{-1} = \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix}^{-1} = \begin{bmatrix} PIN11 & PIN12 \\ PIN21 & PIN22 \end{bmatrix};$$
 (5B)

and letting

$$\underline{P} = \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix}, \quad \underline{D} = \begin{bmatrix} D11 & D12 \\ D21 & D22 \end{bmatrix}, \quad \text{and} \quad \underline{Q} = \begin{bmatrix} Q11 & Q12 \\ Q21 & Q22 \end{bmatrix}$$

as before. Since the \underline{P} matrix is symmetric (as shown in Appendix A), \underline{P}^{-1} is also symmetric.

To solve for the scalar smoothing equations, equation (8) on page 11 was solved for P(k-1/k-1) giving

$$\underline{\mathbf{P}}(\mathbf{k}-1/\mathbf{k}-1) = \underline{\boldsymbol{\Phi}}(\mathbf{k},\mathbf{k}-1)^{-1} \left[\underline{\mathbf{P}}(\mathbf{k}/\mathbf{k}-1) - \underline{\mathbf{Q}}(\mathbf{k})\right] \underline{\boldsymbol{\Phi}}^{\mathrm{T}}(\mathbf{k},\mathbf{k}-1)^{-1}. \tag{6B}$$

This equation was then substituted into equation (2B) giving

$$\underline{D}(1/k) = \underline{D}(1/k-1)\underline{\Phi}(k,k-1)^{-1} \left[\underline{P}(k/k-1) - \underline{Q}(k)\right]\underline{P}(k/k-1)^{-1}, \qquad (7B)$$

which written in matrix form yields

$$\begin{bmatrix} D11 & D12 \\ D21 & D22 \end{bmatrix} = \begin{bmatrix} D11 & D12 \\ D21 & D22 \end{bmatrix} \begin{bmatrix} 1 & -T \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} Q11 & Q12 \\ Q12 & Q22 \end{bmatrix} \begin{bmatrix} PIN11 & PIN12 \\ PIN12 & PIN22 \end{bmatrix}.$$

This reduces to

$$\begin{bmatrix} D11 & D12 \\ D21 & D22 \end{bmatrix} = \begin{bmatrix} D11 & D12 \\ D21 & D22 \end{bmatrix} \begin{bmatrix} 1 & -T \\ 0 & 1 \end{bmatrix} \begin{pmatrix} 1.0 & -QP11 & -QP12 \\ -QP21 & 1.0 & -QP22 \end{pmatrix}$$

where
$$QP11 = Q11(PIN11) + Q12(PIN12)$$

$$QP12 = Q11(PIN12) + Q12(PIN22)$$

$$QP21 = Q12(PIN11) + Q22(PIN12)$$

$$QP22 = Q12(PIN12) + Q22(PIN22)$$

and D(1/0) was initialized as the identity matrix. These are the terms used in the Kalman filter program to solve for sequential values of the D matrix as seen on page 84.

APPENDIX C

DERIVATION OF PLANE TRIANGULATION SOLUTION

From Figure 1 it can be seen that

$$\tan \theta_i = \frac{X_T - X_i}{X_T - Y_i}$$
 (1C)

and
$$\tan \hat{\theta}_f = \frac{X_T - X_f}{X_T - Y_f}$$
 (2C)

Substituting latitude (L)/longitude (λ) coordinates for the X/Y axes of the flat earth model results in angular measurement of distance in the plane. Since equal angles of latitude and longitude yield equal distances of movement only at the equator, a correction factor must be applied to longitude angular measure as either pole is approached from the equator.

At 60° North latitude, for example, one degree of movement measures sixty nautical miles while one degree of movement in longitude only measures thirty nautical miles. This requires that the longitude angular measure be corrected by the cosine of the local latitude coordinate so that the tangent ratio terms will have units of the same dimensional size. This results in the equations given below:

$$tan \hat{\theta}_{i} = \frac{(\lambda_{T} - \lambda_{i})cosL_{T}}{L_{T} - L_{i}}$$
(3C)

$$\tan \hat{\theta}_{f} = \frac{(\lambda_{T} - \lambda_{f}) \cos L_{T}}{L_{T} - L_{f}}$$
 (4C)

These two equations could be solved simultaneously for the two unknowns L_T and λ_T , except that the unknown L_T also appears as a cosine function. The method chosen to resolve this difficulty allowed a known midlatitude function

$$WAV = \frac{L_i + L_f}{2}$$
 (5C)

to be used in approximating an intermediate value of emitter target latitude TILA where

$$TILA = \frac{(\lambda_f - \lambda_i)\cos(WAV) + L_i \tan \hat{\theta}_i - L_f \tan \hat{\theta}_f}{\tan \hat{\theta}_i - \tan \hat{\theta}_f}$$
(6C)

results from substituting $\cos(\text{WAV})$ for $\cos L_T$ in a simultaneous solution of equations (3C) and (4C) for L_T . It should be noted that this midlatitude approximation is only valid if the aircraft position latitude is within a few degrees of the emitter location latitude. This approximation would not be valid for HF sky wave radiation transmitted from an emitter at great distances from the receiving aircraft. For situations in which the approximation is valid, equations (28) and (29) give a corrected solution for emitter latitude and longitude based on the intermediate latitude function TILA.

APPENDIX D

DERIVATION OF SPHERICAL EARTH TRIANGULATION SOLUTION

Figure 4a describes the geometry of the spherical triangulation solution, in which the arc length sides of the spherical triangles must be measured from the angle subtended at the center of the sphere. Since latitude (L) and longitude (λ) lines meet at right angles and θ_i and θ_f are known, then two angles of the spherical triangles are known. This allows the side opposite and the side adjacent to the emitter bearaing angle θ to be computed.

Using the Law of Sines and the Angle Law of Cosines for spherical triangles, the general solution equations may be derived as follows from Figures 4b and 4c:

Sine Law:
$$\frac{\sin(A)}{\sin(a)} = \frac{\sin(B)}{\sin(b)} \text{ or } \sin(A) = \frac{\sin(B)\sin(a)}{\sin(b)}$$
 (1D)

Cosine Law:
$$cos(A) = -cos(B)cos(C) + sin(B)sin(C)cos(a)$$
. (2D)

For angle $C = 90^{\circ}$ equation (2D) reduces to

$$cos(A) = sin(B)cos(a)$$
. (3D)

Dividing equation (1D) by (3D) yields

$$tan(A) = \frac{sin(A)}{cos(A)} = \frac{tan(a)}{sin(b)}$$
(4D)

where a and b are angular arcs of the spherical triangles measured at the center of the sphere. From Figures 4b and 4c, b is an arc of latitude and a is an arc of longitude which must be corrected for the local latitude at which it is measured. Therefore,

$$a = (\lambda_T - \lambda) \cos L_T \tag{5D}$$

and
$$b = L_T - L$$
. (6D)

Substituting these values of a and b into equation (5D) results in

$$\tan \theta_{i} = \frac{\tan \left[(\lambda_{T} - \lambda_{i}) \cos L_{T} \right]}{\sin(L_{T} - L_{i})}$$
(7D)

and
$$\tan \theta_f = \frac{\tan \left[(\lambda_T - \lambda_f) \cos L_T \right]}{\sin(L_T - L_f)}$$
 (8D)

which give the exact solution for emitter location on a spherical earth of constant radius.

APPENDIX E

COMPUTER ITERATION SOLUTION TO SPHERICAL TRIANGULATION EQUATIONS

No closed-form solution to the two spherical triangulation equations (30) and (31) could be found since the unknowns (L_T , λ_T) are enmeshed within multiple trigonometric relations. Instead, an iterative approximation computer solution was developed which allowed successive numerical updating of previously computed values. This method is only as good as the initial estimate with which it commences, so it is best described as a spherical earth correction procedure to the plane triangulation solution for emitter location. The procedure is concerned with updating values of the following terms

DTLAF =
$$L_T - L_f$$

DTLOI =
$$\lambda_T - \lambda_i$$

DTLOF =
$$\lambda_T - \lambda_f$$

by sequentially recomputing them from rearranged versions of equations (30) and (31). The procedure is repeated and successive terms are compared until a satisfactory solution is reached. See the computer subroutine on the following page for a further explanation of the iterative process.

APPENDIX F

DERIVATION OF EXTENDED KALMAN FILTER SCALAR EQUATIONS

The extended Kalman filter recursion equations (42), (43), and (44) are similar in form to the basic Kalman filter recursion equations on page 11, except that the observation matrix is a non-linear transformation matrix M as defined by equation (41) where

$$M(k) = \frac{\partial T}{\partial X} \Big|_{X} = \left[\frac{\partial \theta}{\partial \lambda_{T}} \frac{\partial \theta}{\partial L_{T}} \right] = \left[DMX DMY \right]. \tag{1F}$$

Taking partial derivatives of equation (24) yields

$$DMX = \frac{\partial \theta}{\partial \lambda_{T}} = \frac{(L_{T} - L_{i})^{2} cosL_{T}}{(L_{T} - L_{i})^{2} + (\lambda_{T} - \lambda_{i})^{2} cos^{2}L_{T}}$$
(2F)

$$DMY = \frac{\partial \theta}{\partial L_T} = \frac{-(\lambda_T - \lambda_i) \left[(L_T - L_i) \sin L_T + \cos L_T \right]}{(L_T - L_i)^2 + (\lambda_T - \lambda_i)^2 \cos^2 L_T}$$
(3F)

These terms are computed numerically and substituted into equations (42) and (43) which are rewritten in matrix form as

$$\begin{bmatrix} GX \\ GY \end{bmatrix} = \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix} \begin{bmatrix} DMX \\ DMY \end{bmatrix} \left\{ \begin{bmatrix} DMX & DMY \end{bmatrix} \begin{bmatrix} P11 & P12 \\ P21 & P22 \end{bmatrix} \begin{bmatrix} DMX \\ DMY \end{bmatrix} + R \right\} (4F)$$

and

$$\begin{bmatrix}
P11 & P12 \\
P21 & P22
\end{bmatrix} = \begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix} \begin{bmatrix}
P11 & P12 \\
P21 & P22
\end{bmatrix} \begin{bmatrix}
1 & 0 \\
0 & 1
\end{bmatrix} + \begin{bmatrix}
Q11 & Q12 \\
Q12 & Q22
\end{bmatrix} (5F)$$

Equation (43) is then substituted into (5F); however, the fact that the Φ matrix reduces to the identity matrix for the stationary filtered states (L_T , λ_T) considerably simplifies the recursion equations since they are no longer dependent on the sampling interval T(k). Since the gain terms are computed as a function of the non-linear terms DMX and DMY, it is essential to have the correct units on the error term E(k). In this case the units of gain are not dimensionless but degrees per radian, so the error term must be given in radians to update position estimates given in degrees of latitude and longitude.

These scalar equations may then be written out and programmed sequentially as was done in Appendix A for the linear recursion equations. See page 92 for a listing of these equations within the extended Kalman filter subroutine.

APPENDIX G

MONTE CARLO SIMULATION ERROR ANALYSIS ESTIMATED EMITTER POSITION ERROR IN NAUTICAL MILES

		EMITTER # 1			EMITTER # 2	
RUN	DLAT1	OLON1	DRNG1	DLAT2	DLON2	DRNG2
12345678901234567890123456789012345678901234567890	0.14932913362456863403357213142246665258897998306299131422466652588979983062991311313131313131313131313131313131313		9339053526223774926814490244929995664958776598509803662 57076133652647226356338580815235497765985098133662 22440235766831764323537256877955133572345825 N GE 1.00.00.00.00.00.00.00.00.00.00.00.00.00	00.05518772602609886374697800541780998863746978005418099886374929723180998863795607929723180998863795607929723180998863795667772423295738647555673997648890338187230449795786673996448390030880339938863748574697951433957886417438811723658683287886417438811723658683287886417438811723658687220128033997880416588657211802222222222222222222222222222222222	0.0.42879305887926410192388363139839552401845542879300585428791000000000000000000000000000000000000	00000011000100000000000000000000000000
RUNS	AVERAGE LATITUD ERROR		RANGE ERROR	LATITUDE - ERROR		RANGE ERROR
50	0.44852	0.23376	0.53692	0.39961	0.23108	0.50758

AIRCRAFT FLIGHT DATA COMPUTER OUTPUT

LISTING OF BEARING ANGLES-OF-ARRIVAL AND AIRCRAFT NAVIGATION DATA

N	HDGD (N)	SLAD(N)	SLOD(N)	BRNGD(N)	THETAD(N)	THETA(N)
12345678901234567890123÷567890123456789012345678901234567890123456789012345	99992066613209920666132099209920661320992099206613209920992066132099209920661320992099206613209920992099209920992099209920992099209	22166634172398181028717777970446476817555440332321075531791862545631249901255777648457109774797044664768177949955777368845719099747558113794862533108866577773738866554281164772796993188552299638858611379486265432221109988111030697397699318452288886111030696933333333333333333333333333333	90.887339259 90.8875239259 90.8875239259 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.887523927555128 90.88752392755512200.45555128 90.88752392755128 90.88752392755128 90.88752392755128 90.8875239275128 90.88752352 90.8875235 90.8875235 90.887523	\$\begin{align*} 500 \\ 550 \\ 640 \\ 645 \\	9969620152910137642209772362715265322259902477211225632644887372065667731376422097723627152653222635944169394743458264929100775653288019768574556047851376560478513765604785137656047851376560478513765604785137656047851376560478513765604785137656047851376458322268851869783122665755852776648323333333333333333333333333333333333	109815569838913298710453725226779669378483212227815311717773811864693338123338477315233486845046053048123336440870315511864698336538877322784806023466933007533644087021551186469333655288513926438684946933296215598851398342970215598821396885139826763329688765272884799090100000000000000000000000000000000

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 1
EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM	FREQ	PRF	PW	HDGD	BRNGD	THETAD
1	2872.0	320.0	2.40	135.52739	283.69995	59.22729
2	2872.0	322.0	2.50	135.64830	282.10010	57.74829
4	2872.0	320.0	2.50	135.41739	280.80005	56.21729
6	2872.0	318.0	2.40	135.30740	279.90015	55.20752
7	2872.0	320.0	2.60	135.00026	278.49976	53.50000
9	2872.0	319.0	2.20	135.28560	277.30005	52.58545
10	2872.0	321.0	2.50	135.51651	275.89966	51.41602
14	2872.0	320.0	2.60	135.43913	275.40015	50.83911
15	2872.0	320.0	2.20	135.38472	273.79980	49.18433
19	2872.0	320.0	2.40	131.33047	276.00000	47.33032
28	2872.0	319.0	2.60	121.44293	280.89990	42.34277
32	2872.0	319.0	2.50	120.89346	279.69995	40.59326
33	2872.0	320.0	2.50	121.27792	278.29980	39.57764
37	2872.0	320.0	2.40	121.12437	276.50024	37.62451
40	2872.0	320.0	2.50	120.72844	275.00000	35.72827
42	2872.0	320.0	2.50	120.65167	274.30005	34.95166
48	2872.0	319.0	2.30	118.74031	272.19995	30.94019
53	2872.0	323.0	2.80	120.64079	267.10010	27.74072
54	2872.0	320.0	2.50	120.65167	267.00000	27.65161
59	2872.0	322.0	2.50	120.75079	263.39990	24.15063

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 2
EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NTAR = 2

NUM	FREQ	PRF	PW	HDGD	BRNGD	THETAD
3	2876.0	306.0	2.40	135.48329	286.89990	62.38306
20	2878.0	304.0	2.40	122.29950	291.79980	54.09912
24	2876.0	305.0	2.40	120.76172	258.39990	19.16162
30	2878.0	306.0	2.30	121.58560	289.60010	51.18555
35	2876.0	306.0	2.30	120.68495	287.29980	47.98462
43	2876.0	306.0	2.40	119.66331	287.19946	46.86255
44	2876.0	306.0	2.30	119.27831	285.69995	44.97803
50	2876.0	306.0	2.50	118.82796	282.99976	41.82764
61	2876.0	306.0	2.30	121.01439	277.89990	38.91406
65	2876.0	306.0	2.40	119.79510	278.09985	37.89478

NTAR = 3

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 3

EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM	FREQ	PRF	PW	HDGD	BRNGD	THETAD
5	2846.0	370.0	2.30	135.31882	300.99976	76.31836
18	2846.0	370.0	2.30	134.45079	299.49976	73.95044
29	2846.0	370.0	2.40	121.57474	308.69971	70.27441
34	2846.0	370.0	2.40	119.94865	309.39990	69.34839
41	2846.0	370.0	2.40	120.66260	306.99976	67.66235

NTAR = 4

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 4

EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM ·	FREQ	PRF	PW	HDGD	BRNGD	THETAD
8	5556.0	320.0	1.00	135.06616	314.80005	89.86621
13	5554.0	640.0	1.10	135.49416	312.29956	87.79370
16	5556.0	320.0	1.00	135.30740	311.60010	86.90747
23	5556.0	640.0	1.00	120.76172	276.30005	37.06177
25	5556.0	320.0	1.10	120.87172	275.19995	36.07153
27	5554.0	640.0	1.10	121.36617	275.09985	36.46582
36	5554.0	640.0	1.10	120.70670	272.19995	32.90649
46	5554.0	640.0	0.90	118.83884	312.00000	70.83862
49	5554.0	320.0	1.00	118.72943	311.09985	69.82910
58	5556.0	640.0	1.00	120.21225	264.00000	24.21216
63	5556.0	640.0	1.00	121.15704	257.09985	18.25684
64	5556.0	320.0	1.10	119.75102	257.00000	16.75098

. NTAR = 5

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 5

EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM	FREQ	PRF	PW	HDGD .	BRNGD	THETAD
11	5506.0	640.0	0.60	135.61504	312.79980	88.41479
17	5504.0	640.0	0.60	135.16473	311.29980	86.46436
21	5504.0	640.0	0.60	120.58583	275.69971	36.28540
26	5504.0	640.0	0.60	120.98116	274.19971	35.18066
38	5504.0	640.0	0.60	121.17940	315.00000	76.17920
45	5506.0	640.0	0.70	118.99298	312.79980	71.79272
51	5504.0	642.0	0.60	119.64102	309.00000	68.64087
55	5504.0	640.0	0.60	120.88258	306.10034	66.98291
60	5504.0	642.0	0.60	120.77260	301.39966	62.17212

NTAR = 6

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 6

EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM	FREQ	PRF	PW	HDGD	BRNGD	THETAD
12	2876.0	359.0	2.20	135.56006	277.30005	52.86011
22	2876.0	359.0	2.20	120.73991	283.30005	44.03979
47	2876.0	359.0	2.40	118.71851	276.50024	35.21875
52	2874.0	359.0	2.40	120.40988	271.89990	32.30957
56	2876.0	359.0	2.20	121.12437	266.80005	27.92432
62	2874.0	359.0	2.40	121.15704	298.39966	59.55664

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 7
EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM	FREQ	PRF	PW	HDGD	BRNGD	THETAD
31	9268.0	500.0	0.20	120.76172	287.60010	48.36182
39	9268.0	499.0	0.20	120.91527	283.39966	44.31470
57	9270.0	500.0	0.30	121.22295	276.59985	37.82275

INITIAL JSET DATA

12467904598237028349304053 4566 1234 112223 \\delta \\delo \delta \delta \\delta \delta \\delta \\delta \delta \\delta \\d

LISTING OF KALMAN FILTER PARAMETERS

I	J	KK	K	G1(K)	G2(K)	T(K)	E(K)	GATE(K)
HHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHH	12345678901234567890	01246790459823702834 1112333344455	12467904598237028349	0.98580 0.99970 0.91377 0.51469 0.512820 0.388391 0.3325443 0.3325443 0.3338143 0.333737 0.333737 0.333837 0.333837 0.333837 0.333837	0.0 0.05132 0.01972 0.01124 0.00711 0.006451 0.004502 0.00314 0.00391 0.00256 0.00255 0.00255 0.00205 0.00244	0.0 19.48511 134.17871 19.57300 14.765814 24.465814 24.80640 24.568640 24.568640 24.5686860 73.93848 34.6116 24.684670 34.4116 24.84670 34.4116 24.84670 34.4116 24.84670 39.44677 39.4235	-0.54543 1.06323 0.15733 -0.812813 -0.520146 -0.71368 -0.823994 -0.823999 -0.129539 -0.129539 -0.123783 -0.22783 -0.66829 -0.145219	974.29150 20.42409 10.30163 8.610278 7.67091 7.73504 7.30789 7.37700 8.49747 8.015553 7.470842 7.70516 7.75814 7.17417 7.41809
222222222	1233456789	0300 2200 355 4450 61	304053 22333440 65015	0.98580 1.00000 0.61665 0.78886 0.66823 0.65760 0.42984 0.42162 0.58649 0.44875	0.0 0.00638 0.00505 0.00532 0.00511 0.00448 0.00325 0.00358 0.00309	0.0 156.82422 24.59717 69.00879 59.16089 68.85083 9.78320 34.46094 83.76660 24.76538	-0.86391 -33.63820 0.73169 -0.15179 2.24968 -0.72655 -2.12259 -0.05772 0.16925	7841. 21484 9.69064 13.05779 10.41681 10.25372 7.94606 7.88938 9.33059 8.08118
3333	12345	0 5 18 29 34	5 18 29 34 41	0.98580 1.00000 0.82455 0.62591 0.57737	0.0 0.00972 0.00552 0.00437 0.00393	0.0 102.83667 88.72388 49.30444 54.13794	1.98647 -1.63304 0.36708 0.09885	5141.83984 14.32434 9.80986 9.22935
44444444444	123444445666	083166166499 116664499	836357669834 12222344566	0.98580 0.99999 0.71699 0.79911 0.83155 0.88655 0.953787 0.9573787 0.74102 0.84422 0.86349	0.0 0.02562 0.01466 0.00917 0.00864 0.00742 0.00351 0.00551 0.00683 0.00651	0.0 39.03418 19.63843 54.08618 63.97461 63.63501 162.59375 251.067150 64.11743 98.58377	-0.70366 0.15646 -47.05379 -47.54167 -45.89462 -45.69676 -3.27011 0.14413 -41.61142 -45.44742 -46.34418	1951.72705 11.27851 13.38674 14.61884 17.81369 27.90096 40.328551 9.38385 11.79007 15.20187 16.23961
555555555	123334567	0 11 17 17 17 345 55 55	117 1216 1223 1550 150	0.98580 0.9999 0.83310 0.92856 0.98610 0.69876 0.57013 0.51924 0.48515	0.0 0.02916 0.01454 0.01037 0.00478 0.00599 0.00525 0.00459 0.00395	0.0 34.29346 34.41162 68.95825 187.19775 54.06836 34.47925 34.46753 34.52881	-0.70633 -48.22182 -47.36174 0.30156 -1.39986 -1.38312 0.18741 -2.30612	1714.69336 14.68663 22.44804 50.88951 10.93197 9.15131 8.65340 8.36201
666666	123456	0 12 22 47 52 56	12 22 47 52 56 62	0.98580 1.00000 0.95323 0.61332 0.56416 0.56612	0.0 0.01272 0.00368 0.00567 0.00613 0.00549	0.0 78.58911 201.91919 34.48628 34.49512 39.41577	-4.96594 13.84093 -0.14958 -2.30110 33.63329	3929.46289 27.74507 9.64890 9.08843 9.10896
7 7 7	1 2 3	0 31 39	31 39 57	0.98580 1.00000 0.91990	0.0 0.01452 0.00521	0.0 68.88379 132.89380	-0.75569 1.31592	3444.19727 21.20036

FINAL JSET DATA

NTAR = 1

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 1

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
1	2872.0	320.0	2.40	59.22729	59.22729	59.22729
2	2872.0	322.0	2.50	5 7. 74829	57.74831	59.22726
4	2872.0	320.0	2.50	56.21729	56.12552	59.07391
6	2872.0	318.0	2.40	55.20752	55.15414	59.04889
7	2872.0	320.0	2.60	53.50000	53.92235	59.18648
9	2872.0	319.0	2.20	52.58545	52.52164	59.16264
10	2872.0	321.0	2.50	51.41602	51.74019	59.24341
14	2872.0	320.0	2.60	50.83911	50.47720	59.14085
15	2872.0	320.0	2.20	49.18433	49.66541	59.24278
19	2872.0	320.0	2.40	47.33032	47.87863	59.36894
28	2872.0	319.0	2.60	42.34277	42.75356	59.51895
32	2872.0	319.0	2.50	40.59326	40.52049	59.50125
33	2872.0	320.0	2.50	39.57764	39.77419	59.53006
37	2872.0	320.0	2.40	37.62451	37.53296	59.51884
40	2872.0	320.0	2.50	35.72827	35.84486	59.52922
42	2872.0	320.0	2.50	34.95166	35.12247	59.53961
48	2872.0	319.0	2.30	30.94019	31.07832	59.54620
53	2872.0	323.0	2.80	27.74072	28.14043	59.55690
54	2872.0	320.0	2.50	27.65161	27.75360	59.55846
59	2872.0	322.0	2.50	24.15063	24.70815	59.56096

SMOOTHED INITIAL BEARING ANGLE = 59.56096 FILTERED FINAL BEARING ANGLE = 24.70815

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.98666N

EMITTER LONGITUDE =-119.60387W

NTAR = 2

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 2

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
3	2876.0	306.0	2.40	62.38306	62.38306	62.38306
20	2878.0	304.0	2.40	54.09912	54.09912	62.38306
30	2878.0	306.0	2.30	51.18555	51.03105	62.29199
35	2876.0	306.0	2.30	47.98462	48.03497	62.32124
43	2876.0	306.0	2.40	46.86255	46.09224	61.92072
44	2876.0	306.0	2.30	44.97803	45.39227	62.00427
50	2876.0	306.0	2.50	41.82764	43.05530	62.22186
61	2876.0	306.0	2.30	38.91406	38.93793	62.22676
65	2876.0	306.0	2.40	37.89478	37.80147	62.21768

SMOOTHED INITIAL BEARING ANGLE = 62.21768 FILTERED FINAL BEARING ANGLE = 37.80147

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 34.12234N

EMITTER LONGITUDE =-119.00533W

NTAR = 3

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 3

MUV	FREQ	PRF	PW	THETAD	THTD	THTD1
5	2846.0	370.0	2.30	76.31836	76.31836	76.31836
18	2846.0	370.0	2.30	73.95044	73.95042	76.31836
29	2846.0	370.0	2.40	70.27441	70.56093	76.49266
34	2846.0	370.0	2.40	69.34839	69.21106	76.45313
41	2846.0	370.0	2.40	67.66235	67.62057	76.44316

SMOOTHED INITIAL BEARING ANGLE = 76.44316 FILTERED FINAL BEARING ANGLE = 67.62057

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.81570N

EMITTER LONGITUDE =-118.15642W

NTAR = 4

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 4

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
8	5556.0	320.0	1.00	89.86621	89.86621	89.86621
13	5554.0	640.0	1.10	87.79370	87.79370	89.86620
16	5556.0	320.0	1.00	86.90747	86.86319	89.84418
46	5554.0	640.0	0.90	70.83862	70.91100	90.19023
49	5554.0	320.0	1.00	69.82910	69.76620	90.18146

SMOOTHED INITIAL BEARING ANGLE = 90.18146 FILTERED FINAL BEARING ANGLE = 69.76620

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.25166N

EMITTER LONGITUDE =-119.49638W

NTAR = 5

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 5

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
11	5506.0	640.0	0.60	88.41479	88.41479	88.41479
17	5504.0	640.0	0.60	86.46436	86.46436	88.41478
38	5504.0	640.0	0.60	76.17920	76.17416	88.38852
45	5506.0	640.0	0.70	71.79272	72.21440	88.49161
51	5504.0	642.0	0.60	68.64087	69.23541	88.57054
55	5504.0	640.0	0.60	66.98291	56.89281	88.56223
60	5504.0	642.0	0.60	62.17212	63.35942	88.63921

SMOOTHED INITIAL BEARING ANGLE = 88.63921 FILTERED FINAL BEARING ANGLE = 63.35942

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.25288N

EMITTER LONGITUDE =-119.45715H

NTAR = 6

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 6

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
12	2876.0	359.0	2.20	52.86011	52.86011	52.86011
22	2876.0	359.0	2.20	44.03979	44.03979	52.86009
47	2876.0	359.0	2.40	35.21875	34.57146	52.04729
52	2874.0	359.0	2.40	32.30957	32.36740	52.05992
56	2876.0	359.0	2.20	27.92432	28.92722	52.26216

SMOOTHED INITIAL BEARING ANGLE = 52.26216 FILTERED FINAL BEARING ANGLE = 28.92722

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.99094N

EMITTER LONGITUDE =-119.53415W

NT - R = 7

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 7

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
31	9268.0	500.0	0.20	48.36182	48.36182	48.36182
39	9268.0	499.0	0.20	44.31470	44.31470	48.36180
57	9270.0	500.0	0.30	37.82275	37.71735	48.22450

SMOOTHED INITIAL BEARING ANGLE = 48.22450 FILTERED FINAL BEARING ANGLE = 37.71735

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 34.14626N

EMITTER LONGITUDE =-119.10936W

MONTE CARLO SIMULATION COMPUTER OUTPUT

LISTING OF BEARING ANGLES-OF-ARRIVAL AND AIRCRAFT NAVIGATION DATA

N	HDGD (N)	SLAD(N)	SLOD(N)	BRNGD(N)	THETAD(N)	THETA(N)
1234567890123456789012345678901234567890123456789012345678901	180.00000 180.00000 180.00000 180.0000000 180.00000 180.00000	0260269279379360260269279379360260269279379360 036036936936036603603693603603693693693693693693693693693693693693693	-118.555500000000000000000000000000000000	977016514463218355883170975374368827233586584213197213558837159464444444453554831312272335865851323586583131311111111111111111111111111111	927166407732705576619729640304995780473168415655699133707557661972991403510127075576619729914035101270755766197299140351012707557661972991403510127072991403510127072991403510127072991403510127072991409111111111111111111111111111111111	63699287712584335466434846766230577423632552924674709094659756313680532003254845151893145043254632353584663707849915813383864431460967658378181351199286323358466376151984151117070786046316955319844453135119928721289323954473761151498411121121212121212121212121212121212121

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 1
EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRE OR PRE MULTIPLE

NUM	FREQ	PRF	₽₩	HDGD	BRNGD	THETAD
1357913579135791357913579	1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0	000000000000000000000000000000000000000	00000000000000000000000000000000000000	180.00000 180.00000	9840 9840	137760 137760 137760 137760 137760 137760 137760 137760 138820
6í	1197.0	150.0	3.50	180.30000	50.73331	51.08221

UNFILTERED EMITTER TARGET DATA INITIALLY ASSOCIATED TO NTAR = 2
EMITTER BEARING LINES OF SIMILAR FREQUENCY AND PRF OR PRF MULTIPLE

NUM FREQ PRF PW HDGD BRNGD	THETAD
2 1212.0 250.0 3.00 180.00000 123.17577 4 1212.0 250.0 3.00 180.00000 120.60915 8 1212.0 250.0 3.00 180.00000 119.27164 10 1212.0 250.0 3.00 180.00000 117.89893 12 1212.0 250.0 3.00 180.00000 116.49051 14 1212.0 250.0 3.00 180.00000 115.04587 16 1212.0 250.0 3.00 180.00000 115.04587 16 1212.0 250.0 3.00 180.00000 115.0731 22 1212.0 250.0 3.00 180.00000 108.92969 24 1212.0 250.0 3.00 180.00000 108.92969 24 1212.0 250.0 3.00 180.00000 107.32155 26 1212.0 250.0 3.00 180.00000 107.32155 26 1212.0 250.0 3.00 180.00000 107.32155 26 1212.0 250.0 3.00 180.00000 107.32155 26 1212.0 250.0 3.00 180.00000 107.32155 34 1212.0 250.0 3.00 180.00000 107.32155 34 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 97.1464 38 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 97.1453 36 1212.0 250.0 3.00 180.00000 91.84023 47 1212.0 250.0 3.00 180.00000 97.1473 36 1212.0 250.0 3.00 180.00000 97.1473 37 1212.0 250.0 3.00 180.00000 97.1473 38 1212.0 250.0 3.00 180.00000 97.1473 39 1212.0 250.0 3.00 180.00000 97.49355 40 1212.0 250.0 3.00 180.00000 86.73131 50 1212.0 250.0 3.00 180.00000 77.478148 51 1212.0 250.0 3.00 180.00000 77.478148	71 122.29161 122.29161 122.9984 119.94717 31 117.06317 37 116.65103 117.0631737 112.10175 37 112.10175 31 112.10175 31 109.82207 108.74931 109.82931 27 108.74931 28 100.94479 28 100.944980 98.43813 96.101837 98.43813 96.101837 98.44980 99.87880 87.80681 88.479225 88.47825 88.47825 88.47825 88.47825 88.48825

INITIAL JSET DATA

11111222222333334444445555556 111112222233333444446555556

LISTING OF KALMAN FILTER PARAMETERS

I	J	KK	Κ	G1(K)	G2(K)	T(K)	E(K)	GATE(K)
	1234567890123456789012345678901	01357913579135791357913579	13579135791357913579135791	0.9999 0.99999 0.99934457 0.999324155 0.653914457 0.653914457 0.653914457 0.653914457 0.653914457 0.653914457 0.653914457 0.6417717 0.6417717 0.6417717 0.6417714 0.6417714 0.6417714 0.6417714 0.641771 0.641771 0.641771 0.641771	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	0.0 12.00000	-0.296147 -2.425347 -1.4252446 -2.4252446 -1.6483641 -0.5264280 -0.209281 -0.209281 -0.209281 -0.209281 -0.209281 -0.209281 -0.42633684 -0.537370 -0.537370 -0.537370 -0.76446593 -1.90929 -1.90	36 7.428 924 948 948 948 948 948 948 948 948 948 94
 	123456789012345678901234567890	024680246802468024680246802468	2468024680246802468024680	0.993740552257671800.9933245576779445760391445764737722086554532147177124417717144417714441771444177144417714441771440.4417714	233 3744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 37744185 3774118 37741	12.000000 12.000000 12.000000 12.000000 12.000000 12.000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.0000000 12.00000000 12.00000000 12.0000000000	0.443889777112779282556890833335548801443622556891044378231355484910443788596715494992000000000000000000000000000000000	6634701903881408894473475333333334663470190388140889447344753333333333333333333333333333333

FINAL JSET DATA

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 1

NUM	FREQ	PRF	PW	THETAD	THTD	THTD1
13579135791357913579135791	1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0 1197.0	00000000000000000000000000000000000000	00000000000000000000000000000000000000	104.00 107.00	98.53.249 99.589 99.	105.22139 105.59840 105.598478 105.598478 105.709504 105.709504 105.997471 105.997471 105.997471 105.997471 106.004293 106.004293 106.004293 106.00151 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993 106.01993

SMOOTHED INITIAL BEARING ANGLE = 106.01910 FILTERED FINAL BEARING ANGLE = 50.46893

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 33.24187N

EMITTER LONGITUDE =-117.42503W

FILTERED AND SMOOTHED EMITTER TARGET DATA CORRELATED TO NTAR = 2

NUM FREQ	PRF	PW	THETAD	THTD	THTD1
2 1212.0 4 1212.0 6 1212.0 10 12 12 12 12 12 12 12 12 12 12 12 12 12	000000000000000000000000000000000000000	00000000000000000000000000000000000000	2.24252 122.29161 122.2929847 129.94717 119.947182 117.665103 112.163103 112.163103 112.163103 112.163103 112.163103 112.163103 112.163103 112.163103 110.	22.122.123.25 122.163.25 122.163.25 123.163.25 124.35.25 1127.663.55 1117.663.55 1117.663.55 1117.663.55 1117.663.55 1117.663.55 1117.663.55 1112.67 1013.65 1	22.4.74.790 24.4.74.74.790 24.4.74.74.74.74.74.74.74.74.74.74.74.74.

SMOOTHED INITIAL BEARING ANGLE = 123.42451 FILTERED FINAL BEARING ANGLE = 76.07950

PLANE TRIANGULATION SOLUTION OF EMITTER LOCATION

EMITTER LATITUDE = 32.78056N

EMITTER LONGITUDE =-117.23346W

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              **********************
                        KALMAN FILTER PROGRAM TO ESTIMATE EMITTER LOCATION FROM FILTERED BEARING ANGLES-OF-ARRIVAL
                                                                                                                                                                                                                 *
              **************************************
             THIS FORTRAN PROGRAM IS DESIGNED TO ANALYZE AND SORT ELECTRONIC EMITTER PARAMETERS AND AIRCRAFT NAVIGATION DATA, TO FILTER EMITTER DATA USING KALMAN FILTER TECHNIQUES TO MINIMIZE BEARING ANGLE-OF-ARRIVAL MEASUREMENT NOISE, TO SMOOTH INITIAL UNFILTERED BEARING ANGLES, AND TO PREDICT EMITTER LOCATIONS USING TRIANGULATION
              METHODS
                DIMENSION ALT(100), BRNG(100), BRNGD(100), D11(100), 1D12(100), D21(100), D22(100), E(100), FREQ(100), G1(100), 2G2(100), GATE(100), HDG(100), HDGD(100), JSET(30,50), LW(2), 3MODEN(100), MODET(100), MW(2), NST(100), P11(100), P12(100), 4P22(100), PIN(2,2), PIN11(100), PIN12(100), PIN22(100), 5PITCH(100), PRF(100), PW(100), Q11(100), Q12(100), Q22(100), 6QP11(100), OP12(100), OP21(100), OP22(100), P(100), RB(100), 7RDLL(100), SLA(100), SLAD(100), SLOD(100), TLOO), 8TT(100), TAT(100), TIMEN(100), TIMET(100), TDT(100), 9TDTD(100), TPTD(100), THT(100), THTTD(100), THTTD(100), THTTD(100), THTTD(100), THTTD(100), THTTD(100), THTTD(100), 1TLA(10), TLAD(10), TLOO(10), TLOO(10), VEL(100), 2VELE(100), VELN(100), WAV(100)
              SUBROUTINE TO READ EMITTER TARGET AND AIRCRAFT NAVIGATION DATA FROM CARD DATA DECK DATA SEQUENCE MUST BE OF FORMAT TGT/NAV
              NUM=65
DD 1 I=1, NUM
TYPE=11.0 FMITTER TARGET DATA
READ(5,51)TGT, TIMET(I), BRNG(I), PRF(I), MODET(I), PW(I),
              1FREQ(I)
TYPE=7.0 AIRCRAFT NAVIGATION DATA
READ(5,52) NAV, TIMEN(I), MODEN(I), ALT(I), SLA(I), SLO(I)
READ(5,56) HDG(I), VELE(I), VELN(I), ROLL('), PITCH(I)
               SUBROUTINE TO COMPUTE AIRCRAFT VELOCITY
                     VELS=VELE(I)**2+VEL**(I)**2
VEL(I)=SQRT(VELS)
CONTINUE
        1
              SUBROUTINE TO CHANGE ANGLES AND LATITUDE/LONGITUDE FROM RADIANS TO DEGREES + TENTHS OF DEGREES
                 WRITE(6,57)
DD 5 J=1, NUM
PIRAD=57.29578
HDGD(J)=HDG(J)*PIRAD
BRNGD(J)=BRNG(J)*PIRAD
SLAU(J)=SLA(J)*PIRAD
SLOU(J)=SLO(J)*PIRAD
THETA(J)=BRNG(J)+HDG(J)
IF(THETA(J).GT.6.283186) THETA(J)=THETA(J)-6.283186
THETAD(J)=BRNGD(J)+HDGO(J)
IF(THETAD(J)=BRNGD(J)+HDGO(J)
IF(THETAD(J)-360.0) THETAD(J)=THETAD(J)-360.0
WRITE(6,58)J, HUGD(J), SLAD(J), SLOD(J), BRNGD(J),
THETAD(J), THETA(J)
CONTINUE
        5
                      CONTINUE
```

```
PROGRAM TO SORT EMITTER TARGET DATA AND ESTIMATE NUMBER OF DISTINCT EMITTER TARGETS DATA IS INITIALLY SORTED BY FREQUENCY AND PRF
            NTAR=0
DD 10 I=1, NUM
IF(I.EQ.1) GD TO 19
            DO 18 KQ=1,L

M=I-KQ

DELTF=FREQ(I)-FREQ(M)

DELTPR=PRF(I)-PRF(M)

IF(ABS(DELTF).LE.4.0.AND.ABS(DELTPR).LE.4.0) GO TO 10
        CHECK FOR MULTIPLES OF PRF
             IF(ABS(DELTF).GT.4.0) GO TO 18
             DO 11 N=2,5
RK=N*PRF(M)
            RK=N*PRF(M)

SK=PRF(I)

TK=PRF(M)/FLOAT(N)

CK=ABS(RK-SK)

DK=ABS(TK-SK)

FK=N*4.0

IF(CK.LT.FK.OR.DK.LT.FK) GO TO 10

CONTINUE

CONTINUE

NTAR=NTAR+1
18
            CONTINUE

NTAR=NTAR+1

WRITE(6,59)NTAR

WRITE(6,61)NTAR

NJ=1

JSET(NTAR,NJ)=I

DO 13 J=1,NUM

IF(J.EQ.I) GO TO 13

DELF=FREQ(J)-FREQ(I)

DELPRF=PRF(J)-PRF(1)

IF(ABS(DELF).LE.4.0)

GO TO 13
                                                                   GO TO 4
        CHECK FOR SUBSEQUENT MULTIPLES OF PRF
             DO 12 N=1,5
RK=N*PRF(J)
SK=PRF(I)
TK=PRF(J)/FLOAT(N)
CK=ABS(RK-SK)
DK=ABS(TK-SK)
   4
             FK=N*4.0
IF(CK.LT.FK.OR.DK.LT.FK) GO TO 14
CONTINUE
GO TO 13
12
             NJ=NJ+1
JSET(NTAR,NJ)=J
CONTINUE
14
13
             NST(NTAR)=NJ
NSTA=NST(NTAR)
IF(NSTA.EQ.1.0) GO TO 17
        SUBROUTINE TO CHECK FOR AND ELIMINATE 360 DEGREE VARIATION IN EMITTER BEARING ANGLE THETA
             00 9 J=2,NSTA
             DO 9 J=2,NSTA

JM=J-1

KK=JSET(NTAR,JM)

K=JSET(NTAR,J)

IF(THETAD(KK).LT.10.0.AND.THETAD(K).GT.350.0) GO TO 15

IF(THETAD(KK).GT.350.0.AND.THETAD(K).LT.10.0) GO TO 16

GO TO 9

DO 6 MM=1,JM

M=JSET(NTAR,MM)

THETAD(M)=THETAD(M)+360.0

THETA(M)=THETA(M)+6.283186
15
```

```
CONTINUE
GO TO 17
DO 7 MN=J,NSTA
M=JSET(NTAR,MN)
THETAD(M)=THETAD(M)+360.0
        6
      16
                  THETA(M)=THETA(M)+6.283186
                  CONTINUE
        7
                 GO TO 17
CONTINUE
   CCCC
             PRINT UNFILTERED BEARING LINES OF SIMILAR FREQUENCY AND PRF
               DD 8 J=1,NSTA
K=JSET(NTAR,J)
WRITE(6,60)K,FREQ(K),PRF(K),PW(K),HDGD(K),BRNGD(K),
THETAD(K)
      17
 CONTINUE
                  CONTINUE
             SUBROUTINE TO COMPUTE INITIAL EMITTER TARGET JSET DATA
                 WRITE(6,62)
DO 25 I=1,NTAR
NSTA=NST(I)
WRITE(6,63)(I,J,JSET(I,J),J=1,NSTA)
CONTINUE
25
CCCCCCC
             PROGRAM TO FILTER NOISY MEASURED BEARING ANGLE THETAD(K: AND AIRCRAFT NAVIGATION DATA FOR EACH SUSPECTED EMITTER TARGET OF SIMILAR FREQUENCY AND PRF
                                                                                                                      THETAD(K)
             INITIALIZATION OF KALMAN FILTER EQUATION PARAMETERS
                 WRITE(6,50)
DO 20 I=1,NTAR
KI=JSET(I,1)
P11(KI)=10000.0
P12(KI)=0.0
P22(KI)=0.0
Q11(KI)=0.0
Q12(KI)=0.0
Q12(KI)=0.0
D11(KI)=1.0
D12(KI)=0.0
D21(KI)=0.0
G1(KI)=144.0
G1(KI)=P11(KI)/(P11(KI)+R(KI))
G2(KI)=P12(KI)/(P11(KI)+R(KI))
THTD(KI)=THETAD(KI)
THTD1(KI)=THETAD(KI)
TDTD1(KI)=VEL(KI)*SIN(BRNG(KI))*PIRAD/600000.0
TDTD1(KI)=TDTD(KI)
                  TOTOI(KI)=TOTO(KI)
                  T(K1)=0.0
K2=0
K3=1
                  WRITE(6,64)I,K3,K2,KI,G1(KI),G2(KI),T(KI)
             START OF KALMAN FILTER PREDICTION PROBLEM
                  NSTA=NST(I)

DO 21 J=2,NSTA

K=JSET(I,J)

KK=JSET(I,J-1)

TKM1=TIMET(KK)

T(K)=TIMET(K)-TKM1

TT(K)=T(K)/1000.0

Q11(K)=TT(K)**4/4.0

Q12(K)=TT(K)**3/2.0

Q22(K)=TT(K)**2

R(K)=144.0
       22
```

```
START OF KALMAN FILTER RECURSION EQUATIONS
                   P11(K)=P11(KK)*(1.0-G1(KK))+2.0*P12(KK)*T(K)-(P12(KK)*
IG1(KK)+P11(KK)*G2(KK))*T(K)+(P22(KK)-P12(KK)*G2(KK))*
2T(K)**2+Q11(K)
P12(K)=P12(KK)*(1.0-G1(KK))+(P22(KK)-P12(KK)*G2(KK))*
                   P12(K)=P12(KK)*(1.0-G1(KK))+(P22(KK)
1T(K)+Q12(K)
P22(K)=P22(KK)-P12(KK)*G2(KK)+Q22(K)
G1(K)=P11(K)/(P11(K)+R(K))
G2(K)=P12(K)/(P11(K)+R(K))
TPTD(K)=THTD(KK)+TDTD(KK)*T(K)
                      E(K)=THETAD(K)-TPTD(K)
THTD(K)=TPTD(K)+G1(K)#E(K)
TDTD(K)=TDTD(KK)+G2(K)#E(K)
                 SMOOTHING EQUATIONS FOR FIRST BEARING LINE ESTIMATE
                  PIN(1,1)=P11(K)
PIN(1,2)=P12(K)
PIN(2,1)=P12(K)
PIN(2,1)=P12(K)
PIN(2,2)=P22(K)
CALL MINV(PIN,2,DET,LW,MW)
PIN11(K)=PIN(1,1)
PIN12(K)=PIN(1,2)
PIN2(K)=PIN(1,2)
PIN22(K)=PIN(2,2)
QP11(K)=Q11(K)*PIN12(K)+Q12(K)*PIN12(K)
QP12(K)=Q11(K)*PIN12(K)+Q12(K)*PIN12(K)
QP21(K)=Q12(K)*PIN11(K)+Q22(K)*PIN12(K)
QP22(K)=Q12(K)*PIN12(K)+Q22(K)*PIN12(K)
QP22(K)=Q12(K)*PIN12(K)+Q22(K)*PIN12(K)
1QP21(K)*T(K)
D12(K)=T(K)
D12(K)*T(K)
D21(K)=T(K)
P1(K)=T(K)
D21(K)=D22(KK)*(1.0-QP11(K))-D22(KK)*QP12(K)+D21(KK)*
1QP21(K)=T(K)
TQP1(K)=T(K)
T(K)
TQP1(K)=T(K)=T(K)
                CORRELATION GATING SCHEME TO ESTIMATE WHETHER FILTERED BEARING ANGLE THTD(K) IS AN EMISSION FROM EMITTER TARGET NTAR = I OR A SPURIOUS EMISSION
                      GATE(K)=SQRT(P11(K)+F(K))/2.0
WRITE(6,65)I,J,KK,K,G1(K),G2(K),T(K),E(K),GATE(K)
IF(ABS(E(K)).LT.GATE(K)) GO TO 21
    CCC
                 THTD(K) IS A SPURIOUS BEARING LINE AND IS DISCARDED
                     NST(I)=NST(I)-1
NSTA=NST(I)
IF(NSTA.LT.J) GO TO 2
DO 27 L=J.NSTA
JSET(I,L)=JSET(I,L+1)
CONTINUE
GO TO 22
CONTINUE
       28
                                                               GO TO 20
       27
C C C C
                FILTERED BEARING ANGLE THTD(K) IS CORRELATED TO EMITTER TARGET NTAR = I
20
C
C
                       CONTINUE
                 SUBROUTINE TO COMPUTE FINAL EMITTER TARGET JSET DATA
                      WRITE(6,66)
DO 30 I=1,NTAR
NSTA=NST(I)
                       WŘÍŤE(6,63)(I,J,JSET(I,J),J=1,NSTA)
CONTINUÉ
       30
```

```
PROGRAM TO COMPUTE TARGET POSITION BASED ON A PLANE TRIANGULATION SOLUTION:
              TRIANGULATION SOLUTION

DO 35 I=1,NTAR
WRITE(6,59)I
WRITE(6,57)I
NSTA=NST(I)
KI=JSET(I,1)
KF=JSET(I,1)
WRITE(6,60)K,FREQ(K),PRF(K),PW(K),THETAD(K),THTD(K),
1THTD1(K)
CONTINUE
THT(KF)=THTD(KF)/PIRAD
TAT(I)=TAN(THT1(KF))+TAN(THT(KF))
WAV(I)=(SLA(KI)+SLA(KF))/2.0
TILA(I)=((SLO(KF)-SLC(KI))*COS(WAV(I))+SLA(KI)*
1TAN(THT1(KF))-SLA(KF)*TAN(THT(KF))/TAT(I)
TLA(I)=((SLO(KF)-SLC(KI))*COS(TILA(I))+SLA(KI)*
1TAN(THT1(KF))-SLA(KF)*TAN(THT(KF))/TAT(I)
TLAD(I)=TLA(I)*PIRAD
TLO(I)=SLO(KF)+(TLA(I)-SLA(KF))*TAN(THT(KF))/
1COS(TILA(I))
TLOD(I)=TLO(I)*PIRAD
WRITE(6,68)THTDI(KF),THTD(KF)
WRITE(6,70) TLAD(I),TLOD(I)
CONTINUE
FORMAT(*1*.14X.*LISTING OF KALMAN FILTER PARAMETERS*
             35
50
54
55
56
57
58
59
62
63
64
65
66
68
70
                2//, TOP
```

```
**********
               *
                              MONTE CARLO SIMULATION RUN OF KALMAN FILTER PROGRAM
                *******************************
                   THIS FORTRAN PROGRAM IS DESIGNED TO ANALYZE AND SORT ELECTRONIC EMITTER PARAMETERS AND AIRCRAFT NAVIGATION DATA, TO FILTER EMITTER DATA USING KALMAN FILTER TECHNIQUES TO MINIMIZE BEARING ANGLE-OF-ARRIVAL MEASUREMENT NOISE, TO SMOOTH INITIAL UNFILTERED BEARING ANGLES, AND TO PREDICT EMITTER LOCATIONS USING TRIANGULATION
                        DIMENSION ALT(100), BRNG(100), BRNGD(100), D11(100), 1D12(100), D21(100), D22(100), E(100), FREQ(100), G1(100), 2G2(100), GATE(100), HDG(100), HDGD(100), JSET(5,40), LW(2), 3MODEN(100), MDDET(100), MN(2), NST(100), P11(100), P12(100), 4P22(100), PIN(2,2), P1P11(100), P1N12(100), P1N2(100), P1N2(100), G22(100), G22(
                    SUBROUTINE TO READ EMITTER TARGET AND AIRCRAFT NAVIGATION DATA FROM CARD DATA DECK DATA SEQUENCE MUST BE OF FORMAT TGT/NAV
                               NUM=61
IX=1257
STDEV=1.0
                    WRITE(6,48)
DD 1 I=1,NUM
READ EMITTER TARGET DATA
READ(5,51)TGT,TIMET(1),BRNGD(1),PRF(1),PW(1),FREQ(1)
READ AILCRAFT NAVIGATION DATA
READ(5,51)HDGD(1),SLAD(1),SLOD(1),ALT(1),VELN(1),
1VELE(1)
                                AVE=0.0
 C
 C
                    GENERATE RANDOM NOISE TO ADD TO KNOWN EMITTER BEARING ANGLES OF ARRIVAL
                         CALL GAUSS(IX,STDEV,AVE,V)
THETAD(I)=BRNGD(I)+V
WRITE(6,49)1,TIMET(I),FREQ(I),PRF(I),PW(I),BRNGD(I),
1V,THETAD(I),HDGD(I),SLAD(I),SLOD(I),ALT(I),VELN(I),
2VELE(I)
           2
 S
                      SUBROUTINE TO COMPUTE AIRCRAFT VELOCITY
                               VELS=VELE(I)**2+VELN(I)**2
VEL(I)=SQRT(VELS)
                                CONT I NUE
           1
                      SUBROUTINE TO CHANGE ANGLES FROM DEGREES TO RADIANS
                              WRITE(6,57)
DO 5 J=1,NUM
PIRAD=57.29578
HDG(J)=HDGD(J)/PIRAD
BRNG(J)=BRNGD(J)/PIRAD
SLA(J)=SLAD(J)/PIRAD
SLO(J)=SLOD(J)/PIRAD
THETA(J)=THETAD(J)/PIRAD
RB(J)=BRNG(J)-HDG(J)
IF(THETA(J).GT.6.283186) THETA(J)=THETA(J)-6.283186
```

```
IF(THETAD(J).GT.360.0) THETAD(J)=THETAD(J)-360.0
WRITE(6,58)J,HDGD(J),SLAD(J),SLOD(J),BRNGD(J),
THETAD(J),THETA(J)
CONTINUE
      PROGRAM TO SORT EMITTER TARGET DATA AND ESTIMATE NUMBER OF DISTINCT EMITTER TARGETS DATA IS INITIALLY SORTED BY FREQUENCY AND PRE
            NTAR=0
           DO 10 I=1, NUM
IF(I.EQ.1) GO TO 19
L=I-1
            DO 18 KQ=1,L
           M=I-KQ
DELTF=FREQ(I)-FREQ(M)
DELTPR=PRF(I)-PRF(M)
            IF(ABS(DELTF).LE.4.0.AND.ABS(DELTPR).LE.4.0) GO TO 10
      CHECK FOR MULTIPLES OF PRF
            IF(ABS(DELTF).GT.4.0) GO TO 18
           DO 11 N=2,5
RK=N*PRF(M)
          RK=N*PRF(M)

SK=PRF(I)

TK=PRF(M)/FLOAT(N)

CK=ABS(RK-SK)

DK=ABS(TK-SK)

FK=N*4.0

IF(CK.LT.FK.OR.DK.LT.FK) GO TO 10

CONTINUE

CONTINUE

CONTINUE

NTAR=NTAR+1

WRITE(6,59)NTAR

WRITE(6,61)NTAR

NJ=1

JSET(NTAR,NJ)=1

DO 13 J=1,NUM

IF(J.EQ.I) GO TO 13

DELF=FREQ(J)-FREQ(I)

DELPRF=PRF(J)-PRF(I)

IF(ABS(DELF).LE.4.0) GO TO 4

GO TO 13
11
18
19
       CHECK FOR SUBSEQUENT MULTIPLES OF PRF
           DO 12 N=1,5
RK=N*PRF(J)
SK=PRF(I)
TK=PRF(J)/FLOAT(N)
CK=ABS(RK-SK)
DK=ABS(TK-SK)
  4
           FK=N*4.0
IF(CK.LT.FK.OR.DK.LT.FK) GO TO 14
CONTINUE
GO TO 13
12
           NJ=NJ+1
JSET(NTAR,NJ)=J
14
           CONTINUE
NST(NTAR)=NJ
NSTA=NST(NTAR)
IF(NSTA.EQ.1.0) GO TO 17
13
       SUBROUTINE TO CHECK FOR AND ELIMINATE 360 DEGREE VARIATION IN EMITTER BEARING ANGLE THETA
            DO 9 J=2,NSTA
           JM=J-1
KK=JSET(NTAR,JM)
K=JSET(NTAR,J)
IF(THETAD(KK).LT
            IF (THETAD (KK) LT .10.0 .4ND .THETAD(K) .GT .350.0) IF (THETAD (K) .GT .353.0 .AND .THETAD (K) .LT .13.0)
```

```
GO TO 9
DO 6 MM=1, JM
M=JSET(NTAR, MM)
THETAD(M)=THETAD(M)+360.0
THETA(M)=THETA(M)+6.283186
CONTINUE
GO TO 17
DO 7 MN=J, NSTA
M=JSET(NTAR, MN)
THETAD(M)=THETAD(M)+360.0
THETA(M)=THETAD(M)+6.283186
CONTINUE
GO TO 17
CONTINUE
         15
            6
         16
            7
                 PRINT UNFILTERED BEARING LINES OF SIMILAR FREQUENCY AND PRF
                    DO 8 J=1,NSTA
K=JSET(NTAR,J)
WRITE(6,60)K,FREQ(K),PRF(K),PW(K),HDGD(K),BRNGD(K),
1THETAD(K)
CONTINUE
CONTINUE
        17
  SUBROUTINE TO COMPUTE INITIAL EMITTER TARGET JSET DATA
                       WRITE(6,62)
DO 25 I=1,NTAR
NSTA=NST(I)
WRITE(6,63)(I,J,JSET(I,J),J=1,NSTA)
CONTINUE
        25
                  SUBROUTINE TO CHECK MONTE CARLO SIMULATION ACCURACY IN THE ABSENCE OF RANDOM NOISE CONTAMINATION OF EMITTER
                  BEARING ANGLES
                       IF(V.EQ.O.O) GO TO 23
GD TO 24
WRITE(6,46)
DO 26 I=1,NTAR
NSTA=NST(I)
KI=JSET(I,1)
KF=JSET(I,NSTA)
THTD1(KF)=BRNGD(KI)
THTD(KF)=BRNGD(KF)
CONTINUE
GO TO 29
        23
        26
CCCCCCC 24
                 PROGRAM TO FILTER NOISY MEASURED BEARING ANGLE THETAD(K) AND AIRCRAFT NAVIGATION DATA FOR EACH SUSPECTED EMITTER TARGET OF SIMILAR FREQUENCY AND PRE
                  INITIALIZATION OF KALMAN FILTER EQUATION PARAMETERS
                      WRITE(6,50)
DO 20 I=1,NTAR
KI=JSET(I,1)
P11(KI)=100.0
P12(KI)=0.0
P22(KI)=0.0
Q11(KI)=0.0
Q12(KI)=0.0
D11(KI)=0.0
D11(KI)=1.0
D12(KI)=0.0
D22(KI)=1.0
R(KI)=1.0
R(KI)=1.0
G1(KI)=1.0
```

```
THTD1(KI)=THETAD(KI)
TDTD(KI)=VEL(KI)*SIN(RB(KI))/500000.0
TDTD1(KI)=TDTD(KI)
T(KI)=0.0
K2=0
K3=1
HBITE(4 44)I K3 K3 KI CL(KI) C2(KI) T
                       WRITE(6,64)I,K3,K2,KI,G1(KI),G2(KI),T(KI)
               START OF KALMAN FILTER PREDICTION PROBLEM
                     NSTA=NST(I)
DO 21 J=2,NSTA
K=JSET(I,J)
KK=JSET(I,J-1)
TKM1=TIMET(K)-TKM1
TT(K)=TIMET(K)-TKM1
TT(K)=TT(K)/1000.0
Q11(K)=TT(K)**4/4.0
Q12(K)=TT(K)**3/2.0
Q22(K)=TT(K)**2
R(K)=1.0
   22
               START OF KALMAN FILTER RECURSION EQUATIONS
                  P11(K)=P11(KK)*(1.0-G1(KK))+2.0*P12(KK)*T(K)-(P12(KK)*
1G1(KK)+P11(KK)*G2(KK))*T(K)+(P22(KK)-P12(KK)*G2(KK))*
2T(K)**2+Q11(K)
P12(K)=P12(KK)*(1.0-G1(KK))+(P22(KK)-P12(KK)*G2(KK))*
1T(K)+Q12(K)
P22(K)=P22(KK)-P12(KK)*G2(KK)+Q22(K)
G1(K)=P11(K)/(P11(K)+R(K))
G2(K)=P12(K)/(P11(K)+R(K))
TPTD(K)=THTD(KK)+TDTD(KK)*T(K)
E(K)=THETAD(K)-TPTD(K)
THTD(K)=TPTD(K)+G1(K)*E(K)
TDTD(K)=TDTD(KK)+G2(K)*E(K)
               SMOOTHING EQUATIONS FOR FIRST BEARING LINE ESTIMATE
                 PIN(1,1)=P11(K)
PIN(1,2)=P12(K)
PIN(2,1)=P12(K)
PIN(2,1)=P12(K)
PIN(2,2)=P22(K)

CALL MINV(PIN,2,DET,1W,MW)
PIN11(K)=PIN(1,1)
PIN12(K)=PIN(1,2)
PIN2(K)=PIN(2,2)
QP11(K)=Q11(K)*PIN11(K)+Q12(K)*PIN12(K)
QP12(K)=Q11(K)*PIN11(K)+Q22(K)*PIN12(K)
QP21(K)=Q12(K)*PIN11(K)+Q22(K)*PIN12(K)
QP22(K)=Q12(K)*PIN11(K)+Q22(K)*PIN12(K)
QP22(K)=Q12(K)*PIN12(K)+Q22(K)*PIN22(K)
D11(K)=D11(KK)*(1.0-QP11(K))-D12(KK)*QP12(K)+D11(KK)*
1QP21(K)+T(K)
D12(K)=D12(KK)*(1.0-QP11(K))-D22(KK)*QP12(K)+D11(KK)*
1QP22(K)-1.0)*T(K)
D21(K)=D22(KK)*(1.0-QP11(K))-D22(KK)*QP12(K)+D21(KK)*
1QP21(K)+T(K)
D22(K)=D22(KK)*(1.0-QP11(K))-D21(KK)*QP12(K)+D21(KK)*
1QP22(K)-1.0)*T(K)
THID1(K)=THID1(KK)+(D11(K)*G1(K)+D12(K)*G2(K))*E(K)
TOTD1(K)=TDTD1(KK)+(D21(K)*G1(K)+D22(K)*G2(K))*E(K)
               CORRELATION GATING SCHEME TO ESTIMATE WHETHER FILTERED BEARING ANGLE THIO(K) IS AN EMISSION FROM EMITTER TARGET NTAR = I OR A SPURIOUS EMISSION
                       GATE(K)=3.0*SQRT(P11(K)+R(K))
WRITE(6.65)I,J,KK,K,G1(K),G2(K),T(K),E(K),GATE(K)
IF(ABS(E(K)).LT.GATE(K)) GO TO 21
CCC
               THIO(K) IS A SPURIOUS REARING LINE AND IS DISCARDED
```

```
NST(I)=NST(I)-1
NSTA=NST(I)
IF(NSTA.LT.J) GO TO 2
DO 27 L=J.NSTA
JSET(I,L)=JSET(I,L+1)
CONTINUE
     28
                                                                                                GD TO 20
     27
                               GO TO 22
 C
      21
                               CONTINUE
 CCCC
                     FILTERED BEARING ANGLE THTD(K) IS CORRELATED TO EMITTER
                     TARGET NTAR = I
      20
                              CONTINUE
 CCC
                     SUBROUTINE TO COMPUTE FINAL EMITTER TARGET J'SET DATA
                              WRITE(6,66)
DO 30 I=1,NTAR
NSTA=NST(I)
                              WRITE(6,63)(I,J,JSET(I,J),J=1,NSTA)
CONTINUE
C 29
                     PROGRAM TO COMPUTE TARGET POSITION BASED ON A PLANE TRIANGULATION SOLUTION
                       TRIANGULATION SOLUTION

DO 35 I=1,NTAR
WRITE(6,59)I
WRITE(6,67)I
NSTA=NST(I)
KI=JSET(I,1)
KF=JSET(I,NSTA)
DO 34 J=1,NSTA
K=JSET(I,J)
CONTINUE
THTO(K)
CONTINUE
THTO(K)
THTO(K)
TAT(I)=THTO(KF)/PIRAD
TAT(I)=THTO(KF)/PIRAD
TAT(I)=TAN(THTI(KF))-TAN(THT(KF))
WAV(I)=(SLA(K1)+SLA(KF))/2.0
TILA(I)=((SLC(KF)-SLO(KI))*COS(WAV(I))+SLA(KI)*
1TAN(THTI(KF))-SLA(KF)*TAN(THT(KF)))/TAT(I)
TLA(I)=((SLO(KF)-SLO(KI))*COS(TILA(I))+SLA(KI)*
1TAN(THTI(KF))-SLA(KF)*TAN(THT(KF))/TAT(I)
TLAO(I)=TLA(I)*PIRAD
TLO(I)=TLA(I)*PIRAD
WRITE(6,68)THTOI(KF),THTO(KF)
WRITE(6,68)THTOI(KF),THTO(KF)
WRITE(6,70) TLAO(I),TLOO(I)
CONTINUE
      34
C
35
                               CONTINUE
                        FORMAT('O THIS MONTE CARLO SIMULATION HAS NO EMITTER',

1' BEARING ANGLE MEASUREMENT NOISE', //, 'EXECUTE A',

2'CALIBRATION CHECK FOR SIMULATION ACCURACY IN ABSENCE'

3' OF RANDOM NOISE')
FORMAT('1', 36X, 'LISTING OF EMITTER TARGET DATA AND ',

1'AIRCRAFT NAVIGATION DATA', //, 12X, 'TARGET PARAMETERS',

2,37X, 'AIRCRAFT PARAMETERS', //, 'JSET TIMET FREQ

3'PRF PW BRNGD V THETAD HOGD ',

4'SLAD SLOD ALT VELN VELE', //)
FORMAT('', 13, 2X, F5.1, 2X, F6.1, 2X, F5.1, 2X, F4.2, 2X, F9.5,

12X, F8.5, 2X, F9.5, 2X, F5.1, 2X, F8.5, 2X, F10.5, 2X, 3F7.1)
FORMAT('', 14X, 'LISTING OF KALMAN FILTER PARAMETERS',

1//, 'IJKK G1(K) G2(K) T(K)

2'E(K)', 6X, 'GATE(K)', /)
FORMAT('1 LISTING OF BFARING ANGLES-OF-ARRIVAL AND ',

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)

1'AIRCRAFT NAVIGATION DATA', //, 'N HOGD(N)
      46
      48
       49
       50
```

```
*********************
COCCOCCOCC
                                         EXTENDED KALMAN FILTER COMPUTER SUBROUTINE
            ******************
               EMITTER LOCATION SOLUTION USING EXTENDED KALMAN FILTERING TECHNIQUES EXTENDED KALMAN FILTER INITIALIZATION
                  DO 20 I=1,NTAR

KI=JSET(I,1)

P11(KI)=2.0

P12(KI)=0.5

P22(KI)=1.0

XID(KI)=-117.42503

YTD(KI)=33.24187

XT(KI)=XTD(KI)/PIRAD

YT(KI)=YTD(KI)/PIRAD

AT1=(XT(KI)-SLU(KI))*COS(YT(KI))

AT2=YT(KI)-SLU(KI)-TX(KI)

ER(KI)=THETA(KI)-TX(KI)

DM=(YTD(KI)-SLAD(KI))**2+(XTD(KI)-SLOD(KI))**2

1*(COS(Y'(KI)))**2

DMX(KI)=(YTD(KI)-SLAD(KI))*COS(YT(KI))/DM

DMY(KI)=-(XTD(KI)-SLOD(KI))*((YTD(KI)-SLAD(KI))*

1SIN(YT(KI))+COS(YT(KI)))/DM

DG=P11(KI)*DMX(KI)**2+2.0*P12(KI)*DMX(KI)*DMY(KI)

1+P22(KI)*DMY(KI)**2+R(KI)

GX(KI)=(P11(KI)*DMX(KI)+P12(KI)*DMY(KI))/DG

GY(KI)=(P12(KI)*DMX(KI)+P22(KI)*DMY(KI))/DG
                 START OF KALMAN FILTER PREDICTION PROBLEM
                        NSTA=NST(I)

DO 21 J=2,NSTA

K=JSET(I,J)

KK=JSET(I,J-1)

Q11(K)=0.10

Q12(K)=0.0

Q22(K)=0.10

R(K)=1.0
     22
                 START OF EXTENDED KALMAN FILTER RECURSION EQUATIONS
                   P11(K)=P11(KK)*(1.0-GX(KK)*DMX(KK))-P12(KK)*GX(KK)*

1DMY(KK)+Q11(K)

P12(K)=P12(KK)*(1.0-GX(KK)*DMX(KK))-P22(KE)*GX(KK)*

1DMY(KK)+Q12(K)

P22(K)=P22(KK)*(1.0-GY(KK)*DMY(KK))-P12(KK)*GY(KK)*

1DMX(KK)+Q12(K)

DM=(YTD(KK)-SLAD(K))**2+(XTD(KK)-SLDD(K))**2

1*(COS(YT(KK)))**2

DMX(K)=(YTD(KK)-SLAD(K))*COS(YT(KK))/DM

DMY(K)=-(XTD(KK)-SLDD(K))*(YTD(KK)-SLAD(K))*

1SIN(YT(KK))+CDS(YT(KK)))/DM

DG=P11(K)*DMX(K)**2+2.0*P12(K)*DMX(K)*DMY(K)+

1P22(K)*DMY(K)**2+R(K)

GX(K)=(P11(K)*DMX(K)+P12(K)*DMY(K))/DG

AT1=(XT(KK)-SLO(K))*COS(YT(KK))

AT2=YT(KK)-SLO(K))*COS(YT(KK))

XT(K)=ATAN2(AT1,AT2)

ER(K)=THETA(K)-TX(K)

XT(K)=XTD(K)+GY(K)*ER(K)

YTD(K)=YTD(KK)+GY(K)*ER(K)

YT(K)=YTD(K)/PIRAD

CONTINUE

CONTINUE
                         CONTINUE
```

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