



Fortran programs for aircraft
parameter identification using the
estimation-before-modelling technique

J.C.Hoff

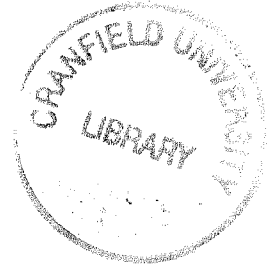
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NOTATION

$a_{x_m}, a_{y_m}, a_{z_m}$	Measured accelerations on axes x, y and z respectively.
b_i	Bias terms of the observation model.
$F(t)$	Gradient of the state matrix.
g	Acceleration of gravity.
h, h_m	Height and its measured value, respectively.
$H_{k/k-1}$	Gradient of the observation matrix calculated for $\hat{x}_{k/k-1}$.
$H_{i/i}$	Gradient of the observation matrix calculated for $\hat{x}_{i/i}$.
I	Unity matrix
I_x, I_y, I_z	Moment of inertia referred to axes x, y and z respectively.
I_{xz}, I_{yz}	Product of inertia xz and yz , respectively.
K_k	Matrix of Kalman filter gains.
L	Normalised roll moment
M	Normalised pitch moment.
N	Normalised yaw moment
$P_{k-1/k-1}$	Filter covariance matrix at instant (sample) $k-1$.
$P_{k/k-1}$	Filter covariance matrix propagated from instant $k-1$ to k .
$P_{k/k}$	Filter covariance matrix updated at instant k .
Q	Process noise covariance matrices.
R	Measurement noise covariance matrix.
u, v, w	Velocity components on axes x, y and z , respectively.
p, q, r	Roll, pitch and yaw body rates.
p_m, q_m, r_m	Roll, pitch and yaw measured body rates.
V, V_m	True and measured true airspeed.
$x(t)$	State vector whose terms are the states of the dynamic model.
$\hat{x}_{k/k}$	Estimated value of state x at instant (or sample) k .
$\hat{x}_{k/k-1}$	Propagated value of state x from instant (sample) $k-1$ to instant k .
$x_{k-1/k}$	Estimated of state x from instant (or sample) k back to instant $k-1$.
x_1, y_1, z_1	Body axes coordinates of accelerometers package - relative to cg .
x_2, y_2, z_2	Body axes coordinates of pitot probe - relative to cg .
x_3, y_3, z_3	Body axes coordinates of incidence vane - relative to cg .
x_4, y_4, z_4	Body axes coordinates of sideslip vane - relative to cg .
α, α_m	Incidence angle and measured incidence angle, respectively.
β, β_m	Sideslip angle and measured sideslip angle, respectively.
ϕ, ϕ_m	Attitude roll angle and its measured value
θ, θ_m	Attitude pitch angle and its measured value.
Δt	Time interval between samples.

1. INTRODUCTION

This report describes five Fortran programs formulated to be used in the Estimation-Before-Modelling (E-B-M) methodology for aircraft parameter estimation.

The E-B-M technique is a two step estimation process. In the first step, as formulated in this report, the aircraft states are estimated by using Extended Kalman Filter techniques. In the second step the unknowns aerodynamic derivatives are estimated by linear regression formulated as the Stepwise Regression [1].

The five program comprise two different techniques for state estimation, two algorithms for linear regression and a fixed-lag smoother-differentiator.

The programs are identified as;

- (i) IEKF.FOR.
- (ii) EKFMBF.FOR
- (iii) EKFDER.FOR
- (iv) MSR.FOR
- (v) MSR.H.FOR

(i) IEKF.FOR - is a simplified Iterated Extended Kalman Filter (IEKF) for the estimation of aircraft states.

(ii) EKFMBF.FOR - is an Extended Kalman Filter (EKF) associated with the Modified Bryson-Frazier (MBF) smoother and is used for aircraft state estimation.

(iii) EKFDER.FOR - is a Fixed Lag (FL) Smoother-Differentiator formulated specifically to smooth and differentiate the output of the program IEKF.FOR.

(iv) MSR.FOR - is the Stepwise Regression program with the linear regression formulated by the normal equation solution approach.

(v) MSR.H.FOR - is the Stepwise Regression program with the linear regression solution formulated by the Householder Transformation approach.

The state estimation algorithms are formulated in terms of inertial and gravitational models, therefore independent of definition of aerodynamic models. The force components X, Y, Z and the moment components L, M and N are modelled as second order Gauss-Markov.

The programs have been formulated in Microsoft Fortran 5.1 and operate in DOS.

2. METHODOLOGY

2.1 State Estimation

The programs follow the methodology described in chapter 3 of reference [2], i.e., Extended Kalman Filter (EKF) methodology. The ordinary EKF is used associated with the MBF smoother and the Iterated Extended Kalman Filter, in its simplified version (i.e., iteration in the measurement model only), is used associated with the Fixed-Lag smoother.

In formulating the EKF and IEKF algorithms the following models have been used:

(i) The dynamic model ($\dot{\mathbf{x}} = f[\mathbf{x}, t]$) is composed by;

$$\begin{aligned}\dot{u} &= rv - qw - g \sin \theta + X_1 \\ \dot{v} &= pw - ru + g \cos \theta \sin \phi + Y_1 \\ \dot{w} &= qu - pv + g \cos \theta \cos \phi + Z_1 \\ \dot{p} &= pqC_{11} + qrC_{12} + qC_{13} + L_1 + N_1C_{14} \\ \dot{q} &= prC_{21} + (r^2 - p^2)C_{22} - rC_{23} + M_1 \\ \dot{r} &= pqC_{31} + qrC_{32} + qC_{33} + L_1C_{34} + N_1 \\ \dot{\theta} &= q \cos \phi - r \sin \phi \\ \dot{\phi} &= p + q \tan \theta \sin \phi + r \tan \theta \cos \phi \\ \dot{h} &= u \sin \theta - v \cos \theta \sin \phi - w \cos \theta \cos \phi\end{aligned}$$

with

$$\begin{aligned}C_{11} &= [I_{XZ}(I_{ZZ} + I_{XX} - I_{YY})] / I^2 & C_{31} &= [I_{XX}(I_{XX} - I_{YY}) + I_{XZ}^2] / I^2 \\ C_{12} &= [I_{ZZ}(I_{YY} - I_{ZZ}) - I_{XZ}^2] / I^2 & C_{32} &= I_{XZ}(I_{YY} - I_{ZZ} - I_{XX}) / I^2 \\ C_{13} &= I_{XZ}I_{EX} / I^2 & C_{33} &= I_{XX}I_{EX} / I^2 \\ C_{14} &= I_{XZ} / I_{XX} & C_{34} &= I_{XZ} / I_{ZZ} \\ C_{21} &= (I_{ZZ} - I_{XX}) / I_{YY} & & \\ C_{22} &= I_{XZ} / I_{YY} & \text{where} & \\ C_{23} &= I_{EX} / I_y & I^2 &= (I_{XX}I_{ZZ} - I_{XZ}^2)\end{aligned}$$

$X_1, Y_1, Z_1, \dots, N_1$ are modelled as 2nd order Gauss-Markov, for example,

$$\begin{aligned}\dot{X}_1 &= X_2 + w_1 \\ \dot{X}_2 &= X_3 + w_2, & \text{where } w_i & \text{ are noise terms.} \\ \dot{X}_3 &= 0 + w_3\end{aligned}$$

(ii) The measurement model ($\mathbf{z} = f_h[\mathbf{x}, t]$) is composed by;

$$\begin{aligned}
\alpha_{x_m} &= \frac{X}{m} + \frac{T}{m} - g \sin \theta - (r^2 + q^2)x_1 + (pq - \dot{r})y_1 + (pr + \dot{q})z_1 + b_{\alpha_x} \\
\alpha_{y_m} &= \frac{Y}{m} + g \cos \theta \sin \phi - (p^2 + r^2)y_1 + (pq + \dot{r})x_1 + (qr - \dot{p})z_1 + b_{\alpha_y} \\
\alpha_{z_m} &= \frac{Z}{m} + g \cos \theta \cos \phi - (p^2 + q^2)z_1 + (pr - \dot{q})x_1 + (qr + \dot{p})y_1 + b_{\alpha_z} \\
p_m &= p \\
q_m &= q + b_q \\
r_m &= r \\
V_m &= \sqrt{u_1^2 + v_1^2 + w_1^2} \\
\alpha_m &= \tan^{-1} \left[\frac{w}{u} \right] - \frac{x_3 \cdot q}{V} + b_{\alpha} \\
\beta_m &= \sin^{-1} \left[\frac{v}{V} \right] - \frac{x_4 \cdot r}{V} + b_{\beta} \\
\theta_m &= \theta + b_{\theta} \\
\phi_m &= \phi \\
h_m &= h
\end{aligned}
\quad \text{where} \quad
\begin{bmatrix} u_1 \\ v_1 \\ w_1 \end{bmatrix} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} + \begin{bmatrix} 0 & -r & q \\ r & 0 & -p \\ -q & p & 0 \end{bmatrix} \begin{bmatrix} x_2 \\ y_2 \\ z_2 \end{bmatrix}$$

The bias terms (b_i) are applied only to the terms most significant for the longitudinal analysis (exception of the sideslip).

(iii) The Kalman Filter Algorithms;

The Iterated Extended Kalman Filter is formulated as follows;

Given the state vector $\mathbf{x}_{k-1/k-1}$ the matrix \mathbf{F} is calculated by,

$$\mathbf{F}(t) = \left. \frac{\partial \mathbf{f}}{\partial \mathbf{x}} \right|_{\mathbf{x}_{k-1/k-1}} \quad (1)$$

$\mathbf{F}(t)$ corresponds to the derivatives of the dynamic model (f) w.r.t the states.

The covariance matrix \mathbf{P} is propagated from the instant t_{k-1} to t_k by;

$$\mathbf{P}_{k/k-1} = \Phi \mathbf{P}_{k-1/k-1} \Phi^T + \mathbf{Q}$$

where:

$$\Phi = \mathbf{I} + \mathbf{F}(t)\Delta t + \frac{1}{2}(\mathbf{F}(t)\Delta t)^2 \quad ,$$

\mathbf{Q} is the process noise covariance matrix and Δt is the sampling interval.

The state vector \mathbf{x} is propagated from t_{k-1} to t_k by numerical integration (4th order Runge-Kutta).

$$\hat{\mathbf{x}}_{k/k-1} = \int_{t_{k-1}}^{t_k} f(\mathbf{x}(t), t) dt \quad (2)$$

The *Kalman gain matrix* is calculated as follows:

$$\mathbf{K}_k = \mathbf{P}_{k/k-1} \mathbf{H}_{k/k-1}^T [\mathbf{H}_{k/k-1} \mathbf{P}_{k/k-1} \mathbf{H}_{k/k-1}^T + \mathbf{R}]^{-1} \quad \text{with} \quad \mathbf{H}_{k/k-1} = \left. \frac{\partial f_h}{\partial \mathbf{x}} \right|_{\hat{\mathbf{x}}_{k/k-1}} \quad (3)$$

\mathbf{H} are the derivatives of the measurement model (f_h) w.r.t the states and \mathbf{R} is the measurement noise covariance matrix.

- *State Update*;

At time t_k the updated estimate of the state vector is calculated by adding the measurement residual, appropriately weighted by the Kalman gain matrix, to the propagated state vector calculated by equation 2.

$$\hat{\mathbf{x}}_{i/i}^j = \hat{\mathbf{x}}_{i/i-1} + \mathbf{K}_i^j [z_i - f_h(\hat{\mathbf{x}}_{i/i}^{j-1}) - \mathbf{H}_{i/i}(\hat{\mathbf{x}}_{i/i}^{j-1})[\hat{\mathbf{x}}_{i/i-1} - \hat{\mathbf{x}}_{i/i}^{j-1}]] \quad j=1,2 \quad (4)$$

The matrix \mathbf{H} is then recalculated for the new vector \mathbf{x} and a new Kalman Gain is calculated. The vector \mathbf{x} is updated again. Only two iterations are performed.

For the ordinary EKF the states are update by the following expression, replacing equation 4 above, and no iterations are performed updating the matrix \mathbf{H} and the Kalman Gain;

$$\mathbf{x}_{k/k} = \mathbf{x}_{k/k-1} + \mathbf{K}_k (z_j - \mathbf{h}_j)$$

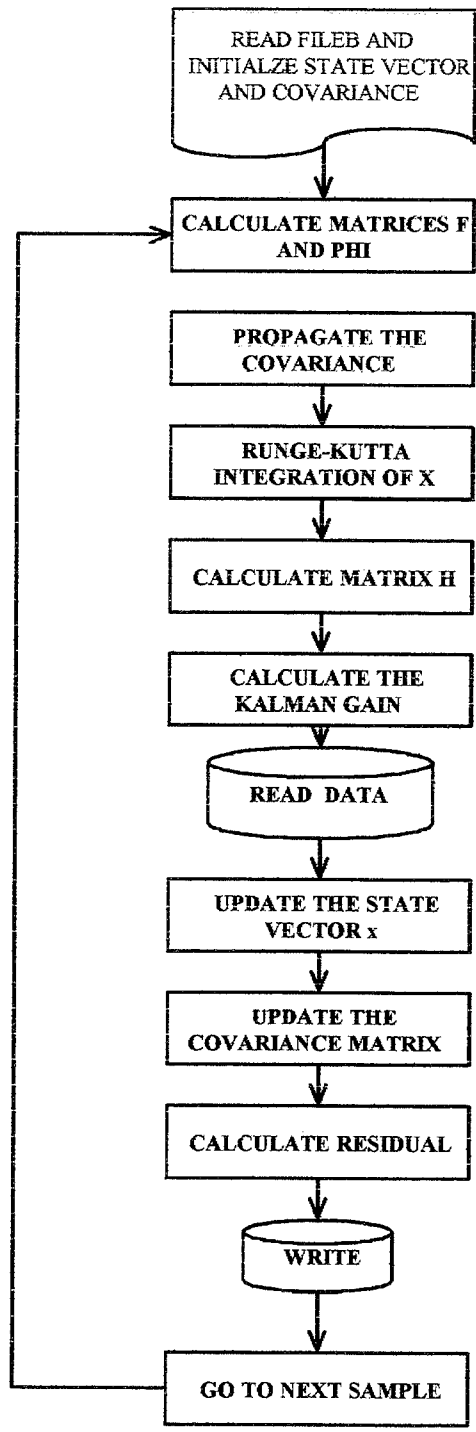
where z_j are the measurements and \mathbf{h} the measurement models estimated for x_{k-1} .

- *Covariance Matrix Update*,

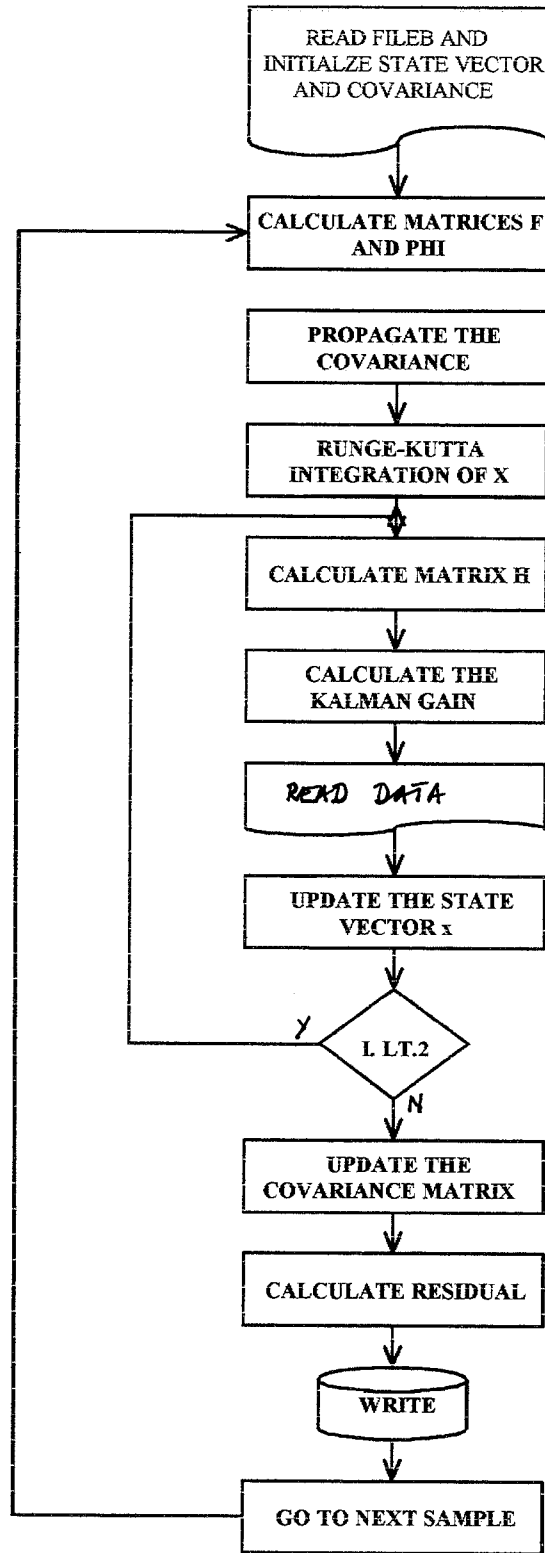
The covariance matrix is updated using the latest calculated Kalman gain and \mathbf{H} matrix,

$$\mathbf{P}_{k/k} = [\mathbf{I} - \mathbf{K}_k \mathbf{H}_{k/k-1}] \mathbf{P}_{k/k-1}$$

The flow diagrams below show the EKF and IEKF mechanization,



Flow diagram 1. Extended Kalman Filter mechanization



Flow diagram 2. Iterated Extended Kalman Filter mechanization

(iv) Modified Bryson-Frazier Smoother

The computational steps involved in its execution are presented below and follows Bierman [3],

- *Initialization:*

The adjoint variables, vector $\hat{\lambda}$ and matrix Λ , are initialized at $t = t_m$, last iteration of the EKF by,

$$\begin{aligned}\hat{\lambda}_{m/m} &= -\mathbf{H}_{m/m-1}^T \mathbf{D}_{m/m-1}^{-1} \mathbf{r}_m \delta_{t_m, t_0} \\ \Lambda_{m/m} &= \mathbf{H}_{m/m-1}^T \mathbf{D}_{m/m-1}^{-1} \mathbf{H}_{m/m-1} \delta_{t_m, t_0}\end{aligned}$$

where:

$$\mathbf{D}_{m/m-1} = [\mathbf{H}_{m/m-1} \mathbf{P}_{m/m-1} \mathbf{H}_{m/m-1}^T + \mathbf{R}]$$

\mathbf{r}_m is the vector of residuals generated by the EKF, $\mathbf{H}_{m/m-1}$ is calculated by equation 3 and δ denotes the Kronecker delta function. $\delta = 0$ if t_m is not an observation time.

- *Adjoint Variables Propagation:*

The adjoint variables are evaluated at time t_k by backward integration of the following equation from time t_{k+1} ,

$$\begin{aligned}\dot{\hat{\lambda}} &= -\mathbf{F}^T \hat{\lambda}_{k+1/k+1} \\ \dot{\Lambda} &= -(\mathbf{F}^T \Lambda_{k+1/k+1}) - (\mathbf{F}^T \Lambda_{k+1/k+1})^T\end{aligned}\quad (5)$$

Here \mathbf{F} is given by equation 1, however it is calculated for $\hat{x}_{k+1/k+1}$. The numerical integration of equations 5 produce $\hat{\lambda}_{k/k+1}$ and $\Lambda_{k/k+1}$,

$$\begin{aligned}\hat{\lambda}_{k/k+1} &= \hat{\lambda}_{k+1/k+1} + \int_{t_{k+1}}^{t_k} \dot{\hat{\lambda}} dt \\ \Lambda_{k/k+1} &= \Lambda_{k+1/k+1} + \int_{t_{k+1}}^{t_k} \dot{\Lambda} dt\end{aligned}$$

- *Adjoint variables matrices update*

The matrices of adjoint variables are updated at time t_k by evaluation of the following equations,

$$\begin{aligned}\hat{\lambda}_{k/k} &= \hat{\lambda}_{k+1/k+1} - \mathbf{H}_{k/k-1}^T \mathbf{D}_{k/k-1}^{-1} [\mathbf{r}_k + \mathbf{D}_{k/k-1} \mathbf{K}_k^T \hat{\lambda}_{k+1/k+1}] \\ \Lambda_{k/k} &= [\mathbf{I} - \mathbf{K}_k \mathbf{H}_{k/k-1}]^T \Lambda_{k+1/k+1} [\mathbf{I} - \mathbf{K}_k \mathbf{H}_{k/k-1}] + \mathbf{H}_{k/k-1}^T \mathbf{D}_{k/k-1}^{-1} \mathbf{H}_{k/k-1}\end{aligned}$$

- *State vector smoothing*

The vector of smoothed state estimates is obtained by correcting the EKF filter state estimates. The vector of smoothed state variable estimates is given by,

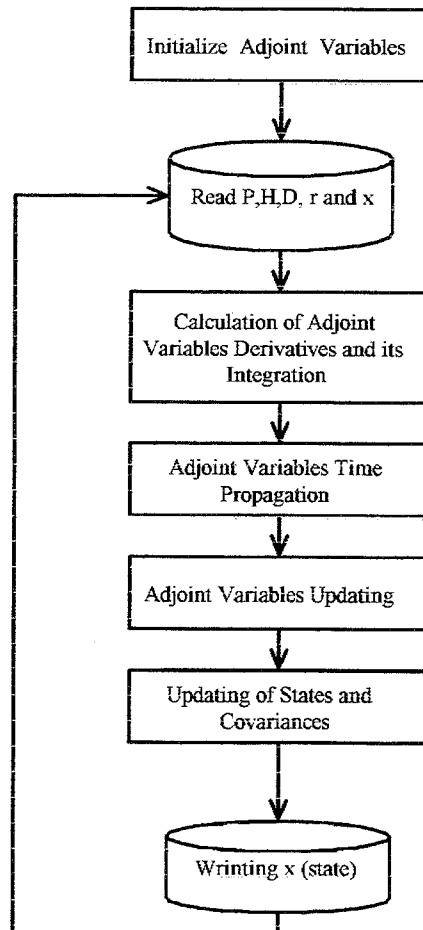
$$\hat{\mathbf{x}}_{k/k|smoother} = \hat{\mathbf{x}}_{k/k|EKF} - \mathbf{P}_{k/k|EKF} \boldsymbol{\lambda}_{k/k}$$

and the corresponding covariance matrix is given by,

$$\mathbf{P}_{k/k|smoother} = \mathbf{P}_{k/k|EKF} - \mathbf{P}_{k/k|EKF} \boldsymbol{\Lambda}_{k/k} \mathbf{P}_{k/k|EKF}$$

In order to reduce computation time $\mathbf{P}_{k/k}$, $\mathbf{H}_{k/k-1}$, $\mathbf{D}_{k/k-1}$, $\mathbf{D}_{k/k-1}^{-1}$, \mathbf{K}_k , \mathbf{r}_k and $\hat{\mathbf{x}}_{k/k}$ computed by the EKF are temporarily stored to be used by the smoother.

Below, it is presented the flow diagram of the Bryson-Frazier smoother;



Flow diagram 3. Bryson-Frazier Smoother

(v) Fixed Lag Smoother-Differentiator;

The Fixed Lag smoother-differentiator formulation follows Fioretti and Jetto [4] and reference [2].

The use of the Fixed Lag smoother requires:

- the definition of a model for the signal.
- the determination of the signal noise characteristics.
- smoothing of the modelled signal for the identified noise characteristics.

The data model is formulated as;

$$\mathbf{x}((k+1)\Delta t) = \mathbf{A}_m \mathbf{x}(k\Delta t) + \sigma_w \underline{\mathbf{w}}((k+1)\Delta t) \quad (6)$$

where

$$\mathbf{A}_m = \begin{bmatrix} 1 & \Delta t & \frac{\Delta t^2}{2!} & \dots & \frac{\Delta t^n}{n!} \\ 0 & 1 & \Delta t & \dots & \frac{\Delta t^{n-1}}{(n-1)!} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix} \quad (7)$$

with $\underline{\mathbf{w}}$ denoting a white noise sequence $\sim N(0, \underline{\mathbf{Q}})$ (Gaussian, zero mean and covariance $\underline{\mathbf{Q}}$). Δt is the sampling interval and the elements of its covariance matrix $\underline{\mathbf{Q}}$ are given by the generic expression,

$$q_{i,j} = \frac{\Delta t^{(2n+3)-(i+j)}}{(n+1-i)!(n+1-j)!(2n+3)-(i+j)}$$

The measured state component is the sampled data, thus the observation equation may be written as,

$$\mathbf{z}(k\Delta t) = \mathbf{C}\mathbf{x}(k\Delta t) + \mathbf{w}_z(k\Delta t) \quad (8)$$

Using the model formed by equations 6, 7 and 8 an ordinary Kalman Filter estimates \mathbf{x} from the measurement \mathbf{z} of \mathbf{x} and an initial guess to the noise. The filter residuals (innovation) are calculated by:

$$r_k = z(k\Delta t) - \mathbf{C}\mathbf{A}_m \hat{\mathbf{x}}_{k-1/k-1}$$

The filter calculates and stores the residuals of the estimation given an initial value for the noise covariance. With the residuals and the **stabilized** filter covariance the theoretical and estimated autocorrelation are determined.

The theoretical autocorrelation function φ of the residuals in steady-state condition is given by;

$$\varphi(\sigma_w, \sigma_\eta, i) = [\mathbf{C}\tilde{\mathbf{P}}(\sigma_w, \sigma_\eta)\mathbf{C}^T + \sigma_\eta^2]\delta_i$$

where δ_i is the Kronecker delta function ($\delta_i = 1$ for $i = 0$, $\delta_i = 0$ for $i \neq 0$) and $\tilde{\mathbf{P}}(\sigma_w, \sigma_\eta)$ is the filter estimated covariance in steady-state condition.

The actual autocorrelation function of the innovation process is calculated by,

$$\hat{\varphi}(\sigma_w, \sigma_\eta, i) = \frac{1}{m-i} \sum_{k=1}^{m-i} r_k r_{k+i}, \quad i=0, 1, \dots, l$$

where:

- r_k ($k=1, 2, \dots, m$) are samples of the innovation process computed with the steady Kalman filter gain.
- m is the total number of observation.
- l is the filter lag measured in iteration steps.

The determination of the noise characteristics of the signal is carried out by minimizing the error between the theoretical and the estimated autocorrelation. The cost function is given by;

$$J(\sigma_w, \sigma_\eta) = \sum_{i=0}^l [(\varphi(\sigma_w, \sigma_\eta, i) - \hat{\varphi}(\sigma_w, \sigma_\eta, i))]^2$$

The optimal values of the process and measurement noise σ_w and σ_η , respectively, are those that minimize the norm of J in a region R of the (σ_w, σ_η) -plane for the autocorrelation functions calculated with the steady-state gain of the Kalman filter. The cost function J is rewritten as,

$$J(\sigma_w, \sigma_\eta) = J_1(\sigma_w, \sigma_\eta) + J_2(\sigma_w, \sigma_\eta)$$

where,

$$J_1(\sigma_w, \sigma_\eta) = [\mathbf{C}\mathbf{P}(\sigma_w, \sigma_\eta)\mathbf{C}^T + \sigma_\eta^2 - \sigma_r^2]^2$$

and,

$$J_2 = \sum_{i=1}^l \hat{\varphi}(\sigma_w, \sigma_\eta, i)^2$$

with,

$$\sigma_r^2 = \frac{1}{m} \sum_{k=1}^m r_k^2$$

Fixing a value $\bar{\sigma}_\eta$ for σ_η , J_2 is minimized by varying σ_w with the help of a simple minimization algorithm. Once $\bar{\sigma}_w$ has been calculated, J_1 is minimized by calculating:

$$\sigma_w^2 = \hat{\sigma}_w^2 = \frac{\bar{\sigma}_r^2 \bar{\sigma}_w^2}{\mathbf{C}\hat{\mathbf{P}}(\bar{\sigma}_w, \bar{\sigma}_\eta)\mathbf{C}^T + \bar{\sigma}_\eta^2}$$

The optimal pair (σ_w, σ_η) is the pair $(\hat{\sigma}_w, \hat{\sigma}_\eta)$ with $\hat{\sigma}_\eta = \frac{\bar{\sigma}_\eta}{\bar{\sigma}_w} \hat{\sigma}_w$ that minimized both $J_1(\sigma_w, \sigma_\eta)$ and $J_2(\sigma_w, \sigma_\eta)$ and consequently $J(\sigma_w, \sigma_\eta)$.

The Fixed Lag Smoother is implemented augmenting the state vector with additional states representing the values of the states in the previous instant of time. The additional states are represented by,

$$\begin{aligned} \mathbf{x}_1(i) &= \mathbf{x}(i-1) \\ \mathbf{x}_2(i) &= \mathbf{x}_1(i-1) \\ &\vdots \\ \mathbf{x}_l(i) &= \mathbf{x}_{l-1}(i-1) \end{aligned}$$

Thus the additional vectors contains the l previous values of the state vector $\mathbf{x}(i)$ where l is the time lag in sampling intervals, resulting in the following augmented state space system:

$$\begin{bmatrix} \mathbf{x}(i) \\ \mathbf{x}_1(i) \\ \mathbf{x}_2(i) \\ \vdots \\ \mathbf{x}_l(i) \end{bmatrix} = \begin{bmatrix} \mathbf{A}_m(i/i-1) & 0 & \cdots & 0 & 0 \\ \mathbf{I} & 0 & \cdots & 0 & 0 \\ 0 & \mathbf{I} & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & \mathbf{I} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x}(i-1) \\ \mathbf{x}_1(i-1) \\ \mathbf{x}_2(i-1) \\ \vdots \\ \mathbf{x}_l(i-1) \end{bmatrix} + \begin{bmatrix} \sigma_w^2 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \mathbf{w}_x(i-1)$$

while the corresponding output equation is,

$$\mathbf{z}(i) = \begin{bmatrix} C & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x}(i) \\ \mathbf{x}_1(i) \\ \mathbf{x}_2(i) \\ \vdots \\ \mathbf{x}_l(i) \end{bmatrix} + \sigma_\eta(i)$$

Fixed-lag smoothing is accomplished by applying Kalman filter equations to the augmented state space model, resulting;

- covariance propagation,

$$\begin{aligned}\mathbf{P}(i/i-1) &= \mathbf{A}\mathbf{P}(i-1/i-1)\mathbf{A}_m^T + \sigma_w^2\overline{\mathbf{Q}} \\ \mathbf{P}_1(i/i-1) &= \mathbf{P}(i-1/i-1)\mathbf{A}_m^T \\ &\vdots \\ \mathbf{P}_l(i/i-1) &= \mathbf{P}_l(i-1/i-1)\mathbf{A}_m^T\end{aligned}$$

- Kalman gain,

$$\begin{aligned}\mathbf{K}(i) &= \mathbf{P}(i/i-1)\mathbf{C}^T[\mathbf{C}\mathbf{P}(i/i-1)\mathbf{C}^T + \sigma_\eta^2]^{-1} \\ \mathbf{K}_1(i) &= \mathbf{P}_1(i/i-1)\mathbf{C}^T[\mathbf{C}\mathbf{P}_1(i/i-1)\mathbf{C}^T + \sigma_\eta^2]^{-1} \\ &\vdots \\ \mathbf{K}_l(i) &= \mathbf{P}_l(i/i-1)\mathbf{C}^T[\mathbf{C}\mathbf{P}_l(i/i-1)\mathbf{C}^T + \sigma_\eta^2]^{-1}\end{aligned}$$

- state update,

$$\begin{aligned}\hat{\mathbf{x}}(i/i) &= \hat{\mathbf{x}}(i/i-1) + \mathbf{K}_k[\mathbf{z}(i) - \mathbf{C}\hat{\mathbf{x}}(i/i-1)] \\ \hat{\mathbf{x}}(i-1/i) &= \hat{\mathbf{x}}(i-1/i-1) + \mathbf{K}_1[\mathbf{z}(i) - \mathbf{C}\hat{\mathbf{x}}(i/i-1)] \\ &\vdots \\ \hat{\mathbf{x}}(i-l/i) &= \hat{\mathbf{x}}(i-l/i-1) + \mathbf{K}_l[\mathbf{z}(i) - \mathbf{C}\hat{\mathbf{x}}(i/i-1)]\end{aligned}$$

Here the prior estimate on the right-hand side of each equation is the previous value of the posterior estimate from the left-hand side of the equation immediately above.

- covariance update,

$$\begin{aligned}\mathbf{P}(i/i) &= [\mathbf{I} - \mathbf{K}_k\mathbf{C}]\mathbf{P}(i/i-1) \\ \mathbf{P}_1(i/i) &= \mathbf{P}_1(i/i-1)[\mathbf{I} - \mathbf{C}^T\mathbf{K}_k^T] \\ &\vdots \\ \mathbf{P}_l(i/i) &= \mathbf{P}_l(i/i-1)[\mathbf{I} - \mathbf{C}^T\mathbf{K}_k^T]\end{aligned}$$

The covariance matrices without subscript such as $\mathbf{P}(i/i-k)$ and $\mathbf{P}(i/i)$ and the Kalman gain matrix \mathbf{K}_k are those of the standard Kalman filter algorithm applied to the original state space system while the remaining \mathbf{K}_1 and \mathbf{P}_l are generated by the Fixed Lag smoother.

The optimal lag is defined as two or three times the dominant time constant of the Kalman filter given by;

$$l \geq \text{int}[2 \tau_f / \Delta t]$$

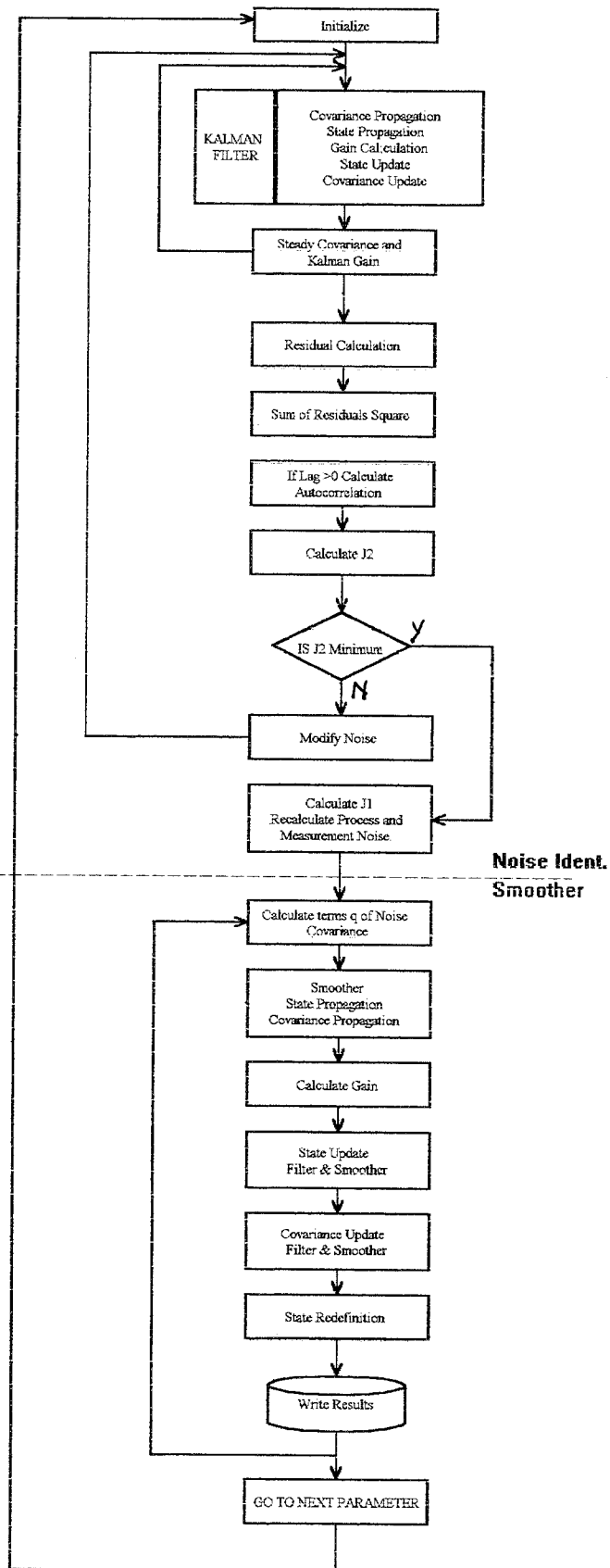
where τ_f is the dominant time constant of the filter and Δt is the sampling interval

$$\tau_f = -\Delta t / \ln \lambda_{\max}$$

λ_{\max} is the dominant eigenvalue of the Kalman filter dynamic matrix

$$[\mathbf{A}_m - \mathbf{K}_k \mathbf{C} \mathbf{A}_m]$$

The eigenvalues are estimated outside of the program (e.g. using Matlab). The dynamic matrix is calculated for every smoothed state and its time constant is determined in order to identify whether the assumed lag satisfy or not the ratio of 2 to 3 time constants. Next, a flow diagram of the program EKFDER.FOR is presented.



Flow Diagram 4. Program EKFDER.FOR

2.2 Linear Regression and Stepwise Regression

- Normal Equation Solution to the Linear Regression;

If \mathbf{A} is the matrix of the independent variables \mathbf{x} (measurements) and \mathbf{y} is the vector of the dependent variable the coefficients of the expansion of \mathbf{y} as a function of \mathbf{x} , i.e., $\mathbf{y} = \mathbf{A}\Theta + e$ are given by:

$$\Theta = [\mathbf{A}^T \mathbf{A}]^{-1} \mathbf{A}^T \mathbf{y} \quad (8)$$

This is the normal equation formulation of the linear regression.

- Linear Regression Solution by Householder Transformation;

The Householder Transformation (an orthogonal transformation) is used to triangularize the matrix formed by augmenting the matrix \mathbf{A} of the independent variables with the vector \mathbf{y} of the dependent variable.

Let \mathbf{T} be the transformation. If \mathbf{A} is $n \times m$, applying m transformation to the augmented matrix the resulting matrix will be triangular;

$$\mathbf{T}[\mathbf{A} \mid \mathbf{y}] = [\bar{\mathbf{A}} \mid \bar{\mathbf{y}}]$$

In this case the solution of $\mathbf{y} = \mathbf{A}\Theta + e$ may be formulated as;

$$\Theta = \bar{\mathbf{A}}^{-1} \bar{\mathbf{y}}$$

This is easily obtained because $\bar{\mathbf{A}}$ is triangular and the inversion may be obtained by back substitution.

Matrix Triangularization by Householder transformation, from Ref. [2] : The basic operation of the triangularization process using the *Householder Transformation* is the construction of the scalar s and the matrix $\tilde{\mathbf{A}}(m, n-1)$ so that,

$$\mathbf{T}_u \mathbf{A} = \begin{bmatrix} s & & \\ & \tilde{\mathbf{A}} & \\ 0 & & \end{bmatrix}$$

where s and $\tilde{\mathbf{A}}$ are computed by;

$$s = -\text{sgn}(A(1,1)) \left[\sum_{i=1}^n [A(i,1)]^2 \right]^{1/2}$$

with

$$u_{(1)} = A(1,1) - s \quad \text{and} \quad u_{(i)} = A(i,1) \quad i=2,3,\dots,m$$

for the application of one elementary transformation. Generalizing for m applications, results;

$$s_j = -\text{sgn}(A(j,j)) \left[\sum_{i=j}^m [A(i,j)]^2 \right]^{1/2} \quad \text{with}$$

$$\tilde{A}(i,j-1) = A(i,j) + \gamma u(i) \quad u_j(i) = 0 \quad i < j$$

$$\gamma = \beta \sum_{i=1}^n u_{(i)} A(i,j) \quad u_j(i) = A(i,j) \quad i > j$$

- Stepwise Regression Procedure

The stepwise regression procedure is a process that includes or excludes independent variables in a regression model by analysing how the variables contribute to the overall fit of the regression model. The exclusion process is carried out by the analysis of the partial 'F' test performed for each variable included in the model while the inclusion process is carried out by residual analysis, as presented in chapter 4 of Reference [2].

Once the regression coefficients were calculated by any one of the regression process above described, the following terms are calculated;

$$Dy = y - \hat{y} \quad \text{Regression residual, where } \hat{y} \text{ is the dependent variable values calculated by the model.}$$

$$RESS = \sum Dy * Dy \quad \text{Residual sum of squares.}$$

$$VAR = RESS / (NAT - IVV) \quad \text{Residual Variance, where NAT is the number of samples and IVV is the number of independent variables in the regression model.}$$

$$SB(I) = \sqrt{VAR * [\mathbf{A}^T(I,I)\mathbf{A}(I,I)]^{-1}} \quad \text{Standard error of the regression coefficients.}$$

$$F = \frac{\Theta \mathbf{A}^T y - NAT y_{AVG}^2}{VAR (IVV - 1)} \quad \text{Regression 'F' value.}$$

y_{AVG} is the average value of the dependent variable.

$$R^2 = \frac{F}{\frac{(NAT - IVV)}{(IVV - 1)} + F} \quad \text{Regression correlation coefficient}$$

The partial correlation coefficient is calculated by;

$$F_p(i) = \frac{\Theta^2(i)}{SB^2(i)}, \quad \text{That is, one coefficient } F_p \text{ for every variable in the regression.}$$

If the calculated value of F_p is small than a specified value (e.g. 7) then the variable is rejected from the model (only one variable is rejected, i.e., the one with smallest F_p value).

To analyse which variable, between the ones not yet included in the model, is the best candidate to be included in the next interaction, the following steps are performed;

- Calculate the regression coefficients of the actual model taken as dependent variables the independent variables not yet included in the model;

$$\Theta_{iN} = [\mathbf{A}^T \mathbf{A}]^{-1} \mathbf{A}^T \mathbf{x}_i \quad (9)$$

- If \hat{y}_{iN} is the estimated value of \mathbf{x}_i , then the residual of the regression is given by;

$$\mathbf{z}_i = \hat{y}_{iN} - \mathbf{x}_i \quad \text{Residual}$$

$$z_{AVG} = \sum z_i / NAT \quad \text{Residual average}$$

- Calculate;

$$SJJ = \sum z_i z_i \quad \text{The residual sum of squares,}$$

$$SJY = \sum (z_i - z_{AVG})(y - \hat{y} - Dy_{AVG}) \quad \text{and} \quad RJY = \frac{SY Y}{\sqrt{SY Y * SJJ}}$$

The process is repeated for all variables not yet included in the model. The next variable to be included in the model will be the one presenting biggest RJY value.

3. INPUT DATA FILES.

3.1 Files for the EKFMFOR and IEKF.FOR programs.

The programs use two files, FileA and FileM.

FileA contains the flight data consisting in lines of measured data containing the following parameters in free format:

α_x - longitudinal acceleration (m/s)
 α_y - lateral acceleration (m/s)
 α_z - normal acceleration (m/s)
 p - roll rate (rad)
 q - pitch rate (rad)
 r - roll rate (rad)
 V - true airspeed (m/s)
 α - incidence angle (rad)
 β - sideslip angle (rad)
 h - altitude (m)
 θ - pitch attitude
 ϕ - roll attitude.
Thrust (2 propellers)
Propeller Normal Force (2 propellers).

Table 3.1 below represents one data sample of FILEA.DAT

9.979670E-01	-2.394202E-03	-5.746084E-02	-6.204185E-03
-8.863121E-04	-8.181342E-04	68.656340	6.889767E-02
1.218210E-02	2085.999000	1.046959E-01	-9.683409E-03
6600.000000	400.000000		

Table 3.1 Example of FILEA.DAT

FileM contains the initial value of the states, covariances, aircraft mass, inertias, etc.

FileM starts with the states initial values (7 values in every line);
By order the states are:

1. u - longitudinal component of the velocity
2. v - lateral component of the velocity.

3. w - normal component of the velocity.
4. p - roll rate
5. q - pitch rate
6. r - roll rate
7. θ - pitch attitude
8. ϕ - roll attitude.
9. h - altitude
10. X1 - corresponds to (T+X)/mass
11. X2 - internal variable (Gauss Markov model) that may be set equal to zero.
12. X3 - internal variable (Gauss-Markov model) that may be set equal to zero.
13. Y1 - is the normalised Lateral Force.
14. Y2 - internal variable (Gauss-Markov model) that may be set equal to zero.
15. Y3 - internal variable (Gauss-Markov model) that may be set equal to zero.
16. Z1 - is the normalised Normal Force.
17. Z2 - internal variable (Gauss-Markov model) that may be set equal to zero.
18. Z3 - internal variable (Gauss-Markov model) that may be set equal to zero.
19. L1 - is the normalised Roll Moment.
20. L2 - internal variable (Gauss-Markov model) that may be set equal to zero.
21. L3 - internal variable (Gauss-Markov model) that may be set equal to zero.
22. M1- is the normalised Pitch Moment.
23. M2 - internal variable (Gauss-Markov model) that may be set equal to zero.
24. M3 - internal variable (Gauss-Markov model) that may be set equal to zero.
25. Z1 - is the normalised Normal Force.
26. Z2 - internal variable (Gauss-Markov model) that may be set equal to zero.
27. Z3 - internal variable (Gauss-Markov model) that may be set equal to zero.
28. Normal acceleration bias
29. Incidence bias.
30. Sideslip bias
31. Pitch rate bias
32. Pitch Attitude bias
33. Longitudinal acceleration bias.

Note that the numbers above represent also the order of the states in the state vector, as used in the programs.

After the state initial values the file contains the covariance of the measurements in the following order (6 values every line);

1. a_x , 2. a_y , 3. a_z , 4. p , 5. q , 6. r
7. V , 8. Incidence, 9. Sideslip, 10. Altitude, 11. Pitch attitude, 12. Attitude Roll.

After that, the file contains the Process Noise of the state variables in the same order as for the initial values of the states and in the same format. After that the file contains the covariance of the initial values of the state variables (same order and format).

The last line of the file contains:

Number of samples, sample interval, CG%, Ix, Iy, Iz, Ixz, Engine z-arm to CG.

Table 3.2 shows an example of FILEB.DAT

68.54	0	4.64	-6.20E-03	-8.86E-04	-8.18E-04	1.04E-01
-9.68E-03	2085	.94	0	0	0	0
0	-9.999	0	0	0	0	0
0	0	0	0	0	0	0
0	0	0	0.34	0.22		
2.1E-04	2.9E-04	1.5E-04	1.0E-08	7.0E-08	2.0E-07	
1.3E-03	1.7E-06	8.0E-07	0.67	8.0E-07	1.0E-08	
0.075	0.05	0.075	0.00001	0.00001	0.00001	0.000001
0.00001	0.75	0.025	0.025	0.01	0.025	0.025
0.001	0.025	0.025	0.01	0.02	0.01	0.001
0.02	0.01	0.001	0.02	0.01	0.001	0.0000000000001
.000000000000001	.000000000000001	.000000000000001	.000000000000001	.000000000000001	.000000000000001	.000000000000001
0.1	0.1	0.1	0.1	0.1	0.1	0.1
0.1	0.1	0.1	0.1	0.1	0.1	0.1
0.1	0.1	0.1	0.1	0.1	0.1	0.1
0.1	0.1	0.1	0.1	0.1	0.1	0.5
0.25	0.25	0.25	0.25	5.0		
1965	0.04	25.59	26682.	42657.	54217.	3304. 0.0633

Table 3.2 Example of FILEB.DAT

This file in fact may have a non fixed format because it depends of the number of states in the program, mainly the bias terms effectively included in the measurement model.

3.2 Files for the EKFDER.FOR program.

The program EKFDER.FOR needs two data files, FILEC and FILED.

FILEC.DAT:

FILEC.DAT: contains the information relative to the parameters that are going to be smoothed-differentiated.

FILEC first line contains the initial values of the measurement and process noise, the last in general formulated as a large number (e.g. 5000 is a good number).

The next four lines contain a 4x4 initial covariance matrix, reflecting an initial guess to the covariances.

It is followed by a line containing the total number of samples in the file, the number of parameters in each sample, the sampling interval, the lag (in sampling intervals) to be used in the smoother, the aircraft MASS and the aircraft inertia I_y .

Next lines contain the parameters to be smoothed-differentiated. Table 3.3 below presents a typical format of FILEC.DAT

0.05	5000.				
0.01	0.1	0.1	0.1	!	
0.1	0.1	0.1	0.1	!	May be used
0.1	0.1	0.1	0.5	!	for any data.
0.1	0.1	0.5	1.0	!	
2015	9	0.04	15	5334.	52600.
69.00		4.551100		-4.6222E-03	7.742999E-02
-6.011E-03		-3.834E-01		2.2265E-03	5.555777E-03
-1.666E-01					!
69.10		4.581100		-4.6322E-03	7.752909E-02
-6.111E-03		-3.854E-01		2.2765E-03	5.565787E-03
-1.766E-01					!
.					! A total of 2015
.					! samples.
.					!
69.10		4.581100		-4.6322E-03	7.752909E-02
-6.111E-03		-3.854E-01		2.2765E-03	5.565787E-03
-1.766E-01					!

Table 3.3 FILEC.DAT

FILED.DAT

Contains the elevator angle in rd (or any other control input). It is not used in the program for calculation purposes. It is used only to include the input in the regression file to be used in the linear regression. Note that if more inputs are added, the program EKFDER has to be revised in order to reflect the additional parameters.

File example:

0.020
0.021
0.021
.
.
.
0.001

3.3 Data File of the Regression Programs MSR.FOR and MSRH.FOR.

One common file is required by the regression programmes MSR.FOR and MSRH.FOR.

The file contains lines of samples taken in the time interval. Each line contains the independent variables plus the dependent variables. The file first line contains the total number of sample, the number of independent variables and the number of dependent variables. The maximum number of independent variables is 11 and the maximum number of dependent variables is 3. The maximum number of samples is 2000.

Table 3.4 below presents a typical format of the data file;

2000	11	3			
68.562580	4.853940	4.829005E-03	7.069317E-02		! 11 independent
-1.054134E-01	4.619305E-03	4700.827000	23.560730		! variables plus 3
-8.899439E-02	-2.015400E-02	4155.000000	-535.373500		! dependent
-52571.430000	-328.451300				! variables.
68.562680	4.847060	3.998138E-03	7.072155E-02		
-1.642646E-01	3.966332E-03	4700.841000	23.493990		
-8.909377E-02	-2.142626E-02	4156.000000	-517.561400		
-52637.130000	-253.778700				
68.562870	4.839545	3.109287E-03	7.072821E-02		
-2.095646E-01	3.879622E-03	4700.867000	23.421190		
-8.929249E-02	-2.291300E-02	4157.000000	-483.274600		
-52673.050000	-221.317600				
68.563100	4.830721	2.182843E-03	7.071401E-02		
-2.468214E-01	3.965823E-03	4700.899000	23.335870		
-8.919313E-02	-2.311341E-02	4158.000000	-445.931600		
-52681.290000	-224.798200				

68.563120	4.818213	1.296546E-03	7.067891E-02
-2.895375E-01	2.652711E-03	4700.901000	23.215180
-8.859694E-02	-2.075625E-02	4159.000000	-437.393500
-52668.930000	-241.696900		
.			
.			
.			
68.521100	4.810721	2.082843E-03	7.061401E-02
-2.468214E-01	3.965823E-03	4700.899000	23.345870
-8.919313E-02	-2.311341E-02	4158.000000	-445.931600
-52691.390000	-225.698200		
Total of 2000 sample			

Table 3.4 Regression data file.

2000 x 12 represents the maximum matrix size (64 K) that can be handled by a PC using DOS.

4. PROGRAM LISTINGS.

The program listings contain notation of the main variables used in the programs.

4.1 Program EKFMBF.FOR

```
C ***** PROGRAM EKFMBF.FOR *****
C                                     Hoff Aug/94
C                                     REV MAR/95, REV APR/P5
C
C Estimation of aircraft states u,v,w,p,q,r,X,Y,Z,L,M,N
C alpha,beta using an Extended Kalman Filter and a Modified
C Bryson-Frazier Smoother applied to an inertial and
C gravitational model.
C
C Bias terms to Airspeed, Incidence, Sideslip and Pitch Rate.
C
C Needs two data files
c (i) File with initial conditions, process and measurement
C     noise covariances, CG position, aircraft Inertias,
C     Mass and engine z arm to CG.
C (ii) File with flight data: Ax, Ay, Az, p, q, r, True airspeed,
C     Alpha, Beta, Altitude, theta, Phi, Thrust, Propeller
C     Normal Force (metric units, angles in rad.).
C
C NAT  Number of samples in the data file
C CG   Aircraft CG %
C RR   Covariance matrix of the measurements
C QQ   Process noise covariance
C PU   Initial covariance of the estimates
C DT   Sampling interval
C Ix, Iy,... Moment of Inertia
C MASS AIRCRAFT MASS
C X(I) MATRIX OF UPDATED STATES
C XM(I) MATRIX OF TIME PROPAGATED STATES
C PM(I,J) TIME PROPAGATED COVARIANCE MATRIX
C PU(I,J) UPDATED COVARIANCE MATRIX
C KK(I,J) KALMAN GAIN MATRIX
C hh(I) VECTOR OF MEASUREMENT MODELS
C ZZ(I) VECTOR OF MEASUREMENTS
C DH(I) RESIDUAL
C F(I,J) GRADIENT OF THE DYNAMIC MODEL
C H(I,J) GRADIENT OF THE MEASUREMENT MODEL
C LAM,LAMB,LAMD SMOOTHER ADJOINT VARIABLE IN ITS DIF. FORMATS
C AA,AAM SMOOTHER ADJOINT VARIABLE IN ITS DIF. FORMATS
C PD,RK,KH AUXILIARY MATRICES
C WORK,IDENT AUXILIARY MATRICES OF INVERSION ROUTINE
C Lze  ENGINE z ARM RELATIVE TO THE CG.

REAL*8 H(12,31),PM(31,31),QQ(31),hh(12),DH(12),
+ LAM(31),LAMB(31),LAMD(31)
REAL*4 X(31),XM(31),ZZ(12),RR(12),Ix,Iy,Iz,Ixz,I2,Lze,MASS
REAL*8 WORK[ALLOCATABLE] (:,:), RK[ALLOCATABLE] (:,:),
+ PD[ALLOCATABLE] (:,:), F[ALLOCATABLE] (:,:),
+ KK[ALLOCATABLE] (:,:), HPHT1[ALLOCATABLE] (:,:),
```

```

+   HPHT[ALLOCATABLE] (:,:), KH[ALLOCATABLE] (:,:),
+   PU[ALLOCATABLE] (:,:), IDENT[ALLOCATABLE] (:,:),
+   AA[ALLOCATABLE] (:,:), AAM[ALLOCATABLE] (:,:)
INTEGER*2 IN,K1,K4
INTEGER*4 MM,K2,K3
OPEN(UNIT=2,FILE='phu809A.DAT',STATUS='OLD')
OPEN(UNIT=6,FILE='FILEC.DAT',STATUS='OLD')
OPEN(UNIT=3,FILE='PMBFD1.DAT')
OPEN(UNIT=4,FILE='PMBFD2.DAT')
c OPEN(UNIT=5,FILE='PMBFD3.DAT')
OPEN(UNIT=7,FILE='PMBFE3.DAT')
OPEN(UNIT=9,FILE='PMBFE1.DAT')
OPEN(UNIT=10,FILE='PMBFE2.DAT')
OPEN(UNIT=8,ACCESS='DIRECT',FILE='DDAT8.DAT',
+ FORM='UNFORMATTED',RECL=248,STATUS='NEW')
OPEN(UNIT=12,ACCESS='DIRECT',FILE='DDAT1.DAT',
+ FORM='UNFORMATTED',RECL=224,STATUS='NEW')
OPEN(UNIT=13,ACCESS='DIRECT',FILE='DDAT2.DAT',
+ FORM='UNFORMATTED',RECL=96,STATUS='NEW')
OPEN(UNIT=14,ACCESS='DIRECT',FILE='DDAT3.DAT',
+ FORM='UNFORMATTED',RECL=96,STATUS='NEW')
OPEN(UNIT=15,ACCESS='DIRECT',FILE='DDAT4.DAT',
+ FORM='UNFORMATTED',RECL=96,STATUS='NEW')
OPEN(UNIT=16,ACCESS='DIRECT',FILE='DDAT5.DAT',
+ FORM='UNFORMATTED',RECL=96,STATUS='NEW')
OPEN(UNIT=17,ACCESS='DIRECT',FILE='DDAT6.DAT',
+ FORM='UNFORMATTED',RECL=248,STATUS='NEW')
OPEN(UNIT=18,ACCESS='DIRECT',FILE='DDAT7.DAT',
+ FORM='UNFORMATTED',RECL=248,STATUS='NEW')
He=0.0
G=9.80665
MM=0
K1=0
K2=0
K3=0
K4=0

C... INITIALIZATION - Matrix in a data file with all initial values
C      obtained averaging the stabilization
READ(6,*) (X(K),K=1,7)      !
READ(6,*) (X(K),K=8,14)    !
READ(6,*) (X(K),K=15,21)   ! STATES INITIAL VALUES
READ(6,*) (X(K),K=22,28)   !
READ(6,*) (X(K),K=29,31)   !

C
READ(6,*) (RR(I),I=1,6)    ! MEASUR. NOISE COVAR. MATRIX
READ(6,*) (RR(I),I=7,12)  ! ASSUMED DIAGONAL

C
READ(6,*) (QQ(J),J=1,7)    !
READ(6,*) (QQ(J),J=8,14)  ! PROCESS NOISE COVAR. MATRIX
READ(6,*) (QQ(J),J=15,21) ! ASSUMED DIAGONAL

```



```

DO I=1,31
DO J=1,31
  F(I,J)=0.D0 ! Zeroing F
ENDDO
DO IJ=1,12
  H(IJ,I)=0.D0 ! Zeroing H
ENDDO
XM(I)=X(I) ! Updating XM
ENDDO

```

```

COST=COS(X(7))
SINT=SIN(X(7))
COSP=COS(X(8))
SINP=SIN(X(8))
TANT=SINT/COST

```

C...

```

F(1,2) = X(6)
F(1,3) = -X(5)
F(1,5) = -X(3)
F(1,6) = X(2)
F(1,7) = -G*COST
F(1,10) = 1.

```

C...

```

F(2,1) = -X(6)
F(2,3) = X(4)
F(2,4) = X(3)
F(2,6) = -X(1)
F(2,7) = -G*SINT*SINP
F(2,8) = G*COST*COSP
F(2,13) = 1.

```

C...

```

F(3,1) = X(5)
F(3,2) = -X(4)
F(3,4) = -X(2)
F(3,5) = X(1)
F(3,7) = -G*SINT*COSP
F(3,8) = -G*COST*SINP
F(3,16) = 1.

```

C...

```

F(4,4) = X(5)*A11
F(4,5) = X(4)*A11 + X(6)*A12 + A13
F(4,6) = X(5)*A12
F(4,19) = 1.
F(4,25) = A14

```

C...

```

F(5,4) = X(6)*A21-2.*X(4)*A22
F(5,6) = X(4)*A21+2.*X(6)*A22-A23
F(5,22) = 1.

```

C...

```

F(6,4) = X(5)*A31
F(6,5) = X(4)*A31+X(6)*A32+A33

```

$$F(6,6) = X(5)*A32$$

$$F(6,19) = A34$$

$$F(6,25) = 1.$$

C...

$$F(7,5) = COSP$$

$$F(7,6) = -SINP$$

$$F(7,8) = -X(5)*SINP-X(6)*COSP$$

C...

$$F(8,4) = 1.$$

$$F(8,5) = TANT*SINP$$

$$F(8,6) = TANT*COSP$$

$$SECTH2 = 1. + TANT*TANT$$

$$F(8,7) = X(5)*SINP*SECTH2 + X(6)*COSP*SECTH2$$

$$F(8,8) = X(5)*TANT*COSP - X(6)*TANT*SINP$$

C...

$$F(9,1) = SINT$$

$$F(9,2) = -COST*SINP$$

$$F(9,3) = -COST*COSP$$

$$F(9,7) = -(-X(1)*COST-X(2)*SINT*SINP-X(3)*SINT*COSP)$$

$$F(9,8) = -(X(2)*COST*COSP-X(3)*COST*SINP)$$

C.....gauss-markov models

$$F(10,11) = 1.$$

$$F(11,12) = 1.$$

$$F(13,14) = 1.$$

$$F(14,15) = 1.$$

$$F(16,17) = 1.$$

$$F(17,18) = 1.$$

$$F(19,20) = 1.$$

$$F(20,21) = 1.$$

$$F(22,23) = 1.$$

$$F(23,24) = 1.$$

$$F(25,26) = 1.$$

$$F(26,27) = 1.$$

C

C** Covariance Time Propagation

C

DO I=1,31

DO J=1,31

FDT = F(I,J)*DT

FDT2 = FDT*FDT

F(I,J)=FDT+0.5*FDT2

ENDDO

F(I,I) = F(I,I) + 1.0

ENDDO

```
ALLOCATE (PD(31,31)) ! PD the derivative of P
```

```
DO I=1,31
DO J=1,31
  PD(I,J)=0.D0
DO K=1,31
  PD(I,J)=PD(I,J)+PU(L,K)*F(J,K) ! P*FT
ENDDO
ENDDO
ENDDO
```

```
DO I=1,31
DO J=1,31
  PM(I,J)=0.D0
DO K=1,31
  PM(I,J)=PM(I,J)+F(I,K)*PD(K,J) ! P=F*P*FT
ENDDO
ENDDO
  PM(I,I) = PM(I,I) + QQ(I) ! PM=F*PU*FT+Q
ENDDO
```

```
DEALLOCATE(F,PD,PU)
ALLOCATE (RK(4,21))
```

C

C RUNGE-KUTTA INTEGRATION - DYNAMIC SYSTEM - time propagation

```
XK=2.
DO K=1,4
SINT=SIN(X(7))
COST=COS(X(7))
SINP=SIN(X(8))
COSP=COS(X(8))
TANT=SINT/COST
RK(K,1)= DT*(X(6)*X(2)-X(5)*X(3)-G*SINT + X(10))
RK(K,2)= DT*(X(4)*X(3)-X(6)*X(1)+G*COST*SINP + X(13))
RK(K,3)= DT*(X(5)*X(1)-X(4)*X(2)+G*COST*COSP + X(16))
RK(K,4)= DT*(X(4)*X(5)*A11+X(5)*X(6)*A12+X(5)*A13+X(19)
+ +A14*X(25))
RK(K,5)= DT*(X(4)*X(6)*A21+(X(6)*X(6)-X(4)*X(4))*A22
+ - X(6)*A23 + X(22))
RK(K,6)= DT*(X(4)*X(5)*A31+X(5)*X(6)*A32+X(5)*A33
+ + X(25)+A34*X(19))
RK(K,7)= DT*(X(5)*COSP - X(6)*SINP)
RK(K,8)= DT*(X(4)+X(5)*TANT*SINP + X(6)*TANT*COSP)
RK(K,9)=-DT*(-X(1)*SINT+X(2)*COST*SINP+X(3)*COST*COSP)
RK(K,10)=X(11)*DT !
RK(K,11)=X(12)*dt !
RK(K,12)=X(14)*DT !
RK(K,13)=X(15)*dt !
RK(K,14)=X(17)*DT !
RK(K,15)=X(18)*dt ! GAUSS-MARKOV PARAMETERS
RK(K,16)=X(20)*DT !
```

```

RK(K,17)=X(21)*dt      !
RK(K,18)=X(23)*DT      !
RK(K,19)=X(24)*dt      !
RK(K,20)=X(26)*DT      !
RK(K,21)=X(27)*dt      !

```

C...

```

IF(K,GE,3) XK=1.
X(1)=X(1)+RK(K,1)/XK
X(2)=X(2)+RK(K,2)/XK
X(3)=X(3)+RK(K,3)/XK
X(4)=X(4)+RK(K,4)/XK
X(5)=X(5)+RK(K,5)/XK
X(6)=X(6)+RK(K,6)/XK
X(7)=X(7)+RK(K,7)/XK
X(8)=X(8)+RK(K,8)/XK

```

```

X(10)=X(10)+RK(K,10)/XK
X(11)=X(11)+RK(K,11)/XK
X(13)=X(13)+RK(K,12)/XK
X(14)=X(14)+RK(K,13)/XK
X(16)=X(16)+RK(K,14)/XK
X(17)=X(17)+RK(K,15)/XK
X(19)=X(19)+RK(K,16)/XK
X(20)=X(20)+RK(K,17)/XK
X(22)=X(22)+RK(K,18)/XK
X(23)=X(23)+RK(K,19)/XK
X(25)=X(25)+RK(K,20)/XK
X(26)=X(26)+RK(K,21)/XK
ENDDO

```

C State Estimate Propagation, x(-) calculation

```

XM(1)= XM(1) + RK(1,1)/6.+RK(2,1)/3.+RK(3,1)/3.+RK(4,1)/6.
XM(2)= XM(2) + RK(1,2)/6.+RK(2,2)/3.+RK(3,2)/3.+RK(4,2)/6.
XM(3)= XM(3) + RK(1,3)/6.+RK(2,3)/3.+RK(3,3)/3.+RK(4,3)/6.
XM(4)= XM(4) + RK(1,4)/6.+RK(2,4)/3.+RK(3,4)/3.+RK(4,4)/6.
XM(5)= XM(5) + RK(1,5)/6.+RK(2,5)/3.+RK(3,5)/3.+RK(4,5)/6.
XM(6)= XM(6) + RK(1,6)/6.+RK(2,6)/3.+RK(3,6)/3.+RK(4,6)/6.
XM(7)= XM(7) + RK(1,7)/6.+RK(2,7)/3.+RK(3,7)/3.+RK(4,7)/6.
XM(8)= XM(8) + RK(1,8)/6.+RK(2,8)/3.+RK(3,8)/3.+RK(4,8)/6.
XM(9)= XM(9) + RK(1,9)/6.+RK(2,9)/3.+RK(3,9)/3.+RK(4,9)/6.
XM(10)=XM(10)+RK(1,10)/6.+RK(2,10)/3.+RK(3,10)/3.+RK(4,10)/6.
XM(11)=XM(11)+RK(1,11)/6.+RK(2,11)/3.+RK(3,11)/3.+RK(4,11)/6.
XM(13)=XM(13)+RK(1,12)/6.+RK(2,12)/3.+RK(3,12)/3.+RK(4,12)/6.
XM(14)=XM(14)+RK(1,13)/6.+RK(2,13)/3.+RK(3,13)/3.+RK(4,13)/6.
XM(16)=XM(16)+RK(1,14)/6.+RK(2,14)/3.+RK(3,14)/3.+RK(4,14)/6.
XM(17)=XM(17)+RK(1,15)/6.+RK(2,15)/3.+RK(3,15)/3.+RK(4,15)/6.
XM(19)=XM(19)+RK(1,16)/6.+RK(2,16)/3.+RK(3,16)/3.+RK(4,16)/6.
XM(20)=XM(20)+RK(1,17)/6.+RK(2,17)/3.+RK(3,17)/3.+RK(4,17)/6.
XM(22)=XM(22)+RK(1,18)/6.+RK(2,18)/3.+RK(3,18)/3.+RK(4,18)/6.
XM(23)=XM(23)+RK(1,19)/6.+RK(2,19)/3.+RK(3,19)/3.+RK(4,19)/6.
XM(25)=XM(25)+RK(1,20)/6.+RK(2,20)/3.+RK(3,20)/3.+RK(4,20)/6.
XM(26)=XM(26)+RK(1,21)/6.+RK(2,21)/3.+RK(3,21)/3.+RK(4,21)/6.

```

```

C...
  DEALLOCATE(RK)

C  DETERMINATION OF H=dt/dx .....
C
  H(1,10) = 1.
C...
  H(2,13) = 1.
C...
  H(3,7) = G*SIN(XM(7))*COS(XM(8))
  H(3,8) = G*COS(XM(7))*SIN(XM(8))
  H(3,16) = -1.
C...
  H(4,4) = 1.0D0
C...
  H(5,5) = 1.0D0
  H(5,31) = 1.0D0
C...
  H(6,6) = 1.0D0
C...
  C1 = XM(1)-XM(6)*Y1+XM(5)*Z1      ! u
  C2 = XM(2)+XM(6)*X1-XM(4)*Z1     ! v
  C3 = XM(3)-XM(5)*X1+XM(4)*Y1     ! w
  C4 = SQRT(C1*C1 + C2*C2 + C3*C3)   ! V (Airspeed)
  H(7,1) = C1/C4
  H(7,2) = C2/C4
  H(7,3) = C3/C4
  H(7,4) = (-Z1*C2+Y1*C3)/C4
  H(7,5) = (Z1*C1-X1*C3)/C4
  H(7,6) = (-Y1*C1+X1*C2)/C4
  H(7,28) = 1.0D0
C...
  C5 = 1. + (XM(3)/XM(1))**2
  C6 = 1./C5
  Vt=SQRT(XM(1)*XM(1)+XM(2)*XM(2)+XM(3)*XM(3))
  Vt3=Vt*Vt*Vt
  H(8,1) = -C6*XM(3)/(XM(1)**2) + d*XM(5)*XM(1)/Vt3
  H(8,2) = d*XM(5)*XM(2)/Vt3
  H(8,3) = C6/XM(1) + d*XM(5)*XM(3)/Vt3
  H(8,5) = -D/Vt
  H(8,29) = 1.0D0
C...
  C7 = 1./SQRT(1.-(XM(2)/Vt)**2)
  H(9,1) = -C7*XM(2)*XM(1)/Vt3
  H(9,2) = C7/Vt-C7*XM(2)*XM(2)/Vt3
  H(9,3) = -C7*XM(2)*XM(3)/Vt3
  H(9,30) = 1.0D0
C...
  H(10,9) = 1.0D0
C...
  H(11,7) = 1.0D0

```

```

C...
  H(12,8) = 1.0D0
C...
  K1=K1+1
  WRITE(12,REC=K1) H(1,10),H(2,13),H(3,7),H(3,8),H(3,16),H(4,4),
+ H(5,5),H(5,31),H(6,6),H(7,1),H(7,2),H(7,3),H(7,4),H(7,5),
+ H(7,6),H(7,28),H(8,1),H(8,2),H(8,3),H(8,5),H(8,29),H(9,1),
+ H(9,2),H(9,3),H(9,30),H(10,9),H(11,7),H(12,8)
C
C** GAIN CALCULATION
C
  ALLOCATE (PD(31,12))
  DO I=1,31
  DO J=1,12
  PD(I,J)=0.D0          ! PARTIAL PRODUCT PD=P*HT
  DO K=1,31
  PD(I,J)=PD(I,J) + PM(I,K)*H(J,K)
  ENDDO
  ENDDO
  ENDDO

  IN=12
  ALLOCATE (HPHT(IN,IN))
  ALLOCATE (HPHT1(IN,IN))

  DO I=1,12
  DO J=1,12
  HPHT(I,J)=0.D0      ! PARTIAL PRODUCT HPHT=H*P*HT
  HPHT1(I,J)=0.D0
  DO K=1,31
  HPHT(I,J)=HPHT(I,J)+H(I,K)*PD(K,J)
  ENDDO
  ENDDO
  HPHT(I,I) = HPHT(I,I) + RR(I)      ! HPHT + RR
  ENDDO

C... MATRIX INVERSION

  ALLOCATE (WORK(IN,2*IN))
  ALLOCATE (IDENT(IN,IN))
  CALL INVMAT(HPHT,IN,IN,HPHT1,WORK,IDENT) ! INVERSE

  DO I=1,12
  K2=K2+1
  WRITE(13,REC=K2) (HPHT(I,N),N=1,12)
  WRITE(14,REC=K2) (HPHT1(I,N),N=1,12)
  ENDDO

  DEALLOCATE (HPHT,WORK,IDENT)
  ALLOCATE (KK(31,12))
C.....

```

```

DO I=1,31
DO J=1,12
  KK(I,J)=0.D0
  DO K=1,12
    KK(I,J)=KK(I,J)+PD(I,K)*HPHT1(K,J) ! GAIN MATRIX
  ENDDO
ENDDO
ENDDO

DO I=1,31
  K3=K3+1
  WRITE(15,REC=K3) (KK(I,J),J=1,12) ! STORE GAIN
ENDDO

DEALLOCATE (PD,HPHT1)
C
READ(2,*) (ZZ(K),K=1,12),Thrust,FN ! MEASUR. READING
pdot= XM(4)*XM(5)*A11+XM(5)*XM(6)*A12+XM(5)*A13+XM(19)+A14*XM(25)
qdot= XM(4)*XM(6)*A21+(XM(6)*XM(6)-XM(4)*XM(4))*A22- XM(6)*A23
+   +XM(22)
rdot= XM(4)*XM((5)*A31+XM(5)*XM(6)*A32+XM(5)*A3+ XM(25)+A34*XM(19)
Dax = -(ZZ(6)**2+ZZ(5)**2)*X0+(ZZ(4)*ZZ(5)-rdot)*Y0
+   +(ZZ(4)*ZZ(6)+qdot)*Z0 ! ax correction to cg
Day = -(ZZ(4)**2+ZZ(6)**2)*Y0+(ZZ(4)*ZZ(5)+rdot)*X0
+   +(ZZ(5)*ZZ(6)-pdot)*Z0 ! ay correction to cg
Daz = -(ZZ(4)**2+ZZ(5)**2)*Z0+(ZZ(4)*ZZ(6)-qdot)*X0
+   +(ZZ(5)*ZZ(6)+pdot)*Y0 ! az correction to cg.
c
hh(1)=XM(10) - Dax
hh(2)=XM(13) - Day
hh(3)=-XM(16)-G*COS(XM(7))*COS(XM(8)) - Daz
hh(4)=XM(4)
hh(5)=XM(5) + XM(31)
hh(6)=XM(6)
hh(7)= C4 + XM(28) ! V
hh(8)= ATAN(XM(3)/XM(1)) - D*XM(5)/Vt + XM(29) ! ALPHA
hh(9)= ASIN(XM(2)/Vt) + XM(30) ! BETA
hh(10)=XM(9) ! ALT
hh(11)=XM(7) ! THETA
hh(12)=XM(8) ! PHI
C... STATES UPDATE AFTER MEASUREMENT - X(+) Calculation
C
DO I=1,31
  X(I)=0.0
  DO J=1,12
    DH(J)=ZZ(J)-hh(J) ! FILTER RESIDUAL
    X(I)=X(I)+KK(I,J)*DH(J)
  ENDDO
  X(I)=X(I)+XM(I) ! STATE UPDATE
ENDDO
K4=K4+1

```



```

K2=K2+1
READ(14,REC=K2) (HPHT1(I,K),K=1,12)
ENDDO
DO I=1,31
DO J=1,12
PD(I,J)=0.0
DO K=1,12
PD(I,J)=PD(I,J)-H(K,I)*HPHT1(K,J) ! -HT*D1
ENDDO
ENDDO
ENDDO
DO I=1,31
LAM(I)=0.0
DO K=1,12
LAM(I)=LAM(I)+PD(I,K)*DH(K) ! INITIAL LAMBDA
ENDDO
ENDDO
DO I=1,31
DO J=1,31
AA(I,J)=0.0
DO K=1,12
AA(I,J)=AA(I,J)-PD(I,K)*H(K,J) ! INITIAL AA
ENDDO
ENDDO
ENDDO
DEALLOCATE (PD,HPHT1)
K1=K1-1
K4=K4-1
WRITE(8,REC=NAT) (X(K),K=1,31)
C
DO ISAM=NAT-1,1,-1 ! .... BACKWARD SMOOTHER .....
WRITE(*,*) 'SMOOTHER - ITERATION',ISAM
ALLOCATE (HPHT1(12,12))

K2=(ISAM-1)*12
DO I=1,12
K2=K2+1
READ(14,REC=K2) (HPHT1(I,K),K=1,12) ! READ HPHT1
ENDDO

DO I=1,31 ! ZEROING 'F' AND 'H'
DO K=1,31
F(I,K)=0.D0
ENDDO
DO J=1,12
H(J,I)=0.D0
ENDDO
ENDDO
READ(12,REC=K1) H(1,10),H(2,13),H(3,7),H(3,8),H(3,16),H(4,4),
+ H(5,5),H(5,31),H(6,6),H(7,1),H(7,2),H(7,3),H(7,4),H(7,5),
+ H(7,6),H(7,28),H(8,1),H(8,2),H(8,3),H(8,5),H(8,29),H(9,1),

```

+ H(9,2),H(9,3),H(9,30),H(10,9),H(11,7),H(12,8)
K1=K1-1

ALLOCATE (PD(31,12))

DO I=1,31
DO J=1,12
PD(I,J)=0.0
DO K=1,12
PD(I,J)=PD(I,J)-H(K,I)*HPHT1(K,J) ! -HT*D1
ENDDO
ENDDO
ENDDO

DEALLOCATE (HPHT1)

C.....

COST=COS(X(7))
SINT=SIN(X(7))
COSP=COS(X(8))
SINP=SIN(X(8))
TANT=SINT/COST
F(1,2) = X(6)
F(1,3) = -X(5)
F(1,5) = -X(3)
F(1,6) = X(2)
F(1,7) = -G*COST
F(1,10) = 1.0D0

C...

F(2,1) = -X(6)
F(2,3) = X(4)
F(2,4) = X(3)
F(2,6) = -X(1)
F(2,7) = -G*SINT*SINP
F(2,8) = G*COST*COSP
F(2,13) = 1.0D0

C...

F(3,1) = X(5)
F(3,2) = -X(4)
F(3,4) = -X(2)
F(3,5) = X(1)
F(3,7) = -G*SINT*COSP
F(3,8) = -G*COST*SINP
F(3,16) = 1.0D0

C...

F(4,4) = X(5)*A11
F(4,5) = X(4)*A11 + X(6)*A12 + A13
F(4,6) = X(5)*A12
F(4,19) = 1.0D0
F(4,25) = A14

C...

F(5,4) = X(6)*A21-2.*X(4)*A22

```

F(5,6) = X(4)*A21+2.*X(6)*A22-A23
F(5,22) = 1.0D0
C...
F(6,4) = X(5)*A31
F(6,5) = X(4)*A31+X(6)*A32+A33
F(6,6) = X(5)*A32
F(6,19) = A34
F(6,25) = 1.0D0
C...
F(7,5) = COSP
F(7,6) = -SINP
F(7,8) = -X(5)*SINP-X(6)*COSP
C...
F(8,4) = 1.
F(8,5) = TANT*SINP
F(8,6) = TANT*COSP
SECTH2 = 1. + TANT*TANT
F(8,7) = X(5)*SINP*SECTH2 + X(6)*COSP*SECTH2
F(8,8) = X(5)*TANT*COSP - X(6)*TANT*SINP
C...
F(9,1) = SINT
F(9,2) = - COST*SINP
F(9,3) = - COST*COSP
F(9,7) = -(-X(1)*COST-X(2)*SINT*SINP-X(3)*SINT*COSP)
F(9,8) = -(X(2)*COST*COSP-X(3)*COST*SINP)
C...
F(10,11) = 1.0D0
F(11,12) = 1.0D0
F(13,14) = 1.0D0
F(14,15) = 1.0D0
F(16,17) = 1.0D0
F(17,18) = 1.0D0
F(19,20) = 1.0D0
F(20,21) = 1.0D0
F(22,23) = 1.0D0
F(23,24) = 1.0D0
F(25,26) = 1.0D0
F(26,27) = 1.0D0

READ(18,REC=K4) (X(K),K=1,31) !.....

READ(16,REC=K4) (DH(K),K=1,12)
K4=K4-1
C
C ADJOINT VARIABLES DERIVATIVES CALCULATION
C
DO I=1,31
  LAMD(I)=0.0
  DO K=1,31
    LAMD(I)=LAMD(I)-F(K,I)*LAM(K) ! LAMD=-FT*LAM
  ENDDO

```

```

ENDDO
DO I=1,31
DO J=1,31
PM(I,J)=0.D0
DO K=1,31
PM(I,J)=PM(I,J)-F(K,I)*AA(K,J)    ! -(FT*A)
ENDDO
ENDDO
ENDDO

DO I=1,31
DO J=1,31
PM(I,J)=PM(I,J)+PM(J,I)          ! AAD=-(FT*A)-(FT*A)T
ENDDO
ENDDO

DEALLOCATE (F)
ALLOCATE (AAM(31,31))
C
C... ADJOINT VARIABLES INTEGRATION
C
DO I=1,31
LAMB(I)=LAM(I)+LAM(D)*(-DT)    ! SIMPLIFIED INTEGRAT.
DO J=1,31
AAM(I,J)=AA(I,J)+PM(I,J)*(-DT) ! SIMPLIFIED INTEGRAT.
ENDDO
ENDDO

DEALLOCATE (AA)
ALLOCATE (KK(31,12))
ALLOCATE (HPHT(12,12))
C...
K3=(ISAM-1)*31
DO I=1,31
K3=K3+1
READ(15,REC=K3) (KK(I,J),J=1,12) ! READ KK - GAIN
ENDDO

K2=(ISAM-1)*12
DO I=1,12
K2=K2+1
READ(13,REC=K2) (HPHT(I,J),J=1,12) ! READ HPHT
ENDDO
C
C... ADJOINT VARIABLES UPDATING
C
DO I=1,12
hh(I)=0.D0
DO K=1,31
hh(I)=hh(I)+KK(K,I)*LAMB(K)    ! KKT*LAMB
ENDDO

```

```

ENDDO
DO I=1,12
  ZZ(I)=0.D0
  DO K=1,12
    ZZ(I)=ZZ(I)+HPHT(I,K)*hh(K)    ! D*KKT*LAMB
  ENDDO
  ZZ(I)=ZZ(I)+DH(I)                ! (DH+D*KKT*LAMB)
ENDDO

```

```

DO I=1,31
  DLAMB=0.0
  DO K=1,12
    DLAMB=DLAMB+PD(I,K)*ZZ(K)
  ENDDO
  LAM(I)=LAMB(I)+DLAMB            ! LAMBDA UPDATE
ENDDO

```

C...

```

ALLOCATE (F(31,31))

```

```

DO I=1,31
  DO J=1,31
    F(I,J)=0.D0
    DO K=1,12
      F(I,J)=F(I,J)- PD(I,K)*H(K,J)    ! HT*D1*H
    ENDDO
  ENDDO
ENDDO

```

```

ALLOCATE (KH(31,31))
DEALLOCATE (PD)

```

```

DO I=1,31
  DO J=1,31
    KH(I,J)=0.D0
    DO K=1,12
      KH(I,J)=KH(I,J)-KK(I,K)*H(K,J)
    ENDDO
  ENDDO
  KH(I,I)=1.0+KH(I,I)              ! I-KH CALCULATION
ENDDO

```

```

DEALLOCATE (KK,HPHT)
ALLOCATE (PD(31,31))

```

```

DO I=1,31
  DO J=1,31
    PD(I,J)=0.D0
    DO K=1,31
      PD(I,J)=PD(I,J)+AAM(I,K)*KH(K,J) ! AAM*(I-KH)
    ENDDO
  ENDDO

```

```

ENDDO
ENDDO

ALLOCATE (AA(31,31))

DO I=1,31
DO J=1,31
AA(I,J)=0.D0
DO K=1,31
AA(I,J)=AA(I,J)+KH(K,I)*PD(K,J)  ! (I-KH)T*A*(I-KH)
ENDDO
AA(I,J)=AA(I,J)+F(I,J)          ! FINAL A UPDATE
ENDDO
ENDDO

DEALLOCATE (KH)
C
C... STATE SMOOTHING, COVARIANCE UPDATE .....
C
MM=(ISAM-1)*31
DO L=1,31
MM=MM+1
READ(17,REC=MM) (PU(L,J),J=1,31)  ! READ PU (COVARIANCE)
ENDDO

DO I=1,31
DDS=0.0
DO K=1,31
DDS=DDS+PU(I,K)*LAMBDM(K)
ENDDO
XM(I)=X(I)-DDS          ! SMOOTHED STATE
ENDDO

DO I=1,31
DO J=1,31
PD(I,J)=0.D0
DO K=1,31
PD(I,J)=PD(I,J)+AAM(I,K)*PU(K,J)  ! A*PU
ENDDO
ENDDO
ENDDO
DO I=1,31
DO J=1,31
F(I,J)=0.D0
DO K=1,31
F(I,J)=F(I,J)+PU(I,K)*PD(K,J)    ! PU*A*PU
ENDDO
PM(I,J)=PU(I,J)-F(I,J)          ! COVARIANCE - SMOOTHER
ENDDO
ENDDO
WRITE(8,REC=ISAM)(XM(I),I=1,31)

```


C ... TO ADJOIN THE A AND IDENT MATRICES

```
MDASH=2*M
DO 40 I=1,N
  DO 30 J=1,M
    WORK(L,J)=A(L,J)
    WORK(I,M+J)=IDENT(L,J)
30  CONTINUE
40  CONTINUE
```

C ... TO MAKE WORK(1,1)=1.0

```
WKDIV=WORK(1,1)

DO 50 J=1,MDASH
  WORK(1,J)=WORK(1,J)/WKDIV
50  CONTINUE
```

C ... TO MAKE ZEROS BELOW DIAGONAL OF LHS OF MATRIX WORK

```
DO 90 I=2,N
  DO 70 K=I,N
    WKMULT=WORK(K,I-1)
    DO 60 J=1,MDASH
      WORK(K,J)=WORK(K,J)-(WKMULT*WORK((I-1),J))
60  CONTINUE
70  CONTINUE
    WKDIV=WORK(I,I)
    DO 80 J=I,MDASH
      WORK(I,J)=WORK(I,J)/WKDIV
80  CONTINUE
90  CONTINUE
```

C ... TO GET THE UPPER LHS TO ZEROS

```
DO 130 K=N,2,-1

  DO 120 I=1,K-1
    WKMULT=WORK(I,K)

    DO 110 J=1,MDASH
      WORK(I,J)=WORK(I,J)-(WKMULT*WORK(K,J))
110  CONTINUE
120  CONTINUE
130  CONTINUE
```

C ... TO EXTRACT INVERSE MATRIX ON RHS AND

C ... MULTIPLY BY ORIGINAL TO SEE ACCURACY OF IDENTITY

```
DO 150 I=1,N
```

```
      DO 140 J=M+1,MDASH
        AINV(L,J-M)=WORK(L,J)
140   CONTINUE
150   CONTINUE
      RETURN
      END
```

4.2 Program IEKF.FOR

```
C ***** PROGRAM IEKF.FOR *****
C                                     Hoff March/94
C                                     REV. NOV/94, APR/95
C
C Estimation of aircraft states u,v,w,p,q,r,X,Y,Z,L,M,N
C alpha,beta using an Iterated Extended Kalman Filter
C applied to an inertial and gravitational model.
C
C Bias terms to Ax, Az, q, alpha, Beta, Theta.
C
C Needs two data files
c (i) File with initial conditions, process and measurement
C     noise covariances, CG position, aircraft Inertias,
C     Mass and engine z arm to CG.
C (ii) File with flight data: Ax, Ay, Az, p, q, r, True airpeed,
C     Alpha, Beta, Altitude, theta, Phi, Thrust, Propeller
C     Normal Force (metric units, angles in rad.).
C Notation:
C NAT - total number of samples
C DT - sample interval in seconds.
C CG - aircraft CG (%).
C X(i) - updated state variable
C XM(i) - time propagated state variable
C PM(i) - time propagated covariance matrix
C PU(i) - updated covariance matrix
C RR(i) - measurement noise covarianance matrix
C QQ(i) - process noise covariance matrix
C H(i,j) - gradient of the observation model matrix
C F(i,j) - gradient of the dynamic model matrix
C KK(i,j) - Kalman gain matrix
C HPHT - matrix product of H*PM*H'
C HPHT1 - inverse of HPHT
C Work, Ident, PD - auxiliary matrices
C RK - auxiliary matrix of Runge-Kutta integration
C KH,VV,PD - auxiliary matrix
C MASS - Aircraft mass
C Lze - Engine z coordinate relative to CG.

REAL*8 H(12,33),PM(33,33),RR(12),QQ(33),hh(12),ZZ(14),XM(33)
REAL*4 X(33),V(33),Ix,Iy,Iz,Ixz,I2,MASS,Lze
REAL*8 WORK[ALLOCATABLE] (:,:), RK[ALLOCATABLE] (:,:),
+ PD[ALLOCATABLE] (:,:), F[ALLOCATABLE] (:,:),
+ KK[ALLOCATABLE] (:,:), HPHT1[ALLOCATABLE] (:,:),
+ HPHT[ALLOCATABLE] (:,:), KH[ALLOCATABLE] (:,:),
+ PU[ALLOCATABLE] (:,:), IDENT[ALLOCATABLE] (:,:)
INTEGER*2 IN

OPEN(UNIT=4,FILE='PHU809A.DAT',STATUS='OLD')
```

```

OPEN(UNIT=6,FILE='PHU809M.DAT',STATUS='OLD')
OPEN(UNIT=5,FILE='IEKFDAT.DAT')
OPEN(UNIT=7,FILE='IEKDAT1.DAT')
OPEN(UNIT=8,FILE='IEKDAT2.DAT')
OPEN(UNIT=9,FILE='IEKDAT3.DAT')
OPEN(UNIT=10,FILE='IEKDAT4.DAT')
OPEN(UNIT=11,FILE='IEKDAT5.DAT')
OPEN(UNIT=12,FILE='IEKDAT6.DAT')
He=0.0
g=9.80665

```

C... INITIALIZATION - Matrix in a data file with all initial values

C obtained averaging the stabilization

```

READ(6,*) (X(K),K=1,7) !
READ(6,*) (X(K),K=8,14) !
READ(6,*) (X(K),K=15,21) ! STATES INITIALIZATION
READ(6,*) (X(K),K=22,28) !
READ(6,*) (X(k),k=29,33) !

```

C

```

READ(6,*) (RR(I),I=1,6) ! MEASUR. NOISE COVAR. MATRIX
READ(6,*) (RR(I),I=7,12) ! ASSUMED DIAGONAL

```

C

```

READ(6,*) (QQ(J),J=1,7) !
READ(6,*) (QQ(J),J=8,14) ! PROCESS NOISE COVAR. MATRIX
READ(6,*) (QQ(J),J=15,21) ! ASSUMED DIAGONAL
READ(6,*) (QQ(J),J=22,28) !
READ(6,*) (QQ(J),J=29,33) !

```

C

```

ALLOCATE (PU(33,33))

```

C INITIAL COVAR. MATRIX

```

DO J=1,33
DO I=1,33
  PU(I,J)=0.D0 ! Zeroing
ENDDO
ENDDO
jk=0
DO I=1,4
  READ(6,*) (PU(J,J),J=1+jk,7+jk) ! Reading initial cov. matrix
  jk=jk+7
ENDDO
READ(6,*) (PU(J,J),J=29,33)

```

```

READ(6,*) NAT,DT,CG,Ix,Iy,Iz,Ixz,MASS,Lze

```

C

```

Xv = 7.1325 + (CG-30)*1.717/100. ! POSITION INCID./SIDES. VANE
Zv = 0.57
D = SQRT(Xv**2 + Zv**2) ! Vane approx. dist to cg.
X1 = Xv-0.35 ! PITOT POSIT. X m
Y1 = 0. ! " " Y m
Z1 = 0.57 ! " " Z m (APPROX.)

```


$$\begin{aligned}
F(2,3) &= X(4) \\
F(2,4) &= X(3) \\
F(2,6) &= -X(1) \\
F(2,7) &= -G*SINT*SINP \\
F(2,8) &= G*COST*COSP \\
F(2,13) &= 1.
\end{aligned}$$

C...

$$\begin{aligned}
F(3,1) &= X(5) \\
F(3,2) &= -X(4) \\
F(3,4) &= -X(2) \\
F(3,5) &= X(1) \\
F(3,7) &= -G*SINT*COSP \\
F(3,8) &= -G*COST*SINP \\
F(3,16) &= 1.
\end{aligned}$$

C...

$$\begin{aligned}
F(4,4) &= X(5)*A11 \\
F(4,5) &= X(4)*A11 + X(6)*A12 + A13 \\
F(4,6) &= X(5)*A12 \\
F(4,19) &= 1. \\
F(4,25) &= A14
\end{aligned}$$

C...

$$\begin{aligned}
F(5,4) &= X(6)*A21-2.*X(4)*A22 \\
F(5,6) &= X(4)*A21+2.*X(6)*A22-A23 \\
F(5,22) &= 1.
\end{aligned}$$

C...

$$\begin{aligned}
F(6,4) &= X(5)*A31 \\
F(6,5) &= X(4)*A31+X(6)*A32+A33 \\
F(6,6) &= X(5)*A32 \\
F(6,19) &= A34 \\
F(6,25) &= 1.
\end{aligned}$$

C...

$$\begin{aligned}
F(7,5) &= COSP \\
F(7,6) &= -SINP \\
F(7,8) &= -X(5)*SINP-X(6)*COSP
\end{aligned}$$

C...

$$\begin{aligned}
F(8,4) &= 1. \\
F(8,5) &= TANT*SINP \\
F(8,6) &= TANT*COSP \\
SECTH2 &= 1. + TANT*TANT \\
F(8,7) &= X(5)*SINP*SECTH2 + X(6)*COSP*SECTH2 \\
F(8,8) &= X(5)*TANT*COSP - X(6)*TANT*SINP
\end{aligned}$$

C...

$$\begin{aligned}
F(9,1) &= SINT \\
F(9,2) &= -COST*SINP \\
F(9,3) &= -COST*COSP \\
F(9,7) &= -(-X(1)*COST-X(2)*SINT*SINP-X(3)*SINT*COSP) \\
F(9,8) &= -(X(2)*COST*COSP-X(3)*COST*SINP)
\end{aligned}$$

C.....gauss-markov models

$$\begin{aligned}
F(10,11) &= 1. \\
F(11,12) &= 1.
\end{aligned}$$

F(13,14) = 1.
F(14,15) = 1.

F(16,17) = 1.
F(17,18) = 1.

F(19,20) = 1.
F(20,21) = 1.

F(22,23) = 1.
F(23,24) = 1.

F(25,26) = 1.
F(26,27) = 1.

C

C** Gradient of the dynamic model.

C

```
DO I=1,33
DO J=1,33
  FDT= F(I,J)*DT
  FDT2= FDT*FDT
  F(I,J)= FDT+0.5*FDT2
ENDDO
F(I,I) = F(I,I) + 1.0
ENDDO
```

C

C... Covariance time propagation

```
ALLOCATE (PD(33,33))
```

```
DO I=1,33
DO J=1,33
  PD(I,J)=0.D0
DO K=1,33
  PD(I,J)=PD(I,J)+PU(I,K)*F(J,K)  ! P*FT
ENDDO
ENDDO
ENDDO
```

```
DO I=1,33
DO J=1,33
  PM(I,J)=0.D0
DO K=1,33
  PM(I,J)=PM(I,J)+F(I,K)*PD(K,J)  ! P=F*P*FT
ENDDO
ENDDO
PM(I,I) = PM(I,I) + QQ(I)  ! PM=F*PU*FT+Q
ENDDO
```

```
DEALLOCATE(F,PD,PU)
ALLOCATE (RK(4,21))
```

C
 C RUNGE-KUTTA INTEGRATION - DYNAMIC SYSTEM - time propagation

```

XK=2.
DO K=1,4
SINT=SIN(X(7))
COST=COS(X(7))
SINP=SIN(X(8))
COSP=COS(X(8))
TANT=SINT/COST
RK(K,1)= DT*(X(6)*X(2)-X(5)*X(3)-G*SINT + X(10))
RK(K,2)= DT*(X(4)*X(3)-X(6)*X(1)+G*COST*SINP + X(13))
RK(K,3)= DT*(X(5)*X(1)-X(4)*X(2)+G*COST*COSP + X(16))
RK(K,4)= DT*(X(4)*X(5)*A11+X(5)*X(6)*A12+X(5)*A13+X(19)
+
+ A14*X(25))
RK(K,5)= DT*(X(4)*X(6)*A21+(X(6)*X(6)-X(4)*X(4))*A22
+
- X(6)*A23 + X(22))
RK(K,6)= DT*(X(4)*X(5)*A31+X(5)*X(6)*A32+X(5)*A33
+
+ X(25)+A34*X(19))
RK(K,7)= DT*(X(5)*COSP - X(6)*SINP)
RK(K,8)= DT*(X(4)+X(5))*TANT*SINP + X(6)*TANT*COSP)
RK(K,9)=-DT*(-X(1)*SINT+X(2)*COST*SINP+X(3)*COST*COSP)
RK(K,10)=X(11)*DT      !
RK(K,11)=X(12)*dt      !
RK(K,12)=X(14)*DT      !
RK(K,13)=X(15)*dt      !
RK(K,14)=X(17)*DT      !
RK(K,15)=X(18)*dt      ! GAUSS-MARKOV PARAMETERS
RK(K,16)=X(20)*DT      !
RK(K,17)=X(21)*dt      !
RK(K,18)=X(23)*DT      !
RK(K,19)=X(24)*dt      !
RK(K,20)=X(26)*DT      !
RK(K,21)=X(27)*dt      !

```

C...

```

IF(K.GE.3) XK=1.
X(1)=X(1)+RK(K,1)/XK
X(2)=X(2)+RK(K,2)/XK
X(3)=X(3)+RK(K,3)/XK
X(4)=X(4)+RK(K,4)/XK
X(5)=X(5)+RK(K,5)/XK
X(6)=X(6)+RK(K,6)/XK
X(7)=X(7)+RK(K,7)/XK
X(8)=X(8)+RK(K,8)/XK

X(10)=X(10)+RK(K,10)/XK
X(11)=X(11)+RK(K,11)/XK
X(13)=X(13)+RK(K,12)/XK
X(14)=X(14)+RK(K,13)/XK
X(16)=X(16)+RK(K,14)/XK
X(17)=X(17)+RK(K,15)/XK
X(19)=X(19)+RK(K,16)/XK

```



```

C...
  H(6,6) = 1.
C...
  C1 = X(1)-X(6)*Y1+X(5)*Z1      ! u
  C2 = X(2)+X(6)*X1-X(4)*Z1      ! v
  C3 = X(3)-X(5)*X1+X(4)*Y1      ! w
  C4 = SQRT(C1*C1 + C2*C2 + C3*C3) ! V (Airspeed)
  H(7,1) = C1/C4
  H(7,2) = C2/C4
  H(7,3) = C3/C4
  H(7,4) = (-Z1*C2+Y1*C3)/C4
  H(7,5) = (Z1*C1-X1*C3)/C4
  H(7,6) = (-Y1*C1+X1*C2)/C4
C...
  C5 = 1. + (X(3)/X(1))**2
  C6 = 1./C5
  Vt=SQRT(X(1)*X(1)+X(2)*X(2)+X(3)*X(3))
  Vt3=Vt*Vt*Vt
  H(8,1) = -C6*X(3)/(X(1)*X(1)) + d*X(5)*X(1)/Vt3
  H(8,2) = d*X(5)*X(2)/Vt3
  H(8,3) = C6/X(1) + D*X(5)*X(3)/Vt3
  H(8,5) = - d/Vt
  H(8,29) = 1.0
C...
  C7 = 1./SQRT(1.-(X(2)/Vt)**2)
  H(9,1) = -C7*X(2)*X(1)/Vt3
  H(9,2) = C7/Vt-C7*X(2)*X(2)/Vt3
  H(9,3) = -C7*X(2)*X(3)/Vt3
  H(9,30) = 1.
C...
  H(10,9) = 1.
C...
  H(11,7) = 1.
  H(11,32) = 1.
C...
  H(12,8) = 1.
C
C** GAIN CALCULATION
C
  ALLOCATE (PD(33,12))
  DO I=1,33
  DO J=1,12
    PD(I,J)=0.D0      ! PARTIAL PRODUCT PD=P*HT
  DO K=1,33
    PD(I,J)=PD(I,J) + PM(I,K)*H(J,K)
  ENDDO
  ENDDO
  ENDDO

  IN=12
  ALLOCATE (HPHT(IN,IN))

```

```

ALLOCATE (HPHT1(IN,IN))

DO I=1,12
DO J=1,12
  HPHT(I,J)=0.D0      ! PARTIAL PRODUCT HPHT=H*P*HT
  HPHT1(I,J)=0.D0
  DO K=1,33
    HPHT(I,J)=HPHT(I,J)+H(L,K)*PD(K,J)
  ENDDO
ENDDO
  HPHT(I,I) = HPHT(I,I) + RR(I)      ! HPHT + RR
ENDDO

```

C... MATRIX INVERSION

```

ALLOCATE (WORK(IN,2*IN))
ALLOCATE (IDENT(IN,IN))
CALL INVMAT(HPHT,IN,IN,HPHT1,WORK,IDENT) ! INVERSE
DEALLOCATE (HPHT,WORK,IDENT)
ALLOCATE (KK(33,12))
C
DO I=1,33
DO J=1,12
  KK(I,J)=0.D0
  DO K=1,12
    KK(I,J)=KK(I,J)+PD(L,K)*HPHT1(K,J) ! GAIN MATRIX
  ENDDO
ENDDO
ENDDO
DEALLOCATE (PD,HPHT1)
IF(ITER.EQ.1) THEN
pdot= X(4)*X(5)*A11+X(5)*X(6)*A12+X(5)*A13+X(19)+A14*X(25)
qdot= X(4)*X(6)*A21+(X(6)*X(6)-X(4)*X(4))*A22- X(6)*A23 + X(22)
rdot= X(4)*X((5)*A31+X(5)*X(6)*A32+X(5)*A3+ X(25)+A34*X(19)
  READ(4,*) (ZZ(L),L=1,12),Thrust,FN ! MEASUREMENTS READING
  Dax =-(zz(6)**2+zz(5)**2)*X0+(zz(4)*zz(5)-rdot)*Y0
+   +(zz(4)*zz(6)+qdot)*Z0      ! ax correction to cg
  Day =-(zz(4)**2+zz(6)**2)*Y0+(zz(4)*zz(5)+rdot)*X0
+   +(zz(5)*zz(6)-pdot)*Z0      ! ay correction to cg
  Daz =-(zz(4)**2+zz(5)**2)*Z0+(zz(4)*zz(6)-qdot)*X0
+   +(zz(5)*zz(6)+pdot)*Y0      ! az correction to cg.
  ZZ(1)=ZZ(1)-Dax
  ZZ(2)=ZZ(2)-DaY
  ZZ(3)=ZZ(3)-DaZ
ENDIF
hh(1)=X(10) + X(33)              ! Ax
hh(2)=X(13)                      ! Ay
hh(3)=-X(16)-G*COS(X(7))*COS(X(8)) + X(28) ! Az
hh(4)=X(4)                        ! p
hh(5)=X(5) + X(31)                ! q
hh(6)=X(6)                        ! r

```

```

      hh(7)= C4 ! + X(29)                    ! V
      hh(8)= ATAN(X(3)/X(1)) - D*X(5)/Vt + X(29) ! ALPHA
      hh(9)= ASIN(X(2)/Vt) + X(30)           ! BETA
      hh(10)=X(9)                            ! ALT
      hh(11)=X(7)+x(32)                      ! THETA
      hh(12)=X(8)                            ! PHI
C
C... STATES UPDATE AFTER MEASUREMENT - X(+) Calculation
C
      DO I=1,12
      V(I)=0.0
      DO K=1,33
      V(I)=V(I)+H(I,K)*(XM(K)-X(K))
      ENDDO
      ENDDO
C
      DO I=1,33
      DDX=0.0
      DO J=1,12
      DDX=DDX+KK(I,J)*(ZZ(J)-hh(J)-V(J))
      ENDDO
      X(I)=DDX+XM(I)          ! STATES UPDATING
      ENDDO
      IF(ITER.EQ.1) DEALLOCATE (KK)
      ENDDO ! >>>>>> END OF IEKF ITERATION >>>>>>>>>
C
C... COVARIANCE UPDATE - P(+) Calculation
C
      ALLOCATE (KH(33,33))

      DO I=1,33
      DO J=1,33
      KH(I,J)=0.0
      DO K=1,12
      KH(I,J)=KH(I,J)+KK(I,K)*H(K,J) ! Calculation of K*H
      ENDDO
      KH(I,J)=-KH(I,J)
      IF(I.EQ.J) KH(I,J)=1.+KH(I,J) ! Calculation of I-K*H
      ENDDO
      ENDDO

      ALLOCATE (PU(33,33))

      DO I=1,33
      DO J=1,33
      PU(I,J)=0.D0
      DO K=1,33
      PU(I,J)=PU(I,J)+KH(I,K)*PM(K,J) ! Updated covariance
      ENDDO
      ENDDO
      ENDDO

```



```

C ... M = NUMBER OF COLUMNS (J)
C ... N = M OR CANNOT INVERT THE MATRIX A
C
  N = IAR
  M = IAC
C
C   TO CREATE THE APPROPRIATE IDENTITY MATRIX In=IDENT(N,M)

  DO 20 I=1,N
    DO 10 J=1,M
      IDENT(I,J)=0.0
10  CONTINUE
      IDENT(I,I)=1.0
20  CONTINUE

C ... TO ADJOIN THE A AND IDENT MATRICES

  MDASH=2*M
  DO 40 I=1,N
    DO 30 J=1,M
      WORK(I,J)=A(I,J)
      WORK(I,M+J)=IDENT(I,J)
30  CONTINUE
40  CONTINUE

C ... TO MAKE WORK(1,1)=1.0

  WKDIV=WORK(1,1)

  DO 50 J=1,MDASH
    WORK(1,J)=WORK(1,J)/WKDIV
50  CONTINUE

C ... TO MAKE ZEROS BELOW DIAGONAL OF LHS OF MATRIX WORK

  DO 90 I=2,N
    DO 70 K=I,N
      WKMULT=WORK(K,I-1)
      DO 60 J=1,MDASH
        WORK(K,J)=WORK(K,J)-(WKMULT*WORK((I-1),J))
60  CONTINUE
70  CONTINUE
      WKDIV=WORK(I,I)
      DO 80 J=I,MDASH
        WORK(I,J)=WORK(I,J)/WKDIV
80  CONTINUE
90  CONTINUE

C ... TO GET THE UPPER LHS TO ZEROS

  DO 130 K=N,2,-1

```

```

DO 120 I=1,K-1
  WKMULT=WORK(I,K)

  DO 110 J=1,MDASH
    WORK(I,J)=WORK(I,J)-(WKMULT*WORK(K,J))
110  CONTINUE
120  CONTINUE
130  CONTINUE

C ... TO EXTRACT INVERSE MATRIX ON RHS AND
C ... MULTIPLY BY ORIGINAL TO SEE ACCURACY OF IDENTITY

DO 150 I=1,N
  DO 140 J=M+1,MDASH
    AINV(I,J-M)=WORK(I,J)
140  CONTINUE
150  CONTINUE
RETURN
END

```

4.3 Program EKFDER.FOR

```
C ***** PROGRAM EKFDER *****
C                                     Hoff August/94
C                                     Rev. A Feb/95
C
C Estimation of noise covariance, smoothed states and state
C derivatives using a Kalman-like Filter approach.
C
C NAT  Number of samples in the data file
C NP   Number of parameters in the data file
C LAG  Smoother lag (in sample intervals)
C SV   Initial estimate of measurem. noise
C SIGW Initial estimate of process noise
C DT   Time interval between samples
C ZZ(i,k) Matrix of data to be analysed or smoother output
C PM   Time Propagated Covariance Matrix
C X    Updated State Matrix
C XM   Time Propagated State Matrix
C PU   Updated Covariance Matrix
C A    Data Model Matrix
C Q    Data Model Noise Covariance Matrix
C C    Output Matrix
C CA   Kalman Filter Dynamic Matrix (to deter. time constant)
C KK   Gain Matrix
C SV   Measurement Noise
C SIGW Process Noise
C KH   Auxiliary Matrix
C FF1,FF2 Cost Functions
C LAMBDA Autocorrelation
C Files - AILRUD: Aileron and Rudder position.
C         PPLORDER1 : Noise statistics and Kalman dynam. matrices,
C         PPLORDER2 : Temporary data storage.
C         REGRES: Regression data file - pre-fixed format.
C         EKFDERDA: Contains the data to be smoothed. Initial
C                 state values, Covariance matrix, constants
C                 and the data composed by: u,w,q,theta,v,
C                 phi,X,Z,M (from IEFK program).

REAL*8 FF1,FF2,LAMBDA(30),DZ(2015)
REAL*4 X(4,31),XM(4,1),C(4),d(4,4),CA(4),KCA(4,4),XN(4,31)
REAL*4 QQ(4,4),ZZ(2015,20),PUI(4,4),A(4,4),Q(4,4),P(4,31)
REAL*4 PU(4,4,31),PM(4,4,31),KK(4,31),KH(4,4),KHT(4,4),MASS,
+  Iy,MT
OPEN(UNIT=6,FILE='EKFDERDA.DAT',STATUS='OLD')
OPEN(UNIT=7,FILE='PPLORDER1.DAT')
OPEN(UNIT=8,FILE='PPLORDER2.DAT')
OPEN(UNIT=9,FILE='ETAT805.DAT',STATUS='OLD')
OPEN(UNIT=10,FILE='REGRES.DAT')
OPEN(UNIT=11,FILE='FILUWQTH.DAT')
OPEN(UNIT=12,FILE='FILUWQD.DAT')
```



```
OPEN(UNIT=13,FILE='FILXZM.DAT')
```

```
C... INITIALIZATION - Initial value of measurement covariance  
C          and an initial covariance matrix
```

```
  write(*,*) 'READING DATA'
```

```
  READ(6,*) SV,SIGW
```

```
C... INITIAL COVAR. MATRIX
```

```
  DO I=1,4
```

```
    READ(6,*) (PUI(I,J),J=1,4) ! COVAR. INITIAL VALUES
```

```
  ENDDO
```

```
  READ(6,*) NAT,NP,DT,LAG,MASS,Iy
```

```
  NAT34=0.75*NAT
```

```
  IF(LAG.GT.30) LAG=30
```

```
  DO J=1,NAT
```

```
    READ(6,*) (ZZ(J,K),K=1,NP) ! Reading the data
```

```
  ENDDO
```

```
C...
```

```
  A(1,1) = 1.0      !
```

```
  A(1,2) = DT      !
```

```
  A(1,3) = DT*DT/2. !
```

```
  A(1,4) = DT*DT*DT/6. !
```

```
  A(2,1) = 0       !
```

```
  A(2,2) = 1.      !
```

```
  A(2,3) = DT      !
```

```
  A(2,4) = DT*DT/2. ! DATA MODEL (3 DERIVATIVES)
```

```
  A(3,1) = 0.      !
```

```
  A(3,2) = 0.      !
```

```
  A(3,3) = 1.0     !
```

```
  A(3,4) = DT      !
```

```
  A(4,1) = 0.      !
```

```
  A(4,2) = 0.      !
```

```
  A(4,3) = 0.      !
```

```
  A(4,4) = 1.0     !
```

```
C...
```

```
  C(1)=1.          !
```

```
  C(2)=0.          ! OUTPUT MATRIX
```

```
  C(3)=0.          !
```

```
  C(4)=0.          !
```

```
C...
```

```
  Q(1,1) = (DT**7)/252. !
```

```
  Q(1,2) = (DT**6)/72.  !
```

```
  Q(1,3) = (DT**5)/30.  !
```

```
  Q(1,4) = (DT**4)/24.  !
```

```
  Q(2,1) = (DT**6)/72.  !
```

```
  Q(2,2) = (DT**5)/20.  !
```

```
  Q(2,3) = (DT**4)/8.   !
```

```
  Q(2,4) = (DT**3)/6.   !
```

```
  Q(3,1) = (DT**5)/30.  ! DATA MODEL COVAR. MATRIX
```

```
  Q(3,2) = (DT**4)/8.   !
```

```
  Q(3,3) = (DT**3)/3.   !
```

```
  Q(3,4) = (DT**2)/2.   !
```

```

Q(4,1) = (DT**4)/24.    !
Q(4,2) = (DT**3)/6.    !
Q(4,3) = (DT**2)/2.    !
Q(4,4) = DT             !
JK=1
C...
777 DSIG=SIGW/5.
   IC=0
   SX=-1.
   FF20=1.
   DO WHILE (SIGW.GT.0)    ! LOOP1 DETERMIN. OF SIGW & SIGV
888 IC=IC+1
   SIGW2=SIGW*SIGW
   write(*,*) 'sigw',sigw
   DO I=1,4
     DO L=1,4
       QQ(L,L)=SIGW2*Q(L,L)
       PU(L,L,1)=PUI(I,L)
     ENDDO
   ENDDO
   X(1,1)=ZZ(1,JK)        !
   X(2,1)=(ZZ(2,JK)-ZZ(1,JK))/DT ! Initialization of X
   X(3,1)=0.              !
   X(4,1)=0.              !
C...
   DO ISAMPLE=1,NAT      !....Kalman Filter Loop.....
C... STATE AND COVARIANCE PROPAGATION
   DO I=1,4
     XM(I,1)=0.0
     DO K=1,4
       XM(I,1)=XM(I,1)+A(L,K)*X(K,1)    ! STATE PROPAGATION
     ENDDO
   ENDDO
   DO I=1,4
     DO J=1,4
       P(I,J)=0.0
       DO K=1,4
         P(I,J)=P(I,J)+A(I,K)*PU(K,J,1) ! AT*P
       ENDDO
     ENDDO
   ENDDO
   DO I=1,4
     DO J=1,4
       PM(I,J,1)=0.0
       DO K=1,4
         PM(I,J,1)=PM(I,J,1)+P(I,K)*A(J,K) ! AT*P*A
       ENDDO
       PM(I,J,1)=PM(I,J,1)+QQ(L,J)        ! AT*P*A+Q
     ENDDO
   ENDDO

```

```

C
C** GAIN CALCULATION
C
DO I=1,4
  P(I,1)=0.0          ! PARTIAL PRODUCT P=P*HT
  DO K=1,4
    P(I,1)=P(I,1) + PM(I,K,1)*C(K)
  ENDDO
ENDDO
HPHT=0.0             ! PARTIAL PRODUCT HPHT=H*P*HT
DO K=1,4
  HPHT=HPHT+C(K)*P(K,1)
ENDDO
HPHT=HPHT+SV*SV      ! HPHT + SV2

C... MATRIX INVERSION
HPHT1=1.0/HPHT
C... GAIN
DO I=1,4
  KK(I,1)=P(I,1)*HPHT1    ! GAIN MATRIX
ENDDO
C
C... STATES UPDATE AFTER MEASUREMENT - X(+) Calculation
DO I=1,4
  X(I,1)=XM(I,1)+KK(I,1)*(ZZ(ISAMPLE,JK)-XM(1,1)) ! STATE UPDATE
ENDDO

C... COVARIANCE UPDATE - P(+) Calculation
DO I=1,4
  DO J=1,4
    KH(I,J)=-KK(I,1)*C(J)  ! Calculation of K*H
  ENDDO
  KH(I,I)=1.+KH(I,I)      ! Calculation of I-K*H
ENDDO

DO I=1,4
  DO J=1,4
    PU(I,J,1)=0.0
    DO K=1,4
      PU(I,J,1)=PU(I,J,1)+KH(I,K)*PM(K,J,1) ! Covar. Update
    ENDDO
  ENDDO
ENDDO
IF(ISAMPLE.EQ.NAT34) THEN
  WRITE(7,*) JK
  WRITE(7,*) '3*NAT/4 COVAR. & GAIN',PU(1,1,1),KK(1,1)
ENDIF
C
ENDDO    !<<<<<< END OF K-F FILTER LOOP >>>>>>>>
C
WRITE(7,*) ' FINAL COVAR. & GAIN',PU(1,1,1),KK(1,1)

```

```

SIGMAR2=0.0
FF1=0
FF2=0
X(1,1)=ZZ(1,JK)
X(2,1)=(ZZ(2,JK)-ZZ(1,JK))/DT
X(3,1)=0
X(4,1)=0
C..
DO ISAM=1,NAT    ! LOOP - AUTOCORRELATION CALC.
DO I=1,4
  XM(I,1)=0.0
  DO K=1,4
    XM(I,1)=XM(I,1)+A(I,K)*X(K,1)  ! State Propagation
  ENDDO
ENDDO
DZ(ISAM)=ZZ(ISAM,JK)-XM(1,1)    ! Residual using K and P
DO K=1,4                    ! from steady-state K-F
  X(K,1)=XM(K,1)+KK(K,1)*DZ(ISAM) ! above - Need to certify s-s
ENDDO
SIGMAR2=SIGMAR2+1./(FLOAT(NAT))*DZ(ISAM)**2
ENDDO
C.....
IF(LAG.GT.0) THEN
  DO N=1,LAG
    LAMBDA(N)=0.0
    DO ISAM=N+1,NAT
      LAMBDA(N)=LAMBDA(N)+1./(FLOAT(NAT-N))*(DZ(ISAM)*DZ(ISAM-N))
    ENDDO
    FF2=FF2+LAMBDA(N)*LAMBDA(N)
  ENDDO
ELSE
  FF2=SIGMAR2*SIGMAR2
ENDIF
FF1=(PU(1,1,1)+SV*SV-SIGMAR2)**2
C...
SIGWB=SQRT((SIGMAR2*SIGW*SIGW)/(PU(1,1,1)+SV*SV))
XLAM=SV/SIGW
SIGVB=XLAM*SIGWB
IF(IC.GE.2) THEN      !
IF(FF2.LT.FF20) THEN !
IF(SIGW.LT.SIG0) THEN !
  SX=-1.             !
ELSE                 !
  SX=1.              !
ENDIF                !
ELSE                 !
IF(SIGW.LT.SIG0) THEN !
  DSIG=DSIG/2.      ! Determination of minimum
  SX=1.              ! FF2. Search for direction
ELSE                 ! and step. Non-efficient !
  DSIG=DSIG/2.      !

```

```

    SX=-1.          !
    ENDIF           !
    ENDIF           !
    ENDIF           !
    SIG0=SIGW       !
    SIGW=SIGW+SX*DSIG !
    IF(SIGW.LE.0.) THEN !
    SIGW=(SIGW+SIG0)/2. !
    DSIG=SIGW/2.     !
    ENDIF
    IF(ABS((FF2-FF20)/FF20).LT.0.00001) GO TO 999
    FF20=FF2
    IF(ABS(SIGW-SIG0).LT.1.0E-04) GO TO 999
    ENDDO    !<<<<<<<<< LAM LOOP
999 CONTINUE
    WRITE(7,112) SV,SIGW,SIGMAR2,FF1,FF2
112 FORMAT(T05,'SV=',E12.6,2X,', SW'E12.5,' SGR2=',E14.7/
+ T05,'FF1=',E14.7,' FF2=',E14.7)
    WRITE(7,114) JK,SIGVB,SIGWB
114 FORMAT(T05,'VARIABLE=',I2,' SIGVB=',E12.7,' SIGWB=',E12.7)
    WRITE(*,114) JK,SIGVB,SIGWB
    SIGW=SIGWB
    SV=SIGVB
    IF(IC.LT.9999) THEN
    DSIG=SIGWB/100.
    IC=9999
    GO TO 888
    ENDIF
C... END OF PROCESS AND MEASUREMENT NOISE DETERMINATION
C
C... CHARACTERISTIC MATRIX
do l=1,4
  ca(l)=0
  do k=1,4
    ca(l)=ca(l)+c(k)*a(k,l)
  enddo
enddo
do I=1,4
  do j=1,4
    kca(i,j)=0
    d(i,j)=0
    kca(i,j)=kk(i,1)*ca(j)
    d(i,j)=a(i,j)-kca(i,j) ! system matrix to determine
    enddo          ! dominant eigenvalue.
  enddo
  write(7,*) 'MATRIX d, PARAMETER',JK
do i=1,4
  write(7,*) (d(i,k),k=1,4)
enddo
write(7,*)

```

C


```

      DDP=DDP+A(L,K)*PM(K,J,1)      ! COVAR. PROPAGATION
    ENDDO
      PM(L,J,1)=DDP+QQ(L,J)        ! PM(K/K-1)
    ENDDO
  ENDDO
C
C** GAIN CALCULATION
C
  DO LL=1,LAG+1
  DO I=1,4
    P(I,LL)=0.0                    ! PARTIAL PRODUCT P=P*HT
    DO K=1,4
      P(I,LL)=P(I,LL) + PM(I,K,LL)*C(K)
    ENDDO
  ENDDO
  ENDDO
  HPHT=0.0                         ! PARTIAL PRODUCT HPHT=H*P*HT
  DO K=1,4
    HPHT=HPHT+C(K)*P(K,1)
  ENDDO
  HPHT=HPHT+SV2                    ! HPHT + SV2
  HPHT1=1.0/HPHT                   ! INVERSION
C...
  DO LL=1,LAG+1
  DO I=1,4
    KK(I,LL)=P(I,LL)*HPHT1        ! GAIN MATRIX - KALMAN
  ENDDO
  ENDDO
C
C... STATES UPDATE AFTER MEASUREMENT - X(+) Calculation
C
  DZZ=ZZ(ISAMPLE,JK)-XM(1,1)
  IF(LAG.GT.0) THEN
    DO LL=2,LAG+1
    DO I=1,4
      XN(I,LL)=X(I,LL-1)+KK(I,LL)*DZZ ! STATE UPDATE - SMOOTHER
    ENDDO
  ENDDO
  ENDDIF

  DO I=1,4
    X(I,1)=XM(I,1)+KK(I,1)*DZZ    ! STATE UPDATE - K.FILTER
  ENDDO
C...
  DO I=1,4
  DO J=1,4
    KH(I,J)=-KK(I,1)*C(J)
    KHT(I,J)=-C(I)*KK(J,1)
  ENDDO
  KH(I,I)=1. + KH(I,I)
  KHT(I,I)=1. + KHT(I,I)        ! I-K*H

```



```

READ(9,*) ELEV,THR      ! READ ELEVATOR, THRUST
u2=ZZ(I,1)**2
w2=ZZ(I,3)**2
X=ZZ(I,13)*MASS ! X Force non normalised
Z=-ZZ(I,15)*MASS ! Z Force " "
MT=ZZ(I,17)*Iy ! M Moment " "
c write u,w,q,theta,wdot,udot,u2,w2,eta,qdot,shp,x,z,m
WRITE(10,*) ZZ(I,1),ZZ(I,3),ZZ(I,5),ZZ(I,7),ZZ(I,4),ZZ(I,2),
+ u2,w2,ELEV,ZZ(I,6),THR,X,Z,MT ! Final regression
! data file.
WRITE(11,*) ZZ(I,1),ZZ(I,3),ZZ(I,5),ZZ(I,7)
WRITE(12,*) ZZ(I,2),ZZ(I,4),ZZ(I,6)
WRITE(13,*) X,Z,MT
ENDDO
CLOSE(6)
CLOSE(7)
CLOSE(8)
STOP
END

```

4.4 Program MSR.FOR

```
C*** PROGRAM MOD. STEPWISE REGRESSION - MSR.FOR *****
C
C           HOFF JUN/93
C           REV. DEC/93,DEC/94
C
C  NAT = NUMBER OF SAMPLES OF EACH VARIABLE
C  NV  = MAXIMUM NUMBER OF INDEPENDENT VARIABLES
C  IVV = ACTUAL NUMBER OF VARIABLES IN THE MODEL
C  NN  = ACTUAL NUMBER OF VARIABLES IN THE DATA ARCHIVE
C  ISTAT(IV) = DEFINE STATUS OF THE VARIABLE IN THE MODEL
C           IF.EQ.-1 IS NEGLETED.
C  ISTATU(I) = VARIABLE NUMBER
C  X(L,J) = INDEPENDENT VARIABLES READ FROM FLIGHT DATA
C  XWORK(L,J) = THE X(S) ACTUALLY USED BY THE MODEL
C  Y(I)  = DEPENDENT VARIABLE - FROM FLIGHT DATA
C  YN(I) = NEW DEPENDENT VARIABLE FOR TESTING NEW X TO ENTER
C           NEXT ITERATION.
C  YHAT  = ESTIMATED Y FROM THE REGRESSION MODEL
c  FT   = PARTIAL Fp TESTE
C  XTRX  = MATRIX PRODUCT OF X TRANSPOSE TIMES X
C  XTRY  = MATRIX PRODUCT OF X TRANSPOSE TIMES MATRIX Y
C  B     = VECTOR OF REGRESSION COEFFICIENTS
C  WORK, IDENT = AUXILIARY MATRICES
C  DY    = REGRESSION RESIDUAL
C  SB    = ESTIMATED STANDARD ERROR
C  VAR   = RESIDUAL VARIANCE
C
REAL*4 X(2000,12),Y(2000,3),SB(12),Z(12),FP(12),DY(2000),
+ XWORK(2000,12),XTRY(12),B(12),YN(2000)
REAL*4 YHAT(2000),YNHAT(2000),BTXTRY
REAL*8 XTRX[ALLOCATABLE] (:,:), XTRXI[ALLOCATABLE] (:,:),
+ IDENT[ALLOCATABLE] (:,:), WORK[ALLOCATABLE] (:,:)

REAL*8 SYY,SJY,SJJ,YAVER,MODRJY,RJY,ZAV,RMAX,DYAV,
+ FMIN,FPMIN,R2,F,VAR,RESS
INTEGER*2 ISTAT(11),ISTATU(12),LJ,K,L,M,N,IN,NAT,NV,NN,IV,
+ IVV,ITER,NEWVAR,IT
CHARACTER*1 ICHAR
CHARACTER*12 ARQ
CHARACTER*6 IMOD(12)/
*' Y =',' B0 + ',' B1*X1+', 'B2*X2+', 'B3*X3+',
*' B4*X4+', 'B5*X5+', 'B6*X6+', 'B7*X7+', 'B8*X8+', 'B9*X9+',
*' B10X10'/
LOGICAL PEND
FMIN=5.
IN=1
IOLD=12
PEND=.FALSE.
C
```



```

C
NEWVAR=-3
ITER=0

999 CONTINUE
IVV=0
DO M=1,NV
IF(ISTAT(M).EQ.0) THEN
IVV=IVV+1
ISTATU(IVV)=M-1
ENDIF
ENDDO
ITER=ITER+1

C*** PRINTING THE MODEL

PRINT *,'
WRITE(*,199) ITER
199 FORMAT('*****ITERATION No. ',I2,' *****')
WRITE(8,200)
WRITE(*,201)
200 FORMAT(/T05,'** REGRESSION MODEL:/' )
201 FORMAT(/T05,'REGRESSION MODEL:')
WRITE(8,205) IMOD(1),(IMOD(ISTATU(L)+2),L=1,IVV)
WRITE(*,205) IMOD(1),(IMOD(ISTATU(L)+2),L=1,IVV)
205 FORMAT(T02,11A6)

IF(PEND) GO TO 1111
IF(ITER.GT.2) GO TO 1111
DO N=1,NAT
DO L=1,IVV
XWORK(N,L)=X(N,ISTATU(L)+1)
ENDDO
ENDDO

C
ALLOCATE (XTRX(IVV,IVV))
DO I=1,IVV
DO J=1,IVV
XTRX(I,J)=0
DO K=1,NAT
XTRX(I,J)=XTRX(I,J)+XWORK(K,I)*XWORK(K,J) ! XT*X
ENDDO
ENDDO
ENDDO

C
ALLOCATE (WORK(IVV,2*IVV))
ALLOCATE (IDENT(IVV,IVV))
ALLOCATE (XTRXI(IVV,IVV))
CALL INVMAT(XTRX,IVV,IVV,XTRXI,WORK,IDENT) ! INVERSE XTRX

C
DO I=1,IVV

```

```

XTRY(I)=0
DO J=1,NAT
  XTRY(I)=XTRY(I)+XWORK(J,I)*Y(J,IY)    ! XT*Y
ENDDO
ENDDO
C
DO I=1,IVV
  B(I)=0
  DO J=1,IVV
    B(I)=B(I)+XTRXI(L,J)*XTRY(J) ! ESTIMATED COEFFICIENTS
  ENDDO
ENDDO
C
C*** STATISTICS ***
C
DO I=1,NAT
  YHAT(I)=0
  DO J=1,IVV
    YHAT(I)=YHAT(I)+XWORK(L,J)*B(J) ! IS THE Y ESTIMATED
  ENDDO
ENDDO
C
  BTXTRY=0
  DO I=1,IVV
    BTXTRY=BTXTRY+B(I)*XTRY(I)
  ENDDO
C
  DYAV=0.0
  RESS=0.0
  DO L=1,NAT
    DY(L)=Y(L,IY)-YHAT(L)          ! RESIDUE
    RESS = RESS + DY(L)*DY(L)      ! RESIDUAL SUM SQUARES
    DYAV=DYAV + DY(L)
  ENDDO
  DYAV=DYAV/NAT
  SYI=0.0
  DO L=1,NAT
    SYI=SYI+(DY(L)-DYAV)**2
  ENDDO
C
  VAR=RESS/FLOAT(NAT-IVV)          ! RESIDUAL VARIANCE
C
  DO K=1,IVV
    SB(K)=SQRT(VAR*XTRXI(K,K))    ! ESTIMATED STD ERROR
  ENDDO

  F=(BTXTRY-NAT*(YAVR**2))/(VAR*(IVV-1)) ! F VALUE

  R2=F/((NAT-IVV)/(IVV-1)+F)      ! CORRELATION COEF.
C
C  COVARIANCE MATRIX

```

```

IF(ITER.EQ.1) THEN
WRITE(9,*) (B(K),K=1,IVV)
DO I=1,IVV
DO J=1,IVV
XTRX(I,J)=VAR*XTRXI(I,J)
ENDDO
WRITE(9,*) (XTRX(I,K),K=1,IVV) ! FOR FIRST MODEL ONLY
ENDDO
ENDIF
C
C *** PRINTING THE SIGNIFICATIVE PARAMETERS
C
WRITE(8,*) ''
DO K=1,IVV
WRITE(8,210) ISTATU(K),B(K),SB(K)
210 FORMAT(T05,'VARIABLE X',I2,' COEF.Bj =',F14.5,' STD ERROR',
* E12.6)
ENDDO
WRITE(8,215) R2,F,RESS,VAR
215 FORMAT(/T05,'CORRELATION COEF. "R2".... =',F10.6/
* T05,'"F" COEFFICIENT..... =',E12.6/
* T05,'RESIDUAL SUM OF SQUARES... =',E12.6/
* T05,'RESIDUAL VARIANCE..... =',E12.6)
C
C<<<<<< VARIABLE TO BE REJECTED >>>>>>>>>>>>>>>>>>>>>>>>>>>>
C THE NULL CASE :

IF(ITER.EQ.1.AND.F.LT.FMIN) THEN
PRINT *, 'ALL B(J)=0 - REGRESSION ABORTED'
WRITE(8,220)
220 FORMAT(/T05,'* REGRESSION ABORTED: F LOWER THAN Fmin *')
GO TO 1111
ENDIF
WRITE(8,222)
222 FORMAT(/T05,'PARTIAL CORRELATION COEFFICIENTS'/)
C
C*** PARTIAL TEST - FP -- VARIABLE TO BE REJECTED
C
IT=0
FPMIN=FMIN
DO J=1,IVV
FP(J)=(B(J)**2)/(SB(J)**2)
IF(FP(J).LT.FPMIN) THEN
FPMIN=FP(J)
IT=ISTATU(J)+1 ! THE LAST WILL BE THE REJECTED
ENDIF
WRITE(8,224) ISTATU(J),FP(J)
224 FORMAT(T05,'FP(',I2,') ..... =',E10.4)
ENDDO
IF(IT.NE.0) THEN
PRINT *, 'VARIABLE',IT-1, ' WILL BE REJECTED'

```

```

PRINT *, 'DO YOU WANT TO HOLD VAR.', IT-1, ' IN THE REGRESSION'
CALL SREAD(ICHAR)
IF(ICHAR.EQ.'S') GO TO 1111
IF(ICHAR.EQ.'Y') THEN
  PRINT *, 'VARIABLE', IT-1, ' HELD'
  WRITE(8,225) IT-1
225  FORMAT(/T05,'VARIABLE ',I2,' HELD IN THE REGRESSION'/)
  IT=0
  GO TO 230
ENDIF
IF(IT.EQ.NEWVAR-1) THEN
  PRINT *, ' * LAST INTRODUCED VARIABLE WAS REJECTED *'
  WRITE(8,226) IT-1
226  FORMAT(/T05,'LAST INTRODUCED VARIABLE WAS REJECTED..X',I2/)
  ISTAT(IT)=-2
  GO TO 230
ELSE
  PRINT *, ' ONE VARIABLE REJECTED ', IT-1
  WRITE(8,228) IT-1
228  FORMAT(/T05,'ONE VARIABLE REJECTED....X',I2/)
  PRINT *, ' REPROCESS WITHOUT THE REJECTED VARIABLE ? '
  CALL SREAD(ICHAR)
  IF(ICHAR.EQ.'S') GO TO 1111
  ISTAT(IT)=-2      ! RESET STATUS VARIABLE TO REJECT
  IF(ICHAR.EQ.'Y') THEN
    DEALLOCATE(XTRX,XTRXI,WORK,IDENT)
    WRITE(8,*) ' NEW REGRESSION WITHOUT REJEC. VARIABLE'
    WRITE(8,*) ''
    GO TO 999
  ENDIF
ENDIF
ELSE
  PRINT *, ' NO VARIABLE REJECTED'
  WRITE(8,229)
229  FORMAT(/T05,'NO VARIABLE REJECTED')
  IF(IVV.EQ.NV) THEN
    WRITE(*,*) ' NO MORE VARIABLES TO BE INCLUDED'
    GO TO 1111
  ENDIF
ENDIF
C
C<<<<< IDENTIFICATION NEW VARIABLE TO INCLUDE IN THE MODEL >>>>>
C
230 WRITE(8,231)
231 FORMAT(/T05,'ANALYSIS OF NEW VARIABLES '/')
  NEWVAR=-3
  RMAX=0.0
C
DO L=1,NV
  IF(ISTAT(L).EQ.0.OR.ISTAT(L).EQ.-2) GO TO 1000
  DO J=1,NAT

```

```

    YN(J)=X(J,L)    ! NEW INDEPENDENT VARIABLE
  ENDDO

C  REGRESSION
DO J=1,IVV
  XTRY(J)=0
  DO I=1,NAT
    XTRY(J)=XTRY(J)+XWORK(L,J)*YN(I)    ! XT*Y (NEW Y)
  ENDDO
ENDDO
DO I=1,IVV
  B(I)=0
  DO J=1,IVV
    B(I)=B(I)+XTRXI(L,J)*XTRY(J)      ! REGRES. COEFF.
  ENDDO
ENDDO
DO I=1,NAT
  YNHAT(I)=0
  DO J=1,IVV
    YNHAT(I)=YNHAT(I)+XWORK(L,J)*B(J)  ! NEW Y ESTIMATE
  ENDDO
ENDDO

ZAV=0.0
DO I=1,NAT
  Z(I)=YN(I)-YNHAT(I)                ! RESIDUE
  ZAV=Z(I)+ZAV                       ! AVERAGE RESIDUE
ENDDO
ZAV=ZAV/FLOAT(NAT)
SJY=0.0
SJJ=0.0
DO I=1,NAT
  SJY=SJY+(Z(I)-ZAV)*((Y(L,IY)-YHAT(I))-DYAV)
  SJJ=SJJ+(Z(I)-ZAV)**2
ENDDO
RJY=0.0
IF((SYY*SJJ).NE.0.0) RJY=SJY/SQRT(SYY*SJJ)
MODRJY=DABS(RJY)
WRITE(8,232) L-1,MODRJY
232 FORMAT(T05,'VARIABLE X',I2,',... RJY=',E12.6)
IF(MODRJY.GT.RMAX) THEN              ! NEW VARIABLE SELECTION
  RMAX=MODRJY
  NEWVAR=L-1                        ! IDENTIFY THE CHOSEN VARIABLE
ENDIF
1000 CONTINUE
ENDDO ! END OF NEW VARIABLE CHOICE - RETURN TO THE LOOP

IF(NEWVAR.EQ.-3) THEN
  WRITE(8,235)
235 FORMAT(/T05,'NO MORE VARIABLES TO BE INCLUDED - PROGRAM END'//
+ T05,'FINAL MODEL:')

```



```

MDASH=2*M
DO 40 I=1,N
  DO 30 J=1,M
    WORK(I,J)=A(I,J)
    WORK(I,M+J)=IDENT(I,J)
30  CONTINUE
40  CONTINUE

C ... TO MAKE WORK(1,1)=1.0

  WKDIV=WORK(1,1)

  DO 50 J=1,MDASH
    WORK(1,J)=WORK(1,J)/WKDIV
50  CONTINUE

C ... TO MAKE ZEROS BELOW DIAGONAL OF LHS OF MATRIX WORK

  DO 90 I=2,N
    DO 70 K=I,N
      WKMULT=WORK(K,I-1)
      DO 60 J=1,MDASH
        WORK(K,J)=WORK(K,J)-(WKMULT*WORK((I-1),J))
60  CONTINUE
70  CONTINUE
      WKDIV=WORK(I,I)
      DO 80 J=I,MDASH
        WORK(I,J)=WORK(I,J)/WKDIV
80  CONTINUE
90  CONTINUE

C ... TO GET THE UPPER LHS TO ZEROS

  DO 130 K=N,2,-1

    DO 120 I=1,K-1
      WKMULT=WORK(I,K)

      DO 110 J=1,MDASH
        WORK(I,J)=WORK(I,J)-(WKMULT*WORK(K,J))
110  CONTINUE
120  CONTINUE
130  CONTINUE

C ... TO EXTRACT INVERSE MATRIX ON RHS AND
C ... MULTIPLY BY ORIGINAL TO SEE ACCURACY OF IDENTITY

  DO 150 I=1,N
    DO 140 J=M+1,MDASH
      AINV(I,J-M)=WORK(I,J)
140  CONTINUE

```


4.5 Program MSRH.FOR

```
C*** PROGRAM MOD. STEPWISE REGRESSION *****
C
C NEW VERSION WITH HOUSEHOLDER TRANSFORMATION
C
C           HOFF, AUGUST/1993, Rev. Fev.95
C
C NAT = NUMBER OF SAMPLES OF EACH VARIABLE
C NV = MAXIMUM NUMBER OF INDEPENDENT VARIABLES
C IVV = ACTUAL NUMBER OF VARIABLES IN THE MODEL
C NN = ACTUAL NUMBER OF VARIABLES IN THE DATA FILE (MAX=11)
C NY = NUMBER OF DEPENDENT VARIABLES IN THE DATA FILE
C   NY MAX. EQUAL TO 3.
C ISTAT(IV) = DEFINE STATUS OF THE VARIABLE IN THE MODEL
C   IF.EQ.-1 IS NEGLETED.
C ISTATU(I) = VARIABLE NUMBER
C X(L,J) = INDEPENDENT VARIABLES READ FROM FLIGHT DATA
C XWORK(L,J) = THE X(s) ACTUALLY USED BY THE MODEL AND AUGMENTED
C   MATRIX.
C Y(I,k) = DEPENDENT VARIABLE - FROM FLIGHT DATA
C YHAT = ESTIMATED Y VALUE (BY THE REGRESSION MODEL).
C B = REGRESSION COEFFICIENTS
C DY = REGRESSION RESIDUAL
C SB = ESTIMATED STANDARD ERROR
C VAR = RESIDUAL VARIANCE
C U = AUXILIARY MATRIX

REAL*4 X(2000,12),Y(2000,3),Z(12),SB(12),FP(12),DY(2000),
+ XWORK(2000,15),YHAT(2000),XHAT(2000),V(2000),
+ U(12,12),COV(12,12),B(12),XTRY(12)

REAL*8 SYY,SJY,SJJ,YAVER,MODRJY,RJY,ZAV,RMAX,DYAV,
+ FMIN,FPMIN,R2,F,VAR,RESS
INTEGER*2 ISTAT(12),ISTATU(12),L,J,K,L,M,N,NAT,NV,NN,IV,
+ IVV,ITER,NEWVAR,IT
CHARACTER*1 ICHAR
CHARACTER*12 ARQ
CHARACTER*7 IMOD(13)/
*' Y =,' B0 + ',' B1*X1+', ' B2*X2+', ' B3*X3+',
*' B4*X4+', ' B5*X5+', ' B6*X6+', ' B7*X7+', ' B8*X8+', ' B9*X9+',
*' B10X10+', ' B11X11'/
LOGICAL PEND
FMIN=5.
PEND=.FALSE.

C
C*** DATA READING ***
C
PRINT *, 'ENTER DATA FILE NAME'
READ(*,777) ARQ
```



```

999 CONTINUE
  IVV=0
  DO M=1,NV
    IF(ISTAT(M).EQ.0) THEN
      IVV=IVV+1
      ISTATU(IVV)=M-1
    ENDIF
  ENDDO
  MM=IVV
  DO M=1,NV
    IF(ISTAT(M).NE.0) THEN
      MM=MM+1
      ISTATU(MM)=M-1
    ENDIF
  ENDDO
  ITER=ITER+1

C*** PRINTING THE MODEL

  PRINT *,'
  WRITE(*,199) ITER
199 FORMAT('*****ITERATION No. ',I2,' *****')
  WRITE(8,200) IY
  WRITE(*,201) IY
200 FORMAT(/T05,'REGRESSION MODEL FOR INDEPENDENT VARIABLE :,I2,/)
201 FORMAT(/T05,'REGRESSION MODEL FOR INDEPENDENT VARIABLE :,I2,)
  WRITE(8,205) IMOD(1),(IMOD(ISTATU(L)+2),L=1,IVV)
  WRITE(*,205) IMOD(1),(IMOD(ISTATU(L)+2),L=1,IVV)
205 FORMAT(T02,11A7)

  IF(PEND) GO TO 1111
  IF(ITER.GT.50) GO TO 1111

  DO N=1,NAT
    DO L=1,IVV
      XWORK(N,L)=X(N,ISTATU(L)+1)  ! X OF WORK
    ENDDO
    XWORK(N,IVV+1)=Y(N,IY)      ! AUGMENT XWORK WITH Y
    DO K=IVV+1,NV
      XWORK(N,K+1)=X(N,ISTATU(K)+1)  ! AUGMENT WITH REMAINING X
    ENDDO
  ENDDO
C
  DO J=1,IVV
    XTRY(J)=0.0
    DO I=1,NAT
      XTRY(J)=XTRY(J)+XWORK(I,J)*Y(I,IY)  ! XT*Y
    ENDDO
  ENDDO
C
  ZZ=0.

```

```

DO 40 J=1,IVV+1
SIG=ZZ
DO 11 I=J,NAT
V(I)=XWORK(I,J)
XWORK(I,J)=ZZ
11 SIG=SIG+V(I)**2
IF(SIG.LE.ZZ) GO TO 40
SIG=SQRT(SIG)
IF(V(J).GT.ZZ) SIG=-SIG
XWORK(J,J)=SIG
V(J)=V(J)-SIG
SIG=1./(SIG*V(J))
DO 30 K=J+1,NV+1
ALF=ZZ
DO 20 I=J,NAT
20 ALF=ALF+XWORK(I,K)*V(I)
ALF=ALF*SIG
DO 30 I=J,NAT
30 XWORK(I,K)=XWORK(I,K)+ALF*V(I)
40 CONTINUE
C REMOVE THE TRANSFORMED Y (Z) FROM THE A MATRIX
DO L=1,IVV
DO I=1,IVV
U(L,I)=ZZ
ENDDO
ENDDO
C --- INVERSION OF 'A' PRODUCING 'U'-----
U(1,1)=1./XWORK(1,1)
DO 60 L=2,IVV
U(L,L)=1./XWORK(L,L)
JM1=L-1
DO 60 K=1,JM1
SUM=0.0
DO 50 I=K,JM1
50 SUM=SUM-U(K,I)*XWORK(I,L)
60 U(K,L)=SUM*U(L,L)
C ---- SOLUTION OF X=A**-1 * Z OR B=U*Z
DO I=1,IVV
B(I)=0.0
COV(I,I)=0.0
DO L=1,IVV
B(I)=B(I)+U(I,L)*XWORK(L,IVV+1) ! REGRESSION COEFF.
COV(I,I)=COV(I,I)+U(I,L)*U(I,L) ! COV. MATRIX MAIN DIAG.
ENDDO
ENDDO
C
C*** STATISTICS ***
C
DO L=1,IVV
DO N=1,NAT
XWORK(N,L)=X(N,ISTATU(L)+1) ! XWORK REDEFINITION

```

```

        ENDDO
    ENDDO
C
    DO I=1,NAT
        YHAT(I)=0
        DO J=1,IVV
            YHAT(I)=YHAT(I)+XWORK(L,J)*B(J)    ! Y ESTIMATED
        ENDDO
    ENDDO
C
    BTXTRY=0
    DO J=1,IVV
        BTXTRY=BTXTRY+B(J)*XTRY(J)
    ENDDO
C
    DYAV=0.0
    RESS=0.0
    DO L=1,NAT
        DY(L)=Y(L,IY)-YHAT(L)                ! RESIDUE
        RESS = RESS + DY(L)*DY(L)            ! RESIDUAL SUM SQUARES
        DYAV=DYAV + DY(L)
    ENDDO
    DYAV=DYAV/NAT
    SYY=0.0
    DO L=1,NAT
        SYY=SYY+(DY(L)-DYAV)**2
    ENDDO
C
    VAR=RESS/(NAT-IVV)                        ! RESIDUAL VARIANCE
C
    DO K=1,IVV
        SB(K)=SQRT(VAR*COV(K,K))             ! ESTIMATED STD ERROR
    ENDDO

    F=(BTXTRY-NAT*(YAVR**2))/(VAR*(IVV-1)) ! F VALUE

    R2=F/((NAT-IVV)/(IVV-1)+F)              ! CORRELATION COEF.
C
C *** PRINTING THE SIGNIFICANT PARAMETERS/STATISTICS
C
    WRITE(8,*) ''
    WRITE(8,*) ''
    DO K=1,IVV
        WRITE(8,210) ISTATU(K),B(K),SB(K)
210  FORMAT(T05,'VARIABLE X',I2,' COEF.Bj =',F14.5,' STD ERROR',
* E12.6)
    ENDDO
    WRITE(8,215) R2,F,RESS,VAR
215  FORMAT(/T05,'CORRELATION COEF. "R2".... =',F10.6/

```



```

228  FORMAT(/T05,'ONE VARIABLE REJECTED....X',I2/)
      PRINT *, 'REPROCESS WITHOUT THE REJECTED VARIABLE ?'
      CALL SREAD(ICHAR)
      IF(ICHAR.EQ.'S') GO TO 1111
      ISTAT(IT)=-2      ! RESET STATUS VARIABLE TO REJECT
      IF(ICHAR.EQ.'Y') THEN
        WRITE(8,*) ' NEW REGRESSION WITHOUT REJEC. VARIABLE'
        WRITE(8,*) ''
        GO TO 999
      ENDIF
    ENDIF
  ELSE
    PRINT *, ' NO VARIABLE REJECTED'
    WRITE(8,229)
229  FORMAT(/T05,'NO VARIABLE REJECTED')
    ENDIF
C
C<<<<< IDENTIFICATION NEW VARIABLE TO BE INCLUDED TO THE MODEL >>>
C
230  WRITE(8,231)
231  FORMAT(/T05,'ANALYSIS OF NEW VARIABLES '/')
      NEWVAR=-3
      RMAX=0.0
C
C  REGRESSION FOR VARIABLES NO INCLUDED PRESENT MODEL
C  SOLUTION OF  $X=A^{-1} * Z$  OR  $B=U*Z$ 
      DO L=IVV+2,NV+1
      DO I=1,IVV
        B(I)=0.0
        DO J=1,IVV
          B(I)=B(I)+U(I,J)*XWORK(J,L)    ! REGRES. COEFF.
        ENDDO
      ENDDO

      DO I=1,NAT
        XHAT(I)=0
        DO J=1,IVV
          XHAT(I)=XHAT(I)+B(J)*XWORK(L,J) ! NEW Y ESTIMATES
        ENDDO
      ENDDO

      ZAV=0.0
      DO I=1,NAT
        Z(I)=X(L,ISTATU(L-1)+1)-XHAT(I)  ! RESIDUE
        ZAV=Z(I)+ZAV                      ! AVERAGE RESIDUE
      ENDDO
      ZAV=ZAV/FLOAT(NAT)
      SJY=0.0
      SJJ=0.0
      DO I=1,NAT
        SJY=SJY+(Z(I)-ZAV)*((Y(L,IY)-YHAT(I))-DYAV)

```


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