SGP-TR-96

Analysis of Injection-Backflow Tracer Tests

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April 1986

Financial **support** was provided through the Stanford Geothermal Program under Department of Energy Contrac No. DE-AT03-80SF11459 and by the Department of Petrole Engineering, Stanford University



Stanford Geothermal Program Interdisciplinary Research in Engineering and Earth Sciences STANFORD UNIVERSITY Stanford, California

ABSTRACT

Tracer tests have been an important technique for determining the flaw and reservoir characteristics in various rock matrix systems. While the interwell tracer tests are aimed at the characterization of the regions between the wells, single-well injection-backflow tracer tests may be useful tools of preliminary evaluation, before implementing long term interwell tracer tests.

This work is concerned with the quantitative evaluation of the tracer return profiles obtained from single well injection-backflow tracer tests. First, two mathematical models of tracer transport through fractures, have been reviewed. These two model ate based on two different principles: Taylor Dispersion along the fracture and simultaneous diffusion in and out of the adjacent matrix. Then the governing equations for the transport du 3ng the injection backflow tests have been solved. Finally the results were applied to field data obtained from Raft River and East Mesa geothermal fields. In order to determine the values of the parameters of the models that define the transport mechanisms through fractures a nonlinear optimization technique was employed.

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1: INTRODUCTION

Reinjection of waste hot water has been commonly practiced in many **geothermal** reservoirs either **as** a means of disposal or **as** a way to maintain the reservoir **pressure** and liq**uid** volume. In some cases, however, it has been observed that the process had **detrimental** effects such as early breakthrough of the injected fluids and reduction in the enthalpy.

Since both beneficial and detrimental effects are possible, for the design of a successful reinjection program, the mechanisms of the fluid flow in the reservoir have to be understood. Tracer tests have been an important technique for studying the flow mechanisms in the reservoirs. The quantitative analysis of the test results is possible from studies of the mixing curves by using mathematical models describing the transport mechanism in the reservoir.

As far as the flow of tracer through porous media is concerned, a fairly large number of research results have been published. The models developed for the porous media flow, however, are not necessarily applicable to geothermal reservoirs which are usually highly fractured in nature. In search of a description of the tracer transport mechanisms in fractured reservoirs, so far two basic models have been developed. The first model is the convection-dispersion model which is based on purely dispersive flow through the fracture and the second one is the matrix diffusion model which describes a convective flow in the fracture with simultaneous diffusion into the adjacent matrix. Researchers have used both models in the analysis of the return profiles of interwell tracer tests conducted at different geothermal fields.

So far, interwell tracer tests have been a useful technique in determination of the interconnections between the injectors and the producers. The single well (injection-backflow) tests, on the other hand, have been proposed as tools for characterizing the flow field within the radius of influence around the injectors. Even though the injection backflow tests are proposed as preliminary evaluation tools before the employment of long term interwell tests, the amount of information that can be recovered from these tests is potentially **as** much 3 can be obtained from interwell tests.

In this work, a theoretical study of the return profiles from injection-backflow tracer tests is presented. Both convection-dispersion and matrix diffusion models are employed in the analysis of the return profiles of both continuous injection and spike injection cases. In addition the theoretical results of the continuous injection case are applied to the field data from Raft River and East Mesa geothermal fields by using a nonlinear least squares optimization technique, in order to determine the effective parameters of the tracer transport.

2: LITERATURE SURVEY

Tracer tests have been an important technique for the analysis of the flow mechanisms in various rock matrix systems. The quantitative analysis of the test results is possible through the use of mathematical models describing the tracer transport mechanisms in reservoirs. As far as the flow through porous media is concerned the theory is well established.

Johnston and Perkins (1963) summarized the dispersion in porous media and gave the empirical correlations to determine the parameters of the classic convection-dispersion model.

In 1964, *Coats and Smith* modified the one-dimensional convection-dispersion model to obtain a differential capacitance model which allows mass transfer between the mobile and immobile fractions of the liquid phase in the reservoir. With the differential capacitance model they were able to produce the asymmetric concentration profiles observed from the experiments. Later in 1976, van Genuchten and Wierenga, improved the model by including a term for the adsorption taking place in both mobile and immobile regions.

However, the models developed to describe the tracer transport mechanisms in porous media are not applicable to highly fractured geothermal reservoirs. Recent'studies on flow through fractured media, led to the development of new models describing the physics of the tracer transport through fractures.

Grisuk and Pickens (1980) formulated a double porosity model combining a convectivedispersive transport the in fractures and a diffusive transport in the unfractured matrix. A finite element method was developed for simulating non-reactive and reactive solute transport by convection, mechanical dispersion and diffusion in a unidirectional flow field.

In 1981, *Rodriguez and Horne* presented **a** theoretical study of the one-dimensional convective-dispersive flow through fractures. In that work, they derived **an** expression for for the dispersivity in flow through a fracture. *Fossum and Horne* (1982) applied this model to in-

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terwell tracer test data obtained from Wairakei, New Zealand with some success. They used a nonlinear optimization technique to perform matching the model to field data.

In 1982, from the studies of migration of radionuclides in bedrock surrounding, nuclear waste repositories, *Neretnieks, Eriksen and Tahtinen* developed a matrix **diffusion** model describing the tracer movement in a single fissure in granitic rock. Using this matrix **diffusion** model *Jensen and Horne* (1983) were able to obtain a better match model to the data obtained from Wairakei, New Zealand as compared to the dispersion model used by Fossum and Horne (1982).

Walkup and Horne (1985) developed a *two* dimensional model which uses the tracer loss terms of the matrix diffusion model and at the same time allows the direct determination of the fracture aperture. Even though matching to field data with nonlinear curve fitting **was** successful, excessive computer time consumption was a serious limitation,

Single well tracer tests have been used to estimate the immobile fluid saturation to determine the waterflood conformance and to evaluate **the** potential of chemical enhanced oil recovery processes by the oil and gas industry. Recently *Antunez and Brigham* (1983), developed a semi-analytical method to analyze the single well tests in oil reservoirs and presented a set of charts for the estimation of residual oil saturation.

Unlike in oil reservoirs, the use of the single well tracer tests in geothermal reservoirs has started only recently. The interwell tracer tests have been commonly practised to determine the interconnections between the injectors and producers and to characterize the reservoir flow mechanisms. In these interwell tests it has been observed that the tracer breakthrough does not always occur within the expected times. It was also found that the tracers could be physically and chemically restrained by the reservoir matrix rocks or the fluid could be moving through locations where there are no monitor wells. For these reasons the single well (injection-backflow) tracer tests have been proposed as a preliminary evaluation technique before the implementation of the long term interwell tests.

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In 1982, *Downs* and his coworkers presented a preliminary study of the injectionbackflow tests conducted at Raft River Geothermal Field. Later in **1983**, *Capuano et al.* presented the qualitative analysis of the **tests** conducted at both Raft River and East Mesa fields. It was concluded that the injection-backflow tracer **tests** can be successfully used to characterize the flow in the near well-bore environment. In another study of the same tests, Russel et *al.* analyzed the tracer return profiles by using the semi-infinite medium solution to the one-dimensional convection-dispersionmodel.

3: THEORY

Tracer tests have been important tools for the study of the flow mechanisms in geothermal reservoirs. In interwell tests a tracer fluid of constant concentration is injected into the reservoir at one well and the tracer is monitored at other wells. If the fluid arrives at the monitor location then inferences can be made about the reservoir flow conditions between the two wells. The most direct information about the interconnection between the injection and observation wells is obtained from the mean arrival times. **Also**, from the shape of the return profiles, the dispersive characteristics can be analyzed. If the arrival times **are** longer **than** expected, the fluid could be physically and chemically restrained by the matrix rocks, it could be moving through locations where there are no monitor wells, or it could be dispersing in an orderly manner.

In a reference prepared by the University of **Uch** Research Institute Earth Scierces Laboratory, injection-backflow tests have been proposed as a preliminary evaluation technique before the implementation of the long term interwell tests. In the report the following advantages of the injection-backflow tests were listed:

a) A suite of tracers can be easily evaluated and those which interact with the formation can be eliminated and conservative tracers can be established.

b) With the conservative tracer established, the mixing of the injected fluid with the reservoir fluid can be studied. From the analysis of the mixing curves selected reservoir characteristics can be determined.

c) If the injected fluid moves from an area of fractures into a porous matrix or vice versa during the injection flow this may be detectable **as** a discontinuity in the tracer return flow.

d) The chemical reactions between the injected fluid and rock matrix and native fluid can be assessed by analyzing the injected and returning fluids. e) Deposition of the fluids can be assessed by analyzing the solid particles in the injected and returning fluids.

f) With the temperature of **the** injected and returning fluids known, together the mixing of the injected and native fluids and the time in **the** formation, the heat transfer may be **related** to matrix porosity.

g) Regional reservoir flow in the area of the injection well can be assessed by delaying the backflow from the test well. Loss of tracer would indicate a substantial **flow** in the reservoir with the tracer moving beyond the well's radius of influence.

The injection backflow **tests** can be divided into three stages : a) Injection period , b) Shut-in period, c) Backflow **period.** However, if the test is not aimed at the determination of regional flow beyond **the** test well's radius of influence, it may be completed in only **two** stages, injection and immediate backflowing.

Since the transport of tracers through geothermal reservoirs is primarily **through fracture**s, the success of interpretive analysis of the return curves depends on the **understanding the phy**sics of the mixing process during the flow. In this work, two mathematical models based **on** two principal mechanisms, dispersion in fracture and the diffusion into the matrix, were employed to analyze the tracer return profiles from injection and immediate backflowing tests.

The injection-backflow tests can be conducted either injecting a tracer fluid of concentration C_o continuously during the injection period (continuous injection case) or injecting a tracer slug followed by the untraced fluid (slug or spike injection case depending upon the size of the slug). Here, both continuous injection and spike injection cases for both convection-dispersion and matrix diffusion models will be considered.

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3.1 CONVECTION-DISPERSION MODEL

In a fracture, under either laminar or turbulent flow the fluid will be **transported** faster in the center of the fracture than near the walls. The result of this non-uniform "convective" transport is the dispersion of the tracer over the region of the transport. Although this convective smearing of the tracer gives rise to large concentration gradients across the narrow width of the fracture, molecular diffusion tends to rapidly equalize the tracer concentration across the fracture, thus counteracting the effect of convective dispersion.

The combination of the transverse diffusion and convective dispersion in the **flow** channel is known as "Taylor Dispersion" and was derived by Taylor (1953) for pipe flow. The net result of The Taylor Dispersion is that the tracer front propagates with the mean speed of the flow. The net longitudinal dispersivity for the the combination is :

$$\eta = \frac{2}{105} \frac{b^2 u^2}{D}$$
(3.1)

where

b: the fracture aperture

u: the average velocity

D: the molecular diffusion coefficient

as derived by Home and Rodriguez (1983).

Taylor (1953) presents the equation governing the effective longitudinal dispersion in an infinite medium

$$\eta \frac{\partial^2 c}{\partial z^2} = \frac{ac}{\partial t}$$
(3.2)

where

z = x - ut



- u: average velocity
- t: time
- C: concentration
- η : Taylor dispersion coefficient

3.1.1 Continuous Injection Case

For the continuous injection case the boundary and initial conditions are

C=0	at	t = O
$C = C_o$	at	z = - ~
C=0	at	z = ∞

The solution to Eq.3.2 with the above boundary and initial conditions is given by Taylor

$$\frac{C}{C_o} = C_D = \frac{1}{2} \operatorname{erfc}\left(\frac{x - ut}{2\sqrt{\eta t}}\right)$$
(3.3)

Eq. 3.3 represents the concentration profile during the injection period and it is symmetric about x = ut which is the average distance traveled by the front. Also the point x = |ut| corresponds to position of the 50 percent concentration contour. If L_t is the zone of transition in which C_D changes from 0.9 C_D to 0.1 C_D , the expression for L_t , is given by Taylor

$$L_t = 3.62 \sqrt{\eta} t$$

Taylor also mentioned that as t increases L, increase proportionally to $t^{\frac{1}{2}}$ whereas the distance traveled by the particles of fluid are proportional to t. Eventually as t increases L_t become small compared with L = ut which is **the** distance traveled by the moving **plane** traveling with the mean speed of flow, **u**. Therefore there is a minimum injection time requirement for the theory to be applicable. At the end of the injection period of time t_j , the **profile** is given by '

$$\frac{C}{C_o} = \frac{1}{2} erfc \left[\frac{x - utj}{2\sqrt{\eta t_j}} \right]$$
(3.4)

The analysis of the injection backflow tests is not a simple one-dimensional problem, because of the change in flow direction during the backflow period. There is also a possibility of change in the average flow velocity during the backflow period. For these reasons, to obtain the backflow period profile the governing equations have to be solved with appropriate initial and boundary conditions. However, for this specific problem, here we will apply a simpler technique developed by Antunez and Brigham (1983), to obtain the solutions.

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In this case, first with the assumption of equal average flow velocities during injection and backflow periods, the problem is simplified. From Eq. 3.4 the concentration profile at the and of the injection period is given by Fig. 3.1



Fig. 3.1 Dispersion of a Sharp interface Caused by Taylor Dispersion.

The 50 percent concentration point is at a distance of x = L to the injection point x = 0, and the profile is symmetric about this point. At this point, first of all, we have to remember that the front propagates with the mean speed of the flow. To obtain the backflow period solution we utilize both the equal injection and backflow average velocities and the symmetry of the profile. During the backflow period we imagine a pseudo-front going away from the injection point as if the injection period is continuing, while the real front approaches the well. Since the injection and backflow velocities are imagined to be equal for both real and pseudo fronts, the distance traveled by them will also be equal. The schematics of **the** process is **given** in Fig. 3.2



Fig. 3.2 Positions of Real and Pseudo Fronts.

Then the concentration of the pseudo-profile at any distance \mathbf{x} is given by

$$\frac{C_p}{C_o} = \frac{1}{2} \operatorname{erfc}\left(\frac{x - u(t_j + t_p)}{2\sqrt{\eta(t_j + t_p)}}\right)$$
(3.5)

If the concentrations of the pseudo-profile were to be calculated at $x = 2L = 2\mu t_j$, then

$$\frac{C_p}{C_o} = \frac{1}{2} \operatorname{erfc} \left\{ \frac{u(t_j - t_p)}{2\sqrt{\eta(t_j + t_p)}} \right\}$$
(3.6)

When the 50 percent concentration of the pseudo-front reaches to x = 2L, the same concentration of the real front reaches to well which is the measurement point. Therefore, the concentrations measured at the well can be calculated by using the pseudo-front concentrations evaluated at x = 2L. The pseudo concentration C_p is related to the actual concentration C_p by

$$\frac{C_r}{C_o} = 1 - \frac{C_p}{C_o} = 1 - \frac{1}{2} erfc \left\{ \frac{u(t_j - t_p)}{2\sqrt{\eta(t_j + t_p)}} \right\}$$
(3.7)

Therefore

$$\frac{C_r}{C_o} = \frac{1}{2} + \frac{1}{2} erf\left\{\frac{u(t_j - t_p)}{2\sqrt{\eta(t_j + t_p)}}\right\}$$
(3.8)

where

- t_j : injection time
- t_p : production time
- u: the average velocity
- $\eta\,$: the dispersion coefficient

Defining
$$\mathbf{a} = \frac{\mu}{\sqrt{\eta}}$$
 and $C_D = \frac{C_r}{C_o}$ Eq.3.8 can be rewritten as

$$C_{D} = \frac{1}{2} + \frac{1}{2} erf\left(\frac{\alpha(t_{j} + t_{p})}{2\sqrt{t_{j} + t_{p}}}\right)$$
(3.9)

3.1.2 Spike Injection Case

In tracer studies, it is more common to inject small slugs than continuously injaction of the tracer fluid, due to the cost of operation. When a slug is injected into a reservoir the concentration can be expressed as the difference between the two dispersed fronts which are called the leading and the trailing edges. Since the flow of a tracer slug through a fracture is linear one-dimensional flow, the concentration profile can be expressed as,

$$\frac{C}{C_o} = \frac{1}{2} \operatorname{erfc}\left(\frac{x - ut_1}{2\sqrt{\eta t_1}}\right) - \frac{1}{2} \operatorname{erfc}\left(\frac{x - ut_2}{2\sqrt{\eta t_2}}\right)$$
(3.10)

If the size of the slug is very small compared to the distance traveled by it, in the limit the solution can be found as follows. The response to a plane source of unit strength concentrated at x = x' = ut in an infinite medium, is given by the Green's Function Solution,

$$C = \frac{1}{2\sqrt{\pi\eta t}} e^{\frac{-(x-\omega)^2}{4\eta t}}$$
(3.11)

If the source is of strength s (mass/unit area), then the response is

$$C = \frac{S}{2\sqrt{\pi\eta t}} e^{\frac{-(x-\omega)^2}{4\eta t}}$$
(3.12)

The tracers **are** injected in tracer fluids of a specified concentrations C_o . Then if **a** volume of $V = AL_1$ tracer fluid is injected, the mass of the tracer is given by

 $M = C_o V = C_o A L_1$

$$S = \frac{M}{A} = \frac{C_o A L_1}{A} = C_o L_1$$
(3.13)

Finally the concentration distribution is given by

$$\frac{C}{C_o} = \frac{L_1}{2\sqrt{\pi\eta t}} e^{\frac{-(x-w)^2}{4\eta t}}$$
(3.14)

1

where L_1 is the length of the slug.

However, since the fracture dimensions **are** unknown we do not **know** the slug length entering the reservoir. Therefore we rewrite **Eq.3.14** by using more explicit variables. The **to**-tal volume of tracer fluid can be represented by

V = injection rate*time of injection of the traced fluid

Then, source strength can be represented as

$$S = \frac{C_o Q t_i}{A} = C_o t_i \frac{Q}{A} = C_o t_i u$$

Now, we can rewrite Eq.3.14 as

$$\frac{C}{C_{o}} = \frac{ut_{i}}{2\sqrt{\pi\eta t}} e^{\frac{-(x-ut)^{2}}{4\eta t}}$$
(3.15)

Eq. 3.15 represents the concentration profile during the injection period for the spike injection case. An important feature of Eq. 3.15 is that it is symmetric about x = ut, which corresponds to the peak concentration. To obtain the concentration profile at the end of the injection period the variable t in Eq. 3.15 is replaced by t_{j} .

$$\frac{C}{C_o} = \frac{ut_i}{2\sqrt{\pi\eta t_j}} e^{\frac{-(x-u_j)^2}{4\eta t_j}}$$
(3.16)

The schematics of the Eq. 3.15 is given in Fig. 3.3



Fig. 3.3 Dispersion of a Slug Caused by Taylor Dispersion.

The concentrations measured at the well during the backflow period **can** be calculated as follow. During the backflow period we assume **-as** we did for the continuous injection casea pseudo-slug going away from the injection point as if the injection process is continuing. Assuming the injection and backflow velocities are equal, the distances traveled by the peak concentrations of real and pseudo slugs will be the same; as it is shown in Fig. 3.4



Then the concentration profile of the pseudo slug at any backflow time is given by

$$\frac{C}{C_o} = \frac{ut_i}{2\sqrt{\pi\eta(t_j + t_p)}} e^{\frac{-(x - u(t_j + t_p))^2}{4\eta(t_j + t_p)}}$$
(3.17)

If the distance is set to $x = 2L = 2ut_j$, the expression becomes

$$\frac{C}{C_o} = \frac{ut_i}{2\sqrt{\pi\eta(t_j + t_p)}} e^{\frac{-u^2(t_j - t_p)^2}{4\eta(t_j + t_p)}}$$
(3.18)

Since the profiles **are** symmetric about the **peak** point concentrations, the calculated values of pseudo slug concentration at x = 2L, will be equal exactly to the concentrations measured at the well. Hence, the solution becomes,

$$\frac{C}{C_o} = \frac{ut_i}{2\sqrt{\pi\eta(t_j + t_p)}} e^{\frac{-u^2(t_j - t_p)^2}{4\eta(t_j + t_p)}}$$
(3.19)

where

- t_i: tctal injection time
- t_p : production time
- **u** : the average velocity
- η : the dispersion coefficient
- t_i : length of the slug injection time

Defining a new parameter $\mathbf{a} = \frac{\mu}{\sqrt{\eta}}$ and $C_D = \frac{C}{C_o}$, Eq.3.19 becomes

$$C_D = \frac{C}{C_o} = \frac{\alpha t_i}{2\sqrt{\pi(t_j + t_p)}} e^{\frac{-\alpha^2 (t_j - t_p)^2}{4(t_j + t_p)}}$$
(3.20)

3.2 MATRIX DIFFUSION MODEL

When a tracer fluid flows in a fracture, the tracer will diffuse **into** the porous matrix adjacent to the fracture. For very large spacing between **the** fissures, the one-dimensional form of the equation of the diffusion into the porous matrix is given by

$$\frac{\partial}{\partial y} \left[\phi D_a \frac{\partial C_p}{\partial y} \right] = \phi \frac{\partial C_p}{\partial t}$$
(3.21)

Assuming the porosity ϕ , and the apparent diffusion coefficient D_a are constant throughout the matrix contacted by the fluid the **Eq. 3.21** takes the form,

$$D_a \frac{\partial^2 C_p}{\partial y^2} = \frac{\partial C_p}{\partial t}$$
(3.22)

When **the** source of the tracer fluid is discontinued the effect will be to flush the fracture and reverse the concentration gradient causing tracer to migrate from the matrix **into** the fracture.

Assuming the concentration profile across the fracture is evened out **due** to molecular diffusion, the flow and sorption from the water in fracture is represented by

$$\frac{\partial C_f}{\partial t} + u \frac{\partial C_f}{\partial x} = 2 \frac{D_{\iota}}{\delta} \frac{\partial C_f}{\partial y} \bigg|_{y=0}$$

The two diffusion coefficients D, and D_e in Eq. 3.22 and 3.23 respectively, are related as follow:

$$D_a = \frac{D_a}{K_d \rho_b} \tag{3.24}$$

The effective diffusion coefficient $D_{r,r}$ is dependent on temperature, porosity, molecular diffusivity and the geometry of the rock. $K_{a}\rho_{b}$ is a volumetric sorption equilibrium constant, and is related to porosity ϕ , the solid rock density ρ_{s} and the adsorption distribution coefficient k by,

$$K_d \rho_b = \phi + (1 - \phi)k\rho_s \tag{3.25}$$

If the solids are inert, i.e k = 0 which we assumed in this work, the volumetric sorption equilibrium constant of the mamx becomes equal to its porosity. Then, the **Eq.3.25** reduces to

$$D_a = \frac{D_a}{\phi} \tag{3.26}$$

Equations 3.22 and 3.23 are the system of equations describing the physical situation of one-dimensional convective flow through a fracture with simultaneous tracer diffusion into the surrounding porous matrix. To be able to determine the concentration profile\$, Equations 3.22 and 3.23 have to be solved simultaneously. Here the solutions for both continuous injection and spike injection cases will be presented.

3.2.1 Continuous Injection Case

For a constant solute source of C_o at x = 0, initially the media are saturated with fluids free from the tracer, the boundary and initial conditions are given as,

$C_f = C_p = 0$	at	<i>t</i> = 0
$C_f = C_0$	at	<i>x</i> = 0
$C_p = C_f$	at	<i>y</i> = <i>o</i>
$C_p \rightarrow 0$	as	y → ∞

If we assume $C_o = 1$ then the solutions we will obtain will be in terms of normalized concentrations. By using the Laplace transform method, we obtain the solution in Laplace space

$$\overline{C}_{f} = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\frac{2\sqrt{D}\epsilon^{\phi}}{\delta u}x\sqrt{s}}$$
(3.27)

$$\overline{C}_{p} = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\left[\frac{2\sqrt{D_{e}\phi}x}{\delta u} + \sqrt{\frac{\phi}{D_{e}}}y\right]\sqrt{s}}$$
(3.28)

1

where s is the Laplace parameter, and Equations **3.26** and **3.27** are the solutions in Laplace space for the concentrations in fracture and porous matrix respectively.

The solutions in real space are

$$C_{f} = erfc \left| \frac{\sqrt{D_{e} \phi}}{\delta} \frac{x}{u} \frac{1}{\sqrt{t - \frac{x}{u}}} \right| \qquad \text{for} \quad t > \frac{x}{2}$$
(3.28)

$$C_{p} = erfc \left\{ \left[2 \frac{\sqrt{D_{e}\phi}}{\delta} \frac{x}{u} + \sqrt{\frac{\phi}{D_{e}}} y \right] \frac{1}{2\sqrt{t - \frac{x}{u}}} \right\} \qquad \text{for} \quad t > \frac{x}{u} \qquad (3.29)$$

and both C_f and C_p are zero for $t < \overline{x}$.

Equations 3.28 and 3.29 represent the injection period concentration profiles in the fracture and in the porous matrix respectively. At the end of the injection period of time t_j , the solutions are,

$$C_{f} = erfc \left[\frac{\sqrt{D_{e}\phi}}{\delta} \frac{x}{u} \frac{1}{\sqrt{t_{j} - \frac{x}{u}}} \right] \qquad \text{for} \quad t_{j} \ge \frac{x}{u} \tag{3.30}$$

for $t_j < \frac{x}{u}$

and

 $C_f = 0$

$$C_{p} = erfc \left\{ \left[2 \frac{\sqrt{D_{e}\phi}}{\delta} \frac{x}{u} + \sqrt{\frac{\phi}{D_{e}}} y \right] \frac{1}{2\sqrt{t_{j} - \frac{x}{u}}} \right\} \quad \text{for} \quad t_{j} \ge \frac{x}{u} \quad (3.31)$$

and

$$C_p = 0$$
 for $t_j < \frac{x}{u}$

Since the flow direction changes, the backflow period concentrations have to be obtained by solving the governing system of equations with appropriate initial and boundary conditions.

The injection period profile in the fracture will be as in Fig. 3.5



Fig. 3.5 Dispersion of a Sharp Interface Caused by Matrix Diffusion.

In the backflow period, to take the change in the direction of the velocity vector into account, the governing differential equations have to be modified. Here the modification will be done by utilizing the nature of the injection period solution and making a simple coordinate change.

Looking at the profile at the end of the injection period it is seen that the concentration in the fracture is zero after a distance of $x = ut_j = L$, as seen from Fig. 3.5 When the origin of the new coordinate system is chosen at x = L and the new space variable is defined as z, the injection period profile can be expressed in terms of z, by simply replacing x by L - z.

$$C_{f} = erfc \left[\sqrt{\frac{D_{\phi}\Phi}{\delta}} \frac{L-z}{u} \frac{1}{\sqrt{t_{j} - \frac{L-z}{u}}} \right] \qquad \text{for} \quad t_{j} \ge \frac{L-z}{u} \qquad (3.32)$$

$$C_{f} = 0 \qquad \text{for} \quad t_{j} < \frac{L-z}{u}$$

$$C_{p} = erfc \left\{ \left[\frac{2\sqrt{D_{\phi}\Phi}}{\delta} \frac{L-z}{u} + \sqrt{\frac{\Phi}{D_{\phi}}}y \right] \frac{1}{2\sqrt{t_{j} - \frac{L-z}{u}}} \right\}$$

$$\text{for} \quad t_{j} \ge \frac{L-z}{u} \qquad (3.33)$$

$$C_{p} = 0 \qquad \text{for} \quad t_{j} < \frac{L-z}{u}$$

Now with the new coordinate system, the governing differential equation of flow in the fracture becomes

$$\frac{\partial C_f}{\partial \tau} + u \frac{\partial C_f}{\partial z} = \frac{2D_e}{\delta} \frac{\partial C_p}{\partial y} \bigg|_{y=0}$$
(3.34)

where τ is the time coordinate, starting **from** the beginning of the backflow period.

As far as the diffusion of tracer into or out of the matrix is concerned, there is no change in the conditions for constructing the governing equation. Hence, the equation remains the same.

$$D_a \frac{\partial^2 C_p}{\partial y^2} = \frac{\partial C_p}{\partial \tau}$$
(3.35)

the boundary conditions are

$$C_{f} = 0 \qquad at \quad z = 0 \quad \tau \ge 0$$

$$C_{p} = C_{f} \qquad at \quad y = 0 \quad \tau \ge 0$$

$$C_{p} \to \infty \qquad as \quad y \to \infty \quad \tau \ge 0$$

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and the initial conditions are

$$C_{f} = erfc \left[\frac{\sqrt{D_{e}\phi}}{\delta} \frac{L-z}{u} \frac{1}{\sqrt{t_{j} - \frac{L-z}{u}}} \right]$$
(3.36)

at $\tau = 0$ and $0 \le z \le L$

$$C_{p} = erfc \left\{ \left[2 \frac{\sqrt{D_{e} \Phi}}{\delta} \frac{L-z}{u} + \sqrt{\frac{\Phi}{D_{e}}} y \right] \frac{1}{2\sqrt{t_{j} - \frac{L-z}{u}}} \right\}$$
(3.37)

at $\tau = 0$ and $0 \le z \le L$

However, the attempts to obtain the real space solutions to Equations 3.34 and 3.35 failed because of the complexity induced by the initial conditions. For **this** reason, the Laplace transformed forms of the initial conditions will be preserved, and a solution in Laplace space will be obtained.

The initial conditions in Laplace space are

$$\overline{C}_{f} = \frac{1}{s} e^{-\frac{x}{u}s} e^{\frac{-2\sqrt{D_{e}\phi}}{\delta u}x\sqrt{s}}$$
(3.38)

$$\overline{C}_{p} = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\left[\frac{2\sqrt{D_{e}\phi}}{\delta} \frac{x}{u} + \sqrt{\frac{\phi}{D_{e}}}y\right]\sqrt{s}}$$
(3.39)

where s is the Laplace parameter corresponding to time $t = t_i$

Now the initial conditions are given by Equations 3.38 and 3.39. When we take the Laplace transform of Eq. 3.35 according to the time variable, τ ,

$$\frac{\partial^2 \tilde{C}_p(z,s,p)}{\partial y^2} - \frac{p}{D_a} \tilde{C}_f(z,s,p) = -\frac{1}{s D_a} e^{-\beta \sqrt{s}} e^{-xs/u} e^{-\sqrt{\frac{s}{D_a}y}}$$
(3.40)

where $\beta = \frac{2\sqrt{D_{e}\phi} x}{\delta u}$ and p is the Laplace parameter corresponding to time τ . The result is a linear nonhomogeneous differential equation and the solution is given by

$$\tilde{C}_{p}(z,s,p) = \tilde{C}_{f}(z,s,p) e^{-\sqrt{\frac{p}{D_{a}}y}} - e^{-xs/u} \frac{e^{-\beta\sqrt{s}}}{(p-s)} \left[e^{-\sqrt{\frac{p}{D_{a}}y}} - e^{-\sqrt{\frac{s}{D_{a}}y}} \right]$$
(3.41)

using the property that the transform of the derivative of a function is equal to the derivative of the transform of the function in Laplace space,

$$L\left\{\frac{\partial C_p}{\partial y}\right\}_{y=0} = \left\{\frac{\partial \overline{C_p}}{\partial y}\right\}_{y=0}$$
(3.42)

The Laplace transform of the Eq. 3.34 can be written as

$$p\tilde{C}_{f}(z,s,p) - \overline{C}_{f}(z,s) + u \frac{\partial \tilde{C}_{f}(z,s,p)}{\partial z} =$$

$$\frac{2D_{e}}{\delta} \left\{ -\tilde{C}_{f}(z,s,p) \sqrt{\frac{p}{D_{a}}} - \frac{\sqrt{s} - \sqrt{p}}{\sqrt{D_{a}}(p-s)} e^{-\frac{x}{u}s} e^{-\beta\sqrt{s}} \right\} \qquad (3.43)$$

Eq. 3.43 is also a linear nonhomogeneous differential differential equation representing the flow in the fracture. The solution is given by

$$\tilde{C}_{f} = \frac{1}{\frac{S}{S}} \left[1 + \frac{\sqrt{-2\alpha}\sqrt{-s}}{\sqrt{p} + \sqrt{s}} \right] \left[s + p + 2\mathbf{d}(\sqrt{p} + \sqrt{s}) \right]$$

$$\left\{ e^{-2\alpha \frac{(L-s)}{u}\sqrt{s}} - e^{-\frac{(L-s)s}{u}} - e^{-\frac{L}{u}s} - e^{-\frac{s}{u}p} - \frac{2\alpha}{u}(\sqrt{p}s + \sqrt{s}L) \right\}$$
(3.44)

where

$$\alpha = \frac{\sqrt{D_e \phi}}{\delta} \tag{3.45}$$

Since the measurements are made at the well which corresponds to $z = L_f C_f$ takes the form,

$$\tilde{C}_{f} = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right] \left\{ 1 - e^{-\frac{L}{u}p} e^{-\frac{L}{u}s - 2\alpha\frac{L}{u}(\sqrt{p} + \sqrt{s})} \right\}$$
(3.46)

The terms $e^{-\frac{L}{u}p}$ and $e^{-\frac{L}{u}s}$ in Eq. 2.38 will cause Heaviside step function effects $H(t_p - \frac{L}{u})$, and $H(t_p - \frac{L}{u})$ respectively,

where

$$H(x) = \begin{cases} 1 & x > 0 \\ 0 & x < 0 \end{cases}$$
(3.47)

These effects are investigated **as** follows. The solution is in the Laplace space and is two transformations away from the real space. In this space the above solution can be exp sed **as** following

$$\tilde{C}_f = F(s,p,\alpha) - G(s,p,\alpha) e^{-\frac{L}{u}s} e^{-\frac{L}{u}p}$$
(3.48)

In the Laplace space which is one transformation away from the real space the solution must be

$$\overline{C}_{f} = F(s,\tau,\alpha) - G(s,\tau - \frac{L}{v},\alpha) e^{\frac{L}{u}s} \qquad \text{for} \quad \tau > \frac{L}{v} \qquad (3.49)$$

and

$$\overline{C}_f = F(s,\tau,\alpha) - 0 \qquad \text{for} \quad \tau < \frac{L}{u} \tag{3.50}$$

In the real space the form of the solution has to be

$$\overline{C}_{f} = F(t_{j},\tau,\alpha) - G(t_{j} - \frac{L}{v},\tau - \frac{L}{v},\alpha) \quad \text{for} \quad \tau > \frac{L}{v} \quad \text{and} \quad t_{j} > \frac{L}{v} \quad (3.51)$$

and

Since $t_j = \frac{L}{u}$ the second **term** will always be zero, therefore the solution becomes

$$C_f = F(t_j, t_p, \alpha)$$

or in the Laplace space,

$$C_f = F(s, p, \alpha)$$

and in complete form

$$\tilde{C}_{f} = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
(3.53)

3.22 Spike Injection Case

In tracer **tests**, injection of a **finite** size tracer slug is more common practice than the cc **n**tinuous tracer fluid injection. The solution for a slug injection can be obtained by superposing the continuous injection solution. In the matrix diffusion model for the **cont**inuous injection case the concentration profile in the fracture is given by Eq. 3.30

Defining

$$F = erfc \left[\frac{\alpha x}{u\sqrt{t-\frac{x}{u}}} \right]$$
(3.54)

and

 $C_f = C_o F$

the slug injection solution can be written as

$$C_{fs} = C_o \bigg[F(t + \Delta t) - F(\Delta t) \bigg]$$

$$C_o = \frac{M}{Q\Delta t} \tag{3.56}$$

Eq. 3.55 becomes

$$C = \frac{M}{Q\Delta t} \left[F(t + At) - F(t) \right]$$
(3.57)

Taking the limit, with respect to time, to obtain a spike solution,

$$C_{f} = \frac{M}{Q} \lim_{A \to 1} \frac{F(t + At) - F(t)}{At}$$
(3.58)

$$C_f = \frac{M}{Q} \frac{\partial F}{\partial t}$$
(3.59)

from **Eq. 3.54**

$$\frac{\partial F}{\partial t} = -\frac{\alpha x}{u\sqrt{\pi(t_j - \frac{x}{u})^3}} e^{-\frac{a}{u^2(t_j - \frac{x}{u})}}$$
(3.60)

Thus the concentration profile of spike injection in the fracture at the end of the jection period (time t_j), is

$$C_{f} = \frac{M}{Q} \frac{\alpha x}{u \sqrt{\pi(t_{j} - \frac{x}{u})^{3}}} e^{-\frac{\alpha^{2} x^{2}}{u^{2}(t_{j} - \frac{x}{u})}}$$
(3.61)

Similarly the profile in the matrix can be found as

$$C_{f} = \frac{M}{Q} \left[\frac{2\alpha x}{u} + \sqrt{\frac{\Phi}{D_{e}}} y \right] \frac{1}{2\sqrt{\pi(t_{j} - \frac{x}{u})^{3}}} e^{-\frac{\left[\frac{2\alpha x}{u} + \sqrt{\frac{\Phi}{D_{e}}} y \right]^{2}}{4(t_{j} - \frac{x}{u})}}$$
(3.62)

A schematic of the profile is given in Fig. 3.6



Fig. 3.6 Dispersion of a Slug Caused by Matrix Diffusion.

For the backflow period, the governing equations of the transport are the same as the equations for the continuous injection case. The boundary conditions are also the same, bu the initial conditions are different. The solution for this case will also be presented in Laplac space, due to the complexity induced by the initial conditions.

The initial conditions in the Laplace space are

$$\overline{C}_{f} = \frac{M}{Q} e^{-\frac{xs}{u}} e^{-\left[\frac{2\alpha x}{u}\right]\sqrt{s}}$$
(3.63)

$$\overline{C_p} = \frac{M}{Q} \qquad e^{-\left[\frac{2\alpha x}{\mu} + \sqrt{\frac{\phi}{D_e}y}\right] \sqrt{a}}$$
(3.64)

Since the governing equations and boundary conditions are the same, and the only difference between the initial condition Equations 3.38 and 3.63 and 3.39 and 3.64 is that $\frac{1}{s}$ term in the

former equations is replaced by $\frac{M}{Q}$ in the later ones, the solution can be written immediately from the continuous injection case.

$$\tilde{C}_{f} = \frac{M}{Q} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
(3.65)

4: ANALYSIS TECHNIQUE

The goal of **the** tracer tests analysis is to determine the parameters defining the flow and reservoir characteristics. In a tracer test, a fluid of constant concentration is **injected** into the reservoir at one well for **a** period of time and produced back from other wells or the same well. Then the return profiles are analyzed both qualitatively and **quantitatively**. This is an inverse type problem in which the input (injection) and output (return profiles) are used in interpreting the system (reservoir), whereas in **a** forward type problem the input and the systems characteristics **are** utilized to predict the output.

The quantitative analysis of an inverse problem is done by using **an** optimization technique to fit the model to data. In **this** work a nonlinear least squares curve fitting method was employed to estimate the parameters. The curve fitting was done by using a **computer program VARPRO**, (from the Stanford Center for Information Technology) which computes the optimal values for both linear and nonlinear parameters of a given function.

The method of optimization in **VARPRO** is based **on** a paper by Go**lub** and Pereya. They showed that the fit of a nonlinear model of the form

$$C(E,\alpha_{j},t) = \sum_{j=1}^{M} E_{j} \phi_{j}(\alpha_{i};t) + \phi_{M+1}(\alpha_{i};t) \qquad i = 1,2,3,....,$$
(4.1)

where

- C: the model to be fit
- E: linear parameters
- α : vector of nonlinear parameters
- t: independent variable
- M : number of linear parameters
- **•** : nonlinear functions

can be done by a nonlinear least squares method by optimizing the linear and **nonlinear** param
eters separately.

The objective function, $R(E_j, \alpha_j)$

$$R(E_{j},\alpha_{j}) = \sum_{i=1}^{N} \left\{ C_{i} - \tilde{C}(E_{j},\alpha_{j},t) \right\}^{2}$$

$$(4.2)$$

where

- C_i: ob erved concentrations
- \bar{C} : calculated concentrations
- N : number of observations

is first minimized by using the initial estimates of the non-linear parameters. Then, the linear least squares procedure is applied to the residual R to obtain the values of the linear parameters. The next step is the modification of the residual by substituting the values of linear parameters. After the modification, the residual is minimized with respect to the nonlinear parameters. When the optimal values of the nonlinear parameters are computed, the linear parameters are recovered immediately. The details and the proof of the technique are discussed by *Fossum* (1982).

Since the routine uses **a** Taylor expansion, the derivatives of the objective function with respect to the nonlinear parameters have to be evaluated.

In the analysis the following input data are required :

- 1) N observed concentrations (C_i)
- 2) time values, t, corresponding to each observation
- 3) initial guesses of the nonlinear parameters

All these input data are called by the main program, which also calls for VARPRO and print the final results. The subroutine ADA called by VARPRO, calculates the solution with respect to the nonlinear parameters.

The solution for the continuous injection case of convection-dispersion model in the form of Eq. 4.1 is,

$$C(\alpha,t) = C_o F(t,\alpha) = C_o \frac{1}{2} \operatorname{erfc}\left(\frac{\alpha(t_j - t_p)}{2\sqrt{t_j + t_p}}\right)$$
(4.3)

where

C (a,t): the concentration at a time t,

 C_o : the concentration at inlet during the injection period

- F: the nonlinear function
- **a** : the nonlinear parameter

If we normalize **the** concentration dividing by C_o , we obtain

$$C_D(\alpha,t) = E F(t,\alpha) \tag{4.4}$$

where E is a normalization factor that normalizes the concentration to 1. **The** program which performs the the nonlinear least squares fitting is listed in appendix E.

Similarly the solution for the continuous injection case of the matrix diffusion model is given by

$$C_{f}(\alpha,t_{i},t_{p}) = C_{o}F(\alpha,s,p)$$
(4.

where

 C_f : concentration at time t

$$C_{o}$$
: concentration at the inlet during the injection period

F: solution in (z,s,p) space (nonlinear function)

Since the solution is available only in (z,s,p)-space the calculation of the solution and the derivatives in real space must be done numerically. Therefore, ADA needs to perform a dc ble numerical inversion

Here the procedure of the numerical evaluation technique of the solutions and derivativ will be presented by utilizing the continuos injection solution. The solution obtained at z = L, i.e at the wellbore was

$$\tilde{C}_f = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right] \qquad \text{for} \quad t_p < t_j \tag{4.6}$$

where the parameters, s, and p are the corresponding Laplace parameters of the real variables

 $t_j = \frac{L}{u}$, t_p , respectively. Eq. **4.6** is analytic only in p-space. Since the p-space is two Laplace transformations away from the real space, **ADA** utilizes **three** function subprograms to perform a double numerical inversion process by using Stehfest algorithm(1970).

The main program CURFIT reads the values of the nonlinear parameters, the time steps at which the solution is to be evaluated, number of terms, **N**, that **are** going to be used in Stehfest algorithm and calls function **COEFF** to provide the vector v of **N** elements. Then it transfers all these parameters to subroutine **ADA**, where **the** solution is evaluated and **ADA** uses the Stehfest algorithm for the inversion of the solution from (z,s,t_p) space to **real** space. To perform the inversion, **ADA** needs the evaluation of the solution in (z,s,t_p) space and it calls the function **CDS**. The direct evaluation of the solution in (z, s, t_p) space can not be done because the solution is analytic only in (z,s,p) space. Therefore **CDS** performs another humerical inversion from (z,s,p) space to (z,s,t_p) space to (z,s,t_p) space. The evaluation of the solution of the solution of the solution in (z,s, p) space. The same procedure is applied for the evaluation of the derivative\$ except **DCDS** and **DCDSP** are used for the evaluations in (z,s,t_p) and (z,s,p) spaces respectively.

5: APPLICATIONS

The quantitative interpretation of the tracer return profiles can be accomplished through the application of the mathematical models to field data. **So** far, three different mathematical models based on different principles of dispersion in fractured media have been developed and employed in the analysis of interwell tracer tests, by other researchers. In the previous sections, two of these models have been discussed and extended for the analysis of the return profiles from the single well injection-backflow tracer tests.

Now, the results from the application of the models to field data will be presented. Both models were curve fitted to four sets of data obtained from two wells in two different geothermal fields. The curve fitting was done by using a nonlinear optimization program. Each model has only one nonlinear parameter to be determined through curve fitting.

The first two sets of the data were from a well in **Reft** River geothermal field. One of the sets was a 4-hour injection test and the other was a 48.5-hour injection test. The other two sets of data were from a well in East Mesa geothermal field. The first one of these sets was a 7.22-hour injection test and the injection period for the second one is 14.22 hours. From the analysis of these sets we were not only able to compare the ability of the mddels to represent the flow in the reservoir, but also analyze the effect of the injection period on the return profiles.

The the results of the curve fittings are shown in figures from Fig. 5.1 to Fig. 5.8. The first observation that we made is the matrix diffusion model gives better fits to all available data than the convection-dispersion model does. As far as the convection-dispersionmodel \mathbf{k} -self is concerned it gives far better results on the small injection period tests. While the duration of injection increases fits get worse, for the convection-dispersion model.

The mamx diffusion model, on the other hand, fits all sets of the data equally well and the fits are excellent However, there is one important point has to be paid attention. **The** nonlinear parameter of the matrix diffusion model is given by,

$$\alpha = \frac{\sqrt{D_e \phi}}{\delta} \tag{5.1}$$

In section 3.2 it was stated that D_{e} is a function of temperature, porosity, molecular diffusivity and the geometry of the rock. It was also assumed that the porosity and the fracture aperture is constant along the path traveled by the tracer fluid. Therefore, the values of the **nonlinear** parameters obtained from the analysis of the data sets of the same well have to be the same. The reason for the differing numerical values can be found in the effects of the Injection periods. Since the assumption of constant fracture aperture and uniform porosity is **not** absolutely true, what we obtain is **an** average value of those physical properties over the distance traveled by the tracer fluid. Therefore, the longer the injection period the longer the distance

The poor fits obtained from the application of the convection-dispersion model can be explained **as** flow. If the injection time is **short**, then the amount of the tracer diffusing into the fracture will not be high, so the length of diffusion. Hence, the contribution to the dispersion **within** the fracture will come from the Taylor Dispersion discussed earlier in section **3.1.** As the injection period increases, the effect of the interaction between the adjacent matrix and the fluid in the fracture becomes the dominant mechanism of dispersion. For this reason, convection-dispersionmodel fails to give a good fit to the data obtained from the long Injection period tests.

The last point to be considered here is the non-unit normalized concentration value even at, $t_p = 0$, for the fit of convection-dispersion model to the data of well 2C which is shown in Fig. 5.4. In order to explain this we need to go check the injection time constraint explained in section 3.1.1. The length of the transition zone was,

$$L_{r} = 3.62 \sqrt{\eta t} \tag{5.2}$$

It was also stated that as t increases L_r proportionally to $t^{\frac{1}{2}}$ whereas the distance traveled by the particles of fluid are proportional to t. Eventually L_r will become small cdmpared with L. Now let's look at the condition for the injection time to be satisfied so that the infinite medium solution can be applied. From the point of view of numerical calculations the argument of the error function must be greater than or equal to 2, for the value of the function to be 1. Therefore, to be able to get a unit C_D at the injection point, the following has to be satisfied. At the end of the injection period, the argument was

$$\frac{x-ut_j}{2\sqrt{\eta t_j}}$$

and at x = 0, we want the following inequality to be satisfied

$$\frac{L}{2\sqrt{\eta t_j}} \ge 2 \tag{5.3}$$

since it was defined that $a = \frac{u}{\sqrt{\eta}}$ and $L=ut_j$. We obtain

$$\alpha \sqrt{t_j} \ge 4 \tag{5.4}$$

If we look at the values given in Fig. 5.4 we see that

$$\alpha\sqrt{t_i} = 1.33 < 4 \tag{5.5}$$

Now there is two possible explanations can be given for this result:

1) The injection time is not enough for the theory to applicable

2) The recovered value of the nonlinear parameter is not correct. In other **words** the model itself is not applicable.

Of course the second explanation is the logical one because of the inability of the model to represent the long injection period tests **as** explained above.







Fig. 5.1 The Result of Curve Fitting the Convection-Dispersion Model to the

Data From well #19 test-4



Fig. 6.3 The Result of Curve Fitting the Convection-Dispersion Model to the

Data From well #2 test-a



Fig. 5.4 The Result of Curve Fitting the Convection-Dispersion Model to the Data From well #2 test-c



Fig. 5.5 The Result of Curve Fitting the Matrix Diffusion Model to the Data From well #19 test-4





From well #19 test-6



Fig. 5.7 The Result of Curve Fitting the Matrix Diffusion Model to the Data

From well #2 test-a





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6: CONCLUSIONS

This study **has** shown that the injection-backflow tracer tests can be used in determining the dispersion characteristics of the area within the radius of influence of the test well. Two mathematical models describing the tracer **transport** in fractured medium have been extended to the quantitative analysis of the return profiles of injection-backflow tracer **tests**. Then the models were used to match six sets of field data. From these fits it was seen that most of the profiles can be successfully matched by the **matrix** diffusion model, whereas only short **term** injection tests could be fitted by the convection-dispersionmodel. One suspects that the reason for this that in a short injection backflow test, the time might not be sufficient for the tracer to diffuse far enough into the porous matrix to produce the long tails in the return profile. In the case of relatively high porosity or long injection periods the return profiles are expected to be long tailed, and can be matched well by the matrix-diffusion model.

It was also observed that the parameters of models determined from **fits** to different data sets obtained from the same well gave different results. The differences between the two values of the same parameter were small if the injection periods were not large. In other words, the larger the difference between the injection periods the higher the difference between the determined values of the same parameter. Therefore, the injection periods should made as long **as** possible to obtain better average values for the governing parameters of the transport models.

7: NOMENCLATURE

CONVECTION-DISPERSION MODEL

- δ : average velocity
- D: molecular diffusivity
- η : dispersivity coefficient
- C: concentration (mass of the tracer per unit volume of the traced fluid)

 $C_D = \frac{C}{C_o}$: normalized concentration

- *x* : distance along the **flow** direction
- t_j : total injection time
- t_p : time variable of production period
- t_i : slug injection time
- z = x ut: moving space coordinate
- $\alpha = \frac{u}{\sqrt{\eta}}$; nonlinear parameter of the solution equation

MATRIX DIFFUSION MODEL

- C_p : concentration in matrix adjacent to fracture
- C_f : concentration in fracture
- C_o : initial concentration of the traced fluid
- D_a : apparent diffusion coefficient

- D_e : effective diffusion coefficient
- $K_d \rho_b$: volumetric sorption equilibrium constants
- ϕ : porosity of adjacent matrix
- ρ_s : solid rock density
- k: adsorption distribution coefficient
- x: distance along the flow direction during injection period
- y: distance normal to the flow direction
- *z* : distance along the flow direction during backflow period
- $L = ut_i$: distance of the front from injection point at the end of the injection period
- M: mass of tracer material
- A: area open to flow
- Q: volumetric injection rate of traced fluid
- t_j : total injection time
- τ : time variable of production period
- *p* : Laplace parameter corresponding to production time
- **s** : Laplace parameter corresponding to time t_j

 $a = \sqrt{\frac{D_e \phi}{\delta}}$: nonlinear parameter of the solution equation

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APPENDIXES

APPENDIX A: DERIVATION OF THE CONTINUOUS INJECTION CASE SOLUTION TO MATRIX-DIFFUSION MODEL

The system of governing differential equations are,

$$\frac{\partial C_f}{\partial t} - \frac{2D_e}{\delta} \left[\frac{\partial C_p}{\partial y} \right]_{y=0} + u \left[\frac{\partial C_f}{\partial x} \right] = 0$$
(A1)

$$D_a \frac{\partial^2 C_p}{\partial y^2} = \frac{\partial C_p}{\partial t} \tag{A2}$$

The boundary and initial conditions are,

$C_{f} = C_{p} = 0$	at	t = o	
$C_f = C_0$	at	<i>x</i> = 0	
$C_p = C_f$	at	y = 0	
$C_p \rightarrow 0$	as	y → ∞	

If we take $C_o = 1$ then the solutions we will obtain will be the concentrations normalized by C_o . Taking the Laplace Transform of Eq. (A2)

$$D_{a}\frac{\partial \overline{C}_{p}}{\partial y^{2}} = s\overline{C}_{p} - C_{p}(x,0)$$
(A3)

Since $C_p(x,0) = 0$, Eq. (A3) takes the form

$$D_a \frac{\partial^2 \overline{C}_p}{\partial y^2} - s \overline{C}_p = 0 \tag{A4}$$

The solution of the above homogeneous differential equation is given by

$$\overline{C}_{p} = A \ e^{-\sqrt{\frac{s}{D_{a}}}y} + Be^{\sqrt{\frac{s}{D_{a}}}y}$$
(A5)

Applying the boundary conditions

$$C_p \to 0$$
 as $y \to \infty$ implies $B = O$

and

$$C_{p} = C_{f} \qquad at \qquad y = 0 \quad implies \qquad A = \overline{C}_{f}$$

$$\overline{C}_{p} = \overline{C}_{f} e^{-\sqrt{\frac{s}{D_{a}}}y} \qquad (A6)$$

$$L\left\{\frac{\partial C_p}{\partial y}\right\} = \frac{\partial \overline{C}_p}{\partial y} = \frac{\partial \overline{C}_f}{\partial y} e^{-\sqrt{\frac{s}{D_a}}y} - \overline{C}_f \sqrt{\frac{s}{D_a}} e^{-\sqrt{\frac{s}{D_a}}y}$$
(A7)

where L is the Laplace operator. Since

$$\frac{\partial \overline{C}_f(x,s)}{\partial y} = 0 \tag{A8}$$

Eq.(A7) becomes

$$\frac{\partial \overline{C}_p}{\partial y} = -\overline{C}_f \sqrt{\frac{s}{D_a}} e^{-\sqrt{\frac{s}{D_a}y}}$$
(A9)

Taking the Laplace transform of the Eq. (A2)

•

$$s\overline{C}_{f} - C_{f}(x,0) - \frac{2D_{e}}{\delta} \frac{\partial \overline{C}_{p}}{\partial y} \bigg|_{y=0} + u \frac{\partial \overline{C}_{f}}{\partial x} = 0$$
(A10)

since $C_f(x,0) = 0$, and from Eq.(A8)

$$\frac{\partial \overline{C}_p}{\partial y} \bigg|_{y=0} = -\sqrt{\frac{s}{D_a}} \overline{C}_f$$
(A11)

Eq. (A9) becomes

$$u\frac{\partial \overline{C}_{f}}{\partial x} + \left[s + \frac{2D_{e}}{\delta}\sqrt{\frac{s}{D_{a}}}\right]\overline{C}_{f} = 0$$
(12)

If the tracer is assumed to **be** inert, the relationship between the apparent **and** the effective diffusion coefficients is given by,

 $D_a = \frac{D_e}{\phi}$

then

$$\frac{2D_e}{\delta}\sqrt{\frac{s}{D_a}} = \frac{2\sqrt{D_e\phi}}{\delta}\sqrt{s}$$

Let **a** be

$$\alpha = \frac{\sqrt{D_e \phi}}{\delta}$$

Then the equation becomes

$$u \frac{\partial \overline{C}_f}{\partial x} + \left[s + 2\alpha \sqrt{s} \right] \ \overline{C}_f = 0 \tag{A13}$$

rearranging

$$\frac{\partial \overline{C}_f}{\partial x} + \left[\frac{s}{u} + \frac{2\alpha x}{u}\sqrt{s}\right] \ \overline{C}_f = 0$$
(A14)

The solution to this linear homogeneous equation is given by,

$$\overline{C}_f = A \ e^{-\frac{x}{u}s} \ e^{-\frac{2\alpha x}{u}\sqrt{s}}$$
(A15)

Apply the boundary condition

$$\overline{C}_{f} = 1.0 \qquad \text{at} \qquad x = 0 \qquad \text{implies} \qquad A = \frac{1}{s}$$

$$\overline{C}_{f} = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}} \qquad (A16)$$

Thus when it is inverted to real space

$$C_{f} = erfc\left[\frac{\alpha x}{u\sqrt{t-\frac{x}{u}}}\right] \qquad \text{for} \qquad t > \frac{x}{v} \qquad (A17)$$

and

$$C_f = 0$$
 for $t < \frac{x}{\mu}$

since the solution to Eq. (A2) was

$$\overline{C}_{p} = \overline{C}_{f} e^{-\sqrt{\frac{s}{D_{a}}} y}$$
(A18)

Inserting the expression for \overline{C}_f

$$\overline{C}_{p} = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}} e^{-\sqrt{\frac{\Phi}{D_{e}}}y\sqrt{s}}$$
(A19)

and the inversion is given as

$$C_{p} = erfc \left\{ \left[\frac{2\alpha x}{u} + \sqrt{\frac{\Phi}{D_{e}}} \right] \frac{1}{2\sqrt{t - \frac{x}{u}}} \right\} \qquad \text{for} \qquad t > \frac{x}{u} \qquad (A20)$$

and

$$C_p = 0$$
 for $t < \frac{x}{u}$

At the end of the injection period (time t_j), the concentration profiles are given by

$$C_{f} = erfc\left[\frac{\alpha x}{u\sqrt{t_{j} - \frac{x}{u}}}\right] \qquad \text{for} \qquad t_{j} > \frac{x}{u} \qquad (A21)$$

and

$$C_{f} = 0 \qquad \text{for} \qquad t_{j} < \frac{\chi}{u}$$

$$C_{p} = erfc \left\{ \frac{2\alpha x}{u} + \sqrt{\frac{\Phi}{D_{e}}} \right\} - \frac{1}{\sqrt{t_{j} - \frac{x}{u}}} \qquad \text{for} \qquad t_{j} > \frac{1}{u} \qquad (A22)$$

$$C_p = 0$$
 for $t_j < \frac{x}{v}$

At the end of the injection period the profile in the fracture will be as in Fig.(A1).



Fig. (A1) Dispersion of a Sharp Interface Caused by Matrix Diffusion.

To obtain the solutions to this period, we will work with the coordinate of space variable z and time of variable τ . We will relate x to z by $x=L-z = ut_j - z$ and take $\tau = 0$ at the beginning of the backflow period.

In the new coordinates with the assumption of equal injection and backflow rates, the governing equation of the flow in the fracture is,

$$\frac{\partial C_f}{\partial \tau} - \frac{2D_r}{\delta} \frac{\partial C_p}{\partial y} + u \frac{\partial C_f}{\partial z} = 0$$
(A23)

$$D_a \frac{\partial^2 C_p}{\partial y^2} = \frac{\partial C_p}{\partial \tau}$$
(A24)

and

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The boundary conditions are

and

$C_f = 0$	at	<i>z</i> = 0
$C_p = C_f$	at	<i>y</i> = 0
$C_p \rightarrow 0$	as	y → ∞

Here the initial conditions are the injection period solutions evaluated at time t_j . However the attempts to obtain the solutions in real space failed because of the complexity induced by the initial conditions. For this reason the Laplace transformed forms of the initial conditions were be preserved during the solutions of the equations. The initial conditions in Laplace space are given by

$$\overline{C}_f = \frac{1}{s} e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}}$$
(A26)

$$\overline{C}_{p} = \frac{1}{s} e^{-\frac{x}{\mu}s} e^{-\frac{2\alpha x}{\mu}\sqrt{s}} e^{-\sqrt{\frac{\phi}{D_{e}}} y\sqrt{s}}$$
(A27)

for the fracture flow and matrix-diffusion respectively. In Equations (A26) and (A27) the Laplace parameter, s, corresponds to the time of t_j .

Take the Laplace transform of the Eq. (A24) with respect to the variable τ ,

$$\frac{\partial^2 \tilde{C}_f}{\partial y^2} - \frac{p}{D_a} \tilde{C}_p = \frac{-1}{s} \frac{e^{-\frac{x}{u}s}}{D_a} e^{-\frac{2\alpha x}{u}\sqrt{s}} e^{-\sqrt{\frac{s}{D_a}y}}$$
(A28)

Eq. (A28) is a linear nonhomogeneous differential equation and the solution is the linear combination of the homogeneous and the particular solutions. The homogeneous solution is

$$\tilde{C}_{P_{h}} = C_{1} e^{\sqrt{\frac{P}{D_{a}}}y} + C_{2} e^{-\sqrt{\frac{P}{D_{a}}}y}$$
(A29)

The particular solution may be found by the method of undetermined coefficients **as** follows. Let the particular solution be

$$\tilde{C}_{p_p} = A \ e^{-\sqrt{\frac{s}{D_a}} \ y} \tag{A30}$$

From Eq. (A30)

$$\frac{d^2 \tilde{C}_{p_p}}{dy^2} = A \frac{s}{D_a} e^{-\sqrt{\frac{s}{D_a}y}}$$
(A31)

Substituting Equations (A30) and (A31) into Eq. (A28) we obtain

$$A \frac{s}{D_a} e^{-\sqrt{\frac{s}{D_a}}y} - A \frac{p}{D_a} e^{-\sqrt{\frac{s}{D_a}}y} = -\frac{e^{-\frac{x}{u}s}}{sD_a} e^{-\frac{2\alpha x}{u}\sqrt{s}} e^{-\sqrt{\frac{s}{D_a}}y}$$
(A32)

Solving for A

$$A = \frac{1}{s(p-s)} e^{-\frac{x}{\mu}s} e^{-\frac{2\alpha x}{\mu}\sqrt{s}}$$
(A33)

Inserting the value of A in Eq. (A30) the particular solution can be written as,

$$\tilde{C}_{P_p} = \frac{1}{s(p-s)} e^{-\frac{s}{m}s} e^{-\frac{2\alpha x}{m}\sqrt{s}} e^{-\sqrt{\frac{s}{D_a}y}}$$
(A34)

Then the complete solution becomes,

$$\tilde{C}_{p} = C_{1} e^{\sqrt{\frac{p}{D_{a}}}y} + C_{2} e^{-\sqrt{\frac{p}{D_{a}}}y} + \frac{1}{s(p-s)} e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}} e^{-\sqrt{\frac{s}{D_{a}}}y}$$
(A35)

Apply the boundary conditions

 $C_p \to 0$ as $y \to \infty$ implies $C_1 = 0$

and

$$C_p = C_f$$
 at $y = 0$

implies

$$C_2 = \tilde{C}_f - \frac{1}{s(p-s)} e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}}$$

Let

$$\beta = e^{-\frac{\pi}{u}s} e^{-\frac{\pi}{u}s}$$

Then

$$\tilde{C}_{p} = \tilde{C}_{f} e^{-\sqrt{\frac{s}{D_{a}}}y} - \frac{\beta}{s(p-s)} \left[e^{-\sqrt{\frac{p}{D_{a}}}y} - e^{-\sqrt{\frac{s}{D_{a}}}y} \right]$$
(A36)

and

$$\frac{\partial \tilde{C}_{p}}{\partial y}\Big|_{y=0} = -\sqrt{\frac{p}{D_{a}}}\tilde{C}_{f} - \frac{s(p\beta_{a})}{s(p\beta_{a})}\left[\sqrt{\frac{s}{D_{a}}} - \sqrt{\frac{p}{D_{a}}}\right]$$
(A37)

Now, if we take the Laplace transform of Eq. (A23) and insert Eq.(A37) we obtain,

$$p\tilde{C}_{f} - \overline{C}_{f}(z,s) - \frac{2D_{e}}{\delta} \left[-\tilde{C}_{f}\sqrt{\frac{p}{D_{a}}} - \frac{\sqrt{s} - \sqrt{p}}{\sqrt{D_{a}}(p-s)} \frac{\beta}{s} \right] + u\frac{\partial\tilde{C}_{f}}{\partial z} = 0$$
(A38)

where $\overline{C}_{f}(z,s)$ is given by Eq. (A26). Rearranging Eq. (A38)

$$\frac{\partial \tilde{C}_f}{\partial z} + \left[\frac{p}{u} + \frac{2\alpha}{u}\sqrt{s}\right]\tilde{C}_f = \left[\frac{1}{su} + \frac{2\alpha}{su(\sqrt{s} + \sqrt{p})}\right] e^{-\frac{x}{u}s} e^{-\frac{2\alpha x}{u}\sqrt{s}}$$
(A39)

Since x = L - z, Eq. (A39) can be rewritten as

$$\frac{\partial \tilde{C}_f}{\partial z} + \left[\frac{p}{u} + \frac{2\alpha}{u}\sqrt{s}\right]\tilde{C}_f = \omega \quad e^{\frac{z}{u}s} \quad e^{\frac{2\alpha z}{u}\sqrt{s}} \tag{A40}$$

| .

where

$$\omega = \frac{1}{s} \left[\frac{1}{u} + \frac{2\alpha}{u(\sqrt{s} + \sqrt{p})} \right] e^{-\frac{L}{u}s} e^{-\frac{2\alpha L}{u}\sqrt{s}}$$

Equation (A40) is a linear nonhomogeneous first order equation and its solution can be found **by** separation of variables

$$\tilde{C}_f \quad e^{\frac{z}{u}p} \quad e^{\frac{2\alpha z}{u}\sqrt{p}} = \omega \quad \int \quad e^{\frac{z}{u}s} \quad e^{\frac{2\alpha z}{u}\sqrt{s}} \quad e^{\frac{z}{u}p} \quad e^{\frac{2\alpha z}{u}\sqrt{p}} \quad dz + A$$
(A41)

where A is a constant, and the solution can be rewritten as

$$\tilde{C}_{f} = \frac{\omega}{\frac{s+p}{u} + \frac{2\alpha}{u}(\sqrt{s} + \sqrt{p})} \quad e^{\frac{s}{u}s} e^{\frac{2\alpha s}{u}\sqrt{s}} + A \quad e^{-\frac{s}{u}p} e^{-\frac{2\alpha s}{u}\sqrt{p}}$$
(A42)

Applying the boundary condition, $C_f = 0$ at z = 0

$$A = -\frac{\omega}{\frac{s+p}{u} + \frac{2\alpha}{u}(\sqrt{s} + \sqrt{p})}$$

inserting the value of A into Eq.(A42), the backflow period solution is obtained

$$C_{f} = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{s} + \sqrt{p}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{s} + \sqrt{p})} \right]$$
$$\left\{ e^{-\frac{L}{u}s} e^{-\frac{2\alpha L}{u}\sqrt{s}} e^{\frac{z}{u}s} e^{\frac{2\alpha u}{u}\sqrt{s}} - e^{-\frac{L}{u}s} e^{-\frac{2\alpha L}{u}\sqrt{s}} e^{-\frac{z}{u}p} e^{-\frac{2\alpha z}{u}\sqrt{p}} \right\}$$
(A43)

Since the concentration measurements are made at the well which corresponds to z = L

$$\tilde{C}_{f} = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{s} + \sqrt{p}} \right] \left[\frac{1}{s + p + 2\alpha\sqrt{s} + \sqrt{p}} \right] \\ \left\{ \begin{array}{c} -\frac{\pi}{s}(p + s) & -\frac{2\alpha L}{s}(\sqrt{p} + \sqrt{s}) \\ 1 - e^{-\frac{\pi}{u}} & e^{-\frac{\pi}{u}} \end{array} \right]$$
(A44)

In Eq. (A44) the terms, $e^{-\frac{L}{u}p}$ and $e^{-\frac{L}{u}s}$ will cause Heaviside step function effects, $H(t_p - \frac{L}{u})$, and $H(t_j - \frac{L}{u})$ respectively,

where

$$H(x) = \begin{cases} 1 & x > 0 \\ 0 & x < 0 \end{cases}$$

These effects **are** investigated **as** follows. The solution is in the Laplace space and it is two transformations away from the real space. In this space the above solution cat be expressed in a general form **as** following.

$$\tilde{C}_f = F(s,p,\alpha) - G(s,p,\alpha) \quad e^{-\frac{L}{u}s} \quad e^{-\frac{L}{u}p}$$

$$\overline{C}_f = F(s,\tau,\alpha) - G(s,\tau - \frac{L}{u},\alpha) e^{-\frac{L}{u}s} \qquad \text{for} \quad \tau > \frac{L}{u}$$

and

$$\overline{C}_f = F(s,\tau,\alpha) - 0 \qquad \text{for} \quad \tau < \frac{L}{u}$$

In the real space the form of the solution has to be

$$\overline{C}_{t} = F(t_{j},\tau,\alpha) - G(t_{j} - \frac{L}{u},\tau - \frac{L}{u},\alpha) \qquad \text{for} \quad \tau > \frac{L}{u} \qquad \text{and} \qquad t_{j} > \frac{L}{u}$$

and

$$\overline{C}_{f} = F(t_{j}, \tau, \alpha) \qquad \text{for} \quad t_{j} < \frac{L}{u}$$

Since $t_j = \frac{L}{u}$ the second term will always be zero, therefore the solution becomes

$$C_f = F(t_j, \tau, \alpha)$$

or in the Laplace space,

$$\tilde{C}_f = F(s,p,\alpha)$$

and in complete form

$$\tilde{C}_{f} = \frac{1}{s} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
(A45)

APPENDIX B : DERIVATIVES OF THE SOLUTIONS OF THE MATRIX DIFFUSION MODEL WITH RESPECT TO THE NONLINEAR PARAMETERS

The solution found for the continuous injection case is :

$$\tilde{C}_{f} = \frac{1}{s} \left[1 \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
(B1)

The equation (B1) has only one parameter a, and the derivative with respect to a is given by

$$\frac{\partial \tilde{C}_f}{\partial \alpha} = \frac{1}{s} \left[\frac{2}{\sqrt{s} + \sqrt{p}} \frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
$$- \frac{1}{s} \left[\left[\frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \frac{2(\frac{8}{s + p} + 2\alpha(\sqrt{p} + \sqrt{s})^2}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})^2} \right]$$
(B2)

The solution for the spike injection case is :

$$\tilde{C}_{f} = \frac{M}{Q} \left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \left[\frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
(B3)

Equation (B3) has only one parameter a, and derivative with respect to α is given by

$$\frac{\partial \tilde{C}_f}{\partial \alpha} = \frac{M}{Q} \left[\frac{2}{\sqrt{s} + \sqrt{p}} \frac{1}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})} \right]$$
$$- \frac{M}{Q} \left[\left[1 + \frac{2\alpha}{\sqrt{p} + \sqrt{s}} \right] \frac{2(\sqrt{s} + \sqrt{p})}{s + p + 2\alpha(\sqrt{p} + \sqrt{s})^2} \right]$$
(B4)

APPENDIX C : DERIVATIVES OF THE SOLUTIONS OF THE CONVECTION-DISPERSION MODEL WITH RESPECT TO THE NONLINEAR PARAMETERS

The solution found for the continuous injection case is :

$$C = \frac{1}{2} + \frac{1}{2} \operatorname{erf}\left[\frac{\alpha(t_j - t_p)}{2\sqrt{t_j + t_p}}\right]$$
(C1)

The partial derivative of the erf (**a** c) with respect to **a** is given by

$$\frac{\partial erf(\alpha c)}{\partial \alpha} = \frac{2c}{\sqrt{\pi}} e^{-(\alpha c)^2}$$
(C2)

Then the derivative of equation (C1) with respect to a can be found as

$$\frac{\partial C}{\partial \alpha} = \frac{1}{2\sqrt{\pi}} \frac{t_j - t_p}{\sqrt{t_j + t_p}} e^{\frac{-\alpha^2 (t_j - t_p)^2}{4(t_j + t_p)}}$$
(C3)

The solution for the spike injection case was found as

$$C = \frac{\alpha t_i}{2\sqrt{\pi(t_j + t_p)}} e^{\frac{-\alpha^2 (t_j - t_p)^2}{4(t_j + t_p)}}$$
(C4)

Here the nonlinear parameter is also \mathbf{Q} and the derivative with respect to it is given by

$$C = \frac{t_i}{2\sqrt{\pi(t_j + t_p)}} \left[1 - \frac{\alpha^2(t_j - t_p)}{2(t_j + t_p)} \right] e^{\frac{-\alpha^2(t_j - t_p)^2}{4(t_j + t_p)}}$$
(C5)

APPENDIX D: Listing of Programs to Perform Nonlinear Curve Fitting of the Matrix Diffusion Model With a Sample Input and a Corresponding Output

LISTING OF THE CURFITTING PROGRAM FOR THE MATRIX DIFFUSION MODEL PROGRAM REGINS ****** MAIN PROGRAM (PROGRAM CURFIT) *************** DEFINITION OF THE PRARAMETERS NI : NUMBER OF TERMS THAT WILL BE USED IN STEHFEST ALGORITHM V : THE VECTOR GENERATED AND USED BY THE STEHFEST ALGORITHM FOR INVERSION TJ : INJECTION PERIOD T(I): TIME STEPS AT WHICH THE CONCENTRATIONS ARE MEASURED DURING BACKFLOW PERIOD C : CONCENTRATIONS CALCULATED BY USING THE OPTIMUM VALUE OF THE NONLINEAR PARAMETER INPUT DATA ARE STORED IN FILE "LAPLS" OUTPUT DATA ARE STORED IN FILE "OUTPUT" VALUES OF THE THE NONLINEAR PARAMETER AND OTHER STATISTICAL DATA ARE DISPLAYED ON THE SCREEN TOO IMPLICIT REAL*8(A-H,O-Z) DIMENSION Y(100),T(100),ALF(2),BETA(2),W(100),A(100,6) *,C(100),INC(3,3),V(50) EXTERNAL ADA COMMON NI.V,TJ DATA NI/10/ OPEN(UNIT=8,FILE='LAPLS') OPEN(UNIT=9,FILE='OUTFILE') SET PARAMETERS FOR VARPRO NMAX=1ØØ LPP2=3IPRINT=1 READ(8,*)TJ WRITE(6,121) TJ FORMAT(1HØ,1ØX,'INJECTION TIME'//(F7.2)) 121

.

00000

-65-

```
C
C
C
          CALL COEFF(NI,V)
С
       READ DATA
C
C
C
       NL IS THE NUMBER OF NONLINEAR PARAMETERS
С
С
         READ(8.10)NL
FORMAT(I3)
WRITE(9,12)NL
   10
  12
           FORMAT(1HØ,1ØX, 'NUMBER OF NONLINEAR PARAMETERS'//(13))
000000
       ESTIMATES OF THE NONLINEAR PARAMETERS
         READ(8,15) ALF(1)
         FORMAT(F7.3)
WRITE(9,20)ALF(1)
FORMAT(1HØ,10X,'INITIAL ESTIMATES OF NONLIN. PARAM. '//(F8.4))
   15
  20
C
C
C
C
C
C
       L IS THE NUMBER OF LINEAR PARAMETERS
Č
         READ(8,22) L
           FORMAT(13)
    22
          WRITE(9,25)L
           FORMAT(1HØ,1ØX, 'NUMBER OF LINEAR PARAMETERS'//(13))
    25
C
C
С
С
С
       N IS THE NUMBER OF OBSERVATIONS
С
         READ(8,3Ø)N
FORMAT(13)
    30
          WRITE(9.35)N
FORMAT(1HØ,1ØX,'NUMBER OF OBSERVATIONS'//(I3))
    35
          DO 1 I=1,N
    1
          \forall (I) = 1.\emptyset
С
C
C
C
C
C
C
       IV IS THE NUMBER OF INDEPENDENT VERIABLES
         READ(8,4Ø)IV
FORMAT(13)
   40
          WRITE(9,45)IV
   45
           FORMAT(1HØ,1ØX,'NUMBER OF INDEPENDENT VARIABLES '//(I3))
C
C
C
           READ THE TIME AND MEASURED CONCENTRATION DATA T(I) AND Y(I)
C
C
          READ(8,55)(T(I),Y(I),I=1,N)
           FORMAT(2(F7.3,2X))
    55
          WRITE(9,6Ø)(T(I),Y(I),I=1,N)
```

.
```
60 FORMAT(1HØ,1ØX,'INDEPENDENT VARIABLES'//,2(F7.3,3X))
C
CALL VARPRO(L,NL,N,NMAX,LPP2,IV,T,Y,W,ADA,A,IPRINT,ALF,BETA,
*IERR)
C
WRITE(9,21)
21 FORMAT(1HØ,'CALCULATED CONCENTRATIONS',/,T5,'TIME',T15,
*'CONCENTRATION',//)
D0 27 I-1.N
TD=T(I)
CD=Ø,Ø
AS=DLOG(2.DØ)/TJ
D0 37 J=1,NI
ARG=AS*J
CD=CD+V(J)*CDS(ARG,J,TD,ALF)
37 CONTINUE
C(I)=CD*AS
WRITE(9,23)T(I),C(I)
23 FORMAT(2(F7.3,2X))
27 CONTINUE
STOP
END
```

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****** PROGRAM BEGINS SUBROUTINE ADA ********** THE MATRIX A IS ARRANGED ACCORDING TO NO LINEAR PARAMETER ASSUMPTION IF TAKE L=1 THEN HAVE TO REARRANGE IT. ***** SUBROUTINE ADA(L,NL,N,NMAX,LPP2,IV,A,INC,T,ALF,ISEL) IMPLICIT REAL*8(A-H,O-Z) DIMENSION ALF(2), A(100,6), T(100), INC(3,3), B(100,6), Y(100) DIMENSION V(50) COMMON NI,V,TJ 0000 SKIP UNLESS ISEL=1 С IF(ISEL.EQ.2)GO TO 90 IF(ISEL.EQ.3)GO TO 165 C C C SET INCIDENCE MATRIX, INC(K,J)=1 IF ALF(K) APPEARS IN PHI(J) INC(1,1)=1.0 WRITE(9,70) INC(1,1) FORMAT(1H0,'INCIDENCE MATRIX INC(1,1)= '//(I3)) 70 00000 CALCULATE MATRIX A STORE VALUES OF THE FUNCTION IN FIRST COLUMN OF A DO 81 I=1,N TD=T(I) 90 ĊD=Ø.Ø $AS=DLOG(2.D\emptyset)/TJ$ DO 15 J=1.NI ARG=AS*J CD=CD+V(J)*CDS(ARG,J,TD,ALF) CONTINUE CD=CD*AS 15 A(∎,1)=CD CONTINUE 01 WRITE(9,103)(A(I.1),I=1,N) FORMAT(1H0,'COLUMN #1 OF A(I,J) MATRIX'//(F9.5)) 0000000 103 SKIP EVALUATION OF DERIVATIVES IF ISEL=2 STORE VALUES OF THE DERIVATIVES IN SECOND COLUMN OF A IF(ISEL.EQ.2)GO TO 200 165 DO 170 I=1.N TD = T(I) $DCD = \emptyset . \emptyset$ AS=DLOG(2.DØ)/TJ

DO 175 J=1,NI

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```
ARG=AS*J
DCD=DCD+V(J)*DCDS(ARG,J,TD,ALF)
   175
         CONTINUE
           DCD=DCD*AS
         A(I,2)=DCD
CONTINUE
   170
C
C
           WRITE(9,180)(A(I,2),I=1,N)
FORMAT(1H0,'COLUMN #2 OF A(I,J) MATRIX'//(F9.5,2X))
  18Ø
   200
         CONTINUE
           RETURN
           END
000000
           EVALUATION OF THE SOLUTION IN ( Z,S,PRODUTION TIME)-SPACE
           FUNCTION CDS(S, J, TD, ALF)
IMPLICIT REAL*8(A-H, 0-Z)
DIMENSION V(50), ALF(2)
           COMMON NI,V,TJ
           CDS = \emptyset . \emptyset
           AP = DLOG(2.DØ)/TD
           DO 20 I=1,NI
ARG=AP*I
           CDS=CDS+V(I)*CDSP(ARG,I,S,ALF)
   20
           CONTINUE
           CDS=CDS*AP
           RETURN
           END
CCCCC
           EVALUATION OF THE SOLUTION IN (Z,S,P)-SPACE
           FUNCTION CDSP(P, I, S, ALF)
IMPLICIT REAL*8(A-H.O-Z)
DIMENSION ALF(2),V(50)
           COMMON NI,V,TJ
           F1=1.\emptyset/S*(1.\emptyset+2.\emptyset*ALF(1)/(DSQRT(S)+DSQRT(P)))
           F2=1.Ø/(S+P+2.Ø*ALF(1)*(DSQRT(S)+DSQRT(P)))
           CDSP=F1*F2
           RETURN
           END
00000000
           EVALUATION OF THE DERIVATIVE OF THE SOLUTION WITH RESPECT TO
           THE NONLINEAR PARAMETER IN (Z,S, PRODUCTION TIME) - SPACE .
           FUNCTION DCDS(S, J, TD, ALF)
IMPLICIT REAL*8(A-H, O-Z)
           DIMENSION V(5Ø)
           COMMON NI,V,TJ
           DCDS=Ø.Ø
           AS=DLOG(2.ØDØ)/TD
           DO 20 I=1,NI
ARG=AS*I
           DCDS=DCDS+V(I)*DCDSP(ARG,I,S,ALF)
  20
           CONTINUE
           DCDS=DCDS*AS
           RETURN
           END
С
```

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EVALUATION OF THE DERIVATIVE OF THE SOLUTION WITH **RESPECT** TO THE NONINEAR SOLUTION IN (**Z,S,P)-SPACE** FUNCTION DCDSP(P,I,S,ALF) IMPLICIT REAL*8(A-H,O-Z) DIMENSION ALF(2),V(5Ø) COMMON NI,V,TJ FA=DSQRT(S)+DSQRT(P) FB=(S+P+2.Ø*ALF(1)*FA) FC=(1.Ø+2.Ø*ALF(1)/FA) DCDSP=2.Ø/S*(1.Ø/(FA*FB)-FC*FA/FB**2) RETURN END

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SUBROUTINE VARPRO (L, NL, N, NMAX, LPP2, IV. T, Y, W, ADA, A, X IPRINT, ALF, BETA, IERR)

GIVEN A SET OF N OBSERVATIONS, CONSISTING OF VALUES Y(1), Y(2), ..., Y(N) OF A DEPENDENT VARIABLE Y, WHERE Y(I) CORRESPONDS TO THE IV INDEPENDENT VARIABLE(S) T(I,1), T(I,2), T(I,IV). VARPRO ATTEMPTS TO COMPUTE A WEIGHTED LEAST SQUARES FIT TO A FUNCTION ETA (THE 'MODEL') WHICH IS A LINEAR COMBINATION

ETA(ALF, BETA: T) = SUM BETA * PHI (ALF; T) + PHI (ALF; T) J-1 J J L+I

OF NONLINEAR FUNCTIONS PHI(J) (E.G., A SUM OF EXPONENTIALS AND/ OR GAUSSIANS). THAT IS, DETERMINE THE LINEAR PARAMETERS BETA(J) AND THE VECTOR OF NONLINEAR PARAMETERS ALF BY MINIMIZ-ING

THE (L+1)-ST TERM IS OPTIONAL, AND IS USED WHEN IT IS DESIRED TO FIX ONE OR MORE OF THE BETA'S (RATHER THAN LET THEM BE DETERMINED). VARPRO REQUIRES FIRST DERIVATIVES OF THE PHI'S.

NOTES :

A) THE ABOVE PROBLEM **IS** ALSO REFERRED TO AS 'MULTIPLE NONLINEAR REGRESSION'. FOR USE IN STATISTICAL ESTIMATION, VARPRO RETURNS THE RESIDUALS, THE COVARIANCE MATRIX OF THE LINEAR AND NONLINEAR PARAMETERS, AND THE ESTIMATED VARIANCE OF THE OBSERVATIONS.

B) AN ETA OF THE ABOVE FORM ${\bf IS}$ CALLED 'SEPARABLE'. THE CASE OF A NONSEPARABLE ETA CAN BE HANDLED BY SETTING L = O AND USING PHI(L+1).

C) VARPRO MAY ALSO BE USED TO SOLVE LINEAR LEAST SQUARES PROBLEMS (IN THAT CASE NO ITERATIONS ARE PERFORMED). SET NL = O.

D) THE MAIN ADVANTAGE OF VARPRO OVER OTHER LEAST SQUARES PROGRAMS IS THAT NO INITIAL GUESSES ARE NEEDED FOR THE LINEAR PARAMETERS. NOT ONLY DOES THIS MAKE IT EASIER TO USE, BUT IT OFTEN LEADS TO FASTER CONVERGENCE.

DESCRIPTION OF PARAMETERS

L	NUMBER OF LINEAR PARAMETERS BETA (MUST BE .GE. OJ.
NL	NUMBER OF NONLINEAR PARAMETERS ALF (MUST BE .GE. Ø).
N	NUMBER OF OBSERVATIONS. N MUST BE GREATER THAN L + NL
	(I.E., THE NUMBER OF OBSERVATIONS MUST EXCEED THE
	NUMBER OF PARAMETERS).
IV	NUMBER OF INDEPENDENT VARIABLES T.
Т	REAL N BY IV MATRIX OF INDEPENDENT VARIABLES. $T(I, J)$
	CONTAINS THE VALUE OF THE I- TH OBSERVATION OF THE J-TH
	INDEPENDENT VARIABLE.
Y	14-VECTOR OF OBSERVATIONS. ONE FOR EACH ROW OF T.
W	N-VECTOR OF NONNEGATIVE WEIGHTS. SHOULD BE SET TO 1'S
	IF WEIGHTS ARE NOT DESIRED. IF VARIANCES OF THE
	INDIVIDUAL OBSERVATIONS ARE KNOWN, W(1) SHOULD BE SET
	TO 1./VARIANCE(I).

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	INC	NL X (L+1) INTEGER INCIDENCE MATRIX. INC(K, J) = 1 IF NON-LINEAR PARAMETER ALF(K) APPEARS IN THE J-TH FUNCTION PHI(J). (THE PROGRAM SETS ALL OTHER INC(K, J) TO ZERO.) IF PHI(L+1) IS INCLUDED IN THE MODEL. THE APPROPRIATE ELEMENTS OF THE (L+1)-ST COLUMN SHOULD BE SET TO 1'S. INC IS NOT NEEDED WHEN L = O OR NL = O. CAUTION: THE DECLARED ROW DIMENSION OF INC (IN ADA)
	NMAX	MUSI CURRENILY BE SET TO 12. SEE 'RESTRICTIONS' BELOW. THE DECLARED ROW DIMENSION OF THE MATRICES A AND T. IT MUST BE AT LEAST MAY (N 24NL 42)
	LPP2	L+P+2, WHERE P IS THE NUMBER OF ONES IN THE MATAIX INC. THE DECLARED COLUMN DIMENSION OF A MUST BE AT LAST LPPZ. (IF L = O, SET LPPZ = NL+2. IF NL = O, SVT LPP2
	A	REAL MATRIX OF SIZE MAX(N, 2*NL+3) BY L+P+2. ON INPUT IT CONTAINS THE PHI(J)'S AND THEIR DERIVATIVES (SEE BELOW). ON OUTPUT, THE FIRST L+NL ROWS AND COLUMNS OF A WILL CONTAIN AN APPROXIMATION TO THE (WEIGHTED) COVARIANCE MATRIX AT THE SOLUTION (THE FIRST L ROWS CORRESPOND TO THE LINEAR PARAMETERS, THE LAST NU TO THE NONLINEAR ONES), COLUMN L+NL+1 WILL CONTAIN THE WEIGHTED RESIDUALS (Y - ETA), A(1, L+NL+2) WILL CONTAIN THE (EUCLIDEAN) NORM OF THE WEIGHTED RESIDUAL, AND A(2, L+NL+2) WILL CONTAIN AN ESTIMATE OF THE (WIIGHTED) VARIANCE OF THE OBSERVATIONS, NORM(RESIDUAL)**20
	IPRINT	INPUT INTEGER CONTROLLING PRINTED OUTPUT. IF IFRINT IS POSITIVE, THE NONLINEAR PARAMETERS, THE NORM OF THE RESIDUAL, AND THE MARQUARDT PARAMETER WILL BE OUTPUT EVERY IPRINT-TH ITERATION (AND INITIALLY, AND AT THE FINAL ITERATION). THE LINEAR PARAMETERS WILL BE PRINTED AT THE FINAL ITERATION. ANY ERROR MESSAGES WILL ALSO BE PRINTED. (IPRINT = 1 IS RECOMMENDED AT FIRST.) IF IPRINT = 0. ONLY THE FINAL QUANTITIES WILL BE PRINTED, AS WELL AS ANY ERROR MESSAGES. IF [PRINT = -1. NO PRINTING WILL BE DONE. THE USER IS THEN RESPONSIBLE FOR CHECKING THE PARAMETER IERR FOR'ERRORS.
	ALF	NL-VECTOR OF ESTIMATES OF NONLINEAR PARAMETERS (INPUT). ON OUTPUT IT WILL CONTAIN OPTIMAL VALUES OF
	BETA ∎ERR	L-VECTOR OF LINEAR PARAMETERS. L-VECTOR OF LINEAR PARAMETERS (OUTPUT ONLY). INTEGER ERROR FLAG (OUTPUT): .GT. O - SUCCESSFUL CONVERGENCE, IERR IS THE NUMBER OF ITERATIONS TAKEN.
		 TERMINATED FOR TOO MANY ITERATIONS. TERMINATED FOR ILL-CONDITIONING (MARQUARDT PARAMETER TOO LARGE.) ALSO SEE IERR = -8 BELOW. INPUT ERROR IN PARAMETER N, L, NL. LPP2, OR NMAX. INC MATRIX IMPROPERLY SPECIFIED, OR P DISAGREES WITH LPPZ.
		 -6 A WEIGHT WAS NEGATIVE. -7 'CONSTANT' COLUMN WAS COMPUTED MORE THAN ON €E. -8 CATASTROPHIC FAILURE - A COLUMN OF THE A MATRIX HAS BECOME ZERO. SEE 'CONVERGENCE FAILURES' BELOW.
		(IF IERR .LE4. THE LINEAR PARAMETERS, COVARIANCE MATRIX, ETC. ARE NOT RETURNED.)
SUE	BROUTINES	S REQUIRED
		CURROUTINES DRA OPEACI OPEACZ RACCUR POSTOR COV

NINE SUBROUTINES, DPA. ORFAC1, ORFACZ. BACSUB. POSTPR, COV, XNORM, INIT, AND VARERR ARE PROVIDED. IN ADDITION, THE USER MUST PROVIDE A SUBROUTINE (CORRESPONDING TO THE ARGUMENT ADA) WHICH, GIVEN ALF, WILL EVALUATE THE FUNCTIONS PHI(J) AND THEIR PARTIAL DERIVATIVES D PHI(J)/D ALF(K), AT THE SAMPLE POINTS T(1). THIS ROUTINE MUST BE DECLARED 'EXTERNAL' IN THE CALL IN THE CALLING ITS CALLING SEQUENCE IS PROGRAM.

SUBROUTINE ADA (L+1, NL, N, NMAX, LPP2, IV, A, INC, T, ALF, ISEL)

THE USER SHOULD MODIFY THE EXAMPLE SUBROUTINE 'ADA' (GIVEN ELSEWHERE) FOR HIS OWN FUNCTIONS.

THE VECTOR SAMPLED FUNCTIONS PHI(J) SHOULD BE STORED IN THE FIRST N ROWS AND FIRST L+I COLUMNS OF THE MATRIX A, I.E., A(I, J) SHOULD CONTAIN PHI(J, ALF; T(I,1), T(I,2), ..., T(I,IV)), I = 1, ..., N: J = I, ..., L (OR L+1). THE (L+1)-ST COLUMN OF A CONTAINS PHI(L+1) IF PHI(L+1) IS IN THE MODEL, OTHERWISE IT IS RESERVED FOR WORKSPACE. THE 'CONSTANT' FUNC-TIONS (THESE ARE FUNCTIONS PHI(J) WHICH DO NOT DEPEND UPON ANY NONLINEAR PARAMETERS ALF, E.G., T(I)**J) (IF ANY) MUST APPEAR FIRST, STARTING IN COLUMN 1. THE COLUMN N-VECTORS OF NONZERO PARTIAL DERIVATIVES D PHI(J) / D ALF(K) SHOULD BE STORED SEQUENTIALLY IN THE MATRIX A IN COLUMNS L+2 THROUGH L+P+1. THE ORDER IS THE ORDER IS

D PHI(1)	D PHI(2)	D PHI(L) D PHI(L+1)	D PHI(1)
D ALF(1)	D ALF(1)	DALF(1) DALF(1)	D ALF(2)
D PHI(2)	D PHI(L+1)	D PHI(1)	D PHI(L+1)
D ALF(2)	D ALF(2)	D ALF(NL)	D ALF(NL)

OMITTING COLUMNS OF DERIVATIVES WHICH ARE ZERO, AND OMITTING PHI(L+1) COLUMNS IF PHI(L+1) IS NOT IN THE MODEL. NOTE THE LINEAR PARAMETERS BETA ARE NOT USED IN THE MATRIX A. COLUMN L+P+2 IS RESERVED FOR WORKSPACE. NOTE THAT

THE CODING OF ADA SHOULD BE ARRANGED SO THAT:

(WHICH OCCURS THE FIRST TIME ADA IS CALLED) MEANS: |SEL = 1|FILL IN THE INCIDENCE MATRIX INC STORE ANY CONSTANT PHI'S IN A. COMPUTE NONCONSTANT PHI'S AND PARTIAL DERIVA-À.

- Β.
- С TIVES
- 2 3 MEANS COMPUTE ONLY THE NONCONSTANT FUNCTIONS PHI = MEANS COMPUTE ONLY THE DERIVATIVES =

(WHEN THE PROBLEM IS LINEAR (NL = Ø) ONLV ISEL = 1 IS USED, AND DERIVATIVES ARE NOT NEEDED.)

RESTRICTIONS

THE SUBROUTINES DPA, INIT (AND ADA) CONTAIN THE LOCALLY DIMENSIONED MATRIX INC, WHOSE DIMENSIONS ARE CURRENTLY SET FOR MAXIMA OF L+I = 8, NL = 12. THEY MUST BE CHANGED FOR LARGER PROBLEMS. DATA PLACED IN ARRAY A IS OVERWRITTEN ('DESTROYED') DATA PLACED IN ARRAYS T, Y AND INC IS LEFT INTACT. THE PROGRA RUNS IN WATFIV, EXCEPT WHEN L = O OR NL = O. THE PROGRAM

IT IS ASSUMED THAT THE MATRIX PHI(J, RLF: T(I)) HAS FULL COLUMN RANK. THIS MEANS THAT THE FIRST L COLUMNS OF THE MATRIX A MUST BE LINEARLY INDEPENDENT.

OPTIONAL NOTE: AS WILL BE NOTED FROM THE SAMPLE SUBPROGRAM ADA, THE DERIVATIVES D PHI(J)/D ALF(K) (ISEL = 3) MUST BE

COMPUTED INDEPENDENTLY OF THE FUNCTIONS PHI(J) (ISEL = 2), SINCE THE FUNCTION VALUES ARE OVERWRITTEN AFTER ADA IS CALLED WITH ISEL = 2. THIS IS DONE TO MINIMIZE STORAGE. AT THE POS-SIBLE EXPENSE OF SOME RECOMPUTATION (SINCE THE FUNCTIONS AND DERIVATIVES FREQUENTLY HAVE SOME COMMON SUBEXPRESSIONS). TO REDUCE THE AMOUNT OF COMPUTATION AT THE EXPENSE OF SOME STORAGE, CREATE A MATRIX B OF DIMENSION NMAX BY L+I IN ADA, AND AFTER THE COMPUTATION OF THE PHI'S (ISEL = 2), COPY THE VALUES INTO B. THESE VALUES CAN THEN BE USED TO CALCULATE THE DERIV-ATIVES (ISEL = 3), (THIS MAKES USE OF THE FACT THAT WHEN A CALL TO ADA WITH ISEL = 3 FOLLOWS A CALL WITH ISEL = 2, THE ALFS ARE THE SAME.)

TO CONVERT TO OTHER MACHINES, CHANGE THE OUTPUT UNIT IN THE DATA STATEMENTS IN VARPRO, DPA, POSTPR, AND VARERR. THE PROGRAM HAS BEEN CHECKED FOR PORTABILITY BY THE BELL LABS PFORT VERIFIER. FOR MACHINES WITHOUT DOUBLE PRECISION HARDWARE, IT MAY BE DESIRABLE TO CONVERT TO SINGLE PRECISION. THIS CAN BE DONE BY CHANGING (A) THE DECLARATIONS 'DOUBLE PRECISION' TO 'REAL', (B) THE PATTERN '.D' TO '.E' IN THE 'DATA' STATEMENT IN VARPRO, (C) DSIGN, DSQRT AND DABS TO SIGN, SQRT AND ABS, RESPECTIVELY, AND (D) DEXP TO EXP IN THE SAMPLE PROGRAMS ONLY.

NOTE ON INTERPRETATION OF COVARIANCE MATRIX

FOR USE IN STATISTICAL ESTIMATION (MULTIPLE NONLINEAR REGRESSION) VARPRO RETURNS THE COVARIANCE MATRIX OF THE LINEAR AND NONLINEAR PARAMETERS. THIS MATRIX WILL BE USEFUL ON Y IF THE USUAL STATISTICAL ASSUMPTIONS HOLD: AFTER WEIGHTING, THE ERRORS IN THE OBSERVATIONS ARE INDEPENDENT AND NORMALLY DISTRI-BUTED, WITH MEAN ZERO AND THE SAME VARIANCE. IF THE ERRORS DO NOT HAVE MEAN ZERO (OR ARE UNKNOWN), THE PROGRAM WILL ISSUE A WARNING MESSAGE (UNLESS IPRINT .LT. Ø) AND THE COVARIANCE MATRIX WILL NOT BE VALID. IN THAT CASE, THE MODEL SHOULD BE ALTERED TO INCLUDE A CONSTANT TERM (SET PHI(1) = 1.).

NOTE ALSO THAT, IN ORDER FOR THE USUAL ASSUMPTIONS TO HOLD, THE OBSERVATIONS MUST ALL BE OF APPROXIMATELY THE SAME MAGNITUDE (IN THE ABSENCE OF INFORMATION ABOUT THE ERROR OF EACH OBSERVATION), OTHERWISE THE VARIANCES WILL NOT BE THE SAME. IF THE OBSERVATIONS ARE NOT THE SAME SIZE, THIS CAN BE CURED BY WEIGHTING.

IF THE USUAL ASSUMPTIONS HOLD, THE SQUARE ROOTS OF THE DIAGONALS OF THE COVARIANCE MATRIX A GIVE THE STANDARD ERROR S(I) OF EACH PARAMETER. DIVIDING A(I,J) BY S(I)*S(J) YIELDS THE CORRELATION MATRIX OF THE PARAMETERS. PRINCIPAL AXES AND CONFIDENCE ELLIPSOIDS CAN BE OBTAINED BY PERFORMING AN EL GEN-VALUE/EIGENVECTOR ANALYSIS ON A. ONE SHOULD CALL THE EI PACK PROGRAM TRED2, FOLLOWED BY TQLP (OR USE THE EISPAC CONTROL PROGRAM).

CONVERGENCE FAILURES

IF CONVERGENCE FAILURES OCCUR, FIRST CHECK FOR INCORRECT CODING OF THE SUBROUTINE ADA. CHECK ESPECIALLY THE ACTION OF ISEL, AND THE COMPUTATION OF THE PARTIAL DERIVATIVES. IF THESE ARE CORRECT. TRY SEVERAL STARTING GUESSES FOR ALF. IF A0A IS CODED CORRECTLY, AND IF ERROR RETURNS IERR = -2 OR -8 PERSISTENTLY OCCUR, THIS IS A SEN OF LLCONDITIONING, WHICH MAY BE CAUSED BY SEVERAL THINGS. ONE IS POOR SCALING OF THE PARAMETERS: ANOTHER IS AN UNFORTUNATE INITIAL GUESS FOR THE PARAMETERS, STILL ANOTHER IS A POOR CHOICE OF THE MODEL.

ALGORITHM

THE RESIDUAL R IS MODIFIED TO INCORPORATE, FOR ANY FIXED ALF, THE OPTIMAL LINEAR PARAMETERS FOR THAT ALF. IT IS THEN POSSIBLE TO MINIMIZE ONLY ON THE NONLINEAR PARAMETERS. AFTER THE OPTIMAL VALUES OF THE NONLINEAR PARAMETERS HAVE BEEN DETER-MINED. THE LINEAR PARAMETERS CAN BE RECOVERED BY LINEAR LEAST SQUARES TECHNIQUES (SEE REF. 1).

THE MINIMIZATION **IS** BY A MODIFICATION OF OSBORNE'S (REF. MODIFICATION OF THE LEVENBERG-MARQUARDT ALGORITHM. INSTEAD 3) INSTEAD OF SOLVING THE NORMAL EQUATIONS WITH MATRIX

 $\begin{bmatrix} T & 2 \\ (J J + NU * D), \end{bmatrix}$ WHERE J = D(ETA)/D(ALF),

STABLE ORTHOGONAL (HOUSEHOLDER) REFLECTIONS ARE USED ON A MODIFICATION OF THE MATRIX

(J)	
()	•
(NU*D)	-

WHERE D IS A DIAGONAL MATRIX CONSISTING OF THE LENGTHS OF THE COLUMNS OF J. THIS MARQUARDT STABILIZATION ALLOWS THE ${\tt ROUTINE}$ TO RECOVER FROM SOME RANK DEFICIENCIES IN THE JACOBIAN OSBORNE'S EMPIRICAL STRATEGY FOR CHOOSING THE MARQUARDT PARAM-ETER HAS PROVEN KEASONABLY SUCCESSFUL IN PRACTICE. (GAUSS-NEWTON WITH STEP CONTROL CAN BE OBTAINED BY MAKING THE CHANGE INDICATED BEFORE THE INSTRUCTION LABELED 5). A DESCRIPTION CAN BE FOUND IN REF. (3), AND A FLOW CHART IN (2), P. 22.

FOR REFERENCE, SEE

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- COMM. ACM 17. PP. 167-169 (MARCH, 1974). KAUFMAN. LINDA, 'A VARIABLE PROJECTION METHOD FOR SOLVING 5. SEPARABLE NONLINEAR LEAST SQUARES PROBLEMS', B.I.T. 15, 49-57 (1975).
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JOHN BOLSTAD COMPUTER SCIENCE DEPT., SERRA HOUSE STANFORD UNIVERSITY **JANUARY**, **1977**

DOUBLE PRECISION A(NMAX, LPP2), BETA(L), ALF(NL), T(NMAX, IV), 2 W(N), Y(N), ACUM, EPSI, GNSTEP, NU, PRJRES, R, RNEW, XNORM INTEGER B1, OUTPUT LOGICAL SKIP EXTERNAL ADA

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C		I F IER I F	(IERF R = 1 (NU	2 • EQ	. ø.	ĝ, (GO T O TC	09 30 30	9)	L, .		., р р		- -				, ,			
	25		CALL	OR	ESTI FAC2	NG (NL)	ΝΕΚ ΙΤ F P1,	OR NMA	ACC ACC	EPTA NU,	A (1	, B	1))	3 E IN I		TING				AND	
00000				S(TH W	OLVE HE T RITT	A RANS EN	NL X SFOR BY T	NL MED HE	UP RE RES	PER SIDU ULT	TRIA JAL DEL	NG (IN TA-	ULAF COI ALF	R SN L. L	YSTI .NL 1	EM F 2 OF	OR A)	DEI IS	_TA= 5 OV	/A⊔LF. ER−	
-	30 35		CALL DO 3 A	ВА(5 К (К,	CSUB = 1 B1)	(NM , N	MAX. L ALF	NL	, А +	(1, A(K,	B1)	, A L2)	(1,	LNI	.2)	>					

DATA EPSI /1.D-6/, ITMAX /4Ø/, OUTPUT /6/

```
NEW ALF(K) = ALF(K) + DELTA ALF(K)
00000
                    STEP TO THE NEW POINT NEW ALF, AND COMPUTE THE NEW NORM OF RESIDUAL. NEW ALF IS STORED IN COLUMN B1 OF A.
                                                                                                 1
            CALL DPA (L, NL, N, NMAX. LPP2, IV, T, Y, W, A(1, B1), ADA,
IERR, JPRINT, A, BETA, A(1, LP1), RNEW)
IF (IERR .NE. 2) GO TO 99
    40
      Х
            ITER = ITER +
                                1
            ITERIN = ITERIN +
                                     1
            SKIP = MOD(ITER, MODIT) .NE. O
            I F
                (SKIP) GO TO 45
                WRITE (OUTPUT. 203) ITER
WRITE (OUTPUT, 216) (A(K
                                    216) (A(K, B1), K = 1, NL)
                WRITE (OUTPUT. 207) ITERIN, RNEW
С
                (ITER .LT. ITMAX) GO TO 50
IERR = -1
CALL VARERR (IPRINT, IERR, 1)
    45
            ΙF
               GO TO 95
(RNEW - R .LT. EPS1*(R + 1.DØ)) GO TO 75
    50
            IF
С
C
C
                    RETRACT THE STEP JUST TAKEN
                IF (NU .NE. Ø.) GO TO 60
С
                                                               GAUSS-NEWTON OPTION ONLY
                GNSTEP = Ø.5*GNSTEP
                IF (GNSTEP .LT. EPS1) GO TO 95
DO 55 K = 1, NL
A(K, B1) = ALF(K) + GNSTEP*A(K, LNL2)
    55
                GO TO 40
С
                                                        ENLARGE THE MARQUARDT PARAMETER
    60
                NU = 1.5 * NU
                    (.NOT. SKIP) WRITE (OUTPUT. 206) NU
                1 F
                     (NU .LE. 1110.) GO TO 65
IERR = -2
                1 F
                     CALL VARERR (IPRINT, IERR, 1)
                     GO TO 95
                                                        RETRIEVE UPPER TRIANGULAR FORM
C
C
                                                        AND RESIDUAL OF FIRST STAGE.
                                                                                                  1
            DO 70 K ≈ 1, NL
    65
                                                                                                  ł
                KSUB = LP1 + K
                                 NLP1
                DO 70 J = K.
                     JSUB = LP1 + J
                     ISUB = NLPI + J
                    A(K, JSUB) = A(ISUB, KSUB)
    70
            GO TO 25
                                                        END OF INNER ITERATION LOOP
C
C
C
                ACCEPT THE STEP JUST TAKEN
    75 R ■ RNEW
        DO 80 K = 1, NL
ALF(K) = A(K, B1)
    80
С
                                                        CALC. NORM(DELTA ALF)/NORM(ALF)
        ACUM = GNSTEP*XNORM(NL, A(1, LNL2))/XNORM(NL, ALF)
0
0
0
0
0
0
0
                IF ITERIN IS GREATER THAN 1, A STEP WAS RETRACTED DURING THIS OUTER ITERATION.
            ULERIN .EQ. 1) NU = Ø.5*NU
(SKIP) GO TO 85
WRITE (OUTPUT, 2110) NU
WRITE (OUTPUT 2110) NU
R = 3
         1 F
        ΙF
    85 IERR = 3
```

IF (PRJRES .GT. EPS1*(R + 1.DØ)) GO TO 5 END OF OUTER ITERATION LOOP CCCCC CALCULATE FINAL QUANTITIES -- LINEAR PARAMETERS, RESIDUALS, COVARIANCE MATRIX, ETC. 90 IERR = ITER95 IF (NL .GT. Ø) CALL DPA(L, NL, N, NMAX, LPP2, IV, T. Y, W, ALF, X ADA, 4, IPRINT, A, BETA, A(1, LP1), R) CALL POSTPR(L, NL, N, NMAX, LNL2, EPS1, R, IPRINT, ALF, W, A, X A(1, LP1), BETA, IERR) 99 RETURN С 200 FORMAT (9H NU =, E15.7) ITERATION, 14, 24H 203 FORMAT (12HØ NONLINEAR PARAMETERS) 206 FORMAT (25H STEP RETRACTED, NU = E15.7) 207 FORMAT (1HØ, I5, 2ØH NORM OF RESIDUAL =, E15.7) NORM(DELTA-ALF) / NORM(ALF) =, E12.3) 208 FORMAT (34H FORMAT (1HØ, 7E15.7) 216 END С SUBROUTINE ORFACI(NLP1, NMAX, N, L, IPRINT, B, PRJRES, IERR) С C C HOUSEHOLDER REDUCTION OF STAGE 1: (DR'. R3) NL (DR - R2) ТΟ R4) 0 N-L-NL NL NL 1 1 WHERE DR = -D(Q2)*Y IS THE DERIVATIVE OF THE MODIFIED RESIDUAL PRODUCED BY DPA, R2 IS THE TRANSFORMED RESIDUAL FROM DPA, AND DR' IS IN UPPER TRIANGULAR FORM (AS IN REF. (2), P. 18) DR IS STORED IN ROWS L+1 TO N AND COLUMNS L+2 TO L + NL + 1 OF THE MATRIX A (I.E., COLUMNS 1 TO NL OF THE MATRIX B). STORED IN COLUMN L + NL + 2 OF THE MATRIX A (COLUMN NL B). FOR K = 1, 2, ..., NL, FIND REFLECTION I = U U' WHICH ZEROES B(I, K), I = L+K+1, ..., N. DOUBLE PRECISION ACUM, ALPHA, B(NMAX, NLP1), BETA. DSIGN, PRJRES, X U, XNORM С NL = NLPI - 1 NL23 = **2*NL** + 3 LP1 = L + 1С DO 30 K = 1, NL LPK = L + K ALPHA = DSIGN(XNORM(N+1-LPK, B(LPK, K)), B(LPK, K))ALPHA = DSIGN(ANORNAMU = B(LPK, K) + ALPHAB(LPK, K) = UBETA = ALPHA UIF (ALPHA .NE. O.O) GO TO 13 COLUMN WAS ZERO С IERR = -8 CALL VARERR (IPRINT, IERR, LP1 + K) GO TO 99 C C APPLY REFLECTIONS TO REMAINING COLUMNS OF B AND TO RESIDUAL VECTOR. KP1 = K + 113 DO 25 J = KP1, NLPI

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. .

. ACUM = O.O DO 20 ∎ ≖ LPK, Ν 20 25 30 С PRJRES = XNORM(NL, B(LP1, NLP1)) 0000 SAVE UPPER TRIANGULAR FORM AND TRANSFORMED RESIDUAL, FOR USE IN CASE A STEP IS RETRACTED. ALSO COMPUTE COLUMN LENGTHS. IF (IERR .EQ. 4) GO TO 99 DO 58 K = 1, NL LPK = L + KDO 40 J = KNLPI JSUB ≈ NLPI + J B(K, J) = B(LPK, J) B(JSUB, K) = B(LPK, J) B(NL23, K) = XNORM(K, B(LP1, K)). 40 5Ø 5 С 99 RETURN END С SUBROUTINE ORFAC2(NLP1, NMAX, NU, B) SPECIAL HOUSEHOLDER REDUCTION OF STAGE 2: - R5) DR' R3 J (DR'' NI Ł ----) ~ -. R4 N-L-NL ø R 4 то 0) . - --1 NL (NU*D 0) C 0 R 6 .) . NL 1 NL 1 WHERE **DR**', R3, **AND** R4 ARE AS IN **ORFAC1**, NU **IS** THE MARQUARDI PARAMETER, D **IS** A DIAGONAL MATRIX CONSISTING OF THE LENGTHS OF THE COLUMNS OF DR', AND DR'' IS IN UPPER TRIANGULAR FORM. DETAILS IN (1), PP. 423-424. NOTE THAT THE (**N-L-NL**) BAND OF TROOPS AND P4 ARE OMITTED IN STORAGE. DOUBLE PRECISION ACUM, ALPHA, B(NMAX, NLPI). BETA, DSIGN, NU, U, X XNORM С NL = NLPI = 7 NL2 = 2*NL 1 NL23 = NL2 + 3DO 30 K = 1. NL KP1 = K + 1 NLPK = NL + K NLPKM1 = NLPK - 1 B(NLPK, K) = NU * B(NL23, K)B(NL, K) = B(K, K)ALPHA = DSIGN(XNORM(K+1, B(NL, K)), B(K, K)) U = B(K, K) + ALPHABETA = ALPHA U B(K, K) = -ALPHATHE K-TH REFLECTION MODIFIES ONLY ROWS K, C C NL+1, NL+2, ..., NL+K, AND COLUMNS K TO NL+1. DO 30 J = KP1, NLP1

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B(NLPK, J) = 8. ACUM = U * B(K,J) DO 20 ■ = NLP1, NLPKM1 ACUM = ACUM + B(I,K) * B(I,J) 20 ACUM = ACUM / BETA B(K,J) = B(K,J) = U * ACUMDO 30 I = NLP1, NLPK B(I,J) = B(I,J) = B(I,K) * ACUM30 С RETURN END С SUBROUTINE DPA (L, NL, N. NMAX, LPP2, IV, T. V, W, ALF, ADA, ISEL, X IPRINT, A, U, R, RNORM) COMPUTE THE NORM OF THE RESIDUAL (IF ISEL = 1 OR 2), OR THE $(N-L) \times NL$ der vat ve of the mod feed rest dual (N-L) vector Q2*Y (IF ISEL = 1 OR 3). Here Q * PHI = S, I.E., L (Q1) () (S . R1 . F 1 PHI Y D(PHI) = (----{ 0 {-<u>a</u>2-} {) R 2 ∎ F2) N-L .) L Ρ Ν 1 Ρ 1 Т WHERE ρ is N x N ORTHOGONAL, AND S IS L X L UPPER TRIANGULAR. THE NORM OF THE RESIDUAL = NORM(R2), AND THE DESIRED DERIVATIVE ACCORDING TO REF. (5), IS D(Q2 * Y) = -Q2 * D(PHI) * S * Q1 * Y.DOUBLE PRECISION A(NMAX, LPP2), ALF(NL), T(NMAX, IV), W(N), Y(N), X ACUM, ALPHA, BETA, RNORM, DSIGN, DSQRT, SAVE, R(N), U(L), XNORM INTEGER FIRSTC, FIRSTR, INC(12, 8) LOGICAL NOWATE, PHILP1 EXTERNAL ADA С (ISEL .NE. 1) GO TO 3 LP1 = L + 1 IF LNLZ = L + 2 + NL LP2 = L + 2 LP2 = L + 2 LPPI = LPP2 - 1FIRSTC = 1LASTC = LPPJ FIRSTR = LP1 CALL INIT(L, NL, N, NMAX, LPP2, IV, T, W, ALF, ADA, ISEL. IPRINT. A. INC. NCON. NCONP1, PHILP1, NOWATE) IF (ISEL .NE. 1) GO TO 99 Х GO TO 30 С 3 CALL ADA (LP1, NL, N, NMAX, LPP2, IV, A, INC, T, ALF, MINØ(ISEL, X 3)) IF (ISEL .EQ. 2) GO TO 6 ISEL = 3 OR 4С FIRSTC = LP2 LASTC = LPPIFIRSTR = (4 - ISEL)*L + 1 GO TO 50 С ISEL = 2 6 FIRSTC = NCONP1 LASTC = LP1 IF (NCON .EQ. Ø) GO TO 30

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IF (A(1, NCON) .EQ. SAVE) GO TO 30 ISEL = -7 CALL VARERR (IPRINT, ISEL, NCON) GO TO 99 С ISEL = 1 OR 230 IF (PHILP1) GO TO 48 DO 35 I = 1, N 35 R(I) = Y(I)GO TÒ 50 DO 45 I = 1, NR(I) = Y(I) - R(I) 40 45 WEIGHT APPROPRIATE COLUMNS С 50 IF (NOWATE) GO TO 58 DO 55 I = 1, N ACUM = W(I) DO 55 J = FIRSTC, LASTC A(I, J) = A(I, J) ACUM 55 COMPUTE ORTHOGONAL FACTORIZATIONS BV HOUSEHOLDER REFLECTIONS. IF ISEL = 1 OR 2, REDUCE PHI (STORED IN THE FIRST L COLUMNS OF THE MATRIX A) TO UPPER TRIANGULAR FORM, (Q*PHI = S), AND TRANSFORM Y (STORED IN COLUMN L+1), GETTING Q*V = R. IF ISEL = 1. ALSO TRANSFORM J = D PHI (STORED IN COLUMNS L+2 THROUGH L+P+1 OF THE MATRIX A). GETTING Q*J = F. IF ISEL = 3 OR 4. PHI HAS ALREADY BEEN REDUCED, TRANSFORM ONLY J. S, R, AND F OVERWRITE PHI, Y, AND J, RESPECTIVELY, AND A FACTORED FORM OF Q IS SAVED IN U AND THE LOWER TRIANGLE OF PHI. 58 IF (L .EO. Ø) GO TO 75 DO 70 K = 1, L KP1 = K + 1 IF (ISEL .GE. 3 .OR. (ISEL .EQ. 2 .AND. K .LT.NCONP1)) GO TO 66<math>ALPHA = DSIGN(XNORM(N+1-K, A(K, K)), A(K, K))U(K) = A(K, K) + ALPHAA(K, K) = -ALPHAFIRSTC = KP1IF (ALPHA .NE. O.O) GO TO 66 ISEL = -8 Call Varerr (Iprint, ISEL, K) GO TO 99 C C APPLY REFLECTIONS TO COLUMNS FIRSTC TO LASTC. BETA = -A(K, K) * U(K) DO 70 J = FIRSTC, LASTC ACUM = U(K)*A(K, J) 66 DO 68 I = KP1, Ń ACUM = ACUM + A(I, K) * A(I, J)68 ACUM = ACUM / BETA A(K,J) = A(K,J) - U(K)*ACUM DO 70 I = KP1, N A(I, J) = A(I, J) = A(I, K) * ACUM70 С 75 IF (ISEL .GE. 3) GO TO 85 RNORM = XNORM(N-L, R(LP1)) IF (ISEL .EO. 2) GO TO 99 IF (NCON .GT. \emptyset) SAVE = A(1, NCON) С F2 IS NOW CONTAINED IN ROWS L+I TO N AND COLUMNS L+2 TO L+P+1 OF THE MATRIX A. NOW SOLVE THE L X L UPPER TRIANGULAR SYSTEM S*BETA = R1 FOR THE LINEAR PARAMETERS BETA. BETA CCCCC OVERWRITES R1. С 85 IF (L.GT. Ø) CALL BACSUB (NMAX, L, A, R)

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MAJOR PART OF KAUFMAN'S SIMPLIFICATION OCCURS HERE. COMPUTE THE DERIVATIVE OF ETA WITH RESPECT TO THE NONLINEAR PARAMETERS T* DETA Т D PHI(J) D PHI(L+1)(SUM BETA(J) Q F2*BETA 22 3 D ALF(K) D ALF(K) J=1 D ALF(K) AND STORE THE RESULT IN COLUMNS L+2 TO L+NL+1. IF ISEL NOT = 4, THE FIRST L ROWS ARE OMITTED. THIS IS -D(Q2)*Y. IF ISEL NOT = 4 THE RESIDUAL R2 = Q2*Y (IN COL. L+1) IS COPIED IF ISEL NOT TO COLUMN L+NL+2. OTHERWISE ALL OF COLUMN L+1 IS COPIED. DO 95 I ≖ FIRSTR, N IF (L .EQ. NCON) GO TO 95 M = LP1 DO 90 K = 1, NL ACUM = O. DO 88 J = NCONP1, L IF (INC(K, J) .EQ. 8) GO TO 88 M = M + 1ACIJM = ACUM + A(I, M) * R(J) CONT INUE 88 KSUB = LP1 + KIF (INC(K, LP1) .EQ. Ø) GO TO 90 M = M + 1 ACUM = ACUM + A(I, M)A(I, KSUB) = ACUMA(I, LNL2) = R(I)90 95 С 99 RETURN END С SUBROUTINE INIT(L, NL, N, NMAX, LPP2, IV, T, W, ALF, ADA, ISEL, X IPRINT. A, INC, NCON, NCONP1, PHILP1, NOWATE) С CHECK VALIDITY OF INPUT PARAMETERS, AND DETERMINE NUMBER OF 0000 CONSTANT FUNCTIONS. Ċ DOUBLE PRECISION A(NMAX, LPP2), ALF(NL), T(NMAX, IV), W(N), X DSQRT INTEGER OUTPUT, P, INC(12, 8) LOGICAL NOWATE, PHILP1 DATA OUTPUT /6/ С LP1 = L + 1LNL2 = L + 2 + NL CHECK FOR VALID INPUT IF (L.GE, O.AND. NL.GE. I.AND. L+NL.LT. N.AND. LNL2.LE. X LPPE AND. 2*NL + 3 .LE. NMAX AND. N.LE. NMAX AND. X IV.GT.O.AND. .NOT. (NL.EQ.O.AND. L.EQ.Ø)) GO TO 1 С |SEL = -4CALL VARERR (IPRINT, ISEL, 1) GO TO 99 С 1 | F (L.EQ. I.OR. NL.) DO 2 J = 1, LP1 DO 2 K = 1, NL 2 INC(K, J) = 0 NL .EQ. Ø) GO TO 3 С 3 CALL ADA (LP1, NL, N, NMAX, LPPZ, IV, A, INC, T, ALF, ISEL)

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С NOWATE = .TRUE. DO 9 ■ = 1, N NOWATE = NOWATE .AND. (W(I) .EQ. 1.Ø) IF (W(I) .GE. Ø.) GO TO 9 С ERROR IN WEIGHTS ISEL = -6 CALL VARERR (IPRINT, ISEL, I) GO TO 99 W(I) = DSQRT(W(I))9 С NCOH = LNCONP1 = LP1PHILPI = L .EQ. O IF (PHILP1 .OR. NL .EQ. Ø) GO TO 99 CHECK INC MATRIX FOR VALID INPUT AND DETERMINE NUMBER OF CONSTANT FCNS. C C P = ODO 11 J = 1, LP1 IF (P.EQ. Ø) NCONPI = J DO 11 K = 1, NL INCKJ = INC(K, J) IF (INCKJ .NE. O .AND. INCKJ .NE. 1) GO TO 15 IF (INCKJ .EQ. 1) P = P + 1CONTINUE 11 С NCON = NCONPI - 1 IF (IPRINT .GE. Ø) WRITE (OUTPUT, 21Ø) NCON IF (L+P+2 .EQ. LPP2) GO TO 20 С INPUT ERROR IN INC MATRIX 15 ISEL. = -5 CALL VARERR (IPRINT, ISEL, 1) GO TO 99 С DETERMINE IF PHI(L+1) IS IN THE MODEL. 20 DO 25 K = 1. NL 25 IF (INC(K, LP1), EQ. 1) PHILPI = .TRUE. С 99 RETURN 210 FORMAT (33HØ NUMBER OF CONSTANT FUNCTIONS ™, 14 /) END SUBROUTINE BACSUB (NMAX, N, A, X) CCCC BACKSOLVE THE N X N UPPER TRIANGULAR SYSTEM $A^*X = B$. THE SOLUTION X OVERWRITES THE RIGHT SIDE B. DOUBLE PRECISION A(NMAX, N), X(N), ACUM С X(N) = X(N) / A(N, N) IF (N .EQ. 1) GO TO 30 NP1 = N + 1DO 20 IBACK = 2, N = NP1 - IBACK I = N-1, N-2, ..., 2, 1 IP1 = I + 1 С ACUM = X(I)DO 10 J = IP1, N ACUM = ACUM = A(I,J)*X(J) 10 X(I) = ACUM / A(I,I)20 С 30 RETURN END SUBROUTINE POSTPR(L, NL, N, NMAX, LNL2, EPS, RNORM, IPRINT, ALF, X W, A, R, U. IERR) С

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CALCULATE RESIDUALS, SAMPLE VARIANCE, AND COVARIANCE MATRIX. ON INPUT, U CONTAINS INFORMATION ABOUT HOUSEHOLDER REFLECTIONS C C Č C FROM DPA. ON OUTPUT. IT CONTAINS THE LINEAR PARAMETERS. DOUBLE PRECISION A(NMAX, LNL2), ALF(NL), R(N), U(L), W(N), ACUM, X EPS, PRJRES, RNORM. SAVE, DABS INTEGER OUTPUT DATA OUTPUT /6/ С LP1 = L + 1LPNL ≖ LNL2 - 2 LNL1 = LPNL + 1 DO 10 I = 1, N W(I) = W(I)**2 10 С С UNWIND HOUSEHOLDER TRANSFORMATIONS TO GET RESIDUALS. Č C AND MOVE THE LINEAR PARAMETERS FROM R TO U. IF (L .EQ. Ø) GO TO 30 DO 25 KBACK = 1, L K = LP1 - KBACK $\mathsf{KP1} = \mathsf{K} + \mathsf{1}$ ACUM = O. DO 20 I = KP1, N ACUM = ACUM + A(I, K) * R(I) 20 SAVE = R(K)R(K) = ACUM / A(K, K) ACUM = -ACUM / (U(K) * A(K, K))U(K) = SAVEDO 25 I = KP125 I = KP1, N R(I) = R(I) - A(I, K)*ACUM 25 COMPUTE MEAN ERROR С 30 ACUM **≍** O. DO 35 I = 1. N ACUM = ACUM + R(I) 35 SAVE = ACUM / N С THE FIRST L COLUMNS OF THE MATRIX HAVE BEEN REDUCED TO CCCCC UPPER TRIANGULAR FORM IN DPA. FINISH BY REDUCING ROWS L+I TO N AND COLUMNS L+2 THROUGH L+NL+1 TO TRIANGULAR FORM. THEN SHIFT COLUMNS OF DERIVATIVE MATRIX OVER ONE TO THE LEFT TO BE ADJACENT TO THE FIRST L COLUMNS. С IF (NL .EQ. Ø) GO TO 45 CALL ORFACI(NL+1, NMAX, N, L, IPRINT, A(1, L+2), PRJRES, 4) DO 40° I = 1, N A(I, LNL2) = R(I) DO 40 K = LP1, LNL1 A(I, K) = A(I, K+1) 45 С COMPUTE COVARIANCE MATRIX 45 A(1, LNL2) = RNORM ACUM = RNORM*RNORM/(N = L - NL) A(2, LNL2) = ACUM CALL COV(NMAX, LPNL, ACUM, A) С IF (IPRINT .LT. \emptyset) GO TO 99 WRITE (OUTPUT. $2\emptyset$ 9) IF (L.GT. \emptyset) WRITE (OUTPUT. $21\emptyset$) (U(J), J = 1, L) IF (NL.GT. \emptyset) WRITE (OUTPUT, 211) (ALF(K), K = 1, NL) WRITE (OUTPUT, 214) RNORM. SAVE, ACUM IF (DABS(SAVE) .GT. EPS) WRITE (OUTPUT, 215) WRITE (OUTPUT, $2\emptyset$ 9) RETURN 99 RETURN С

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209 FORMAT (1HØ, **5Ø(1H'))** 210 FORMAT (2ØHØ LINEAR 211 FORMAT (23HØ NONLINE) LINEAR PARAMETERS // (7E15.7)) NONLINEAR PARAMETERS // (7E15.7)) NORM OF RESIDUAL =, E15.7, 33H EXPECTED ERROR OF OBS 214 FORMAT (21HØ XERVATIONS =, E15.7, / 39H ESTIMATED VARIANCE OF OBSERVATIONS =. X E15.7) 215 FORMAT (95H WARNING -- EXPECTED ERROR OF OBSERVATIONS IS NOT ZERO X. COVARIANCE MATRIX MAY BE MEANINGLESS. /) Х. END SUBROUTINE COV(NMAX, N, SIGMA2, A) COMPUTE THE SCALED COVARIANCE MATRIX OF THE L + NL PARAMETERS. THIS INVOLVES COMPUTING 2 -1 -T SIGMA * T * T WHERE THE (L+NL) X (L+NL) UPPER TRIANGULAR MATRIX T IS DESCRIBED IN SUBROUTINE POSTPR. THE RESULT OVERWRITES THE FIRST L+NL ROWS AND COLUMNS OF THE MATRIX A. THE RESULTING MATRIX IS SYMMETRIC. SEE REF. 7, PP. 67-70, 281. DOUBLE PRECISION A (NMAX, N), SUM, SIGMA2 С DO 10 J = 1, N A(J, J) = 1./A(J, J) 10 C C C INVERT T UPON ITSELF IF (N .EQ. 1) GO TO 70 NM1 = N - 1 $DO \ 6 \ 0 \ I = 1. \ NM1$ SUM = O.DO 50 M = I. JM1 SUM = SUM + A(I, M) * A(M, J) A(I, J) = -SUM * A(J, J) 50 60 C C NOW FORM THE MATRIX PRODUCT С 70 DO 90 ∎ = 1, N DO 90 J = I, N SUM = O. DO 80 M = J, N SUM = SUM + A(I, M) * A(J, M) SUM = SUM * SIGMA2 8Ø . A(I, J) = SUM A(J, I) = SUM 90 С RETURN END SUBROUTINE VARERR (IPRINT, IERR, K) **C** C C PRINT ERROR MESSAGES INTEGER ERRNO, OUTPUT DATA OUTPUT /6/ С IF (IPRINT .LT. Ø) GO TO 99 ERRNO = IABS(IERR)

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GO TO (1, 2, 99, 4, 5, 6, 7, 8), ERRNO

С 1 WRITE (OUTPUT, 1Ø1) GO TO 99 WRITE (O (OUTPUT, 102) 2 GO TO 99 WRITE (OUTPUT, 1Ø4) GO TO 99 WRITE 4 WRITE (OUTPUT, 105) 5 GO TO 99 WRITE (OUTPUT, 106) K 6 GO TO 99 7 WRITE (OUTPUT, 107) K GO TO 99 WRITE (OUTPUT, 108) K 8 С 99 RETURN PROBLEM TERMINATED FOR EXCESSIVE ITERATIONS //) 101 FORMAT (46HØ 102 FORMAT (49HØ PROBLEM TERMINATED BECAUSE OF ILL-CONDITIONING //) 104 FORMAT (/ 5ØH INPUT ERROR IN PAR METER L, NL, N, LPP2, OR NMAX. / 105 FORMAT (68HØ ERROR -- INC MATRIX IMPROPERLY SPECIFIED, OR DISAGRE XES WITH LPP2. 106 FORMAT (19HØ 1) ERROR -- WEIGHT(, I4, 14H) IS NEGATIVE. /) ERROR -- CONSTANT COLUMN , I3, 37H MUST BE COMPUTED . = 1. /) 107 FORMAT (28HØ XONLY WHEN ISEL = 1. /) 3 FORMAT (33HØ CATASTROPHIC FAILURE ⁻⁻ COLUMN , 14, 28H IS ZERO, SE 108 FORMAT (33HØ XE DOCUMENTATION. /) END DOUBLE PRECISION FUNCTION XNORM(N, X) 00000 COMPUTE THE L2 (EUCLIDEAN) NORM OF A VECTOR, MAKING SURE TO AVOID UNNECESSARY UNDERFLÓWS. NO ATTEMPT IS MADE TO SUPPRESS OVERFLOWS. DOUBLE PRECISION X(N), RMAX. SUM, TERM, DABS, DSQRT C C FIND LARGEST (IN ABSOLUTE VALUE) ELEMENT RMAX = O.DO 10 I = 1, IF (DABS(X(I)) .GT. RMAX) RMAX = DABS(X(I)) CONTINUE 10 С SUM = O. IF (RMAX .EO. Ø.) GO TO 30 DO 2Ø I = 1. N $T \in RM = O.$ IF (RMAX + DABS(X(I)) .NE. RMAX) TERM ≈ X(I)/RMAX SUM = SUM + TERM*TERM 20 С 30 XNORM = RMAX*DSQRT(SUM) 99 RETURN END

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SAMPLE	INP'UT
48.5 1 0.022 0 50	
0.022 O 59 1 0.117 0.167 0.667 1.167 2.017 2.167 2.417 2.667 2.917 3.167 3.917 4.167 4.417 4.667 5.167 5.417 5.667 5.417 5.667 5.417 5.667 6.167 6.677 6.767 6.767 7.167 7.417 7.917 8.917 10.917 10.917 10.917 12.917 12.917 12.917 12.917 12.917 12.917 12.917 22.917 23.917 4.917 15.917 16.917 16.917 17.917 12.917 12.917 22.917 23.917 24.917 22.917 36.917 40.917 4.917 4.917 5.917 5.917 16.917 17.917 12.917 22.917 36.917 40.917 4.917 4.917 5.917 5.917 5.917 5.917 5.917 5.917 5.917 5.917 7.417 7.917	0.738 1.900 1.000 1.008 0.825 0.923 1.000 0.996 1.000 0.9988 0.945 0.956 0.932 0.901 0.885 0.873 0.838 0.841 0.830 0.794 0.794 0.794 0.794 0.773 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.749 0.734 0.619 0.615 0.559 0.5531 0.527 0.488 0.448 0.460 8.472 0.331 0.250 0.333 0.250 0.234
52.917 60.917 68.91 7	0.182 0.151 8.873

76.917	0.105
84.917	0.081
92.917	0.063

*******	**SAMPLE OUTPUT*******
0	INJECTION TIME
48.50 O	NUMBER OF NONLINEAR PARAMETERS
0	INITIAL ESTIMATES OF NONLIN. PARAM.
0.0220 O	NUMBER OF LINEAR PARAMETERS
0 0	NUMBER OF OBSERVATIONS
59 O	NUMBER OF INDEPENDENT VARIABLES
$ \begin{array}{c} 1\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\ 0\\$	INDEPENDENT VARIABLES DEPENDENT VARIRBLES 0.117 0.730 0.167 1.000 0.667 1.000 1.167 1.000 1.667 0.825 2.017 0.923 2.167 1.000 2.417 0.996 2.667 1.000 2.917 0.988 3.167 0.945 3.417 0.996 3.667 0.932 3.917 0.901 4.167 0.885 4.417 0.873 4.667 0.830 4.917 0.841 5.167 0.794 5.917 0.794 5.917 0.782 6.667 0.785 6.917 0.773 7.167 0.749 8.417 0.749 8.917 0.734 9.917 0.734 9.917 0.734 9.917 0.615 14.917 0.619

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0	20.917	0.468
0	21,917	0.472
õ	22 917	8 448
0		0.440
0	23.917	0.400
0	24.917	0.409
0	28.917	0.381
Ō	32.917	0.333
0	36.917	0.273
Ō	40.917	0.250
Ō	44.917	0.234
Ō	48.917	0.234
0	52.917	0.182
0	60.917	0.151
0	68.917	0.073
0	76.917	0.105
0	84.917	0.081
0	92.917	0.063
ØINCIDENCE	MATRIX INC(1,1)=	
1		

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1.

O NORM OF RESIDUAL = Ø.1757343e+Ø1 0 $\mathsf{NU} = \emptyset.1\emptyset\emptyset\emptyset\emptyset\emptyset\emptyset\emptysete+\emptyset1$ ITERATION NONLINEAR PARAMETERS o 1 Ø.51511Ø9e-Ø1 0 1 NORM OF RESIDUAL = Ø.114912Øe+Ø1 NU = Ø.5ØØØØØØe+ØØ 0 NORM(DELTA-ALF) / NORM(ALF) = Ø.573e+ØØ NONLINEAR PARAMETERS 0 ITERATION 2 Ø.1Ø24895e+ØØ O 1 NORM OF RESIDUAL = Ø.6024982e+00 NU = Ø.2500000e+00 0 NORMIDELTA-ALF) / NORM(ALF) = Ø.497e+ØØ ITERATION 3 NONLINEAR PARAMETERS 0 0 Ø.151Ø6Ø7e+ØØ 1 NORM OF RESIDUAL = Ø.4143Ø13e+ØØ O NU = $\emptyset.125\emptyset\emptyset\emptyset\emptysete+\emptyset\emptyset$ NORMIDELTA-ALF) / NORM(ALF) = Ø.34 ERATION 4 NONLINEAR PARAMETERS Ø.322e+ØØ ITERATION 0 0 Ø.1682783e+ØØ 1 NORM OF RESIDUAL = Ø.4030864e+00 0 NU = Ø.625ØØØØe-Ø1 NORM(DELTA-ALF) / NORM(ALF) # ERATION 5 NONLINEAR PARAM Ø.102e+00 NONLINEAR PARAMETERS 0 ITERATION Ø.16873Ø9e+ØØ 0 1 NORM OF RESIDUAL = Ø.4030805e+00 1 NORM OF REGIONAL NU = Ø.3125ØØØ€-Ø1 NORMIDELTA-ALF) / NORM(ALF) = Ø.20 FRATION 6 NONLINEAR PARAMETERS 0 Ø.268e-Ø2 0 ITERATION Ø.1687Ø41e+ØØ 0 1 NORM OF RESIDUAL = Ø.4030805e+00 NU = Ø.1562500e-01 0 NORM(DELTA-ALF) / NORM(ALF) = Ø.1 ERATION 7 NONLINEAR PARAMETERS Ø.159e-Ø3 ITERATION 0 Ø.1687Ø58e+ØØ 0 1 NORM OF RESIDUAL = Ø.4Ø3Ø8Ø6e+ØØ NU = Ø.78125ØØe-Ø2 NORMIDELTA-ALF) / NORM(ALF) = Ø.1Ø3 0 Ø.102e-04 NONLINEAR PARAMETERS 0 ITERATION 8 Ø.1687Ø57e+ØØ 0 1 NORM OF RESIDUAL = Ø.4Ø3Ø8Ø4e+ØØ NU = Ø.39Ø625Øe-Ø2 0 Ø Ø.76Øe-Ø6 **O** NONLINEAR PARAMETERS Ø.1687Ø57e+ØØ O NORM OF RESIDUAL = Ø.4Ø3Ø8Ø4e+ØØ EXPECTED ERROR OF OBSERVATIONS = -μ.16323Ø9e-Ø2 ESTIMATED VARIANCE OF OBSERVATIONS = Ø.28Ø1272e-Ø2 WARNING -- EXPECTED ERROR OF OBSERVATIONS IS NOT ZERO. COVARIANCE MATRIX MAY BE MI COVARIANCE MATRIX MAY BE MEANINGLES

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ØMEASURMENT	CALCULATED
TIME	CONCENTRATION
0.117 0.667 1.167 2.667 2.817 2.167 2.417 2.667 2.917 3.167 3.417 3.667 3.917 4.167 4.417 4.667 5.167 5.167 5.167 5.167 5.417 5.667 5.917 6.6677 5.917 7.167 7.417 8.917 7.917 8.917 9.917 10.917 12.917 13.917 15.917 15.917 16.917 15.917 16.917 17.917 12.917 22.917 22.917 23.917 22.917 22.917 22.917 32.9	0.994 0.992 0.970 0.950 0.932 0.919 0.914 0.906 0.897 0.889 0.881 0.874 0.866 0.858 0.851 0.844 0.822 0.822 0.822 0.822 0.822 0.822 0.796 0.776 0.776 0.776 0.776 0.7728 0.776 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.728 0.764 0.663 0.663 0.663 0.6643 0.623 0.604 0.586 0.5518 0.502 0.487 0.472 0.447 0.472

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36.917	0.308
10.917	0.275
44.917	0.246
48.917	0.221
52.917	0.200
60.917	0.166
68.917	0.139
76.917	0.119
04.917	0.103
92.917	0.091

APPENDIX E: Listing of Programs to Perform Nonlinear Curve Fitting of the Convection-Dispersion Model With a Sample Input and a Corresponding output

```
LISTING OF THE CURVEFITING PROGRAMS FOR THE CONVECTION DISPERSION
MODEL CONTINUOUS INJECTON CASE
         ********
                PROGRAM BEGINS
      *******
                           *******
                MAIN PROGRAM ( PROGRAM CURFIT )
      ********
                DEFINITION OF THE PARAMETERS
             TJ : INJECTION PERIOD
             T(I): TIME STEPS AT WHICH THE CONCENTRATIONS ARE MEASURED
                  DURING BAKFLOW PERIOD
              C : CONCENTRATIONS CALCULATED BY USING THE OPTIMUM VALUE OF
                 THE NONLINEAR PARAMETER
      **********************
              INPUT DATA ARE STORED IN THE FILE "INFILE"
              OUTPUT ARE STORED IN FILE "OUTFILE"
              VALUES OF THE NONLINEAR PARAMETER AND OTHER STATISTICAL
              INFORMATION OF THE CURFIT ARE DISPLAYED ON THE SCREEN TOO
      *****
      IMPLICIT REAL*8(A-H, 0-Z)
    DIMENSION Y(100), T(100), ALF(2), BETA(2), W(100), A(100,6)
*, C(100), INC(3,3)
      EXTERNAL ADA
      COMMON TJ
      OPEN(UNIT=8,FILE='INFILE')
OPEN(UNIT=9,FILE='OUTFILE')
C C C C C C C
    SET PARAMETERS FOR VARPRO
      READ(8,*)TJ
WRITE(9,113)
      FORMAT(1HØ,1Øx,'INJECTION TIME'//)
WRITE(9, )TJ
 113
        NMAX=1ØØ
        LPP2=3
        IPRINT=1
        IV = 1
        N=53
        L=Ø
      DO 1 I=1.N
       W(I)=1.Ø
   1
C
C
```

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```
READ DATA
000000
       NL IS THE NUMBER OF NONLINEAR PARAMETERS
         READ(8,10) NL
FORMAT(13)
  10
         WRITE(9,12) NL
          FORMAT(1HØ,10X,'NUMBER OF NONLINEAR PARAMETERS'//(13))
  12
000000
       INITIAL ESTIMATES OF THE NONLINEAR PARAMETERS
         READ(8,15) ALF(1)
  15
          FORMAT(F7.3)
         WRITE(9,20) ALF(1)
          FORMAT(1HØ,1ØX,'INITIAL ESTIMATES OF NONLIN. PARAM. '//(F7.3))
  20
CCCCCC
       L IS THE NUMBER OF LINEAR PARAMETERS
         READ(8,22) L
    22
          FORMAT(13)
         WRITE(9,25) L
          FORMAT(1HØ,10X, 'NUMBER OF LINEAR PARAMETERS'//(I3))
    25
C
C
C C C C
       N IS THE NUMBER OF OBSERVATIONS
         READ(8,30) N
FORMAT(13)
    30
         WRITE(9,35) N
           FORMAT(1HØ,1ØX, 'NUMBER OF OBSERVATIONS'//(I3))
    35
00000
       IV IS THE NUMBER OF INDEPENDENT VERIABLES
         READ(8,4Ø)IV
          FORMAT(13)
   40
          WRITE(9,45)IV
           FORMAT(1HØ,10X,'NUMBER OF INDEPENDENT VARIABLES '//(I3))
   45
00000000
          READ THE TIME AND MEASURED CONCENTRATION DATA T(I) and Y(I)
          READ(8,55) (T(I), Y(I), I=1, N)
    55
           FORMAT(2(F7.3.2X))
          WRITE(9.59)
         FORMAT(1HØ,1ØX,'INDEPENDENT VARIABLES '//,2(F7.3,3X))
WRITE(9,6Ø) (T(I),Y(I),I=1,N)
FORMAT(1HØ,'TIME vs MEASURED CONCENTRATIONS'//,2(F7.3,8X))
    59
    60
C
```

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00000 00000 0 **** SUBROUTINE ADA SUBROUTINE ADA(L,NL,N,NMAX,LPP2,IV,A,INC,T,ALF,ISEL) IMPLICIT REAL*8(A-H,O-Z) DIMENSION ALF(2),A(100,6),T(100),INC(3,3),B(100,6),Y(100) COMMON TJ 000000000 SKIP UNLESS ISEL IS EQUAL TO 1 SET INCIDENCE MATRIX, INC(K,J)=1 IF ALP(K) APPEARS IN PHI(J) IF(ISEL.EQ.2)GO TO 90 IF(ISEL.EQ.3)GO TO 165 (13E1.1)=1.Ø WRITE(9,7Ø) INC(1,1) FORMAT(1HØ,'INCIDENCE MATRIX INC(1,1)= '//(I3)) 70 C C CALCULATION OF THE MATRIX A C C C STOKE VALUES OF THE FUNCTION IN COLUMN NUMBER 1 OF A DO 81 I=1,N G=ALF(1)*(TJ-T(I))/(2.Ø*DSQRT(TJ+T(I))) 90 RESULT=DERF(G) $A(I,1) = \emptyset.5 * (1.\theta + RESULT)$ CONTINUE 81 WRITE(9,1Ø3)(A(I,1),I=1,N) FORMAT(1HØ,'COLUMN #1 OF A(I,J) MATRIX'//(F9.5)) 00**0**00 103 SKIP CALCULATION OF THE DEIVATIVES IF ISEL-2 IF(ISEL.EQ.2)GO TO 200 C C STORE VALUES OF THE DERIVATIVES IN COLUMN NUMBER 2 OF A С DO 170 I=1,N 165 B(I.1)=DEXP(-ALF(1)**2*(TJ**2-2.Ø*TJ*T(I)+T(I)**2)/(4.Ø*(TJ+T(I)))) A(I,2)=1./(2.*3.1415**Ø.5)*B(I,1)*(TJ-T(I))/DSQRT(TJ+T(I)) 170 CONTINUE WRITE(9,180)(A(I,2),I=1,N) FORMAT(1H0,'COLUMN #2 OF A(I,J) MATRIX'//(F9.5)) С С 180 200 CONTINUE RETURN END

SUBROUTINE VARPRO (L, NL, N. NMAX, LPP2, IV. T, V, W, ADA, A, X IPRINT, ALF, BETA, IERR)

GIVEN A SET OF N OBSERVATIONS. CONSISTING OF VALUES Y(1), Y(2), ..., Y(N) OF A DEPENDENT VARIABLE V, WHERE Y(I) CORRESPONDS TO THE IV INDEPENDENT VARIABLE(S) T(I,1), T(I,2), ..., T(I.IV), VARPRO ATTEMPTS TO COMPUTE A WEIGHTED LEAST SQUARES FIT TO A FUNCTION ETA (THE 'MODEL') WHICH IS A LINEAR COMBINATION

ETA(ALF, BETA: T) = SUM BETA * PHI (ALF; T) + PHI (ALF; T) J-1 J J J L+I

OF NONLINEAR FUNCTIONS PHI(J) (E.G., A SUM OF EXPONENTIALS AND/ OR GAUSSIANS). THAT IS, DETERMINE THE LINEAR PARAMETERS BETA(J) AND THE VECTOR OF NONLINEAR PARAMETERS ALF BY MINIMIZ-ING

THE **(L+1)-ST** TERM IS OPTIONAL, **AND** IS USED WHEN IT IS DESIRED TO FIX ONE OR MORE OF THE BETA'S (RATHER THAN LET THEM BE DETERMINED). VARPRO REQUIRES FIRST DERIVATIVES OF THE PHI'S.

NOTES :

A) THE ABOVE PROBLEM IS ALSO REFERRED TO AS 'MULTIPLE NONLINEAR REGRESSION'. FOR USE IN STATISTICAL ESTIMATION, VARPRO RETURNS THE RESIDUALS, THE COVARIANCE MATRIX OF THE LINEAR AND NONLINEAR PARAMETERS, AND THE ESTIMATED VARIANCE OF THE OBSERVATIONS.

B) AN ETA OF THE ABOVE FORM IS CALLED 'SEPARABLE'. THE CASE OF A NONSEPARABLE ETA CAN BE HANDLED BY SETTING L = O AND USING PHI(L+1).

C) VARPRO MAY ALSO BE USED TO SOLVE LINEAR LEAST SQUARES PROBLEMS (IN THAT CASE NO ITERATIONS ARE PERFORMED). SET NL = O.

()) THE MAIN ADVANTAGE **OF** VARPRO OVER OTHER LEAST SQUARES PROGRAMS **IS** THAT NO INITIAL GUESSES ARE NEEDED FOR THE LINEAR PARAMETERS. NOT ONLY DOES THIS MAKE IT EASIER TO USE, BUT IT OFTEN LEADS TO FASTER CONVERGENCE.

DESCRIPTION OF PARAMETERS

L	NUMBER OF LINEAR PARAMETERS BETA (MUST BE .GE. Ø).
NL	NUMBER OF NONLINEAR PARAMETERS ALF (MUST BE ,GE. Ø).
N	NUMBER OF OBSERVATIONS. N MUST RE GREATER THAN L + NL
	(I.E., THE NUMBER OF OBSERVATIONS MUST EXCEED THE
	NUMBER OF PARAMETERS).
IV	NUMBER OF INDEPENDENT VARIABLES T.
Т	REAL N BY IV MATRIX OF INDEPENDENT VARIABLES. T(I, J)
	CONTAINS THE VALUE OF THE I- TH OBSERVATION OF THE J-TH
	INDEPENDENT VARIABLE.
Y	fI-VECTOR OF OBSERVATIONS, ONE FOR EACH ROW OF T.
W	Id-VECTOR OF NONNEGATIVE WEIGHTS. SHOULD BE SET TO 1'S
	IF WEIGHTS ARE NOT DESIRED. IF VARIANCES OF THE
	INDIVIDUAL OBSERVATIONS ARE KNOWN, W(I) SHOULD BE SET
	TO 1./VARIANCE(I).

NC NL X (L+1) INTEGER INCIDENCE MATRIX. INC(K, J) = 1 | FNON-LINEAR PARAMETER ALF(K) APPEARS IN THE J-TH FUNCTION PHI(J). (THE PROGRAM SETS ALL OTHER INC(K, J) TO ZERO.) IF PHI(L+1) IS INCLUDED IN THE MODEL, THE APPROPRIATE ELEMENTS OF THE (L+1)-ST COLUMN SHOULD BE SET TO 1'S. INC IS NOT NEEDED WHEN L = O OR NL = O. CAUTION: THE DECLARED ROW DIMENSION OF INC (IN ADA) MUST CURRENTLY BE SET TO 12. SEE 'RESTRICTIONS' BE THE DECLARED ROW DIMENSION OF THE MATRICES A AND T. BEĹOW. NMAX IT MUST BE AT LEAST MAX(N, 2*NL+3). L+P+2, WHERE P IS THE NUMBER OF ONES IN THE MATRIX INC. THE DECLARED COLUMN DIMENSION OF A MUST BE AT LEAST LPPE. (IF L = O. SET LPP2 = NL+2. IF NL = O, SET LPP2 LPP2 L+2.) REAL MATRIX OF SIZE MAX(N, 2*NL+3) BY L+P+2. ON INPUT IT CONTAINS THE PHI(J)'S AND THEIR DERIVATIVES (SEE BELOW). ON OUTPUT, THE FIRST L+NL ROWS AND COLUMNS OF A WILL CONTAIN AN APPROXIMATION TO THE (WEIGHTED) А A WILL CONTAIN AN APPROXIMATION TO THE (WEIGHTED) COVARIANCE MATRIX AT THE SOLUTION (THE FIRST L ROWS CORRESPOND TO THE LINEAR PARAMETERS, THE LAST NL TO THE NONLINEAR ONES), COLUMN L+NL+1 WILL CONTAIN THE WEIGHTED RESIDUALS (Y - ETA), A(1, L+NL+2) WILL CONTAIN THE (EUCLIDEAN) NORM OF THE WEIGHTED RESIDUAL, AND A(2, L+NL+2) WILL CONTAIN AN ESTIMATE OF THE (WEIGHTED) VARIANCE OF THE OBSERVATIONS, NORM(RESIDUAL)**2/ $|N - L - NL\rangle$. INPUT-INTEGER CONTROLLING PRINTED OUTPUT. IF IPRINT POSITIVE, THE NONLINEAR PARAMETERS, THE NORM OF THE RESIDUAL, AND THE MARQUARDT PARAMETER WILL BE OUTPUT PR NT F PR NT IS EVERY IPRINT- TH ITERATION (AND INITIALLY. AND AT THE FINAL ITERATION). THE LINEAR PARAMETERS WILL BE PRINTED AT THE FINAL ITERATION. ANY ERROR MESSAGES WILL ALSO BE PRINTED. (IPRINT = 1 IS RECOMMENDED AT WILL ALSO BE PRINTED. (IPRINT = 1 IS RECOMMENDED AT FIRST.) IF IPRINT = 0. ONLY THE FINAL QUANTITIES WILL BE PRINTED. AS WELL AS ANY ERROR MESSAGES. IF IPRINT = -1, NO PRINTING WILL BE DONE. THE USER IS THEN RESPONSIBLE FOR CHECKING THE PARAMETER IERR FOR ERRORS. NL-VECTOR OF ESTIMATES OF NONLINEAR PARAMETERS (INPUT). ON OUTPUT IT WILL CONTAIN OPTIMAL VALUES OF IF IPRINT = ALF THE NONLINEAR PARAMETERS. BETA L-VECTOR OF LINEAR PARAMETERS (OUTPUT ONLY). INTEGER ERROR FLAG (OUTPUT): .GT. O - SUCCESSFUL CONVERGENCE, IERR IS THE NUMBER OF ERR ITERATIONS TAKEN. TERMINATED FOR TOO MANY ITERATIONS. TERMINATED FOR ILL-CONDITIONING (MAROUARDT PARAMETER TOO LARGE.) ALSO SEE IERR = -8 BELOW. INPUT ERROR IN PARAMETER N, L, NL, LPPE, OR NMAX. INC MATRIX IMPROPERLY SPECIFIED, OR P DISAGREES - 1 - 2 - 4 - 5 WITH LPP2. A WEIGHT WAS NEGATIVE. - 6 'CONSTANT' COLUMN WAS COMPUTED MORE THAN ONCE. CATASTROPHIC FAILURE - A COLUMN OF THE A MATRIX HAS - 7 - 8 SEE 'CONVERGENCE FAILURES' BECOME ZERO. BELOW. (IF IERR .LE. -4, THE LINEAR PARAMETERS, COVARIANCE MATRIX, ETC. ARE NOT RETURNED.) SUBROUTINES REQUIRED

NINE SUBROUTINES, DPA, ORFAC1, ORFAC2, BACSUB, POSTPR, COV, XNORM, INIT, AND VARERR ARE PROVIDED. IN ADDITION, THE USER MUST PROVIDE A SUBROUTINE (CORRESPONDING TO THE ARGUMENT ADA) WHICH, GIVEN ALF, WILL EVALUATE THE FUNCTIONS PHI(J) AND THEIR

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PARTIAL DERIVATIVES D PHI(J)/D ALF(K), AT THE SAMPLE POINTS T(I). THIS ROUTINE MUST BE DECLARED 'EXTERNAL' IN THE CALL IN THE CALLING ITS CALLING SEQUENCE IS PROGRAM. SUBROUTINE ADA (L+1, NL, N, NMAX. LPPZ, IV, A, INC, T, ALF, ISEL) THE USER SHOULD MODIFY THE EXAMPLE SUBROUTINE 'ADA' (GIVEN ELSEWHERE) FOR HIS OWN FUNCTIONS. THE VECTOR SAMPLED FUIJCTIONS PHI(J) SHOULD BE STORED IN THE THE VECTOR SAMPLED FOIDCHTONS FHI(J) SHOULD BE STORED TN THE FIRST N ROWS AND FIRST L+I COLUMNS OF THE MATRIX A, I.E., A(I, J) SHOULD CONTAIN PHI(J, ALF; T(I,1), T(I,2), ..., T(I,IV)), I = 1, ..., N: J = 1, ..., L (OR L+1). THE (L+1)-ST COLUMN OF A CONTAINS PHI(L+1) IF PHI(L+1) IS IN THE MODEL, OTHERWISE IT IS RESERVED FOR WORKSPACE. THE 'CONSTANT' FUNC-TIONS (THESE ARE FUNCTIONS PHI(J) WHICH DO NOT DEPEND UPON ANY NONLINEAR DARAMETERS ALE C T(I)*1) (LE ANY) MUST ADDEAP NONLINEAR PARAMETERS ALF. E.G., T(I)**J) (IF ANY) MUST APPEAR FIRST, STARTING IN COLUMN 1. THE COLUMN N-VECTORS OF NONZERO PARTIAL DERIVATIVES D PHI(J) / D ALF(K) SHOULD BE STORED SEQUENTIALLY IN THE MATRIX A IN COLUMNS L+2 THROUGH L+P+1. THE ORDER IS D PHI(1) D PHI(2) D PHI(L+1) D PHI(1) D PHI(L) ..., D ALF(1) D ALF(2) D ALF(1) D ALF(1)D ALF(1)D PHI(2) D PHI(L+1) D PHI(1) D PHI(L+1). . . . , , D ALF(NL) D ALF(2) D ALF(2) D ALF(NL) OMITTING COLUMNS OF DERIVATIVES WHICH ARE ZERO, AND OMITTING PHI(L+1) COLUMNS IF PHI(L+1) IS NOT IN THE MODEL. NOTE THAT THE LINEAR PARAMETERS BETA ARE NOT USED IN THE MATRIX A. COLUMN L+P+2 IS RESERVED FOR WORKSPACE. THE CODING OF ADA SHOULD BE ARRANGED SO THAT: (WHICH OCCURS THE FIRST TIME ADA IS CALLED) MEANS: ISEL = 1STORE ANY CONSTANT PHI'S IN A. COMPUTE NONCONSTANT PHI'S AND PARTIAL DERIVA-Α. Β. C. TIVES = 2 MEANS COMPUTE ONLY THE NONCONSTANT FUNCTIONS PHI MEANS COMPUTE ONLY THE DERIVATIVES 3 = (WHEN THE PROBLEM IS LINEAR (NL ≈ Ø) ONLY ISEL = 1 IS USED, AND DERIVATIVES ARE NOT NEEDED.) RESTRICTIONS THE SUBROUTINES DPA, INIT (AND ADA) CONTAIN THE LOCALLY DIMENSIONED MATRIX INC, WHOSE DIMENSIONS ARE CURRENTLY SET FOR MAXIMA OF L+I = 8, NL = 12. THEY MUST BE CHANGED FOR LARGER PROBLEMS. DATA PLACED IN ARRAY A IS OVERWRITTEN ('DESTROYED'). DATA PLACED IN ARRAYS T, Y AND INC IS LEFT INTACT. THE PROGRAM THE PROGRAM RUNS IN UATFIV, EXCEPT WHEN L = O OR NL = O. IT IS ASSUMED THAT THE MATRIX PHI(J, ALF: T(I)) has full column rank. This means that the first L columns of the matrix a must be linearly independent. OPTIONAL NOTE: AS WILL BE NOTED FROM THE SAMPLE SUBPROGRAM

ADA, THE DERIVATIVES D PHI(J)/D ALF(K) (ISEL = 3) MUST BE

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COMPUTED INDEPENDENTLY OF THE FUNCTIONS PHI(J) (ISEL ± 2), SINCE THE FUNCTION VALUES ARE OVERWRITTEN AFTER ADA IS CALLED WITH ISEL ± 2 . THIS IS DONE TO MINIMIZE STORAGE, AT THE POS-SIBLE EXPENSE OF SOME RECOMPUTATION (SINCE THE FUNCTIONS AND DERIVATIVES FREQUENTLY HAVE SOME COMMON SUBEXPRESSIONS). TO REDUCE THE AMOUNT OF COMPUTATION AT THE EXPENSE OF SOME STORAGE, CREATE A MATRIX B OF DIMENSION NMAX BY L+I IN ADA, AND AFTER THE COMPUTATION OF THE PHI'S (ISEL ± 2), COPY THE VALUES INTO B. THESE VALUES CAN THEN BE USED TO CALCULATE THE DERIV-ATIVES (ISEL ± 3). (THIS MAKES USE OF THE FACT THAT WHEN A CALL TO ADA WITH ISEL ± 3 FOLLOWS A CALL WITH ISEL ± 2 , THE ALFS ARE THE SAME.)

TO CONVERT TO OTHER MACHINES, CHANGE THE OUTPUT UNIT IN THE DATA STATEMENTS IN VARPRO, DPA, POSTPR, AND VARERR. THE PROGRAM HAS BEEN CHECKED FOR PORTABILITY BY THE BELL LABS PFORT VERIFIER. FOR MACHINES WITHOUT DOUBLE PRECISION HARDWARE, IT MAY BE DESIRABLE TO CONVERT TO SINGLE PRECISION. THIS CAN BE DONE BY CHANGING (A) THE DECLARATIONS 'DOUBLE PRECISION' TO 'REAL'. (B) THE PATTERN '.D' TO '.E' IN THE 'DATA' STATEMENT IN VARPRO, (C) DSIGN, DSQRT AND DABS TO SIGN, SQRT AND ABS, RESPECTIVELY, AND (D) DEXP TO EXP IN THE SAMPLE PROGRAMS ONLY.

NOTE ON INTERPRETATION OF COVARIANCE MATRIX

FOR USE IN STATISTICAL ESTIMATION (MULTIPLE NONLINEAR REGRESSION) VARPRO RETURNS THE COVARIANCE MATRIX OF THE LINEAR AND NONLINEAR PARAMETERS. THIS MATRIX WILL BE USEFUL ONLY IF THE USUAL STATISTICAL ASSUMPTIONS HOLD: AFTER WEIGHTING, THE ERRORS IN THE OBSERVATIONS ARE INDEPENDENT AND NORMALLY DISTRI-BUTED, WITH MEAN ZERO AND THE SAME VARIANCE. IF THE ERRORS DO NOT HAVE MEAN ZERO (OR ARE UNKNOWN), THE PROGRAM WILL ISSUE A WARNING MESSAGE (UNLESS IPRINT .LT. Ø) AND THE COVARIANCE MATRIX WILL NOT BE VALID. IN THAT CASE, THE MODEL SHOULD BE ALTERED TO INCLUDE A CONSTANT TERM (SET PHI(1) = 1.).

NOTE ALSO THAT, IN ORDER FOR THE USUAL ASSUMPTIONS TO HOLD, THE OBSERVATIONS MUST ALL BE OF APPROXIMATELY THE SAME MAGNITUDE (IN THE ABSENCE OF INFORMATION ABOUT THE ERROR OF EACH OBSERVATION), OTHERWISE THE VARIANCES WILL NOT BE THE SAME. IF THE OBSERVATIONS ARE NOT THE SAME SIZE, THIS CAN BE CURED BY WEIGHTING.

IF THE USUAL ASSUMPTIONS HOLD, THE SQUARE ROOTS OF THE DIAGONALS OF THE COVARIANCE MATRIX A GIVE THE STANDARD ERROR S(I) OF EACH PARAMETER. DIVIDING A(I,J) BY S(I)*S(J) YIELDS THE CORRELATION MATRIX OF THE PARAMETERS. PRINCIPAL AXES AND CONFIDENCE ELLIPSOIDS CAN BE OBTAINED BY PERFORMING AN EIGEN-VALUE/EIGENVECTOR ANALYSIS ON A. ONE SHOULD CALL THE EISPACK PROGRAM TRED2, FOLLOWED BY TQL2 (OR USE THE EISPAC CONTROL PROGRAM).

CONVERGENCE FAILURES

IF CONVERGENCE FAILURES OCCUR, FIRST CHECK FOR INCORRECT CODING OF THE SUBROUTINE ADA. CHECK ESPECIALLY THE ACTION OF ISEL, AND THE COMPUTATION OF THE PARTIAL DERIVATIVES. IF THESE ARE CORRECT, TRY SEVERAL STARTING GUESSES FOR ALF. IF ADA IS CODED CORRECTLY, AND IF ERROR RETURNS IERR = -2 OR -8 PERSISTENTLY OCCUR, THIS IS A SIGN OF ILL-CONDITIONING, WHICH MAY BE CAUSED BY SEVERAL THINGS. ONE IS POOR SCALING OF THE PARAMETERS: ANOTHER IS AN UNFORTUNATE INITIAL GUESS FOR THE PARAMETERS, STILL ANOTHER IS A POOR CHOICE OF THE MODEL.

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ALGORITHM

THE RESIDUAL **R** IS MODIFIED TO INCORPORATE, FOR ANY FIXED ALF, THE OPTIMAL LINEAR PARAMETERS FOR THAT ALF. IT IS THEN POSSIBLE TO MINIMIZE ONLY ON THE NONLINEAR PARAMETERS. AFTER THE OPTIMAL VALUES OF THE NONLINEAR PARAMETERS HAVE BEEN DETER-MINED, THE LINEAR PARAMETERS CAN BE RECOVERED BY LINEAR LEAST SQUARES TECHNIQUES (SEE REF. 1).

THE MINIMIZATION IS BY A MODIFICATION OF OSBORNE'S (REF. MODIFICATION OF THE LEVENBERG-MARQUARDT ALGORITHM. INSTEAD 3) INSTEAD OF SOLVING THE NORMAL EQUATIONS WITH MATRIX

$$T$$
 2
(J J + NU * D), WHERE J = D(ETA)/D(ALF),

STABLE ORTHOGONAL (HOUSEHOLDER) REFLECTIONS ARE USED ON A MODIFICATION OF THE MATRIX

J) (----) (NU*D)

WHERE D IS A DIAGONAL MATRIX CONSISTING OF THE LENGTHS OF THE COLUMNS OF J. THIS MARQUARDT STABILIZATION ALLOWS THE ROUTINE TO RECOVER FROM SOME RANK DEFICIENCIES 11.1 THE JACOBIAN. OSBORNE'S EMPIRICAL STRATEGY FOR CHOOSING THE MARQUARDT PARAM-ETER HAS PROVEN REASONABLY SUCCESSFUL IN PRACTICE. (GAUSS-NEWTON WITH STEP CONTROL CAN BE OBTAINED BY MAKING THE CHANGE A DESCRIPTION CAN INDICATED BEFORE THE INSTRUCTION LABELED 5). BE FOUND IN REF. (3), AND A FLOW CHART IN (2), P. 22.

FOR REFERENCE, SEE

GENE H. GOLUR AND V. PEREYRA, 'THE DIFFERENTIATION OF PSEUDO-INVERSES AND NONLINEAR LEAST SQUARES PROBLEMS WHOSE' 1. VARIABLES SEPARATE, SIAM J. NUMER. ANAL. 10, 413-432 (1973).

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4. JECTION ALGORITHM FOR NONLINEAR LEAST SQUARES PROBLEMS,

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- 49-57 (1975). DRAPER, N., AND SMITH, H., APPLIED REGRESSION ANALYSIS, WILEY, N.Y., 1966 (FOR STATISTICAL INFORMATION ONLY). C. LAWSON AND R. HANSON, SOLVING LEAST SQUARES PROBLEMS, PRENTICE-HALL, ENGLEWOOD CLIFFS, N. J., 1974. 6. 7.

JOHN BOLSTAD COMPUTER SCIENCE DEPT., SERRA HOUSE STANFORD UNIVERSITY **JANUARY.** 1977

DOUBLE PRECISION A(NMAX, LPPZ). BETA(L), ALF(NL), TINMAX, IV 2 W(N), Y(N), ACUM, EPSI, GNSTEP, NU, PRJRES, R, RNEW, XNORM INTEGER B1, OUTPUT IV), LOGICAL SKIP EXTERNAL ADA

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DATA EPSI /1.D-6/, ITMAX /48/, OUTPUT /6/ 0000000 THE FOLLOWING TWO PARAMETERS ARE USED IN THE CONVERGENCE TEST: EPSI IS AN ABSOLUTE AND RELATIVE TOLERANCE FOR THE NORM OF THE PROJECTION OF THE RESIDUAL ONTO THE RANGE OF THE JACOBIAN OF THE VARIABLE PROJECTION FUNCTIONAL ITMAX IS THE MAXIMUM NUMBER OF FUNCTION AND DERIVATIVE EVALUATIONS ALLOWED. CAUTION: EPSI MUST NOT BE Ĉ SET SMALLER THAN 10 TIMES THE UNIT ROUND-OFF OF THE MACHINE. č-005 CALL LIB MONITOR FROM VARPRO, MAINTENANCE NUMBER 509, DATE 77178, C***PLEASE DON'T REMOVE OR CHANGE THE ABOVE CALL. IT IS YOUR ONLY 006 C***PLEASE DON'T REMOVE OR CHANGE THE ABOVE CALL. IT IS YOUR C***PROTECTION AGAINST YOUR USING AN OUT-OF-DATE OR INCORRECT 088 009 C***VERSION OF THE ROUTINE. THE LIBRARY MONITOR REMOVES THIS CALL, C***SO IT ONLY OCCURS ONCE, ON THE FIRST ENTRY TO THIS ROUTINE. 01 011 012 IERR = 1ITER = OLP1 = L +1 **B**1 = L + 2 LNLZ = L + NL + 2 NLPI = NL + 1SKIP = .FALSE MODIT = IPRINT IF (IPRINT .LE. Ø) MODIT = ITMAX + 2 NU \doteq O. С IF GAUSS-NEWTON IS DESIRED REMOVE THE NEXT STATEMENT. NU = 1. 000000 BEGIN OUTER ITERATION LOOP TO UPDATE ALF CALCULATE THE NORH OF THE RESIDUAL AND THE DERIVATIVE OF THE MODIFIED RESIDUAL THE FIRST TIME, BUT ONLY THE DERIVATIVE IN SUBSEQUENT ITERATIONS. 5 CALL DPA (L, NL, N, NMAX, LPP2, IV, T, Y, W, ALF, ADA, IERR, X IPRINT. A, BETA, A(1, LP1), R) IPRINT. A, GNSTEP = 1.0|TERIN = 0IF (ITER .GT. Ø) GO TO 10 IF (NL .EO. Ø) GO TO 90 IF (IERR .NE. 1) GO TO 99 С IF (IPRINT LE. Ø) GO TO 10 WRITE (OUTPUT, 2077) ITERIN, R WRITE (OUTPUT, 2007) NU BEGIN TWO-STAGE ORTHOGONAL FACTORIZATION С 10 CALL ORFACI(NLP1, NMAX. N IF (IERR .LT. Ø) GO TO 99 N, L, IPRINT, A(1, B1), PRJRES, IERR) IERR = 2IF (NU .EO. Ø.) GO TO 30 C C C C C BEGIN INNER ITERATION LOOP FOR GENERATING NEW ALF AND TESTING IT FOR ACCEPTANCE. 25 CALL ORFAC2(NLP1, NMAX, NU, A(1, B1)) 00000 SOLVE A NL X NL UPPER TRIANGULAR SYSTEM FOR DELTA-ALF. THE TRANSFORMED RESIDUAL (IN COL. LNLZ OF A) IS OVER-WRITTEN BY THE RESULT DELTA-ALF. 30 CALL BACSUB (NMAX, NL, A(1, B1), A(1, LNL2)) DO 35 K = 1, NL A(K, B1) = ALF(K) + A(K, LNL2) 35

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NEW ALF(K) = ALF(K) + DELTA ALF(K)CCCCC STEP TO THE NEW POINT NEW ALF, AND COMPUTE THE NEW NORM OF RESIDUAL. NEW ALF IS STORED IN COLUMN B1 OF A. CALL DPA (L, NL, N, NMAX. LPP2, IV, T, Y, W, A(1, B1), ADA, IERR, IPRINT, A, BETA, AII, LP1), RNEW) IF (IERR .NE. 2) GO TO 99 40 Х ITER = ITER + 1|TERIN = |TERIN + 1|SKIP = MOD(ITER, MODIT) .NE. O IF (SKIP) GO TO 45 WRITE (OUTPUT, 203) ITER WRITE (OUTPUT, 216) (A(K, B1), K = 1, NL) WRITE (OUTPUT, 207) ITERIN, RNEW С 45 IF (ITER .LT. ITMAX) GO TO 50 ÌERR = −1 CALL VARERR (IPRINT, IERR, 1) GO TO 95 IF (RNEW - R .LT. EPS1*(R + 1.DØ)) GO TO 75 50 C C C RETRACT THE STEP JUST TAKEN IF (NU .NE. Ø.) GO TO 60 С GAUSS-NEWTON OPTION ONLY GNSTEP = $\emptyset.5 \times GNSTEP$ IF (GNSTEP .LT. EPS1) GO TO 95 DO 55 K = 1, NL A(K, B1)' = ALF(K) + GNSTEP*A(K, LNL2)55 GO TO 40 С ENLARGE THE MARQUARDT PARAMETER NU = 1.5 * NU60 IF (NOT. SKIP) WRITE (OUTPUT. 206) NU (NU .LE. 180.) GO TO 65 IERR = -2 1 F CALL VARERR (IPRINT, IERR, 1) GO TO 95 С RETRIEVE UPPER TRIANGULAR FORM С AND RESIDUAL OF FIRST STAGE. 65 DO 70 K = 1, NL KSUB = LP1 + KDO 70 J = K, NLPI JSUB = LP1 + JISUB = NLPI + J A(K, JSUB) = A(ISUB, KSUB)70 GO TO 25 С END OF INNER ITERATION LOOP ACCEPT THE STEP JUST TAKEN C C 75 R = RNEWDO 80 K = 1, NL 8Ø ALF(K) = A(K, B1)С CALC. NORM(DELTA ALF)/NORM(ALF) ACUM = GNSTEP*XNORM(NL, A(1, LNL2))/XNORM(NL, ALF) С C C IF ITERIN IS GREATER THAN 1, A STEP WAS RETRACTED DURING THIS OUTER ITERATION. Č (ITERIN .EQ. 1) NU = Ø.5*NU (SKIP) GO TO 85 WRITE (OUTPUT. 2ØØ) NU WRITE (OUTPUT, 2Ø8) ACUM 1 F 1 F 85 IERR = 3

IF (PRJRES .GT. EPS1*(R + 1.DØ)) GO TO 5 END OF OUTER ITERATION LOOP 00000 CALCULATE FINAL QUANTITIES -- LINEAR PARAMETERS, RESIDUALS, COVARIANCE MATRIX, ETC. 90 IERR = ITER 95 IF (NL .GT. Ø) CALL DPA(L, NL, N, NMAX, LPPZ, IV. T. Y, W, ALF, X ADA, 4, IPRINT, A, BETA, A(1, LP1), R) CALL POSTPR(L, NL, N, NMAX, LNL2, EPS1, R, IPRINT, ALF, W, A, X A(1, LP1), BETA, IERR) С 200 FORMAT (9H NU =, E15.7) 203 FORMAT (12HØ ITERATION, I4, 24H NONLINEAR PARAMET 206 FORMAT (25H STEP RETRACTED, NU = E15.7) 207 FORMAT (1HØ, I5, 2ØH NORM OF RESIDUAL =, E15.7) 208 FORMAT (34H NORM (DELTA-ALF) / NORM(ALF) =, E12.3) 209 FORMAT (440 TELECT) NONLINEAR PARAMETERS) 216 FORMAT (1HØ, 7E15.7) END С SUBROUTINE ORFACI(NLP1, NMAX, N, L, IPRINT, B, PRJRES, IERR) С STAGE 1: HOUSEHOLDER REDUCTION OF (DR'. R3) NL DR R 2 TO) 0 R4) N-L-NL (NL 1 NL 1 WHERE DP, = -D(Q2)*Y IS THE DERIVATIVE OF THE MODIFIED RESIDUAL PRODUCED BY DPA, R2 IS THE TRANSFORMED RESIDUAL FROM DPA, AND' DR' IS IN UPPER TRIANGULAR FORM (AS IN REF. (2), P. 18). DR IS STORED IN ROWS L+1 TO N AND COLUMNS L+2 TO L + NL + 1 OF THE MATRIX A (I.E., COLUMNS 1 TO NL OF THE MATRIX B). R2 IS STORED IN COLUMN L + NL + 2 OF THE MATRIX A (COLUMN NL + 1 OF B). FOP, K = 1, 2, ..., NL. FIND REFLECTION I - U UI / BETA WHICH ZEROES B(I, K), I = L+K+1, ..., N. DOUBLE PRECISION ACUM, ALPHA, B(NMAX, NLP1), BETA, DSIGN, PRJRES, X U, XNORM С NL ≖ NLPI -1 $NL23 = 2 \times NL + 3$ LP1 = L + 1 С DO 30 K = 1. NL LPK = L + KALPHA = DSIGN(XNORM(N+1-LPK, B(LPK, K)), B(LPK, K)) U = B(LPK, K) + ALPHA B(LPK, K) = UBETA = ALPHA UIF (ALPHA .NE. O.O) GO TO 13 С COLUMN WAS ZERO IERR ≠ -8 CALL VARERR (IPRINT, IERR, LP1 + K) GO TO 99 APPLY REFLECTIONS TO REMAINING COLUMNS OF **B** AND TO RESIDUAL VECTOR. C C 13 KP1 = K + 1DO 25 J = KP1, NLPI

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ACUM = 0.0 DO 20 I = LPK, N ACUM = ACUM + B(I, K) * B(I, J) 20 ACUM = ACUM / BETA DO 25 I = LPK, N $\vec{B}(I, J) = \vec{B}(I, J) = B(I, K) * ACUM B(LPK, K) = -ALPHA$ 25 30 С PRJRES = XNORM(NL, B(LP1, NLP1))C C C C C SAVE UPPER TRIANGULAR FORM AND TRANSFORMED RESIDUAL, FOR USE IN CASE A STEP IS RETRACTED. ALSO COMPUTE COLUMN LENGTHS. IF (IERR .EQ. 4) GO TO 99 DO **50** K = 1, NL LPK = L + KDO 40 **J** = K, NLPI JSUB = NLPI + J B(K, J) = B(LPK, J) B(JSUB, K) = B(LPK, J) B(NL23, K) = XNORM(K, B(LP1, K))40 5 58 С 99 RETURN END С SUBROUTINE ORFAC2(NLP1, NMAX, NU, B) STAGE 2: SPECIAL HOUSEHOLDER REDUCTION OF (DR'' (DR' _ R3) NL R5) ~ ~ ~ ~ ~) . R4 ø R 4 TO N-L-NL (.) 0) ____ -) (NU*D . O 0 R 6 NL) (.) NL 1 NL 1 WHERE DR', R3, AND R4 ARE AS IN ORFAC1, NU IS THE MARQUARDT PARAMETER, D IS A DIAGONAL MATRIX CONSISTING OF THE LENGTHS OF THE COLUMNS OF DR', AND DR'' IS IN UPPER TRIANGULAR FORM. DETAILS IN (1), PP. 423-424. NOTE THAT THE (N-L-NL) BAND OF ZEROES, AND R4, ARE OMITTED IN STORAGE. DOUBLE PRECISION ACUM. ALPHA, B(NMAX, NLP1), BETA, DSIGN, NU, U, X XNORM С NL 🖛 NLPI = 1 NL2 = 2*NLNL23 = NL2 + 3DO 30 K = 1, NL KP1 = K + 1NLPK = NL + K NLPKM1 = NLPK - 1U = B(K, K) + ALPHABETA = ALPHA * U B(K, K) = -ALPHAС С THE K-TH REFLECTION MODIFIES ONLY ROWS K, NL+1, NL+2, ..., NL+K, AND COLUMNS K TO NL+1. DO 30 J = KP1, NLPI

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B(NLPK, J) = Ø. ACUM = U * B(K,J) DO 20 I = NLP1, NLPKM1 ACUM = ACUM + B(I,K) * B(I,J) 20 ACUM = ACUM / BETAACUM = ACUM / BETA B(K,J) = B(K,J) - U + ACUM DO 30 = NLP1, NLPK B(I,J) = B(I,J) - B(I,K) + ACUM30 С RETURN END С SUBROUTINE DPA (L, NL, N, NMAX, LPP2, IV, T, Y, W, ALF, ADA, ISEL, X IPRINT. A, U, R, RNORM) COMPUTE THE NORM OF THE RESIDUAL (IF ISEL = 1 OR 2), OR THE (N-L) X NL DERIVATIVE OF THE MODIFIED RESIDUAL (N-L) VECTOR Q2*Y (IF ISEL = 1 OR 3). HERE Q * PHI = S, I.E., (Q1) ((----) ((Q2) (PHI Y D(PHI)) = (S . R1 . F1 (Ø . R2 . F2) L) N-L) Ρ Ν L Ρ 1 1 L WHERE Q IS N X N ORTHOGONAL, AND S IS L X L UPPER TRIANGULAR. THE NORM OF THE RESIDUAL = NORM(R2), AND THE DESIRED DERIVATIVE ACCORDING TO REF. (5), IS $D(Q2 * Y) = -Q2 * D(PHI) * s^{-1} * Q1 * Y.$ DOUBLE PRECISION A(NMAX, LPP2), ALF(NL), T(NMAX, IV), W(N), Y(N), X ACUM, ALPHA, BETA, RNORM, DSIGN, DSQRT, SAVE, R(N), U(L), XNORM INTEGER FIRSTC, FIRSTR, INC(12, 8) LOGICAL NOWATE, PHILPI EXTERNAL ADA С IF (ISEL .NE. 1) GO TO 3 I.P1 = L + 1 LNLZ = L + 2 + NL LP2 = L + 2 LPPI = LPPZ = 1 FIRSTC = 1 LASTC = LPPI FIRSTR = LP1 CALL INIT(L, NL, N, NMAX, LPP2, IV, T, W, ALF, ADA, ISEL. IPRINT, A, INC, NCON, NCONPI, PHILPI, NOWATE) IF (ISEL .NE. 1) GO TO 99 GO TO 30 Х С 3 CALL ADA (LP1, NL, N, NMAX, LPP2, IV, A, INC, T, ALF, MINØ(ISEL, X 3)) IF (ISEL .EQ. 2) GO TO 6 С ISEL = 3 OR 4FIRSTC = LP2LASTC = LPPI FIRSTR = (4 = ISEL)*L + 1 GO TO 50 С ISEL = 26 FIRSTC = NCONPI LASTC = LP1 IF (NCON .EQ. Ø) GO TO 30

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IF (A(1, NCON) .EO. SAVE) GO TO 30 ISEL = −7 CALL VARERR (IPRINT, ISEL, NCON) GO TO 99 С ISEL = 1 OR 2 30 IF (PHILP1) GO TO 48 DO 35 I = 1, N 35 R(I) = Y(I)GO TO 50 DO 45 ■ = 1, N R(I) = Y(I) - R(I) 40 45 WEIGHT APPROPRIATE COLUMNS С 50 IF [NOWATE) GO TO 58 DO 55 ■ = 1, N ACUM = W(I) DO 55 J = FIRSTC, LASTC A(I, J) = A(I, J) * ACUM 55 С 000000000000 COMPUTE ORTHOGONAL FACTORIZATIONS BY HOUSEHOLDER REFLECTIONS. IF ISEL = 1 OR 2, REDUCE PHI (STORED IN THE FIRST L COLUMNS OF THE MATRIX \mathbf{A}) TO UPPER TRIANGULAR FORM, (Q*PHI = S), AND TRANSFORM Y (STORED IN COLUMN L+1), GETTING Q*Y = R. IF ISEL = 1. ALSO TRANSFORM J = D PHI (STORED IN COLUMNS L+2 THROUGH L+P+1 OF THE MATRIX A), GETTING Q*J = #. IF ISEL = 3 OR 4. PHI HAS ALREADY BEEN REDUCED, TRANSFORM ONLY J. S, R, AND F OVERWRITE PHI, Y. AND J, RESPECTIVELY, AND A FACTORED FORM OF Q IS SAVED IN U AND THE LOWER TRIANGLE OF PHI. 58 IF (L.EO. Ø) GO TO 75 DO 70 K = 1. L KP1 = K + 1IF (ISEL .GE. 3 .OR. (ISEL .EO. 2 .AND. E .LT.NCONP1)) GO TO 66 ALPHA = DSIGN(XNORM(N+1-K, A(K, K)), A(K, K)) U(K) = A(K, K) + ALPHAA(K, K) = -ALPHAA(K, K) = -AFIRSTC = KPI IF (ALPHA .NE. O.O) GO TO 66 ISEL = -8CALL VARERR (IPRINT, ISEL, K) GO TO **99** C C APPLY REFLECTIONS TO COLUMNS FIRSTC TO LASTC. BETA = -A(K, K) * U(K)DO 70 J = FIRSTC, LASTC 66 ACUM = U(K) * A(K, J)DO 68 $\blacksquare = KP1$, N ACUM = ACUM + A(I, K)*A(I, J) 68 ACUM = ACUM / BETA A(K,J) = A(K,J) - U(K)*ACUM DO 70 \blacksquare = KP1, N A(I, J) = A(I, J) - A(I, K) * ACUM70 С 75 IF (ISEL .GE. 3) GO TO 85 RNORM = XNORM(N-L, R(LP1)) IF (ISEL .EQ. 2) GO TO 99 IF (NCON .GT. Ø) SAVE = A(1, NCON) CCCCCC F2 IS NOW CONTAINED IN ROWS L+I TO N AND COLUMNS L+2 TO L+P+1 OF THE MATRIX A. NOW SOLVE THE L X L UPPER TRIANGULAR SYSTEM S*BETA = R1 FOR THE LINEAR PARAMETERS BETA. BETA OVERWRITES R1. 85 IF (L.GT. Ø) CALL BACSUB (NMAX. L, A, R)

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С
Ċ
               MAJOR PART OF KAUFMAN'S SIMPLIFICATION OCCURS HERE.
                                                                                     COMPUTE
THE DERIVATIVE OF ETA WITH RESPECT TO THE NONLINEAR
               PARAMETERS
                           Т*
                                                D PHI(J)
                                                               D PHI(L+1)
          D TA
                                  1
                               (SUM BÉTA(J)
    0 *
                          Q
                                                                         F2*BETA
                      =
                                                                                .
                                                D ALF(K)
         D ALF(K)
                                                                 D ALF(K)
                                J=1
               AND STORE THE RESULT IN COLUMNS L+2 TO L+NL+1.
                                                                              IF ISEL NOT
               = 4, THE FIRST L ROWS ARE OMITTED. THIS IS -D(Q2)*Y, IF
ISEL NOT = 4 THE RESIDUAL R2 = Q2*Y (IN COL. L+1) IS COPIED
TO COLUMN L+NL+2. OTHERWISE ALL OF COLUMN L+1 IS COPIED.
        DO 95 I = FIRSTR, N
IF (L.EO. NCON) GO TO 95
            M = LP1
            DO 90 K = 1, NL
                ACUM \Rightarrow O.
                DO 88 J = NCONP1, L
IF (INC(K, J) .EQ. Ø) GO TO 88
                    M = M + 1
                    ACUM = ACUM + A(I, M) * R(J)
    88
                    CONTINUE
                KSUB = LP1 + K
                IF (INC(K, LP1) .EQ. Ø) GO TO 90
                M = M +
                          1
            ACUM = ACUM + A(I, M)

A(I, KSUB) = ACUM

A(I, LNLZ) = R(I)
    90
    95
С
    99 RETURN
        END
С
      SUBROUTINE INIT(L, NL, N, NMAX, LPP2, IV, T, W, ALF. ADA, ISEL, X IPRINT, A, INC, NCON, NCONP1, PHILPI, NOWATE)
000000
            CHECK VALIDITY OF INPUT PARAMETERS, AND DETERMINE NUMBER OF
            CONSTANT FUNCTIONS.
        DOUBLE PRECISION A(NMAX, LPP2), ALF(NL), T(NMAX, IV), W(N),
       X DSQRT
        INTEGER OUTPUT, P, INC(12, 8)
        LOGICAL NOWATE,
                             PHILPI
        DATA OUTPUT /6/
C
        LP1 ≈ L + 1
LNL2 ≈ L + 2 + NL
С
                                                        CHECK FOR VALID INPUT
       IF (L.GE. Ø AND. NL.GE. Ø AND. L+NL.LT. N AND. LNL2 .LE.
X LPP2 AND. 2*NL + 3 .LE. NMAX AND. N .LE. NMAX AND.
X IV.GT. O AND. .NOT. (NL .EQ. O AND. L .EQ. Ø)) GO TO 1
ISEL = -4
            CALL VARERR (IPRINT, ISEL, 1)
            GO TO 99
С
      1 IF (L .EQ. Ø .OR. NL
DO 2 J = 1. LP1
DO 2 K = 1, NL
                               NL .EQ. Ø) GO TO 3
      2
                    INC(K, J) = \emptyset
С
      3 CALL ADA (LP1, NL, N, NMAX, LPP2, IV, A, INC, T, ALF, ISEL)
```

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```
C
         NOWATE = .TRUE.
        DO 9 I = 1, N
NOWATE = NOWATE .AND. (W(I) .EQ. 1.Ø)
IF (W(I) .GE. Ø.) GO TO 9
С
                                                                       ERROR IN WEIGHTS
             ISEL = -6
             CALL VARERR (IPRINT, ISEL, I)
             GO TO 99
     9
             W(I) = DSQRT(W(I))
С
        NCOH = L
        NCONP1 = LP1
        PHILPI = L .EQ. O
IF (PHILP1 .OR. NL .EQ. Ø) GO TO 99
С
                                                    CHECK INC MATRIX FOR VALID INPUT AND
C
                                                    DETERMINE NUMBER OF CONSTANT FCNS.
        P = \emptyset
        P = \emptyset

DO 11 J = 1, LP1

IF (P.EQ. Ø) NCONPI = J

DO 11 K = 1, NL

INCKJ = INC(K, J)

IF (INCKJ .NE. O .AND. INCKJ .NE. 1) GO TO 15

IF (INCKJ .EQ. 1) P = P + 1

CONTINUE
                 CONTINUE
     11
С
        NCON = NCONPI - 1
IF (IPRINT .GE. Ø) WRITE (OUTPUT, 21Ø) NCON
IF (L+P+2 .EQ. LPP2) GO TO 20
С
                                                                    INPUT ERROR IN INC MATRIX
     15 ISEL = -5
        CALL VARERR (IPRINT. ISEL. 1)
         GO TO 99
С
                                                 DETERMINE IF PHI(L+1) IS IN THE MODEL.
    20 DO 25 K = 1, NL
25 IF (INC(K, LP1) .EQ. 1) PHILPI = .TRUE.
С
    99 RETURN
   21Ø FORMAT (33HØ NUMBER OF CONSTANT FUNCTIONS =, 14 /)
         FND
         SUBROUTINE BACSUB (NMAX, N, A, X)
С
             BACKSOLVE THE N X N UPPER TRIANGULAR SYSTEM A*X = B. THE SOLUTION X OVERWRITES THE RIGHT SIDE B.
C
C
C
         DOUBLE PRECISION A(NMAX, N), X(N), ACUM
С
        X(N) = X(N) / A(N, N)
IF IN .EQ. 1) GO TO 30
        NP1 = N + 1

DO 20 IBACK = 2, N

I = NP1 - IBACK
                 I = N-1, N-2, ..., 2, 1
I = I + 1
С
             IP1 = I +
             ACUM = X(I)
             ACUM = IP1, N
ACUM = ACUM - A(I,J)*X(J)
X(I) = ACUM / A(I,I)
     10
    20
С
     30 RETURN
         END
         SUBROUTINE POSTPR(L, NL, N, NMAX, LNL2, EPS, RNORM, IPRINT, ALF,
       XW, A, R, U. IERR)
С
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CALCULATE RESIDUALS, SAMPLE VARIANCE, AND COVARIANCE MATRIX. ON INPUT, U CONTAINS INFORMATION ABOUT HOUSEH**OLDER REFLE**CTIONS 0000 ON OUTPUT, IT CONTAINS THE LINEAR PARAMETERS. FROM DPA. DOUBLE PRECISION A(NMAX, LNL2), ALF(NL), R(N), U(L), W(N), ACUM, X EPS, PRJRES, RNORM, SAVE, DABS INTEGER OUTPUT DATA OUTPUT /6/ С LP1 = L + 1 LPNL = LNLZ = 2 LNL1 = LPNL + 1DO 10 I = 1, N W(I) = W(I)**2 10 0000 UNWIND HOUSEHOLDER TRANSFORMATIONS TO GET RESIDUALS, AND MOVE THE LINEAR PARAMETERS FROM R TO U. IF (L .EQ. Ø) GO TO 38 DO 25 KBACK = 1, L K = LP1 - KBACK KP1 = K + 1ACUM = \mathbf{O} . DO 20 I = KP1, N ACUM = ACUM + A(I, K) = R(I) 2Ø SAVE = R(K) $\begin{array}{rcl} R(K) &= & ACUM \ / \ A(K, \ K) \\ ACUM &= & -ACUM \ / \ (U(K) \ \ast \ A(K, \ K)) \end{array}$ U(K) = SAVEDO 25 I = KP1, N R(I) = R(I) - A(I, K)*ACUM 25 COMPUTE MEAN ERROR С **30** ACUM **= 0.** DO 35 I **=** 1, N 35 ACUM **=** ACUM + R(I) SAVE = ACUM / N 000000 THE FIRST L COLUMNS OF THE MATRIX HAVE BEEN REDUCED TO UPPER TRIANGULAR FORM IN DPA. FINISH BY REDUCING ROWS L+I TO N AND COLUMNS L+2 THROUGH L+NL+1 TO TRIANGULAR FORM. THEN SHIFT COLUMNS OF DERIVATIVE MATRIX OVER ONE TO THE LEFT TO BE ADJACENT TO THE FIRST L COLUMNS. С IF (NL .EQ. Ø) GO TO 45 CALL_ORFAC1(NL+1, NMAX, N, L, IPRINT, A(1, L+2), PRJRES, 4) DO 40 I = 1, N A(I, LNL2) = R(I) DO 40 K = LP1, LNL1 A(I, K) = A(I, K+1)40 С COMPUTE COVARIANCE MATRIX 45 A(1, LNL2) = RNORM ACUM = RNORM*RNORM/(N - L - NL) A(2, LNL2) = ACUM CALL COV(NMAX, LPNL, ACUM, A) С IF (IPRINT .LT. Ø) GO TO 99
WRITE (OUTPUT, 2Ø9)
IF (L .GT. Ø) WRITE (OUTPUT. 210) (U(J), J = 1, L)
IF (NL .GT. Ø) WRITE (OUTPUT, 211) (ALF(K), K = 1, NL)
WRITE (OUTPUT, 214) RNORM, SAVE, ACUM
IF (DABS(SAVE) .GT. EPS) WRITE (OUTPUT, 215)
WRITE (OUTPUT, 2Ø9)
99 RETURN С

209 FORMAT (1HØ, 5Ø(1H')) 210 FORMAT (2ØHØ LINEAR LINEAR PARAMETERS // (7E15.7)) NONLINEAR PARAMETERS // (7E15.7)) 211 FORMAT (23HØ NORM OF RESIDUAL =, E15.7, 33H EXPECTED ERROR OF OS E15.7, / 39H ESTIMATED VARIANCE OF OBSERVATIONS =, 214 FORMAT (21HØ 33H EXPECTED ERROR OF OSS XERVATIONS ■, E15.7, X E15.7) 215 FORMAT (95H RMAT (95H WARNING -- EXPECTED ERROR OF OBSERVATIONS IS NOT ZERO COVARIANCE MATRIX MAY BE MEANINGLESS. /) X. U SUBROUTINE COV(NMAX, N, SIGMA2, A) COMPUTE THE SCALED COVARIANCE MATRIX OF THE L + NL RAMETERS. THIS INVOLVES COMPUTING PARAMETERS. 2 -1 -T SIGMA * T * T WHERE THE (L+NL) X (L+NL) UPPER TRIANGULAR MATRIX T IS DESCRIBED IN SUBROUTINE POSTPR. THE RESULT OVERWRITES THE FIRST L+NL ROWS AND COLUMNS OF THE MATRIX A. THE RESULTIN MATRIX IS SYMMETRIC. SEE REF. 7, PP. 67-70, 281. THE RESULTING DOUBLE PRECISION A (NMAX, N), SUM, SIGMA2 С DO 10 J = 1, N A(J, J) = 1./A(J, J)10 С С С INVERT T UPON ITSELF IF (N .EQ. 1) GO TO 70 NM1 = N = 1 DO 60 I = 1. NMI IP1 = I + 1DO 60 J = IP1, NJM1 = J -1 SUM = O.DO 50 M = I, JM1 SUM = SUM + A(I, M) * A(M, J) A(I, J) = -SUM * A(J, J) 50 ĞØ C C C NOW FORM THE MATRIX PRODUCT 70 DO 90 I = 1. N DO 90 J ≖ I,N SUM = 0.DO 80 M = J, N SUM = SUM + A(I, M) * A(J, M) 80 SUM = SUM * SIGMA2 A(I, J) = SUM A(J, I) = SUM 90 С RETURN FND SUBROUTINE VARERR (IPRINT, IERR, K) С C PRINT ERROR MESSAGES INTEGER ERRNO, OUTPUT DATA OUTPUT /6/ С IF (IPRINT .LT. Ø) GO TO 99 ERRNO = IABS(IERR)

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GO TO (1, 2, 99, 4, 5, 6, 7, 8), ERRNO С 1 WRITE (OUTPUT, 101) GO TO 99 2 WRITE (OUTPUT, 102) GO TO 99 4 WRITE (OUTPUT, 1Ø4) GO TO 99 WRITE (OUTPUT, 105) 5 GO TO 99 (OUTPUT, 106) K 6 WRITE GO TO 99 7 WRITE (OUTPUT, 107) K GO TO 99 8 WRITE (OUTPUT, 108) K С 99 RETURN 99 RETURN 101 FORMAT (46HØ PROBLEM TERMINATED FOR EXCESSIVE IFERATIONS /// 102 FORMAT (49HØ PROBLEM TERMINATED BECAUSE OF ILL-CONDITIONING //) 104 FORHAT (/ 5ØH INPUT ERROR IN PARAMETER L. NL, N. LPPZ, OR NMAX. /) 105 FORMAT (68HØ ERROR -- INC MATRIX IMPROPERLY SPECIFIED, OR DISAGRE ÉRROR -- WEIGHT(, I4, 14H) IS NEGATIVE. /) ERROR -- CONSTANT COLUMN , I3, 37H MUST BE COMPUTED L = 1. /) 106 FORMAT (19HØ 107 FORHAT (28HØ XONLY WHEN ISEL = 1. /) 108 FORHAT (33HØ CATASTROPHIC FAILURE -- COLUMN , I4, 28H IS ZERO, SE XE DOCUMENTATION. /) **FND** DOUBLE PRECISION FUNCTION XNORM(N, X) 00000 COMPUTE THE L2 (EUCLIDEAN) NORM OF A VECTOR, MAKING SURE TO AVOID UNNECESSARY UNDEKFLÓWS. NO ATTEMPT IS MADE TO SUPPRESS OVERFLOWS. DOUBLE PRECISION X(N), RMAX. SUM, TERM, DABS, DSQRT С FIND LARGEST (IN ABSOLUTE VALUE) ELEMENT С RMAX = O. DO 10 = 1 N IF (DABS(X(I)) .GT. RMAX) RMAX = DABS(X(I)) 10 CONTINUE С SUM = \emptyset . IF IRMAX .EQ. Ø.) GO TO 30 DO 20 I = 1. N TERM = O.IF (RMAX + DABS(X(I)) .NE. RMAX) TERM = X(I)/RMAX SUM = SUM + TERM*TERM 20 С 30 XNORM = RMAX*DSQRT(SUM) 99 RETURN END

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4.00	NF U I			
1 4.20				1
0 32				
1 0.	1.000			
0.050	1.000			
0.250	1.000			
1.080 1.250	1.000			
1.500 1.750	1.000 1.000			
2.000	0.961 0.795			
2.750 3.000 3.083	0.690 0.567 0.462			
3.250	0.349			
4.000 4.250	0.227 0.187			
4.500 4.750	0.151 0.128			
5.000	0.122			
5.750	0.093			
6.250	0.075 0.087			
7.800 8.583	0.046 0.028			
10.633 20.700	0.025 0.020			

0 INJECTION TIME 4.0000808000088 NUMBER OF NONLINEAR PARAMETERS 0 1 0 INITIAL ESTIMATES OF NONLIN. PARAM. 4.200 0 NUMBER OF LINEAR PARAMETERS 0 0 NUMBER OF OBSERVATIONS 32 0 NUMBER OF INDEPENDENT VARIABLES 0 **1**

INDEPENDENT VARIABLES

ØTIME VS MEASURED CONCENTRATIONS

O.	1.000
0.050	1.000
8.883	1.000
0.258	1.000
8.500	1.000
0.750	1.000
1.000	1.000
1.258	1.000
1.750 2.800 2.250 2.750	1.000 1.000 6.961 8.795 8.690
3.000	8.567
3.883	8.462
3.250	8.349
3.667	8.292
4.000	8.227
4.250	8.187
4.500	6.151
4.750	8.128
5.000	0.122
5.250	8.129
5.500	8.093
5.750	0.093
6.000	8.063
6.250	8.075
6.500	8.087
7.800	8.046
8.583	8.028
10.633	8.025
20.700	8.020
ØINCIDENCE	MATRIX INC(1,1)=

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O NORM OF RESIDUAL = ∅.9∅59386e+∅∅ ο NU = $\emptyset.1\emptyset\emptyset\emptyset\emptyset\emptyset\emptyset\emptysete+\emptyset1$ ITERATION NONLINEAR PARAMETERS ο 1 0 Ø.2916745e+Ø1 1 NORM OF RESIDUAL = Ø.8017488e+00 NU = Ø.5000000e+00 Ο NORM(DELTA-ALF) / NORM(ALF) = Ø.44Øe+ØØ NONLINEAR PARAMETERS ITERATION 0 2 Ø.24746Ø9e+Ø1 0 1 NORM OF RESIDUAL = Ø.7871255e+ØØ NU = Ø.25ØØØØØe+ØØ Ο NORM(DELTA-ALF) / NORM(ALF) = Ø.179e+ØØ NONLINEAR PARAMETERS 0 ITERATION 3 Ø.245411Øe+Ø1 0 1 NORM OF RESIDUAL ≖ Ø.787Ø9Ø9e+ØØ NU = Ø.125ØØØØe+ØØ O NORM(DELTA-ALF) / NORM(ALF) = Ø.835e-Ø2 NONLINEAR PARAMETERS ITERATION 0 4 Ø.245437Øe+Ø1 0 1 NORM OF RESIDUAL = Ø.787Ø9Ø9e+ØØ NU = Ø.625ØØØØe-Ø1 O NORM(DELTA-ALF) / NORM(ALF) = Ø.1Ø6e-Ø3 NONLINEAR PARAMETERS ITERATION 0 5 Ø.2454356e+Ø1 0 1 NORM OF RESIDUAL = Ø.787Ø9Ø9e+ØØ NU = Ø.3125ØØØe-Ø1 0 NORM(DELTA-ALF) / NORM(ALF) = Ø.568e~Ø5 ο ITERATION 6 NONLINEAR PARAMETERS Ø.2454357e+Ø1 0 1 NORM OF RESIDUAL = Ø.787Ø9Ø9e+ØØ NU = Ø.15625ØØe-Ø1 O NORM(DELTA-ALF) / NORM(ALF) = Ø.373e-Ø6 **O** NONLINEAR PARAMETERS Ø.2454357e+Ø1 NORM OF RESIDUAL = Ø.787Ø9Ø9e+ØØ EXPECTED ERROR OF OBSERVATIONS = -Ø.7967799e-Ø1 ESTIMATED VARIANCE OF OBSERVATIONS = Ø.1998425e-Ø1 WARNING -- EXPECTED ERROR OF OBSERVATIONS IS NOT ZERO. COVARIANCE MATRIX MAY BE M ø COVARIANCE MATRIX MAY BE MEANINGLES:

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ØTIME VS CALCULATED CONCENTRATION

O. 0.050 0.083 0.250 0.580 0.750 1.0000 1.250	1.000 1.000 8.999 17.998 17.995 8.990 17.981
1.500	8.968
1.750 2.ØØØ	8.948 Ø.922
2.250	8.888
3.000	8.744
3.083	8.725
3.250	8.686 Ø.583
4.000	8.500
4.250	8.440
4.750	8.330
5.000	8.281
5.250	8.238 8.199
5.750	8.165
6.000	8.136
6.500	8.090
7.800	8.027
8.583	8.012
20.700	8.001

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