

HYPOTHESIS TESTING AND REGION ESTIMATION
IN NONLINEAR REGRESSION

BY

A.H. POOI

THESIS SUBMITTED FOR THE DEGREE OF DOCTOR
OF PHILOSOPHY IN THE UNIVERSITY OF LONDON

ACKNOWLEDGEMENTS

I would like to express my gratitude to my supervisor, Professor E.M.L. Beale, for his guidance and encouragement. I would also like to thank the University of Malaya for its financial assistance.

Abstract

The present work is concerned with the problems of hypothesis testing and region estimation concerning subsets of components of the parameter vector in nonlinear regression models.

A fundamental approach to these problems is to devise methods for indicating when the use of usual linear theory results as approximations is justified. Measures of nonlinearity are proposed in Beale (1960) for this purpose. In the present work, the problem of finding bounds for these measures of nonlinearity within which it is justifiable to use linear theory results is investigated. The use of nonlinear transformations of the parameter vector for making a model more nearly linear is also discussed.

The main approach considered here is based on general maximum likelihood (m.l.) ratios. The derivation of truncated series expansions of the significance probabilities and power functions of the general m.l. ratio tests is considered. The use of a computer to do the algebraic manipulation involved in this derivation is also illustrated.

The above approaches are then compared by means of numerical examples.

Contents

	<u>Page</u>
CHAPTER 1. INTRODUCTION	1
Section 1.1 Nonlinear regression model	1
Section 1.2 The problems of hypothesis testing and region estimation	4
Section 1.3 Layout of the thesis	8
 CHAPTER 2. MEASURES OF NONLINEARITY	 9
Section 2.1 Definitions of measures of nonlinearity	9
Section 2.2 Computation of theoretical measures of nonlinearity	13
Section 2.3 Significance of measures of nonlinearity	18
Section 2.4 Reduction of nonlinearity for inference purposes	19
Section 2.5 Region estimate of a different subset of components of the parameter vector	25
Section 2.6 Conditions for the existence of the power transformation which achieves maximum reduction of the nonlinearity associated with an individual parameter	27
Section 2.7 Alternative transformations to reduce nonlinearity	34
Section 2.8 Effects of design of experiments on nonlinearity	36

	<u>Page</u>
CHAPTER 3. HYPOTHESIS TESTING AND REGION ESTIMATION BASED ON GENERAL MAXIMUM LIKELIHOOD RATIOS	41
Section 3.1 Introduction	41
Section 3.2 Hypothesis testing in unconstrained nonlinear models	41
Section 3.3 Significance probabilities of the general m.l. ratio tests	45
Section 3.4 Power functions of the general m.l. ratio tests	75
Section 3.5 Region estimation in unconstrained nonlinear models	90
Section 3.6 Inference of functions of the parameter vector based on general maximum likelihood ratios	92
Section 3.7 Hypothesis testing and region estimation in constrained nonlinear models	96
 CHAPTER 4. DERIVATION OF APPROXIMATIONS OF THE POWER OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TESTS USING A COMPUTER	97
Section 4.1 Introduction	97
Section 4.2 Representation of algebraic expressions by one dimensional arrays in a computer	99
Section 4.3 Algebraic manipulation done on a computer	99
Section 4.4 Representation of the equation $S_1^{DA}(z) = d_1^{+2}$ in a computer	100

Section 4.5 Representation of the equation

$$S_1^{DA}(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{r_1^{(+s)} - (s)} z_{p-k^*+1},$$

$$s_{p-k^*+2} \sqrt{r_1^{(+s)} - (s)} z_{p-k^*+2}, \dots, s_p \sqrt{r_1^{(+s)} - (s)} z_p,$$

$$z_{p+1}, z_{p+2}, \dots, z_n) = d_1^{+2}$$

in a computer

101

Section 4.6

Representation of $\frac{\partial r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+}$, $a^+ = 0$

$$\frac{\partial r_1^{(+s)}}{\partial a_{i_2 j_2 k_2}^+}, a^+ = 0 \quad \text{and} \quad \frac{\partial^2 r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+}, a^+ = 0$$

in a computer

102

Section 4.7

Computation of $\beta_{1a_{i_1 j_1 k_1}^+}$ and $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$

in a computer

102

Section 4.8

Partition of the set of all possible a_{ijk}^+
into subsets such that in each subset,

different a_{ijk}^+ have similar expressions of

$$\beta_{1a_{ijk}^+}$$

103

Section 4.9

Partition of the set of all possible $(a_{i_1 j_1 k_1}^+,$
 $a_{i_2 j_2 k_2}^+)$ into subsets such that in each subset,
different $(a_{i_1 j_1 k_1}^+, a_{i_2 j_2 k_2}^+)$ have similar
expressions of $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$

105

Section 4.10 Programs for deriving $\left[\frac{\partial r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+} \times \frac{\partial r_1^{(+s)}}{\partial a_{i_2 j_2 k_2}^+} \right] a^+ = 0$ and $\left[\frac{\partial^2 r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+} \right] a^+ = 0$	108
Section 4.11 Expressions of $\left[\frac{\partial r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+} \times \frac{\partial r_1^{(+s)}}{\partial a_{i_2 j_2 k_2}^+} \right] a^+ = 0$ $\left[\frac{\partial^2 r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+} \right] a^+ = 0$	and 108
Section 4.12 Programs for deriving $\beta_1(\theta_A, \sigma_A)$ Section 4.13 Numerical examples	130 130
CHAPTER 5. COMPARISON OF VARIOUS METHODS OF OBTAINING REGION ESTIMATES BY MEANS OF NUMERICAL EXAMPLES 147	
Section 5.1 Introduction Section 5.2 Estimation of coverage probability and nonlinearity Section 5.3 Region estimates of θ Section 5.4 Interval estimates of θ_i	147 148 176 190
Appendix 1 Appendix 2 Appendix 3 Appendix 4 References	216 218 222 245 266

CHAPTER 1

INTRODUCTION

Section 1.1 Nonlinear regression model

The present work is concerned with hypothesis testing and region estimation concerning one or more components of the parameter vector in the nonlinear regression model which can be described as follows.

Suppose that we have a set of observations y_u ($u = 1, 2, \dots, n$) and a set of corresponding theoretical mean values which we may write as $\eta(\xi_u, \theta)$. This notation indicates that the theoretical mean values depend on the conditions under which the u^{th} observation was taken, represented by a vector ξ_u of independent variables, and also on the parameter vector $\theta = (\theta_1, \theta_2, \dots, \theta_p)^T$ which is assumed to lie in a certain set Ω , the parameter space. Then the model can be expressed as

$$(1.1.1) \quad y_u = \eta(\xi_u, \theta) + \varepsilon_u,$$

where ε_u are random errors with zero means and some statistical distribution.

It is convenient to consider a set of observations y_u as a point $y = (y_1, y_2, \dots, y_n)^T$ in an n -dimensional Euclidean space. Within this space there is a subset consisting of points each of which fits the theoretical model exactly for some value of θ . We call this subset the Solution Locus. In symbols a point in the solution locus has its u^{th} coordinate η_u defined by

$$(1.1.2) \quad \eta_u = \eta(\xi_u, \theta) \text{ for all } u,$$

for some values of $\underline{\theta}$. Let $P(\underline{\theta})$ denote the point whose coordinates are defined by (1.1.2). The least squares estimate for $\underline{\theta}$ is then the value of $\underline{\theta}$ for which $P(\underline{\theta})$ is nearest to the observed sample point y . In the present work we make the assumption that the ε_u are independently normally distributed with a common variance σ^2 . This assumption implies that the least squares estimate of the parameter vector $\underline{\theta}$ is also the maximum likelihood estimate $\hat{\underline{\theta}}$.

Each of the function $\eta(\xi_u, \underline{\theta})$ can be either linear or nonlinear in $\underline{\theta}$. If

- (i) all $\eta(\xi_u, \underline{\theta})$ are linear in $\underline{\theta}$,
- (ii) all real $\underline{\theta}$ are in Ω ,
- (iii) the $(n \times p)$ matrix $\{c_{uj}\} = \left\{ \frac{\partial \eta(\xi_u, \underline{\theta})}{\partial \theta_j} \right\}$ is of rank p ,

then the corresponding solution locus is a p -dimensional hyperplane in sample space. Furthermore the components of $\underline{\theta}$ define a Cartesian, i.e. uniform, system of coordinates in this hyperplane. If one or more $\eta(\xi_u, \underline{\theta})$ are nonlinear in $\underline{\theta}$, then the derivatives c_{uj} may become functions $c_{uj}(\underline{\theta})$ of $\underline{\theta}$. The solution locus may then be a distorted hyperplane. We refer to this solution locus as an "unconstrained" solution locus, and the corresponding model as an "unconstrained" model. This solution locus is to be distinguished from one which has boundary points, in which case the solution locus and the model are "constrained". It is important to realize that not all models are unconstrained. Typically parameters must lie between zero and infinity, and the solution locus may have a definite boundary where a function of $\underline{\theta}$ tends to a limit. For example consider the models with $\eta(\xi_u, \underline{\theta})$ given by

$$(A) \quad \eta(\xi_u, \underline{\theta}) = \frac{\theta_1}{\theta_1 - \theta_2} (e^{-\theta_2 \xi_u} - e^{-\theta_1 \xi_u}),$$

where $0 < \theta_1 < \infty$, $0 < \theta_2 < \infty$, and

$$\xi_u = 0.25, 0.5, 1, 1.5, 2, 4,$$

$$(B) \quad n(\xi_u, \theta) = 1 - \frac{1}{\theta_1 - \theta_2} (\theta_1 e^{-\theta_2 \xi_u} - \theta_2 e^{-\theta_1 \xi_u}),$$

where $0 < \theta_1 < \infty$, $0 < \theta_2 < \infty$, and

$$\xi_u = 1, 2, 3, 4, 5, 6.$$

(c.f. Guttman and Meeter (1965)).

We observe that as θ_1 or θ_2 tends to zero, the $n(\xi_u, \theta)$ in models (A) and (B) tend to finite limits. This implies that the solution loci of these models are constrained. In particular as θ_2 tends to zero, the $n(\xi_u, \theta)$ in model (A) tend to finite limits which depend on θ_1 . Thus the solution locus of model (A) has a boundary where θ_2 tends to zero. An important feature of this boundary is that the $c_{uj}(\theta)$ tend to zeros as θ_2 tends to zero. We next note that as θ_1 tends to infinity, the $n(\xi_u, \theta)$ in models (A) and (B) tend to finite limits which depend on θ_2 . This implies that each of the corresponding solution loci has a boundary where θ_1 tends to infinity. For each of these models, the $c_{uj}(\theta)$ tend to zeros as $P(\theta)$ approaches this boundary. We also observe that $P(\theta)$ in the solution locus of model (B) remains the same if we interchange θ_1 and θ_2 . This suggests that we may impose the constraint that $\theta_1 \geq \theta_2$. Furthermore we note that as $(\theta_1 - \theta_2)$ tends to zero, the $n(\xi_u, \theta)$ tend to finite limits which depend on θ_1 . Therefore the solution locus has a boundary where $(\theta_1 - \theta_2)$ tends to zero. Finally we note that the matrix $\{c_{uj}(\theta)\}$ becomes of rank one, which is less than p , as $(\theta_1 - \theta_2)$ tends to zero. We can examine the solution loci of these models more closely if we apply an orthogonal transformation of coordinates in sample space. The details of this transformation will be described in Chapter 2.

The effects of this transformation are that the point $P(\theta_T)$ associated with the true value θ_T of θ , where for these models we choose $\theta_T = (1.4, 0.4)^T$, becomes the new origin corresponding to the transformed coordinates $\underline{z} = 0$, and the plane tangent to the solution locus at $P(\theta_T)$ consists of points for which $z_i = 0$ for $i = 3, 4, 5, 6$. In Fig. (1.1.1) and (1.1.2) we display the first two coordinates z_1 and z_2 of a number of points in the solution loci of these models. Each line in these figures corresponds to some constant value of θ_1 or θ_2 . The values of R are for indicating the severity of an aspect of nonlinearity of the model.

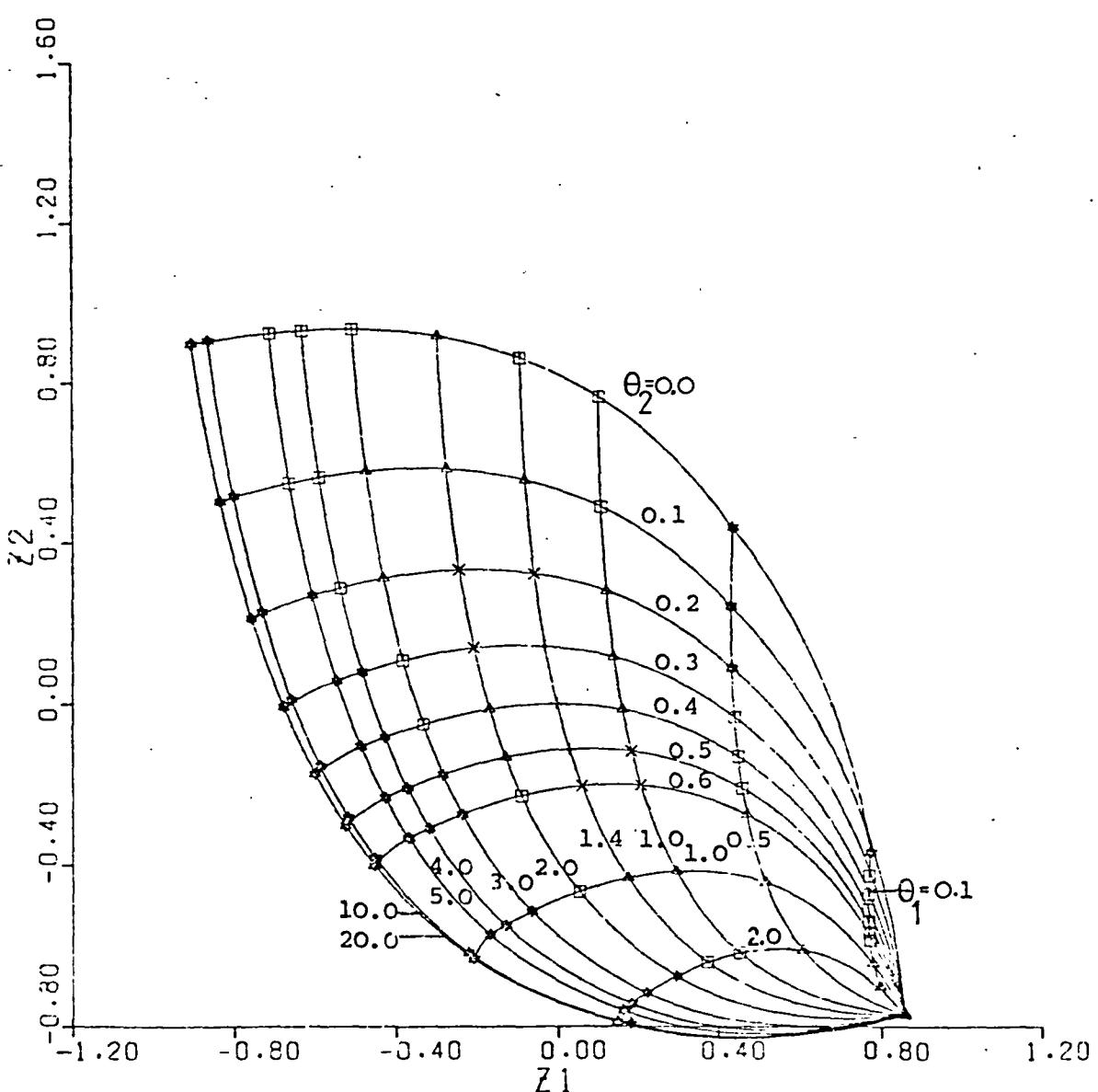
The problems of hypothesis testing and region estimation for a model with a constrained solution locus are still open questions. Similar problems for a model with an unconstrained solution locus are less difficult. The present work is mainly concerned with the discussion of methods appropriate to the latter problem, and situations under which these methods may be applied if the model is constrained.

Section 1.2 The problems of hypothesis testing and region estimation

Many computer programs, using a variety of numerical methods, have been written to find point estimates of the parameter vector θ using the least squares criterion. Attention has also been paid to the problems of hypothesis testing and region estimation concerning k^* ($1 \leq k^* \leq p$) components of interest in the parameter vector, treating the other components, if any, as nuisance parameters. Although large sample methods have been proposed, and justified asymptotically as n tends to infinity, these problems are known to be rather intractable when n is small and the functions $n(\xi_u, \theta)$ have no special properties that simplify the analysis.

FIGURE (J.1.1)
 ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
 MODEL IS
 $E(Y) = (\Theta_{T1}/(\Theta_{T1}-\Theta_{T2}))$
 $\times (\exp(-\Theta_{T2} \cdot X)) - \exp(-\Theta_{T1} \cdot X))$

XI = 0.25 0.5 1.0 1.5 2.0 4.0
 THETAI TRUE ARE 1.4000 0.4000



R=SQUARE ROOT ((SUM FROM P+1 TO N OF ZI SQUARE)/
 (SUM FROM 1 TO P OF ZI SQUARE))
 $+ : 0 \leq R \leq 0.05 ; X : 0.05 < R \leq 0.1 ; \Delta : 0.1 < R \leq 0.2 ; \square : 0.2 < R \leq 0.3 ; \star : R > 0.3$

FIGURE (1.1.2)

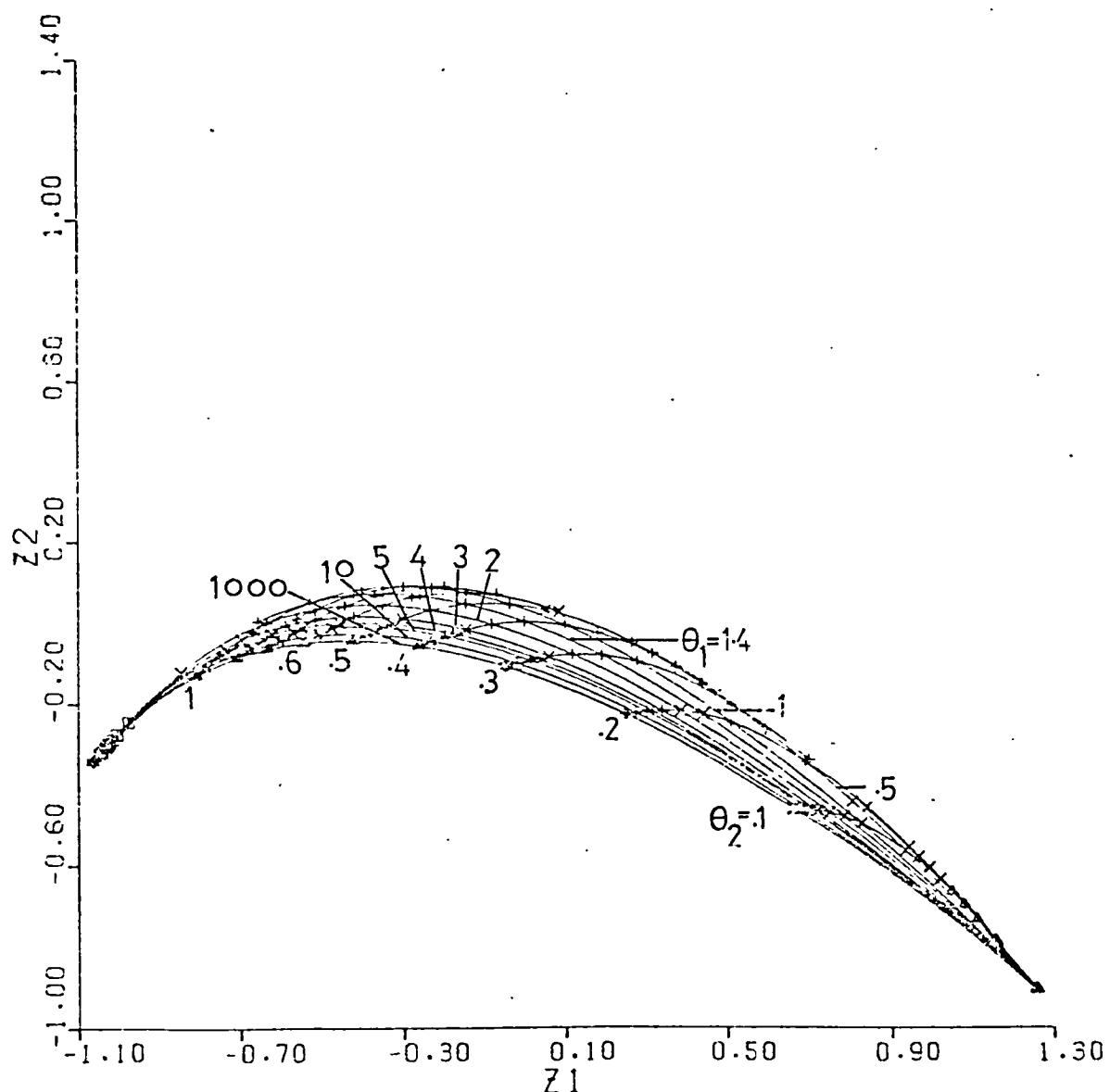
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

$$E(Y) = I - (\Theta_{11} \cdot \exp(-\Theta_{12} \cdot X_1) - \Theta_{21} \cdot \exp(-\Theta_{22} \cdot X_1)) / (\Theta_{11} - \Theta_{21})$$

 $X_1 = 1, 2, 3, 4, 5, 6$

THETA1 TRUE ARE 1.4000 0.4000



$R = \text{SQUARE ROOT } ((\text{SUM FROM } P+1 \text{ TO } N \text{ OF } Z_i^2) / (\text{SUM FROM } 1 \text{ TO } P \text{ OF } Z_i^2))$

+: $0 \leq R \leq 0.05$; x: $0.05 < R \leq 0.1$; Δ: $0.1 < R \leq 0.2$; □: $0.2 < R \leq 0.3$; ★: $R > 0.3$

We first explain the following phrases before outlining some approaches to these problems. We refer to a model as being "approximately linear in the parameter vector θ " if and only if the solution locus can be approximated, for statistical purposes, by a p-dimensional linear manifold in which the components of θ define a uniform system of coordinates. We next refer to a model as being "approximately intrinsically linear" if and only if the solution locus can be approximated, for statistical purposes, by a p-dimensional linear manifold in sample space. We further refer to a model as being "approximately intrinsically linear in the parameters $\theta_{i_1}, \theta_{i_2}, \dots, \theta_{i_{p-k^*}}$ ", where $k^* < p$, if and only if the set of points in solution locus such that $\theta_{i_j} = \theta_{Ti_j}$ for all $j > p-k^*$ can be approximated, for statistical purposes, by a $(p-k^*)$ -dimensional linear manifold in sample space.

If the model is approximately linear in the parameter vector θ , these problems are straightforward. If not there are two approaches that can be considered:

- either make nonlinear transformations of the original parameters in such a way that there are at least k^* of the transformed parameters which depend only on the original parameters of interest,
- or carry out general maximum likelihood ratio tests and derive region estimates based on general maximum likelihood ratio criterion.

The former approach is based implicitly on the assumption that the model is approximately linear in the transformed parameter vector. The latter is based implicitly on the assumptions that

- (a) if $k^* = p$, then the model is approximately intrinsically linear,

- (b) if $k^* < p$, then the model apart from being approximately intrinsically linear, is also approximately intrinsically linear in the nuisance parameters.

.....
Section 1.3 Layout of the thesis

In Chapter 2, we investigate the former approach. A more efficient and more illuminating method of computing Beale's measures of nonlinearity using Householder transformations is first described. Then nonlinear transformations of the parameters, in particular power transformations, are used to reduce the nonlinearity for inference purposes.

In Chapter 3, we investigate the latter approach. General m.l. ratio tests are used to test a number of composite nonlinear hypotheses for making inference about subsets of components of the parameter vector. The derivation of the significance probabilities and power functions of the tests for these hypotheses is considered. The estimation of the coverage probabilities of the region estimates based on these tests is also discussed.

In Chapter 4, the derivation of series expansions of the power functions of the tests in Chapter 3 using a computer is discussed. Computer programs for deriving these series expansions truncated after some finite number of terms are presented.

Chapter 5 is devoted to the comparison of the methods of constructing interval and region estimates in Chapters 2 and 3 by means of numerical examples.

CHAPTER 2

MEASURES OF NONLINEARITY

Section 2.1 Definitions of measures of nonlinearity

Throughout this chapter, we make the assumptions that the particular observation y that we have obtained is such that $\hat{\theta}$ obtained by minimizing the residual sum of squares

$$S(\underline{\theta}) = \sum_{u=1}^n \{y_u - \eta(\xi_u, \underline{\theta})\}^2$$

is the unique unconstrained minimum, and for each ξ_u , $\eta(\xi_u, \underline{\theta})$ is continuous in $\underline{\theta}$ at $\underline{\theta} = \hat{\theta}$.

A point $P(\underline{\theta}^*)$ associated with a feasible $\underline{\theta}^*$ will be referred to in the present work as a "nonsingular point" in the solution locus if and only if

- (a) $P(\underline{\theta}^*) = P(\underline{\theta}_A)$ implies that $\underline{\theta}^* = \underline{\theta}_A$
- (b) there exists an $\epsilon > 0$ such that

$$|\underline{\theta} - \underline{\theta}^*| < \epsilon$$

implies that $\underline{\theta} \in \Omega$

- (c) for each ξ_u , $\eta(\xi_u, \underline{\theta})$ is a function of $\underline{\theta}$ differentiable up to the second order at $\underline{\theta} = \underline{\theta}^*$
- and (d) the $(n \times p)$ matrix $\{c_{uj}(\underline{\theta}^*)\}$ is of rank p .

Thus the point $P(\hat{\theta})$ associated with the $\hat{\theta}$ is a nonsingular point in the solution locus if and only if (b), (c) and (d) with $\underline{\theta}^*$ replaced by $\hat{\theta}$ hold.

In Beale (1960), numerical measures of nonlinearity in a sufficiently small neighbourhood of a nonsingular point $P(\hat{\theta})$ in the solution locus are introduced as functions of the experimental design, the parameterization of the model and the least-squares estimate $\hat{\theta}$.

The original and revised versions of the definitions of the measures shall now be introduced. First we note that the linear approximations of $\eta(\xi_u, \theta)$ as a function of θ , valid in the neighbourhood of $\hat{\theta}$, can be written as

$$(2.1.1) \quad \eta(\xi_u, \theta) = \eta(\xi_u, \hat{\theta}) + \sum_{j=1}^p c_{uj} t_j + o(t^2),$$

where $c_{uj} = c_{uj}(\hat{\theta})$,

$$(2.1.2) \quad t_j = \theta_j - \hat{\theta}_j$$

and $t^2 = \sum_{j=1}^p t_j^2$.

The plane tangent to the solution locus at $P(\hat{\theta})$ is then defined parametrically by

$$(2.1.3) \quad \eta_u = \eta(\xi_u, \hat{\theta}) + \sum_{j=1}^p c_{uj} t_j .$$

Let $T(\theta)$ be the point whose coordinates are given by (2.1.3) when $t_j = \theta_j - \hat{\theta}_j$. Further, let $T^*(\theta)$ be the point on the tangent plane such that the line joining $T^*(\theta)$ and $P(\hat{\theta})$ is perpendicular to the tangent plane. Next, consider a set of parameter values $\theta_1, \theta_2, \dots, \theta_w$ near $\hat{\theta}$. Let t_{wj} ($j = 1, 2, \dots, p$) be the j^{th} component of $t_w = \theta_w - \hat{\theta}$ and $\eta_{uw}, \hat{\eta}_u$ ($u = 1, 2, \dots, n$) be the values $\eta(\xi_u, \theta_w), \eta(\xi_u, \hat{\theta})$ respectively. We can regard the expression

$$(2.1.4) \quad Q_{\theta} = \sum_{w=1}^W \sum_{u=1}^n (\eta_{uw} - \hat{\eta}_u - \sum_{j=1}^p c_{uj} t_{wj})^2 = \sum_{w=1}^W |P(\theta_w) - T(\theta_w)|^2$$

as a crude measure of the total nonlinearity of the model in terms of the parameter vector $\underline{\theta}$ in the neighbourhood of $P(\hat{\theta})$. This Q_{θ} is essentially the sum of squares of distances from the points $P(\theta_w)$ in the solution locus to the associated point $T(\theta_w)$ on the tangent plane at $P(\hat{\theta})$.

As the measure Q_{θ} depends on the number of points $P(\theta_w)$ that one uses and on their distances from $P(\hat{\theta})$, it is necessary to normalize this measure. In the neighbourhood of $P(\hat{\theta})$, $\eta_{uw} - \hat{\eta}_u - \sum_{j=1}^p c_{uj} t_{wj}$ can be expected to be roughly proportional to the square of the distance of $P(\theta_w)$ from $P(\hat{\theta})$ i.e. proportional to $\sum_{u=1}^n (\eta_{uw} - \hat{\eta}_u)^2$. So it is natural to divide Q_{θ} by

$$(2.1.5) \quad D = \sum_{w=1}^W \left\{ \sum_{u=1}^n (\eta_{uw} - \hat{\eta}_u)^2 \right\}^2 = \sum_{w=1}^W |P(\theta_w) - P(\hat{\theta})|^4.$$

As Q_{θ} has the dimensions of the square of an observation, and D has the dimensions of the fourth power of an observation, the quantity

$$(2.1.6) \quad \hat{N}_{\theta} = ps^2 Q_{\theta} / D,$$

where s^2 is an estimate of σ^2 , is a dimensionless quantity, and can be regarded as an estimated normalized measure of the total nonlinearity of the model in terms of the parameter vector $\underline{\theta}$ in the neighbourhood of $P(\hat{\theta})$. The reason for the factor p in (2.1.6) is given in Beale (1960).

The empirical measure of nonlinearity \hat{N}_{θ} given by (2.1.6) has the theoretical measure of nonlinearity N_{θ} as its counterpart. In Beale (1960), N_{θ} is derived from \hat{N}_{θ} by altering s^2 to σ^2 , and changing the finite set of values of θ_w to an infinite set of values of $\underline{\theta}$ such that

the points $T(\underline{\theta})$ have a p -dimensional spherical normal distribution about $P(\hat{\underline{\theta}})$ with an arbitrarily small variance. But it now seems preferable to replace the set of points $T(\underline{\theta})$ by the points $T^*(\underline{\theta})$. This ensures that no transformation of the parameter vector $\underline{\theta}$ can change the chosen set of points $P(\underline{\theta})$. Both $N_{\underline{\theta}}$ and $\hat{N}_{\underline{\theta}}$ are invariant under any linear transformation of $\underline{\theta}$ (for $\hat{N}_{\underline{\theta}}$, the values $\underline{\theta}_1, \underline{\theta}_2, \dots, \underline{\theta}_W$ are held constant in these transformations), and also invariant under any orthogonal transformation of coordinates in sample space.

Now suppose we fix the model and the experimental design and then make arbitrary transformations of $\underline{\theta}$, say $\psi = \psi(\underline{\theta})$. Suppose the minimum value of $N_{\underline{\theta}}$ under these transformations is attained by using the transformation $\phi = \phi(\underline{\theta})$, and is denoted by N_{ϕ} . In Beale (1960), N_{ϕ} is referred to as the intrinsic nonlinearity of the model in the neighbourhood of $P(\hat{\underline{\theta}})$. The geometrical interpretation of N_{ϕ} is that it is the value of $N_{\underline{\theta}}$ when the parameter vector $\underline{\theta}$ is transformed in such a way that $T(\underline{\theta})$ is always at the point $T^*(\underline{\theta})$. The difference $N_{\underline{\theta}} - N_{\phi}$ may be called the removable nonlinearity of the model in terms of parameter vector $\underline{\theta}$ in the neighbourhood of $P(\hat{\underline{\theta}})$.

The theoretical measures of nonlinearity $N_{\underline{\theta}}$ and N_{ϕ} are derived in Beale (1960) from the first and second partial derivatives of $n(\xi_u, \underline{\theta})$ with respect to the θ_j . These theoretical measures together with the empirical measures were investigated by Guttman and Meeter (1965). These authors concluded that the empirical measures may significantly underestimate the nonlinearity of the model, and are therefore unreliable, but that the corresponding theoretical measures give an indication of the severity of the nonlinearity - although the interpretation of the measures suggested in Beale (1960) is unduly conservative. In latter sections and in Chapter 5 we will continue the investigation of the theoretical measures of nonlinearity.

Section 2.2 Computation of theoretical measures of nonlinearity

A method based on Householder transformations for computing the theoretical measures of nonlinearity for a given model in the neighbourhood of a nonsingular point $P(\hat{\theta})$ will now be described.

As each $\eta(\xi_u, \theta)$ as a function of θ is differentiable up to the second order at $\theta = \hat{\theta}$, we can obtain second order approximations to $\eta(\xi_u, \theta)$ as a function of θ , valid in the neighbourhood of $P(\hat{\theta})$, as

$$(2.2.1) \quad \eta(\xi_u, \theta) = \eta(\xi_u, \hat{\theta}) + \sum_{j=1}^p c_{uj} t_j + \sum_{j=1}^p \sum_{k=1}^p c_{ujk} t_j t_k + o(t^2),$$

$$\text{where } c_{ujk} = \frac{1}{2} \left[\frac{\partial^2 \eta(\xi_u, \theta)}{\partial \theta_j \partial \theta_k} \right]_{\theta=\hat{\theta}}.$$

Now let $\underline{\eta}$, $\hat{\underline{\eta}}$, \underline{c}_j and \underline{c}_{jk} denote the $(n \times 1)$ vectors whose u^{th} components are $\eta(\xi_u, \theta)$, $\eta(\xi_u, \hat{\theta})$, c_{uj} and c_{ujk} respectively. Further, let \underline{C} be the $(n \times p)$ matrix $\{c_{uj}\}$, and \underline{H} an $(n \times n)$ orthogonal matrix such that $\underline{H}\underline{C}$ is an upper triangular $(p \times p)$ nonsingular matrix \underline{D} with an $((n-p) \times p)$ zero matrix beneath it. \underline{H} can be written as a product of p orthogonal $(n \times n)$ matrices $\underline{H}^{(p)}, \underline{H}^{(p-1)}, \dots, \underline{H}^{(2)}, \underline{H}^{(1)}$ corresponding to p Householder transformations. Each $\underline{H}^{(j)}$ can be written as

$$\underline{H}^{(j)} = \underline{I} - [\underline{v}^{(j)}][\underline{v}^{(j)}]^T,$$

where the $(n \times 1)$ vectors $\underline{v}^{(j)}$ are computed as shown in Appendix 1.

We then apply an orthogonal transformation

$$(2.2.2) \quad \underline{H}(\underline{y} - \hat{\underline{\eta}}) = \underline{z}$$

of coordinates in sample space such that the point $P(\hat{\theta})$ in the solution locus becomes the new origin $\underline{z} = \underline{0}$ and the plane tangent to the solution

locus at $\hat{P}(\underline{\theta})$ consists of points for which $z_i = 0$ for $i = p+1, p+2, \dots, n$.

We refer to \underline{z} as the rotated coordinates of the sample point \underline{y} .

For a point $P(\underline{\theta})$ in the solution locus, the rotated coordinates are given by

$$(2.2.3) \quad z_i = \begin{cases} \sum_{j=i}^p d_{ij} t_j + \sum_{j=1}^p \sum_{k=1}^p d_{ijk} t_j t_k + o(t^2), & (i = 1, 2, \dots, p) \\ \sum_{j=1}^p \sum_{k=1}^p d_{ijk} t_j t_k + o(t^2), & (i = p+1, p+2, \dots, n), \end{cases}$$

where d_{ij} and d_{ijk} are the i^{th} components of the $(n \times 1)$ vectors $\underline{H} \underline{c}_j$ and $\underline{H} \underline{c}_{jk}$ respectively.

If we apply the linear transformation

$$(2.2.4) \quad \underline{\tau} = \underline{D} \underline{t}$$

of parameter vector \underline{t} , then (2.2.3) becomes

$$(2.2.5) \quad z_i = \begin{cases} \tau_i + \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} \tau_{\ell} \tau_m + o(\tau^2), & (i = 1, 2, \dots, p) \\ \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} \tau_{\ell} \tau_m + o(\tau^2), & (i = p+1, p+2, \dots, n), \end{cases}$$

$$\text{where } \tau^2 = \sum_{i=1}^p \tau_i^2,$$

$$(2.2.6) \quad f_{ilm} = f_{iml} = \sum_{j=1}^p \sum_{k=1}^m d_{ijk} d^{jl} d^{km},$$

and d^{jl} is the (j, l) entry of the inverse of D .

For the purpose of deriving the measures, we consider that u_i given by

$$u_i = z_i = \tau_i + \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} \tau_\ell \tau_m + o(\tau^2)$$

are independently normally distributed with mean zero and variance V where V is arbitrarily small. As Q_θ is given by

$$\begin{aligned} Q_\theta &= \sum_{i=1}^n \left(\sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} \tau_\ell \tau_m \right)^2 + o(\tau^4) \\ &= \sum_{i=1}^n \left(\sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} u_\ell u_m \right)^2 + o(u^4), \end{aligned}$$

where

$$u^4 = \left(\sum_{i=1}^p u_i^2 \right)^2,$$

the mean value of Q_θ is given by

$$\begin{aligned} \bar{Q}_\theta &= \int_{u_1=-\infty}^{\infty} \int_{u_2=-\infty}^{\infty} \dots \int_{u_p=-\infty}^{\infty} \left\{ \sum_{i=1}^n \left(\sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} u_\ell u_m \right)^2 + o(u^4) \right\} \left\{ \prod_{j=1}^p \frac{1}{\sqrt{2\pi V}} e^{-u_j^2/(2V)} \right\} \\ &\quad du_1 du_2 \dots du_p \\ &= E_u \left\{ \sum_{i=1}^n \left(\sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} u_\ell u_m \right)^2 + o(u^4) \right\} \end{aligned}$$

i.e.

$$(2.2.7) \quad \bar{Q}_\theta = V^2 \sum_{i=1}^n \sum_{\ell=1}^p \sum_{m=1}^p (f_{i\ell\ell} f_{imm} + 2f_{ilm}^2) + o(V^2).$$

Next, as the fourth power of the distance from $P(\theta)$ to $\hat{P}(\theta)$ is given by

$$\begin{aligned} |P(\theta) - \hat{P}(\theta)|^4 &= \left(\sum_{i=1}^p \tau_i^2 \right)^2 + o(\tau^4) \\ &= (u^4)^2 + o(u^4), \end{aligned}$$

the mean value of this fourth power is given by

$$\begin{aligned}\bar{D} &= \underset{\tilde{u}}{E} \left\{ \left(\sum_{i=1}^p u_i^2 \right)^2 + o(u^4) \right\} \\ &= p(p+2)v^2 + o(v^2).\end{aligned}$$

Then we see that

$$(2.2.8) \quad N_\theta = \frac{\sigma^2}{p+2} \sum_{i=1}^n \sum_{l=1}^p \sum_{m=1}^p (f_{ill} f_{imm} + 2f_{ilm}^2).$$

We next derive N_ϕ . As N_θ is based only on second order approximations of $\eta(\xi_u, \theta)$ as functions of θ , valid in the neighbourhood of $\hat{\theta}$, the terms of order higher than two in the expansions of the arbitrary transformations $\psi_i = \psi_i(\theta)$, valid in the neighbourhood of $\hat{\theta}$, are not relevant as far as reduction of total nonlinearity by means of the transformation ψ is concerned. Therefore, in general, we can write ψ_i as

$$\psi_i = \psi_{io} + \sum_{j=1}^p \left[\frac{\partial \psi_i}{\partial \theta_j} \right]_{\theta=\hat{\theta}} t_j + \frac{1}{2} \sum_{j=1}^p \sum_{k=1}^p \left[\frac{\partial^2 \psi_i}{\partial \theta_j \partial \theta_k} \right]_{\theta=\hat{\theta}} t_j t_k + o(t^2).$$

As N_θ is not changed by any linear transformation of parameter vector, we can restrict our attention to ψ such that

$$\psi_{io} = 0$$

and

$$\left[\frac{\partial \psi_i}{\partial \theta_j} \right]_{\theta=\hat{\theta}} = d_{ij}.$$

Then ψ_i can be written as

$$\psi_i = \sum_{j=1}^p d_{ij} t_j + \frac{1}{2} \sum_{j=1}^p \sum_{k=1}^p \left[\frac{\partial^2 \psi_i}{\partial \theta_j \partial \theta_k} \right]_{\theta=\hat{\theta}} t_j t_k + o(t^2).$$

After applying the transformation $\underline{\tau} = D\underline{t}$, we can write ψ_i as

$$(2.2.9) \quad \psi_i = \tau_i - \sum_{\ell=1}^p \sum_{m=1}^p g_{ilm} \tau_\ell \tau_m + o(\tau^2),$$

where $g_{ilm} = g_{iml}$.

If we apply this transformation, then (2.2.5) becomes

$$(2.2.10) \quad z_i = \begin{cases} \psi_i + \sum_{\ell=1}^p \sum_{m=1}^p (f_{ilm} + g_{ilm}) \psi_\ell \psi_m + o(\psi^2), & (i = 1, 2, \dots, p) \\ \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm} \psi_\ell \psi_m + o(\psi^2), & (i = p+1, p+2, \dots, n) \end{cases}$$

$$\text{where } \psi^2 = \sum_{i=1}^p \psi_i^2.$$

The theoretical measure of total nonlinearity in terms of parameter vector ψ in the neighbourhood of $P(\hat{\theta})$ is

$$(2.2.11) \quad N_\psi = \frac{\sigma^2}{p+2} \left\{ \sum_{i=1}^p \sum_{\ell=1}^p \sum_{m=1}^p [(f_{ill} + g_{ill})(f_{imm} + g_{imm}) + 2(f_{ilm} + g_{ilm})^2] \right. \\ \left. + \sum_{i=p+1}^n \sum_{\ell=1}^p \sum_{m=1}^p [f_{ill} f_{imm} + 2f_{ilm}^2] \right\}.$$

Minimizing N_ψ with respect to g_{ilm} ($i, \ell, m = 1, 2, \dots, p$), we obtain the theoretical measure of intrinsic nonlinearity in the neighbourhood of $P(\hat{\theta})$ as

$$(2.2.12) \quad N_\phi = \frac{\sigma^2}{p+2} \sum_{i=p+1}^n \sum_{\ell=1}^p \sum_{m=1}^p (f_{ill} f_{imm} + 2f_{ilm}^2).$$

The theoretical measure of removable nonlinearity in terms of parameter $\underline{\theta}$ in the neighbourhood of $P(\hat{\underline{\theta}})$ is now given by

$$(2.2.13) \quad N_{\underline{\theta}} - N_{\underline{\phi}} = \frac{\sigma^2}{p+2} \sum_{i=1}^p \sum_{l=1}^p \sum_{m=1}^p (f_{ill} f_{imm} + 2f_{ilm}^2).$$

If we set g_{plm} ($l, m = 1, 2, \dots, p$) in (2.2.9) to be zeros so that the last component ψ_p of the transformed parameter vector ψ is a linear function of θ_p , and choose g_{ilm} ($i = 1, 2, \dots, p-1; l, m = 1, 2, \dots, p$) such that the total nonlinearity is minimized, then the resulting minimum value of the total nonlinearity may be called the measure of nonlinearity associated with θ_p in the neighbourhood of $P(\hat{\underline{\theta}})$ and is given by

$$(2.2.14) \quad N_{\theta_p} = \frac{\sigma^2}{p+2} \sum_{i=p}^n \sum_{l=1}^p \sum_{m=1}^p (f_{ill} f_{imm} + 2f_{ilm}^2).$$

By permuting the positions of the components of $\underline{\theta}$, each component θ_i ($i = 1, 2, \dots, p-1$) can be in the last position and N_{θ_i} ($i = 1, 2, \dots, p-1$) can be obtained in a similar way as N_{θ_p} .

Section 2.3 Significance of measures of nonlinearity

We now investigate how the total nonlinearity in terms of the parameter vector ψ , where ψ is either the original or the transformed parameter vector, can be used to conclude, or to indicate, that the model is approximately linear in the parameter vector ψ .

Suppose σ^2 is known and all $P(\underline{\theta})$ are nonsingular points. Further, let $N_{\psi_{crit}}$ be the value such that if N_{ψ} at the true value ψ_T of ψ is less than $N_{\psi_{crit}}$, then the use of linear theory results as approximations is justified. Now if all N_{ψ} evaluated at feasible ψ is less than $N_{\psi_{max}}$, then N_{ψ} at ψ_T is less than $N_{\psi_{max}}$. Thus if $N_{\psi_{max}}$ is less than $N_{\psi_{crit}}$,

then we can conclude that the model is approximately linear in the parameter vector ψ . The problem of finding this N_{ψ} will be considered in Chapter 5.

In practice, we may use N_{ψ} evaluated at the least squares estimate $\hat{\psi}$ of ψ , and $\sigma^2 = s^2$ to estimate N_{ψ} at ψ_T . Information concerning the reliability of this estimate may be derived from N_{ψ} evaluated at feasible ψ in the neighbourhood of $\hat{\psi}$, and σ^2 in the neighbourhood of s^2 . Suppose this estimate is small enough and is reliable. Then it is plausible to believe that linear theory results can be applied, with negligible errors, to make inference about ψ . The estimation of N_{ψ} at ψ_T when ψ is the original parameter vector θ will be investigated in Chapter 5.

Section 2.4 Reduction of nonlinearity for inference purposes

The total nonlinearity N_{θ} in the neighbourhood of $P(\hat{\theta})$ may be large but the corresponding N_{ϕ} of intrinsic nonlinearity may be fairly small. In these circumstances, we can apply nonlinear transformation of the parameter vector θ to reduce the total nonlinearity. The choice of transformation depends on the type of inference that we want to make about θ . If we want to obtain region estimate for the last k^* ($k^* = 1, 2, \dots, p$) components of the parameter vector θ , then it is convenient to use a one to one transformation $\gamma = \gamma(t)$ of the form

(2.4.1) $\gamma_i = \gamma_i(t)$, $(i = 1, 2, \dots, p)$ in the case when $k^* = p$,
and in the case when $k^* < p$,

$$(2.4.2) \quad \gamma_i = \gamma_i(t), \quad (i = 1, 2, \dots, p-k^*),$$

$$(2.4.3) \quad \gamma_i = \gamma_i(t_{p-k^*+1}, t_{p-k^*+2}, \dots, t_p), \quad (i = p-k^*+1, p-k^*+2, \dots, p),$$

where γ_i are differentiable up to the second order.

We choose $\gamma = \gamma(t)$ to be such that the nonlinearity N_γ in terms of the transformed parameter γ is minimized. For reasons similar to those given in the derivation of N_ϕ in section 2.2, we can restrict our attention to γ_i of the form

$$(2.4.4) \quad \gamma_i = t_i - \sum_{\ell=1}^p \sum_{m=1}^p g_{ilm} t_\ell t_m + o(t^2),$$

where $g_{ilm} = g_{iml}$. Suppose $g_{ilm}^{(k*)}$ are such that the corresponding γ achieves maximum reduction of nonlinearity. Then we refer to $g_{ilm}^{(k*)}$ as the optimal g_{ilm} . To find $g_{ilm}^{(k*)}$, we first write (2.4.4) as

$$(2.4.5) \quad t_i = \gamma_i + \sum_{\ell=1}^p \sum_{m=1}^p g_{ilm} \gamma_\ell \gamma_m + o(\gamma^2),$$

where $\gamma^2 = \sum_{i=1}^p \gamma_i^2$.

Substituting these expressions for t_i into (2.2.3), we obtain

$$(2.4.6) \quad z_i = \begin{cases} \sum_{j=i}^p d_{ij} \gamma_j + \sum_{j=1}^p \sum_{k=1}^p d_{ijk}^+ \gamma_j \gamma_k + o(\gamma^2), & (i = 1, 2, \dots, p) \\ \sum_{j=1}^p \sum_{k=1}^p d_{ijk}^+ \gamma_j \gamma_k + o(\gamma^2), & (i = p+1, p+2, \dots, n), \end{cases}$$

where

$$(2.4.7) \quad d_{ijk}^+ = \begin{cases} d_{ijk} + \sum_{\ell=i}^p d_{il} g_{\ell j k}, & (i = 1, 2, \dots, p) \\ d_{ijk}, & (i = p+1, p+2, \dots, n). \end{cases}$$

We now apply the transformation $\xi^+ = D\gamma$ so that (2.4.6) becomes

$$(2.4.8) \quad z_i = \begin{cases} \tau_i^+ + \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm}^+ \tau_\ell^+ \tau_m^+ + o((\tau^+)^2), & (i = 1, 2, \dots, p) \\ \sum_{\ell=1}^p \sum_{m=1}^p f_{ilm}^+ \tau_\ell^+ \tau_m^+ + o((\tau^+)^2), & (i = p+1, p+2, \dots, n), \end{cases}$$

where

$$(2.4.9) \quad f_{ilm}^+ = \sum_{j=1}^l \sum_{k=1}^m d_{ijk}^+ d^{jl} d^{km},$$

and

$$(\tau^+)^2 = \sum_{i=1}^p (\tau_i^+)^2.$$

The total nonlinearity in terms of parameter vector γ in the neighbourhood of $P(\hat{\theta})$ is then given by

$$(2.4.10) \quad N_\gamma = \frac{\sigma^2}{p+2} \sum_{i=1}^n \sum_{\ell=1}^p \sum_{m=1}^p [f_{ill}^+ f_{imm}^+ + 2(f_{ilm}^+)^2].$$

N_γ is now minimized with respect to g_{ijk} ($i, j, k = 1, 2, \dots, p$). It can be shown that if $k^* = p$, $g_{ijk}^{(p)}$ are given by

$$(2.4.11) \quad d_{ijk} + \sum_{\ell=i}^p d_{il\ell} g_{\ell j k}^{(p)} = 0, \quad (i, j, k = 1, 2, \dots, p),$$

and if $k^* < p$, $g_{ijk}^{(k*)}$ are given by

$$(2.4.12) \quad g_{ijk}^{(k*)} = \begin{cases} \sum_{\ell=p-k^*+1}^j \sum_{m=p-k^*+1}^k s_{ilm} d_{\ell j} d_{mk}, & (i, j, k = p-k^*+1, p-k^*+2, \dots, p) \\ 0 & , \quad (i = p-k^*+1, p-k^*+2, \dots, p; \\ & j, k = 1, 2, \dots, p-k^*), \end{cases}$$

$$(2.4.13) \quad d_{ijk} + \sum_{\ell=i}^p d_{i\ell} g_{\ell jk}^{(k*)} = 0, \quad (i = 1, 2, \dots, p-k*; j, k = 1, 2, \dots, p),$$

$$(2.4.14) \quad 2(f_{ijj} + \sum_{k=i}^p d_{ik} s_{kj}^{(k)}) + \sum_{\ell=1}^p f_{i\ell\ell} + \sum_{m=p-k*+1}^p \sum_{k=i}^p d_{ik} s_{kmm} = 0,$$

$$(i, j = p-k*+1, p-k*+2, \dots, p),$$

and

$$(2.4.15) \quad f_{ilm} + \sum_{k=i}^p d_{ik} s_{klm} = 0, \quad (i, l, m = p-k*+1, p-k*+2, \dots, p \text{ and } l \neq m).$$

We note that if $k^* = 1$, then from (2.4.12) and (2.4.14), we obtain

$$(2.4.16) \quad g_{ppp}^{(1)} = -\frac{d_{pp}}{3} \left(\sum_{m=1}^{p-1} f_{pmm} + 3f_{ppp} \right),$$

and the nonlinearity associated with γ_p in the neighbourhood of $P(\hat{\theta})$ is given by

$$(2.4.17) \quad N_{\gamma_p} = \frac{\sigma^2}{p+2} \left[\frac{2}{3} \left(\sum_{\ell=1}^{p-1} f_{p\ell\ell} \right)^2 + 2 \sum_{\ell=1}^p \sum_{m=1}^p f_{p\ell m}^2 - 2f_{ppp}^2 \right.$$

$$\left. + \sum_{i=p+1}^n \sum_{\ell=1}^p \sum_{m=1}^p (f_{i\ell\ell} f_{imm} + 2f_{ilm}^2) \right].$$

Now the task of finding γ will be complete if the terms of order higher than two in (2.4.4) can be found. For the case when $k^* = 1$ and θ_p is non-negative, it is convenient to use a power transformation of the form

$$(2.4.18) \quad \psi_p = \begin{cases} \frac{\theta_p^{\lambda_p - 1}}{\lambda_p} & \text{if } \lambda_p \neq 0 \\ \ln \theta_p & \text{if } \lambda_p = 0, \end{cases}$$

where the parameter of transformation λ_p is defined in terms of $g_{ppp}^{(1)}$ by the equation

$$(2.4.19) \quad \lambda_p = 1 - 2g_{ppp}^{(1)}.$$

We need a further linear transformation to derive γ_p from ψ_p . The parameters γ_i for $i < p$ are yet not found. But as far as finding an interval estimate for θ_p is concerned, we need not find these γ_i explicitly. In fact an interval estimate of θ_p is readily seen to be the set of feasible values of θ_p which lie in the following interval:

$$(2.4.20) \quad \left\{ \begin{array}{l} [\hat{\theta}_p^{\lambda_p - k_\alpha} | \lambda_p \hat{\theta}_p^{\lambda_p - 1} | \sigma_{\theta_p}]^{1/\lambda_p} \leq \theta_p \leq [\hat{\theta}_p^{\lambda_p + k_\alpha} | \lambda_p \hat{\theta}_p^{\lambda_p - 1} | \sigma_{\theta_p}]^{1/\lambda_p}, \\ (\lambda_p \neq 0; \sigma^2 \text{ is known}) \\ \exp[\ln \hat{\theta}_p^{-k_\alpha} \sigma_{\theta_p} / \hat{\theta}_p] \leq \theta_p \leq \exp[\ln \hat{\theta}_p^{+k_\alpha} \sigma_{\theta_p} / \hat{\theta}_p], \\ (\lambda_p = 0; \sigma^2 \text{ is known}) \end{array} \right.$$

or

$$(2.4.21) \quad \left\{ \begin{array}{l} [\hat{\theta}_p^{\lambda_p - t_\alpha(n-p)} | \lambda_p \hat{\theta}_p^{\lambda_p - 1} | \hat{\sigma}_{\theta_p}]^{1/\lambda_p} \leq \theta_p \leq [\hat{\theta}_p^{\lambda_p + t_\alpha(n-p)} | \lambda_p \hat{\theta}_p^{\lambda_p - 1} | \hat{\sigma}_{\theta_p}]^{1/\lambda_p}, \\ (\lambda_p \neq 0; \sigma^2 \text{ is unknown}) \\ \exp[\ln \hat{\theta}_p^{-t_\alpha(n-p)} \hat{\sigma}_{\theta_p} / \hat{\theta}_p] \leq \theta_p \leq \exp[\ln \hat{\theta}_p^{+t_\alpha(n-p)} \hat{\sigma}_{\theta_p} / \hat{\theta}_p], \\ (\lambda_p = 0; \sigma^2 \text{ is unknown}), \end{array} \right.$$

where

$$\sigma_{\theta_p} = |d^{pp}| \sigma,$$

$$\hat{\sigma}_{\theta_p} = |d^{pp}| \sqrt{s(\hat{\theta})/(n-p)},$$

k_α is the $100(1 - \frac{1}{2}\alpha)$ percentage point of a standard normal distribution, and $t_\alpha(n-p)$ is the $100(1 - \frac{1}{2}\alpha)$ percentage point of a t-distribution with $n-p$ degrees of freedom.

For the case when the θ_i are not non-negative, the terms of order higher than two in (2.4.4) are not yet found. However if we are satisfied with the second order approximations of γ , then a region estimate of $\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p$ is seen to be the set of feasible values of these components which lie in the following region:

$$(2.4.22) \quad \sum_{i=p-k^*+1}^p \left\{ \sum_{j=i}^p d_{ij} [(\theta_j - \hat{\theta}_j) - \sum_{l=p-k^*+1}^p \sum_{m=p-k^*+1}^p g_{jlm}^{(k^*)} (\theta_l - \hat{\theta}_l) (\theta_m - \hat{\theta}_m)] \right\}^2 \leq \sigma^2 \chi_{k^*, \alpha}^2 \quad \text{if } \sigma^2 \text{ is known,}$$

or

$$(2.4.23) \quad \sum_{i=p-k^*+1}^p \left\{ \sum_{j=i}^p d_{ij} [(\theta_j - \hat{\theta}_j) - \sum_{l=p-k^*+1}^p \sum_{m=p-k^*+1}^p g_{jlm}^{(k^*)} (\theta_l - \hat{\theta}_l) (\theta_m - \hat{\theta}_m)] \right\}^2 \leq \frac{k^*}{n-p} s(\hat{\theta}) F_{\alpha}(k^*, n-p) \quad \text{if } \sigma^2 \text{ is unknown,}$$

where

$\chi_{k^*, \alpha}^2$ is the $100(1-\alpha)$ percentage point of a χ^2 -distribution with k^* degrees of freedom

and

$F_{\alpha}(k^*, n-p)$ is the $100(1-\alpha)$ percentage point of an F-distribution with k^* and $n-p$ degrees of freedom.

Whether an interval or region estimate derived in this section will cover the true values of the corresponding components of θ with the nominal probability $(1-\alpha)$ depends on the adequacy of the approximations that the model is linear in the corresponding y .

In Chapter 5 we shall investigate the estimates in this section by means of some numerical examples.

Section 2.5 Region estimate of a different subset of components of the parameter vector

Suppose now we are interested in the region estimate of a different subset of k^* ($1 \leq k^* < p$) components $\theta_{i_1}, \theta_{i_2}, \dots, \theta_{i_{k^*}}$. For any component for which i_s does not satisfy $p-k^*+1 \leq i_s \leq p$, we interchange its position in the vector $(\theta_1, \theta_2, \dots, \theta_p)^T$ with another component θ_j (where $p-k^*+1 \leq j \leq p$) which are not of interest so that in the resulting vector, $\theta_{i_1}, \theta_{i_2}, \dots, \theta_{i_{k^*}}$ form the last k^* components, and the methods in section 2.4 for obtaining region estimate can be applied. A method for doing the above interchanging of the position of θ_{i_s} with θ_j will now be described.

Define

$$(2.5.1) \quad \underline{H}^{[i_s]} = \underline{I} - [\underline{y}^{[i_s]}][\underline{y}^{[i_s]}]^T$$

to be a Householder transformation such that

$$\underline{H}^{[i_s]} \begin{bmatrix} 0 \\ \vdots \\ 0 \\ d_{i_s} & i_s+1 \\ d_{i_s+1} & i_s+1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

is a column vector whose only nonzero entry is at i_s^{th} position.

Let $\underline{z}^{(i_s)}$ be given by

$$(2.5.2) \quad z_i^{(i_s)} = \begin{cases} \sum_{j=i}^p d_{ij} t_j + \sum_{j=1}^p \sum_{k=1}^p d_{ijk} t_j t_k + o(t^2), & (i = 1, 2, \dots, p) \\ \sum_{j=1}^p \sum_{k=1}^p d_{ijk} t_j t_k + o(t^2), & (i = p+1, p+2, \dots, n) \end{cases}$$

If we apply the transformation

$$\underline{H}^{[i_s]} \underline{z}^{(i_s)} = \underline{z}^{(i_s+1)}$$

of coordinates in sample space, then (2.5.2) becomes

$$(2.5.3) \quad z_i^{(i_s+1)} = \begin{cases} \sum_{j=i}^p d_{ij}^{(i_s+1)} t_j^{(i_s+1)} + \sum_{j=1}^p \sum_{k=1}^p d_{ijk}^{(i_s+1)} t_j^{(i_s+1)} t_k^{(i_s+1)} + o(t^2), & (i = 1, 2, \dots, p), \\ \sum_{j=1}^p \sum_{k=1}^p d_{ijk}^{(i_s+1)} t_j^{(i_s+1)} t_k^{(i_s+1)} + o(t^2), & (i = p+1, p+2, \dots, n), \end{cases}$$

where

$$(2.5.4) \quad t_j^{(i_s+1)} = \begin{cases} t_{i_s+1}, & j = i_s \\ t_{i_s}, & j = i_s + 1 \\ t_j, & \text{otherwise.} \end{cases}$$

We note that an effect of this transformation is to interchange the position of t_{i_s} with the next component t_{i_s+1} in $(t_1 t_2 \dots t_{i_s} t_{i_s+1} \dots t_p)$. We then repeat the above process until t_{i_s} is at the j^{th} position.

Section 2.6 Conditions for the existence of the power transformation which achieves maximum reduction of the nonlinearity associated with an individual parameter

In section 2.4 we have shown that the power transformation given by (2.4.18) achieves maximum reduction of the nonlinearity associated with the p^{th} parameter in the neighbourhood of a nonsingular point $P(\hat{\theta})$. And in section 2.5 we have discussed how the corresponding power transformation for reducing nonlinearity associated with the i^{th} parameter, where $i < p$, can be found. As for some models, e.g. models (A) and (B) in Chapter 1, the solution loci are bounded, it is of interest to investigate whether these power transformations for reducing the nonlinearity in the neighbourhood of $P(\hat{\theta})$ will exist as the parameter vector $\underline{\theta}$ tends to a value which may correspond to a point on a boundary of a solution locus.

Let $\{\underline{\theta}_m\}$ be a sequence of feasible values of the parameter vector converging to $\underline{\theta}_B$. It can be shown that the following conditions are a set of sufficient conditions for the existence of $\lambda_i = \lambda_i(\underline{\theta}_m)$ of the power transformation which achieves maximum reduction of the nonlinearity

associated with the i^{th} parameter in the neighbourhood of $P(\theta_m)$, as m tends to infinity:

- [1] the $c_{uk}(\theta_m)$ can be expressed as products of two functions of θ_m as follows:

$$c_{uk}(\theta_m) = c_{uk}^*(\theta_m) h_k(\theta_m), \quad (u = 1, 2, \dots, n; k = 1, 2, \dots, p),$$

where $c_{uk}^*(\theta_m)$ are functions with

$$\lim_{m \rightarrow \infty} c_{uk}^*(\theta_m) = c_{uk},$$

and the c_{uk} are finite numbers,

- [2] the $(n \times p)$ matrix $C = \{c_{uk}\}$ is of rank p ,

$$[3] \lim_{m \rightarrow \infty} \frac{c_{ujk}(\theta_m)}{h_j(\theta_m) h_k(\theta_m)} h_i(\theta_m) \theta_{mi} = q_{uijk}, \quad (u = 1, 2, \dots, n; j, k = 1, 2, \dots, p),$$

$$\text{where } c_{ujk}(\theta_m) = \frac{1}{2} \left[\frac{\partial^2 n(\xi_u, \theta)}{\partial \theta_j \partial \theta_k} \right]_{\theta=\theta_m},$$

θ_{mi} is the i^{th} component of θ_m ,

and the q_{uijk} are finite numbers.

To show that these are the sufficient conditions, we first let H_m be the $(n \times n)$ matrix such that $H_m C_m$, where $C_m = \{c_{uk}(\theta_m)\}$, is a $(p \times p)$ upper triangular matrix $D_m = \{d_{ik}(\theta_m)\}$ with an $[(n-p) \times p]$ zero matrix beneath it (c.f. (2.2.2)). H_m can be written as

$$H_m = \prod_{j=1}^p (I - [v^{(p+1-j)}(\theta_m)] [v^{(p+1-j)}(\theta_m)]^T),$$

where $\underline{v}^{(j)}(\theta_m)$ are $(nx1)$ vectors whose u^{th} components are $v_u^{(j)}(\theta_m)$.

We note that (c.f. Appendix 1)

$$v_1^{(1)}(\theta_m) = \sqrt{1 + \frac{c_{11}(\theta_m)}{r_1(\theta_m)}}$$

and

$$v_u^{(1)}(\theta_m) = \frac{c_{ul}(\theta_m)}{v_1^{(1)}(\theta_m)r_1(\theta_m)}, \quad (u = 2, 3, \dots, n),$$

where

$$r_1(\theta_m) = \sqrt{\sum_{u=1}^n [c_{ul}(\theta_m)]^2}$$

is chosen to have the same sign as $c_{11}(\theta_m)$. Then because of condition [1], we have

$$(2.6.1) \quad \lim_{m \rightarrow \infty} v_1^{(1)}(\theta_m) = \sqrt{1 + \frac{c_{11}}{r_1}}$$

and

$$(2.6.2) \quad \lim_{m \rightarrow \infty} v_u^{(1)}(\theta_m) = \frac{c_{ul}}{\sqrt{1 + \frac{c_{11}}{r_1} r_1}}, \quad (u = 2, 3, \dots, n),$$

where

$$r_1 = \sqrt{\sum_{u=1}^n c_{ul}^2}.$$

$r_1 \neq 0$ because C is of rank p . Therefore the limits in (2.6.1) and (2.6.2) are finite. Similarly because of conditions [1] and [2],

$\lim_{m \rightarrow \infty} v_u^{(j)}(\theta_m)$ for $j = 2, 3, \dots, p$ and $u = 1, 2, \dots, n$ are finite.

Next we have

$$d_{11}(\theta_m) = c_{11}(\theta_m) - v_1^{(1)}(\theta_m) \sum_{u=1}^n c_{u1}(\theta_m) v_u^{(1)}(\theta_m).$$

This together with (2.6.1), (2.6.2) and condition [1] imply that $d_{11}(\theta_m)$ can be expressed in the form

$$(2.6.3) \quad d_{11}(\theta_m) = d_{11}^*(\theta_m) h_1(\theta_m),$$

where $d_{11}(\theta_m)$ is a function with

$$(2.6.4) \quad \lim_{m \rightarrow \infty} d_{11}^*(\theta_m) = d_{11}$$

and d_{11} is a finite number. Similarly it can be shown that $d_{ik}(\theta_m)$ can be expressed in the form

$$(2.6.5) \quad d_{ik}(\theta_m) = d_{ik}^*(\theta_m) h_k(\theta_m), \quad (i = 1, 2, \dots, p, k = 2, 3, \dots, p, \text{ and } i \leq k),$$

where $d_{ik}^*(\theta_m)$ are functions with

$$(2.6.6) \quad \lim_{m \rightarrow \infty} d_{ik}^*(\theta_m) = d_{ik}$$

and the d_{ik} are finite numbers. Further because of [2], D is nonsingular and consequently $d_{ii} \neq 0$ for all $i = 1, 2, \dots, p$.

Now for any θ_m such that C_m is of rank p , D_m is non-singular and its inverse $D_m^{-1} = \{d_{ik}(\theta_m)\}$ is given by

$$d_{ii}^{ii}(\theta_m) = \frac{1}{d_{ii}(\theta_m)}, \quad (i = 1, 2, \dots, p),$$

and

$$d_{ik}^{ik}(\theta_m) = -\frac{1}{d_{ii}(\theta_m)} \sum_{j=i}^k d_{ij}(\theta_m) d_{jk}^{jk}(\theta_m), \quad (i = 1, 2, \dots, p-1, k = 2, 3, \dots, p \\ \text{and } i < k).$$

Then because of (2.6.3) and (2.6.4), $d_{ik}^{ik}(\theta_m)$ can be expressed in the form

$$(2.6.7) \quad d_{ik}^{ik}(\theta_m) = d_{ik}^+(\theta_m) [h_i(\theta_m)]^{-1}, \quad (i, k = 1, 2, \dots, p),$$

where $d_{ik}^+(\theta_m)$ are functions with

$$(2.6.8) \quad \lim_{m \rightarrow \infty} d_{ik}^+(\theta_m) = s_{ik}$$

and the s_{ik} are finite numbers.

Define

$$\tilde{d}_{jk}(\theta_m) = H_m c_{jk}(\theta_m), \quad (j, k = 1, 2, \dots, p)$$

where $c_{jk}(\theta_m)$ is the $(n \times 1)$ vector whose u^{th} component is $c_{ujk}(\theta_m)$.

Then because $\lim_{m \rightarrow \infty} v_u^{(j)}(\theta_m)$ are finite, the i^{th} component of $\tilde{d}_{jk}(\theta_m)$ can be expressed in the form

$$(2.6.9) \quad d_{ijk}(\theta_m) = \sum_{u=1}^n v_{ijk}^{(u)}(\theta_m) c_{ujk}(\theta_m),$$

where $v_{ijk}^{(u)}(\theta_m)$ are functions with

$$(2.6.10) \quad \lim_{m \rightarrow \infty} v_{ijk}(\theta_m) = v_{ijk}$$

and the v_{ijk} are finite numbers.

Applying the method in section 2.4 for finding λ_p at θ to find the corresponding $\lambda_p(\theta_m)$ at θ_m for which C_m is of rank p , we obtain

$$(2.6.11) \quad \begin{aligned} \lambda_p(\theta_m) &= 1 + \frac{2}{3} d_{pp}(\theta_m) \left[\sum_{\ell=1}^{p-1} \sum_{j=1}^{\ell} \sum_{k=1}^{\ell} d_{pjk}(\theta_m) d^{j\ell}(\theta_m) d^{k\ell}(\theta_m) \right. \\ &\quad \left. + 3 \sum_{j=1}^p \sum_{k=1}^p d_{pjk}(\theta_m) d^{jp}(\theta_m) d^{kp}(\theta_m) \right] \theta_{mp}. \end{aligned}$$

Then by using (2.6.3)-(2.6.10), we obtain

$$(2.6.12) \quad \begin{aligned} \lim_{m \rightarrow \infty} \lambda_p(\theta_m) &= 1 + \frac{2}{3} d_{pp} \lim_{m \rightarrow \infty} \left\{ h_p(\theta_m) \left[\sum_{\ell=1}^{p-1} \sum_{j=1}^{\ell} \sum_{k=1}^{\ell} \sum_{u=1}^n v_{pjk} c_{ujk}(\theta_m) \right. \right. \\ &\quad \times \frac{s_{jl}}{h_j(\theta_m)} \frac{s_{kl}}{h_k(\theta_m)} \theta_{mp} + 3 \sum_{j=1}^p \sum_{k=1}^p \sum_{u=1}^n v_{pjk} c_{ujk}(\theta_m) \frac{s_{jp}}{h_j(\theta_m)} \frac{s_{kp}}{h_k(\theta_m)} \theta_{mp} \left. \right] \right\}. \end{aligned}$$

From (2.6.12) we see that $\lim_{m \rightarrow \infty} \lambda_p(\theta_m)$ is finite if

$$\lim_{m \rightarrow \infty} \frac{c_{ujk}(\theta_m)}{h_j(\theta_m) h_k(\theta_m)} h_p(\theta_m) \theta_{mp}$$

is finite for $u = 1, 2, \dots, n$ and $j, k = 1, 2, \dots, p$.

Similarly $\lim_{m \rightarrow \infty} \lambda_i(\theta_m)$, where $i < p$, is finite if

$$\lim_{m \rightarrow \infty} \frac{c_{ujk}(\theta_m)}{h_j(\theta_m) h_k(\theta_m)} h_i(\theta_m) \theta_{mi}$$

is finite for $u = 1, 2, \dots, n$ and $j, k = 1, 2, \dots, p$.

We now show that in models (A) and (B), the sufficient conditions for the existence of the power transformation which achieves maximum reduction of the nonlinearity associated with the first parameter are satisfied when θ_1 tends to infinity. By differentiating the $n(\xi_u, \theta)$ of these models with respect to θ_1 and θ_2 , we see that [1] and [2] are satisfied. Next we can show that for model (A),

$$h_1(\theta_m) = \frac{1}{\theta_{ml}^2},$$

$$h_2(\theta_m) = 1,$$

$$c_{ull}(\theta_m) = \frac{v_{ull}(\theta_m)}{\theta_{ml}^3},$$

$$c_{ul2}(\theta_m) = \frac{v_{ul2}(\theta_m)}{\theta_{ml}^2},$$

$$c_{u22}(\theta_m) = v_{u22}(\theta_m),$$

where $v_{uij}(\theta_m)$ are functions which converge to finite limits as m tends to infinity, and for model (B),

$$h_1(\theta_m) = \frac{1}{\theta_{ml}^2},$$

$$h_2(\theta_m) = 1,$$

$$c_{ull}(\theta_m) = \frac{w_{ull}(\theta_m)}{\theta_{ml}^3},$$

$$c_{u12}(\theta_m) = \frac{w_{u12}(\theta_m)}{\theta_{ml}^2},$$

$$c_{u22}(\theta_m) = w_{u22}(\theta_m),$$

where $w_{uij}(\theta_m)$ are functions which converge to finite limits as m tends to infinity. Condition [3] can then be readily shown to be satisfied in each of these models.

For model (B), the matrix C for the case when θ_B is such that $\theta_{B1} - \theta_{B2} = 0$ is of rank one, which is less than p . However, if we let

$$\begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \end{bmatrix},$$

then the corresponding sufficient conditions for the existence of the power transformation of β_2 are satisfied. This indicates that whenever [2] is not satisfied, there may exist a linear transformation $\beta = L\theta$, where L is a (pxp) non-singular matrix, such that the corresponding sufficient conditions for the existence of the power transformation of β_i are satisfied.

Section 2.7 Alternative transformations to reduce nonlinearity

In many situations, it will be much easier to appreciate the physical significance of transformations of individual parameters rather than transformations such that the new parameters are functions of more than one original parameter. We therefore investigate transformation β of the form

$$\beta_i = \beta_i(t_i), \quad (i = 1, 2, \dots, p).$$

We choose $\beta_i = \beta_i(t_i)$ to be one such that the total nonlinearity is reduced as much as possible. For reasons similar to those given in the derivation of N_ϕ in section 2.2, we can restrict our attention to β_i of the form

$$(2.7.1) \quad \beta_i = t_i - g_i t_i^2 + o(t_i^2), \quad (i = 1, 2, \dots, p).$$

Suppose g_i^* are such that the corresponding β achieves the maximum reduction of the total nonlinearity. To find g_i^* , we first apply the orthogonal transformation H (c.f. (2.2.2)) of coordinates in sample space and then use the linear transformation $\xi^{(\beta)} = D\beta$ to obtain

$$(2.7.2) \quad N_\beta = \frac{\sigma^2}{p+2} \sum_{i=1}^n \sum_{\ell=1}^p \sum_{m=1}^p [f_{ilm}^+ f_{imm}^+ + 2(f_{ilm}^+)^2]$$

where

$$(2.7.3) \quad f_{ilm}^+ = \begin{cases} f_{ilm} + \sum_{i \leq j \leq \min[\ell, m]} d_{ij} g_j d^{j\ell} d^{jm}, & (i = 1, 2, \dots, p) \\ f_{ilm} & , \quad (i = p+1, p+2, \dots, n). \end{cases}$$

By differentiating (2.7.2) with respect to the g_j , it can be shown that the g_j^* are given by

$$(2.7.4) \quad \sum_{j=1}^p \left[\sum_{i=1}^{\min[j, s]} d_{is} \left[\left(\sum_{k=j}^p d_{ik} (d^{jk})^2 \right) \left(\sum_{\ell=s}^p (d^{s\ell})^2 \right) \right. \right. \\ \left. \left. + 2 d_{ij} \left(\sum_{\ell=\max[j, s]}^p d^{j\ell} d^{s\ell} \right)^2 \right] \right] g_j^*$$

$$+ \sum_{i=1}^s d_{is} [(\sum_{l=1}^p f_{ill}) (\sum_{k=s}^p (d^{sk})^2) + 2 \sum_{l=s}^p \sum_{m=s}^p f_{ilm} d^{sl} d^{sm}] = 0$$

$$(s = 1, 2, \dots, p).$$

For the case when the θ_i are non-negative, the power transformations

$$(2.7.5) \quad \psi_i = \begin{cases} \frac{\theta_i^{\lambda_i} - 1}{\lambda_i}, & \lambda_i \neq 0, \\ \ln \theta_i, & \lambda_i = 0 \end{cases}, \quad (i = 1, 2, \dots, p),$$

where $\lambda_i = 1 - 2g_i^* \hat{\theta}_i$, and the transformations β_i given by (2.7.1) with g_i changed to g_i^* are equivalent in the sense that they reduce the total nonlinearity by the same amount.

Interval and region estimates based on these power transformations are investigated numerically in Chapter 5.

Section 2.8 Effects of design of experiments on nonlinearity

This section is concerned with the effects of design of experiments on nonlinearity.

It is geometrically obvious that if the number, n , of observations is less than or equal to the number, p , of components of the parameter vector θ , then the solution locus of the model is a subset of an n -dimensional linear manifold. If n is larger than p , but the number, s , of distinct experimental conditions adopted in the model is less than or equal to p , then the solution locus can be shown as follows to be a subset of an s -dimensional linear manifold.

Let $y_{u_1 u_2}$ be the observed value of the u_2^{th} of r_{u_1} responses in the u_1^{th} of s ($s \leq p$) distinct experiments. We then have

$$(2.8.1) \quad E(y_{u_1 u_2}) = \eta(\xi_{u_1}, \theta), \quad (u_1 = 1, 2, \dots, s; \quad u_2 = 1, 2, \dots, r_{u_1};$$

$$\sum_{u_1=1}^s r_{u_1} = n).$$

Let \underline{H}_k and \underline{H}^+ be the $(k \times k)$ and $(n \times n)$ orthogonal matrices given by

$$(2.8.2) \quad \underline{H}_k = \begin{bmatrix} \frac{1}{\sqrt{k}} & \frac{1}{\sqrt{k}} & \frac{1}{\sqrt{k}} & \cdots & \frac{1}{\sqrt{k}} & \frac{1}{\sqrt{k}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 & \cdots & 0 & 0 \\ \frac{1}{\sqrt{2 \times 3}} & \frac{1}{\sqrt{2 \times 3}} & -\frac{2}{\sqrt{2 \times 3}} & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ \frac{1}{\sqrt{(k-1)k}} & \frac{1}{\sqrt{(k-1)k}} & \frac{1}{\sqrt{(k-1)k}} & \cdots & \frac{1}{\sqrt{(k-1)k}} & \frac{-(k-1)}{\sqrt{(k-1)k}} \end{bmatrix}$$

and

$$(2.8.3) \quad \underline{H}^+ = \begin{bmatrix} \underline{H}_{r_1} & & & \\ & \ddots & & \\ & & \underline{H}_{r_2} & \\ & & & \ddots \\ & & & & \ddots & \underline{H}_{r_s} \end{bmatrix}.$$

Further, let y^* be the $(n \times 1)$ vector given by

$$y^* = [y_{11} y_{12} \cdots y_{1r_1} : y_{21} y_{22} \cdots y_{2r_2} : \cdots : y_{s1} y_{s2} \cdots y_{sr_s}]^T.$$

If we apply the orthogonal transformation

$$(2.8.4) \quad \underline{H}^+ \underline{y}^* = \underline{y}^+$$

of coordinates in sample space, then (2.8.1) becomes

$$(2.8.5) \quad E(y_{u_1 u_2}^+) = \begin{cases} \sqrt{r_{u_1}} \eta(\xi_{u_1}, \hat{\theta}), & (u_1 = 1, 2, \dots, s; u_2 = 1) \\ 0, & (u_1 = 1, 2, \dots, s; u_2 = 2, 3, \dots, r_{u_1}) \end{cases}$$

where $y_{u_1 u_2}^+$ is the $(\sum_{i=1}^{u_1} r_i - r_{u_1} + u_2)^{\text{th}}$ component of \underline{y}^+ .

From (2.8.5), it is clear that the solution locus of the model is a subset of an s -dimensional linear manifold.

We next investigate the effects of replication of experiments on the measures of nonlinearity in the neighbourhood of a non-singular point $P(\hat{\theta})$.

Let s, r_{u_1} and n be changed to n, r and nr respectively. We have

$$(2.8.6) \quad E(y_{u_1 u_2}^+) = n(\xi_{u_1}, \hat{\theta}) + \sum_{j=1}^P c_{u_1 j} t_j + \sum_{j=1}^P \sum_{k=1}^P c_{u_1 j k} t_j t_k + o(t^2),$$

$$(u_1 = 1, 2, \dots, n; u_2 = 1, 2, \dots, r).$$

Let η^* be the $(nr \times 1)$ vector whose $((u_1 - 1)r + u_2)^{\text{th}}$ component is $n(\xi_{u_1}, \hat{\theta})$.

After applying the orthogonal transformation

$$\underline{H}^+ (\underline{y}^* - \eta^*) = \underline{y}^{(r+)}$$

of coordinates in sample space, (2.8.6) becomes

$$(2.8.7) \quad E(y_{u_1 u_2}^{(r+)}) = \begin{cases} \sum_{j=1}^p \sqrt{r} c_{u_1 j} t_j + \sum_{j=1}^p \sum_{k=1}^p \sqrt{r} c_{u_1 j k} t_j t_k + o(t^2), & (u_1 = 1, 2, \dots, n; u_2 = 1) \\ 0 & , (u_1 = 1, 2, \dots, n; u_2 = 2, 3, \dots, r), \end{cases}$$

where $y_{u_1 u_2}^{(r+)}$ is the $((u_1-1)r + u_2)^{\text{th}}$ component of $\underline{y}^{(r+)}$.

Let $I_{n(r-1)}$ be an $(n(r-1) \times n(r-1))$ identity matrix and $\underline{y}^{(r)}$ be the $(nr \times 1)$ vector given by

$$\underline{y}^{(r)} = [y_{11}^{(r+)} \ y_{21}^{(r+)} \ \dots \ y_{n1}^{(r+)} \ 0 \ 0 \ \dots \ 0]^T.$$

If we apply the orthogonal transformation

$$(2.8.8) \quad \left[\begin{array}{c|c} H & \text{ } \\ \hline & I_{n(r-1)} \end{array} \right] \underline{y}^{(r)} = \underline{z}^* \quad (\text{c.f. (2.2.2)})$$

of coordinates in sample space, then (2.8.7) becomes

$$(2.8.9) \quad E(z_{i_1 i_2}^{*}) = \begin{cases} \sum_{j=1}^p \sqrt{r} d_{i_1 j} t_j + \sum_{j=1}^p \sum_{k=1}^p \sqrt{r} d_{i_1 j k} t_j t_k + o(t^2), & (i_1 = 1, 2, \dots, p; \\ & \quad i_2 = 1) \\ \sum_{j=1}^p \sum_{k=1}^p \sqrt{r} d_{i_1 j k} t_j t_k + o(t^2), & (i_1 = p+1, p+2, \dots, n; i_2 = 1) \\ 0 & , (i_1 = 1, 2, \dots, n; i_2 = 2, 3, \dots, r), \end{cases}$$

where $z_{i_1 i_2}^{*}$ is the $(i_1 + (i_2-1)n)^{\text{th}}$ component of \underline{z}^* .

After applying the linear transformation

$$(2.8.10) \quad \underline{\tau}^{(r)} = \sqrt{r} \underline{\tau},$$

(2.8.9) becomes

$$(2.8.11) \quad E(z_{i_1 i_2}) = \begin{cases} \tau_{i_1}^{(r)} + \sum_{\ell=1}^p \sum_{m=1}^p \frac{1}{\sqrt{r}} f_{i_1 \ell m} \tau_{\ell}^{(r)} \tau_m^{(r)} + o((\tau^{(r)})^2), & (i_1 = 1, 2, \dots, p; i_2 = 1) \\ \sum_{\ell=1}^p \sum_{m=1}^p \frac{1}{\sqrt{r}} f_{i_1 \ell m} \tau_{\ell}^{(r)} \tau_m^{(r)} + o((\tau^{(r)})^2), & (i_1 = p+1, p+2, \dots, n; i_2 = 1) \\ \vdots & \\ 0 & (i_1 = 1, 2, \dots, n; i_2 = 2, 3, \dots, r). \end{cases}$$

We then have

$$(2.8.12) \quad N_\theta = \frac{1}{r} \left[\frac{\sigma^2}{p+2} \sum_{i_1=1}^n \sum_{\ell=1}^p \sum_{m=1}^p (f_{i_1 \ell \ell} f_{i_1 m m} + 2f_{i_1 \ell m}^2) \right]$$

and

$$(2.8.13) \quad N_\phi = \frac{1}{r} \left[\frac{\sigma^2}{p+2} \sum_{i_1=p+1}^n \sum_{\ell=1}^p \sum_{m=1}^p (f_{i_1 \ell \ell} f_{i_1 m m} + 2f_{i_1 \ell m}^2) \right].$$

Therefore replication of each of the experiments r times shrinks the measures of total nonlinearity and intrinsic nonlinearity in the neighbourhood of $P(\hat{\theta})$ by a factor of r .

We next see from (2.8.10) and (2.8.11) that the values of $g_{ijk}^{(k*)}$ in section 2.4 and $g_i^{(*)}$ in section 2.7 do not change due to replication of each of the experiments r times.

CHAPTER 3

HYPOTHESIS TESTING AND REGION ESTIMATION BASED
 ON GENERAL MAXIMUM LIKELIHOOD RATIOS

Section 3.1 Introduction

In this chapter, we consider the problem of testing a number of hypotheses for making inference about subsets of components of the parameter vector in nonlinear models. General maximum likelihood (m.l.) ratio tests are used for testing these hypotheses. The estimation of the coverage probabilities of the region estimates based on these tests is also discussed.

Section 3.2 Hypothesis testing in unconstrained nonlinear models

We shall restrict our attention to unconstrained nonlinear models and consider the problem of testing the null hypothesis

H_i that $(\theta, \sigma) \in \Omega_{H_i}$

against the alternative hypothesis

K_i that $(\theta, \sigma) \in \Omega_{K_i}$, ($i = 1, 2, \dots, 5$),

where

$$\Omega_{H_1} = \{(\theta, \sigma): \theta_j = \theta_{0j} \text{ for } j = p-k^*+1, p-k^*+2, \dots, p, \text{ and } \sigma = \sigma_0\}, \\ (k^* = 1, 2, \dots, p-1),$$

$$\Omega_{K_1} = \{(\underline{\theta}, \sigma) : \theta_j \neq \theta_{0j} \text{ for } j = p-k^*+1, p-k^*+2, \dots, p, \text{ and } \sigma = \sigma_0\},$$

$$\Omega_{H_2} = \{(\underline{\theta}, \sigma) : \underline{\theta} = \underline{\theta}_0, \text{ and } \sigma = \sigma_0\},$$

$$\Omega_{K_2} = \{(\underline{\theta}, \sigma) : \underline{\theta} \neq \underline{\theta}_0, \text{ and } \sigma = \sigma_0\},$$

$$\Omega_{H_3} = \{(\underline{\theta}, \sigma) : \theta_j = \theta_{0j} \text{ for } j = p-k^*+1, p-k^*+2, \dots, p\},$$

$$\Omega_{K_3} = \{(\underline{\theta}, \sigma) : \theta_j \neq \theta_{0j} \text{ for } j = p-k^*+1, p-k^*+2, \dots, p\},$$

$$\Omega_{H_4} = \{(\underline{\theta}, \sigma) : \underline{\theta} = \underline{\theta}_0\},$$

$$\Omega_{K_4} = \{(\underline{\theta}, \sigma) : \underline{\theta} \neq \underline{\theta}_0\},$$

$$\Omega_{H_5} = \{(\underline{\theta}, \sigma) : \sigma = \sigma_0\},$$

$$\Omega_{K_5} = \{(\underline{\theta}, \sigma) : \sigma \neq \sigma_0\},$$

$\underline{\theta}_0$ and σ_0 are particular values of $\underline{\theta}$ and σ respectively, and θ_{0j} is the j^{th} component of $\underline{\theta}_0$.

The usual monotonic functions of the general m.l. ratios for testing these hypotheses are given respectively by

$$T_1(\underline{y}) = [s^M(\theta_{0p-k^*+1}, \theta_{0p-k^*+2}, \dots, \theta_{0p}) - s(\hat{\theta})],$$

$$T_2(\underline{y}) = [s(\underline{\theta}_0) - s(\hat{\theta})],$$

$$T_3(\underline{y}) = [s^M(\theta_{0p-k^*+1}, \theta_{0p-k^*+2}, \dots, \theta_{0p}) - s(\hat{\theta})]/s(\hat{\theta}),$$

$$T_4(\underline{y}) = [s(\underline{\theta}_0) - s(\hat{\theta})]/s(\hat{\theta})$$

and

$$T_5(\underline{y}) = s(\hat{\theta}),$$

$$\text{where } S(\theta) = \sum_{u=1}^n \{y_u - n(\xi_u, \theta)\}^2,$$

and $S^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op})$ is the minimum value of $S(\theta)$ with respect to θ where θ are such that $(\theta, \sigma_0) \in \Omega_{H_1}$.

Acceptance regions of the general m.l. ratio tests are

$$\omega_i = \{z : s_i^D(z) \leq d_i^* \}, \quad (i = 1, 2, \dots, 5),$$

where

$$s_1^D(z) = [S^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}) - S(\hat{\theta})]/\sigma_0^2,$$

$$s_2^D(z) = [S(\theta_0) - S(\hat{\theta})]/\sigma_0^2,$$

$$s_3^D(z) = [S^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}) - (1 + \frac{k^* F_\alpha(k^*, n-p)}{n-p}) S(\hat{\theta})]/\sigma_0^2$$

$$+ \frac{k^* F_\alpha(k^*, n-p)}{n-p} \sum_{j=p+1}^n z_j^2/\sigma_0^2,$$

$$s_4^D(z) = [S(\theta_0) - (1 + \frac{p F_\alpha(p, n-p)}{n-p}) S(\hat{\theta})]/\sigma_0^2 + \frac{p F_\alpha(p, n-p)}{n-p} \sum_{j=p+1}^n z_j^2/\sigma_0^2,$$

$$s_5^D(z) = S(\hat{\theta})/\sigma_0^2,$$

$$(\theta_0, \sigma_0) \in \Omega_{H_i}, \quad (i = 1, 2, 3, 4, 5),$$

$$d_i^{*2} = (d_i^*)^2 = \begin{cases} \chi_{k^*, \alpha}^2 & , (i = 1) \\ \chi_{p, \alpha}^2 & , (i = 2) \\ \frac{k^* F_\alpha(k^*, n-p)}{n-p} \sum_{j=p+1}^n z_j^2 / \sigma_0^2 & , (i = 3) \\ \frac{p F_\alpha(p, n-p)}{n-p} \sum_{j=p+1}^n z_j^2 / \sigma_0^2 & , (i = 4) \\ \chi_{n-p, \alpha}^2 & , (i = 5), \end{cases}$$

and the z_i are rotated coordinates of a point \underline{z} in sample space with $P(\theta_0)$ as origin (c.f. section 2.2).

A number which is not larger than the probability

$$\Gamma_i(\theta_0, \sigma_0) = \Pr\{\underline{z} \in \omega_i \mid \theta = \theta_0 \text{ and } \sigma = \sigma_0\}$$

for all $(\theta_0, \sigma_0) \in \Omega_{H_i}$ is the significance probability of the test for the hypothesis H_i , and the probability

$$\beta_i(\theta_A, \sigma_A) = 1 - \Pr\{\underline{z} \in \omega_i \mid \theta = \theta_A \text{ and } \sigma = \sigma_A\},$$

where $(\theta_A, \sigma_A) \in \Omega_{K_i}$, is the power function of this test.

Note that we have chosen $\chi_{k^*, \alpha}^2$, $\chi_{p, \alpha}^2$, $F_\alpha(k^*, n-p)$, $F_\alpha(p, n-p)$ and $\chi_{n-p, \alpha}^2$ to be the corresponding constants appearing in the expressions for d_i^{*2} . A reason for choosing them is that if the model is linear, then α is the level of significance of each of the tests.

Section 3.3 Significance probabilities of the general m.l. ratio tests

In this section we restrict our attention to the $\eta(\xi_u, \theta)$ which are functions of θ differentiable up to the third order and consider the derivation of approximations of the probabilities $I_i(\theta_0, \sigma_0)$ (c.f. section 3.2), where θ_0 is such that $P(\theta_0)$ is a non-singular point in the solution locus.

The functions $\eta(\xi_u, \theta)$ can be written as

$$(3.3.1) \quad \eta(\xi_u, \theta) = \eta(\xi_u, \theta_0) + \sum_{j=1}^p c_{uj} t_j + \tilde{t}_u^T C_u \tilde{t}_u + \sum_{j=1}^p [t_u^T C_u t_u] t_j + o(t^3), \quad (u = 1, 2, \dots, n),$$

where

\tilde{t} is a $(p \times 1)$ vector whose j^{th} component is $t_j = \theta_j - \theta_{0j}$,

C_u is a $(p \times p)$ symmetric matrix whose (j, k) entry is c_{ujk}

C_{uj} is a $(p \times p)$ symmetric matrix whose (k, l) entry is c_{ujkl}

and

$t^3 = (\sum_{j=1}^p t_j^2)^{3/2}$. Though t_j , c_{uj} , c_{ujk} are also used to denote

the corresponding terms that arise in expanding $\eta(\xi_u, \theta)$ in the neighbourhood of $\hat{\theta}$ in Chapter 2, we hope that no confusion should arise.

Let η_0 be the $(n \times 1)$ vector whose u^{th} component is $\eta(\xi_u, \theta_0)$. Further, let H (c.f. section 2.2) be an $(n \times n)$ orthogonal matrix such that HC , where $C = \{c_{uj}\}$, is an upper triangular $(p \times p)$ matrix D with an $((n-p) \times p)$ zero matrix beneath it.

As in section 2.2, we can apply the orthogonal transformation

$$H(y - \eta_0) = z$$

in sample space and the linear transformation

$$\underline{\tau} = D\underline{t}$$

of parameter vector \underline{t} . After these transformations, the rotated coordinates z_i^* of a point $P(\theta)$ in the solution locus can be written as

$$(3.3.2) \quad z_i^* = \begin{cases} \tau_i + \underline{\tau}^T F_i \underline{\tau} + \sum_{j=1}^p [\underline{\tau}^T F_{ij} \underline{\tau}] \tau_j + o(\tau^3), & (i = 1, 2, \dots, p) \\ \underline{\tau}^T F_i \underline{\tau} + \sum_{j=1}^p [\underline{\tau}^T F_{ij} \underline{\tau}] \tau_j + o(\tau^3), & (i = p+1, p+2, \dots, n) \end{cases}$$

where

$$\tau^3 = \left(\sum_{i=1}^p \tau_i^2 \right)^{3/2},$$

and

$$F_i = \{f_{ijk}\},$$

$$F_{ij} = \{f_{ijkl}\}$$

are $(p \times p)$ symmetric matrices.

After applying the nonlinear transformations

$$(3.3.3) \quad \phi_i = \tau_i + \underline{\tau}^T F_i \underline{\tau} + \sum_{j=1}^p [\underline{\tau}^T F_{ij} \underline{\tau}] \tau_j + o(\tau^3), \quad (i = 1, 2, \dots, p),$$

of the parameters τ_i , (3.3.2) becomes

$$(3.3.4) \quad z_i^* = \begin{cases} \phi_i & , \quad (i = 1, 2, \dots, p) \\ \phi^T A_i \phi + \sum_{j=1}^p [\phi^T A_{ij} \phi] \phi_j + o(\phi^3), & (i = p+1, p+2, \dots, n), \end{cases}$$

where $A_{ij} = F_{ij}$,

$$A_{ij} = F_{ij} - 2 \sum_{k=1}^p f_{ijk} F_k$$

$$\text{and } \phi^3 = \left(\sum_{i=1}^p \phi_i^2 \right)^{3/2}.$$

The sum of squares of the residuals of an observed sample point z is given by

$$(3.3.5) \quad S = \sum_{i=1}^p (z_i - \phi_i)^2 + \sum_{i=p+1}^n \{ z_i - \phi_i^T A_i \phi - \sum_{j=1}^p [\phi_j^T A_{ij} \phi] \phi_j + o(\phi^3) \}^2.$$

Approximations of the least squares estimates $\hat{\phi}_m$ for the components ϕ_m of the parameter vector ϕ can be obtained by minimizing S in (3.3.5) with respect to the ϕ_i , and it is found that

$$(3.3.6) \quad \begin{aligned} \hat{\phi}_m &= z_m + 2 \sum_{i=p+1}^n \sum_{j=1}^p a_{ijm} z_i z_j \\ &+ 4 \sum_{i=p+1}^n \sum_{h=p+1}^n \sum_{j=1}^p \sum_{k=1}^p a_{ijm} a_{hjk} z_i z_h z_k \\ &- 2 \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p a_{ijk} a_{ilm} z_j z_k z_l \\ &+ 3 \sum_{i=p+1}^n \sum_{j=1}^p \sum_{l=1}^p a_{ijml} z_i z_j z_l + o(z^3), \quad (m = 1, 2, \dots, p), \end{aligned}$$

where $z^3 = \left(\sum_{i=1}^p z_i^2 \right)^{3/2}$. The corresponding sum of squares of residuals is

$$\begin{aligned}
 (3.3.7) \quad S(\hat{\theta}) = & \sum_{j=p+1}^n z_j^2 - 2 \sum_{i=p+1}^n \sum_{j=1}^n \sum_{k=1}^p \sum_{l=1}^p a_{ijk} z_i z_j z_k \\
 & + \sum_{i=p+1}^n \sum_{j=1}^n \sum_{k=1}^p \sum_{l=1}^p \sum_{m=1}^p a_{ijk} a_{ilm} z_j z_k z_l z_m \\
 & - 4 \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j=1}^n \sum_{k=1}^p \sum_{l=1}^p a_{hjl} a_{ikl} z_h z_i z_j z_k \\
 & - 2 \sum_{i=p+1}^n \sum_{j=1}^n \sum_{k=1}^p \sum_{l=1}^p a_{ijk} z_i z_j z_k z_l + o(z^4).
 \end{aligned}$$

Under the hypothesis H_1 or H_3 , the components τ_j ($j = p-k^*+1, p-k^*+2, \dots, p$) are zero, and consequently (3.3.2) becomes

$$(3.3.8) \quad z_i^* = \begin{cases} \tau_i + \tau_M^T F_{Mi} \tau_M + \sum_{j=1}^{p-k^*} [\tau_{M-Mij}^T \tau_M] \tau_j + o(\tau_M^3), & (i = 1, 2, \dots, p-k^*) \\ \tau_M^T F_{Mi} \tau_M + \sum_{j=1}^{p-k^*} [\tau_{M-Mij}^T \tau_M] \tau_j + o(\tau_M^3), & (i = p-k^*+1, p-k^*+2, \dots, n) \end{cases}$$

where

$$\tau_M = [\tau_1 \tau_2 \dots \tau_{p-k^*}]^T,$$

$$\tau_M^3 = (\sum_{i=1}^{p-k^*} \tau_i^2)^{3/2},$$

and

$$F_{Mi} = \{f_{ijk}\},$$

$$F_{Mij} = \{f_{ijkl}\}$$

are $((p-k^*) \times (p-k^*))$ symmetric matrices.

After applying the nonlinear transformations

$$\phi_{Mi} = \tau_i + \tau_M^T F_{Mi} \tau_M + \sum_{j=1}^{p-k^*} [\tau_M^T F_{Mij} \tau_M] \tau_j + o(\tau_M^3), \quad (i = 1, 2, \dots, p-k^*)$$

of the parameters τ_i , (3.3.8) becomes

$$(3.3.9) \quad z_i^* = \begin{cases} \phi_{Mi} & , \quad (i = 1, 2, \dots, p-k^*) \\ \phi_M^T A_{Mi} \phi_M + \sum_{j=1}^{p-k^*} [\phi_M^T A_{Mij} \phi_M] \phi_{Mj} + o(\phi_M^3), & (i = p-k^*+1, p-k^*+2, \dots, n), \end{cases}$$

where

$$A_{Mi} = F_{Mi},$$

$$A_{Mij} = F_{Mij} - 2 \sum_{k=1}^{p-k^*} f_{ijk} F_{Mk}$$

and

$$\phi_M^3 = \left(\sum_{i=1}^{p-k^*} \phi_{Mi}^2 \right)^{3/2}.$$

For the same observed sample point z , we can find the values $\hat{\phi}_{Mm}$ of the ϕ_{Mm} such that the sum of squares

$$\sum_{i=1}^{p-k^*} (z_i - \phi_{Mi})^2 + \sum_{i=p-k^*+1}^n \{z_i - \phi_M^T A_{Mi} \phi_M - \sum_{j=1}^{p-k^*} [\phi_M^T A_{Mij} \phi_M] \phi_{Mj} + o(\phi_M^3)\}^2$$

is minimized, and it is found that

$$\begin{aligned}
 (3.3.10) \quad & \hat{\phi}_{Mm} = z_m + 2 \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} a_{ijm} z_i z_j \\
 & + 4 \sum_{i=p-k^*+1}^n \sum_{h=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijm} a_{hjk} z_i z_h z_k \\
 & - 2 \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{\ell=1}^{p-k^*} a_{ijk} a_{ilm} z_j z_k z_\ell \\
 & + 3 \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{\ell=1}^{p-k^*} a_{Mijml} z_i z_j z_\ell + o(z^3), \\
 & \quad (m = 1, 2, \dots, p-k^*),
 \end{aligned}$$

and

$$\begin{aligned}
 (3.3.11) \quad & S^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}) \\
 & = \sum_{j=p-k^*+1}^n z_j^2 - 2 \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijk} z_i z_j z_k \\
 & + \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{\ell=1}^{p-k^*} \sum_{m=1}^{p-k^*} a_{ijk} a_{ilm} z_j z_k z_\ell z_m \\
 & - 4 \sum_{h=p-k^*+1}^n \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{\ell=1}^{p-k^*} a_{hjl} a_{ikl} z_h z_i z_j z_k \\
 & - 2 \sum_{i=p-k^*+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{\ell=1}^{p-k^*} a_{Mijkl} z_i z_j z_k z_\ell + o(z^4).
 \end{aligned}$$

Provided that σ_0 , the a_{hjk} , and the a_{hjkl} or a_{Mhjkl} are sufficiently small, the quartic approximations of $S(\hat{\theta})$ and $S^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op})$ are adequate for most of the z_i , and $s_i^D(z)$ ($i = 1, 2, \dots, 5$) can thus be approximated by the following expressions for most of the z_i :

$$(3.3.12) \quad s_1^D(z) = \sum_{j=p-k^*+1}^p z_j^2 / \sigma_0^2$$

$$- \frac{2}{\sigma_0^3} \sum_{i=p-k^*+1}^p \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijk}^* z_i z_j z_k$$

$$+ \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j,k=1,2,\dots,p} a_{ijk}^* z_i z_j z_k$$

and at least one of j, k is equal to $p-k^*+1$ or $p-k^*+2, \dots$ or p

$$+ \frac{1}{\sigma_0^4} \sum_{i=p-k^*+1}^p \left(\sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijk}^* z_j z_k \right)^2$$

$$- \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j,k,l,m=1,2,\dots,p} a_{ijk}^* a_{ilm}^* z_j z_k z_l z_m$$

and at least one of j, k, l, m is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$- \frac{4}{\sigma_0^4} \sum_{h,i=p-k^*+1, p-k^*+2, \dots, n} \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{l=1}^{p-k^*} a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k$$

and at least one of h, i is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$+ \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j,k,l=1,2,\dots,p} a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k$$

and at least one of j, k, l is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$-\frac{2}{\sigma_0^4} \sum_{i=p-k^*+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p a_{MCijkl} z_i z_j z_k z_l,$$

where

$$a_{ijk}^* = a_{ijk} \sigma_0^2,$$

$$a_{MCijkl} = a_{Mijkl} \sigma_0^2, \quad (i = p-k^*+1, p-k^*+2, \dots, p; j, k, l = 1, 2, \dots, p-k^*),$$

$$a_{MCijkl} = 0 \quad , \quad (i = p-k^*+1, p-k^*+2, \dots, p; j, k, l = 1, 2, \dots, p \\ \text{and at least one of } j, k, l \text{ is equal to } p-k^*+1 \text{ or } p-k^*+2, \dots, \text{ or } p)$$

$$a_{MCijkl} = a_{Mijkl} \sigma_0^2 - a_{ijkl} \sigma_0^2, \quad (i = p+1, p+2, \dots, n; j, k, l = 1, 2, \dots, p-k^*)$$

and

$$a_{MCijkl} = -a_{ijkl} \sigma_0^2, \quad (i = p+1, p+2, \dots, n; j, k, l = 1, 2, \dots, p \text{ and at least one of } j, k, l \text{ is equal to } p-k^*+1 \text{ or } p-k^*+2, \dots, \text{ or } p);$$

$$(3.3.13) \quad S_2^D(z) = \sum_{j=1}^p z_j^2 / \sigma_0^2$$

$$+ \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p a_{ijk}^* z_i z_j z_k$$

$$- \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p \sum_{m=1}^p a_{ijk}^* a_{ilm}^* z_j z_k z_l z_m$$

$$+ \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p \sum_{m=1}^p a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k$$

$$+ \frac{2}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p a_{ijkl}^* z_i z_j z_k z_l,$$

where

$$a_{ijkl}^* = a_{ijkl} \sigma_0^2 ;$$

$$(3.3.14) \quad S_3^D(z) = \sum_{j=p-k^*+1}^p z_j^2 / \sigma_0^2$$

$$- \frac{2}{\sigma_0^3} \sum_{i=p-k^*+1}^p \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijk}^* z_i z_j z_k$$

$$+ \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \frac{k^* F}{n-p} a_{ijk}^* z_i z_j z_k$$

$$+ \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j,k=1,2,\dots,p} (1 + \frac{k^* F}{n-p}) a_{ijk}^* z_i z_j z_k$$

and at least one of j, k is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$+ \frac{1}{\sigma_0^4} \sum_{i=p-k^*+1}^p \left(\sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} a_{ijk}^* z_j z_k \right)^2$$

$$- \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \left(\sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sqrt{\frac{k^* F}{n-p}} a_{ijk}^* z_j z_k \right)^2$$

$$- \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j,k,l,m=1,2,\dots,p} (1 + \frac{k^* F}{n-p}) a_{ijk}^* a_{ilm}^* z_j z_k z_l z_m$$

and at least one of j, k, l, m is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$- \frac{4}{\sigma_0^4} \sum_{h,i=p-k^*+1, p-k^*+2, \dots, n} \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{l=1}^{p-k^*} a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k$$

and at least one of h, i is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p

$$+ \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} \sum_{l=1}^{p-k^*} \frac{k^*F}{n-p} a_{hj\ell}^* a_{ik\ell}^* z_h z_i z_j z_k$$

$$+ \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j,k,\ell=1,2,\dots,p} (1 + \frac{k^*F}{n-p}) a_{hj\ell}^* a_{ik\ell}^* z_h z_i z_j z_k$$

and at least one of j, k, ℓ is equal to $p-k^*+1$ or $p-k^*+2, \dots$, or p .

$$- \frac{2}{\sigma_0^4} \sum_{i=p-k^*+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{\ell=1}^p a_{MFijk\ell}^* z_i z_j z_k z_\ell,$$

where

$$F = F_\alpha(k^*, n-p),$$

$$a_{MFijk\ell} = a_{Mijk\ell} \sigma_0^2, \quad (i = p-k^*+1, p-k^*+2, \dots, p; j, k, \ell = 1, 2, \dots, p-k^*),$$

$$a_{MFijk\ell} = 0 \quad , \quad (i = p-k^*+1, p-k^*+2, \dots, p; j, k, \ell = 1, 2, \dots, p \text{ and} \\ \text{at least one of } j, k, \ell \text{ is equal to } p-k^*+1 \text{ or} \\ p-k^*+2, \dots, \text{ or } p),$$

$$a_{MFijk\ell} = a_{Mijk\ell} \sigma_0^2 - (1 + \frac{k^*F}{n-p}) a_{ijk\ell} \sigma_0^2, \quad (i = p+1, p+2, \dots, n; \\ j, k, \ell = 1, 2, \dots, p-k^*),$$

$$a_{MFijk\ell} = -(1 + \frac{k^*F}{n-p}) a_{ijk\ell} \sigma_0^2, \quad (i = p+1, p+2, \dots, n; j, k, \ell = 1, 2, \dots, p \\ \text{and at least one of } j, k, \ell \text{ is equal to} \\ p-k^*+1 \text{ or } p-k^*+2, \dots, \text{ or } p);$$

$$(3.3.15) \quad S_4^D(z) = \sum_{j=1}^p z_j^2$$

$$+ \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p (1 + \frac{pF}{n-p}) a_{ijk}^* z_i z_j z_k$$

$$- \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{\ell=1}^p \sum_{m=1}^p (1 + \frac{pF}{n-p}) a_{ijk}^* a_{ilm}^* z_i z_j z_k z_\ell z_m$$

$$\begin{aligned}
& + \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p (1 + \frac{pF}{n-p}) a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k \\
& + \frac{2}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p (1 + \frac{pF}{n-p}) a_{ijk\ell}^* z_i z_j z_k z_\ell,
\end{aligned}$$

where

$$F = F_\alpha(p, n-p);$$

and

$$\begin{aligned}
(3.3.16) \quad S_5^D(z) = & \sum_{j=p+1}^n z_j^2 \\
& - \frac{2}{\sigma_0^3} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p a_{ijk}^* z_i z_j z_k \\
& + \frac{1}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p \sum_{m=1}^p a_{ijk}^* a_{ilm}^* z_j z_k z_\ell z_m \\
& - \frac{4}{\sigma_0^4} \sum_{h=p+1}^n \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p a_{hjl}^* a_{ikl}^* z_h z_i z_j z_k \\
& - \frac{2}{\sigma_0^4} \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p \sum_{l=1}^p a_{ijk\ell}^* z_i z_j z_k z_\ell.
\end{aligned}$$

Given that $\theta = \theta_0$ and $\sigma = \sigma_0$, z_j / σ_0 ($j = 1, 2, \dots, n$) are independently distributed as $N(0, 1)$. Therefore as far as deriving approximations of the probabilities $I_i(\theta_0, \sigma_0)$ is concerned, we can set σ_0 appearing in the expressions of $S_i^D(z)$ given by (3.3.12)-(3.3.16) to be one while leaving the σ_0 appearing in the expressions of the $a_i^{*2}, a_{ijk}^*, a_{ijk\ell}^*, a_{MCijkl}^*$ and a_{MFijkl}^* unchanged, and treat the z_j as being distributed as $N(0, 1)$. Let $z_j^{(s)}$ denote z_j^2 . Each of $z_j^{(s)}$ is then distributed as χ_1^2 , and conditional on z_j being negative (or non-negative), $z_j^{(s)}$ is still

distributed as χ^2_1 . Further, let $E_{z_{i_1}, z_{i_2}, \dots, z_{i_k}} \{f(z_{i_1}, z_{i_2}, \dots, z_{i_k})\}$ denote the expectation of $f(z_{i_1}, z_{i_2}, \dots, z_{i_k})$. The probabilities $I_i(\theta_0, \sigma_0)$ can now be written as

$$(3.3.17) \quad I_i(\theta_0, \sigma_0) = E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right. \\ \left[\frac{1}{2^{k^*}} \int_{z_{p-k^*+1}^{(s)}} \int_{z_{p-k^*+2}^{(s)}} \dots \int_{z_p^{(s)}} \chi_1^2(z_{p-k^*+1}^{(s)}) \chi_1^2(z_{p-k^*+2}^{(s)}) \dots \chi_1^2(z_p^{(s)}) \right. \\ \left. (z_{p-k^*+1}^{(s)}, z_{p-k^*+2}^{(s)}, \dots, z_p^{(s)}) e z_i^* \right. \\ \left. \times dz_{p-k^*+1}^{(s)} dz_{p-k^*+2}^{(s)} \dots dz_p^{(s)} \right] \} , \quad (i = 1 \text{ and } 3),$$

$$(3.3.18) \quad I_i(\theta_0, \sigma_0) = E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1} \right. \\ \left[\frac{1}{2^p} \int_{z_1^{(s)}} \int_{z_2^{(s)}} \dots \int_{z_p^{(s)}} \chi_1^2(z_1^{(s)}) \chi_1^2(z_2^{(s)}) \dots \chi_1^2(z_p^{(s)}) \right. \\ \left. (z_1^{(s)}, z_2^{(s)}, \dots, z_p^{(s)}) e z_i^* \right. \\ \left. \times dz_1^{(s)} dz_2^{(s)} \dots dz_p^{(s)} \right] \} , \quad (i = 2 \text{ and } 4),$$

and

$$(3.3.19) \quad I_5(\theta_0, \sigma_0) = E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\frac{1}{2^{n-p}} \int_{z_{p+1}^{(s)}} \int_{z_{p+2}^{(s)}} \dots \int_{z_n^{(s)}} \chi_1^2(z_{p+1}^{(s)}) \chi_1^2(z_{p+2}^{(s)}) \dots \chi_1^2(z_n^{(s)}) dz_{p+1}^{(s)} dz_{p+2}^{(s)} \dots dz_n^{(s)} \right] \\ (z_{p+1}^{(s)}, z_{p+2}^{(s)}, \dots, z_n^{(s)}) \in Z_5^*,$$

where

$$(3.3.20) \quad Z_i^* = \left\{ (z_{p-k^*+1}^{(s)}, z_{p-k^*+2}^{(s)}, \dots, z_p^{(s)}) : \right.$$

$$s_i^D(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{z_{p-k^*+1}^{(s)}}, s_{p-k^*+2} \sqrt{z_{p-k^*+2}^{(s)}} \dots,$$

$$s_p \sqrt{z_p^{(s)}}, z_{p+1}, z_{p+2}, \dots, z_n) \leq d_i^{*2} \right\},$$

$$(i = 1 \text{ and } 3),$$

$$(3.3.21) \quad Z_i^* = \left\{ (z_1^{(s)}, z_2^{(s)}, \dots, z_p^{(s)}) : \right.$$

$$s_i^D(s_1 \sqrt{z_1^{(s)}}, s_2 \sqrt{z_2^{(s)}}, \dots, s_p \sqrt{z_p^{(s)}}, z_{p+1}, z_{p+2}, \dots, z_n) \leq d_i^{*2} \right\},$$

$$(i = 2 \text{ and } 4),$$

$$(3.3.22) \quad z_5^* = \left\{ (z_{p+1}^{(s)}, z_{p+2}^{(s)}, \dots, z_n^{(s)}) : \right.$$

$$\left. s_5^D(z_1, z_2, \dots, z_p, s_{p+1}\sqrt{z_{p+1}^{(s)}}, s_{p+2}\sqrt{z_{p+2}^{(s)}}, \dots, s_n\sqrt{z_n^{(s)}}) \leq d_5^{*2} \right\},$$

$$(3.3.23) \quad s_i = \begin{cases} -1 & \text{if } z_i < 0 \\ +1 & \text{if } z_i \geq 0 \end{cases}, \quad (i = 1, 2, \dots, n),$$

and for $k \geq 1$,

$$x_k^2(t) = \frac{e^{-t/2} t^{\frac{k}{2}-1}}{2^{k/2} \Gamma(k/2)}, \quad (t \geq 0).$$

For $i = 1$ and 3 , we apply the transformations

$$(3.3.24) \quad r^{(s)} = \sum_{j=p-k^*+1}^p z_j^{(s)}$$

and

$$(3.3.25) \quad \bar{z}_j^{(s)} = \frac{1}{r^{(s)}} z_j^{(s)} \quad \text{for } j = p-k^*+1, p-k^*+2, \dots, p-1.$$

For $i = 2$ and 4 , we apply the transformations

$$(3.3.26) \quad r^{(s)} = \sum_{j=1}^p z_j^{(s)}$$

and

$$(3.3.27) \quad \bar{z}_j^{(s)} = \frac{1}{r^{(s)}} z_j^{(s)} \quad \text{for } j = 1, 2, \dots, p-1.$$

For $i = 5$, we apply the transformations

$$(3.3.28) \quad r^{(s)} = \sum_{j=p+1}^n z_j^{(s)}$$

and

$$(3.3.29) \quad \frac{z_j(s)}{z_r(s)} = \frac{1}{z_j(s)} \quad \text{for } j = p+1, p+2, \dots, n-1.$$

The probabilities $I_{ij}(\theta_0, \sigma_0)$ can then be written as

$$(3.3.30) \quad I_i(\theta_0, \sigma_0) = E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1}, \sum_{s_{p-k^*+2}=1,+1}, \dots, \sum_{s_p=-1,+1} \right.$$

$$\left[\frac{1}{2^{k^*}} \int_0^1 z_{p-k^*+1}^{-}(s) = 0 \right] \quad \left[\int_0^1 z_{p-k^*+2}^{-}(s) = z_{p-k^*+1}^{-}(s) \right] \quad \dots \quad \left[\int_0^1 z_{p-1}^{-}(s) = z_{p-2}^{-}(s) \right]$$

$$\chi_1^2(z_{p-k^*+1}^{(s)})\chi_1^2(z_{p-k^*+2}^{(s)})\dots\chi_1^2(z_{p-1}^{(s)})\chi_1^2(l-\sum_{j=p-k^*+1}^{p-1}z_j^{(s)})[\chi_{k^*}^2(l)]^{-1}$$

$$\times \left(\int_{R_i^*} x_{k*}^2(r^{(s)}) d(r^{(s)}) \right)$$

$$\times \left. \frac{dz^{(s)}}{dp-k^*+1} \frac{dz^{(s)}}{p-k^*+2} \dots \frac{dz^{(s)}}{p-1} \right\} , \quad (i = 1 \text{ and } 3),$$

(3.3.31)

$$I_i(\theta_0, \sigma_0) = E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\frac{1}{2^p} \int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 x_1^2(\bar{z}_1^{(s)}) x_1^2(\bar{z}_2^{(s)}) \dots x_1^2(\bar{z}_{p-1}^{(s)}) \right.$$

$$\times x_1^2(1 - \sum_{j=1}^{p-1} \bar{z}_j^{(s)}) [x_p^2(1)]^{-1}$$

$$\times \left(\int_{R_i^*} x_p^2(r^{(s)}) d(r^{(s)}) \right)$$

$$\left. \times d\bar{z}_1^{(s)} d\bar{z}_2^{(s)} \dots d\bar{z}_{p-1}^{(s)} \right] \left. \right\} , \quad (i = 2 \text{ and } 4),$$

and

(3.3.32)

$$I_5(\theta_0, \theta_0) = E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\frac{1}{2^{n-p}} \int_{\bar{z}_{p+1}^{(s)}=0}^1 \int_{\bar{z}_{p+2}^{(s)}=\bar{z}_{p+1}^{(s)}}^1 \int_{\bar{z}_{n-1}^{(s)}=\bar{z}_{n-2}^{(s)}}^1 x_1^2(\bar{z}_{p+1}^{(s)}) x_1^2(\bar{z}_{p+2}^{(s)}) \dots x_1^2(\bar{z}_{n-1}^{(s)}) \right.$$

$$\times x_1^2(1 - \sum_{j=p+1}^{n-1} \bar{z}_j^{(s)}) [x_{n-p}^2(1)]^{-1}$$

$$\times \left(\int_{R_5^*} \chi_{n-p}^2(r^{(s)}) d(r^{(s)}) \right)$$

$$\times dz_{p+1}^{-(s)} dz_{p+2}^{-(s)} \dots dz_{n-1}^{-(s)} \Big] \Big\} ,$$

where

(3.3.33)

$$R_i^* = \{r^{(s)} : s_i^D(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{r^{(s)} z_{p-k^*+1}}, s_{p-k^*+2} \sqrt{r^{(s)} z_{p-k^*+2}}, \dots, s_{p-1} \sqrt{r^{(s)} z_{p-1}}, s_p \sqrt{r^{(s)} (1 - \sum_{j=p-k^*+1}^{p-1} \frac{z_j^{-(s)}}{z_j})}, z_{p+1}, z_{p+2}, \dots, z_n) \leq d_i^* \}$$

$$, (i = 1 \text{ and } 3),$$

(3.3.34)

$$R_i^* = \{r^{(s)} : s_i^D(s_1 \sqrt{r^{(s)} z_1}, s_2 \sqrt{r^{(s)} z_2}, \dots, s_{p-1} \sqrt{r^{(s)} z_{p-1}}, s_p \sqrt{r^{(s)} (1 - \sum_{j=1}^{p-1} \frac{z_j^{-(s)}}{z_j})}, z_{p+1}, z_{p+2}, \dots, z_n) \leq d_i^* \}, (i = 2 \text{ and } 4),$$

and

(3.3.35)

$$R_5^* = \{r^{(s)} : s_5^D(z_1, z_2, \dots, z_p, s_{p+1} \sqrt{r^{(s)} z_{p+1}}, s_{p+2} \sqrt{r^{(s)} z_{p+2}}, \dots, s_{n-1} \sqrt{r^{(s)} z_{n-1}}, s_n \sqrt{r^{(s)} (1 - \sum_{j=p+1}^{n-1} \frac{z_j^{-(s)}}{z_j}))} \leq d_5^* \} .$$

To the extent that the approximations of $s_i^D(z)$ given by (3.3.12)-(3.3.16) are adequate, we can obtain a truncated series expansion of the probability $I_i(\theta_0, \sigma_0)$ as follows:

(3.3.36)

$$I_i(\theta_0, \sigma_0) \approx (1-\alpha) + \sum_{j=1}^{n^*} I_{ia_j^*} a_j^* + \frac{1}{2} \sum_{j=1}^{n^*} \sum_{k=1}^{n^*} I_{ia_j^* a_k^*} a_j^* a_k^*$$

where $n^* = \frac{1}{2} p(p+1)n$,

$\{a_1^*, a_2^*, \dots, a_{n^*}^*\}$ is the set containing all the a_{hjk}^* for which $j \leq k$,

$$I_{ia_j^*} = \left[\frac{\partial I_i(\theta_0, \sigma_0)}{\partial a_j^*} \right]_{a^*=0},$$

$$I_{ia_j^* a_k^*} = \left[\frac{\partial^2 I_i(\theta_0, \sigma_0)}{\partial a_j^* \partial a_k^*} \right]_{a^*=0}$$

and \underline{a}^* is a column vector whose components are the a_{hjk}^* , and the a_{hjkl}^* , a_{MChjkl} or a_{MFhjkl} for which $j \leq k \leq l$.

We now consider how $I_{ia_j^*}$ and $I_{ia_j^* a_k^*}$ can be derived. We first note that provided that the magnitude $|\underline{a}^*|$ of \underline{a}^* is sufficiently small, the sets R_i^* can be written as

$$R_i^* = \{r_i^{(s)} : 0 \leq r_i^{(s)} \leq r_i^{(*s)}\},$$

where $r_i^{(*s)}$ are such that

(3.3.37)

$$\begin{aligned} S_i^D(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{r_i^{(*s)} - z_{p-k^*+1}}, s_{p-k^*+2} \sqrt{r_i^{(*s)} - z_{p-k^*+2}}, \dots \\ \dots, s_{p-1} \sqrt{r_i^{(*s)} - z_{p-1}}, s_p \sqrt{r_i^{(*s)} (1 - \sum_{j=p-k^*+1}^{p-1} z_j^{(s)})}, z_{p+1}, z_{p+2}, \dots, z_n) = d_i^{*2} \\ , \quad (i = 1, 3), \end{aligned}$$

(3.3.38)

$$s_i^D(s_1 \sqrt{r_i^{(s)} - z_1^{(s)}}, s_2 \sqrt{r_i^{(s)} - z_2^{(s)}}, \dots, s_{p-1} \sqrt{r_i^{(s)} - z_{p-1}^{(s)}})$$

$$s_p \sqrt{r_i^{(s)} (1 - \sum_{j=1}^{p-1} \frac{z_j^{(s)}}{z_p^{(s)}})}, z_{p+1}, z_{p+2}, \dots, z_n) = d_i^{*2}, \\ (i = 2, 4),$$

and

(3.3.39)

$$s_5^D(z_1, z_2, \dots, z_p, s_{p+1} \sqrt{r_5^{(s)} - z_{p+1}^{(s)}}, s_{p+2} \sqrt{r_5^{(s)} - z_{p+2}^{(s)}}, \dots, s_{n-1} \sqrt{r_5^{(s)} - z_{n-1}^{(s)}}) \\ s_n \sqrt{r_5^{(s)} (1 - \sum_{j=p+1}^{n-1} \frac{z_j^{(s)}}{z_n^{(s)}})} = d_5^{*2}, \quad (i = 5).$$

The equations (3.3.37), (3.3.38) and (3.3.39) are of the form

$$(3.3.40) \quad r_i^{(s)} + \sum_{j=1}^{n^*} h_j^*(r_i^{(s)}) a_j^* + \sum_{j=1}^{n^*} \sum_{k=1}^{n^*} h_{jk}^*(r_i^{(s)}) a_j^* a_k^* + b^* = d_i^{*2},$$

where $h_j^*(r_i^{(s)})$ and $h_{jk}^*(r_i^{(s)})$ are functions of $r_i^{(s)}$, the z_ℓ and $\bar{z}_m^{(s)}$, and b^* is a function of the z_ℓ and $\bar{z}_m^{(s)}$, and the a_{ijkl}^* , a_{MCijkl} or a_{MFijkl} . Note that as b^* does not depend on the a_j^* , we have

$$\left[\frac{\partial b^*}{\partial a_j^*} \right]_{a^*=0} = 0, \quad (j = 1, 2, \dots, n^*),$$

and

$$\left[\frac{\partial^2 b^*}{\partial a_j^* \partial a_k^*} \right]_{a^*=0} = 0, \quad (j, k = 1, 2, \dots, n^*).$$

By differentiating both sides of (3.3.40) with respect to a_j^* , we get

$$(3.3.41) \quad \frac{\partial r_i^{(*s)}}{\partial a_j^*} + h_j^*(r_i^{(*s)}) + \sum_{j_1=1}^{n^*} \frac{\partial h_{j_1}^*(r_i^{(*s)})}{\partial r_i^{(*s)}} \frac{\partial r_i^{(*s)}}{\partial a_j^*} a_{j_1}^* \\ + 2 \sum_{k=1}^{n^*} h_{jk}^*(r_i^{(*s)}) a_k^* + \sum_{j_1=1}^{n^*} \sum_{k=1}^{n^*} \frac{\partial h_{j_1 k}^*(r_i^{(*s)})}{\partial r_i^{(*s)}} \frac{\partial r_i^{(*s)}}{\partial a_j^*} a_{j_1}^* a_k^* \\ + \frac{\partial b^*}{\partial a_j^*} = 0.$$

Hence $\left[\frac{\partial r_i^{(*s)}}{\partial a_j^*} \right]_{a^*=0} = -h_j^*(d_i^{*2})$.

By differentiating both sides of (3.3.41) with respect to a_k^* and subsequently setting a^* to 0, we get

(3.3.42)

$$\left[\frac{\partial^2 r_i^{(*s)}}{\partial a_j^* \partial a_k^*} \right]_{a^*=0} = - \left[\frac{\partial h_j^*(r_i^{(*s)})}{\partial r_i^{(*s)}} \frac{\partial r_i^{(*s)}}{\partial a_k^*} \right]_{a^*=0} - \left[\frac{\partial h_k^*(r_i^{(*s)})}{\partial r_i^{(*s)}} \frac{\partial r_i^{(*s)}}{\partial a_j^*} \right]_{a^*=0} - 2h_{jk}^*(d_i^{*2}).$$

From the integrability and continuity of the function $x_k^2(r^{(s)})$ and the differentiability of $r_i^{(*s)}$ with respect to the a_j^* , we have

$$\left[\frac{\partial}{\partial a_j^*} \int_{r(s)=0}^{r_i^{(*s)}} x_k^2(r^{(s)}) d(r^{(s)}) \right]_{a^*=0} = x_k^2(d_i^{*2}) \left[\frac{\partial r_i^{(*s)}}{\partial a_j^*} \right]_{a^*=0}$$

and

$$\begin{aligned} & \left[\frac{\partial^2}{\partial a_j^* \partial a_k^*} \int_{r(s)=0}^{r_i^{(s)}} x_k^2(r(s)) d(r(s)) \right]_{a^*=0} \\ &= x_k^2(d_i^{*2}) \left[\left[\frac{1}{2} - 1 \right] \frac{1}{d_i^{*2}} - \frac{1}{2} \right] \left[\frac{\partial r_i^{(s)}}{\partial a_j^*} \frac{\partial r_i^{(s)}}{\partial a_k^*} \right]_{a^*=0} + \left[\frac{\partial^2 r_i^{(s)}}{\partial a_j^* \partial a_k^*} \right]_{a^*=0}, \end{aligned}$$

where $k = k^*$, p or $n-p$.

Then by showing that conditions similar to those which justify the validity of

$$\left[\frac{\partial}{\partial a} \int_{-\infty}^{\infty} g(z, a) dz \right]_{a=a_0} = \int_{-\infty}^{\infty} \left[\frac{\partial g(z, a)}{\partial a} \right]_{a=a_0} dz,$$

where $\int_{-\infty}^{\infty} g(z, a) dz$ exists, are satisfied, we get

(3.3.43)

$$I_{ia_j^*} = E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\begin{aligned} & \left[\frac{1}{2^{k^*}} \int_{\bar{z}_{p-k^*+1}^{(s)}=0}^1 \int_{\bar{z}_{p-k^*+2}^{(s)}=\bar{z}_{p-k^*+1}^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 x_1^2(\bar{z}_{p-k^*+1}^{(s)}) x_1^2(\bar{z}_{p-k^*+2}^{(s)}) \right. \\ & \quad \left. \dots x_1^2(\bar{z}_{p-1}^{(s)}) \right] \end{aligned}$$

$$\times x_1^2(1 - \sum_{j_1=p-k^*+1}^{p-1} \bar{z}_{j_1}^{(s)}) [x_{k^*}^2(1)]^{-1} x_{k^*}^2(d_i^{*2}) \left[\frac{\partial r_i^{(s)}}{\partial a_j^*} \right]_{a^*=0}$$

$$\times \left[dz_{p-k^*+1}^{(s)} dz_{p-k^*+2}^{(s)} \dots dz_{p-1}^{(s)} \right] \} , \quad (i = 1 \text{ and } 3),$$

(3.3.44)

$$I_{ia_j^*} = E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1} \right. \\ \left[\frac{1}{2^p} \int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \chi_1^2(\bar{z}_1^{(s)}) \chi_1^2(\bar{z}_2^{(s)}) \dots \chi_1^2(\bar{z}_{p-1}^{(s)}) \right. \\ \left. \times \chi_1^2(1 - \sum_{j_1=1}^{p-1} \bar{z}_{j_1}^{(s)}) [\chi_p^2(1)]^{-1} \chi_p^2(d_i^{*2}) \left[\frac{\partial r_i^{(s)}}{\partial a_j^*} \right]_{a^*=0} \right. \\ \left. \times dz_1^{(s)} dz_2^{(s)} \dots dz_{p-1}^{(s)} \right\} , \quad (i = 2 \text{ and } 4),$$

(3.3.45)

$$I_{5a_j^*} = E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\frac{1}{2^{n-p}} \int_{\bar{z}_{p+1}^{(s)}=0}^1 \int_{\bar{z}_{p+2}^{(s)}=\bar{z}_{p+1}^{(s)}}^1 \dots \int_{\bar{z}_{n-1}^{(s)}=\bar{z}_{n-2}^{(s)}}^1 \chi_1^2(\bar{z}_{p+1}^{(s)}) \chi_1^2(\bar{z}_{p+2}^{(s)}) \dots \chi_1^2(\bar{z}_{n-1}^{(s)}) \right.$$

$$\left. \times \chi_1^2(1 - \sum_{j_1=p+1}^{n-1} \bar{z}_{j_1}^{(s)}) [\chi_{n-p}^2(1)]^{-1} \chi_{n-p}^2(d_5^{*2}) \left[\frac{\partial r_5^{(s)}}{\partial a_j^*} \right]_{a^*=0} \right]$$

$$\left. \times dz_{p+1}^{-(s)} dz_{p+2}^{-(s)} \dots dz_{n-1}^{-(s)} \right] \} , \quad (i = 5),$$

(3.3.46)

$$I_{i a^* j a^*} = E_{z_{p+1} z_{p+2} \dots z_n, z_1 z_2 \dots z_{p-k^*}} \left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\frac{1}{2^{k^*}} \int_{\bar{z}_{p-k^*+1}^{(s)}=0}^1 \int_{\bar{z}_{p-k^*+2}^{(s)}=\bar{z}_{p-k^*+1}^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \dots \right]$$

$$\chi_1^2(\bar{z}_{p-k^*+1}^{(s)}) \chi_1^2(\bar{z}_{p-k^*+2}^{(s)}) \dots \chi_1^2(\bar{z}_{p-1}^{(s)})$$

$$\begin{aligned} & \times \chi_1^2(1 - \sum_{j_1=p-k^*+1}^{p-1} \bar{z}_{j_1}^{(s)}) [\chi_{k^*}^2(1)]^{-1} \chi_{k^*}^2(d_i^{*2}) \left[\left(\frac{k^*}{2} - 1 \right) \frac{1}{d_i^{*2}} - \frac{1}{2} \right] \left[\frac{\partial r_i^{(*s)}}{\partial a_j^*} \frac{\partial r_i^{(*s)}}{\partial a_k^*} \right]_{a^*=0} \\ & + \left[\frac{\partial^2 r_i^{(*s)}}{\partial a_j^* \partial a_k^*} \right]_{a^*=0} \end{aligned}$$

$$\left. \times dz_{p-k^*+1}^{-(s)} dz_{p-k^*+2}^{-(s)} \dots dz_{p-1}^{-(s)} \right] \} , \quad (i = 1 \text{ and } 3),$$

(3.3.47)

$$I_{i a^* j a^*} = E_{z_{p+1} z_{p+2} \dots z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\frac{1}{2^p} \int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \chi_1^2(\bar{z}_1^{(s)}) \chi_1^2(\bar{z}_2^{(s)}) \dots \chi_1^2(\bar{z}_{p-1}^{(s)}) \right]$$

$$\begin{aligned}
& x_1^2 \left(1 - \sum_{j_1=1}^{p-1} \bar{z}_{j_1}^{(s)} \right) [x_p^2(1)]^{-1} x_p^2(d_i^{*2}) \left[\left[\left(\frac{p}{2} - 1 \right) \frac{1}{d_i^{*2}} - \frac{1}{2} \right] \left[\frac{\partial r_i^{(s)}}{\partial a_j^*} \frac{\partial r_i^{(s)}}{\partial a_k^*} \right]_{a^*=0} \right. \\
& \quad \left. + \left[\frac{\partial^2 r_i^{(s)}}{\partial a_j^* \partial a_k^*} \right]_{a^*=0} \right] \\
& \times \left. \bar{d}z_{p+1}^{(s)} \bar{d}z_{p+2}^{(s)} \dots \bar{d}z_{n-1}^{(s)} \right\} , \quad (i = 2 \text{ and } 4),
\end{aligned}$$

and

(3.3.48)

$$I_{5,j,k}^{a^*a^*} = E_{z_1, z_2, \dots, z_p}$$

$$\begin{aligned}
& \left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right. \\
& \left[\frac{1}{2^{n-p}} \int_{\bar{z}_{p+1}^{(s)}=0}^1 \int_{\bar{z}_{p+2}^{(s)}=\bar{z}_{p+1}^{(s)}}^1 \dots \int_{\bar{z}_{n-1}^{(s)}=\bar{z}_{n-2}^{(s)}}^1 x_1^2(\bar{z}_{p+1}^{(s)}) x_1^2(\bar{z}_{p+2}^{(s)}) \dots x_1^2(\bar{z}_{n-1}^{(s)}) \right. \\
& \times x_1^2 \left(1 - \sum_{j_1=p+1}^{n-1} \bar{z}_{j_1}^{(s)} \right) [x_{n-p}^2(1)]^{-1} x_{n-p}^2(d_5^{*2}) \left[\left[\left(\frac{n-p}{2} - 1 \right) \frac{1}{d_5^{*2}} - \frac{1}{2} \right] \right. \\
& \quad \left. \times \left[\frac{\partial r_5^{(s)}}{\partial a_j^*} \frac{\partial r_5^{(s)}}{\partial a_k^*} \right]_{a^*=0} + \left[\frac{\partial^2 r_5^{(s)}}{\partial a_j^* \partial a_k^*} \right]_{a^*=0} \right] \\
& \times \left. \bar{d}z_{p+1}^{(s)} \bar{d}z_{p+2}^{(s)} \dots \bar{d}z_{n-1}^{(s)} \right\} , \quad (i = 5).
\end{aligned}$$

In deriving $I_{ia_j^*}$ and $I_{ia_j^*a_k^*}$, we can make use of

$$(3.3.49) \quad \int_0^1 \int_0^1 \cdots \int_0^1 \left[\frac{\bar{z}_{i_1}^{(s)}}{z_1} \right]^{\frac{m_1}{2}} \left[\frac{\bar{z}_{i_2}^{(s)}}{z_2} \right]^{\frac{m_2}{2}} \cdots \left[\frac{\bar{z}_{i_q}^{(s)}}{z_q} \right]^{\frac{m_q}{2}}$$

$$\begin{aligned} & \bar{z}_1^{(s)} = 0 \quad \bar{z}_2^{(s)} = z_1^{(s)} \quad \cdots \quad \bar{z}_{m-1}^{(s)} = z_{m-2}^{(s)} \\ & \times x_1^2(\bar{z}_1^{(s)}) x_1^2(\bar{z}_2^{(s)}) \cdots x_1^2(\bar{z}_m^{(s)}) dz_1^{(s)} dz_2^{(s)} \cdots dz_{m-1}^{(s)} \\ & = \left[\prod_{k=1}^q \frac{\frac{m_{k_1}+1}{2} \Gamma\left(\frac{m_{k_1}+1}{2}\right)}{\sqrt{2\pi}} \right] x_{m+\sum_{l=1}^q m_l}^2 \quad (1), \end{aligned}$$

where

$$m = 2, 3, \dots,$$

$$q = 1, 2, \dots, m,$$

$$m_l = 0, 1, 2, \dots, \text{for } l = 1, 2, \dots, q,$$

$$i_l = 1, 2, \dots, m \text{ for } l = 1, 2, \dots, q \text{ and these } i_l \text{ are all distinct,}$$

and

$$\bar{z}_m^{(s)} = 1 - \sum_{l=1}^{m-1} \bar{z}_l^{(s)}.$$

The derivation of $I_{ia_j^*}$ and $I_{ia_j^*a_k^*}$ is shown in Appendix 2. From the expressions of $I_{ia_j^*}$ and $I_{ia_j^*a_k^*}$, it is found that $I_1(\theta_0, \sigma_0)$ given by (3.3.36) is

(3.3.50)

$$\begin{aligned} I_1(\theta_0, \sigma_0) & \approx (1-\alpha) + \sigma_0^2 x_3^2 [x_{1,\alpha}^2] \left[\sum_{j=1}^{p-1} \sum_{k=1}^{p-1} (2a_{pjk}^2 - a_{pj} a_{pk}) \right. \\ & \left. - \sum_{i=p+1}^n [a_{ipp}^2 + 2(\sum_{j=1}^{p-1} (2a_{ijp}^2 - a_{ij} a_{ipp}))] \right], \quad (p \geq 2; k^* = 1), \end{aligned}$$

(3.3.51)

$$\begin{aligned}
 I_1(\theta_0, \sigma_0) &\approx (1-\alpha) + \sigma_0^2 \chi_{k^*+2}^2 [\chi_{k^*, \alpha}^2] \left[\sum_{i=p-k^*+1}^p \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} (2a_{ijk}^2 - a_{ijj} a_{ikk}) \right. \\
 &- \sum_{i=p+1}^n \left[\sum_{j=p-k^*+1}^p \sum_{k=p-k^*+1}^p (2a_{ijk}^2 - a_{ijj} a_{ikk}) \right. \\
 &+ 2 \left. \sum_{j=1}^{p-k^*} \sum_{k=p-k^*+1}^p (2a_{ijk}^2 - a_{ijj} a_{ikk}) \right] \left. \right] \\
 &\quad (p \geq 3; k^* \geq 2).
 \end{aligned}$$

The derivation of $I_i(\theta_0, \sigma_0)$ ($i = 2, 3, 4, 5$) given by (3.3.36) is similar to that of $I_1(\theta_0, \sigma_0)$ and it can be shown that

(3.3.52)

$$I_2(\theta_0, \sigma_0) \approx (1-\alpha) - \sigma_0^2 \chi_{p+2}^2 [\chi_p^2, \alpha] \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p (2a_{ijk}^2 - a_{ijj} a_{ikk}),$$

(3.3.53)

$$\begin{aligned}
 I_3(\theta_0, \sigma_0) &\approx (1-\alpha) + \sigma_0^2 U_{0,3,n-p}(F_1) \sum_{j=1}^{p-1} \sum_{k=1}^{p-1} (2a_{pjk}^2 - a_{pj} a_{pk}) \\
 &- \sigma_0^2 \frac{1}{n-p} U_{0,3,n-p}(F_1) \sum_{i=p+1}^n \sum_{j=1}^{p-1} \sum_{k=1}^{p-1} (2a_{ijk}^2 - a_{ijj} a_{ikk}) \\
 &+ \sigma_0^2 \sum_{i=p+1}^n \left[[(1+F_1)(3F_1-1)U_{2,3,n-p}(F_1) - 3(1+F_1)^2 U_{2,5,n-p}(F_1) \right. \\
 &\quad \left. + 3(1+F_1)U_{0,5,n-p}(F_1)] a_{ipp}^2 \right. \\
 &\quad \left. + [4(1+F_1)F_1 U_{2,1,n-p}(F_1) - 4(1+F_1)(2+F_1)U_{2,3,n-p}(F_1) \right. \\
 &\quad \left. + 4(1+F_1)U_{0,3,n-p}(F_1)] \sum_{j=1}^{p-1} a_{ijp}^2 \right]
 \end{aligned}$$

$$+ [2F_1(1+F_1)U_{2,1,n-p}(F_1) + 2(1+F_1)U_{0,3,n-p}(F_1) \\ - 2F_1(1+F_1)U_{2,3,n-p}(F_1)] \sum_{j=1}^{p-1} a_{ijj} a_{ipp}]$$

, (p ≥ 2; k* = 1),

(3.3.54)

$$\begin{aligned} I_3(\theta_0, \sigma_0) &\approx (1-\alpha) + \sigma_0^2 U_{0,k^*+2,n-p}(F_2) \sum_{i=p-k^*+1}^p \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} (2a_{ijk}^2 - a_{ijj} a_{ikk}) \\ &- \sigma_0^2 \frac{k^*}{n-p} U_{0,k^*+2,n-p}(F_2) \sum_{i=p+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} (2a_{ijk}^2 - a_{ijj} a_{ikk}) \\ &+ \sigma_0^2 \sum_{i=p+1}^n \left[[(1+F_2)(3F_2-1)U_{2,k^*+2,n-p}(F_2) - 3(1+F_2)^2 U_{2,k^*+4,n-p}(F_2) \right. \\ &\quad \left. + 3(1+F_2)U_{0,k^*+4,n-p}(F_2)] \sum_{j=p-k^*+1}^p a_{ijj}^2 \right. \\ &\quad \left. + [2(1+F_2)(F_2-1)U_{2,k^*+2,n-p}(F_2) - 2(1+F_2)^2 U_{2,k^*+4,n-p}(F_2) \right. \\ &\quad \left. + 2(1+F_2)U_{0,k^*+4,n-p}(F_2)] \sum_{\substack{j,k=p-k^*+1, p-k^*+2, \dots, p \\ \text{and } j \neq k}} a_{ijk}^2 \right] \\ &+ [(1+F_2)^2 U_{2,k^*+2,n-p}(F_2) - (1+F_2)^2 U_{2,k^*+4,n-p}(F_2) \\ &\quad + (1+F_2)U_{0,k^*+4,n-p}(F_2)] \sum_{\substack{j,k=p-k^*+1, p-k^*+2, \dots, p \\ \text{and } j \neq k}} a_{ijj} a_{ikk} \\ &+ [4(1+F_2)F_2 U_{2,k^*,n-p}(F_2) - 4(1+F_2)(2+F_2)U_{2,k^*+2,n-p}(F_2) \\ &\quad + 4(1+F_2)U_{0,k^*+2,n-p}(F_2)] \sum_{j=1}^{p-k^*} \sum_{k=p-k^*+1}^p a_{ijk}^2 \end{aligned}$$

$$\begin{aligned}
& + [2F_2(1+F_2)U_{2,k^*,n-p}(F_2) + 2(1+F_2)U_{0,k^*+2,n-p}(F_2) \\
& - 2F_2(1+F_2)U_{2,k^*+2,n-p}(F_2)] \sum_{j=1}^{p-k^*} \sum_{k=p-k^*+1}^p a_{ijj} a_{ikk}] \\
& , \quad (p \geq 3; \quad k^* \geq 2),
\end{aligned}$$

(3.3.55)

$$\begin{aligned}
I_4(\theta_0, \sigma_0) & \approx (1-\alpha) + \sigma_0^2 \sum_{i=p+1}^n \left[(1+F_3)(3F_3-1)U_{2,p+2,n-p}(F_3) - 3(1+F_3)^2 U_{2,p+4,n-p}(F_3) \right. \\
& \quad \left. + 3(1+F_3)U_{0,p+4,n-p}(F_3) \right] \sum_{j=1}^p a_{ijj}^2 \\
& + [2(1+F_3)(F_3-1)U_{2,p+2,n-p}(F_3) - 2(1+F_3)^2 U_{2,p+4,n-p}(F_3) \\
& \quad + 2(1+F_3)U_{0,p+4,n-p}(F_3)] \sum_{\substack{j,k=1,2,\dots,p \\ \text{and } j \neq k}} a_{ijk}^2 \\
& + [(1+F_3)^2 U_{2,p+2,n-p}(F_3) - (1+F_3)^2 U_{2,p+4,n-p}(F_3) \\
& \quad + (1+F_3)U_{0,p+4,n-p}(F_3)] \sum_{\substack{j,k=1,2,\dots,p \\ \text{and } j \neq k}} a_{ijj} a_{ikk}]
\end{aligned}$$

and

(3.3.56)

$$I_5(\theta_0, \sigma_0) \approx (1-\alpha) + \sigma_0^2 \chi_{n-p+2}^2 [\chi_{n-p,\alpha}^2] \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p [2a_{ijk}^2 - a_{ijj} a_{ikk}] ,$$

where

$$F_1 = \frac{F_\alpha(1, n-p)}{n-p} ,$$

$$F_2 = \frac{k^* F_\alpha(k^*, n-p)}{n-p} ,$$

$$F_3 = \frac{p F_\alpha(p, n-p)}{n-p} ,$$

$$U_{k_1, k_2, k_3}(v) = E_{z_{i_1}, z_{i_2} \dots z_{i_{k_3}}} [z_{i_1}^{k_1} z_{i_2}^{k_2} \dots z_{i_{k_3}}^{k_3} (v \sum_{j^*=1}^{k_3} z_{i_j}^{k_3})] ,$$

$$(k_1 = 0, 1, 2, \dots; k_2, k_3 = 1, 2, \dots; j = 1, 2, \dots, k_3) ,$$

$$(3.3.57) \quad U_{0, k_2, k_3}(v) = \frac{\frac{k_2-2}{2} v^{\frac{k_2+k_3-2}{2}} \Gamma(\frac{k_2+k_3-2}{2})}{2(1+v)^{\frac{k_2+k_3-2}{2}} \Gamma(\frac{k_2}{2}) \Gamma(\frac{k_3}{2})} ,$$

$$(3.3.58) \quad U_{2, k_2, k_3}(v) = \frac{\frac{k_2-2}{2} v^{\frac{k_2+k_3-2}{2}} \Gamma(\frac{k_2+k_3-2}{2})}{k_3(1+v)^{\frac{k_2+k_3}{2}} \Gamma(\frac{k_2}{2}) \Gamma(\frac{k_3}{2})} ,$$

and v is a constant.

We further find, after some reduction, that

(3.3.59)

$$I_3(\theta_0, \sigma_0) \approx (1-\alpha) + \sigma_0^2 U_{0, 3, n-p}(F_1) \left[\sum_{j=1}^{p-1} \sum_{k=1}^{p-1} (2a_{pjk}^2 - a_{pj} a_{pk}) \right]$$

$$- \frac{1}{n-p} \sum_{i=p+1}^n \sum_{j=1}^{p-1} \sum_{k=1}^{p-1} (2a_{ijk}^2 - a_{ij} a_{ik})$$

$$- \frac{n-p+1}{n-p} \sum_{i=p+1}^n [a_{ipp}^2 + 2(\sum_{j=1}^{p-1} 2a_{ijp}^2 - a_{ij} a_{ip})] ,$$

$(p \geq 2; k^*=1)$

(3.3.60)

$$\begin{aligned}
 I_3(\theta_0, \sigma_0) &\approx (1-\alpha) + \sigma_0^2 U_{0,k^*+2,n-p}(F_2) \left[\sum_{i=p-k^*+1}^p \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} (2a_{ijk}^2 - a_{ijj}a_{ikk}) \right. \\
 &\quad - \frac{k^*}{n-p} \sum_{i=p+1}^n \sum_{j=1}^{p-k^*} \sum_{k=1}^{p-k^*} (2a_{ijk}^2 - a_{ijj}a_{ikk}) \\
 &\quad - \frac{n-p+k^*}{n-p} \sum_{i=p+1}^n \left[\sum_{j=p-k^*+1}^p \sum_{k=p-k^*+1}^p (2a_{ijk}^2 - a_{ijj}a_{ikk}) \right. \\
 &\quad \left. \left. + 2 \sum_{j=1}^{p-k^*} \sum_{k=p-k^*+1}^p (2a_{ijk}^2 - a_{ijj}a_{ikk}) \right] \right] \\
 &\quad , \quad (p \geq 3; k^* \geq 2),
 \end{aligned}$$

and

(3.3.61)

$$I_4(\theta_0, \sigma_0) \approx (1-\alpha) - \frac{n}{n-p} \sigma_0^2 U_{0,p+2,n-p}(F_3) \sum_{i=p+1}^n \sum_{j=1}^p \sum_{k=1}^p (2a_{ijk}^2 - a_{ijj}a_{ikk}).$$

Note that the approximations of $I_2(\theta_0, \sigma_0)$, $I_4(\theta_0, \sigma_0)$ and $I_5(\theta_0, \sigma_0)$ given by (3.3.52), (3.3.61) and (3.3.56) respectively agree with the corresponding approximations derived in Beale (1960).

Section 3.4 Power functions of the general m.l. ratio tests

This section is concerned with the derivation of approximations of the power functions $\beta_i(\theta_A, \sigma_A)$ (c.f. section 3.2). Note that $\sigma_A = \sigma_0$ if $i = 1$ and 2 , and $\sigma_A \neq \sigma_0$ if $i = 5$.

The acceptance regions

$$\omega_i = \{z : s_i^D(z) \leq d_i^{*2}\} , \quad (i = 1, 2, \dots, 5)$$

defined in section 3.2 can also be written as

$$\omega_i = \{z : s_i^{DA}(z) \leq d_i^{+2}\} , \quad (i = 1, 2, \dots, 5),$$

where

$$s_1^{DA}(z) = [s^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}) - s(\hat{\theta})]/\sigma_A^2,$$

$$s_2^{DA}(z) = [s(\theta_0) - s(\hat{\theta})]/\sigma_A^2,$$

$$s_3^{DA}(z) = [s^M(\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}) - (1 + \frac{k^* F_\alpha(k^*, n-p)}{n-p}) s(\hat{\theta})]/\sigma_A^2$$

$$+ \frac{k^* F_\alpha(k^*, n-p)}{n-p} \sum_{j=p+1}^n z_j^2/\sigma_A^2,$$

$$s_4^{DA}(z) = [s(\theta_0) - (1 + \frac{p F_\alpha(p, n-p)}{n-p}) s(\hat{\theta})]/\sigma_A^2 + \frac{p F_\alpha(p, n-p)}{n-p} \sum_{j=p+1}^n z_j^2/\sigma_A^2 ,$$

$$s_5^{DA}(z) = s(\hat{\theta})/\sigma_A^2 ,$$

and

$$d_i^+ = (d_i^+)^2 = \begin{cases} \chi_{k^*, \alpha}^2 & , (i = 1) \\ \chi_{p, \alpha}^2 & , (i = 2) \\ \frac{k^* F_\alpha(k^*, n-p)}{n-p} \sum_{j=p+1}^n z_j^2 / \sigma_A^2 & , (i = 3) \\ \frac{p F_\alpha(p, n-p)}{n-p} \sum_{j=p+1}^n z_j^2 / \sigma_A^2 & , (i = 4) \\ \frac{\sigma_0^2}{\sigma_A^2} \chi_{n-p, \alpha}^2 & , (i = 5). \end{cases}$$

The corresponding approximations of $S_i^{DA}(z)$ ($i = 1, 2, \dots, 5$) can be expressed respectively in terms of the right hand side of (3.3.12)-(3.3.16) with σ_0 , a_{ijk}^* , a_{ijkl}^* , a_{MCijkl} and a_{MFijkl} changed to σ_A , a_{ijk}^+ , a_{ijkl}^+ , $a_{MCAijkl}$ and $a_{MFAijkl}$ respectively.

Let η_A be the $(nx1)$ vector whose u^{th} component is $\eta(\xi_u, \theta_A)$. Given that $\theta = \theta_A$ and $\sigma = \sigma_A$, z_j / σ_A ($j = 1, 2, \dots, n$) are independently distributed as $N(z_{Aj}, 1)$, where z_{Aj} is the j^{th} component of $\frac{1}{\sigma_A} H(\eta_A - \eta_0)$. As far as obtaining approximations of the $\beta_i(\theta_A, \sigma_A)$ is concerned, we can set σ_A appearing in the expressions of the $S_i^{DA}(z)$ to be one while leaving the σ_A appearing in the expressions of the $d_i^+, a_{ijk}^+, a_{ijkl}^+, a_{MCAijkl}$ and $a_{MFAijkl}$ unchanged, and treat the z_j as being distributed as $N(z_{Aj}, 1)$. Each of the $z_j^{(s)}$ is then distributed as non-central χ^2 with one degree of freedom and parameter $\lambda_j = z_{Aj}^2$. Conditional on z_j being negative, i.e. $s_j = -1$ (or non-negative, i.e. $s_j = +1$), the probability density function (p.d.f.) of $z_j^{(s)}$ is

$$\frac{1}{2\sqrt{2\pi}} \frac{\exp[-\frac{1}{2}(s_j \sqrt{z_j(s)} - z_{Aj})^2]}{\sqrt{z_j(s)}} \left/ \int_{z_j(s)=0}^{\infty} \frac{1}{2\sqrt{2\pi}} \frac{\exp[-\frac{1}{2}(s_j \sqrt{z_j(s)} - z_{Aj})^2]}{\sqrt{z_j(s)}} dz_j(s) \right.$$

The power functions $\beta_i(\theta_A, \sigma_A)$ can now be written as

(3.4.1)

$$\beta_i(\theta_A, \sigma_A) = 1 - E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{z_{p-k^*+1}(s)} \int_{z_{p-k^*+2}(s)} \dots \int_{z_p(s)} \left[\prod_{j=p-k^*+1}^p \left[\frac{1}{2\sqrt{2\pi}} \frac{\exp[-\frac{1}{2}(s_j \sqrt{z_j(s)} - z_{Aj})^2]}{\sqrt{z_j(s)}} \right] \right] \right]$$

$$(z_{p-k^*+1}^{(s)}, z_{p-k^*+2}^{(s)}, \dots, z_p^{(s)}) \in Z_i^+$$

$$\left. dz_{p-k^*+1}^{(s)} dz_{p-k^*+2}^{(s)} \dots dz_p^{(s)} \right\}$$

, (i = 1 and 3),

(3.4.2)

$$\beta_i(\theta_A, \sigma_A) = 1 - E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{z_1^{(s)}} \int_{z_2^{(s)}} \dots \int_{z_p^{(s)}} \left[\prod_{j=1}^p \left[\frac{1}{2\sqrt{2\pi}} \frac{\exp[-\frac{1}{2}(s_j \sqrt{z_j^{(s)}} - z_{Aj})^2]}{\sqrt{z_j^{(s)}}} \right] dz_1^{(s)} dz_2^{(s)} \dots dz_p^{(s)} \right] \right], \quad (i = 2 \text{ and } 4),$$

and

(3.4.3)

$$\beta_5(\theta_A, \sigma_A) = 1 - E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \left[\int_{z_{p+1}^{(s)}} \int_{z_{p+2}^{(s)}} \dots \int_{z_n^{(s)}} \left[\prod_{j=p+1}^n \left[\frac{1}{2\sqrt{2\pi}} \frac{\exp[-\frac{1}{2}(s_j \sqrt{z_j^{(s)}} - z_{Aj})^2]}{\sqrt{z_j^{(s)}}} \right] dz_{p+1}^{(s)} dz_{p+2}^{(s)} \dots dz_n^{(s)} \right] \right] \right\},$$

where z_i^+ are given by the right hand side of (3.3.20)-(3.3.22) with s_i^D and d_i^* changed to s_i^{DA} and d_i^+ respectively.

After applying the transformations given by (3.3.24)-(3.3.29), the power functions $\beta_i(\theta_A, \sigma_A)$ can be written as

(3.4.4)

$$\beta_i(\theta_A, \sigma_A) = 1 - E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1}$$

$$\left[\int_{\bar{z}_{p-k^*+1}^{(s)}=0}^1 \int_{\bar{z}_{p-k^*+2}^{(s)}=\bar{z}_{p-k^*+1}^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1$$

$$\chi_1^2(\bar{z}_{p-k^*+1}^{(s)}) \chi_1^2(\bar{z}_{p-k^*+2}^{(s)}) \dots \chi_1^2(\bar{z}_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=p-k^*+1}^{p-1} \bar{z}_j^{(s)})$$

$$\times \left[\int_{R_i^+} e^{1/2} e^{-r(s)/2} [r(s)]^{\frac{k^*}{2}-1} \prod_{k=p-k^*+1}^p \frac{1}{2} e^{-z_{Ak}^2/2} e^{s_k \sqrt{r(s) - z_k^{(s)}}} z_{Ak} \right] dr(s)$$

$$\times \left. \frac{dz_{p-k^*+1}^{(s)}}{dz_{p-k^*+2}^{(s)}} \frac{dz_{p-k^*+2}^{(s)}}{dz_{p-1}^{(s)}} \dots \frac{dz_{p-1}^{(s)}}{} \right] \left. \right\}$$

(i = 1 and 3),

(3.4.5)

$$\beta_i(\theta_A, \sigma_A) = 1 - E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_p=-1,+1}$$

$$\left[\int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \dots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \chi_1^2(\bar{z}_1^{(s)}) \chi_1^2(\bar{z}_2^{(s)}) \dots \chi_1^2(\bar{z}_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=1}^{p-1} \bar{z}_j^{(s)}) \right]$$

$$\times \left[\int_{R_i^+} e^{1/2} e^{-r(s)/2} [r(s)]^{\frac{p}{2}-1} \left[\prod_{k=1}^p \frac{1}{2} e^{-z_{Ak}^2/2} s_k \sqrt{r(s) z_k(s)} z_{Ak} \right] dr(s) \right]$$

$$\times dz_1^{(s)} dz_2^{(s)} \dots dz_{p-1}^{(s)} \Bigg] \Bigg\}, \quad (i = 2 \text{ and } 4),$$

(3.4.6)

$$\beta_5(\theta_A, \sigma_A) = 1 - E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\int_{\bar{z}_{p+1}(s)=0}^1 \int_{\bar{z}_{p+2}(s)=z_{p+1}}^1 \dots \int_{\bar{z}_{n-1}(s)=z_{n-2}}^1$$

$$x_1^2(z_{p+1}) x_1^2(z_{p+2}) \dots x_1^2(z_{n-1}) x_1^2(1 - \sum_{j=p+1}^{n-1} \bar{z}_j(s))$$

$$\times \left[\int_{R_5^+} e^{1/2} e^{-r(s)/2} [r(s)]^{\frac{n-p}{2}-1} \left[\prod_{k=p+1}^n \frac{1}{2} e^{-z_{Ak}^2/2} s_k \sqrt{r(s) z_k(s)} z_{Ak} \right] dr(s) \right]$$

$$\times dz_{p+1}^{(s)} dz_{p+2}^{(s)} \dots dz_{n-1}^{(s)} \Bigg] \Bigg\}$$

where R_i^+ are given by the right hand side of (3.3.33)-(3.3.35) with s_i^D and d_i^* changed to s_i^{DA} and d_i^+ respectively.

We can obtain a truncated series expansion of the power function $\beta_i(\theta_A, \sigma_A)$ as follows:

$$(3.4.7) \quad \beta_i(\theta_A, \sigma_A) \approx \alpha_A + \sum_{j=1}^{n^*} \beta_{ia_j^+} a_j^+ + \frac{1}{2} \sum_{j=1}^{n^*} \sum_{k=1}^{n^*} \beta_{ia_j^+ a_k^+} a_j^+ a_k^+,$$

where $\{a_1^+, a_2^+, \dots, a_{n^*}^+\}$ is the set containing all the a_{hjk}^+ ,

$$\alpha_A = [\beta_i(\theta_A, \sigma_A)]_{\underline{a}^+ = 0},$$

$$\beta_{ia_j^+} = \left[\frac{\partial \beta_i(\theta_A, \sigma_A)}{\partial a_j^+} \right]_{\underline{a}^+ = 0},$$

$$\beta_{ia_j^+ a_k^+} = \left[\frac{\partial^2 \beta_i(\theta_A, \sigma_A)}{\partial a_j^+ \partial a_k^+} \right]_{\underline{a}^+ = 0}$$

and \underline{a}^+ is a column vector whose components are the a_{hjk}^+ , and the $a_{hjk\ell}^+$, $a_{MCAhjk\ell}$ or $a_{MFAhjk\ell}$ for which $j \leq k \leq \ell$.

We now consider how $\beta_{ia_j^+}$ and $\beta_{ia_j^+ a_k^+}$ can be derived. We first note that provided that the magnitude $|\underline{a}^+|$ of \underline{a}^+ is sufficiently small, the sets R_i^+ can be written as

$$R_i^+ = \{r^{(s)} = 0 \leq r^{(s)} \leq r_i^{(+s)}\},$$

where $r_i^{(+s)}$ are given by (3.3.37)-(3.3.39) with s_i^D , $r_i^{(*s)}$ and d_i^* changed to s_i^{DA} , $r_i^{(+s)}$ and d_i^+ respectively. We next note that the equations defining the $r_i^{(+s)}$ are of the form

$$(3.4.8) \quad r_i^{(+s)} + \sum_{j=1}^{n^k} h_j^+(r_i^{(+s)}) a_j^+ + \sum_{j=1}^{n^*} \sum_{k=1}^{n^*} h_{jk}^+(r_i^{(+s)}) a_j^+ a_k^+ + b^+ = d_i^{+2},$$

where $h_j^+(r_i^{(+s)})$ and $h_{jk}^+(r_i^{(+s)})$ are functions of $r_i^{(+s)}$, the z_ℓ and $\bar{z}_m^{(s)}$, and b^+ is a function of the z_ℓ and $\bar{z}_m^{(s)}$, and the a_{ijkl}^+ , $a_{MCAijkl}$ or $a_{MFAijkl}$.

As for $r_i^{(*s)}$ in section 3.3, we have

$$(3.4.9) \quad \left[\frac{\partial r_i^{(+s)}}{\partial a_j^+} \right]_{a^+=0} = -h_j^+(d_i^{+2})$$

and

$$(3.4.10) \quad \begin{aligned} \left[\frac{\partial^2 r_i^{(+s)}}{\partial a_j^+ \partial a_k^+} \right]_{a^+=0} &= - \left[\frac{\partial h_j^+(r_i^{(+s)})}{\partial r_i^{(+s)}} \frac{\partial r_i^{(+s)}}{\partial a_k^+} \right]_{a^+=0} \\ &- \left[\frac{\partial h_k^+(r_i^{(+s)})}{\partial r_i^{(+s)}} \frac{\partial r_i^{(+s)}}{\partial a_j^+} \right]_{a^+=0} - 2h_{jk}^+(d_i^{+2}). \end{aligned}$$

Using the arguments similar to those in deriving $I_{ia_j^*}$ and $I_{ia_j^* a_k^*}$, we obtain

(3.4.11)

$$\beta_{ia_j^*} = -E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k^*}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{\bar{z}_{p-k^*+1}^{(s)}=0}^1 \int_{\bar{z}_{p-k^*+2}^{(s)}=\bar{z}_{p-k^*+1}}^1 \cdots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}}^1 \right]$$

$$\chi_1^2(\bar{z}_{p-k^*+1}^{(s)}) \chi_1^2(\bar{z}_{p-k^*+2}^{(s)}) \cdots \chi_1^2(\bar{z}_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=p-k^*+1}^{p-1} \bar{z}_j^{(s)})$$

$$\times e^{1/2} e^{-d_i^{+2}/2} [d_i^{+2}]^{\frac{k^*}{2}-1} \left[\prod_{k=p-k^*+1}^p \frac{1}{2} e^{-z_{Ak}^2/2} s_k d_i^+ \sqrt{z_k^{(s)}} z_{Ak} \right] \left[\frac{\partial r_i^{(+s)}}{\partial a_j^+} \right]_{a^+=0}$$

$$\times dz_{p-k^*+1}^{(s)} dz_{p-k^*+2}^{(s)} \cdots dz_{p-1}^{(s)} \Bigg\} , \quad (i = 1 \text{ and } 3),$$

(3.4.12)

$$B_{ia_j^+} = - E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \cdots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \cdots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \right]$$

$$\chi_1^2(z_1^{(s)}) \chi_1^2(z_2^{(s)}) \cdots \chi_1^2(z_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=1}^{p-1} \bar{z}_j^{(s)})$$

$$\times e^{1/2} e^{-d_i^{+2}/2} [d_i^{+2}]^{\frac{p}{2}-1} \left[\prod_{k=1}^p \frac{1}{2} e^{-z_{Ak}^2/2} s_k d_i^+ \sqrt{z_k^{(s)}} z_{Ak} \right] \left[\frac{\partial r_i^{(+s)}}{\partial a_j^+} \right]_{a^+=0}$$

$$\times dz_1^{(s)} dz_2^{(s)} \cdots dz_{p-1}^{(s)} \Bigg\} , \quad (i = 2 \text{ and } 4),$$

(3.4.13)

$$\beta_{5a_j^+} = - E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\int_{\frac{-z(s)}{z_{p+1}}=0}^1 \int_{\frac{-z(s)}{z_{p+2}}=z_{p+1}}^1 \dots \int_{\frac{-z(s)}{z_{n-1}}=z_{n-2}}^1 \right]$$

$$x_1^2(z_{p+1}^{-}(s)) x_1^2(z_{p+2}^{-}(s)) \dots x_1^2(z_{n-1}^{-}(s)) x_1^2(1 - \sum_{j=p+1}^{n-1} \frac{-z(s)}{z_j})$$

$$\times e^{1/2} e^{-d_5^{+2}/2} [d_5^{+2}]^{\frac{n-p}{2}-1} \left[\prod_{k=p+1}^n \frac{1}{2} e^{-z_{Ak}^2/2} e^{s_k d_5^{+} \sqrt{z_k^{-}(s)} z_{Ak}} \right] \left[\frac{\partial r_5^{(+s)}}{\partial a_j^+} \right]_{a_j^+=0}$$

$$\left. \times \frac{dz_{p+1}^{-}(s)}{dz_{p+2}^{-}(s)} \dots \frac{dz_{n-1}^{-}(s)}{} \right] \left. \right\}$$

(3.4.14)

$$\beta_{1a_j^+ a_k^+} = - E_{z_{p+1}, z_{p+2}, \dots, z_n, z_1, z_2, \dots, z_{p-k}}$$

$$\left\{ \sum_{s_{p-k^*+1}=-1,+1} \sum_{s_{p-k^*+2}=-1,+1} \dots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{\bar{z}_{p-k^*+1}^{(s)}=0}^1 \int_{\bar{z}_{p-k^*+2}^{(s)}=\bar{z}_{p-k^*+1}}^1 \cdots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}}^1 \right]$$

$$\chi_1^2(\bar{z}_{p-k^*+1}^{(s)}) \chi_1^2(\bar{z}_{p-k^*+2}^{(s)}) \cdots \chi_1^2(\bar{z}_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=p-k^*+1}^{p-1} \bar{z}_j^{(s)})$$

$$\times e^{1/2} e^{-d_i^{+2}/2} [d_i^{+2}]^{\frac{k^*}{2}-1} \left[\prod_{k=p-k^*+1}^p \frac{1}{2} e^{-z_{Ak}^2/2} s_k d_i^{+\sqrt{z_k^{(s)}} z_{Ak}} \right]$$

$$\times \left[\left[\left(\frac{k^*}{2} - 1 \right) \frac{1}{d_i^{+2}} - \frac{1}{2} + \frac{1}{2d_i^+} \right] \sum_{\ell=p-k^*+1}^p s_\ell \sqrt{z_\ell^{(s)}} z_{A\ell} \right] \left[\frac{\partial r_i^{(+s)}}{\partial a_j^+} \frac{\partial r_i^{(+s)}}{\partial a_k^+} \right]_{\underline{a}^+=0}$$

$$+ \left[\frac{\partial^2 r_i^{(+s)}}{\partial a_j^+ \partial a_k^+} \right]_{\underline{a}^+=0}$$

$$\times dz_{p-k^*+1}^{(s)} dz_{p-k^*+2}^{(s)} \cdots dz_{p-1}^{(s)} \Bigg] \Bigg\} , \quad (i = 1 \text{ and } 3),$$

(3.4.15)

$$\beta_{ia_j^+ a_k^+} = - E_{z_{p+1}, z_{p+2}, \dots, z_n}$$

$$\left\{ \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \cdots \sum_{s_p=-1,+1} \right.$$

$$\left[\int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \cdots \int_{\bar{z}_{p-1}^{(s)}=\bar{z}_{p-2}^{(s)}}^1 \chi_1^2(z_1^{(s)}) \chi_1^2(z_2^{(s)}) \cdots \chi_1^2(z_{p-1}^{(s)}) \chi_1^2(1 - \sum_{j=1}^{p-1} \bar{z}_j^{(s)}) \right]$$

$$\times e^{1/2} e^{-d_i^{+2}/2} [d_i^{+2}]^{\frac{p}{2}-1} \left[\prod_{k=1}^{\frac{p}{2}} \frac{1}{2} e^{-z_{Ak}^2/2} e^{s_k d_i^{+} \sqrt{-z_k^{(s)}} z_{Ak}} \right]$$

$$\times \left[\left[\left(\frac{p}{2} - 1 \right) \frac{1}{d_i^{+2}} - \frac{1}{2} + \frac{1}{2d_i^{+}} \right] \sum_{\ell=1}^{\frac{p}{2}} s_{\ell} \sqrt{z_{\ell}^{(s)}} z_{A\ell} \right] \left[\frac{\partial r_i^{(+s)}}{\partial a_j^*} \frac{\partial r_i^{(+s)}}{\partial a_k^*} \right]_{a^+=0}$$

$$+ \left[\frac{\partial^2 r_i^{(+s)}}{\partial a_j^+ \partial a_k^+} \right]_{a^+=0}$$

$$\times dz_1^{-(s)} dz_2^{-(s)} \dots dz_{p-1}^{-(s)} \left. \right\} , \quad (i = 2 \text{ and } 4)$$

and

(3.4.16)

$$\beta_{5a_j^+ a_k^+} = - E_{z_1, z_2, \dots, z_p}$$

$$\left\{ \sum_{s_{p+1}=-1,+1} \sum_{s_{p+2}=-1,+1} \dots \sum_{s_n=-1,+1} \right.$$

$$\left[\int_{\bar{z}_{p+1}^{(s)}=0}^1 \int_{\bar{z}_{p+2}^{(s)}=\bar{z}_{p+1}^{(s)}}^1 \dots \int_{\bar{z}_{n-1}^{(s)}=\bar{z}_{n-2}^{(s)}}^1 \chi_1^2(\bar{z}_{p+1}^{(s)}) \chi_1^2(\bar{z}_{p+2}^{(s)}) \dots \chi_1^2(\bar{z}_{n-1}^{(s)}) \chi_1^2(1 - \sum_{j=p+1}^{n-1} \bar{z}_j^{(s)}) \right]$$

$$\times e^{1/2} e^{-d_5^{+2}/2} [d_5^{+2}]^{\frac{n-p}{2}-1} \left[\prod_{k=p+1}^n \frac{1}{2} e^{-z_{Ak}^2/2} e^{s_k d_5^{+} \sqrt{-z_k^{(s)}} z_{Ak}} \right]$$

$$\times \left[\left[\left(\frac{n-p}{2} - 1 \right) \frac{1}{d_5^{+2}} - \frac{1}{2} + \frac{1}{2d_5^{+}} \right] \sum_{\ell=p+1}^n s_{\ell} \sqrt{z_{\ell}^{(s)}} z_{A\ell} \right] \left[\frac{\partial r_5^{(+s)}}{\partial a_j^+} \frac{\partial r_5^{(+s)}}{\partial a_k^+} \right]_{a^+=0}$$

$$+ \left[\frac{\partial^2 r_5^{(+s)}}{\partial a_j^+ \partial a_k^+} \right]_{a^+=0}$$

$$\times \left. dz_{p+1}^{(s)} dz_{p+2}^{(s)} \dots dz_{n-1}^{(s)} \right] \}.$$

In deriving $\beta_{ia_j^+}$ and $\beta_{ia_j^+ a_k^+}$, we can make use of

$$(3.4.17) \quad E_{z_j}(z_j) = z_{Aj},$$

$$(3.4.18) \quad E_{z_j}(z_j^2) = 1 + z_{Aj}^2,$$

$$(3.4.19) \quad E_{z_j}(z_j^3) = z_{Aj}(3 + z_{Aj}^2),$$

$$(3.4.20) \quad E_{z_j}(z_j^4) = 3 + 6z_{Aj}^2 + z_{Aj}^4,$$

$$(3.4.21) \quad \sum_{s_1=-1,+1} \sum_{s_2=-1,+1} \dots \sum_{s_k=-1,+1} \int_{\bar{z}_1^{(s)}=0}^1 \int_{\bar{z}_2^{(s)}=\bar{z}_1^{(s)}}^1 \dots \int_{\bar{z}_{k-1}^{(s)}=\bar{z}_{k-2}^{(s)}}^1$$

$$x_1^2(\bar{z}_1^{(s)}) x_2^2(\bar{z}_2^{(s)}) \dots x_k^2(\bar{z}_k^{(s)})$$

$$\times e^{1/2} e^{-d^{+2}/2} [d^{+2}]^{\frac{k}{2}-1} \left[\prod_{j=1}^k \frac{1}{2} e^{-z_{Aj}^2/2} e^{s_j d^+ \sqrt{\bar{z}_j^{(s)}} z_{Aj}} \right]$$

$$\times \left[\prod_{a=1}^b \left[\bar{z}_{i_a}^{(s)} \right]^{\frac{m_a}{2}} \right] \left[\prod_{c=1}^d \left[s_{j_c} \sqrt{\bar{z}_{j_c}^{(s)}} \right]^{\frac{n_c}{2}} \right]$$

$$\times dz_1^{(s)} dz_2^{(s)} \dots dz_{k-1}^{(s)}$$

$$\begin{aligned}
 &= \sum_{i_1^*=0}^{\infty} \sum_{i_2^*=0}^{\infty} \dots \sum_{i_k^*=0}^{\infty} \left[\prod_{c=1}^d (z_{A_j c})^{1-\delta_{0n_c}} \right] [d^+] - \sum_{a=1}^b m_a - \sum_{c=1}^d n_c \\
 &\quad \times \left[\prod_{e=1}^k 2x_{2+2i_e^*}^2 (z_{A_e}^2) \right] \\
 &\quad \times \left[x_{k+\sum_{a=1}^b m_a + \sum_{c=1}^d (n_c + 1 - \delta_{0n_c}) + 2 \sum_{e=1}^k i_e^*}^{(d+2)} \right] \left[\prod_{a=1}^b p_a \right] \left[\prod_{c=1}^d q_c \right]
 \end{aligned}$$

where

$$\bar{z}_k^{(s)} = 1 - \sum_{j=1}^{k-1} \bar{z}_j^{(s)},$$

$$k = 2, 3, \dots,$$

$$b, d = 1, 2, \dots, k,$$

$i_1, i_2, \dots, i_b, j_1, j_2, \dots, j_d$ are distinct elements of $\{1, 2, \dots, k\}$,

$m_a = 0$ or even positive integer,

$n_c = 0$ or odd positive integer,

$$p_a = \begin{cases} 1 & \text{if } m_a = 0 \\ m_a/2 & \\ \prod_{f=1}^a (2f-1+2i_f^*) & \text{if } m_a = 2, 4, 6, \dots, \end{cases}$$

and

$$q_c = \begin{cases} 1 & \text{if } n_c = 0 \text{ and } 1 \\ (n_c+1)/2 & \\ \prod_{g=2}^{c-1} (2g-1+2i_g^*) & \text{if } n_c = 3, 5, 7, \dots; \end{cases}$$

(3.4.22)

$$\sum_{i_1^*=0}^{\infty} \sum_{i_2^*=0}^{\infty} \dots \sum_{i_k^*=0}^{\infty} \left[\prod_{e=1}^k 2x_{2+2i_e^*}^2 (z_{A_e}^2) \right] x_{k'+2 \sum_{e=1}^k i_e^*}^{(d+2)} = x_{k'}^2, \sum_{e=1}^k z_{A_e}^2 (d+2),$$

where $k = 1, 2, \dots$ and $k' \geq k$;

(3.4.23)

$$\sum_{i_1^*=0}^{\infty} \sum_{i_2^*=0}^{\infty} \dots \sum_{i_k^*=0}^{\infty} \left[\prod_{s=1}^l i_{j_s}^* (i_{j_s}^* - 1) \dots (i_{j_s}^* - r_{j_s}) \right] \left[\prod_{e=1}^k 2x_{2+2i_e^*}^2 (z_{Ae}^2) \right] x_{k'+2}^2 \sum_{e=1}^k i_e^* (d^{+2})$$

$$= \left[\prod_{s=1}^l \frac{z_{Aj_s}^2}{2} \right] x_{k'+2}^2 \sum_{s=1}^l (r_{j_s} + 1), \sum_{e=1}^k z_{Ae}^2 (d^{+2}),$$

where

$$k = 2, 3, \dots,$$

$$k' \geq k,$$

$$l = 1, 2, \dots, k,$$

j_1, j_2, \dots, j_l are distinct elements of $\{1, 2, \dots, k\}$

and $r_{j_s} = 0, 1, 2, \dots ;$

$$(3.4.24) \quad \int_{r(s)=0}^{\infty} [r(s)]^{\frac{k_0}{2}} x_{k_1, \lambda_1}^2 (r(s)) x_{k_2, \lambda_2}^2 (vr(s)) dr(s)$$

$$= \sum_{i_1=0}^{\infty} \sum_{i_2=0}^{\infty} 2^{\frac{k_0+1}{2}} x_{2+2i_1}^2 (\lambda_1) x_{2+2i_2}^2 (\lambda_2) \frac{\frac{k_2+2i_2-2}{2}}{(1+v)^{\frac{k_0+k_1+k_2+2i_1+2i_2-2}{2}}}$$

$$\times \frac{\Gamma(\frac{k_0+k_1+k_2+2i_1+2i_2-2}{2})}{\Gamma(\frac{k_1+2i_1}{2}) \Gamma(\frac{k_2+2i_2}{2})},$$

where k_0 is an integer,

$k_1, k_2 = 1, 2, \dots$

λ_1, λ_2 are the parameters of the non-central χ^2 distributions,

and v is a constant.

The evaluation of the approximations of $\beta_i(\theta_A, \sigma_A)$ using a computer will be considered in Chapter 4.

Section 3.5 Region estimation in unconstrained nonlinear models

Consider the problem of finding region estimates for

- (1) $\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p$, where $1 \leq k^* < p$, when σ is known to be equal to σ_0 ,
 - (2) $\underline{\theta}$ when σ is known to be equal to σ_0 ,
 - (3) $\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p$, where $1 \leq k^* < p$, when σ is unknown,
 - (4) $\underline{\theta}$ when σ is unknown,
- and
- (5) σ^2 when $\underline{\theta}$ is unknown.

For each $i = 1, 2, 3, 4$, let $R_i(z)$ be the totality of values

$\theta_{Op-k^*+1}, \theta_{Op-k^*+2}, \dots, \theta_{Op}$ (where $1 \leq k^* < p$ if $i = 1, 3$ and $k^* = p$ if $i = 2, 4$) for which H_i is accepted when the general m.l. ratio test is carried out for the observed z ,

i.e.

$$\begin{aligned} R_1(z) &= \{\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p : S^M(\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p) - S(\hat{\theta}) \\ &\quad \leq \sigma_0^2 \chi_{k^*, \alpha}^2\}, \\ R_2(z) &= \{\underline{\theta} : S(\underline{\theta}) - S(\hat{\theta}) \leq \sigma_0^2 \chi_{p, \alpha}^2\}, \end{aligned}$$

$$R_3(z) = \{\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p : S^M(\theta_{p-k^*+1}, \theta_{p-k^*+2}, \dots, \theta_p) - s(\hat{\theta})$$

$$\leq \frac{k^*}{n-p} s(\hat{\theta}) F_\alpha(k^*, n-p)\},$$

and

$$R_4(z) = \{\theta : s(\theta) - s(\hat{\theta}) \leq \frac{p}{n-p} s(\hat{\theta}) F_\alpha(p, n-p)\}.$$

We next denote $R_5(z)$ to be the totality of values σ_0^2 for which H_5 is accepted when the general m.l. ratic test is carried out for the observed z , i.e.

$$R_5(z) = \{\sigma^2 : s(\hat{\theta}) \leq \sigma^2 \chi_{n-p, \alpha}^2\}.$$

Then $R_i(z)$ can be regarded as a region estimate for the parameters in (i). If the true value of σ is σ_T , then the region estimate $R_i(z)$ will cover the true values of the corresponding parameters with probability $I_i(\theta_T, \sigma_T)$. This probability is an important quantity associated with the region estimate $R_i(z)$. Usually we do not know this probability as we do not know θ_T and σ_T . In the case when σ_T is known to be equal to σ_0 , we could use $I_i(\hat{\theta}, \sigma_0)$ to estimate this probability. Information concerning the reliability of this estimate could be derived from $I_i(\theta_f, \sigma_0)$ where θ_f are feasible θ in the neighbourhood of $\hat{\theta}$. The estimation of the probability $I_i(\theta_T, \sigma_0)$ will be investigated in Chapter 5.

Section 3.6 Inference of functions of the parameter vector based
on general maximum likelihood ratios

In previous sections, we have considered the problem of making inference about subsets of components of the parameter vector $\underline{\theta}$ in an unconstrained nonlinear model. In practice we may also be interested in making inference about nonlinear functions of this parameter vector $\underline{\theta}$. For example, we may be interested in hypothesis testing and interval estimation concerning $\eta(\xi^*, \underline{\theta})$ where $\xi^* \neq \xi_u$ for all $u = 1, 2, \dots, n$. In general we can denote these functions of $\underline{\theta}$ by $\alpha_{p-k^*+1}(\underline{\theta}), \alpha_{p-k^*+2}(\underline{\theta}), \dots, \alpha_p(\underline{\theta})$, where $1 \leq k^* \leq p$. We shall restrict our attention to the $\alpha_i(\underline{\theta})$ which have the following properties:

- (a) $\alpha_i(\underline{\theta})$ are differentiable up to the third order,
- (b) if $k^* = p$, then $\underline{g} = \underline{g}(\underline{\theta}) = [\alpha_1(\underline{\theta}), \alpha_2(\underline{\theta}), \dots, \alpha_p(\underline{\theta})]^T$ is a one to one function of $\underline{\theta}$, and if $k^* < p$, then there exist differentiable functions $\alpha_i(\underline{\theta})$, where $1 \leq i \leq p-k^*$, such that the resulting \underline{g} is a one to one function of $\underline{\theta}$.

We now consider the problem of testing the hypothesis

$$H_i^{(\alpha)} \text{ that } (\underline{g}, \sigma) \in \Omega_{H_i}^{(\alpha)}$$

against the alternative

$$K_i^{(\alpha)} \text{ that } (\underline{g}, \sigma) \in \Omega_{K_i}^{(\alpha)}, \quad (i = 1, 2, 3, 4),$$

where $\Omega_{H_i}^{(\alpha)}$ and $\Omega_{K_i}^{(\alpha)}$ are respectively the same as Ω_{H_i} and Ω_{K_i} (cf. section 3.2) if we change $\underline{\theta}$ to \underline{g} and $\underline{\theta}_0$ to \underline{g}_0 . If we can find the expressions $\theta_i(\underline{g})$ for the θ_i in terms of \underline{g} , then the model is one in which

the theoretical means of y_u are $\eta(\xi_u, \theta(\alpha))$, where $\theta(\alpha) = [\theta_1(\alpha), \theta_2(\alpha), \dots, \theta_p(\alpha)]^T$, and the problem of testing the hypotheses $H_i^{(\alpha)}$ is similar to that of testing the hypotheses H_i . In practice, it is not necessary to find the expressions $\theta_i(\alpha)$ for the purpose of testing the hypotheses $H_i^{(\alpha)}$. We note that the appropriate quantities which need to be computed are $\theta(\alpha_0)$, $\hat{\theta}$, $S(\theta(\alpha_0))$, $s(\hat{\theta})$, and $S_{\alpha}^M(\alpha_{Op-k^*+1}, \alpha_{Op-k^*+2}, \dots, \alpha_{Op})$ which is the minimum value of $S(\theta(\alpha))$ with respect to α where α are such that $(\alpha, \sigma_0) \in \Omega_{H_1}^{(\alpha)}$. This could be achieved by using some technique of constrained and unconstrained minimization. The remaining quantities which need to be computed are the significance probabilities and the power of the general m.l. ratio tests. To derive approximations of these probabilities, it suffices to express $\eta(\xi_u, \theta(\alpha))$ as cubic functions of the $\alpha_j^* = \alpha_j - \alpha_{0j}$ similar to (3.3.1), as then the problem of deriving these approximations under the parameterization α is similar to that under the parameterization θ .

As $\alpha_i(\theta)$, where $p-k^*+1 \leq i \leq p$, are differentiable up to the third order, we can obtain cubic approximations of the corresponding α_i^* as follows:

$$(3.6.1) \quad \alpha_i^* = \sum_{j=1}^p b_{ij} t_j + t^T B_i t + \sum_{j=1}^p [t^T B_{ij} t] t_j + o(t^3),$$

$(i = p-k^*+1, p-k^*+2, \dots, p),$

where $B_i = \{b_{ijk}\}$ and $B_{ij} = \{b_{ijkl}\}$ are symmetric $(p \times p)$ matrices.

As there exist differentiable functions $\alpha_i(\theta)$, where $1 \leq i \leq p-k^*$, such that the resulting α is a one to one function of θ , we can find the numbers b_{ij} , where $1 \leq i \leq p-k^*$ and $1 \leq j \leq p$, such that the $(p \times p)$ matrix $B = \{b_{ij}\}$ is non-singular. After finding these numbers, we define

$$(3.6.2) \quad \alpha_i^* = \sum_{j=1}^p b_{ij} t_j, \quad (i = 1, 2, \dots, p-k^*).$$

Let $H_B = \{h_{Bjk}\}$ be a $(p \times p)$ orthogonal matrix such that $H_B \tilde{t}$ is an upper triangular $(p \times p)$ matrix $\tilde{E} = \{e_{ij}\}$. Next, let γ_i be the i^{th} component of

$$(3.6.3) \quad \gamma = H_B \alpha^*.$$

Further, let b_{jk} and b_{jkl} be the $(p \times 1)$ column vectors whose i^{th} components are b_{ijk} and b_{ijkl} respectively if $p-k^*+1 \leq i \leq p$, and zeros if $1 \leq i \leq p-k^*$.

Then from (3.6.1)-(3.6.3), we have

$$(3.6.4) \quad \gamma_i = \sum_{j=1}^p e_{ij} t_j + \tilde{t}^T E_i \tilde{t} + \sum_{j=1}^p [\tilde{t}^T E_i \tilde{t}] t_j + o(t^3), \quad (i = 1, 2, \dots, p),$$

where

$$E_i = \{e_{ijk}\},$$

$$\tilde{E}_{ij} = \{e_{ijkl}\}.$$

These e_{ijk} and e_{ijkl} are respectively the i^{th} components of $H_B b_{jk}$ and $H_B b_{jkl}$.

Let $\tau = E^{-1} \tilde{t}$. Then we have

$$(3.6.5) \quad \tau = E^{-1} \tilde{t},$$

where $E^{-1} = \{e^{ij}\}$ is the inverse of E . From (3.6.4), we have

$$(3.6.6) \quad \gamma_i = \tau_i + \tilde{t}^T R_i \tau + \sum_{j=1}^p [\tilde{t}^T R_i \tilde{t}] [\sum_{k=j}^p e^{jk} \tau_k] + o(\tau^3) \quad (i = 1, 2, \dots, p),$$

where

$$(3.6.7) \quad \underline{R}_i = (\underline{E}^{-1})^T \underline{E}_i (\underline{E}^{-1}) = \{r_{ijk}\},$$

$$(3.6.8) \quad \underline{R}_{ij} = (\underline{E}^{-1})^T \underline{E}_{ij} (\underline{E}^{-1}) = \{r_{ijkl}\},$$

and $\tau^3 = (\sum_{j=1}^p \gamma_j^2)^{3/2}$.

From (3.6.6) we have

$$(3.6.9) \quad \tau_i = \gamma_i - \gamma^T \underline{R}_i \gamma - \sum_{k=1}^p [\gamma^T (\sum_{j=1}^k e^{jk} \underline{R}_{ij}) - 2 \sum_{j=1}^p r_{ijk} \underline{R}_j] \gamma_k + o(\gamma^3)$$

$$, \quad (i = 1, 2, \dots, p),$$

where $\gamma^3 = (\sum_{j=1}^p \gamma_j^2)^{3/2}$.

Finally, by using (3.3.1), (3.6.5), (3.6.9) and (3.6.3), we can express the $n(\xi_u, \theta(\alpha))$ as cubic functions of the α_j^* . In what follows, we present these functions truncated after the quadratic terms:

$$(3.6.10) \quad n(\xi_u, \theta(\alpha)) = n(\xi_u, \theta(\alpha_0)) + \sum_{j=1}^p \left[\sum_{j_1=1}^p \sum_{j_2=j_1}^p c_{uj_1} e^{j_1 j_2} h_{Bj_2 j_1} \alpha_j^* \right.$$

$$\left. + (\alpha^*)^T [H_B^T (\underline{E}^{-1})^T C_u E^{-1} H_B - \sum_{j_1=1}^p \sum_{j_2=j_1}^p c_{uj_1} e^{j_1 j_2} H_B^T R_{j_2} H_B] \alpha^* + o((\alpha^*)^3) \right],$$

where $(\alpha^*)^3 = (\sum_{i=1}^p (\alpha_i^*)^2)^{3/2}$.

Section 3.7 Hypothesis testing and region estimation in constrained nonlinear models

In previous sections we have considered the problems of testing nonlinear hypotheses and obtaining region estimates in the case when the solution locus is unconstrained. In some models, for example models (A) and (B) described in Chapter 1, the solution loci are constrained. However, if the point $P(\underline{\theta}_T)$ is sufficiently far away from any boundary of a solution locus of these models, and the standard error σ of the observations is sufficiently small, then we can still treat this solution locus as being unconstrained, and obtain inference based on general maximum likelihood ratios. Thus if the number p of parameters is at most two, then we can first attempt to display the first p rotated coordinates of the points in the solution locus of a constrained model (c.f. Fig. (1.1.1) and (1.1.2)) in order to find out in what way the solution locus is constrained. We can then attempt to judge, from the point $P(\hat{\underline{\theta}})$ and the value $S(\hat{\underline{\theta}})/(n-p)$, whether the model could be treated as being unconstrained for statistical purposes.

In the case when we cannot treat the constrained nonlinear models as being unconstrained for statistical purposes, the problems of hypothesis testing and region estimation are still open questions for most of these models.

CHAPTER 4

DERIVATION OF APPROXIMATIONS OF THE POWER OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TESTS USING A COMPUTER

Section 4.1 Introduction

In Chapter 3, the power functions $\beta_i(\theta_A, \sigma_A)$ of the general maximum likelihood ratio tests for the various nonlinear hypotheses (c.f. section 3.2) are approximated by series expansions truncated after some finite number of terms (c.f. (3.4.7)). These truncated series expansions can also be written as

$$(4.1.1) \quad \beta_i(\theta_A, \sigma_A) \approx a_A + \beta_i^{(1)} \sigma_A + \beta_i^{(2)} \sigma_A^2 ,$$

where

$$\beta_i^{(1)} = \sum_{h=1}^n \sum_{j=1}^p \sum_{k=j}^p \beta_i a_{hjk}$$

and

$$\beta_i^{(2)} = \frac{1}{2} \sum_{h_1=1}^n \sum_{h_2=1}^n \sum_{j_1=1}^p \sum_{k_1=j_1}^p \sum_{j_2=1}^p \sum_{k_2=j_2}^p \beta_{ia} a_{h_1 j_1 k_1} a_{h_2 j_2 k_2}^+$$

In this chapter, we consider the case when $i = 1$ and 2 , i.e. the case in which the corresponding hypotheses are concerned with one or more components of the parameter vector θ when σ is known to be equal to σ_0 . An outline of how the approximate values of the corresponding $\beta_i(\theta_A, \sigma_A)$ can be calculated is as follows. Initially we have the model and the

values k^* , (θ_0, σ_0) and (θ_A, σ_A) , where in this case $\sigma_A = \sigma_0$. We can next calculate the values $\eta(\xi_u, \theta_A) - \eta(\xi_u, \theta_0)$ and the first and second derivatives $c_{uj}(\theta_0)$ and $c_{ujk}(\theta_0)$. From the values $\eta(\xi_u, \theta_A) - \eta(\xi_u, \theta_0)$ we can calculate z_A (c.f. section 3.4), and from the values $c_{uj}(\theta_0)$ and $c_{ujk}(\theta_0)$, we can calculate a_{hjk} (c.f. section 3.3), where $h = 1, 2, \dots, n$ and $j, k = 1, 2, \dots, p$. Then from the values σ_0 , $d_i^{+2} = x_{k^*, \alpha}^2$ or $x_{p, \alpha}^2$, z_A and the a_{hjk} , we can calculate $\beta_i^{(1)}$ and $\beta_i^{(2)}$ (c.f. (3.4.11)-(3.4.16)). With these values of $\beta_i^{(1)}$ and $\beta_i^{(2)}$, the approximate value of $\beta_i(\theta_A, \sigma_A)$ can be found if we further calculate α_A and the right hand side of (4.1.1).

For the purpose of illustrating how we can use a computer to do the algebraic manipulation and calculation involved in finding $\beta_i(\theta_A, \sigma_A)$, we shall not consider all the calculation outlined above. Instead we assume that we already know the values of

- (i) n, p, k^* ,
- (ii) σ_0 ,
- (iii) d_i^{+2}
- (iv) z_A
- and (v) the a_{ijk} ,

and we wish to calculate $\beta_i^{(1)}$ and $\beta_i^{(2)}$. There are two stages in the calculation of $\beta_i^{(1)}$ and $\beta_i^{(2)}$. First we derive algebraic expressions for the first two derivatives of $r_i^{(+s)}$ (c.f. (3.4.9) and (3.4.10)). The details of this derivation are described in sections 4.2-4.11, and the programs for actually deriving the derivatives of $r_i^{(+s)}$ are given in Appendix 3. We then make use of the derivatives of $r_i^{(+s)}$ and the equations (3.4.11)-(3.4.23) to calculate $\beta_i^{(1)}$ and $\beta_i^{(2)}$. The programs for deriving $\beta_i(\theta_A, \sigma_A)$ in terms of $\alpha_A, \beta_i^{(1)}, \beta_i^{(2)}$ and σ_A are given in Appendix 4.

Section 4.2 Representation of algebraic expressions by one dimensional arrays in a computer

We note that the algebraic expressions that we are dealing with in deriving the coefficients $\beta_{ia_j^+}$ and $\beta_{ia_j^+ a_k^+}$ are mainly sums of a number of terms each of which is of the form

$$\text{constant} \times \prod_{i=1}^m \beta_i^{l_i}$$

where m is a positive integer, l_i are integers and β_i are symbols which are common for all the terms. Each term can be represented in a computer by storing the corresponding constant, l_1, l_2, \dots, l_m in one dimensional arrays.

Section 4.3 Algebraic manipulation done on a computer

The addition of two terms with the same values of l_1, l_2, \dots, l_m can be achieved by adding together the constants of these terms and storing the sum of these constants, as well as l_1, l_2, \dots, l_m in one dimensional arrays.

The multiplication of two terms represented by

constant A, $l_{A1}, l_{A2}, \dots, l_{Am}$

and

constant B, $l_{B1}, l_{B2}, \dots, l_{Bm}$

can be achieved by calculating

constant = constant A \times constant B

$l_i = l_{Ai} + l_{Bi}$, ($i = 1, 2, \dots, m$),

and storing constant, $\ell_1, \ell_2, \dots, \ell_m$ in one dimensional arrays.

Section 4.4 Representation of the equation $S_1^{DA}(z) = d_1^{+2}$ in a computer

The case when $i = 1$ shall be chosen to illustrate the ideas behind the derivation of $\beta_{ia_j^+}$ and $\beta_{ia_j^+ a_k^+}$, where $i = 1, 2$. Suppose we are interested in finding $\beta_{1a_{i_1 j_1 k_1}^+}, \beta_{1a_{i_2 j_2 k_2}^+}$ and $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$. To find these coefficients, the equation

$$[S_1^{DA}(z)] = d_1^{+2} \quad \text{all components of } z^+ \text{ other than } a_{i_1 j_1 k_1}^+ \text{ and } a_{i_2 j_2 k_2}^+ \text{ are zero}$$

is first represented in a computer as follows

- (i) set II to be 1,
- (ii) find out the IIth term (in $S_1^{DA}(z)$) containing $a_{i_1 j_1 k_1}^+$, $a_{i_2 j_2 k_2}^+$ or $a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+$,
- (iii) a term found out in (ii) is of the form

$$\text{constant } z_1^{\ell_1} z_2^{\ell_2} \dots z_n^{\ell_n} \times a_{i_1 j_1 k_1}^{m_1} a_{i_2 j_2 k_2}^{m_2}$$

and can be represented in a computer by storing

constant in AC(II),

$\ell_1, \ell_2, \dots, \ell_n$ in AZ(II,1), AZ(II,2), ..., AZ(II,n)

and m_1, m_2 in AA(II,1), AA(II,2),

- (iv) Increase the current value of II by 1 and continue the process same as for II = 1 until all the terms (in $S_1^{DA}(z)$) containing $a_{i_1 j_1 k_1}^+, a_{i_2 j_2 k_2}^+$ or $a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+$ have been found out, and

represented by storing the appropriate values in AC(II),
 AZ(II,1), AZ(II,2), ..., AZ(II,n), AA(II,1), AA(II,2).

Section 4.5 Representation of the equation

$$\underline{s_1^{DA}(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{r_1^{(+s)} z_{p-k^*+1}}, s_{p-k^*+2} \sqrt{r_1^{(+s)} z_{p-k^*+2}}, \dots)}$$

$$\underline{s_p \sqrt{r_1^{(+s)} z_p}, z_{p+1}, z_{p+2}, \dots, z_n) = d_1^{+2}}$$

in a computer

For each II, store 1.0 in each of SIZN(II,1), SIZN(II,2), ..., SIZN(II,p-k*), SIZN(II,p+1), SIZN(II,p+2), ..., SIZN(II,n) and ss_e in SIZN(II,e) for e = p-k*+1, p-k*+2, ..., z_p, where

$$ss_e = \begin{cases} 1.0 & \text{if } AZ(II,e) \text{ is even} \\ -1.0 & \text{if } AZ(II,e) \text{ is odd.} \end{cases}$$

Further, store [AZ(II,p-k*+1) + AZ(II,p-k*+2) + ... + AZ(II,p)] in R(II).

Then the terms in

$$\left[s_1^{DA}(z_1, z_2, \dots, z_{p-k^*}, s_{p-k^*+1} \sqrt{r_1^{(+s)} z_{p-k^*+1}}, s_{p-k^*+2} \sqrt{r_1^{(+s)} z_{p-k^*+2}}, \dots, s_p \sqrt{r_1^{(+s)} z_p}, z_{p+1}, z_{p+2}, \dots, z_n) \right]$$

all components of \underline{a}^+
 other than $a_{i_1 j_1 k_1}^+$ and
 $a_{i_2 j_2 k_2}^+$
 are zero

involving $a_{i_1 j_1 k_1}^+$ or $a_{i_2 j_2 k_2}^+$ are represented by

$AC(II), R(II), SIZN(II,1), AZ(II,1), SIZN(II,2), AZ(II,2), \dots, SIZN(II,n),$
 $AZ(II,n), AA(II,1), AA(II,2)$

where $II = 1, 2, \dots, IIMAX$ and $IIMAX$ is the total number of such terms.

Section 4.6 Representation of $\left[\frac{\partial r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+} \right]_{a^+=0}, \left[\frac{\partial r_1^{(+s)}}{\partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$

and $\left[\frac{\partial^2 r_1^{(+s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$ in a computer

By using (3.4.9) and (3.4.10), we can obtain this representation straightforwardly from the representation of the equation in section 4.5.

Section 4.7 Computation of $\beta_{1a_{i_1 j_1 k_1}^+}$ and $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$ in a computer

By using (3.4.11)-(3.4.23), we can find the values of $\beta_{1a_{i_1 j_1 k_1}^+}$ and $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$ from the representation of the derivatives in section 4.6.

Section 4.8 Partition of the set of all possible a_{ijk}^+ into subsets

such that in each subset, different a_{ijk}^+ have similar

expressions of $\beta_{1a_{ijk}^+}$

It is not necessary to derive all the $\beta_{1a_{ijk}^+}$ because different a_{ijk}^+ may have similar expressions of $\beta_{1a_{ijk}^+}$. For example for the case when $k^* \geq 2$, $p-k^* \geq 2$, i_1, i_2 belong to $ST2 = \{p-k^*+1, p-k^*+2, \dots, p\}$, j_1, j_2 belong to $ST1 = \{1, 2, \dots, p-k^*\}$ (note that $ST3 = \{p+1, p+2, \dots, n\}$), $i_1 \neq i_2$ and $j_1 \neq j_2$, we have

$$(4.8.1) \quad \beta_{1a_{i_1j_1j_1}^+} = -z_{Ai_1}(1 + z_{Aj_1}^2)x_{k^*+2, \lambda}^{(d_1+2)}, \text{ where } \lambda = \sum_{j=p-k^*+1}^p z_{Aj_j}^2,$$

and

$$(4.8.2) \quad \beta_{1a_{i_2j_2j_2}^+} = -z_{Ai_2}(1 + z_{Aj_2}^2)x_{k^*+2, \lambda}^{(d_1+2)},$$

and these expressions can be summarised by

$$(4.8.3) \quad \beta_{1a_{ijk}^+} = f_1(i, j, k), \quad (i = i_1, i_2; j = j_1, j_2; k = j_1, j_2; j = k),$$

$$\text{where } f_1(i, j, k) = -z_{Ai}(1 + z_{Aj}^2)x_{k^*+2, \lambda}^{(d_1+2)}.$$

We refer to $a_{i_1j_1k_1}^+$ and $a_{i_2j_2k_2}^+$ as being in the same subset provided

that i_1, i_2 belong to the same ST_i , ($i = 1, 2, 3$)

j_1, j_2 belong to the same ST_i ,

k_1, k_2 belong to the same ST_i ,

and if the expressions of $\beta_{1a_{i_1j_1k_1}^+}$ and $\beta_{1a_{i_2j_2k_2}^+}$ are represented by

$f_{11}(i_1, j_1, k_1)$ and $f_{12}(i_2, j_2, k_2)$ respectively, then $f_{11}(i, j, k) = f_{12}(i, j, k)$.

It is straightforward to see from the expressions of $s_i^{DA}(z)$ that the sufficient conditions for $a_{i_1 j_1 k_1}^+$ and $a_{i_2 j_2 k_2}^+$ to be in the same subset are that

$$JP(1, i) = JP(2, i), \quad (i = 1, 2, \dots, 6),$$

where

$$JP(m, 1) = \begin{cases} 1 & \text{if } i_m \in ST1 \\ 2 & \text{if } i_m \in ST2, \\ 3 & \text{if } i_m \in ST3 \end{cases} \quad (m = 1, 2),$$

$$JP(m, 2) = \begin{cases} 1 & \text{if } j_m \in ST1 \\ 2 & \text{if } j_m \in ST2, \end{cases}$$

$$JP(m, 3) = \begin{cases} 1 & \text{if } k_m \in ST1 \\ 2 & \text{if } k_m \in ST2, \end{cases}$$

$$JP(m, 4) = \begin{cases} 1 & \text{if } i_m = j_m \\ 0 & \text{if } i_m \neq j_m, \end{cases}$$

$$JP(m, 5) = \begin{cases} 1 & \text{if } i_m = k_m \\ 0 & \text{if } i_m \neq k_m, \end{cases}$$

and

$$JP(m, 6) = \begin{cases} 1 & \text{if } j_m = k_m \\ 0 & \text{if } j_m \neq k_m. \end{cases}$$

Section 4.9 Partition of the set of all possible $(a_{i_1 j_1 k_1}^+, a_{i_2 j_2 k_2}^+)$

into subsets such that in each subset, different

$(a_{i_1 j_1 k_1}^+, a_{i_2 j_2 k_2}^+)$ have similar expressions of $\beta_{1a_{i_1 j_1 k_1}^+ a_{i_2 j_2 k_2}^+}$

As for $\beta_{1a_{ijk}^+}$, we refer to $(a_{i_{11} j_{11} k_{11}}^+, a_{i_{12} j_{12} k_{12}}^+)$ and $(a_{i_{21} j_{21} k_{21}}^+, a_{i_{22} j_{22} k_{22}}^+)$ as being in the same subset provided that if the expressions of $\beta_{1a_{i_{11} j_{11} k_{11}}^+ a_{i_{12} j_{12} k_{12}}^+}$ and $\beta_{1a_{i_{21} j_{21} k_{21}}^+ a_{i_{22} j_{22} k_{22}}^+}$ are represented by $f_{21}(i_{11}, j_{11}, k_{11}, i_{12}, j_{12}, k_{12})$ and $f_{22}(i_{21}, j_{21}, k_{21}, i_{22}, j_{22}, k_{22})$ respectively, then $f_{21}(i_1, j_1, k_1, i_2, j_2, k_2) = f_{22}(i_1, j_1, k_1, i_2, j_2, k_2)$. It is seen from the expressions of $S_i^{DA}(z)$ that sufficient conditions for $(a_{i_{11} j_{11} k_{11}}^+, a_{i_{12} j_{12} k_{12}}^+)$ and $(a_{i_{21} j_{21} k_{21}}^+, a_{i_{22} j_{22} k_{22}}^+)$ to be in the same subset are that

$$KP(1,i) = KP(2,i), \quad (i = 1, 2, \dots, 21),$$

where

$$KP(m,1) = \begin{cases} 1 & \text{if } i_m \in ST1 \\ 2 & \text{if } i_m \in ST2 \\ 3 & \text{if } i_m \in ST3 \end{cases}, \quad (m = 1, 2),$$

$$KP(m,2) = \begin{cases} 1 & \text{if } j_m \in ST1 \\ 2 & \text{if } j_m \in ST2 \end{cases},$$

$$KP(m,3) = \begin{cases} 1 & \text{if } k_m \in ST1 \\ 2 & \text{if } k_m \in ST2 \end{cases},$$

$$KP(m,4) = \begin{cases} 1 & \text{if } i_{m2} \in ST1 \\ 2 & \text{if } i_{m2} \in ST2 \\ 3 & \text{if } i_{m2} \in ST3 \end{cases},$$

$$KP(m,5) = \begin{cases} 1 & \text{if } j_{m2} \in ST1 \\ 2 & \text{if } j_{m2} \in ST2 \end{cases}$$

$$KP(m,6) = \begin{cases} 1 & \text{if } k_{m2} \in ST1 \\ 2 & \text{if } k_{m2} \in ST2 \end{cases},$$

$$KP(m,7) = \begin{cases} 1 & \text{if } i_{ml} = j_{ml} \\ 0 & \text{if } i_{ml} \neq j_{ml} \end{cases},$$

$$KP(m,8) = \begin{cases} 1 & \text{if } i_{ml} = k_{ml} \\ 0 & \text{if } i_{ml} \neq k_{ml} \end{cases},$$

$$KP(m,9) = \begin{cases} 1 & \text{if } i_{ml} = i_{m2} \\ 0 & \text{if } i_{ml} \neq i_{m2} \end{cases},$$

$$KP(m,10) = \begin{cases} 1 & \text{if } i_{ml} = j_{m2} \\ 0 & \text{if } i_{ml} \neq j_{m2} \end{cases},$$

$$KP(m,11) = \begin{cases} 1 & \text{if } i_{ml} = k_{m2} \\ 0 & \text{if } i_{ml} \neq k_{m2} \end{cases},$$

$$KP(m,12) = \begin{cases} 1 & \text{if } j_{ml} = k_{ml} \\ 0 & \text{if } j_{ml} \neq k_{ml} \end{cases},$$

$$KP(m,13) = \begin{cases} 1 & \text{if } j_{m1} = i_{m2} \\ 0 & \text{if } j_{m1} \neq i_{m2} \end{cases},$$

$$KP(m,14) = \begin{cases} 1 & \text{if } j_{m1} = j_{m2} \\ 0 & \text{if } j_{m1} \neq j_{m2} \end{cases},$$

$$KP(m,15) = \begin{cases} 1 & \text{if } j_{m1} = k_{m2} \\ 0 & \text{if } j_{m1} \neq k_{m2} \end{cases},$$

$$KP(m,16) = \begin{cases} 1 & \text{if } k_{m1} = i_{m2} \\ 0 & \text{if } k_{m1} \neq i_{m2} \end{cases},$$

$$KP(m,17) = \begin{cases} 1 & \text{if } k_{m1} = j_{m2} \\ 0 & \text{if } k_{m1} \neq j_{m2} \end{cases},$$

$$KP(m,18) = \begin{cases} 1 & \text{if } k_{m1} = k_{m2} \\ 0 & \text{if } k_{m1} \neq k_{m2} \end{cases},$$

$$KP(m,19) = \begin{cases} 1 & \text{if } i_{m2} = j_{m2} \\ 0 & \text{if } i_{m2} \neq j_{m2} \end{cases}$$

$$KP(m,20) = \begin{cases} 1 & \text{if } i_{m2} = k_{m2} \\ 0 & \text{if } i_{m2} \neq k_{m2} \end{cases},$$

and

$$KP(m,21) = \begin{cases} 1 & \text{if } j_{m2} = k_{m2} \\ 0 & \text{if } j_{m2} \neq k_{m2} \end{cases}.$$

Section 4.10 Programs for deriving $\left[\frac{\partial r_1^{(s)}}{\partial a_{i_1 j_1 k_1}^+} \times \frac{\partial r_1^{(s)}}{\partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$

and $\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$.

The programs for deriving these expressions are programs PARTIT and POWPRO, subroutines POWS0, POWSUA, POWSUB and POWSUC as shown in Appendix 3. Note that these programs can also be used to derive the expressions for the case when $k^* = p$. Further, if POWSUC is replaced by SIGSUC (c.f. Appendix 3), the resulting set of programs can also be used to derive the approximations of the probabilities $I_1(\theta_0, \sigma_0)$ and $I_2(\theta_0, \sigma_0)$. These probabilities are equal to $\beta_1(\theta_0, \sigma_0)$ and $\beta_2(\theta_0, \sigma_0)$ respectively.

Section 4.11 Expressions of $\left[\frac{\partial r_1^{(s)}}{\partial a_{i_1 j_1 k_1}^+} \times \frac{\partial r_1^{(s)}}{\partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$

and $\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{i_1 j_1 k_1}^+ \partial a_{i_2 j_2 k_2}^+} \right]_{a^+=0}$

Each of these expressions has one or more terms. Each term is of the form

constant $[d^+]^{l_0} z_1^{l_1} z_2^{l_2} \dots z_{p-k^*}^{l_{p-k^*}} [s_{p-k^*+1} \sqrt{z_{p-k^*+1}^{(s)}}]^{l_{p-k^*+1}}$
 $[s_{p-k^*+2} \sqrt{z_{p-k^*+2}^{(s)}}]^{l_{p-k^*+2}} \dots [s_p \sqrt{z_p^{(s)}}]^{l_p} z_{p+1}^{l_{p+1}} z_{p+2}^{l_{p+2}} \dots z_n^{l_n}$.

These terms obtained by using the programs in section 4.10 are presented in Table (4.11.1)- Table (4.11.20). The numbers in columns [11] to [31] of a table are the codes $XP(\cdot, \cdot)$ for the subsets containing elements whose corresponding expressions are nonzero. The subscripts $i_1, j_1, k_1, i_2, j_2, k_2$ of a typical element $(a_{i_1 j_1 k_1}^+, a_{i_2 j_2 k_2}^+)$ or a subset are in columns [5] to [10]. Each number in column [2] is the total number of terms of an expression. The numbers in columns [3] and [4] are respectively ℓ_0 and constant. Each number in column $[i_1^+]$ (where $i_1^+ = 32, 34, 36$ and 38) is such that if it is j^+ , then the number in the next column (i.e. column $[i_1^+ + 1]$) of the same row is ℓ_{k^+} of $z_{k^+}^{j^+}$ where

$$(4.11.1) \quad k^+ = \begin{cases} i_1 & \text{if } j^+ = 1 \\ j_1 & \text{if } j^+ = 2 \\ k_1 & \text{if } j^+ = 3 \\ i_2 & \text{if } j^+ = 4 \\ j_2 & \text{if } j^+ = 5 \\ k_2 & \text{if } j^+ = 6 \end{cases},$$

and $i_1, j_1, k_1, i_2, j_2, k_2$ all belong to ST1. All values of $\ell_1, \ell_2, \dots, \ell_{p-k^*}$ other than those in columns [33], [35], [37] and [39] are zeros. These values are not presented in the tables. Each number in column $[i_2^+]$ (where $i_2^+ = 40, 42, 44$ and 46) is such that if it is j^+ , then the number in the next column of the same row is ℓ_{k^+} of $[s_{k^+} \sqrt{z_{k^+}}]^{j^+}$ where k^+ and j^+ are related as shown in (4.11.1), and $i_1, j_1, k_1, i_2, j_2, k_2$ in (4.11.1) all belong to ST2. All values of $\ell_{p-k^*+1}, \ell_{p-k^*+2}, \dots, \ell_p$ other than those in columns [41], [43], [45] and [47] are zeros. These values are not presented in the tables. Each number in column $[i_3^+]$ (where $i_3^+ = 48, 50, 52$ and 54) is such that if it is j^+ , then the number in the next column

of the same row is λ_{k^+} of $z_{k^+}^{k^+}$ where k^+ and j^+ are related as shown in (4.11.1) and $i_1, j_1, k_1, i_2, j_2, k_2$ in (4.11.1) all belong to ST3. All values of $\lambda_{p+1}, \lambda_{p+2}, \dots, \lambda_n$ other than those in columns [49], [51], [53] and [55] are zeros. These values are not presented in the tables. For example from Table (4.11.1), we get

$$\left[\frac{\partial r_2^{(+s)}}{\partial a_{211}^+} \right]_{a^+=0} = 4d_2^*{}^4 \left[s_1 \sqrt{z_1^{(s)}} \right]^4 z_2^2$$

and

$$\left[\frac{\partial r_2^{(+s)}}{\partial a_{211}^+} \frac{\partial r_2^{(+s)}}{\partial a_{311}^+} \right]_{a^+=0} = 4d_2^*{}^4 \left[s_1 \sqrt{z_1^{(s)}} \right]^4 z_2 z_3 .$$

While from Table (4.11.2), we get

$$\left[\frac{\partial^2 r_2^{(+s)}}{\partial a_{211}^+ \partial a_{311}^+} \right]_{a^+=0} = 2d_2^*{}^4 \left[s_1 \sqrt{z_1^{(s)}} \right]^4 - 8d_2^*{}^2 \left[s_1 \sqrt{z_1^{(s)}} \right]^2 z_2^2 + 8d_2^*{}^2 \left[s_1 \sqrt{z_1^{(s)}} \right]^4 z_2^2$$

and

$$\left[\frac{\partial^2 r_2^{(+s)}}{\partial a_{211}^+ \partial a_{311}^+} \right]_{a^+=0} = -8d_2^*{}^2 \left[s_1 \sqrt{z_1^{(s)}} \right]^2 z_2 z_3 + 8d_2^*{}^2 \left[s_1 \sqrt{z_1^{(s)}} \right]^4 z_2 z_3 .$$

Table (4.11.1) Representation of the terms in the expressions of the product of the first partial derivatives of $r_2^{(ts)}$ for the case when $p=1$ and $k^*=1$

1	1	4.0	4.0	211211	322322001001011011001	00000000	24000000	12000000
2	1	4.0	4.0	211311	322322000001011011001	00000000	24000000	11410000

Table (4.11.2) Representation of the terms in the expressions of the second partial derivatives of $r_2^{(ts)}$ for the case when $p=1$ and $k^*=1$

1	3	4.0	2.0	211211	322322001001011011001	00000000	24000000	00000000
2	3	2.0	-8.0	211211	322322001001011011001	00000000	22000000	12000000
3	3	2.0	8.0	211211	322322001001011011001	00000000	24000000	12000000
4	2	2.0	-8.0	211311	322322000001011011001	00000000	22000000	11410000
5	2	2.0	8.0	211311	322322000001011011001	00000000	24000000	11410000

Table (4.11.3) Representation of the terms in the expressions of the product of the first partial derivatives of $r_1^{(ts)}$ for the case when $p=2$ and $k^*=1$

1	1	2.0	4.0	211211	211211001001011011001	24000000	12000000	00000000
2	1	2.0	-8.0	211312	211312000010110010000	23000000	12000000	41000000
3	1	3.0	-4.0	211322	211322000110000000001	22000000	13000000	41000000
4	1	2.0	16.0	312312	31231200100000100010000	22000000	32000000	12000000
5	1	3.0	6.0	312322	3123220010000000001	21000000	33000000	12000000
6	1	3.0	8.0	322312	3223120010000010000000	51000000	23000000	12000000
7	1	4.0	4.0	322322	32232200010010110001	00000000	24000000	12000000
8	1	2.0	16.0	312412	3123120000010001000000	22000000	32000000	11410000
9	1	3.0	8.0	312422	3123220000000000001	21000000	33000000	11410000
10	1	3.0	8.0	322412	3223120000010010000000	51000000	23000000	11410000
11	1	4.0	4.0	322422	322322000001011011001	00000000	24000000	11410000

Table (4.11.4) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(ts)}$ for the case when $p=2$ and $k^*=1$

1	3	0	-2.0	211211	211211001001011011001	24000000	00000000	00000000
2	3	2.0	8.0	211211	211211001001011011001	22000000	12000000	00000000
3	3	0	4.0	211211	211211001001011011001	24000000	12000000	00000000
4	1	1.0	8.0	211311	211312000010110010000	22000000	11000000	41000000
5	1	0	-8.0	211312	211312000010110010000	23000000	12000000	41000000
6	1	1.0	-6.0	211322	211322000011000000001	22000000	13000000	41000000
7	2	1.0	4.0	311312	31131200010010110010000	27000000	61000000	00000000
8	2	1.0	-8.0	311312	31131200010010110010000	21000000	61000000	12000000
9	1	2.0	2.0	311322	311322000100100000001	22000000	52000000	00000000
10	2	1.0	4.0	312311	3123110001000000001	23000000	31000000	00000000
11	2	1.0	-8.0	312311	3123110001000000001	21000000	31000000	12000000
12	4	2.0	8.0	312312	31231200000100010000	22000000	32000000	00000000
13	4	0	-8.0	312312	31231200000100010000	22000000	00000000	12000000
14	4	2.0	-8.0	312312	31231200001100010000	00000000	32000000	12000000
15	4	0	16.0	312312	3123120001000100010000	22000000	32000000	12000000
16	3	3.0	4.0	312322	3123220001000000001	21000000	33000000	00000000
17	3	1.0	-8.0	312322	3123220001000000001	21000000	31000000	12000000
18	3	1.0	12.0	312322	3123220001000000001	21000000	33000000	12000000
19	1	2.0	2.0	322311	3223110001000000001	52000000	22000000	00000000
20	3	3.0	"	322312	3223120001000000001	51000000	23000000	00000000
21	3	1.0	-8.0	322312	3223120001000000001	51000000	21000000	12000000
22	3	1.0	12.0	322312	3223120001000000001	51000000	23000000	12000000
23	3	4.0	2.0	322322	3223220001000000001	00000000	24000000	00000000
24	3	2.0	-8.0	322322	3223220001000000001	00000000	22000000	12000000
25	3	2.0	8.0	322322	3223220001000000001	00000000	24000000	12000000
26	1	1.0	-8.0	311412	31131200000100010000	21000000	61000000	11410000
27	1	1.0	-8.0	312411	31231100000100010000	21000000	31000000	11410000
28	3	0	-8.0	312412	31231200000100010000	22000000	00000000	11410000
29	3	2.0	-8.0	312412	31231200000100010000	00000000	32000000	11410000
30	3	0	16.0	312412	31231200000100010000	22000000	32000000	11410000
31	2	1.0	-8.0	312422	3123220000000000001	21000000	31000000	11410000
32	2	1.0	12.0	312422	3123220000000000001	21000000	33000000	11410000
33	2	1.0	-8.0	322412	3223120000000000001	21000000	21000000	11410000
34	2	1.0	12.0	322412	3223120000000000001	21000000	23000000	11410000
35	2	2.0	-8.0	322422	3223220000000000001	00000000	22000000	11410000
36	2	2.0	8.0	322422	3223220000000000001	00000000	24000000	11410000

Table (4.11.5) Representation of the terms in the expressions of the product of the first partial derivatives of $r_2^{(s)}$ for the case when $p=2$ and $k^*=2$

1	1	4.0	4.0	311311	322322001001011011001	00000000	24000000	12000000
2	1	4.0	8.0	311312	3223220010010110010000	00000000	23610000	12000000
3	1	4.0	4.0	311322	322322001001001000000001	00000000	22520000	12000000
4	1	4.0	8.0	312311	32232200100001000010001	00000000	23310000	12000000
5	1	4.0	16.0	312312	32232200100001000010001	00000000	22320000	12000000
6	1	4.0	8.0	312322	322322001000000011001	00000000	21340000	12000000
7	1	4.0	8.0	322312	3223220010010010010000	00000000	51230000	12000000
8	1	4.0	4.0	311411	322322000001011011001	00000000	24070000	11410000
9	1	4.0	8.0	311412	322322000001010010001000	00000000	23610000	11410000
10	1	4.0	4.0	311422	32232200000100000001	00000000	22520000	11410000
11	1	4.0	8.0	312411	322322000000011000001	00000000	23310000	11410000
12	1	4.0	16.0	312412	322322000000010001000	00000000	22320000	11410000
13	1	4.0	8.0	312422	322322000000000011001	00000000	21330000	11410000
14	1	4.0	8.0	322412	322322000001001001000	00000000	51230000	11410000

Table (4.11.6) Representation of the terms in the expressions of the second partial derivatives of $r_2^{(s)}$ for the case when $p=2$ and $k^*=2$

1	3	4.0	2.0	311311	322322001001011011001	00000000	24070000	00000000
2	3	2.0	-8.0	311311	322322001001011011001	00000000	22000000	12000000
3	3	2.0	8.0	311311	322322001001011011001	00000000	24070000	12000000
4	3	4.0	4.0	311312	322322001001010010000	00000000	23610000	00000000
5	3	2.0	-8.0	311312	322322001001010010000	00000000	21310000	12000000
6	3	2.0	16.0	311312	322322001001010010000	00000000	23610000	12000000
7	2	4.0	2.0	311322	32232200100100000001	00000000	22520000	00000000
8	2	2.0	8.0	311322	32232200100100000001	00000000	22520000	12000000
9	3	4.0	4.0	312311	322322001000011000001	00000000	23310000	00000000
10	3	2.0	-8.0	312311	322322001000011000001	00000000	21310000	12000000
11	3	2.0	16.0	312311	322322001000011000001	00000000	23310000	12000000
12	4	4.0	8.0	312312	3223220010000010001000	00000000	22320000	00000000
13	4	2.0	-8.0	312312	3223220010000010001000	00000000	22320000	12000000
14	4	2.0	-8.0	312312	3223220010000010001000	00000000	32000000	12000000
15	4	2.0	32.0	312312	3223220010000010001000	00000000	22320000	12000000
16	3	4.0	4.0	312322	322322001000000011001	00000000	21330000	00000000
17	3	2.0	-8.0	312322	322322001000000011001	00000000	21310000	12000000
18	3	2.0	16.0	312322	322322001000000011001	00000000	21330000	12000000
19	3	4.0	4.0	322312	3223220010000010001000	00000000	51230000	00000000
20	3	2.0	-8.0	322312	3223220010000010001000	00000000	51210000	12000000
21	3	2.0	16.0	322312	3223220010000010001000	00000000	51230000	12000000
22	2	2.0	-8.0	311411	322322000001011011001	00000000	22000000	11410000
23	2	2.0	8.0	311411	322322000001011011001	00000000	24070000	11410000
24	2	2.0	-8.0	311412	322322000001010010000	00000000	21310000	11410000
25	2	2.0	16.0	311412	322322000001010010000	00000000	23610000	11410000
26	1	2.0	8.0	311422	32232200000100000001	00000000	22520000	11410000
27	2	2.0	-8.0	312411	322322000000001100001	00000000	21310000	11410000
28	2	2.0	16.0	312411	322322000000001100001	00000000	23310000	11410000
29	3	2.0	-8.0	312412	322322000000001100001	00000000	22000000	11410000
30	3	2.0	-8.0	312412	322322000000001100001	00000000	32000000	11410000
31	3	2.0	32.0	312412	322322000000001100001	00000000	22320000	11410000
32	2	2.0	-8.0	312422	322322000000001100001	00000000	21310000	11410000
33	2	2.0	16.0	312422	322322000000001100001	00000000	21330000	11410000
34	2	2.0	-8.0	322412	322322000000001100001	00000000	51210000	11410000
35	2	2.0	16.0	322412	322322000000001100001	00000000	51230000	11410000

Table (4.11.7) Representation of the terms in the expressions of the product of the first partial derivatives of $r_1^{(+s)}$ for the case when $p=3$ and $k*=1$

1	1	2.0	4.0	311311	211211001001011011001	240000000	120000000	000000000
2	1	2.0	8.0	311312	211211001001011011000	236100000	120000000	000000000
3	1	2.0	4.0	311322	211211001001011011000	225200000	120000000	000000000
4	1	2.0	4.0	312311	211211001001011011000	225200000	120000000	000000000
5	1	2.0	16.0	312312	211211001001011011000	225200000	120000000	000000000
6	1	2.0	8.0	312322	211211001001011011000	213300000	120000000	000000000
7	1	2.0	8.0	322312	211211001001011011000	512300000	120000000	000000000
8	1	2.0	-8.0	311413	211312000011010010000	230000000	120000000	410000000
9	1	2.0	-8.0	311423	211312000011010010000	225100000	120000000	410000000
10	1	3.0	-4.0	311433	211322001011000000001	220000000	130000000	410000000
11	1	2.0	-16.0	312413	211312000010010000000	223100000	120000000	410000000
12	1	2.0	-16.0	312423	211312000010000010000	213200000	120000000	410000000
13	1	3.0	-8.0	312433	211322000110000000001	213100000	130000000	410000000
14	1	2.0	16.0	413413	312312001001000000000	220000000	320000000	120000000
15	1	2.0	16.0	413423	312312001000000000000	215100000	320000000	120000000
16	1	3.0	8.0	413433	312322001000000011001	210000000	330000000	120000000
17	1	3.0	8.0	433413	322312001001001000000	510000000	230000000	120000000
18	1	4.0	4.0	433433	322322001001001100000	000000000	240000000	120000000
19	1	2.0	16.0	413513	312312000000001000000	220000000	320000000	114100000
20	1	2.0	16.0	413523	312312000000000000000	215100000	320000000	114100000
21	1	3.0	8.0	413533	312322000000000011001	210000000	330000000	114100000
22	1	3.0	8.0	433513	3223120000001001001000	510000000	230000000	114100000
23	1	4.0	4.0	433533	3223220000001011011001	000000000	240000000	114100000

Table (4.11.8) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(+s)}$ for the case when $p=3$ and $k*=1$

1	3	0	-2.0	311311	211211001001011011001	240000000	000000000	000000000
2	3	2.0	8.0	311311	211211001001011011001	220000000	120000000	000000000
3	3	0	4.0	311311	211211001001011011001	240000000	120000000	000000000
4	3	0	-4.0	311312	211211001001011011000	236100000	000000000	000000000
5	3	2.0	8.0	311312	211211001001011011000	216100000	120000000	000000000
6	3	0	8.0	311312	211211001001011011000	236100000	120000000	000000000
7	2	0	-2.0	311322	211211001001000000001	225200000	000000000	000000000
8	2	0	4.0	311322	211211001001000000001	225200000	120000000	000000000
9	3	0	-4.0	312311	211211001000001100001	223100000	000000000	000000000
10	3	2.0	8.0	312311	211211001000001100001	213100000	120000000	000000000
11	3	0	8.0	312311	211211001000001100001	233100000	120000000	000000000
12	4	0	-8.0	312312	2112110010000010001000	223200000	000000000	000000000
13	4	2.0	8.0	312312	2112110010000010001000	220000000	120000000	000000000
14	4	2.0	8.0	312312	2112110010000010001000	320000000	120000000	000000000
15	4	0	16.0	312312	2112110010000010001000	223200000	120000000	000000000
16	3	0	-4.0	312322	211211001000000011001	213300000	000000000	000000000
17	3	2.0	8.0	312322	211211001000000011001	213100000	120000000	000000000
18	3	0	8.0	312322	211211001000000011001	213300000	120000000	000000000
19	3	0	-4.0	322312	211211001000001001000	512300000	000000000	000000000
20	3	2.0	8.0	322312	211211001000001001000	512100000	120000000	000000000
21	3	0	8.0	322312	211211001000001001000	512300000	120000000	000000000
22	1	1.0	8.0	311411	211311000010110110001	220000000	110000000	410000000
23	1	1.0	8.0	311412	211311000010110110000	213100000	110000000	410000000
24	1	0	-8.0	311413	211312000011010110000	230000000	120000000	410000000
25	1	0	-8.0	311423	211312000011010110000	225100000	120000000	410000000
26	1	1.0	-6.0	311433	211322000111000000001	220000000	130000000	410000000
27	1	1.0	0.0	312411	211311000000110000001	213100000	110000000	410000000
28	2	1.0	0.0	312412	211311000000110000000	220000000	110000000	410000000
29	2	1.0	8.0	312412	211311000000110000000	320000000	110000000	410000000
30	1	0	-16.0	312413	211312000010010000000	223100000	120000000	410000000

Table (4.11.8) contd.

31	1	1.0	8.0	312422	2113110000000000011001	21310000	11000000	41000000
32	1	0	-16.0	312423	2113120000100000100000	21320000	12000000	41000000
33	1	1.0	-12.0	312433	21132?0001100000000001	21310000	13000000	41000000
34	1	1.0	8.0	322412	2113110000100000100001	51210000	11000000	41000000
35	2	1.0	4.0	411413	3113120010010100100000	23000000	61000000	00000000
36	2	1.0	-8.0	411413	3113120010010100100000	21000000	61000000	12000000
37	1	1.0	4.0	411423	3113120010010000000000	22510000	61000000	00000000
38	1	2.0	2.0	411433	3113220010010000000001	22000000	52000000	00000000
39	2	1.0	8.0	412413	3113120010000010000000	22310000	61000000	00000000
40	2	1.0	-8.0	412413	3113120010000010000000	31000000	61000000	12000000
41	2	1.0	8.0	412423	3113120010000000010000	21320000	61000000	00000000
42	2	1.0	-8.0	412423	3113120010000000010000	21000000	61000000	12000000
43	1	2.0	4.0	412433	3113220010000000000001	21310000	52000000	00000000
44	2	1.0	4.0	413411	3123110010000110000001	23000000	31000000	00000000
45	2	1.0	-8.0	413411	3123110010000110000001	21000000	31000000	12000000
46	2	1.0	8.0	413412	3123110010000100000000	22510000	31000000	00000000
47	2	1.0	-8.0	413412	3123110010000100000000	61000000	31000000	12000000
48	4	2.0	8.0	413413	3123120010000100000000	22000000	32000000	00000000
49	4	0	-8.0	413413	3123120010000100000000	22000000	00000000	12000000
50	4	2.0	-8.0	413413	3123120010000100000000	00000000	32000000	12000000
51	4	0	16.0	413413	3123120010000100000000	22000000	32000000	12000000
52	1	1.0	4.0	413422	3123110010000000000001	21520000	31000000	00000000
53	3	2.0	8.0	413423	3123120010000000000001	21510000	32000000	00000000
54	3	0	-8.0	413423	3123120010000000000001	21510000	00000000	12000000
55	3	0	16.0	413423	3123120010000000000001	21510000	32000000	12000000
56	3	3.0	4.0	413433	3123220010000000000001	21000000	33000000	00000000
57	3	1.0	-8.0	413433	3123220010000000000001	21000000	31000000	12000000
58	3	1.0	12.0	413433	3123220010000000000001	21000000	33000000	12000000
59	2	1.0	8.0	423412	3123110010000100000000	51220000	31000000	00000000
60	2	1.0	-8.0	423412	3123110010000100000000	51000000	31000000	12000000
61	1	2.0	2.0	433411	3223110010000000000001	52000000	22000000	00000000
62	1	2.0	4.0	433412	3223110010000000000001	51610000	22000000	00000000
63	3	3.0	4.0	433413	322312001001001001000	51000000	23000000	00000000
64	3	1.0	-8.0	433413	322312001001001001000	51000000	21000000	12000000
65	3	1.0	12.0	433413	322312001001001001000	51000000	23000000	12000000
66	3	4.0	2.0	433433	3223220010010010110001	00000000	24000000	00000000
67	3	2.0	-8.0	433433	3223220010010010110001	00000000	22000000	12000000
68	3	2.0	8.0	433433	3223220010010010110001	00000000	24000000	12000000
69	1	1.0	-8.0	411513	3113120000001000010000	21000000	61000000	11410000
70	1	1.0	-8.0	412513	3113120000001000000000	31000000	61000000	11410000
71	1	1.0	-8.0	412523	3113120000000000000000	21000000	61000000	11410000
72	1	1.0	-8.0	413511	3123110000000000000000	21000000	31000000	11410000
73	1	1.0	-8.0	413512	3123110000000000000000	51000000	31000000	11410000
74	3	0	-8.0	413513	3123120000000000000000	22000000	00000000	11410000
75	3	2.0	-8.0	413513	3123120000000000000000	00000000	32000000	11410000
76	3	0	16.0	413513	3123120000000000000000	22000000	32000000	11410000
77	2	0	-8.0	413523	3123120000000000000000	21510000	00000000	11410000
78	2	0	16.0	413523	3123120000000000000000	21510000	32000000	11410000
79	2	1.0	-8.0	413533	3123220000000000000000	21000000	31000000	11410000
80	2	1.0	12.0	413533	3123220000000000000000	21000000	33000000	11410000
81	1	1.0	-8.0	423512	3123110000000000000000	51000000	31000000	11410000
82	2	1.0	-8.0	433513	3223120000000000000000	51000000	21000000	11410000
83	2	1.0	12.0	433513	3223120000000000000000	51000000	23000000	11410000
84	2	2.0	-8.0	433533	3223220000000000000000	00000000	22000000	11410000
85	2	2.0	8.0	433553	3223220000000000000000	00000000	24000000	11410000

Table (4.11.9) Representation of the terms in the expressions of the product of the first partial derivatives of $r_1^{(+)}$ for the case when $p=3$ and $k^*=2$

1	1	2,0	4,0	211211	211211001001011011001	240000000	124000000	000000000
2	1	2,0	4,0	211311	211211000001011011001	240000000	114100000	000000000
3	1	2,0	-6,0	211412	211312000001101001000	230000000	120000000	410000000
4	1	2,0	-8,0	211413	211312000001010110000	230000000	116100000	410000000
5	1	3,0	-4,0	211422	211322000111000010001	220000000	130000000	410000000
6	1	3,0	-8,0	211423	211322000101000000000	220000000	126100000	410000000
7	1	3,0	-4,0	211433	211322000001000000001	220000000	115200000	410000000
8	1	3,0	-8,0	311423	211322000011000000000	220000000	511200000	410000000
9	1	2,0	16,0	412412	312312001000010001000	220000000	320000000	120000000
10	1	2,0	16,0	412413	312312001000010000000	220000000	316100000	120000000
11	1	3,0	8,0	412422	31232200100000000011001	210000000	330000000	120000000
12	1	3,0	16,0	412423	3123220010000000001000	210000000	326100000	120000000
13	1	3,0	8,0	412432	3123220010000000000001	210000000	315200000	120000000
14	1	3,0	16,0	413423	3123220010000000001000	210000000	513200000	120000000
15	1	3,0	8,0	422412	322312001001001001000	510000000	230000000	120000000
16	1	3,0	8,0	422413	322312001001000000000	510000000	226100000	120000000
17	1	4,0	4,0	422422	3223220010010010011001	000000000	240000000	120000000
18	1	4,0	8,0	422423	322322001001001000000	000000000	236100000	120000000
19	1	4,0	4,0	422433	322322001001000000001	000000000	225200000	120000000
20	1	3,0	16,0	423412	322312001000000000000	510000000	223100000	120000000
21	1	3,0	16,0	423413	3223120010000000001000	510000000	213200000	120000000
22	1	4,0	8,0	423422	32232200100000000011001	000000000	233100000	120000000
23	1	4,0	16,0	423423	3223220010000000001000	000000000	223200000	120000000
24	1	4,0	8,0	423433	32232200100000000011001	000000000	213300000	120000000
25	1	4,0	8,0	433423	322322001001001001000	000000000	512300000	120000000
26	1	2,0	16,0	412512	312312000000001000000	220000000	320000000	114100000
27	1	2,0	16,0	412513	312312000000000000000	220000000	316100000	114100000
28	1	3,0	8,0	412522	3123220000000000000001	210000000	330000000	114100000
29	1	3,0	16,0	412523	3123220000000000000000	210000000	326100000	114100000
30	1	3,0	8,0	412533	3123220000000000000000	210000000	315200000	114100000
31	1	3,0	16,0	413523	3123220000000000000000	210000000	513200000	114100000
32	1	3,0	8,0	422512	322312000001001001000	510000000	230000000	114100000
33	1	3,0	8,0	422513	322312000001000000000	510000000	226100000	114100000
34	1	4,0	4,0	422522	322322000001010010001	000000000	240000000	114100000
35	1	4,0	8,0	422523	3223220000000000001000	000000000	236100000	114100000
36	1	4,0	4,0	422533	3223220000000000000001	000000000	225200000	114100000
37	1	3,0	16,0	423512	322312000000001000000	510000000	223100000	114100000
38	1	3,0	16,0	423513	3223120000000000000000	510000000	213200000	114100000
39	1	4,0	8,0	423522	3223220000000000000000	000000000	233100000	114100000
40	1	4,0	16,0	423523	3223220000000000000000	000000000	223200000	114100000
41	1	4,0	8,0	423533	3223220000000000000000	000000000	213300000	114100000
42	1	4,0	8,0	433523	3223220000001001001000	000000000	512300000	114100000

Table (4.11.10) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(+)}$ for the case when $p=3$ and $k^*=2$

1	3	0	-2,0	211211	211211001001011011001	240000000	000000000	000000000
2	3	2,0	8,0	211211	211211001001011011001	220000000	120000000	000000000
3	3	0	4,0	211211	211211001001011011001	240000000	120000000	000000000
4	2	2,0	8,0	211311	211211000001011011001	220000000	114100000	000000000
5	2	0	4,0	211311	211211000001011011001	240000000	114100000	000000000
6	1	1,0	8,0	211411	211211000001011011001	220000000	110000000	410000000
7	1	0	-8,0	211412	211312000001010010000	230000000	120000000	410000000
8	1	0	-8,0	211413	211312000001010010000	230000000	116100000	410000000
9	1	1,0	-6,0	211422	211322000001010000000	220000000	130000000	410000000
10	1	1,0	-12,0	211423	211322000001010000000	220000000	126100000	410000000
11	1	1,0	-6,0	211433	211322000001010000000	220000000	115200000	410000000
12	1	1,0	-12,0	311423	211322000001010000000	220000000	511200000	410000000
13	2	1,0	4,0	411412	311312000001010000000	230000000	610000000	000000000
14	2	1,0	-3,0	411412	311312000001010000000	210000000	610000000	120000000
15	1	2,0	2,0	411422	311322000010100000000	220000000	520000000	000000000
16	1	2,0	4,0	411423	311322000010100000000	220000000	516100000	000000000
17	2	1,0	4,0	412411	312311000001010000000	230000000	310000000	000000000
18	2	1,0	-8,0	412411	312311000001010000000	210000000	310000000	120000000
19	4	2,0	8,0	412412	312312000001000000000	220000000	320000000	000000000
20	4	0	-8,0	412412	312312000001000000000	220000000	000000000	120000000
21	4	2,0	-8,0	412412	312312000001000000000	000000000	320000000	120000000
22	4	0	16,0	412412	312312000001000000000	220000000	320000000	120400000
23	3	2,0	8,0	412413	312312000001000000000	220000000	316100000	000000000
24	3	2,0	-8,0	412413	312312000001000000000	000000000	316100000	120000000
25	3	0	16,0	412413	312312000001000000000	220000000	316100000	120000000
26	3	3,0	4,0	412422	312322000100000000000	210000000	330000000	000000000
27	3	1,0	-8,0	412422	312322000100000000000	210000000	310000000	120000000
28	3	1,0	12,0	412422	312322000100000000000	210000000	330000000	120000000
29	3	3,0	8,0	412423	312322000100000000000	210000000	326100000	000000000
30	3	1,0	-8,0	412423	312322000100000000000	210000000	610000000	120000000

Table (4.11.10) contd.

Table (4.11.11) Representation of the terms in the expressions of the product of the first partial derivatives of $r_2^{(+)}$ for the case when $p=3$ and $k^*=3$

1	1	4.0	411411	322322001001011011001	00000000	24000000	12000000
2	1	4.0	411412	322322001001010110001	00000000	23610000	12000000
3	1	4.0	411422	322322001001000000001	00000000	22520000	12000000
4	1	4.0	411423	322322001001000000000	00000000	22516100	12000000
5	1	4.0	412411	322322001000011000001	00000000	23310000	12000000
6	1	4.0	412412	322322001000010001000	00000000	22320000	12000000
7	1	4.0	412413	322322001000010000000	00000000	22316100	12000000
8	1	4.0	412422	32232200100000110001	00000000	21330000	12000000
9	1	4.0	412423	322322001000001000000	00000000	21326100	12000000
10	1	4.0	412433	322322001000000000001	00000000	21315200	12000000
11	1	4.0	413423	322322001000000010000	00000000	21513200	12000000
12	1	4.0	422412	322322001001001001000	00000000	51230000	12000000
13	1	4.0	423412	322322001000001000000	00000000	51223100	12000000
14	1	4.0	411511	322322000001011011001	00000000	24000000	11410000
15	1	4.0	411512	3223220000010101001000	00000000	23610000	11410000
16	1	4.0	411522	322322000001000000001	00000000	22520000	11410000
17	1	4.0	411523	322322000000100000000	00000000	22516100	11410000
18	1	4.0	412511	322322000000011000001	00000000	23310000	11410000
19	1	4.0	412512	322322000000010001000	00000000	22320000	11410000
20	1	4.0	412513	322322000000010000000	00000000	22316100	11410000
21	1	4.0	412522	32232200000000110001	00000000	21330000	11410000
22	1	4.0	412523	32232200000000010000	00000000	21326100	11410000
23	1	4.0	412533	322322000000000000001	00000000	21315200	11410000
24	1	4.0	413523	322322000000000010000	00000000	21513200	11410000
25	1	4.0	422512	322322000001001001000	00000000	51230000	11410000
26	1	4.0	423512	322322000000001000000	00000000	51223100	11410000

Table (4.11.12) Representation of the terms in the expressions of the second partial derivatives of $r_2^{(+)}$ for the case when $p=3$ and $k^*=3$

1	3	4.0	2.0	411411	322322001001011011001	00000000	24000000	00000000
2	3	2.0	-8.0	411411	322322001001011011001	00000000	22000000	12000000
3	3	2.0	8.0	411411	322322001001011011001	00000000	24000000	12000000
4	3	4.0	4.0	411412	322322001001010100000	00000000	23610000	00000000
5	3	2.0	-8.0	411412	322322001001010100000	00000000	21610000	12000000
6	3	2.0	16.0	411412	322322001001010100000	00000000	23610000	12000000
7	2	4.0	2.0	411422	322322001001000000001	00000000	22520000	00000000
8	2	2.0	8.0	411422	322322001001000000001	00000000	22520000	12000000
9	2	4.0	4.0	411423	322322001001000000000	00000000	22516100	00000000
10	2	2.0	16.0	411423	322322001001000000000	00000000	22516100	12000000
11	3	4.0	4.0	412411	322322001000101100001	00000000	23310000	00000000
12	3	2.0	-8.0	412411	322322001000101100001	00000000	21310000	12000000
13	3	2.0	16.0	412411	322322001000101100001	00000000	23310000	12000000
14	4	4.0	8.0	412412	322322001000100010000	00000000	22320000	00000000
15	4	2.0	-8.0	412412	322322001000100010000	00000000	22000000	12000000
16	4	2.0	-8.0	412412	322322001000100010000	00000000	32000000	12000000
17	4	2.0	32.0	412412	322322001000100010000	00000000	22320000	12000000
18	3	4.0	8.0	412413	322322001000100010000	00000000	22316100	00000000
19	3	2.0	-8.0	412413	322322001000100010000	00000000	31610000	12000000
20	3	2.0	32.0	412413	322322001000100010000	00000000	22316100	12000000
21	3	4.0	4.0	412422	322322001000100011001	00000000	21330000	00000000
22	3	2.0	-8.0	412422	322322001000100011001	00000000	21310000	12000000
23	3	2.0	16.0	412422	322322001000100011001	00000000	21330000	12000000
24	3	4.0	8.0	412423	322322001000100011000	00000000	21326100	00000000
25	3	2.0	-8.0	412423	322322001000100011000	00000000	21610000	12000000
26	3	2.0	32.0	412423	322322001000100011000	00000000	21326100	12000000
27	2	4.0	4.0	412433	322322001000100011000	00000000	21315200	00000000
28	2	2.0	16.0	412433	322322001000100011000	00000000	21315200	12000000
29	3	4.0	8.0	413423	322322001000100011000	00000000	21513200	00000000
30	3	2.0	-8.0	413423	322322001000100011000	00000000	21510000	12000000

Table (4.11.12) contd.

31	3	2.0	32.0	413423	322322000100000000010000	000000000	215132000	120000000
32	3	4.0	4.0	422412	3223220010010010010000	000000000	512300000	000000000
33	3	2.0	-8.0	422412	3223220010010010010000	000000000	512100000	120000000
34	3	2.0	16.0	422412	3223220010010010010000	000000000	512300000	120000000
35	3	4.0	8.0	423412	3223220010000010000000	000000000	512231000	000000000
36	3	2.0	-8.0	423412	3223220010000010000000	000000000	513100000	120000000
37	3	2.0	32.0	423412	3223220010000010000000	000000000	512231000	120000000
38	2	2.0	-8.0	411511	322322000001011011001	000000000	220000000	114100000
39	2	2.0	8.0	411511	322322000001011011001	000000000	240000000	114100000
40	2	2.0	-8.0	411512	3223220000010100100000	000000000	216100000	114100000
41	2	2.0	16.0	411512	3223220000010100100000	000000000	236100000	114100000
42	1	2.0	8.0	411522	322322000001000000001	000000000	225200000	114100000
43	1	2.0	16.0	411523	322322000001000000000	000000000	225161000	114100000
44	2	2.0	-8.0	412511	322322000001011000001	000000000	213100000	114100000
45	2	2.0	16.0	412511	322322000001011000001	000000000	233100000	114100000
46	3	2.0	-8.0	412512	322322000001000010000	000000000	220000000	114100000
47	3	2.0	-8.0	412512	322322000000100001000	000000000	320000000	114100000
48	3	2.0	32.0	412512	322322000000100001000	000000000	223200000	114100000
49	2	2.0	-8.0	412513	322322000000100000000	000000000	316100000	114100000
50	2	2.0	32.0	412513	322322000000100000000	000000000	223161000	114100000
51	2	2.0	-8.0	412522	3223220000000000011001	000000000	213100000	114100000
52	2	2.0	16.0	412522	3223220000001000011001	000000000	213300000	114100000
53	2	2.0	-8.0	412523	3223220000000000010000	000000000	216100000	114100000
54	2	2.0	32.0	412523	3223220000000000010000	000000000	213261000	114100000
55	1	2.0	16.0	412533	3223220000000000000001	000000000	213152000	114100000
56	2	2.0	-8.0	413523	3223220000000000000001	000000000	215100000	114100000
57	2	2.0	32.0	413523	3223220000000000000001	000000000	215132000	114100000
58	2	2.0	-8.0	422512	32232200000100010000	000000000	512100000	114100000
59	2	2.0	16.0	422512	32232200000100010000	000000000	512300000	114100000
60	2	2.0	-8.0	423512	3223220000000001000000	000000000	513100000	114100000
61	2	2.0	32.0	423512	3223220000000001000000	000000000	512231000	114100000

Table (4.11.13) Representation of the terms in the expressions of the product of the first partial derivatives of $r_1^{(+)}$ for the case when $p=4$ and $k^*=1$

Table (4.11.14) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(+s)}$ for the case when $p=4$ and $k^*=1$

1	3	0	-2.0	411411	211211001091011011001	240000000	000000000	000000000
2	3	2.0	8.0	411411	211211001001011011001	220000000	120000000	000000000
3	3	0	4.0	411411	211211001001011011001	240000000	120000000	000000000
4	3	0	-4.0	411412	211211001071010101000	236100000	000000000	000000000
5	3	2.0	8.0	411412	211211001001010101000	216100000	120000000	000000000
6	3	0	8.0	411412	211211001001010101000	236100000	120000000	000000000
7	2	0	-2.0	411422	211211001001000000001	220000000	000000000	000000000
8	2	0	4.0	411422	211211001001000000001	225100000	120000000	000000000
9	2	0	-4.0	411423	211211001001000000000	225161000	000000000	000000000
10	2	0	8.0	411423	211211001001000000000	225161000	120000000	000000000
11	3	0	-4.0	412411	211211001001000000001	233100000	000000000	000000000
12	3	2.0	8.0	412411	211211001000000000001	213100000	120000000	000000000
13	3	0	8.0	412411	211211001000000000001	233100000	120000000	000000000
14	4	0	-8.0	412412	211211001000000000000	223200000	000000000	000000000
15	4	2.0	8.0	412412	211211001000000000000	220000000	120000000	000000000
16	4	2.0	8.0	412412	211211001000000000000	320000000	120000000	000000000
17	4	0	16.0	412412	211211001000000000000	223200000	120000000	000000000
18	3	0	-8.0	412413	211211001000000000000	223161000	000000000	000000000
19	3	2.0	8.0	412413	211211001000000000000	316100000	120000000	000000000
20	3	0	16.0	412413	211211001000000000000	223161000	120000000	000000000
21	3	0	-4.0	412422	211211001000000000001	213300000	000000000	000000000
22	3	2.0	8.0	412422	211211001000000000001	213100000	120000000	000000000
23	3	0	8.0	412422	211211001000000000001	213300000	120000000	000000000
24	3	0	-8.0	412423	211211001000000000000	213261000	000000000	000000000
25	3	2.0	8.0	412423	211211001000000000000	216100000	120000000	000000000
26	3	0	16.0	412423	211211001000000000000	213261000	120000000	000000000
27	2	0	-4.0	412433	211211001000000000001	213152000	000000000	000000000
28	2	0	8.0	412433	211211001000000000001	213152000	120000000	000000000
29	3	0	-8.0	413423	211211001000000000000	215132000	000000000	000000000
30	3	2.0	8.0	413423	211211001000000000000	215100000	120000000	000000000
31	3	0	16.0	413423	211211001000000000000	215132000	120000000	000000000
32	3	0	-4.0	422412	211211001001001001000	512300000	000000000	000000000
33	3	2.0	8.0	422412	211211001001001001000	512100000	120000000	000000000
34	3	0	8.0	422412	211211001001001001000	512300000	120000000	000000000
35	3	0	-8.0	423412	211211001000000000000	512231000	000000000	000000000
36	3	2.0	8.0	423412	211211001000000000000	513100000	120000000	000000000
37	3	0	16.0	423412	211211001000000000000	512231000	120000000	000000000
38	1	1.0	8.0	411511	211311000001011011001	220000000	110000000	410000000
39	1	1.0	8.0	411512	211311000001010010000	216100000	110000000	410000000
40	1	0	-8.0	411514	2113120000011010010000	230000000	120000000	410000000
41	1	0	-8.0	411524	211312000001100000000	225100000	120000000	410000000
42	1	1.0	-6.0	411544	211322000011000000001	220000000	130000000	410000000
43	1	1.0	8.0	412511	211311000000110000001	213100000	110000000	410000000
44	2	1.0	8.0	412512	211311000000010000000	220000000	110000000	410000000
45	2	1.0	8.0	412512	211311000000010000000	320000000	110000000	410000000
46	1	1.0	8.0	412513	211311000000010000000	316100000	110000000	410000000
47	1	0	-16.0	412514	211312000010000000000	223100000	120000000	410000000
48	1	1.0	8.0	412522	211311000000000000001	213100000	110000000	410000000
49	1	1.0	8.0	412523	211311000000000000000	216100000	110000000	410000000
50	1	0	-16.0	412524	211312000000000000000	213200000	120000000	410000000
51	1	0	-16.0	412534	211312000010000000000	213151000	120000000	410000000
52	1	1.0	-12.0	412544	211322000010000000001	213100000	130000000	410000000
53	1	1.0	8.0	413523	211311000000000000000	215100000	110000000	410000000
54	1	1.0	8.0	422512	211311000000000000001	512100000	110000000	410000000
55	1	1.0	8.0	423512	211311000000000000000	513100000	110000000	410000000
56	2	1.0	4.0	511514	311312000100101000000	230000000	610000000	000000000
57	2	1.0	-8.0	511514	311312000100101000000	210000000	610000000	120000000
58	1	1.0	4.0	511524	311312000100100000000	225100000	610000000	000000000
59	1	2.0	2.0	511544	311322000100100000000	220000000	520000000	000000000
60	2	1.0	8.0	512514	311312000100000000000	223100000	610000000	000000000

Table (4.11.14) contd.

61	2	1.0	-8.0	512514	311312001000100010000000	310000000	610000000	120000000
62	2	1.0	8.0	512524	311312001000000010000	213200000	610000000	000000000
63	2	1.0	-8.0	512524	311312001000000010000	210000000	610000000	120000000
64	1	1.0	8.0	512534	311312001000000000000	21315100	610000000	000000000
65	1	2.0	4.0	512544	311322001000000000001	213100000	520000000	000000000
66	2	1.0	4.0	514511	312311001000011000000000	230000000	310000000	000000000
67	2	1.0	-8.0	514511	312311001000011000000000	210000000	310000000	120000000
68	2	1.0	8.0	514512	312311001000010000000000	226100000	310000000	000000000
69	2	1.0	-8.0	514512	312311001000010000000000	610000000	310000000	120000000
70	4	2.0	8.0	514514	3123120010000100001000	220000000	320000000	000000000
71	4	0	-8.0	514514	3123120010000100001000	220000000	000000000	120000000
72	4	2.0	-8.0	514514	3123120010000100001000	000000000	320000000	120000000
73	4	0	16.0	514514	3123120010000100001000	220000000	320000000	120000000
74	1	1.0	4.0	514522	312311001000000000001	215200000	310000000	000000000
75	1	1.0	8.0	514523	312311001000000000000	21516100	310000000	000000000
76	3	2.0	8.0	514524	3123120010000000001000	215100000	320000000	000000000
77	3	0	-8.0	514524	3123120010000000001000	215100000	000000000	120000000
78	3	0	16.0	514524	3123120010000000001000	215100000	320000000	120000000
79	3	3.0	4.0	514544	31232200100000000011001	210000000	330000000	000000000
80	3	1.0	-8.0	514544	31232200100000000011001	210000000	310000000	120000000
81	3	1.0	12.0	514544	31232200100000000011001	210000000	330000000	120000000
82	2	1.0	8.0	524512	3123110010000100000000	512200000	310000000	000000000
83	2	1.0	-8.0	524512	3123110010000100000000	510000000	310000000	120000000
84	1	2.0	2.0	544511	322311001001000000001	520000000	220000000	000000000
85	1	2.0	4.0	544512	322311001001000000000	516100000	220000000	000000000
86	3	3.0	4.0	544514	3223120010010010001000	510000000	230000000	000000000
87	3	1.0	-8.0	544514	3223120010010010001000	510000000	210000000	120000000
88	3	1.0	12.0	544514	3223120010010010001000	510000000	230000000	120000000
89	3	4.0	2.0	544544	3223220010010011011001	000000000	240000000	000000000
90	3	2.0	-8.0	544544	3223220010010011011001	000000000	220000000	120000000
91	3	2.0	8.0	544544	3223220010010011011001	000000000	240000000	120000000
92	1	1.0	-8.0	511614	311312000001001000000	210000000	610000000	114100000
93	1	1.0	-8.0	512614	311312000000100000000	310000000	610000000	114100000
94	1	1.0	-8.0	512624	311312000000000010000	210000000	610000000	114100000
95	1	1.0	-8.0	514611	312311000000100000001	210000000	310000000	114100000
96	1	1.0	-8.0	514612	312311000000100000000	610000000	310000000	114100000
97	3	0	-8.0	514614	3123120000000010001000	220000000	000000000	114100000
98	3	2.0	-8.0	514614	3123120000000010001000	000000000	320000000	114100000
99	3	0	16.0	514614	3123120000000010001000	220000000	320000000	114100000
100	2	0	-8.0	514624	312312000000000010000	215100000	000000000	114100000
101	2	0	16.0	514624	312312000000000010000	215100000	320000000	114100000
102	2	1.0	-8.0	514644	3123220000000000011001	210000000	310000000	114100000
103	2	1.0	12.0	514644	3123220000000000011001	210000000	330000000	114100000
104	1	1.0	-8.0	524612	312311000000000010000	510000000	310000000	114100000
105	2	1.0	-8.0	544614	322312000000100010001	510000000	210000000	114100000
106	2	1.0	12.0	544614	322312000000100010001	510000000	230000000	114100000
107	2	2.0	-8.0	544644	3223220000001011011001	000000000	220000000	114100000
108	2	2.0	8.0	544644	3223220000001011011001	000000000	240000000	114100000

Table (4.11.15) Representation of the terms in the expressions of the product of the first partial derivatives of $r_1^{(+)s}$ for the case when $p=4$ and $k^*=2$

1	1	2.0	4.0	311311	211211001001011011001	240000000	120000000	000000000
2	1	2.0	8.0	311312	21121100100101010010000	236100000	120000000	000000000
3	1	2.0	4.0	311322	211211001001000000000	225200000	120000000	000000000
4	1	2.0	8.0	312311	21121100100000110000001	233100000	120000000	000000000
5	1	2.0	16.0	312312	21121100100000100010000	223200000	120000000	000000000
6	1	2.0	8.0	312322	21121100100000000011001	213300000	120000000	000000000
7	1	2.0	8.0	322312	21121100100010010010000	512300000	120000000	000000000
8	1	2.0	4.0	311411	21121100000010010110001	240000000	114100000	000000000
9	1	2.0	8.0	311412	21121100000010100100000	236100000	114100000	000000000
10	1	2.0	4.0	311422	21121100000010000000001	225200000	114100000	000000000
11	1	2.0	8.0	312411	21121100000000110000001	233100000	114100000	000000000
12	1	2.0	16.0	312412	21121100000000100010000	223200000	114100000	000000000
13	1	2.0	8.0	312422	21121100000000000011001	213300000	114100000	000000000
14	1	2.0	8.0	322412	21121100000010010010000	512300000	114100000	000000000
15	1	2.0	-8.0	311513	21131200000110100100000	230000000	120000000	410000000
16	1	2.0	-8.0	311514	21131200000101001000000	230000000	116100000	410000000
17	1	2.0	-8.0	311523	21131200001100000000000	225100000	120000000	410000000
18	1	2.0	-8.0	311524	21131200000100000000000	225100000	116100000	410000000
19	1	3.0	-4.0	311533	21132200001110000000001	220000000	130000000	410000000
20	1	3.0	-8.0	311534	21132200001010000000000	220000000	126100000	410000000
21	1	3.0	-4.0	311544	21132200000100000000001	220000000	115200000	410000000
22	1	2.0	-16.0	312513	21131200001001000000000	223100000	120000000	410000000
23	1	2.0	-16.0	312514	21131200000010000000000	223100000	116100000	410000000
24	1	2.0	-16.0	312523	211312000010000010000	213200000	120000000	410000000
25	1	2.0	-16.0	312524	211312000000000010000	213200000	116100000	410000000
26	1	3.0	-8.0	312533	21132200001100000000000	213100000	130000000	410000000
27	1	3.0	-16.0	312534	21132200001000000000000	213100000	126100000	410000000
28	1	3.0	-8.0	312544	21132200000000000000001	213100000	115200000	410000000
29	1	3.0	-8.0	411534	21132200001100000000000	220000000	511200000	410000000
30	1	3.0	-16.0	412534	21132200001000000000000	213100000	511200000	410000000
31	1	2.0	16.0	513513	31231200010000100010000	220000000	320000000	120000000
32	1	2.0	16.0	513514	31231200010000010000000	220000000	316100000	120000000
33	1	2.0	16.0	513523	31231200010000000000000	215100000	320000000	120000000
34	1	2.0	16.0	513524	31231200010000000000000	215100000	316100000	120000000
35	1	3.0	8.0	513533	31232200010000000000001	210000000	330000000	120000000
36	1	3.0	16.0	513534	31232200010000000000000	210000000	326100000	120000000
37	1	3.0	8.0	513544	31232200010000000000000	210000000	315200000	120000000
38	1	3.0	16.0	514534	31232200010000000000000	210000000	513200000	120000000
39	1	3.0	8.0	533513	32231200010000100010000	510000000	230000000	120000000
40	1	3.0	8.0	533514	32231200010000000000000	510000000	226100000	120000000
41	1	4.0	4.0	533533	32232200010000000000000	000000000	240000000	120000000
42	1	4.0	8.0	533534	32232200010000000000000	000000000	236100000	120000000
43	1	4.0	4.0	533544	32232200010000000000000	000000000	225200000	120000000
44	1	3.0	16.0	534513	32231200010000000000000	510000000	223100000	120000000
45	1	3.0	16.0	534514	32231200010000000000000	510000000	213200000	120000000
46	1	4.0	8.0	534533	32232200010000000000000	000000000	233100000	120000000
47	1	4.0	16.0	534534	32232200010000000000000	000000000	223200000	120000000
48	1	4.0	8.0	534544	32232200010000000000000	000000000	213300000	120000000
49	1	4.0	8.0	544534	32232200010000000000000	000000000	512300000	120000000
50	1	2.0	16.0	513613	31231200000010001000000	220000000	320000000	114100000
51	1	2.0	16.0	513614	31231200000001000000000	220000000	315100000	114100000
52	1	2.0	16.0	513623	31231200000000000000000	215100000	320000000	114100000
53	1	2.0	16.0	513624	31231200000000000000000	215100000	316100000	114100000
54	1	3.0	8.0	513633	31232200000000000000000	210000000	330000000	114100000
55	1	3.0	16.0	513634	31232200000000000000000	210000000	326100000	114100000
56	1	3.0	8.0	513644	31232200000000000000000	210000000	315200000	114100000
57	1	3.0	16.0	514634	31232200000000000000000	210000000	513200000	114100000
58	1	3.0	8.0	533613	32231200000010001000000	510000000	230000000	114100000
59	1	3.0	8.0	533614	32231200000000000000000	510000000	226100000	114100000
60	1	4.0	4.0	533633	32232200000010110110001	000000000	240000000	114100000
61	1	4.0	8.0	533634	32232200000010110100000	000000000	236100000	114100000
62	1	4.0	4.0	533644	32232200000010000000000	000000000	225200000	114100000
63	1	3.0	16.0	534613	32231200000010001000000	510000000	225100000	114100000
64	1	3.0	16.0	534614	32231200000000000000000	510000000	213300000	114100000
65	1	4.0	8.0	534633	32232200000000110000000	000000000	233100000	114100000
66	1	4.0	16.0	534634	32232200000000100000000	000000000	223200000	114100000
67	1	4.0	8.0	534644	32232200000000000000000	000000000	213300000	114100000
68	1	4.0	8.0	544634	32232200000010001000000	000000000	512300000	114100000

Table (4.11.16) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(s)}$ for the case when $p=4$ and $k^*=2$

1	3	0	-2.0	311311	211211001001011011001	240000000	000000000	000000000
2	3	2.0	0.0	311311	211211001001011011001	220000000	120000000	000000000
3	3	0	4.0	311311	211211001001011011001	240000000	120000000	000000000
4	3	0	-4.0	311312	211211001001011000000	234100000	000000000	000000000
5	3	2.0	8.0	311312	211211001001011000000	216100000	120000000	000000000
6	3	0	8.0	311312	211211001001011000000	236100000	120000000	000000000
7	2	0	-2.0	311322	211211001001000000001	225200000	000000000	000000000
8	2	0	4.0	311322	211211001001000000001	225200000	120000000	000000000
9	3	0	-4.0	312311	2112110010000011000001	233100000	000000000	000000000
10	3	2.0	8.0	312311	2112110010000011000001	213100000	120000000	000000000
11	3	0	8.0	312311	2112110010000011000001	233100000	120000000	000000000
12	4	0	-8.0	312312	2112110010000010001000	223200000	000000000	000000000
13	4	2.0	8.0	312312	2112110010000010001000	220000000	120000000	000000000
14	4	2.0	8.0	312312	2112110010000010001000	320000000	120000000	000000000
15	4	0	16.0	312312	2112110010000010001000	223200000	120000000	000000000
16	3	0	-4.0	312322	211211001000000110001	213300000	000000000	000000000
17	3	2.0	8.0	312322	211211001000000110001	213100000	120000000	000000000
18	3	0	8.0	312322	211211001000000110001	213300000	120000000	000000000
19	3	0	-4.0	322312	211211001001001000000	512300000	000000000	000000000
20	3	2.0	8.0	322312	211211001001001000000	512100000	120000000	000000000
21	3	0	8.0	322312	211211001001001000000	512300000	120000000	000000000
22	2	2.0	8.0	311411	2112110000001011011001	220000000	114100000	000000000
23	2	0	4.0	311411	2112110000001011011001	240000000	114100000	000000000
24	2	2.0	8.0	311412	2112110000001010010000	216100000	114100000	000000000
25	2	0	8.0	311412	2112110000001010010000	236100000	114100000	000000000
26	1	0	4.0	311422	211211000001000000001	225200000	114100000	000000000
27	2	2.0	8.0	312411	2112110000000011000001	213100000	114100000	000000000
28	2	0	8.0	312411	2112110000000011000001	233100000	114100000	000000000
29	3	2.0	8.0	312412	2112110000000010001000	220000000	114100000	000000000
30	3	2.0	8.0	312412	2112110000000010001000	320000000	114100000	000000000
31	3	0	16.0	312412	2112110000000010001000	223200000	114100000	000000000
32	2	2.0	8.0	312422	2112110000000000110001	213100000	114100000	000000000
33	2	0	8.0	312422	2112110000000000110001	213300000	114100000	000000000
34	2	2.0	8.0	322412	2112110000000010001000	512100000	114100000	000000000
35	2	0	8.0	322412	2112110000000010001000	512300000	114100000	000000000
36	1	1.0	8.0	311511	2113110000001011011001	220000000	110000000	410000000
37	1	1.0	8.0	311512	2113110000001010010000	216100000	110000000	410000000
38	1	0	-8.0	311513	2113120000010100100000	230000000	120000000	410000000
39	1	0	-8.0	311514	2113120000010100100000	230000000	116100000	410000000
40	1	0	-8.0	311523	211312000001000000000	225100000	120000000	410000000
41	1	0	-8.0	311524	211312000000100000000	225100000	116100000	410000000
42	1	1.0	-6.0	311533	21132200000100000000001	220000000	130000000	410000000
43	1	1.0	-12.0	311534	21132200000101000000000	220000000	126100000	410000000
44	1	1.0	-6.0	311544	21132200000010000000000	220000000	115200000	410000000
45	1	1.0	8.0	312511	2113110000000011000001	213100000	110000000	410000000
46	2	1.0	8.0	312512	2113110000000010000000	220000000	110000000	410000000
47	2	1.0	8.0	312512	2113110000000010000000	320000000	110000000	410000000
48	1	0	-16.0	312513	21131200000100000000000	223100000	120000000	410000000
49	1	0	-16.0	312514	21131200000010000000000	223100000	116100000	410000000
50	1	1.0	8.0	312522	2113110000000000110001	213100000	110000000	410000000
51	1	0	-16.0	312523	2113120000000000100000	213200000	120000000	410000000
52	1	0	-16.0	312524	2113120000000000010000	213200000	116100000	410000000
53	1	1.0	-12.0	312533	21132200000100000000001	213100000	130000000	410000000
54	1	1.0	-24.0	312534	21132200000100000000000	213100000	126100000	410000000
55	1	1.0	-12.0	312544	21132200000000000000000	213100000	115200000	410000000
56	1	1.0	8.0	322512	2113110000001001000000	512100000	110000000	410000000
57	1	1.0	-12.0	411534	21132200000110000000000	220000000	511200000	410000000
58	1	1.0	-24.0	412534	21132200000100000000000	213100000	511200000	410000000
59	2	1.0	4.0	511513	311312000100100100000	230000000	610000000	000000000
60	2	1.0	-8.0	511513	311312000100100100000	210000000	610000000	120000000

Table (4.11.16) contd.

61	1	1.0	4.0	511523	31131200100100000000	22510000	61000000	000000000
62	1	2.0	2.0	511533	31132200100100000000	22040000	52000000	000000000
63	1	2.0	4.0	511534	31132200100100000000	22000000	51610000	000000000
64	2	1.0	8.0	512513	31131200100000100000	22310000	61000000	000000000
65	2	1.0	-8.0	512513	31131200100000100000	31000000	61000000	120000000
66	2	1.0	8.0	512523	311312001000101410000	21320000	61000000	000000000
67	2	1.0	-8.0	512523	31131200100000000000	21000000	61000000	120000000
68	1	2.0	4.0	512543	31132200100000000000	21310000	52000000	000000000
69	1	2.0	8.0	512534	31132200100000000000	21310000	51610000	000000000
70	2	1.0	4.0	513511	31231100100001100000	23000000	31000000	000000000
71	2	1.0	-8.0	513511	31231100100001100000	21000000	31000000	120000000
72	2	1.0	8.0	513512	31231100100001000000	22610000	31000000	000000000
73	2	1.0	-8.0	513512	31231100100001000000	61000000	31000000	120000000
74	4	2.0	8.0	513513	31231200100001000100	22000000	32000000	000000000
75	4	0	-8.0	513513	31231200100001000100	22000000	00000000	120000000
76	4	2.0	-8.0	513513	31231200100001000100	00000000	32000000	120000000
77	4	0	16.0	513513	31231200100001000100	22000000	32000000	120000000
78	3	2.0	8.0	513514	31231200100001000000	22000000	31610000	000000000
79	3	2.0	-8.0	513514	31231200100001000000	00000000	31610000	120000000
80	3	0	16.0	513514	31231200100001000000	22000000	31610000	120000000
81	1	1.0	4.0	513522	31231100100000000001	21520000	31000000	000000000
82	3	2.0	8.0	513523	31231200100000000000	21510000	32000000	000000000
83	3	0	-8.0	513523	31231200100000000000	21510000	00000000	120000000
84	3	0	16.0	513523	31231200100000000000	21510000	32000000	120000000
85	2	2.0	8.0	513524	31231200100000000000	21510000	31610000	000000000
86	2	0	16.0	513524	31231200100000000000	21510000	31610000	120000000
87	3	3.0	4.0	513533	3123220010000000011001	21000000	33000000	000000000
88	3	1.0	-8.0	513533	3123220010000000011001	21000000	31000000	120000000
89	3	1.0	12.0	513533	3123220010000000011001	21000000	33000000	120000000
90	3	3.0	8.0	513534	3123220010000000011000	21000000	32610000	000000000
91	3	1.0	-8.0	513534	3123220010000000011000	21000000	61000000	120000000
92	3	1.0	24.0	513534	3123220010000000011000	21000000	32610000	120000000
93	2	3.0	4.0	513544	31232200100000000001	21000000	31520000	000000000
94	2	1.0	12.0	513544	31232200100000000001	21000000	31520000	120000000
95	3	3.0	8.0	514534	31232200100000000000	21000000	51320000	000000000
96	3	1.0	-8.0	514534	31232200100000000000	21000000	51000000	120000000
97	3	1.0	24.0	514534	31232200100000000000	21000000	51320000	120000000
98	2	1.0	8.0	523512	31231100100000000000	51220000	31000000	000000000
99	2	1.0	-8.0	523512	31231100100000000000	51000000	31000000	120000000
100	1	2.0	2.0	533511	32231100100100000001	52000000	22000000	000000000
101	1	2.0	4.0	533512	32231100100100000000	51610000	22000000	000000000
102	3	3.0	4.0	533513	32231200100100100000	51000000	23000000	000000000
103	3	1.0	-8.0	533513	32231200100100100000	51000000	21000000	120000000
104	3	1.0	12.0	533513	32231200100100100000	51000000	23000000	120000000
105	2	3.0	4.0	533514	32231200100100000000	51000000	22610000	000000000
106	2	1.0	12.0	533514	32231200100100000000	51000000	22610000	120000000
107	3	4.0	2.0	533533	32232200100100111001	00000000	24000000	000000000
108	3	2.0	-8.0	533533	32232200100100111001	00000000	22000000	120000000
109	3	2.0	8.0	533533	32232200100100111001	00000000	24000000	120000000
110	3	4.0	4.0	533534	32232200100100111000	00000000	23610000	000000000
111	3	2.0	-8.0	533534	32232200100100111000	00000000	21610000	120000000
112	3	2.0	16.0	533534	32232200100100111000	00000000	23610000	120000000
113	2	4.0	2.0	533544	32232200100100000001	00000000	22520000	000000000
114	2	2.0	8.0	533544	32232200100100000001	00000000	22520000	120000000
115	1	2.0	4.0	534511	32231100100000000001	52000000	21310000	000000000
116	1	2.0	8.0	534512	32231100100000000000	51610000	21310000	000000000
117	3	3.0	8.0	534513	32231200100000001000	51000000	22310000	000000000
118	3	1.0	-8.0	534513	32231200100000001000	51000000	31000000	120000000
119	3	1.0	24.0	534513	32231200100000001000	51000000	22310000	120000000
120	3	3.0	8.0	534514	32231200100000001000	51000000	21320000	000000000

Table (4.11.16) contd.

121	3	1.0	-8.0	534514	322312001000000001000	510000000	210000000	120000000
122	3	1.0	24.0	534514	322312001000000001000	510000000	213200000	120000000
123	3	4.0	4.0	534533	322322001000011000001	000000000	233100000	000000000
124	3	2.0	-8.0	534533	322322001000011000001	000000000	213100000	120000000
125	3	2.0	16.0	534533	322322001000011000001	000000000	233100000	120000000
126	4	4.0	8.0	534534	322322001000011000100	000000000	223200000	000000000
127	4	2.0	-4.0	534534	322322001000011000100	000000000	220000000	120000000
128	4	2.0	-8.0	534534	322322001000011000100	000000000	320000000	120000000
129	4	2.0	32.0	534534	322322001000010001000	000000000	223200000	120000000
130	3	4.0	4.0	534544	322322001000011000100	000000000	213300000	000000000
131	3	2.0	-8.0	534544	322322001000000011001	000000000	213100000	120000000
132	3	2.0	16.0	534544	322322001000000011001	000000000	213300000	120000000
133	3	4.0	4.0	544534	322322001000100100100	000000000	512300000	000000000
134	3	2.0	-8.0	544534	322322001000100100100	000000000	512100000	120000000
135	3	2.0	16.0	544534	322322001000100100100	000000000	512300000	120000000
136	1	1.0	-8.0	511613	311312000001001000000	210000000	610000000	114100000
137	1	1.0	-8.0	512613	311312000000100000000	310000000	610000000	114100000
138	1	1.0	-8.0	512623	311312000000000000000	210000000	610000000	114101000
139	1	1.0	-8.0	513611	312311000000011000001	210000000	310000000	114100000
140	1	1.0	-8.0	513612	312311000000010000000	610000000	310000000	114100000
141	3	0	-8.0	513613	312312000000010001000	220000000	000000000	114100000
142	3	2.0	-8.0	513613	312312000000010001000	000000000	320000000	114100000
143	3	0	16.0	513613	312312000000010001000	220000000	320000000	114100000
144	2	2.0	-8.0	513614	312312000000010000000	000000000	316100000	114100000
145	2	0	16.0	513614	312312000000010000000	220000000	316100000	114100000
146	2	0	-8.0	513623	312312000000000000000	215100000	000000000	114100000
147	2	0	16.0	513623	312312000000000000000	215100000	320000000	114100000
148	1	0	16.0	513624	312312000000000000000	215100000	316100000	114100000
149	2	1.0	-8.0	513633	312322000000000011001	210000000	310000000	114100000
150	2	1.0	12.0	513633	312322000000000011001	210000000	330000000	114100000
151	2	1.0	-8.0	513634	312322000000000010000	210000000	610000000	114100000
152	2	1.0	24.0	513634	312322000000000010000	210000000	326100000	114100000
153	1	1.0	12.0	513644	312322000000000000001	210000000	315200000	114100000
154	2	1.0	-8.0	514634	312322000000000000001	210000000	510000000	114100000
155	2	1.0	24.0	514634	312322000000000000001	210000000	513200000	114100000
156	1	1.0	-8.0	523612	312311000000001000000	510000000	310000000	114100000
157	2	1.0	-8.0	533613	322312000001001001000	510000000	210000000	114100000
158	2	1.0	12.0	533613	322312000001001001000	510000000	230000000	114100000
159	1	1.0	12.0	533614	322312000001000000000	510000000	226100000	114100000
160	2	2.0	-8.0	533633	322322000001011011001	000000000	220000000	114100000
161	2	2.0	8.0	533633	322322000001011011001	000000000	240000000	114100000
162	2	2.0	-8.0	533634	322322000001010010000	000000000	216100000	114100000
163	2	2.0	16.0	533634	322322000001010010000	000000000	236100000	114100000
164	1	2.0	8.0	533644	322322000001000000001	000000000	225200000	114100000
165	2	1.0	-8.0	534613	322312000000100000000	510000000	310000000	114100000
166	2	1.0	24.0	534613	322312000000100000000	510000000	223100000	114100000
167	2	1.0	-8.0	534614	32231200000000001000	510000000	210000000	114100000
168	2	1.0	24.0	534614	32231200000000001000	510000000	213200000	114100000
169	2	2.0	-8.0	534633	322322000000110000001	000000000	213100000	114100000
170	2	2.0	16.0	534633	322322000000110000001	000000000	233100000	114100000
171	3	2.0	-8.0	534634	322322000000100000000	000000000	220000000	114100000
172	3	2.0	-8.0	534634	322322000000010001000	000000000	320000000	114100000
173	3	2.0	32.0	534634	322322000000010001000	000000000	223200000	114100000
174	2	2.0	-8.0	534644	322322000000000011001	000000000	213100000	114100000
175	2	2.0	16.0	534644	322322000000000011001	000000000	213200000	114100000
176	2	2.0	-8.0	544634	322322000001001001000	000000000	512100000	114100000
177	2	2.0	16.0	544634	322322000001001001000	000000000	512300000	114100000

**Table (4.11.17) Representation of the terms in the expressions
of the product of the first partial derivatives of $r_1^{(7s)}$ for
the case when $p=4$ and $k^*=3$**

1	1	2.0	4.0	211211	2112110001001011011001	240000000	120000000	000000000
2	1	2.0	4.0	211511	211511000001011011001	240000000	114100000	000000000
3	1	2.0	-8.0	211512	2115120000011010001000	230000000	120000000	410000000
4	1	2.0	-8.0	211513	2115130000010110001000	230000000	116100000	410000000
5	1	3.0	-4.0	211522	21152200011100000001	220000000	130000000	410000000
6	1	3.0	-4.0	211523	21152300010100000000	220000000	126100000	410000000
7	1	3.0	-4.0	211533	21153300010000000000	220000000	115200000	410000000
8	1	3.0	-8.0	211534	21153400010000000000	220000000	115161000	410000000
9	1	3.0	-8.0	311523	21152300011000000000	220000000	511200000	410000000
10	1	2.0	16.0	512512	312512001000010001000	220000000	320000000	120000000
11	1	2.0	16.0	512513	312513001000010000000	220000000	316100000	120000000
12	1	3.0	8.0	512522	312522001000000010001	210000000	330000000	120000000
13	1	3.0	16.0	512523	312523001000000001000	210000000	326100000	120000000
14	1	3.0	8.0	512533	31253300100000000001	210000000	315200000	120000000
15	1	3.0	16.0	512534	31253400100000000000	210000000	315161000	120000000
16	1	3.0	16.0	513523	31232200100000000000	210000000	513200000	120000000
17	1	3.0	8.0	522512	322512001001001001000	510000000	230000000	120000000
18	1	3.0	8.0	522513	32251300100100000000	510000000	226100000	120000000
19	1	4.0	4.0	522522	322522001001011011001	000000000	240000000	120000000
20	1	4.0	8.0	522523	322523001001010010000	000000000	236100000	120000000
21	1	4.0	4.0	522533	32252300100100000000	000000000	225200000	120000000
22	1	4.0	8.0	522534	32252300100100000000	000000000	225161000	120000000
23	1	3.0	16.0	523512	32235120010000000000	510000000	223100000	120000000
24	1	3.0	16.0	523513	32235130010000000000	510000000	213200000	120000000
25	1	3.0	16.0	523514	32235140010030000000	510000000	213161000	120000000
26	1	4.0	8.0	523522	32235220010000000000	000000000	233100000	120000000
27	1	4.0	16.0	523523	32235230010000000000	000000000	223200000	120000000
28	1	4.0	16.0	523524	32235240010000000000	000000000	223161000	120000000
29	1	4.0	8.0	523533	32235330010000000000	000000000	213300000	120000000
30	1	4.0	16.0	523534	32235340010000000000	000000000	213261000	120000000
31	1	4.0	8.0	523544	32235440010000000000	000000000	213152000	120000000
32	1	4.0	16.0	524534	32245340010000000000	000000000	215132000	120000000
33	1	4.0	8.0	533523	32235230010010010000	000000000	512300000	120000000
34	1	4.0	16.0	534523	32234523001000000000	000000000	512231000	120000000
35	1	2.0	16.0	512612	31261200000000000000	220000000	320000000	114100000
36	1	2.0	16.0	512613	31261300000000000000	220000000	316100000	114100000
37	1	3.0	8.0	512622	31262200000000000000	210000000	330000000	114100000
38	1	3.0	16.0	512623	31262300000000000000	210000000	326100000	114100000
39	1	3.0	8.0	512633	31263300000000000000	210000000	315200000	114100000
40	1	3.0	16.0	512634	31263400000000000000	210000000	315161000	114100000
41	1	3.0	16.0	513523	31235230000000000000	210000000	513200000	114100000
42	1	3.0	8.0	522612	32261200000000000000	510000000	230000000	114100000
43	1	3.0	8.0	522613	32261300000000000000	510000000	226100000	114100000
44	1	4.0	4.0	522622	32262200000000000000	000000000	240000000	114100000
45	1	4.0	8.0	522623	32262300000000000000	000000000	236100000	114100000
46	1	4.0	4.0	522633	32262633000000000000	000000000	225200000	114100000
47	1	4.0	8.0	522634	32262634000000000000	000000000	225161000	114100000
48	1	3.0	16.0	523612	32236120000000000000	510000000	223100000	114100000
49	1	3.0	16.0	523613	32236130000000000000	510000000	213200000	114100000
50	1	3.0	16.0	523614	32236140000000000000	510000000	213161000	114100000
51	1	4.0	8.0	523622	32236220000000000000	000000000	233100000	114100000
52	1	4.0	16.0	523623	32236230000000000000	000000000	223200000	114100000
53	1	4.0	16.0	523624	32236240000000000000	000000000	223161000	114100000
54	1	4.0	8.0	523633	32236330000000000000	000000000	213300000	114100000
55	1	4.0	16.0	523634	32236340000000000000	000000000	213261000	114100000
56	1	4.0	8.0	523644	32236440000000000000	000000000	213162000	114100000
57	1	4.0	16.0	524634	32246340000000000000	000000000	215132000	114100000
58	1	4.0	8.0	533623	32233623000000000000	000000000	512300000	114100000
59	1	4.0	16.0	534623	32234623000000000000	000000000	512231000	114100000

Table (4.11.18) Representation of the terms in the expressions of the second partial derivatives of $r_1^{(+s)}$ for the case when $p=4$ and $k^*=3$

1	3	0	-2.0	211211	211211001001011011001	24000000	00010000	00000000
2	3	2.0	8.0	211211	211211001001011011001	22000000	12000000	00000000
3	3	0	4.0	211211	211211001001011011001	24000000	12000000	00000000
4	2	2.0	8.0	211311	211311000001011011001	22000000	11410000	00000000
5	2	0	4.0	211311	211311000001011011001	24000000	11410000	00000000
6	1	1.0	8.0	211511	211311000001011011001	22000000	11000000	41000000
7	1	0	-8.0	211512	211312000011010010000	23000000	12000000	41000000
8	1	0	-8.0	211513	211312000010100100000	23000000	11610000	41000000
9	1	1.0	-6.0	211522	211322000011100000001	22000000	13000000	41000000
10	1	1.0	-12.0	211523	211322000010100000000	22000000	12610000	41000000
11	1	1.0	-6.0	211533	211322000010000000001	22000000	11520000	41000000
12	1	1.0	-12.0	211534	211322000010000000000	22000000	11516100	41000000
13	1	1.0	-12.0	311523	211322000011000000000	22000000	51120000	41000000
14	2	1.0	4.0	511512	31131200100101001000	23000000	61000000	00000000
15	2	1.0	-8.0	511512	31131200100101001000	21000000	61000000	12000000
16	1	2.0	2.0	511522	31132200100100000001	22000000	52000000	00000000
17	1	2.0	4.0	511523	311322001001000000000	22000000	51610000	00000000
18	2	1.0	4.0	512511	312311001000011000001	23000000	31000000	00000000
19	2	1.0	-8.0	512511	312311001000011000001	21000000	31000000	12000000
20	4	2.0	8.0	512512	312312001000011000000	22000000	32000000	00000000
21	4	0	-8.0	512512	312312001000011000000	22000000	00900000	12000000
22	4	2.0	-8.0	512512	312312001000010000000	00900000	32000000	12000000
23	4	0	16.0	512512	312312001000010000000	22000000	32000000	12000000
24	3	2.0	8.0	512513	312312001000010000000	22000000	31610000	00000000
25	3	2.0	-8.0	512513	312312001000010000000	00900000	31610000	12000000
26	3	0	16.0	512513	312312001000010000000	22000000	31610000	12000000
27	3	3.0	4.0	512522	312322001000000011001	21000000	33000000	00000000
28	3	1.0	-8.0	512522	312322001000000011001	21000000	31000000	12000000
29	3	1.0	12.0	512522	312322001000000011001	21000000	33000000	12000000
30	3	3.0	8.0	512522	312322001000000011000	21000000	32610000	00000000
31	3	1.0	-8.0	512523	312322001000000010000	21000000	61000000	12000000
32	3	1.0	24.0	512523	312322001000000010000	21000000	32610000	12000000
33	2	3.0	4.0	512533	31232200100000000001	21000000	31520000	00000000
34	2	1.0	12.0	512533	31232200100000000001	21000000	31520000	12000000
35	2	3.0	8.0	512534	312322001000000000000	21000000	31516100	00000000
36	2	1.0	24.0	512534	312322001000000000000	21000000	31516100	12000000
37	3	3.0	8.0	513523	312322001000000000000	21000000	51320000	00000000
38	3	1.0	-8.0	513523	312322001000000000000	21000000	51000000	12000000
39	3	1.0	24.0	513523	312322001000000000000	21000000	51320000	12000000
40	1	2.0	2.0	522511	32231100100100000001	52000000	22000000	00000000
41	3	3.0	4.0	522512	322312001001001001000	51000000	23000000	00000000
42	3	1.0	-8.0	522512	322312001001001001000	51000000	21000000	12000000
43	3	1.0	12.0	522512	322312001001001001000	51000000	23000000	12000000
44	2	3.0	4.0	522513	322312001001000000000	51000000	22610000	00000000
45	2	1.0	12.0	522513	322312001001000000000	51000000	22610000	12000000
46	3	4.0	2.0	522522	322322001001001011001	00000000	24000000	00000000
47	3	2.0	-8.0	522522	322322001001001011001	00000000	22000000	12000000
48	3	2.0	8.0	522522	322322001001001011001	00000000	24000000	12000000
49	3	4.0	4.0	522523	322322001001001011000	00000000	23610000	00000000
50	3	2.0	-8.0	522523	322322001001001011000	00000000	21610000	12000000
51	3	2.0	16.0	522523	322322001001001011000	00000000	23610000	12000000
52	2	4.0	2.0	522533	32232200100100000001	00000000	22520000	00000000
53	2	2.0	8.0	522533	32232200100100000001	00000000	22520000	12000000
54	2	4.0	4.0	522534	322322001001000000000	00000000	22516100	00000000
55	2	2.0	16.0	522534	322322001001000000000	00000000	22516100	12000000
56	1	2.0	4.0	523511	32231100100000000001	52000000	21310000	00000000
57	3	3.0	8.0	523512	322312001000000000000	51000000	22310000	00000000
58	3	1.0	-8.0	523512	322312001000000000000	51000000	31000000	12000000
59	3	1.0	24.0	523512	322312001000000000000	51000000	22310000	12000000
60	3	3.0	8.0	523513	322312001000000000000	51000000	21320000	00000000
61	3	1.0	-8.0	523513	322312001000000000000	51000000	21000000	12000000
62	3	1.0	24.0	523513	322312001000000000000	51000000	21320000	12000000
63	2	3.0	8.0	523514	322312001000000000000	51000000	21316100	00000000
64	2	1.0	24.0	523514	322312001000000000000	51000000	21316100	12000000
65	3	4.0	4.0	523522	32232200100000000001	00000000	23310000	00000000

Table (4.11.18) contd.

66	3	2.0	-8.0	523522	322322001000011020001	000000000	21310000	12000000
67	3	2.0	16.0	523522	322322001000011020001	000000000	21310000	12000000
68	4	4.0	8.0	523523	322322001000011020001	000000000	22320000	00000000
69	4	2.0	-8.0	523523	322322001000011020001	000000000	220000000	120000000
70	4	2.0	-8.0	523523	322322001000011020001	000000000	320000000	120000000
71	4	2.0	32.0	523523	322322001000011020001	000000000	22320000	120000000
72	3	4.0	8.0	523524	322322001000011020001	000000000	22316100	000000000
73	3	2.0	-8.0	523524	322322001000011020001	000000000	31610000	120000000
74	3	2.0	32.0	523524	322322001000011020001	000000000	22316100	120000000
75	3	4.0	4.0	523533	322322001000011020001	000000000	21330000	000000000
76	3	2.0	-8.0	523533	322322001000011020001	000000000	21310000	120000000
77	3	2.0	16.0	523533	322322001000011020001	000000000	21330000	120000000
78	3	4.0	8.0	523534	322322001000011020001	000000000	21326100	000000000
79	3	2.0	-8.0	523534	322322001000011020001	000000000	21610000	120000000
80	3	2.0	32.0	523534	322322001000011020001	000000000	21326100	120000000
81	2	4.0	4.0	523544	322322001000011020001	000000000	21315200	000000000
82	2	2.0	16.0	523544	322322001000011020001	000000000	21315200	120000000
83	3	4.0	8.0	524534	322322001000011020001	000000000	21513200	000000000
84	3	2.0	-8.0	524534	322322001000011020001	000000000	21510000	120000000
85	3	2.0	32.0	524534	322322001000011020001	000000000	21513200	120000000
86	3	4.0	4.0	533523	322322001001001001000	000000000	51230000	000000000
87	3	2.0	-8.0	533523	322322001001001001000	000000000	51210000	120000000
88	3	2.0	16.0	533523	322322001001001001000	000000000	51230000	120000000
89	3	4.0	8.0	534523	322322001000010000001	000000000	51223100	000000000
90	3	2.0	-8.0	534523	322322001000010000001	000000000	51310000	120000000
91	3	2.0	32.0	534523	322322001000010000001	000000000	51223100	120000000
92	1	1.0	-8.0	511612	311312000001010010000	210000000	610000000	114100000
93	1	1.0	-8.0	512611	312311000000110000001	210000000	310000000	114100000
94	3	0	-8.0	512612	312312000000010000001	220000000	000000000	114100000
95	3	2.0	-8.0	512612	312312000000010000001	000000000	320000000	114100000
96	3	0	16.0	512612	312312000000010000001	220000000	320000000	114100000
97	2	2.0	-8.0	512613	312312000000010000001	000000000	31610000	114100000
98	2	0	16.0	512613	312312000000010000001	220000000	31610000	114100000
99	2	1.0	-8.0	512622	312322000000000000001	210000000	310000000	114100000
100	2	1.0	12.0	512622	312322000000000000001	210000000	330000000	114100000
101	2	1.0	-8.0	512623	312322000000000000001	210000000	610000000	114100000
102	2	1.0	24.0	512623	312322000000000000001	210000000	326100000	114100000
103	1	1.0	12.0	512633	312322000000000000001	210000000	315200000	114100000
104	1	1.0	24.0	512634	312322000000000000001	210000000	31516100	114100000
105	2	1.0	-8.0	513623	312322000000000000001	210000000	510000000	114100000
106	2	1.0	24.0	513623	312322000000000000001	210000000	513200000	114100000
107	2	1.0	-8.0	522612	322312000001001000000	510000000	210000000	114100000
108	2	1.0	12.0	522612	322312000001001000000	510000000	230000000	114100000
109	1	1.0	12.0	522613	322312000001000000000	510000000	226100000	114100000
110	2	2.0	-8.0	522622	322322000001011010001	000000000	220000000	114100000
111	2	2.0	8.0	522622	322322000001011010001	000000000	240000000	114100000
112	2	2.0	-8.0	522623	322322000001011010001	000000000	215100000	114100000
113	2	2.0	16.0	522623	322322000001011010000	000000000	236100000	114100000
114	1	2.0	8.0	522633	322322000001000000001	000000000	225200000	114100000
115	1	2.0	16.0	522634	322322000001000000000	000000000	22516100	114100000
116	2	1.0	-8.0	523612	322312000000000000001	510000000	310000000	114100000
117	2	1.0	24.0	523612	322312000000000000001	510000000	223100000	114100000
118	2	1.0	-8.0	523613	322312000000000000001	510000000	210000000	114100000
119	2	1.0	24.0	523613	322312000000000000001	510000000	213200000	114100000
120	1	1.0	24.0	523614	322312000000000000000	510000000	21316100	114100000
121	2	2.0	-8.0	523622	322322000000000000001	000000000	213100000	114100000
122	2	2.0	16.0	523622	322322000000000000001	000000000	233100000	114100000
123	3	2.0	-8.0	523623	322322000000000000001	000000000	220000000	114100000
124	3	2.0	-8.0	523623	322322000000000000001	000000000	320000000	114100000
125	3	2.0	32.0	523623	322322000000000000001	000000000	223200000	114100000
126	2	2.0	-8.0	523624	322322000000000000000	000000000	316100000	114100000
127	2	2.0	32.0	523624	322322000000000000000	000000000	22316100	114100000
128	2	2.0	-8.0	523633	322322000000000000001	000000000	213100000	114100000
129	2	2.0	16.0	523633	322322000000000000001	000000000	213300000	114100000
130	2	2.0	-8.0	523634	322322000000000000001	000000000	216100000	114100000
131	2	2.0	32.0	523634	322322000000000000001	000000000	21326100	114100000
132	1	2.0	16.0	523644	322322000000000000001	000000000	21315200	114100000
133	2	2.0	-8.0	524634	322322000000000000001	000000000	215100000	114100000
134	2	2.0	32.0	524634	322322000000000000001	000000000	21513200	114100000
135	2	2.0	-8.0	533623	322322000000000000001	000000000	512100000	114100000
136	2	2.0	16.0	533623	322322000000000000001	000000000	512300000	114100000
137	2	2.0	-8.0	534623	322322000000000000001	000000000	513100000	114100000
138	2	2.0	32.0	534623	322322000000000000001	000000000	51223100	114100000

Table (4.11.19) Representation of the terms in the expressions of the product of the first partial derivatives of $r_2^{(+s)}$ for the case when $p=4$ and $k^*=4$

1	1	4.0	4.0	511511	322322001001011011001	000000000	240000000	120000000
2	1	4.0	8.0	511512	322322001001011001000	000000000	236100000	120000000
3	1	4.0	4.0	511522	322322001001000000001	000000000	225200000	120000000
4	1	4.0	8.0	511523	322322001001000000000	000000000	225141000	120000000
5	1	4.0	8.0	512511	3223220010000011000001	000000000	233100000	120000000
6	1	4.0	16.0	512512	3223220010000010001000	000000000	223200000	120000000
7	1	4.0	16.0	512513	3223220010000010000000	000000000	223161000	120000000
8	1	4.0	8.0	512522	3223220010000000110001	000000000	213300000	120000000
9	1	4.0	16.0	512523	3223220010000000010000	000000000	213261000	120000000
10	1	4.0	8.0	512533	3223220010000000000001	000000000	213142000	120000000
11	1	4.0	16.0	512534	3223220010000000000000	000000000	21315161	120000000
12	1	4.0	16.0	513523	3223220010000000000000	000000000	215132000	120000000
13	1	4.0	8.0	522512	322322001001001001000	000000000	512300000	120000000
14	1	4.0	16.0	522512	3223220010000001000000	000000000	512231000	120000000
15	1	4.0	4.0	511611	3223220000001011011001	000000000	240000000	114100000
16	1	4.0	8.0	511612	3223220000001010010000	000000000	236100000	114100000
17	1	4.0	4.0	511622	3223220000001010000000	000000000	225200000	114100000
18	1	4.0	8.0	511623	3223220000001000000000	000000000	225161000	114100000
19	1	4.0	8.0	512611	3223220000001100000001	000000000	233100000	114100000
20	1	4.0	16.0	512612	3223220000000010001000	000000000	223200000	114100000
21	1	4.0	16.0	512613	3223220000000010000000	000000000	223161000	114100000
22	1	4.0	8.0	512622	3223220000000000110001	000000000	213300000	114100000
23	1	4.0	16.0	512623	3223220000000000010000	000000000	213261000	114100000
24	1	4.0	8.0	512633	3223220000000000000001	000000000	213152000	114100000
25	1	4.0	16.0	512634	3223220000000000000000	000000000	21315161	114100000
26	1	4.0	16.0	513623	3223220000000000000000	000000000	215132000	114100000
27	1	4.0	8.0	522612	3223220000001001001000	000000000	512300000	114100000
28	1	4.0	16.0	523612	3223220000000000100000	000000000	512231000	114100000

Table (4.11.20) Representation of the terms in the expressions of the second partial derivatives of $r_2^{(+s)}$ for the case when $p=4$ and $k^*=4$

1	3	4.0	2.0	511511	322322001001011011001	000000000	240000000	000000000
2	3	2.0	-8.0	511511	322322001001011011001	000000000	221000000	120000000
3	3	2.0	8.0	511511	322322001001011011001	000000000	240000000	120000000
4	3	4.0	4.0	511512	322322001001011001000	000000000	236100000	000000000
5	3	2.0	-8.0	511512	322322001001011001000	000000000	216100000	120000000
6	3	2.0	16.0	511512	322322001001011001000	000000000	236100000	120000000
7	2	4.0	2.0	511522	322322001001000000001	000000000	225200000	000000000
8	2	2.0	8.0	511522	322322001001000000000	000000000	225200000	120000000
9	2	4.0	4.0	511523	322322001001000000000	000000000	225161000	000000000
10	2	2.0	16.0	511523	322322001001000000000	000000000	225161000	120000000
11	3	4.0	4.0	512511	3223220010000011000001	000000000	233100000	000000000
12	3	2.0	-8.0	512511	3223220010000011000001	000000000	213100000	120000000
13	3	2.0	16.0	512511	3223220010000011000001	000000000	233100000	120000000
14	4	4.0	8.0	512512	3223220010000010000000	000000000	223200000	000000000
15	4	2.0	-8.0	512512	3223220010000010000000	000000000	220000000	120000000
16	4	2.0	-8.0	512512	3223220010000010000000	000000000	320000000	120000000
17	4	2.0	32.0	512512	3223220010000010000000	000000000	223200000	120000000
18	3	4.0	8.0	512513	3223220010000010000000	000000000	223161000	000000000
19	3	2.0	-8.0	512513	3223220010000010000000	000000000	316100000	120000000
20	3	2.0	32.0	512513	3223220010000010000000	000000000	223161000	120000000
21	3	4.0	4.0	512522	3223220010000000110001	000000000	213300000	000000000
22	3	2.0	-8.0	512522	3223220010000000010001	000000000	213100000	120000000
23	3	2.0	16.0	512522	3223220010000000010001	000000000	213300000	120000000
24	3	4.0	8.0	512523	3223220010000000000000	000000000	213261000	000000000
25	3	2.0	-8.0	512523	3223220010000000000000	000000000	216100000	120000000
26	3	2.0	32.0	512523	3223220010000000000000	000000000	213261000	120000000
27	2	4.0	4.0	512533	3223220010000000000000	000000000	213152000	000000000
28	2	2.0	16.0	512533	3223220010000000000000	000000000	213152000	120000000
29	2	4.0	8.0	512534	3223220010000000000000	000000000	21315161	000000000
30	2	2.0	32.0	512534	3223220010000000000000	000000000	21315161	120000000

Table (4.11.20) contd.

31	3	4.0	8.0	513523	32232200100000001000	00000000	21513200	000000000
32	3	2.0	-8.0	513523	32232200100000001000	00000000	21510000	120000000
33	3	2.0	32.0	513523	32232200100000001000	00000000	21513200	120000000
34	3	4.0	4.0	522512	322322001001001001000	00000000	51230000	000000000
35	3	2.0	-8.0	522512	322322001001001001000	00000000	51210000	120000000
36	3	2.0	16.0	522512	322322001001001001000	00000000	51230000	120000000
37	3	4.0	8.0	523512	322322001000001000000	00000000	51223100	000000000
38	3	2.0	-8.0	523512	322322001000001000000	00000000	51310000	120000000
39	3	2.0	32.0	523512	322322001000001000000	00000000	51223100	120000000
40	2	2.0	-8.0	511611	32232200001011011001	00000000	22000000	11410000
41	2	2.0	8.0	511611	32232200001011011001	00000000	24000000	11410000
42	2	2.0	-8.0	511612	32232200000101001000	00000000	21610000	11410000
43	2	2.0	16.0	511612	32232200000101001000	00000000	23610000	11410000
44	1	2.0	8.0	511622	3223220000010000001	00000000	22520000	11410000
45	1	2.0	16.0	511623	3223220000010000000	00000000	22516100	11410000
46	2	2.0	-8.0	512611	32232200000011000001	00000000	21310000	11410000
47	2	2.0	16.0	512611	32232200000011000001	00000000	25310000	11410000
48	3	2.0	-9.0	512612	32232200000010001000	00000000	22000000	11410000
49	3	2.0	-8.0	512612	32232200000010001000	00000000	32000000	11410000
50	3	2.0	32.0	512612	32232200000010001000	00000000	22320000	11410000
51	2	2.0	-8.0	512613	32232200000010000000	00000000	31610000	11410000
52	2	2.0	32.0	512613	32232200000010000000	00000000	22316100	11410000
53	2	2.0	-8.0	512622	32232200000000011001	00000000	21310000	11410000
54	2	2.0	16.0	512622	32232200000000011001	00000000	21330000	11410000
55	2	2.0	-8.0	512623	32232200000000010000	00000000	21610000	11410000
56	2	2.0	32.0	512623	32232200000000010000	00000000	21326100	11410000
57	1	2.0	16.0	512633	32232200000000000001	00000000	21315200	11410000
58	1	2.0	32.0	512634	32232200000000000000	00000000	21315161	11410000
59	2	2.0	-8.0	513623	32232200000000010000	00000000	21510000	11410000
60	2	2.0	32.0	513623	32232200000000010000	00000000	21513200	11410000
61	2	2.0	-8.0	522612	3223220000001001001000	00000000	51210000	11410000
62	2	2.0	16.0	522612	3223220000001001001000	00000000	51230000	11410000
63	2	2.0	-8.0	523612	32232200000010000000	00000000	51310000	11410000
64	2	2.0	32.0	523612	32232200000010000000	00000000	51223100	11410000

Section 4.12 Programs for deriving $\beta_1(\theta_A, \sigma_A)$

The programs for deriving $\beta_1(\theta_A, \sigma_A)$ are program POWCAL and subroutines COEF11, COEFL, COEF2, E1000, E2000, etc. as shown in Appendix 4. Note that these programs can also be used to derive $\beta_2(\theta_A, \sigma_A)$.

Section 4.13 Numerical examples

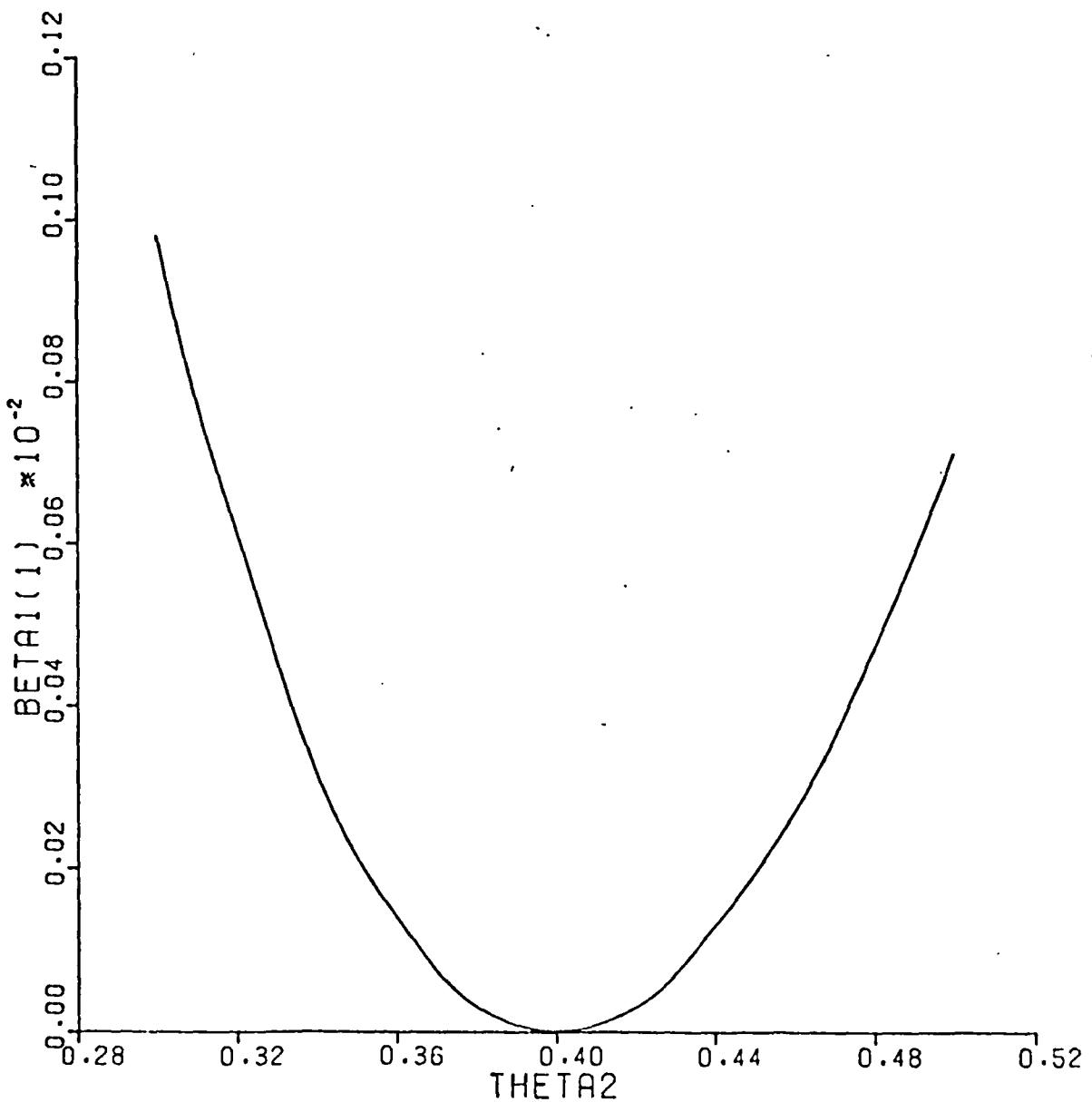
In this section, we make use of the computer programs in section 4.12 to evaluate $\beta_i^{(1)}$ and $\beta_i^{(2)}$ ($i = 1, 2$) in models (A) and (B) described in Chapter 1. For all the corresponding hypotheses in these models, we choose θ_0 to be $(1.4, 0.4)^T$. For the case when the hypothesis is concerned with $\theta_1 = 1.4$, we consider θ_A of the form $(1.4, \theta_{A2})^T$, and for the case when the hypothesis is concerned with $\theta_2 = 0.4$, we consider θ_A of the form $(\theta_{A1}, 0.4)^T$. If the hypothesis is concerned with $\theta = (1.4, 0.4)^T$, we consider θ_A of the forms $(1.4, \theta_{A2})^T$ and $(\theta_{A1}, 0.4)^T$. In Fig. (4.13.1)-(4.13.16), the variation of $\beta_i^{(1)}$ and $\beta_i^{(2)}$ ($i = 1, 2$) with respect to θ_A is shown.

FIGURE(4.13.1)

BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA 1
WHEN SIGMA IS KNOWN
MODEL IS

$$E(Y) = (\text{THETA}1 / (\text{THETA}1 - \text{THETA}2)) \\ * (\exp(-\text{THETA}2 * X) - \exp(-\text{THETA}1 * X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0 \\ \text{THETA}1 \text{ ZERO ARE } 1.4000, 0.4000$$



FIGURE(4.13.2)

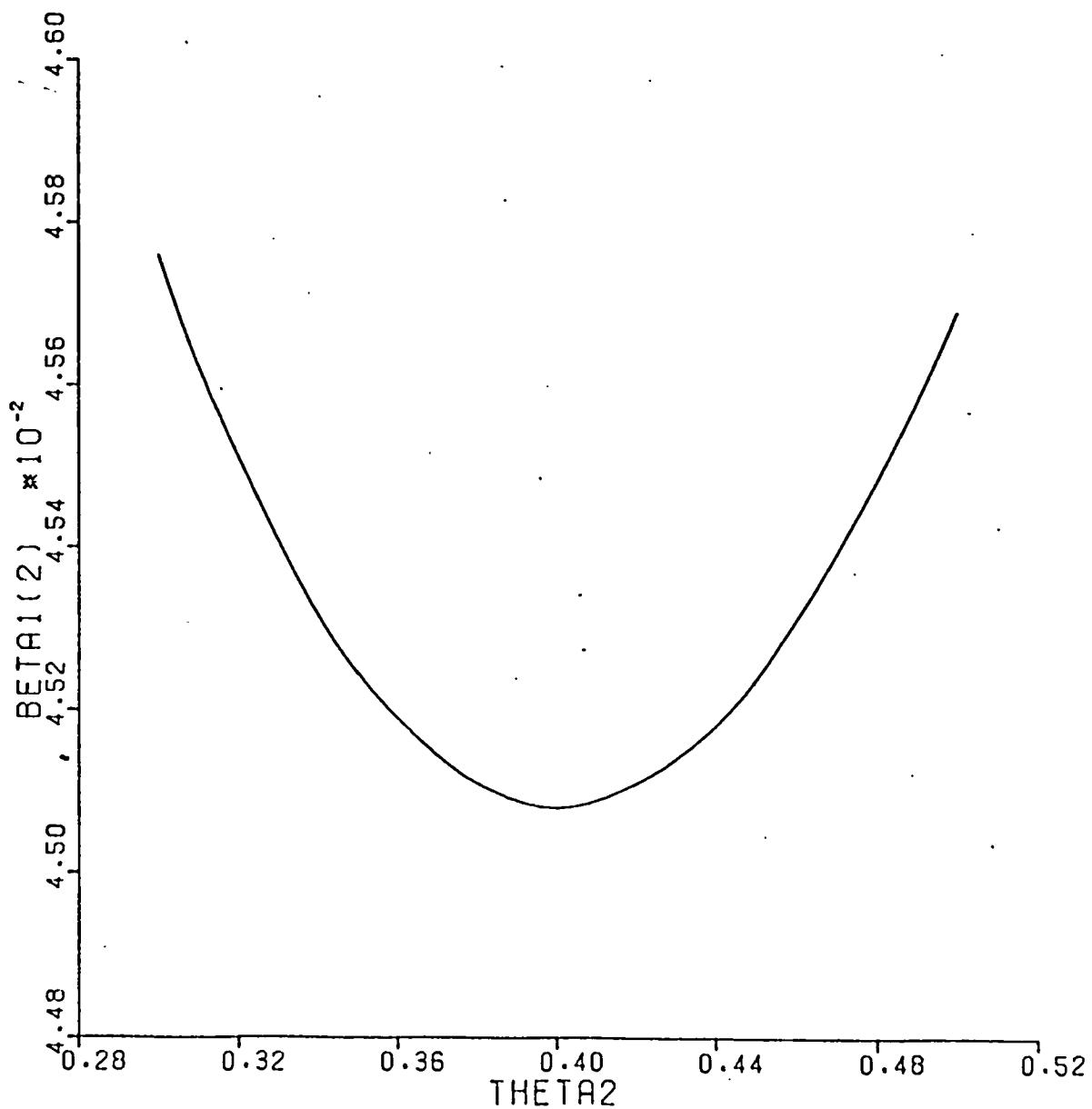
BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA 1
WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\text{THETA}1 / (\text{THETA}1 - \text{THETA}2)) \\ * (\exp(-\text{THETA}2 \cdot X) - \exp(-\text{THETA}1 \cdot X))$$

XI= 0.25.0.5.1.0.1.5.2.0.4.0

THETA1 ZERO ARE 1.4000 0.4000



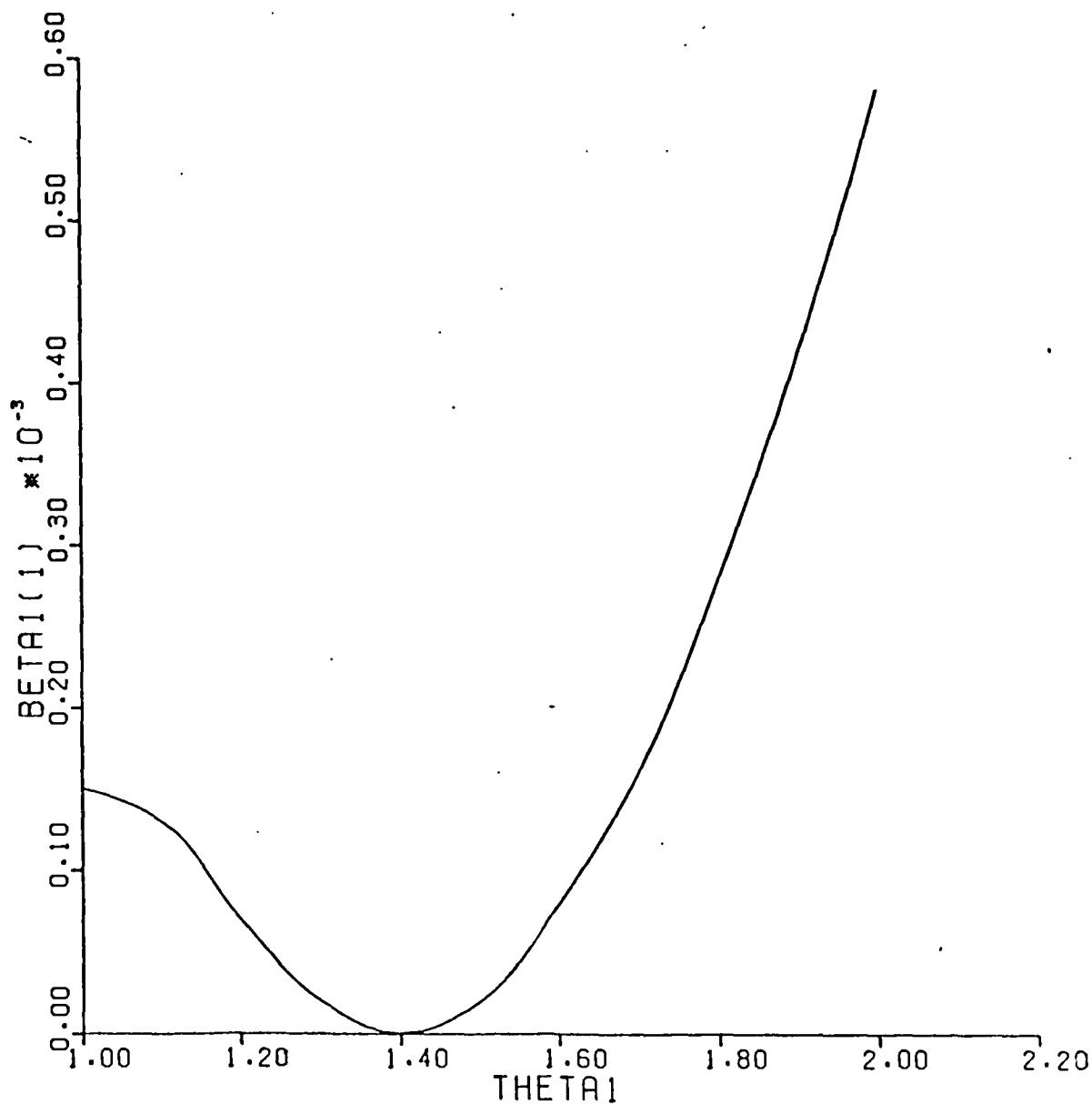
FIGURE(4.13.3)

BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA 2
WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \\ \times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0 \\ \Theta_1 \text{ ZERO ARE } 1.4000, 0.4000$$



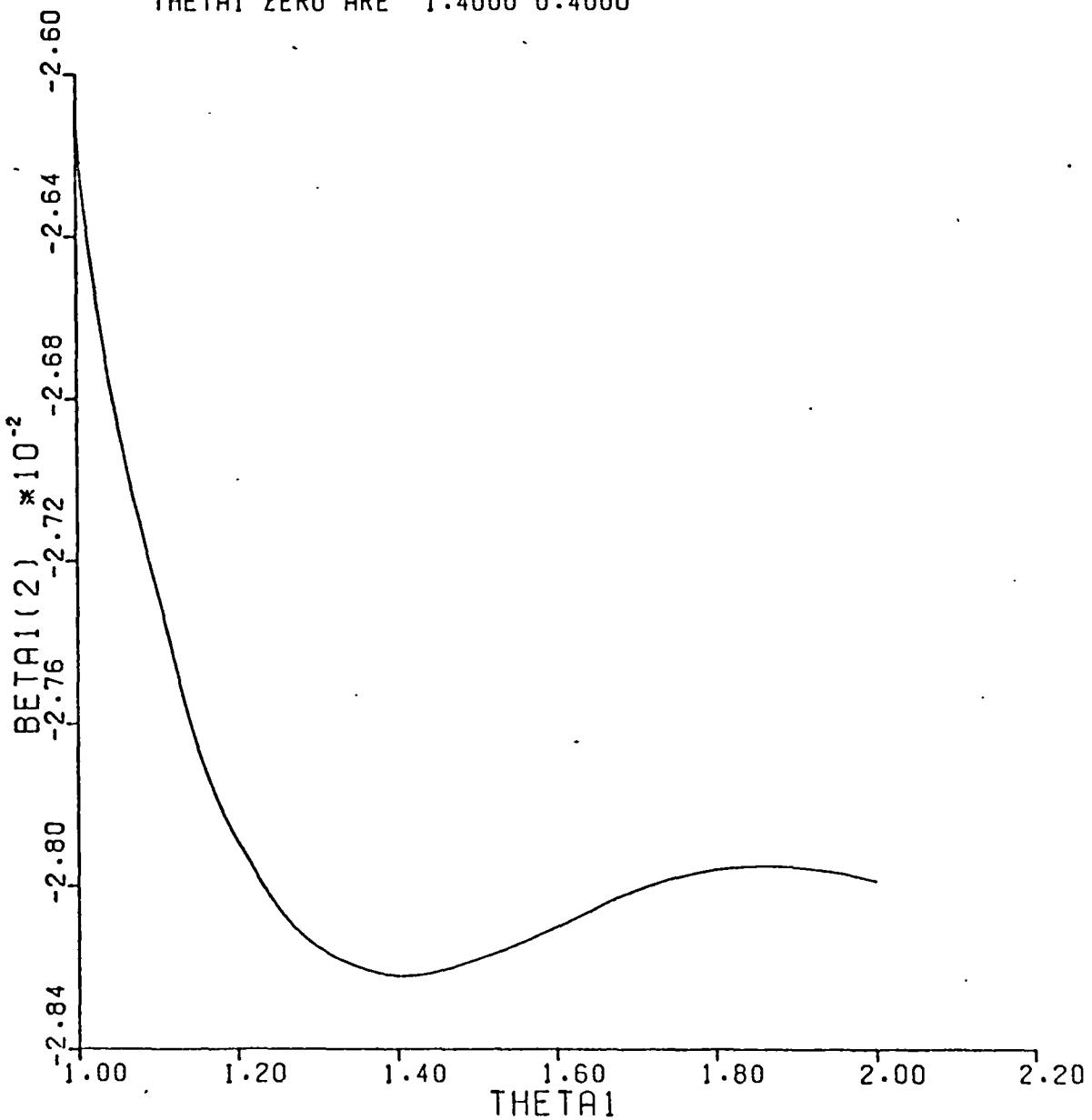
FIGURE(4.13.4)

BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA 2 WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\text{THETA}1 / (\text{THETA}1 - \text{THETA}2)) \cdot (\exp(-\text{THETA}2 \cdot X) - \exp(-\text{THETA}1 \cdot X))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0 \\ \text{THETA}1 \text{ ZERO ARE } 1.4000, 0.4000$$



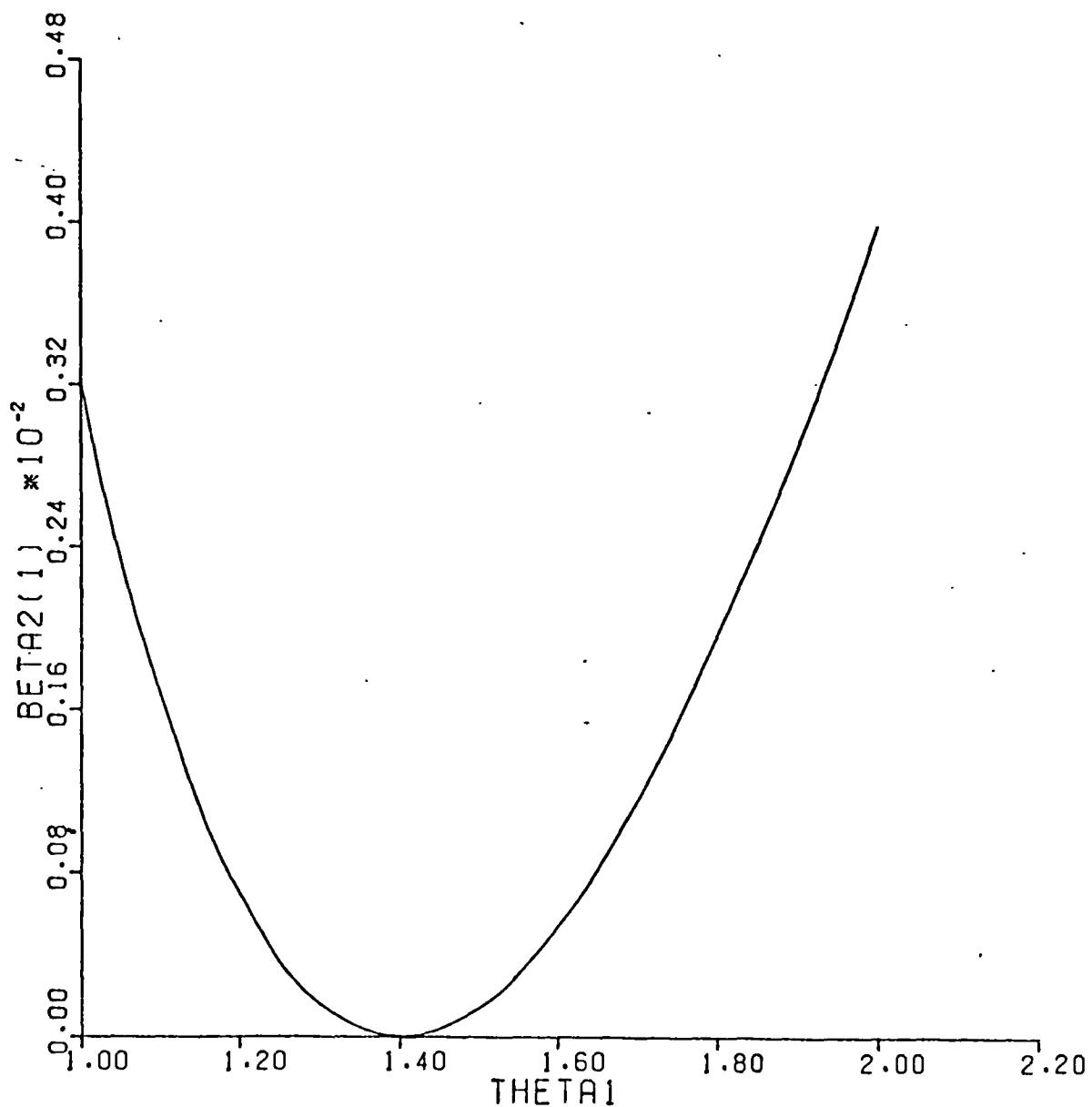
FIGURE(4.13.5)

BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA
WHEN SIGMA IS KNOWN
MODEL IS

$$E(Y) = (\text{THETA}1 / (\text{THETA}1 - \text{THETA}2)) \\ * (\exp(-\text{THETA}2 * X) - \exp(-\text{THETA}1 * X))$$

XI= 0.25 0.5 1.0 1.5 2.0 4.0

THETA1 ZERO ARE 1.4000 0.4000



FIGURE(4.13.5)

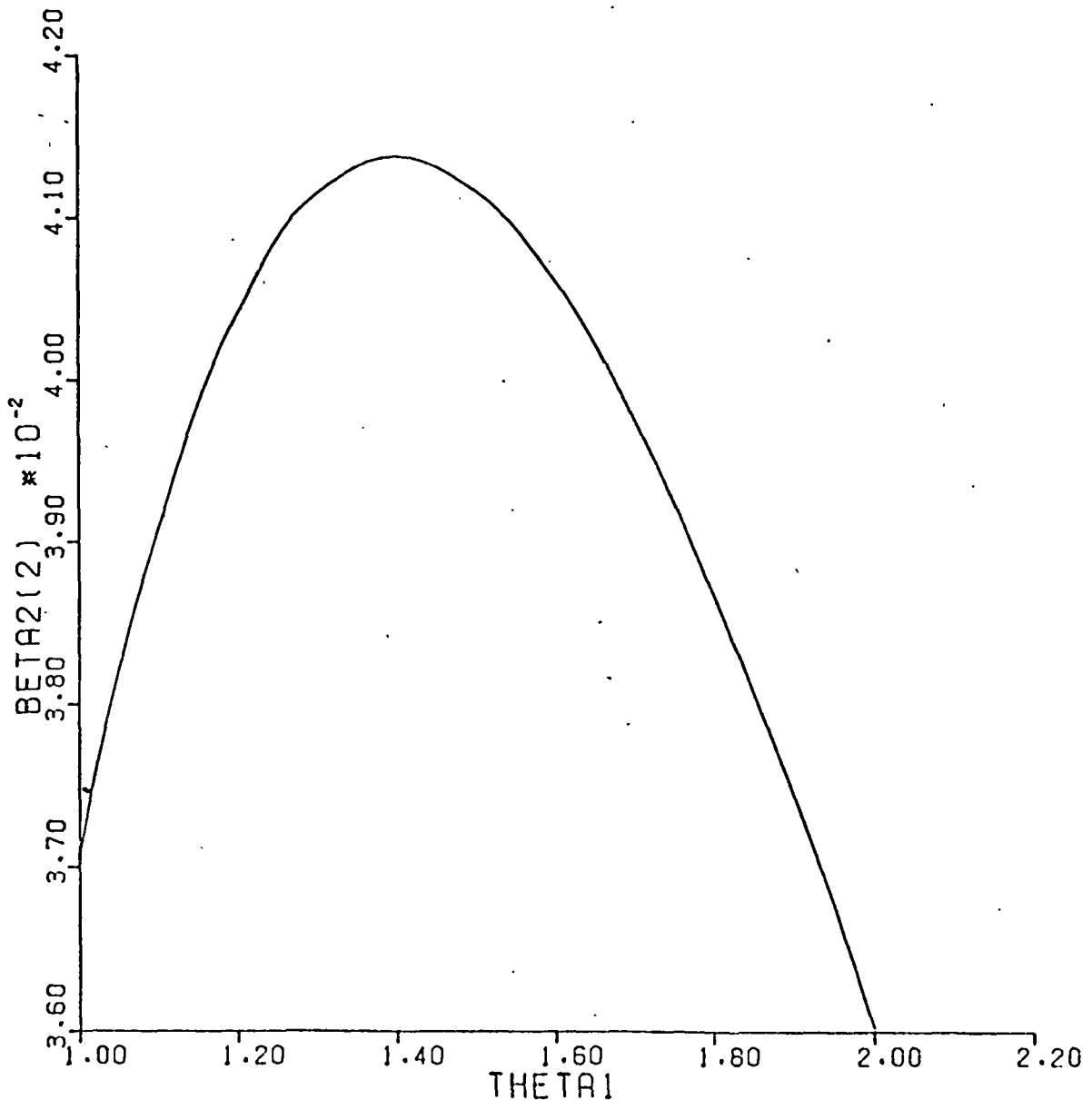
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\text{THETA}1 / (\text{THETA}1 - \text{THETA}2)) * (\exp(-\text{THETA}2 * X) - \exp(-\text{THETA}1 * X))$$

XI = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0

THETA1 ZERO ARE 1.4000 0.4000



FIGURE(4.13.7)

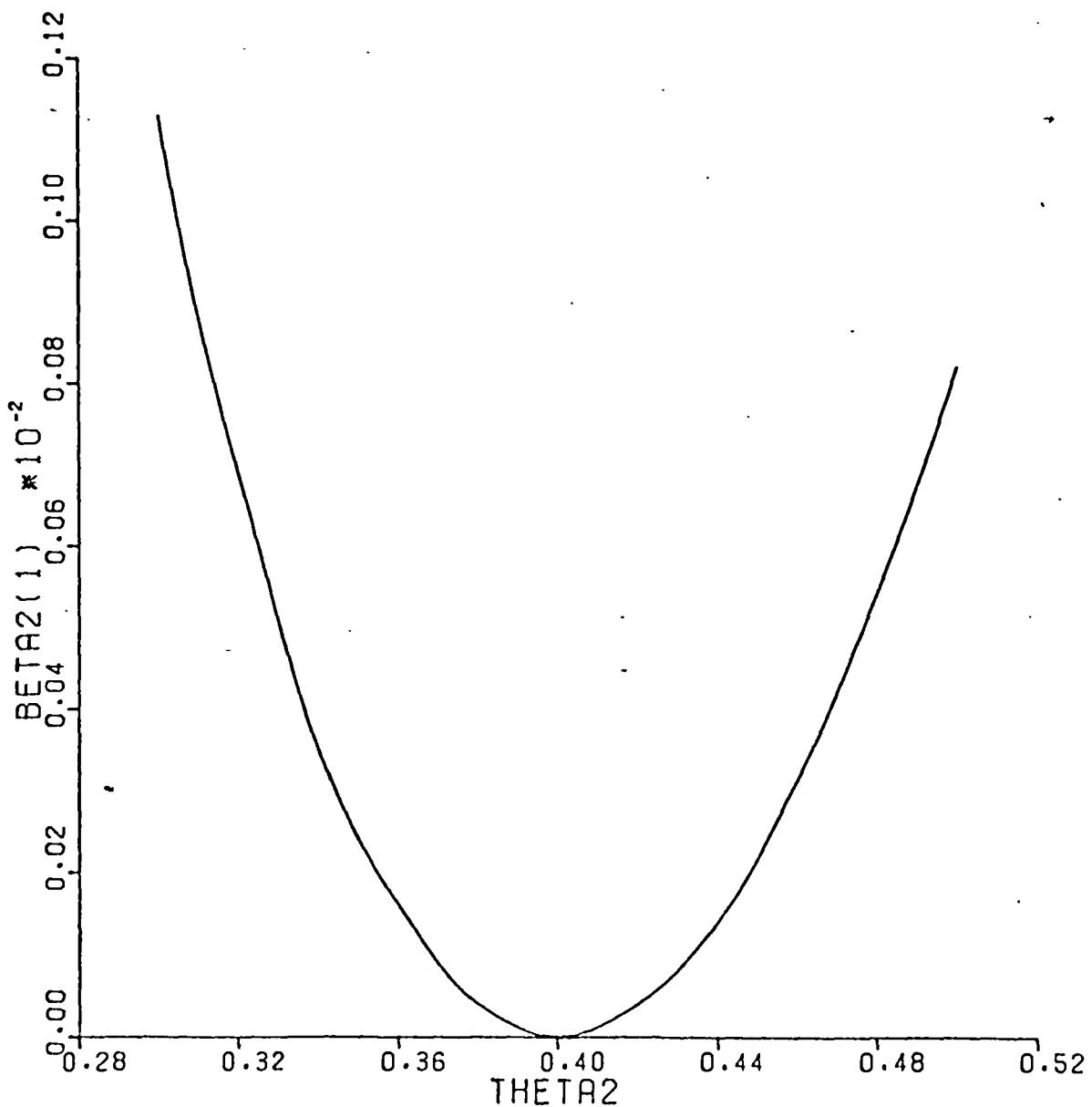
BETA1(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \cdot (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

X1 = 0.25 0.5 1.0 1.5 2.0 4.0

THETA1 ZERO ARE 1.4000 0.4000



FIGURE(4.13.8)

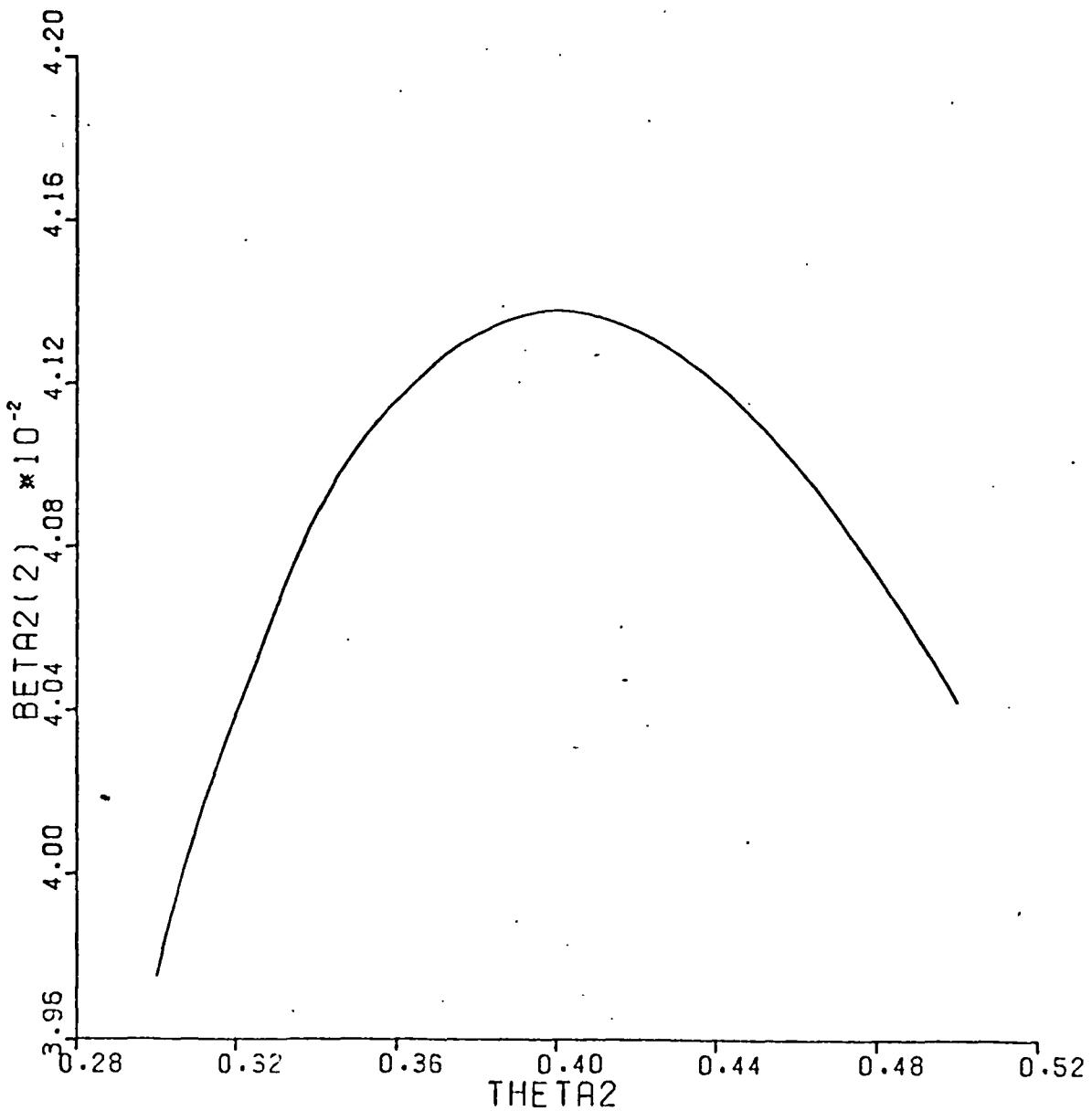
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) * (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0

THETA1 ZERO ARE 1.4000 0.4000



FIGURE(4.13.9)

BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA 1 WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \exp(-\text{THETA}2 \cdot X)) \\ - (\text{THETA}2 \cdot \exp(-\text{THETA}1 \cdot X)) \\ / (\text{THETA}1 - \text{THETA}2)$$

XI= 1.2.3.4.5.6

THETA1 ZERO ARE 1.4000 0.4000

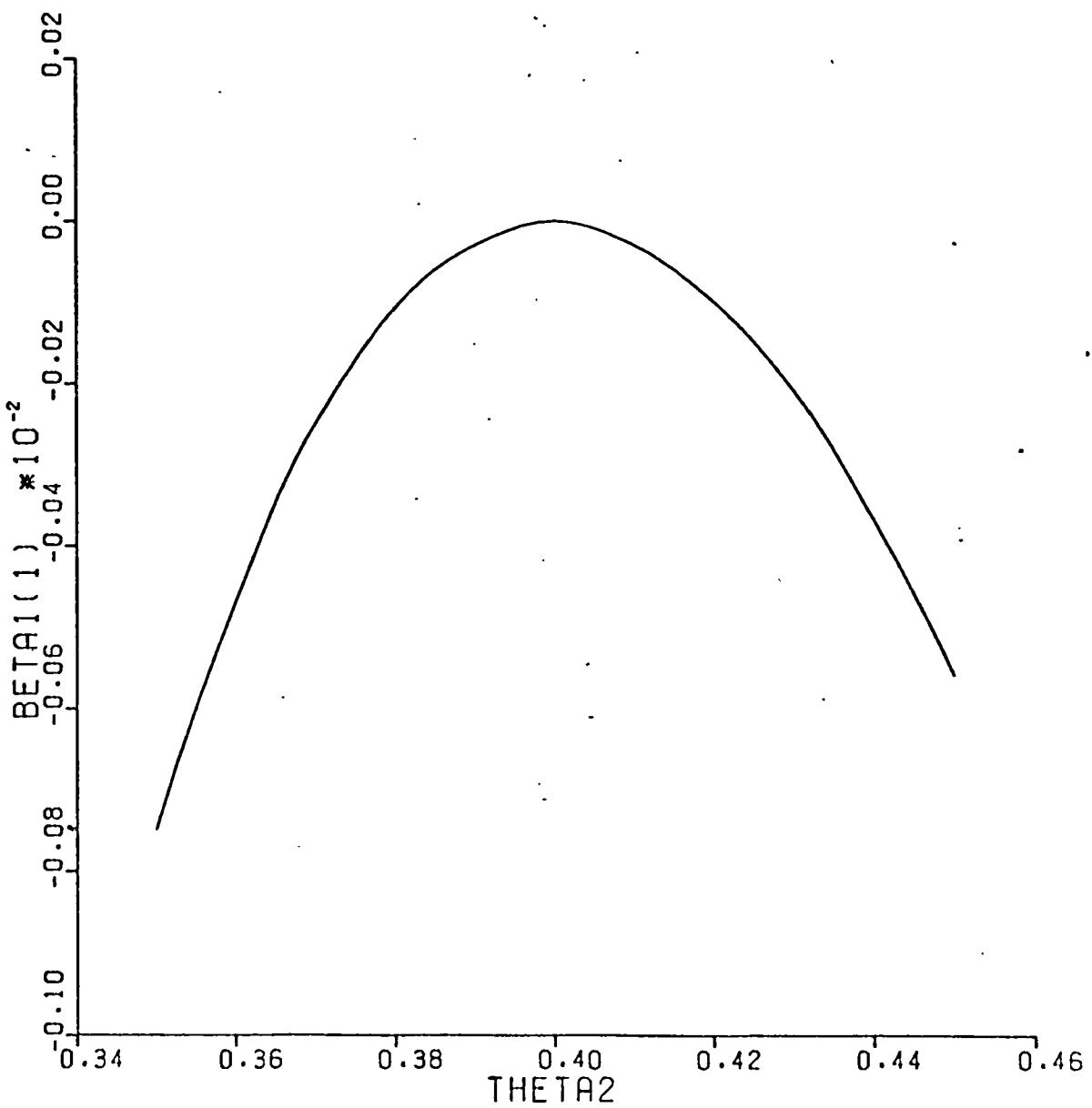


FIGURE (4.13.10)

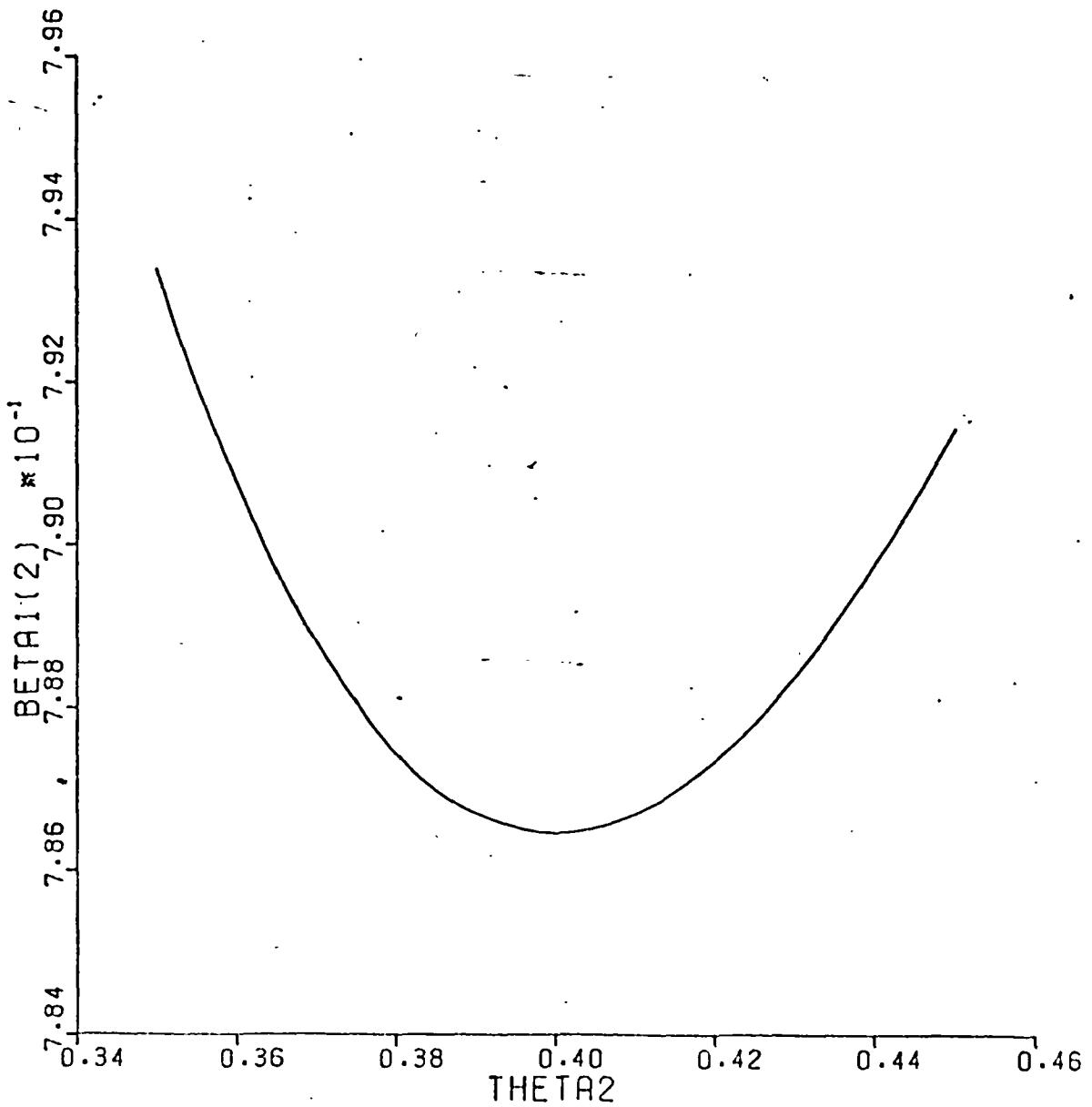
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA 1 WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \text{EXP}(-\text{THETA}2 \cdot \text{XI}) - \text{THETA}2 \cdot \text{EXP}(-\text{THETA}1 \cdot \text{XI})) / (\text{THETA}1 - \text{THETA}2)$$

XI = 1, 2, 3, 4, 5, 6

THETA1 ZERO ARE 1.4000 0.4000



FIGURE(4.13.11)
BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA 2
WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - \Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

X_I = 1.2, 3, 4, 5, 6

THETA₁ ZERO ARE 1.4000 0.4000

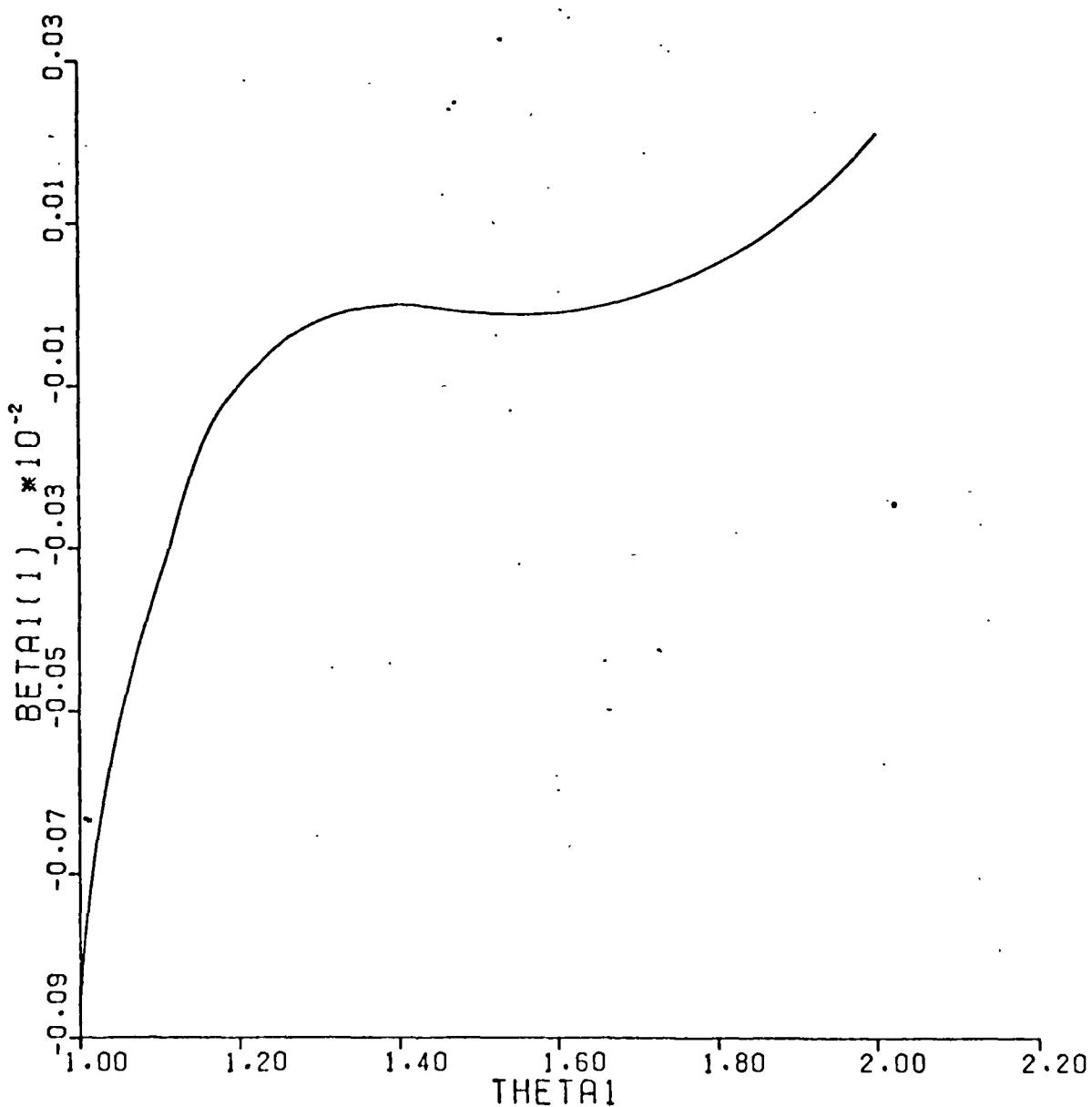


FIGURE (4.13.12)

BETAI(J) IN THE SERIES EXPANSION OF THE
POWER FUNCTION OF THE GENERAL MAXIMUM
LIKELIHOOD RATIO TEST CONCERNING THETA 2
WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \text{EXP}(-\text{THETA}2 \cdot X)) \\ - \text{THETA}2 \cdot \text{EXP}(-\text{THETA}1 \cdot X)) \\ / (\text{THETA}1 - \text{THETA}2)$$

XI = 1.2.3.4.5.6

THETA1 ZERO ARE 1.4000 0.4000

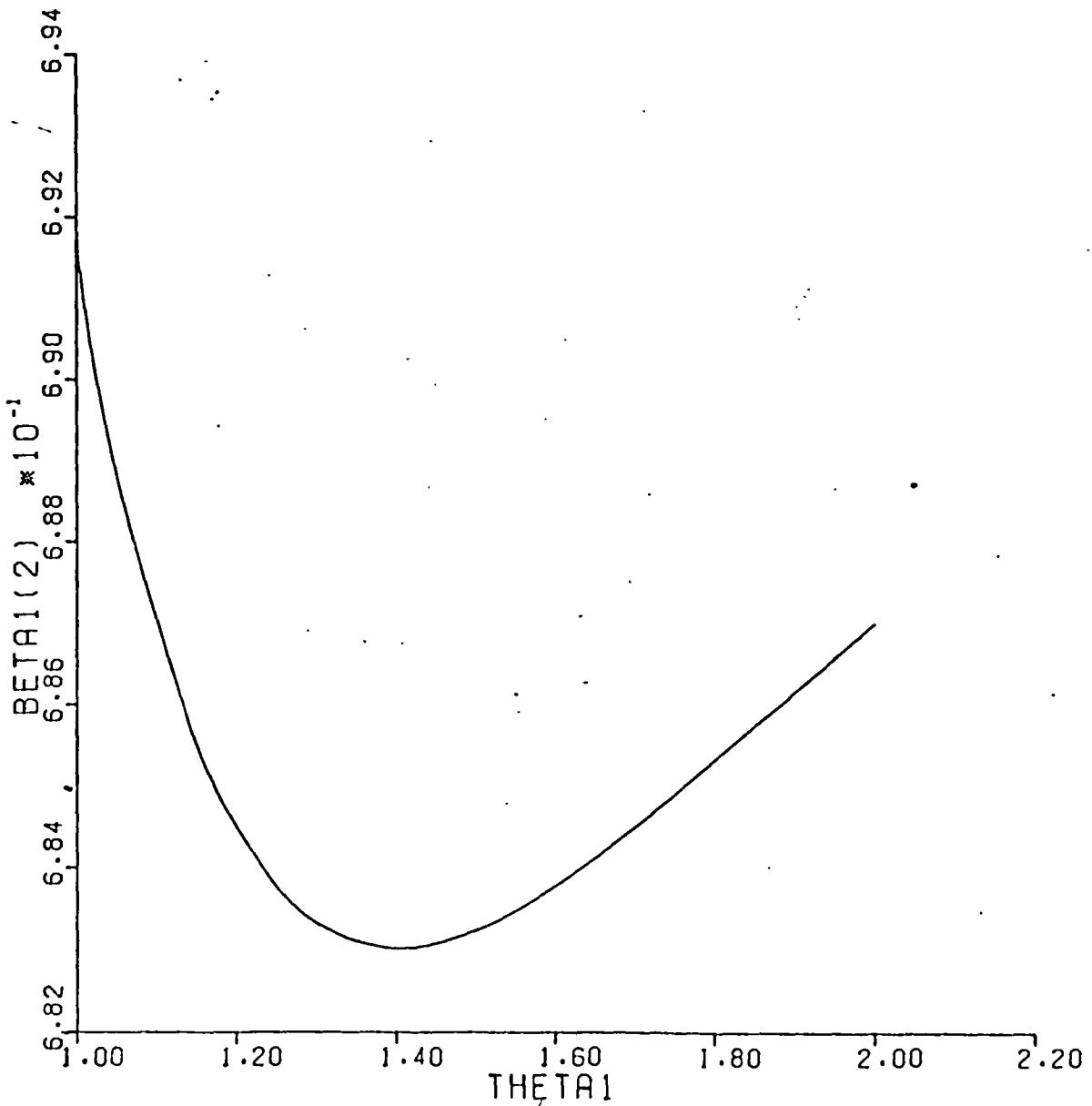


FIGURE (4.13.13)

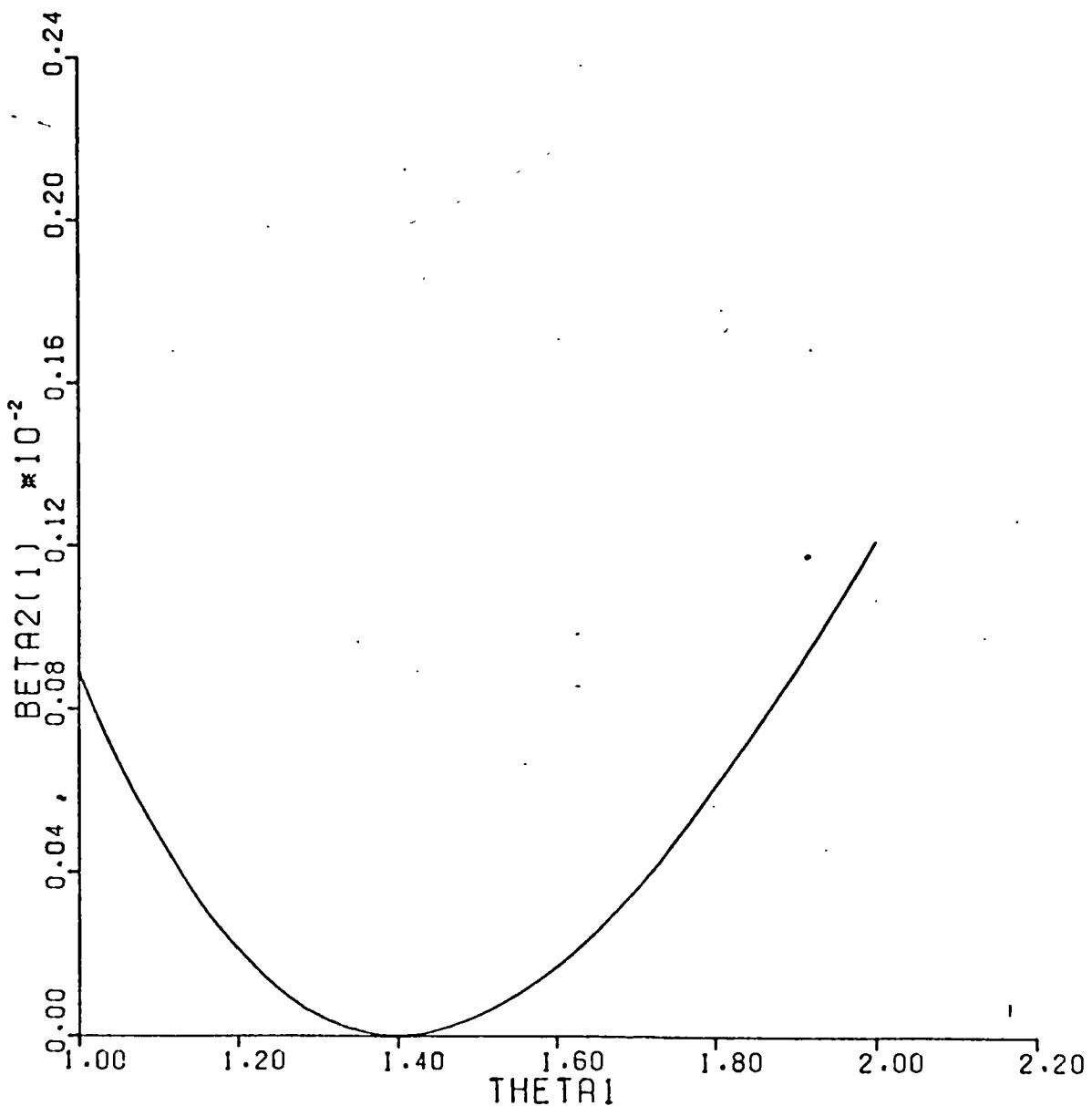
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \exp(-\text{THETA}2 \cdot X)) \\ - (\text{THETA}2 \cdot \exp(-\text{THETA}1 \cdot X)) \\ / (\text{THETA}1 - \text{THETA}2)$$

X_I = 1, 2, 3, 4, 5, 6

THETA₁ ZERO ARE 1.4000 0.4000



FIGURE(4.13.14)

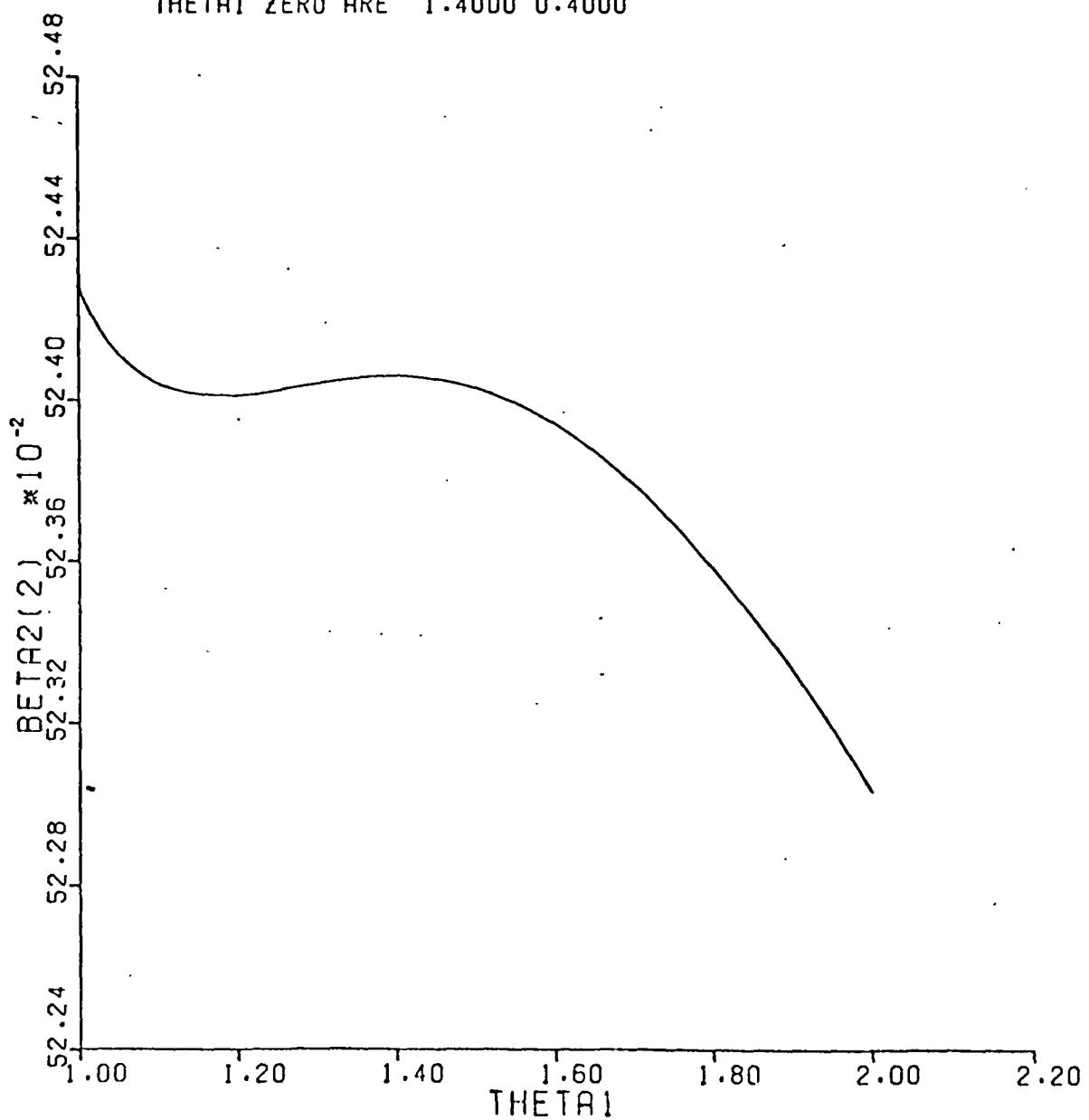
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \exp(-\text{THETA}2 \cdot X)) \\ - (\text{THETA}2 \cdot \exp(-\text{THETA}1 \cdot X)) \\ / (\text{THETA}1 - \text{THETA}2)$$

X_I = 1.2.3.4.5.6

THETA₁ ZERO ARE 1.4000 0.4000



FIGURE(4.13.15)

BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \cdot \exp(-\text{THETA}2 \cdot X)) \\ - \text{THETA}2 \cdot \exp(-\text{THETA}1 \cdot X) \\ / (\text{THETA}1 - \text{THETA}2)$$

XI= 1.2.3.4.5.6

THETAI ZERO ARE 1.4000 0.4000

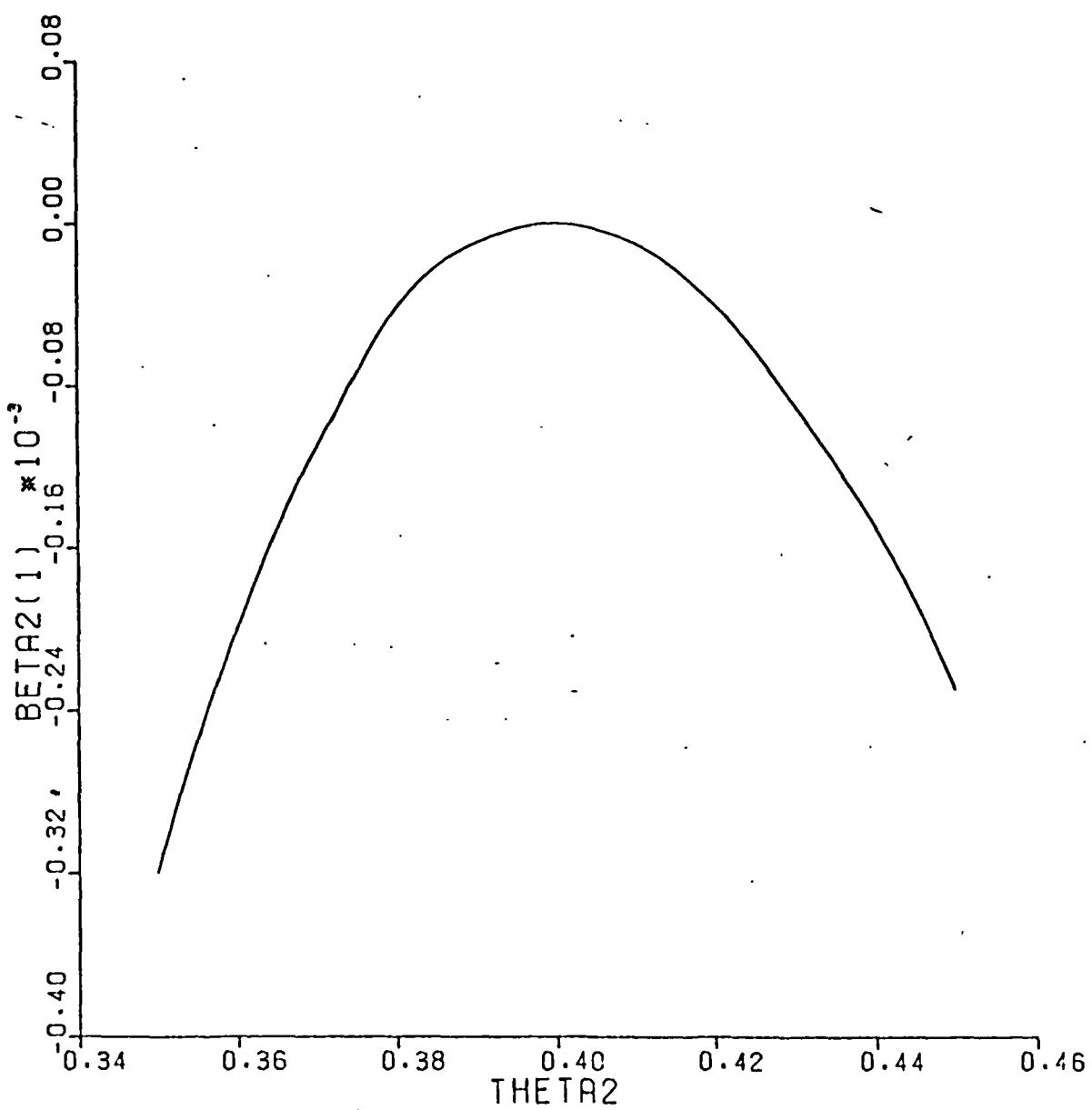


FIGURE (4.13.16)

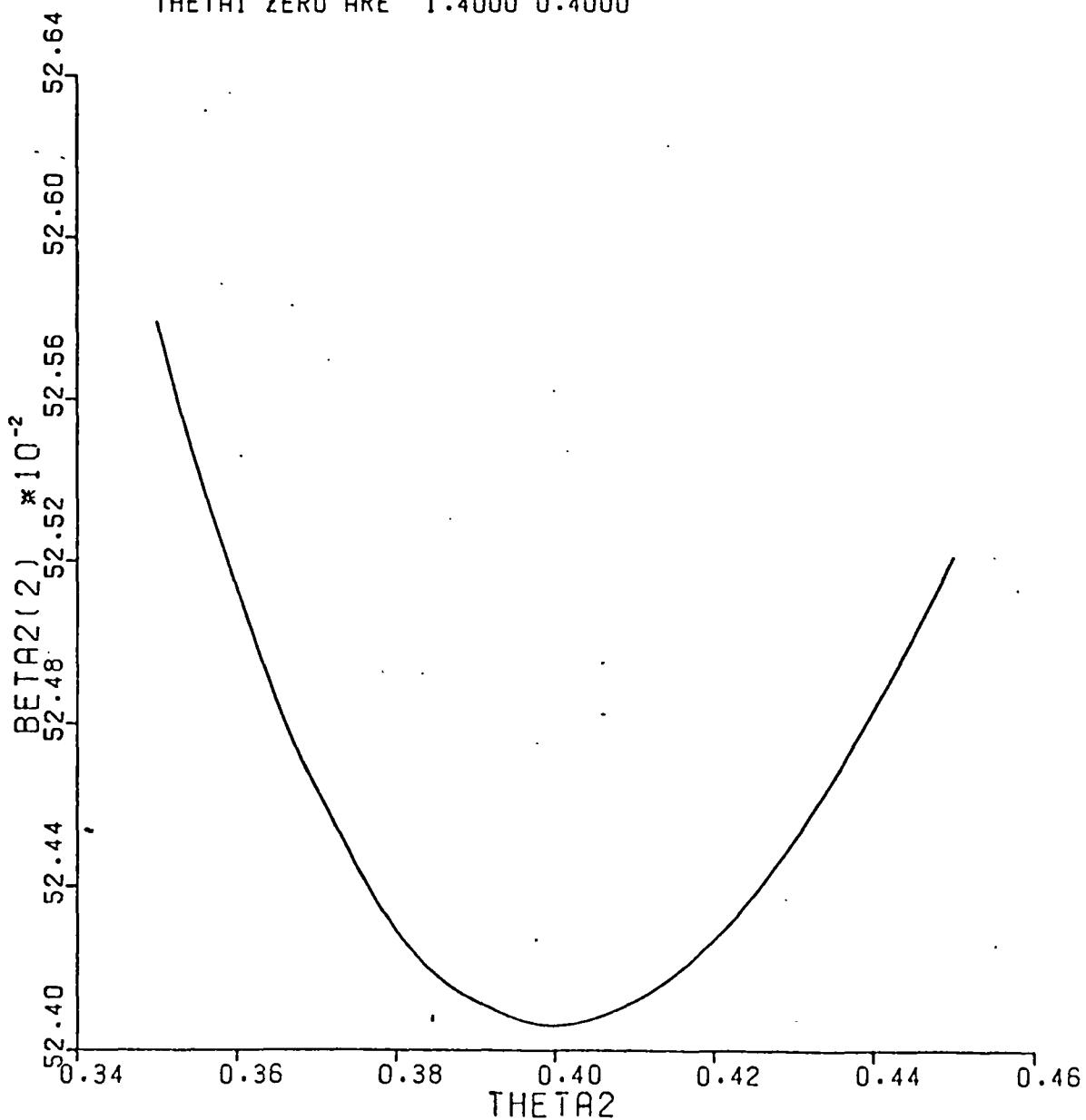
BETAI(J) IN THE SERIES EXPANSION OF THE POWER FUNCTION OF THE GENERAL MAXIMUM LIKELIHOOD RATIO TEST CONCERNING THETA WHEN SIGMA IS KNOWN

MODEL IS

$$E(Y) = 1 - (\text{THETA}1 \times \exp(-\text{THETA}2 \times \xi)) \\ - (\text{THETA}2 \times \exp(-\text{THETA}1 \times \xi)) \\ / (\text{THETA}1 - \text{THETA}2)$$

$\xi = 1, 2, 3, 4, 5, 6$

THETA1 ZERO ARE 1.4000 0.4000



CHAPTER 5

COMPARISON OF VARIOUS METHODS OF OBTAINING REGION ESTIMATES BY MEANS OF NUMERICAL EXAMPLES

Section 5.1 Introduction

In Chapters 2 and 3, various methods of obtaining region estimates for a subset of k^* ($1 \leq k^* \leq p$) components of the parameter vector θ have been described. We refer to these methods as methods 1-4 as follows:

Method 1 is based on the approximations that the model is linear in the original parameter vector θ .

Method 2 is based on power transformations of all the individual parameters (c.f. section 2.7).

Method 3 is based on general transformations in which the transformed parameters of interest depend only on the original parameters of interest, but the remaining transformed parameters may depend on all the original parameters (c.f. section 2.4).

Method 4 is based on general maximum likelihood ratios (c.f. Chapter 3).

In this chapter we restrict our attention to the case when σ is known, and apply the four methods to obtain region estimates in models (A) and (B) described in Chapter 1. These regions are compared in two aspects:

- (i) the estimation of the coverage probability and the values of nonlinearity associated with the regions,
- (ii) the boundaries of these regions.

In comparing (ii) we see how we can suggest bounds for the values of nonlinearity within which the use of linear theory to obtain the corresponding region estimates is justifiable.

Section 5.2 Estimation of coverage probability and nonlinearity

The coverage probability of the region estimate given by method 4 for k^* ($1 \leq k^* \leq p$) components of the parameter vector θ is an important quantity associated with this region. As θ_T is not known, we usually do not know the value of the coverage probability. In this section, we shall investigate the feasibility of estimating the actual coverage probability by using the coverage probability evaluated at $\theta = \hat{\theta}$.

We choose $\theta_T = (1.4, 0.4)^T$. We then set $\alpha = 0.05$ and use (3.3.50) and (3.3.52) to calculate the coverage probabilities $I_1(\theta, \sigma)$ and $I_2(\theta, \sigma)$ for various values of θ . In Fig. (5.2.1)-(5.2.6) we display the absolute values of

$$[I_i(\theta, \sigma) - I_i(\theta_T, \sigma)]/[I_i(\theta_T, \sigma) - 0.95]$$

for $i = 1, 2$. These absolute values are classified into five categories each of which is represented by a symbol (c.f. footnotes of the figures).

FIGURE (5.2.1)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

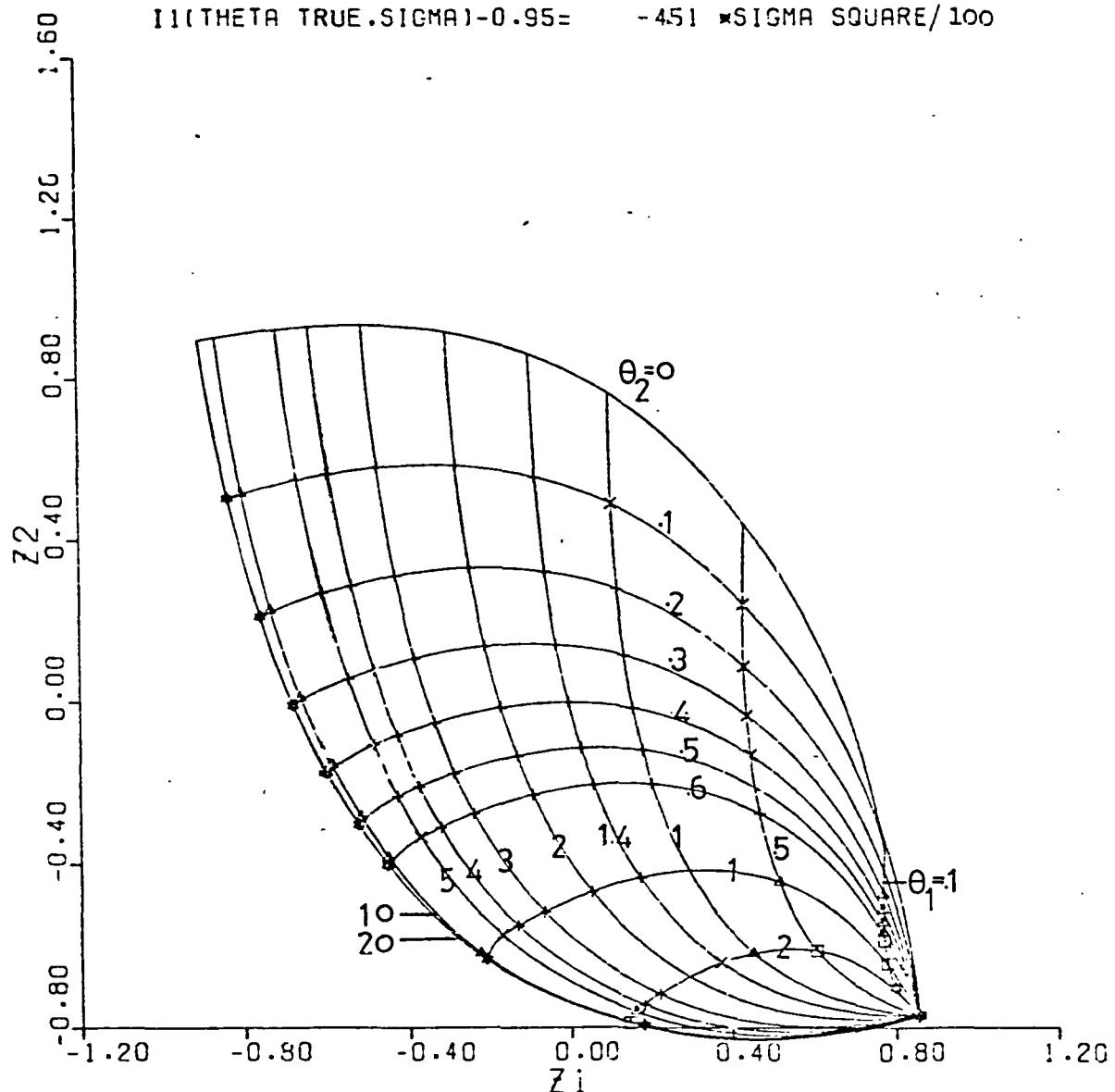
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$III(\Theta_1 \text{ TRUE}, \Sigma) - 0.95 = -451 \times \Sigma^2 / 100$$



$R = \text{ABSOLUTE VALUE OF } III(\Theta_1 \text{ TRUE}, \Sigma) - III(\Theta_1 \text{ TRUE}, \Sigma) / III(\Theta_1 \text{ TRUE}, \Sigma) - 0.95$
 PARAMETER OF INTEREST IS Θ_1

+ : $0 \leq R \leq 0.5$; X : $0.5 < R \leq 1$; △ : $1 < R \leq 10$; □ : $10 < R \leq 100$; ♦ : $R > 100$

FIGURE (5.2.2)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

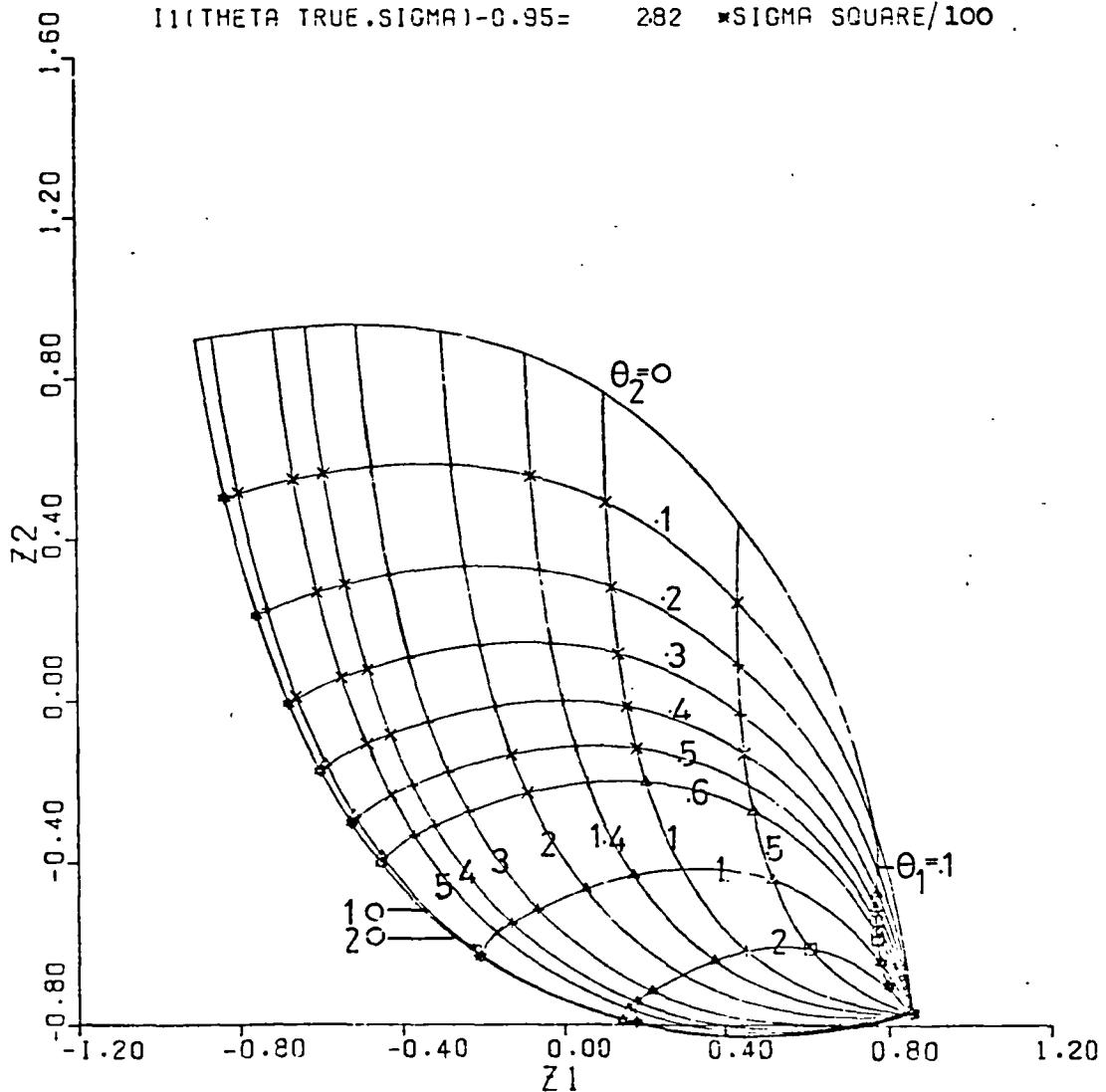
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$|I_1(\Theta_1 \text{ TRUE}, \Sigma)| - 0.95 = \quad 282 \quad \Sigma \text{ SQUARE/100}$$



R=ABSOLUTE VALUE OF ((|I_1(\Theta_1, \Sigma)| - |I_1(\Theta_1 \text{ TRUE}, \Sigma)|) / (|I_1(\Theta_1 \text{ TRUE}, \Sigma)| - 0.95))
 PARAMETER OF INTEREST IS $\Theta_1 \text{ AND } \Theta_2$

+ : 0 < R < 0.5 : X : 0.5 < R < 1 : △ : 1 < R < 10 : □ : 10 < R < 100 : * : R > 100

FIGURE (5.2.3)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

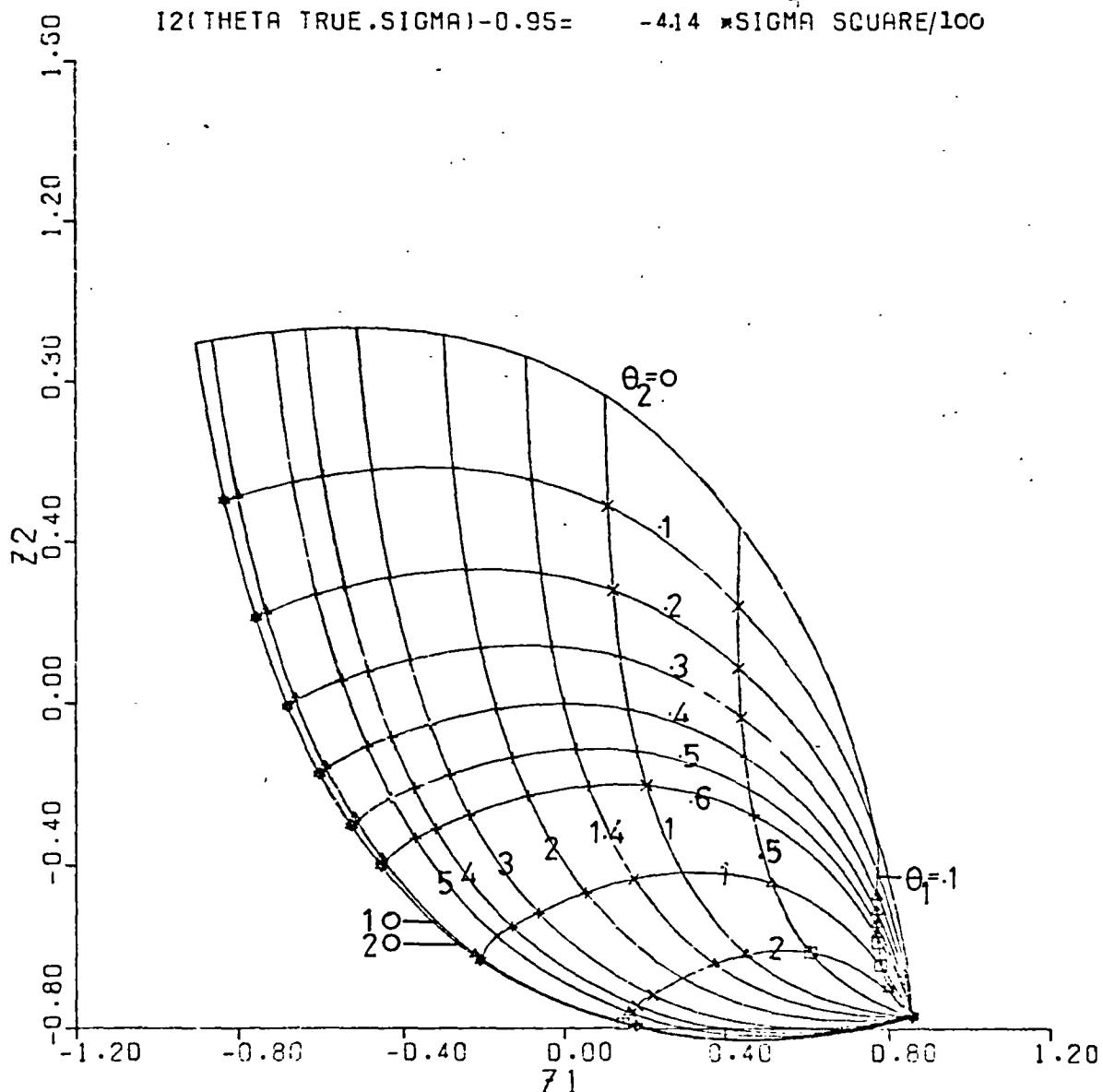
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$I_2(\Theta \text{ TRUE}, \Sigma) - 0.95 = -4.14 \text{ } \Sigma^2 / 100$$



$$R = \text{ABSOLUTE VALUE OF } (I_2(\Theta, \Sigma) - I_2(\Theta \text{ TRUE}, \Sigma)) / (I_2(\Theta \text{ TRUE}, \Sigma) - 0.95)$$

$$+ : 0 \leq R \leq 0.5 \quad : X : 0.5 < R \leq 1 \quad : \Delta : 1 < R \leq 10 \quad : \square : 10 < R \leq 100 \quad : \star : R > 100$$

FIGURE (5.2.4)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

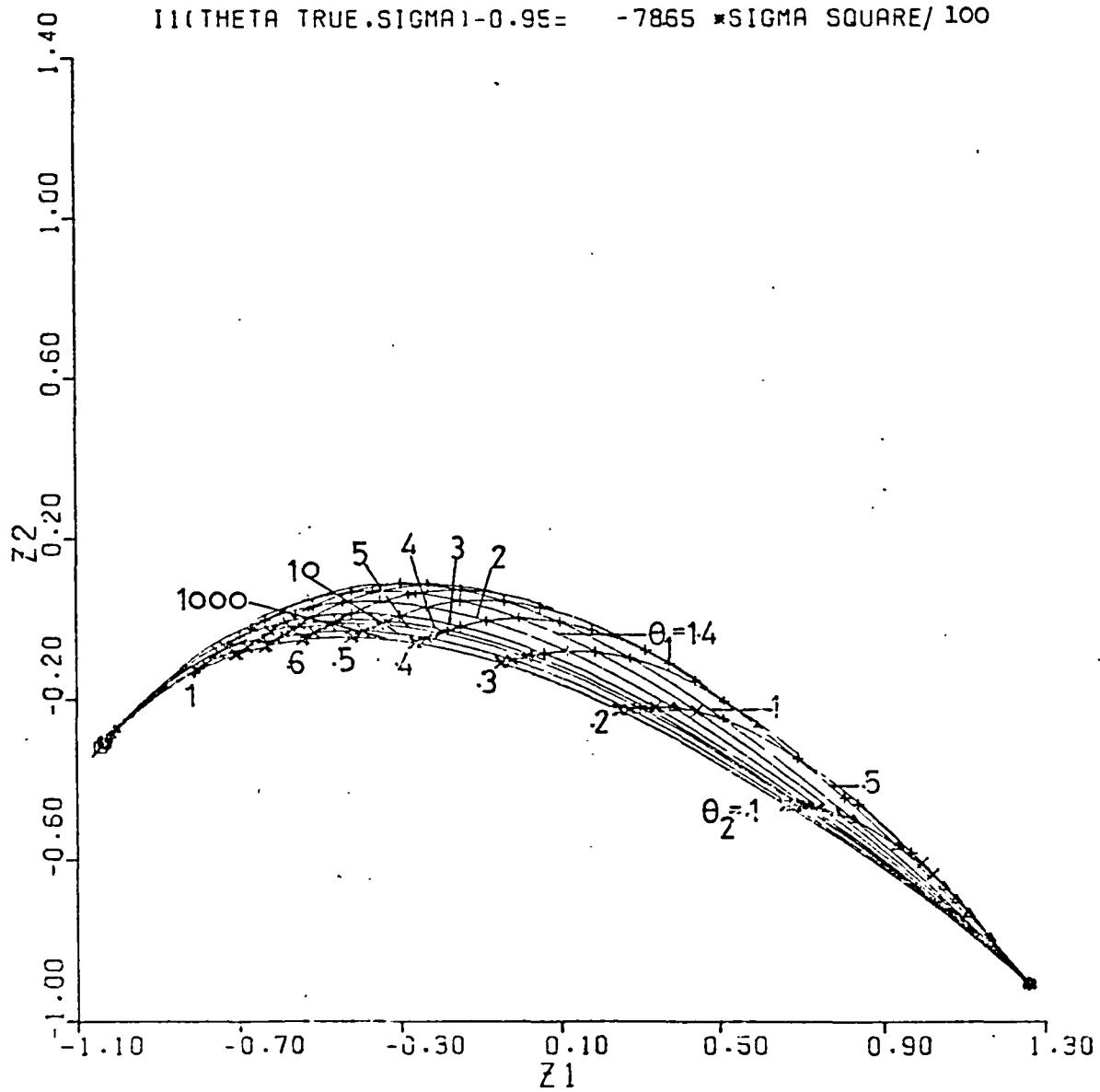
MODEL IS

$$E(Y) = 1 - (\Theta_1 \exp(-\Theta_2 \cdot X_1) - \Theta_2 \exp(-\Theta_1 \cdot X_1)) / (\Theta_1 - \Theta_2)$$

 $X_1 = 1.2 \cdot 3.4.5.6$

THETA1 TRUE ARE 1.4000 0.4000

II(THETA TRUE, SIGMA) - 0.95 = -7865 * SIGMA SQUARE/ 100



R=ABSOLUTE VALUE OF (II(THETA, SIGMA) - II(THETA TRUE, SIGMA)) / II(THETA TRUE, SIGMA) - 0.95
 PARAMETER OF INTEREST IS THE THETA1

+ : 0 ≤ R ≤ 0.5 ; X : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ★ : R > 100

FIGURE (5.2.5)

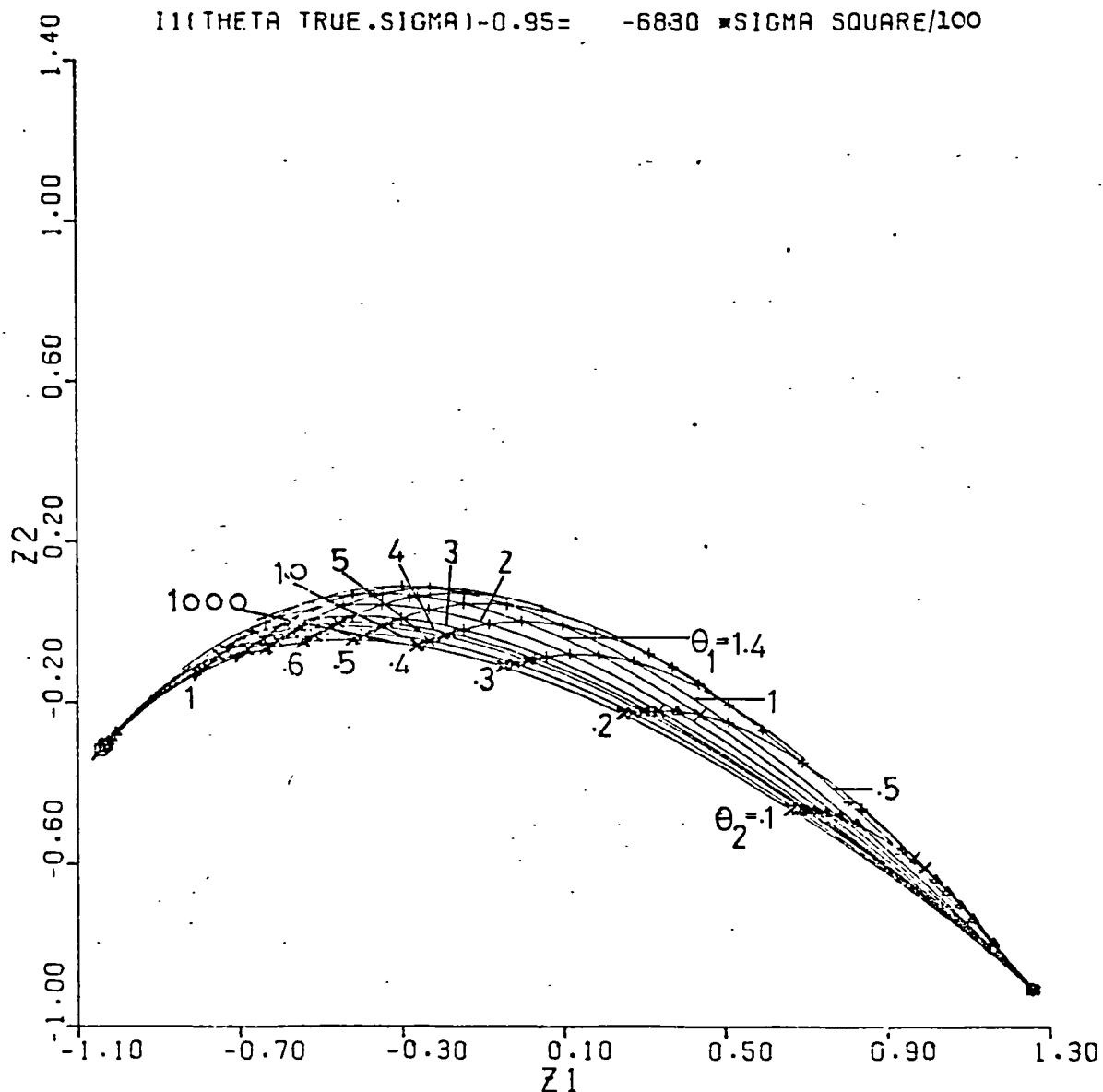
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$\begin{aligned} E(Y_i) = & 1 - (\theta_1 \exp(-\theta_2 x_i)) \\ & - \theta_2 \exp(-\theta_1 x_i) \\ & / (\theta_1 - \theta_2) \end{aligned}$$

$x_i = 1, 2, 3, 4, 5, 6$

θ_1 TRUE ARE 1.4000 0.4000

$I_1(\theta_1 \text{ TRUE}, \sigma) - 0.95 = -6830 \cdot \sigma^2 / 100$



R=ABSOLUTE VALUE OF: $(I_1(\theta_2, \sigma) - I_1(\theta_1 \text{ TRUE}, \sigma)) / (I_1(\theta_1 \text{ TRUE}, \sigma) - 0.95)$
PARAMETER OF INTEREST IS θ_2

+: $0 \leq R \leq 0.5$; X: $0.5 < R \leq 1$: A: $1 < R \leq 10$: D: $10 < R \leq 100$: *: $R > 100$

FIGURE (5.2.6)

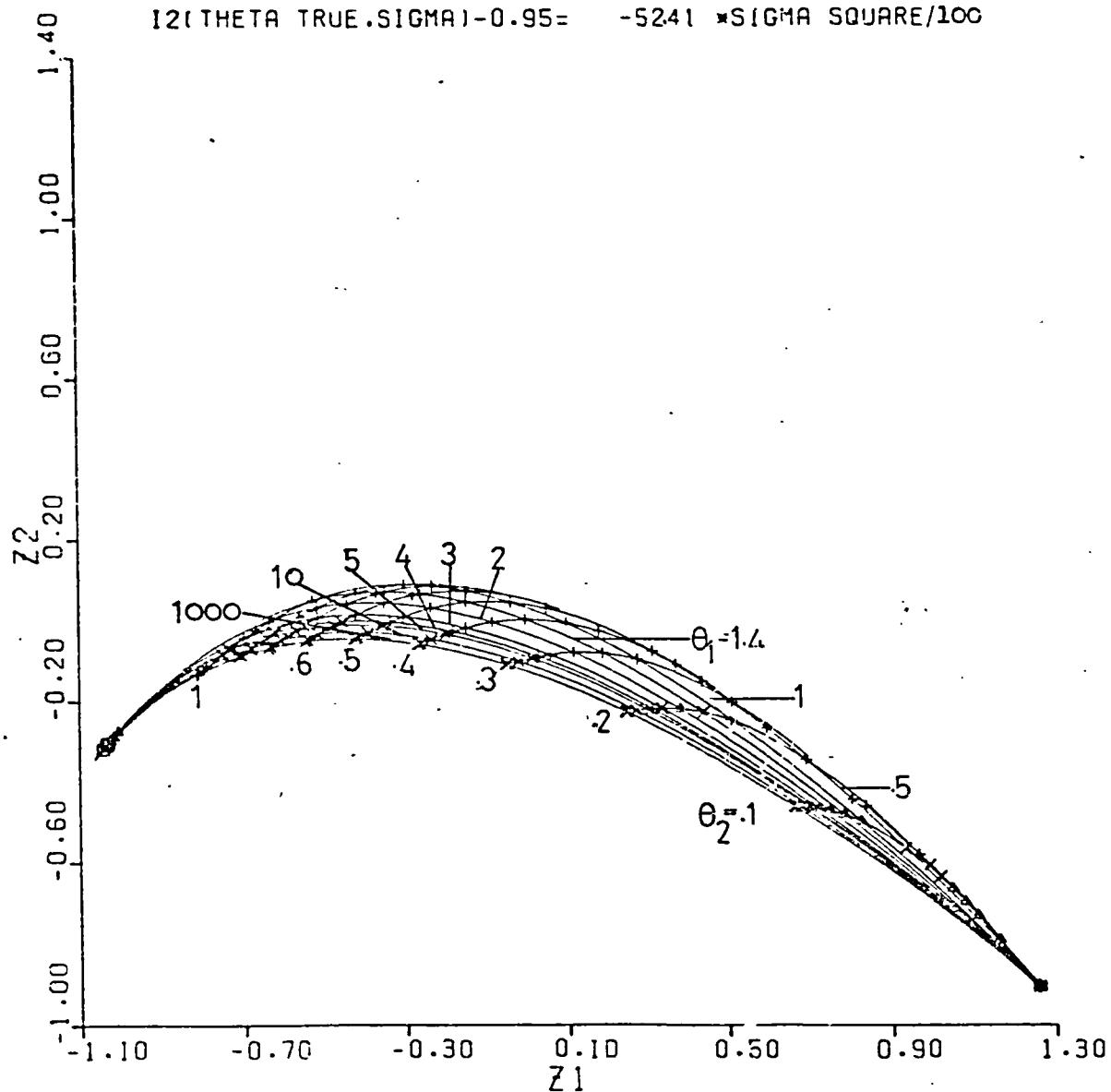
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

 $X_i = 1, 2, 3, 4, 5, 6$

THETA1 TRUE ARE 1.4000 0.4000

 $I_2(\Theta \text{ TRUE}, \sigma) - 0.95 = -5241 \cdot \sigma^2 / 100$ 

$R = \text{ABSOLUTE VALUE OF } (I_2(\Theta, \sigma) - I_2(\Theta \text{ TRUE}, \sigma)) / (I_2(\Theta \text{ TRUE}, \sigma) - 0.95)$

+ : $0 \leq R \leq 0.5$; x : $0.5 < R \leq 1$; △ : $1 < R \leq 10$; □ : $10 < R \leq 100$; * : $R > 100$

These figures indicate that provided that σ is sufficiently small, the difference between $I_i(\theta_T, \sigma)$ and $I_i(\hat{\theta}, \sigma)$ is small for most of the $\hat{\theta}$, and the estimation of $I_i(\theta_T, \sigma)$ using $I_i(\hat{\theta}, \sigma)$ can thus be regarded as feasible. For example consider model (A) with $\sigma = 0.1$ and model (B) with $\sigma = 0.02$. These figures indicate that for most of the $\hat{\theta}$,

$$|I_i(\hat{\theta}, \sigma) - I_i(\theta_T, \sigma)| \leq 0.00005$$

for model (A), and

$$|I_i(\hat{\theta}, \sigma) - I_i(\theta_T, \sigma)| \leq 0.00032$$

for model (B). Thus if we use $I_i(\hat{\theta}, \sigma)$ to estimate $I_i(\theta_T, \sigma)$, the errors involved are small for most of the $\hat{\theta}$.

In practice, after calculating $\hat{\theta}$ and $I_i(\hat{\theta}, \sigma)$, we may wish to get some indication of whether σ is small enough for the estimation of the unknown value of $I_i(\theta_T, \sigma)$ to be feasible. One suggestion is to examine the values of $I_i(\theta, \sigma)$ evaluated at θ which are such that the distance between the point $P(\theta)$ and $P(\hat{\theta})$ is less than δ , where $\delta > 0$. A plausible value of δ is 2σ .

Suppose we have obtained a particular $\hat{\theta}$ and there is indication that σ is small enough for the estimation of $I_i(\theta_T, \sigma)$ to be feasible. Then we may refer to the region estimate based on this particular $\hat{\theta}$ as an "approximately $100 I_i(\hat{\theta}, \sigma) - c\%$ " region estimate, where c is any number which can be regarded as negligible, in particular, $c = 0$.

We next consider the estimation of the following values of nonlinearity which are multiples of the measures of nonlinearity:

$$(5.2.1) \quad M_{\theta} = 100(p+2)\chi^2_{p+2}(\chi^2_{p,\alpha})N_{\theta},$$

$$(5.2.2) \quad M_{\psi} = 100(p+2)\chi^2_{p+2}(\chi^2_{p,\alpha})N_{\psi},$$

$$(5.2.3) \quad M_{\phi} = 100(p+2)\chi^2_{p+2}(\chi^2_{p,\alpha})N_{\phi},$$

$$(5.2.4) \quad M_{\theta_i} = 100(p+2)\chi^2_3(\chi^2_{1,\alpha})N_{\theta_i},$$

and

$$(5.2.5) \quad M_{\psi_i} = 100(p+2)\chi^2_3(\chi^2_{1,\alpha})N_{\psi_i},$$

where ψ is the transformation based on method 2 and the ψ_i are transformations based on methods 2 and 3. Note that the values of M_{θ} , M_{ψ} and M_{ϕ} evaluated at $\theta = \theta_A$ are the upper bounds of $|J_2(\theta_A, \sigma)|$, and the values of M_{θ_i} evaluated at $\theta = \theta_A$ are the upper bounds of the corresponding $|J_1(\theta_A, \sigma)|$, where

$$(5.2.6) \quad J_i(\theta_A, \sigma) = 100(I_i(\theta_A, \sigma) - (1 - \alpha)), \quad (i = 1, 2),$$

and the $I_i(\theta_A, \sigma)$ are evaluated by using (3.3.50) and (3.3.52). In fact

$$(5.2.7) \quad M_{\theta} \geq M_{\psi} \geq M_{\phi} \geq |J_2(\theta_A, \sigma)|.$$

Furthermore we have

$$(5.2.8) \quad M_{\theta_i} \geq M_{\psi_i} \geq |J_1(\theta_A, \sigma)|$$

if the transformation ψ_i is based on method 3.

We choose $\theta_T = (1.4, 0.4)^T$. We then set $\alpha = 0.05$ and calculate these values of nonlinearity at various values of θ . Let these values of nonlinearity be denoted by $M_\theta(\theta)$, $M_\psi(\theta)$, $M_\phi(\theta)$, $M_{\theta_i}(\theta)$ and $M_{\psi_i}(\theta)$. In Fig. (5.2.7) to (5.2.24), we display the absolute values of

$$\frac{M_\beta(\theta) - M_\beta(\theta_T)}{M_\beta(\theta_T)},$$

where $\beta = \theta, \psi, \phi, \theta_i$ and ψ_i :

These figures indicate that, for a given β , provided that σ is sufficiently small, the difference between $M_\beta(\hat{\theta})$ and $M_\beta(\theta_T)$ is small for most of the $\hat{\theta}$, and the estimation of $M_\beta(\theta_T)$ using $M_\beta(\hat{\theta})$ can thus be regarded as feasible. Note that the value of σ needs to be much smaller in the present case than in the case when we estimate $I_i(\theta_T, \sigma)$. Further, the larger the value of $M_\beta(\theta_T)/\sigma^2$, the smaller the value of σ should be for the estimation of $M_\beta(\theta_T)$ to be feasible.

FIGURE (5.2.7)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

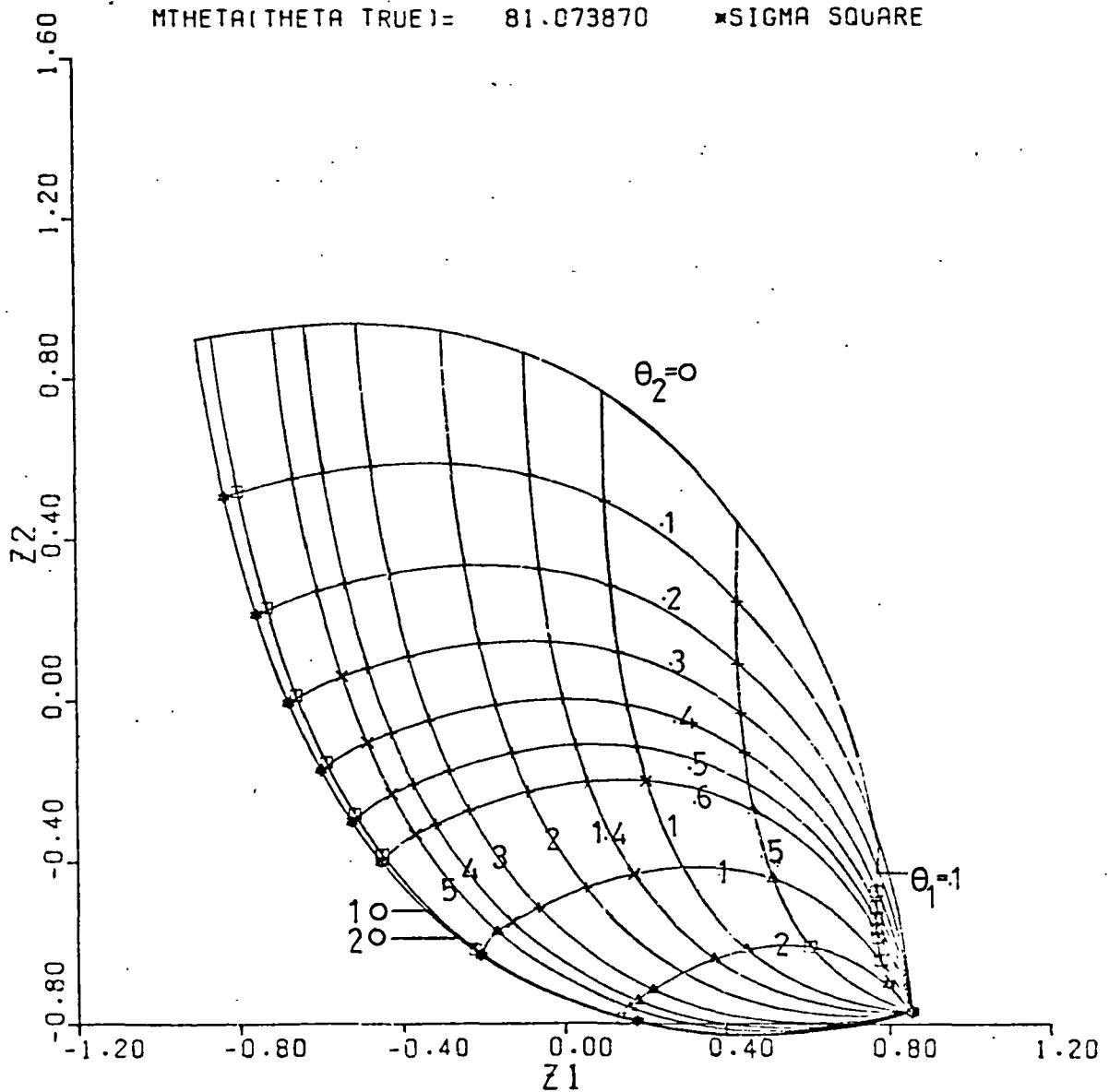
$$F(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\cdot (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

$$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$M\Theta_1(\Theta_1 \text{ TRUE}) = 81.073870 \quad * \text{SIGMA SQUARE}$$



R=ABSOLUTE VALUE OF ((M\Theta_1(\Theta_1)-M\Theta_1(\Theta_1 \text{ TRUE}))/M\Theta_1(\Theta_1 \text{ TRUE}))

+:0≤R≤0.5 ; X:0.5 < R ≤1 ; △:1 < R ≤10 ; □:10 < R ≤100 ; ☆:R>100

FIGURE (5.2.8)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

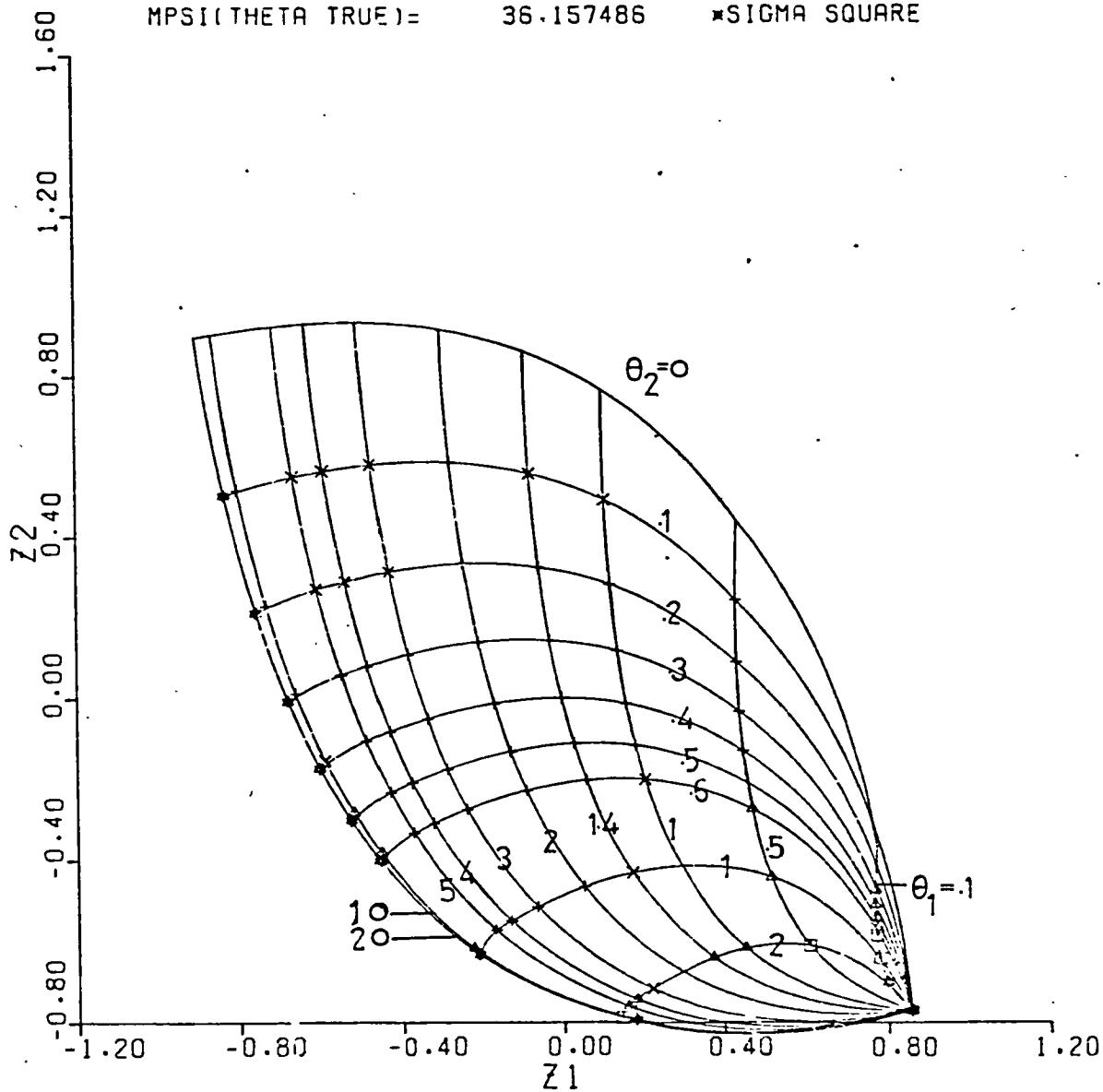
MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \\ * (\exp(-\Theta_2 * X) - \exp(-\Theta_1 * X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0, 0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$MPSI(\Theta \text{ TRUE}) = 36.157486 \quad * \text{SIGMA SQUARE}$$



R=ABSOLUTE VALUE OF ((MPSI(THETA)-MPSI(THETA TRUE))/MPSI(THETA TRUE))
 PSI IS TRANSFORMATION BASED ON METHOD 2

+ : 0 ≤ R ≤ 0.5 ; X : 0.5 < R ≤ 1 ; Δ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; * : R > 100

FIGURE (5.2.9)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

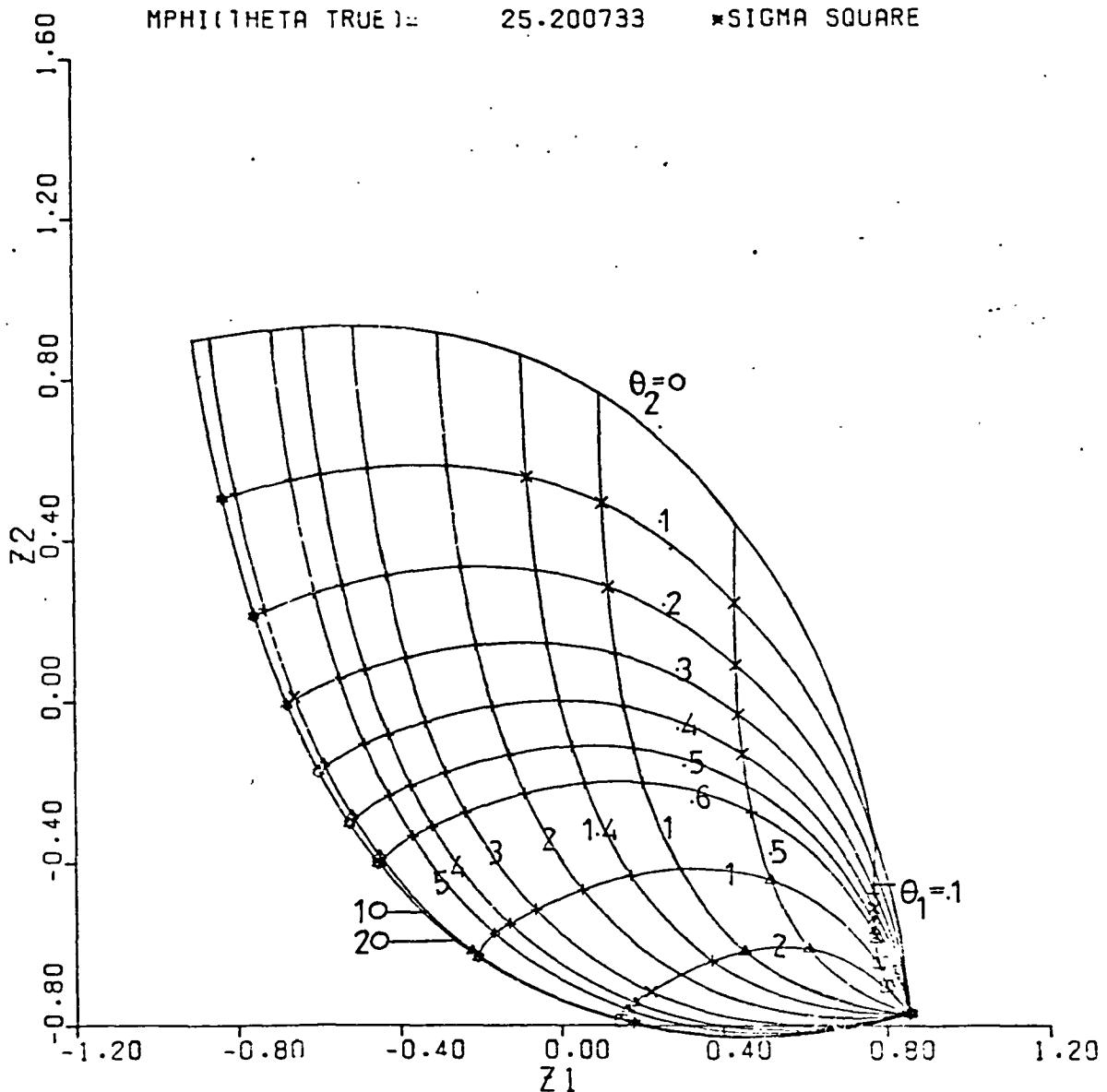
$$E(\xi) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot \xi) - \exp(-\Theta_1 \cdot \xi))$$

$$\xi = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad \Theta_2 \text{ TRUE }$$

$$MPHI(\Theta_1 \text{ TRUE}) = 25.200733 \quad * \text{SIGMA SQUARE}$$



$$R = \text{ABSOLUTE VALUE OF} (MPHI(\Theta) - MPhi(\Theta \text{ TRUE})) / MPhi(\Theta \text{ TRUE})$$

$$+ : 0 \leq R \leq 0.5 \quad : \times : 0.5 < R \leq 1 \quad : \Delta : 1 < R \leq 10 \quad : \square : 10 < R \leq 100 ; \star : R > 100$$

FIGURE (5.2.10)

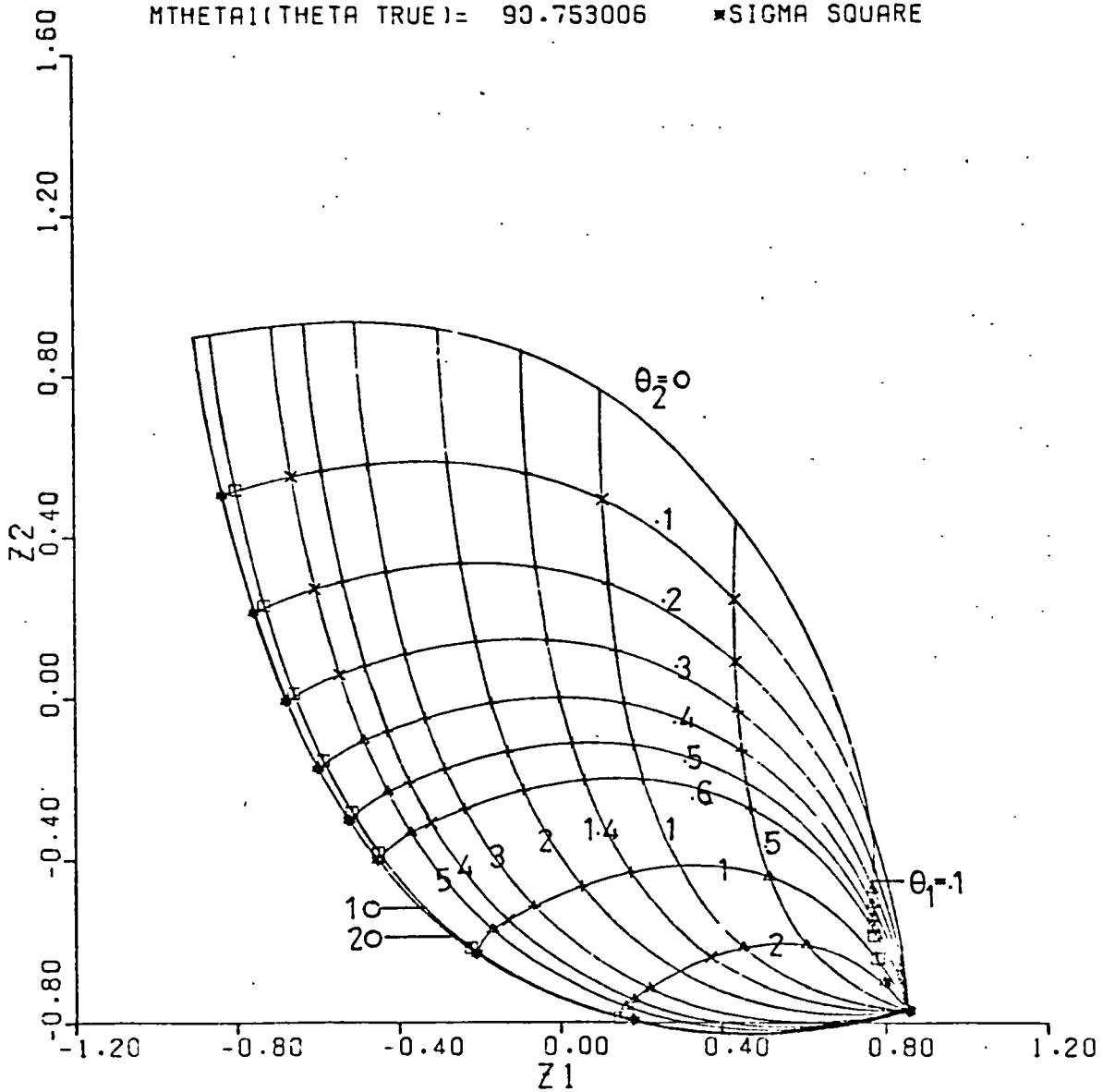
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

Θ_1 TRUE ARE 1.4000 C.4000

MTHETA1(THETA TRUE) = 90.753006 *SIGMA SQUARE



R=ABSOLUTE VALUE OF ((MTHETA1(THETA)-MTHETA1(THETA TRUE))/MTHETA1(THETA TRUE))

+ : 0 ≤ R ≤ 0.5 : X : 0.5 < R ≤ 1 : △ : 1 < R ≤ 10 : □ : 10 < R ≤ 100 : ♦ : R > 100

FIGURE (5.2.11)

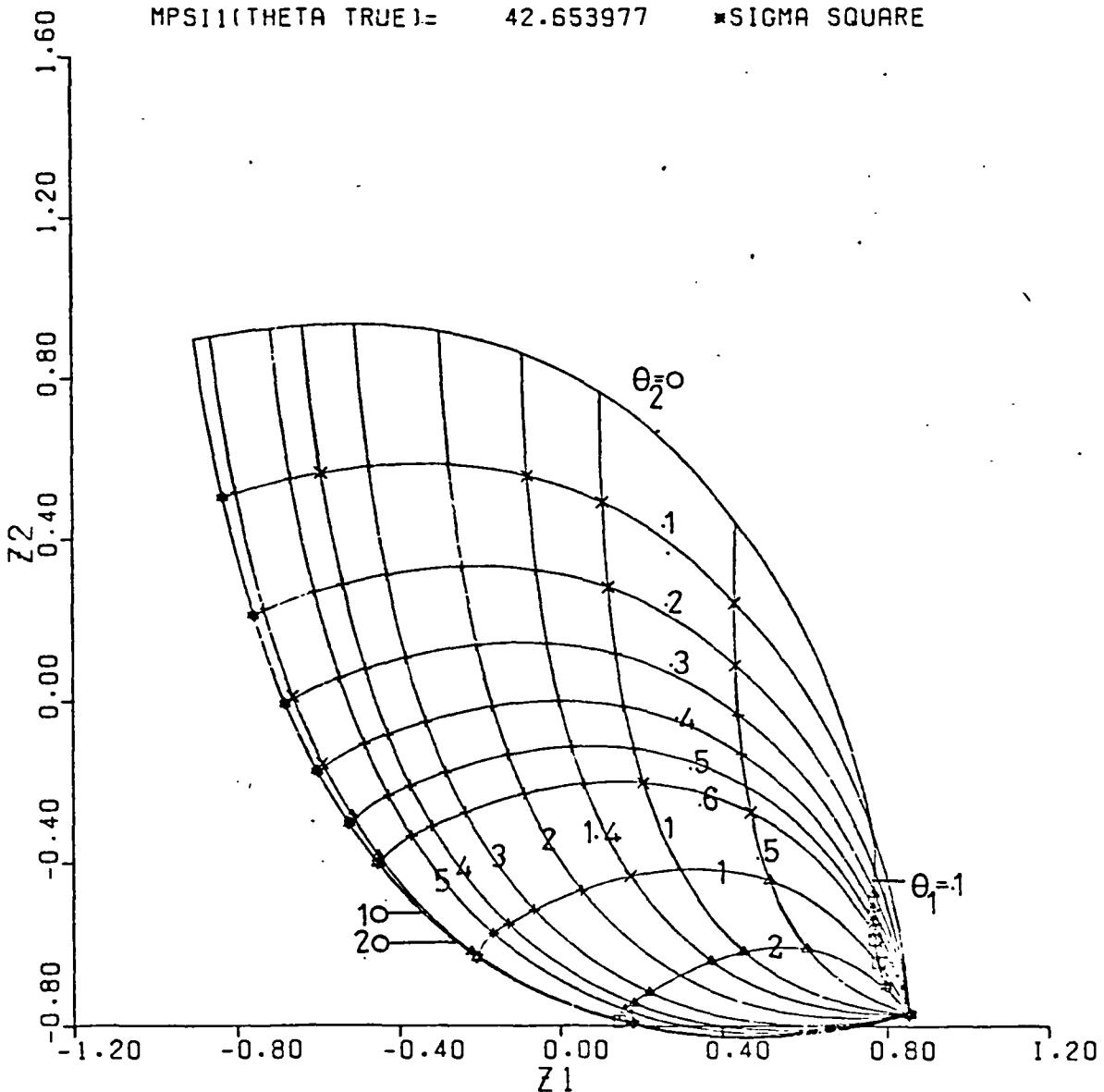
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \cdot (e^{-(\Theta_2 \cdot X)} - e^{-(\Theta_1 \cdot X)})$$

$$X_I = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$MPSII(\Theta \text{ TRUE}) = 42.653977 \quad * \text{SIGMA SQUARE}$$



R=ABSOLUTE VALUE OF (MPSII(THETA)-MPSII(THETA TRUE))/MPSII(THETA TRUE)
PSII IS POWER TRANSFORMATION BASED ON METHOD 2

+ : 0 ≤ R ≤ 0.5 ; . : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ☆ : R > 100

FIGURE (5.2.12)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

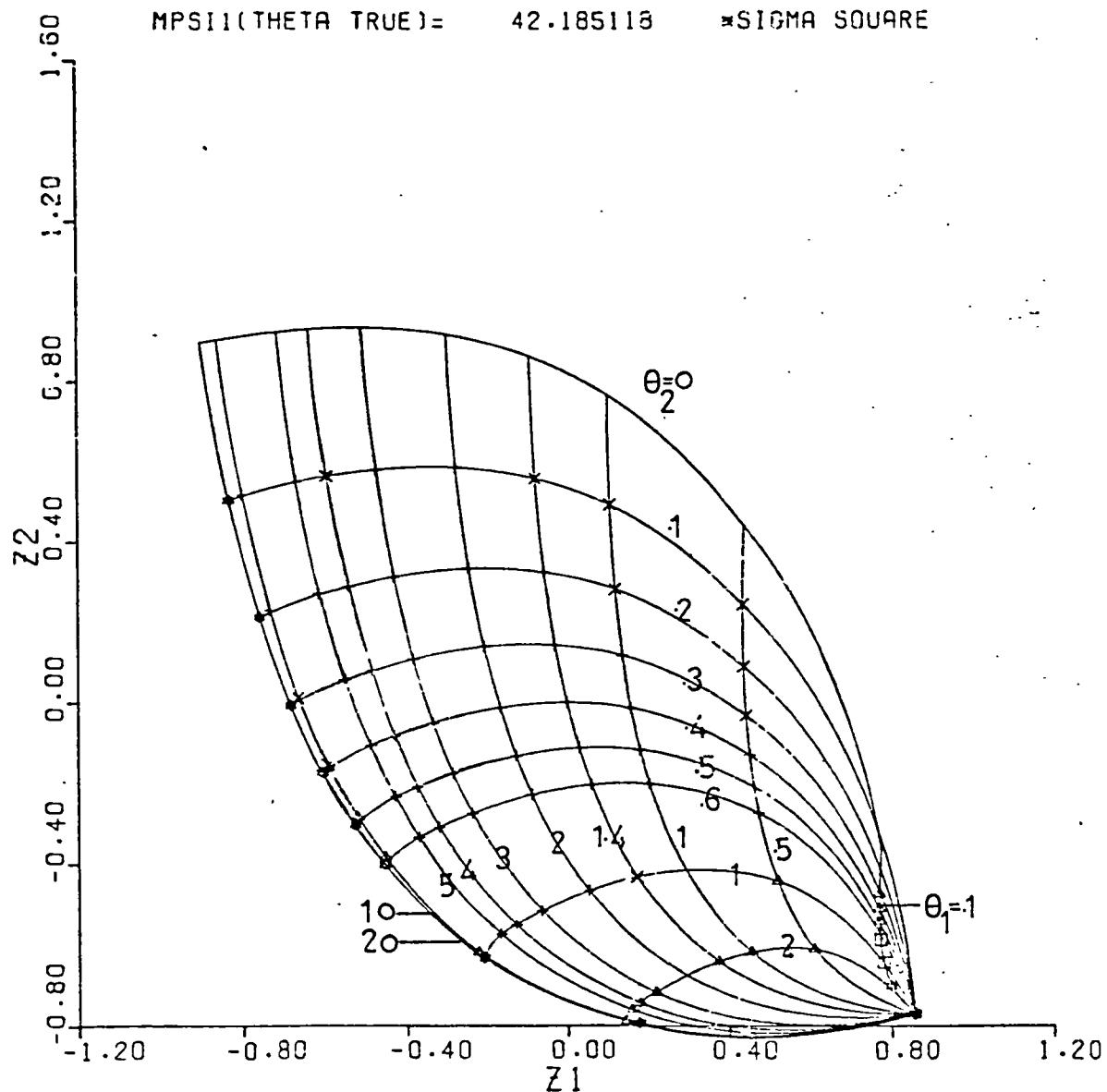
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$MPSII(\Theta \text{ TRUE}) = 42.185118 \quad \star \text{SIGMA SQUARE}$$



R=ABSOLUTE VALUE OF ((MPSII(THETA)-MPSII(THETA TRUE))/MPSII(THETA TRUE))
 PSII IS POWER TRANSFORMATION BASED ON METHOD 3

+: 0 ≤ R ≤ 0.5 ; X: 0.5 < R ≤ 1 ; △: 1 < R ≤ 10 ; □: 10 < R ≤ 100 ; ☆: R > 100

FIGURE (5.2.13)

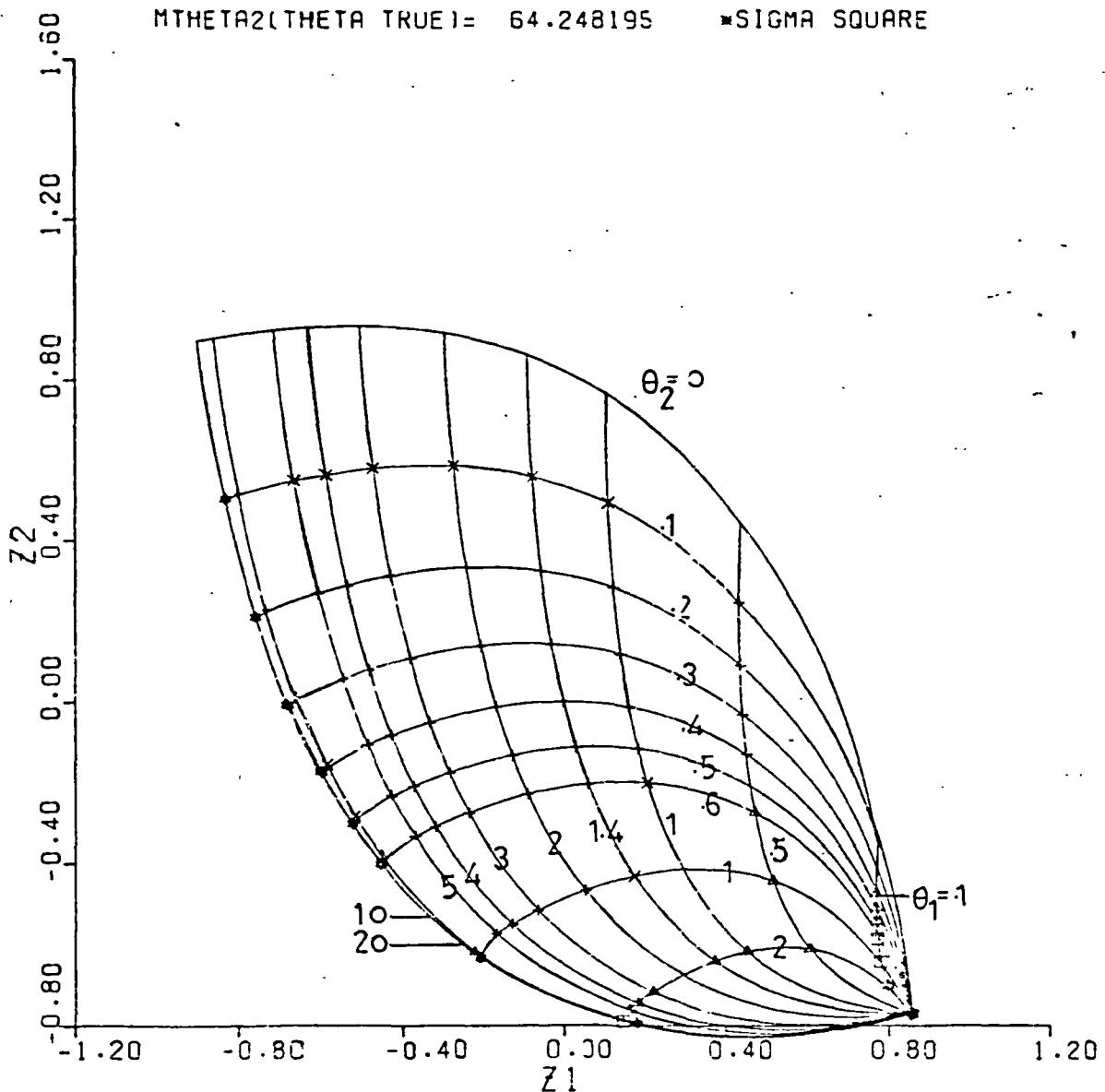
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \\ * (\exp(-\Theta_2 * X) - \exp(-\Theta_1 * X))$$

$$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$M\Theta_2(\Theta \text{ TRUE}) = 64.248195 \quad * \text{SIGMA SQUARE}$$



$$R = \text{ABSOLUTE VALUE OF } ((M\Theta_2(\Theta) - M\Theta_2(\Theta \text{ TRUE})) / M\Theta_2(\Theta \text{ TRUE}))$$

$$+ : 0 \leq R \leq 0.5 ; \times : 0.5 < R \leq 1 ; \Delta : 1 < R \leq 10 ; \square : 10 < R \leq 100 ; \star : R > 100$$

FIGURE (5.2.14)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

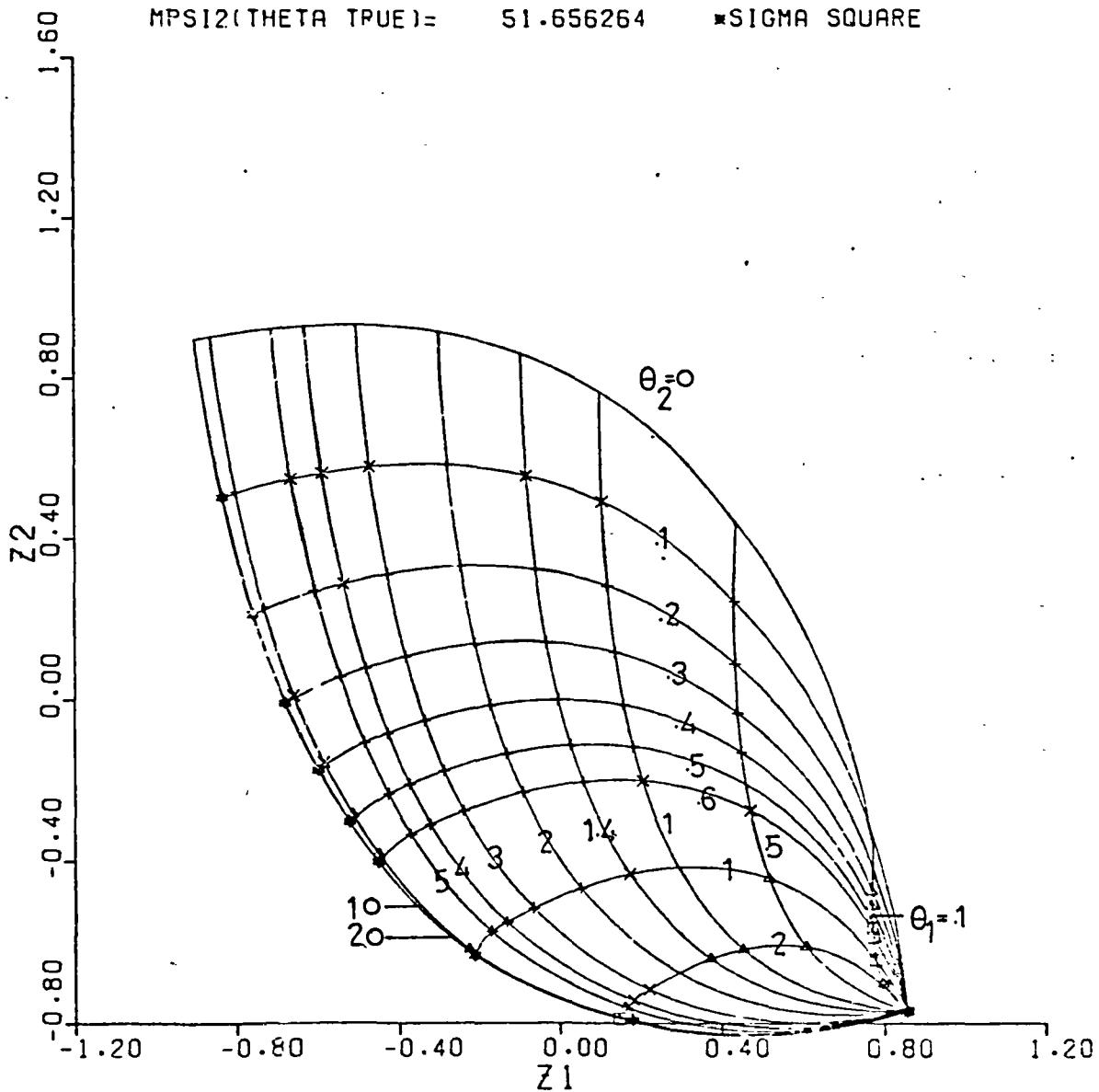
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 0.4, 0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$MPS12(\Theta \text{ TRUE}) = 51.656264 \quad * \text{SIGMA SQUARE}$$



R=ABSOLUTE VALUE OF (MPS12(THETA)-MPS12(THETA TRUE))/MPS12(THETA TRUE)
MPS12 IS POWER TRANSFORMATION BASED ON METHOD 2

+ : 0 ≤ R ≤ 0.5 : X : 0.5 < R ≤ 1 : △ : 1 < R ≤ 10 : □ : 10 < R ≤ 100 : * : R > 100

FIGURE (5.2.15)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

MODEL IS

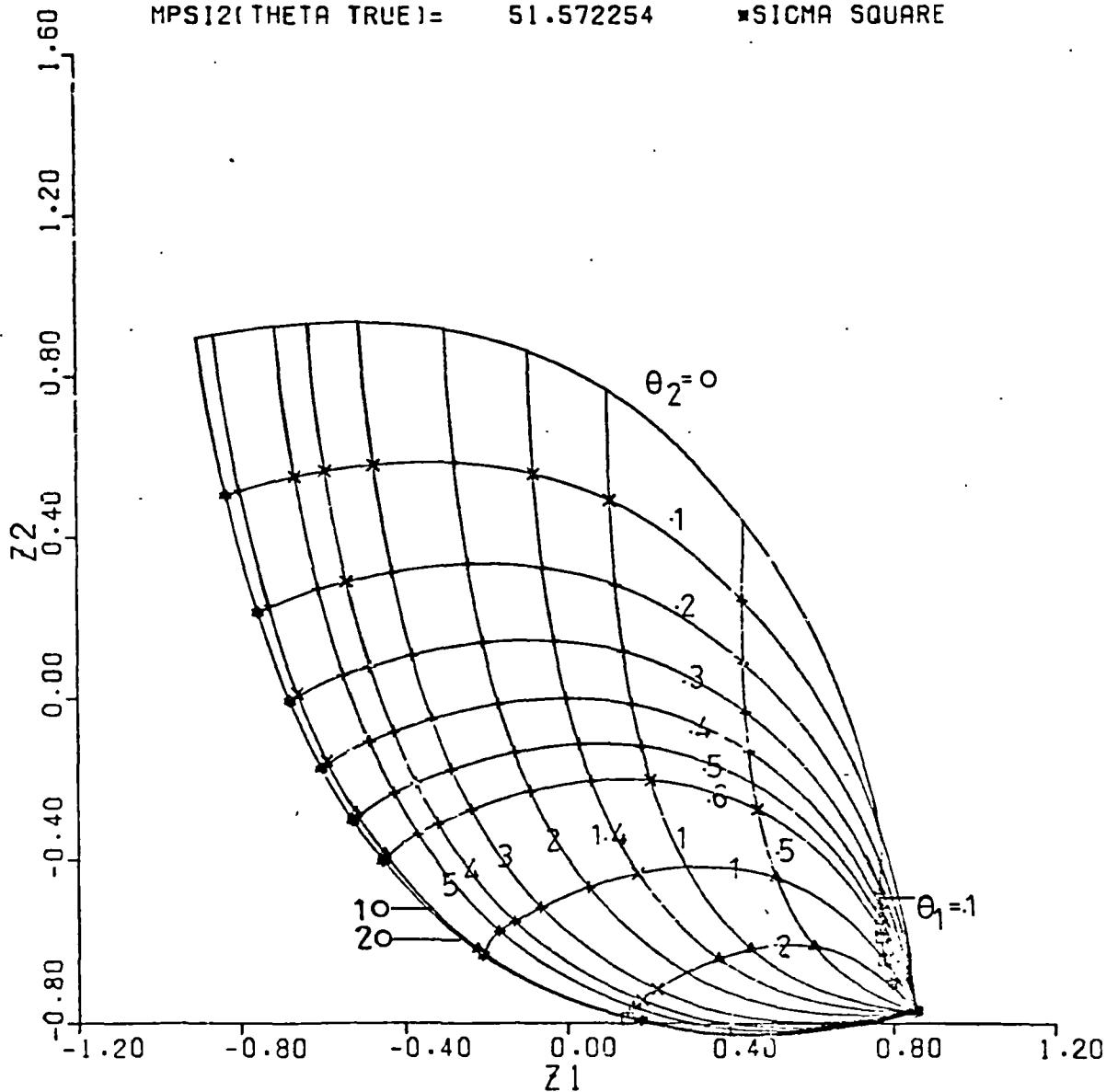
$$E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$$

$$\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$$

$$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$$

$$\Theta_1 \text{ TRUE ARE } 1.4000 \quad 0.4000$$

$$MPSI2(\Theta_1 \text{ TRUE}) = 51.572254 \quad \sigma^2$$



$R = \text{ABSOLUTE VALUE OF } ((MPSI2(\Theta) - MPSI2(\Theta_1 \text{ TRUE})) / MPSI2(\Theta_1 \text{ TRUE}))$
 $MPSI2 \text{ IS POWER TRANSFORMATION BASED ON METHOD 3}$

$+ : 0 \leq R \leq 0.5 \quad \times : 0.5 < R \leq 1 \quad \Delta : 1 < R \leq 10 \quad \square : 10 < R \leq 100 \quad \star : R > 100$

FIGURE (5.2.16)

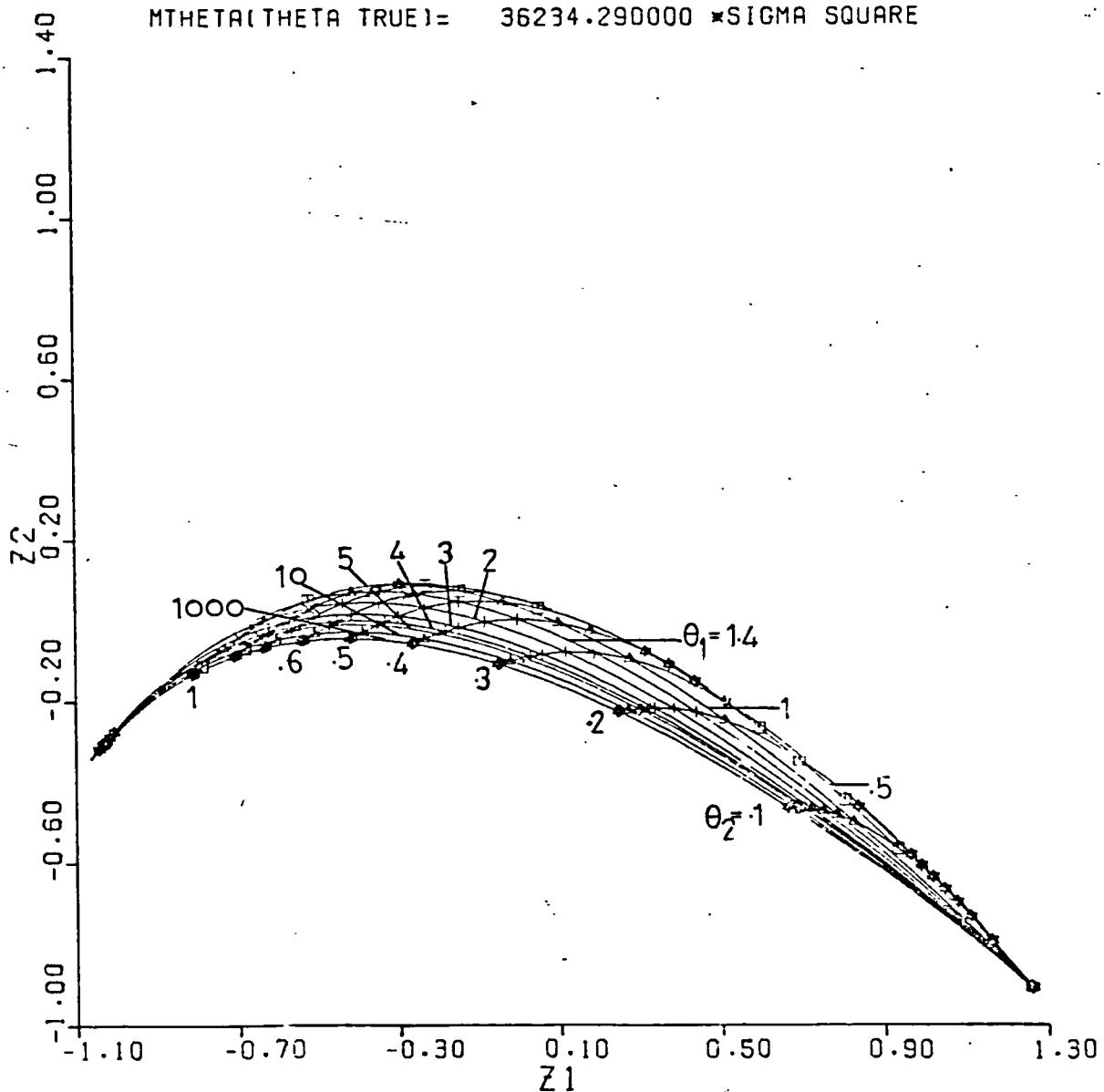
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = 1 - (\Theta_1 \times \exp(-\Theta_2 \times X)) \\ - (\Theta_2 \times \exp(-\Theta_1 \times X)) \\ / (\Theta_1 - \Theta_2)$$

$X_i = 1, 2, 3, 4, 5, 6$

Θ_1 TRUE ARE 1.4000 0.4000

$M\Theta_1(\Theta_1 \text{ TRUE}) = 36234.290000 * \text{SIGMA SQUARE}$



R = ABSOLUTE VALUE OF (MΘ1(Θ1) - MΘ1(Θ1 TRUE)) / MΘ1(Θ1 TRUE)

+ : 0 ≤ R ≤ 0.5 : × : 0.5 < R ≤ 1 : △ : 1 < R ≤ 10 : □ : 10 < R ≤ 100 : ♦ : R > 100

FIGURE (5.2.17)

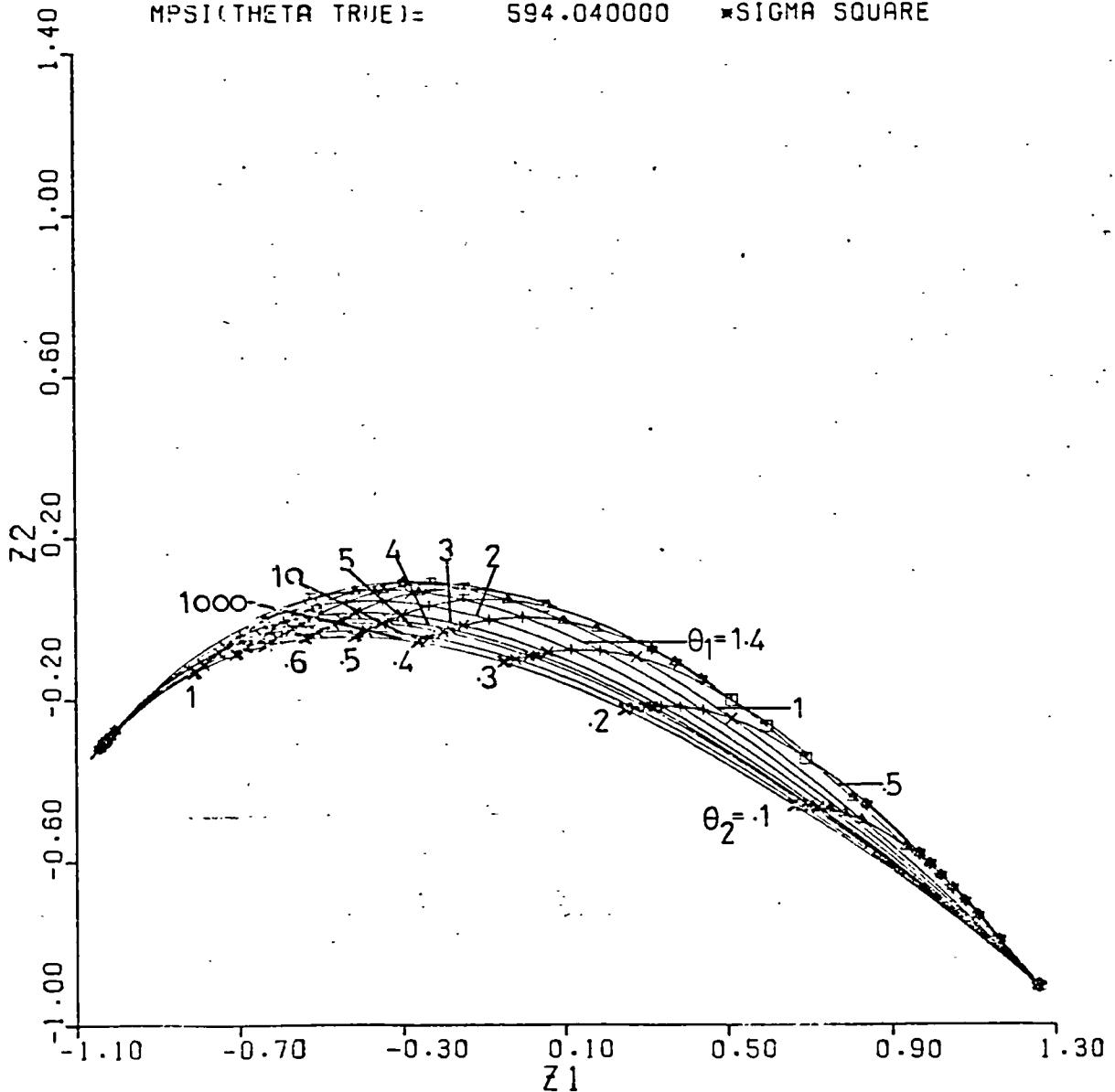
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X_1)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X_1)) \\ / (\Theta_1 - \Theta_2)$$

$X_1 = 1, 2, 3, 4, 5, 6$

Θ_1 TRUE ARE 1.4000 0.4000

MPSI(Θ TRUE) = 594.040000 *SIGMA SQUARE



R = ABSOLUTE VALUE OF (MPSI(Θ) - MPSI(Θ TRUE)) / MPSI(Θ TRUE)
PSI IS TRANSFORMATION BASED ON METHOD 2

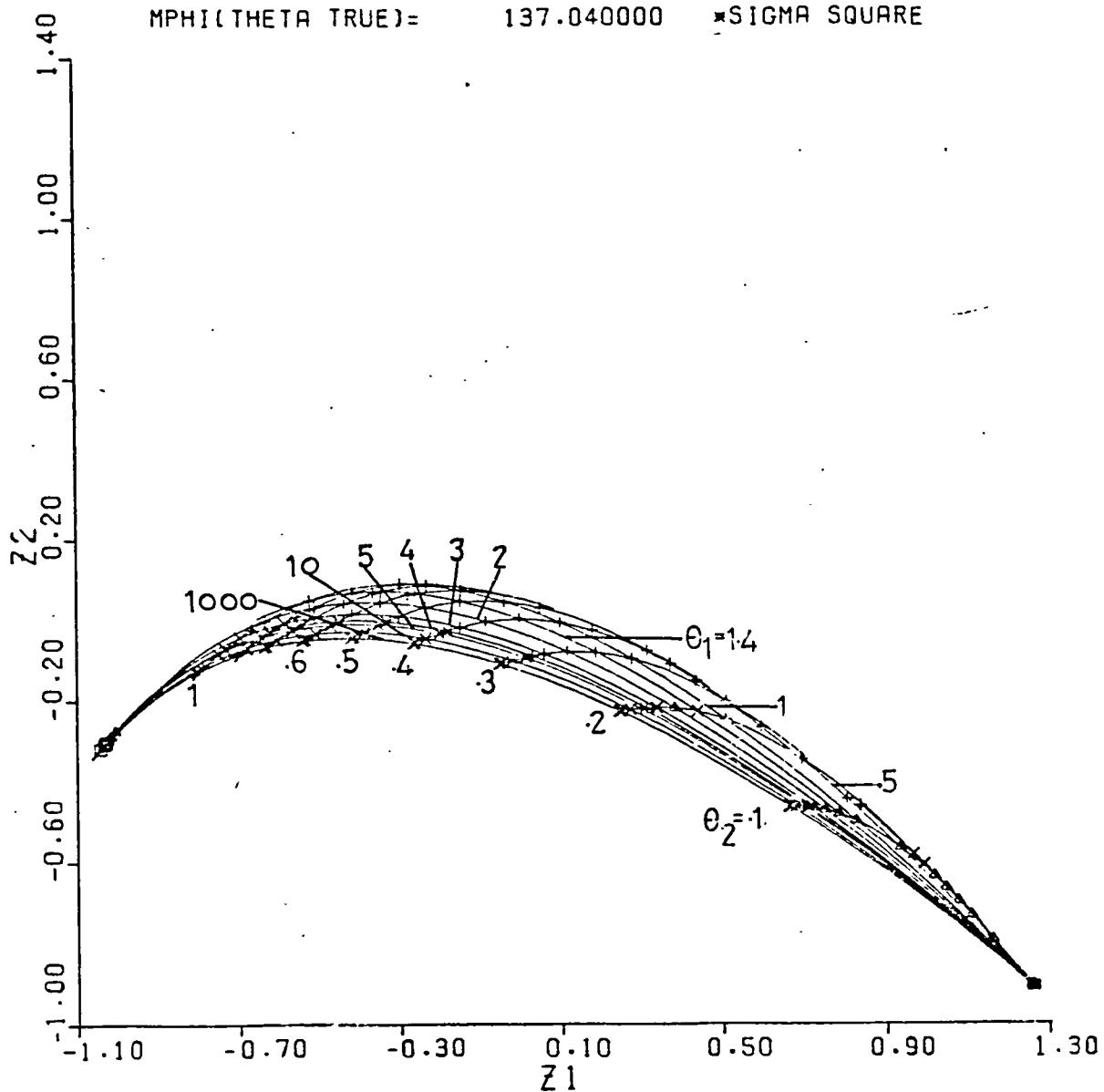
+ : 0 ≤ R ≤ 0.5 ; x : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ♦ : R > 100

FIGURE (5.2.18)
 ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
 MODEL IS
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X_1))$
 $\quad \quad \quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X_1)))$
 $\quad \quad \quad / (\Theta_1 - \Theta_2)$

$X_1 = 1.2.3.4.5.6$

THETA1 TRUE ARE 1.4000 0.4000

MPHI(THETA TRUE) = 137.040000 * SIGMA SQUARE



R=ABSOLUTE VALUE OF [(MPHI(I,THETA)-MPHI(THETA TRUE))/MPHI(THETA TRUE)]

+ : 0 ≤ R ≤ 0.5 ; X : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ♦ : R > 100

FIGURE (5.2.19)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

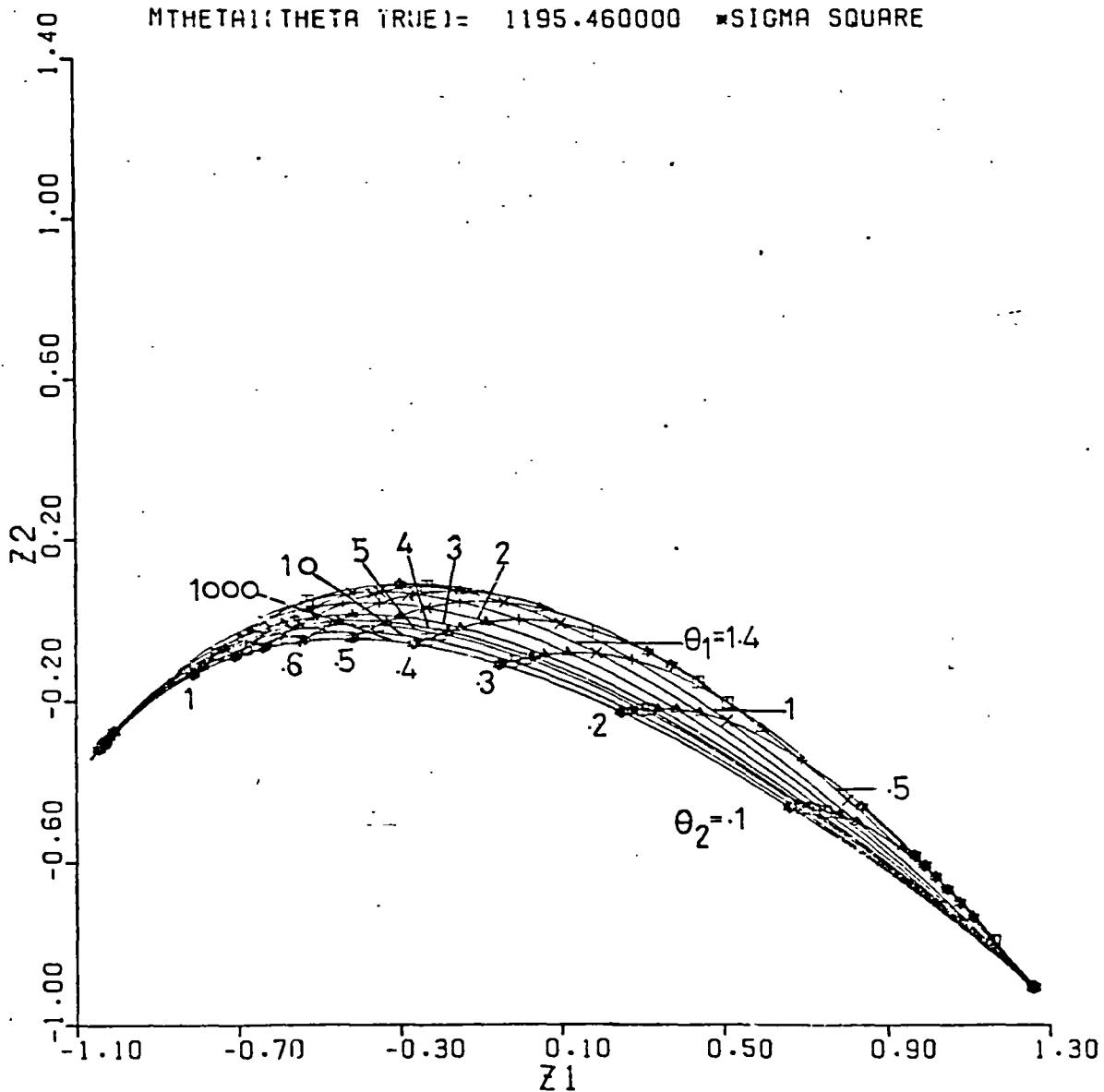
MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

 $X_i = 1, 2, 3, 4, 5, 6$

THETA1 TRUE ARE 1.4000 0.4000

MTHETA1(THETA TRUE) = 1195.460000 * SIGMA SQUARE



R=ABSOLUTE VALUE OF ((MTHETA1(THETA)-MTHETA1(THETA TRUE))/MTHETA1(THETA TRUE))

+ : 0 ≤ R ≤ 0.5 ; × : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ♦ : R > 100

FIGURE (5.2.20)

ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS

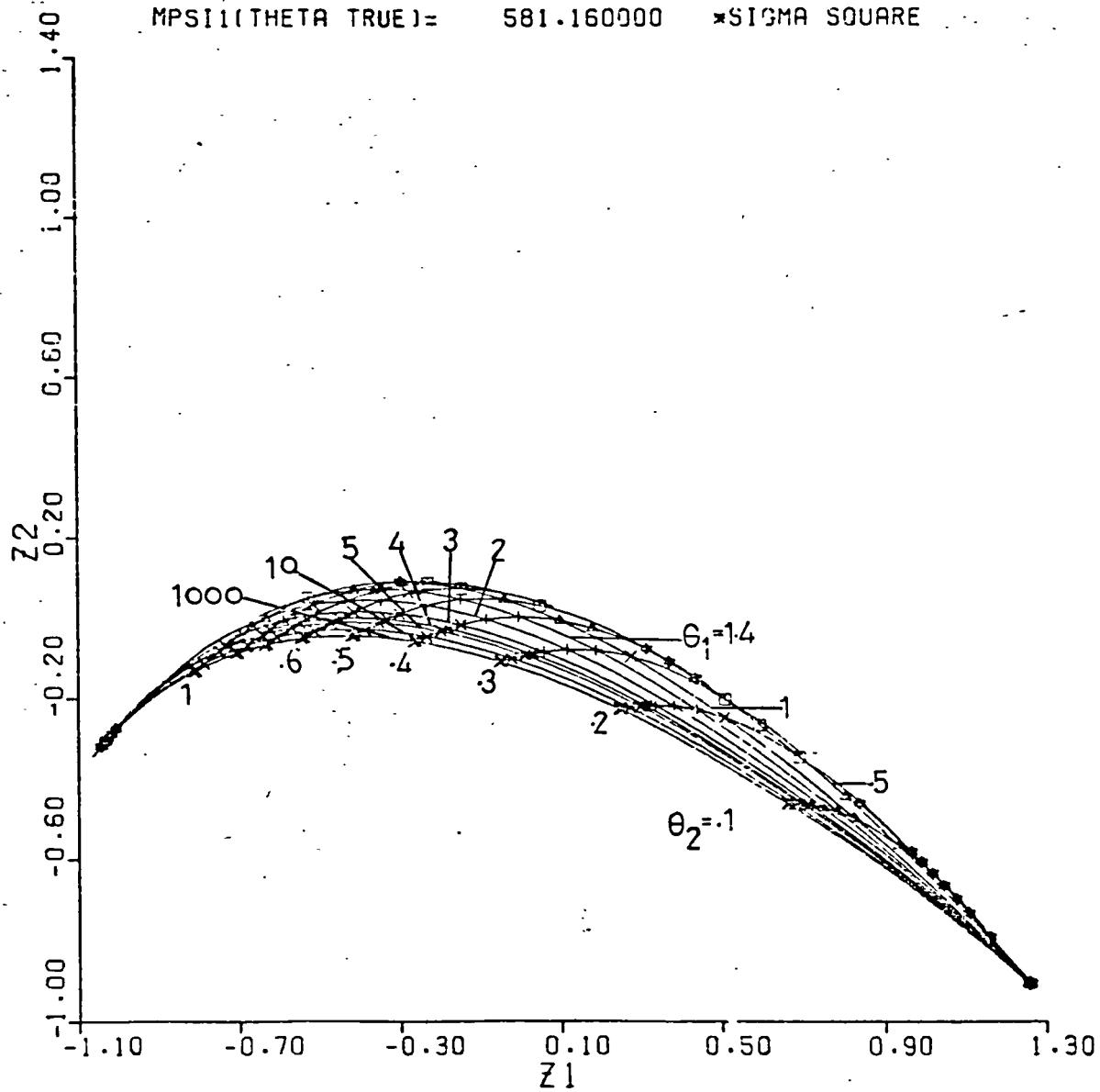
MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

 $X_i = 1, 2, 3, 4, 5, 6$

THETA1 TRUE ARE 1.4000 0.4000

MPSII(THETA TRUE) = 581.160000 *SIGMA SQUARE



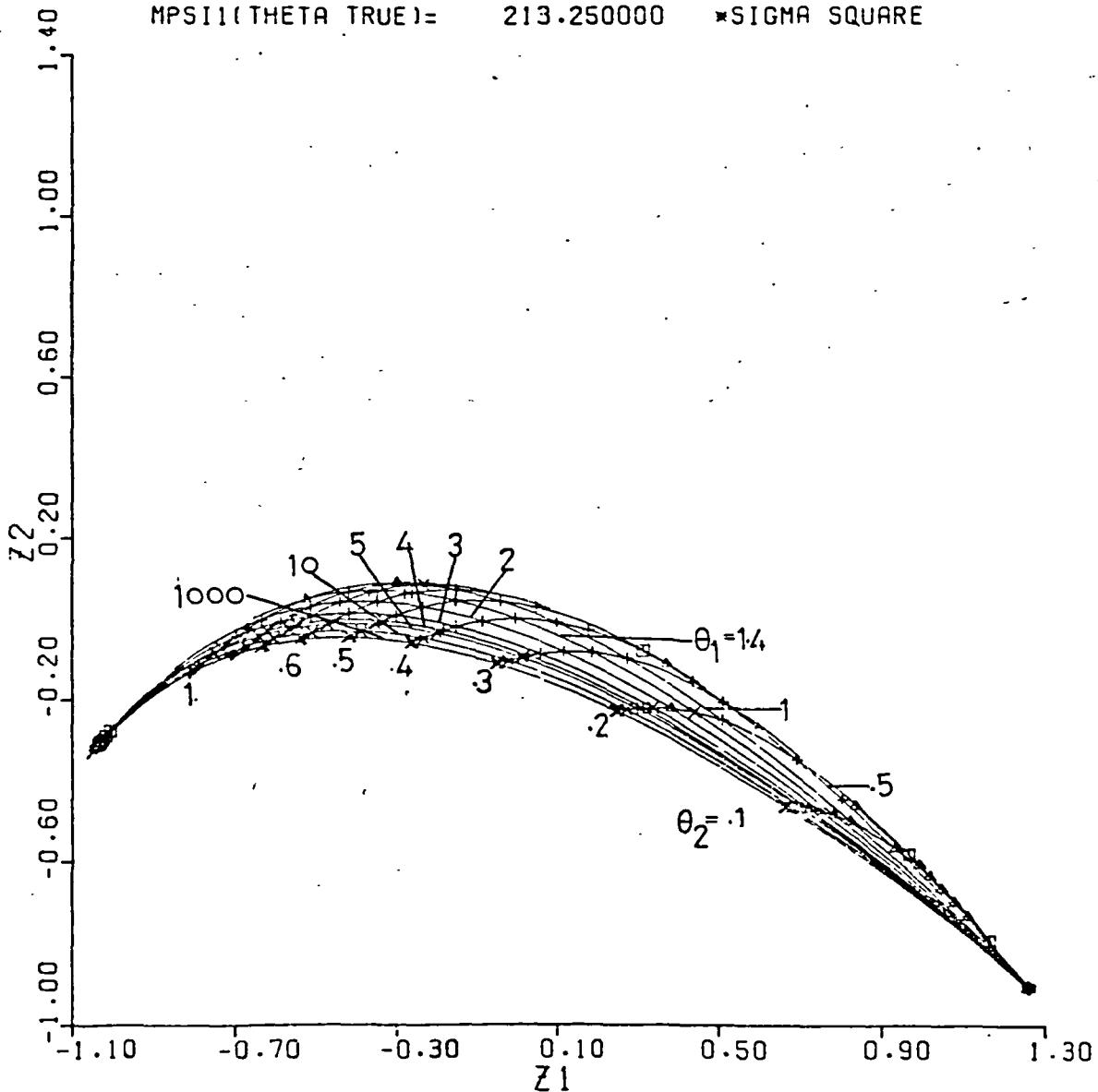
R=ABSOLUTE VALUE OF (MPSII(THETA)-MPSII(THETA TRUE))/MPSII(THETA TRUE)
 MPSII IS POWER TRANSFORMATION BASED ON METHOD 2

+:0≤R≤0.5 ;X:0.5 < R ≤ 1 ;Δ:1 < R ≤ 10 ;□:10 < R ≤ 100 ;◊:R > 100

FIGURE (5.2.21)
 ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
 MODEL IS
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X_1))$
 $\quad \quad \quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X_1)))$
 $\quad \quad \quad / (\Theta_1 + \Theta_2)$
 $X_1 = 1, 2, 3, 4, 5, 6$

THETA1 TRUE ARE 1.4000 0.4000

MPSII(THETA TRUE)= 213.250000 *SIGMA SQUARE



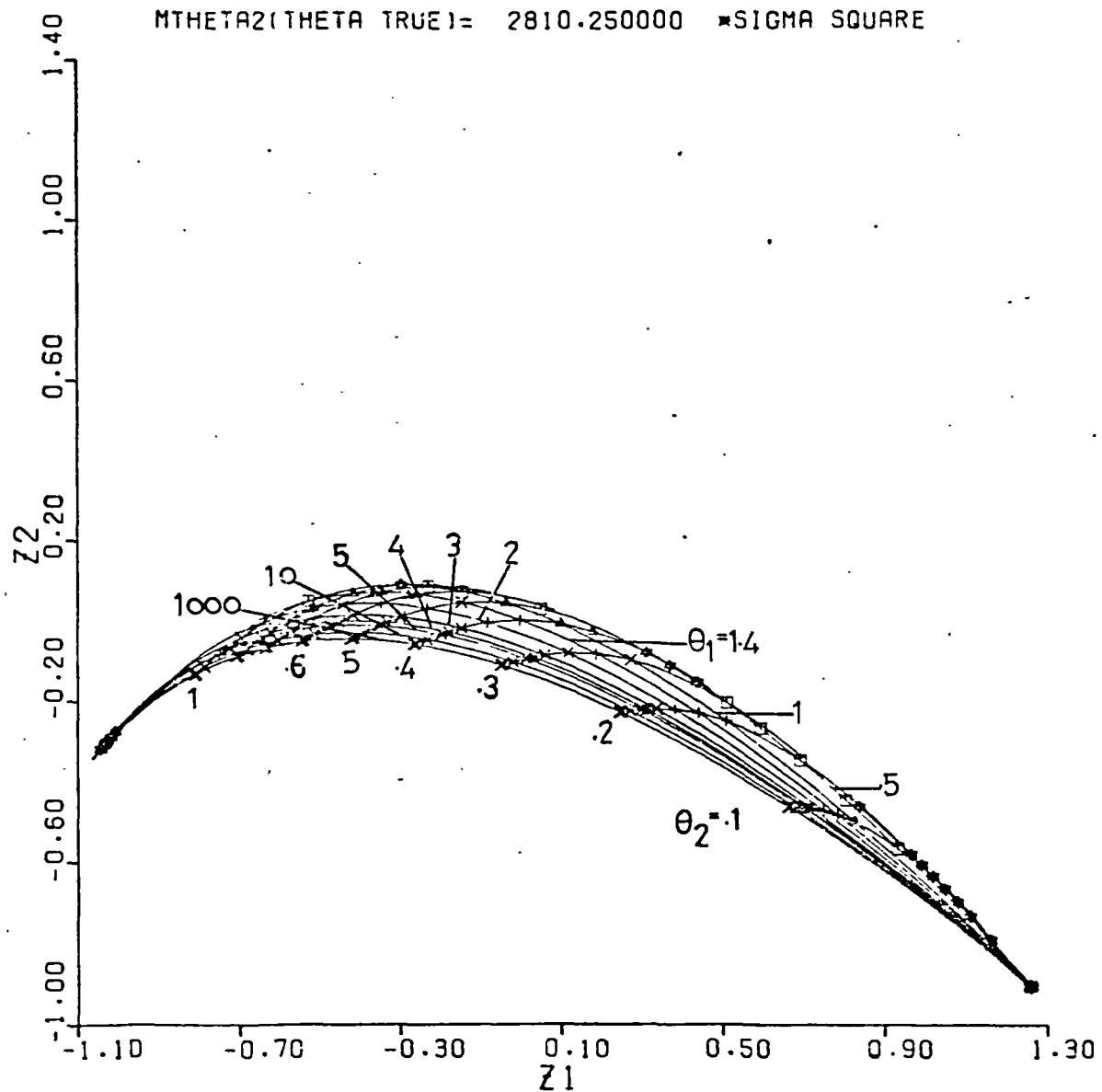
R=ABSOLUTE VALUE OF ((MPSII(THETA)-MPSII(THETA TRUE))/MPSII(THETA TRUE))
 MPSII IS POWER TRANSFORMATION BASED ON METHOD 3

+:0≤R≤0.5 ; X:0.5 < R ≤ 1 ; Δ:1 < R ≤ 10 ; □:10 < R ≤ 100 ; ♦:R > 100

FIGURE (5.2.22)
 ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
 MODEL IS
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $- (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $/(\Theta_1 - \Theta_2)$
 $X = 1.2.3.4.5.6$

THETA1 TRUE ARE 1.4000 0.4000

MTHETA2(THETA TRUE) = 2810.250000 *SIGMA SQUARE



R=ABSOLUTE VALUE OF (MTHETA2(THETA)-MTHETA2(THETA TRUE))/MTHETA2(THETA TRUE)

+ : 0 ≤ R ≤ 0.5 ; X : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; ♦ : R > 100

FIGURE (5.2.23)

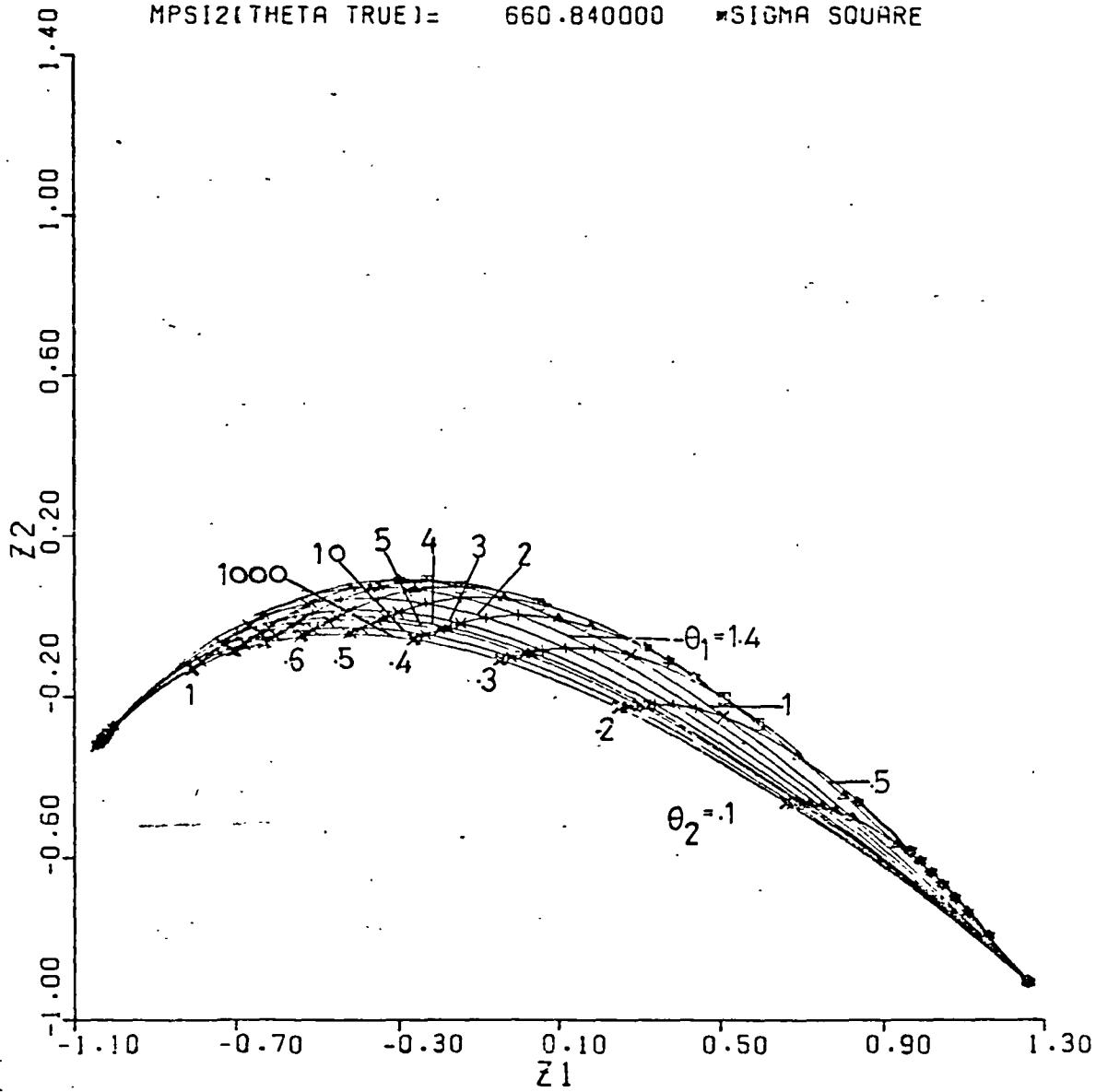
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = 1 - (\theta_1 \cdot \exp(-\theta_2 \cdot X_1)) \\ - (\theta_2 \cdot \exp(-\theta_1 \cdot X_1)) \\ / (\theta_1 - \theta_2)$$

$X_1 = 1, 2, 3, 4, 5, 6$

θ_1 TRUE ARE 1.4000 0.4000

MPSI2(θ TRUE) = 660.840000 σ^2 SIGMA SQUARE



R = ABSOLUTE VALUE OF (MPSI2(θ) - MPSI2(θ TRUE)) / MPSI2(θ TRUE)
PSI2 IS POWER TRANSFORMATION BASED ON METHOD 2

+ : 0 ≤ R ≤ 0.5 ; x : 0.5 < R ≤ 1 ; △ : 1 < R ≤ 10 ; □ : 10 < R ≤ 100 ; * : R > 100

FIGURE (5.2.24)

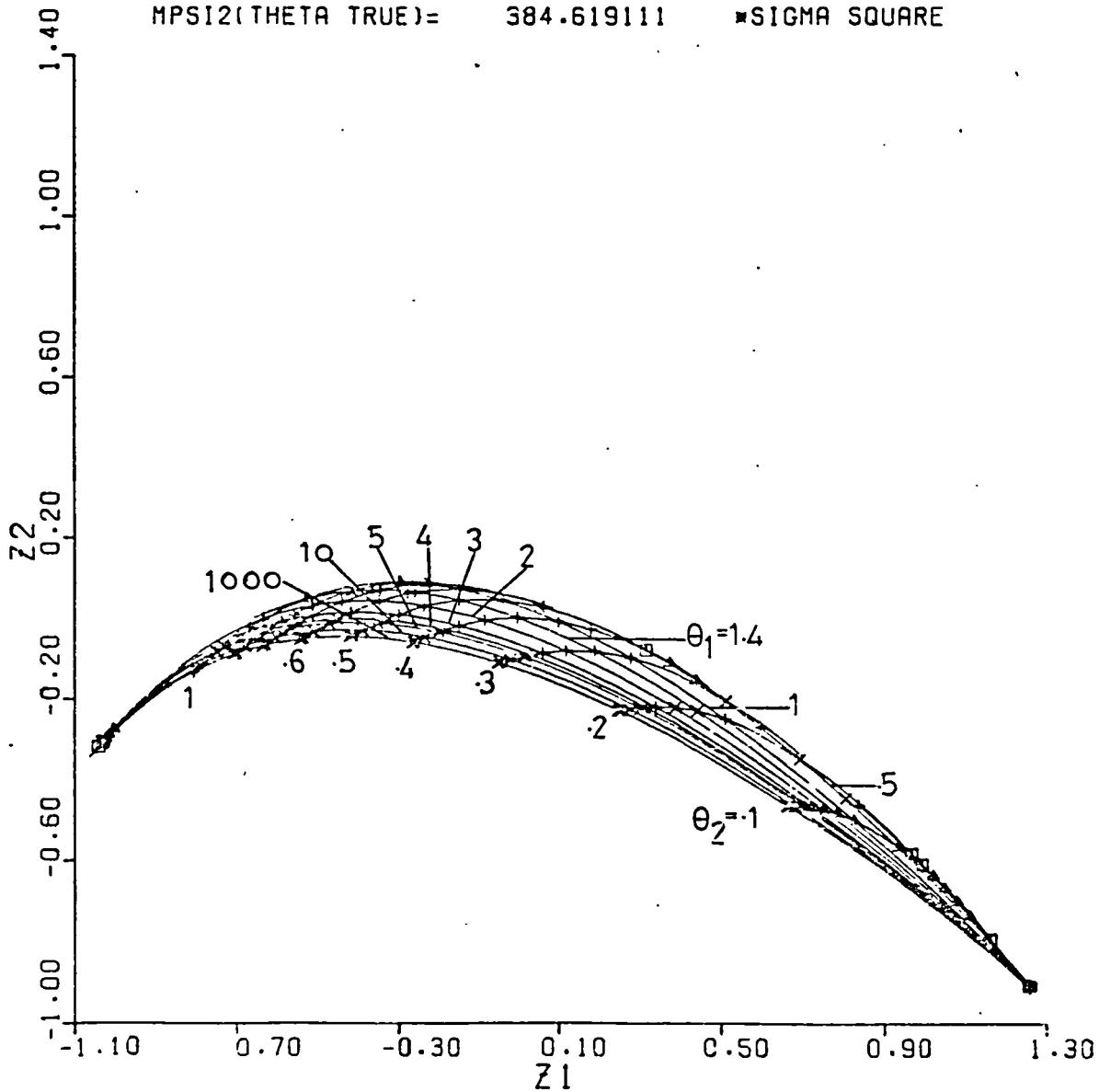
ROTATED COORDINATES OF POINTS IN SOLUTION LOCUS
MODEL IS

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

$X_i = 1, 2, 3, 4, 5, 6$

Θ_1 TRUE ARE 1.4000 0.4000

MPSI2(Θ TRUE) = 384.619111 *SIGMA SQUARE



R=ABSOLUTE VALUE OF ((MPSI2(Θ) - MPSI2(Θ TRUE)) / MPSI2(Θ TRUE))
PSI2 IS POWER TRANSFORMATION BASED ON METHOD 3

+ : 0 < R < 0.5 ; X : 0.5 < R < 1 ; Δ : 1 < R < 10 ; □ : 10 < R < 100 ; ♦ : R > 100

Section 5.3 Region estimates of $\underline{\theta}$

In this section we compare the boundaries of the region estimates given by the four methods. The levels of σ are set to be sufficiently small to ensure that models (A) and (B) with $\underline{\theta}_T = (1.4, 0.4)^T$ can be treated as unconstrained models for statistical purposes. For model (A), we consider three values of $\hat{\underline{\theta}}$, namely $(1.4, 0.4)^T$, $(2.0, 0.2)^T$ and $(1.0, 0.8)^T$, while for model (B), we consider $(1.4, 0.4)^T$, $(2.0, 0.4)^T$ and $(1.0, 0.35)^T$. For a value of $\hat{\underline{\theta}}$, we find two observations \underline{y}_i ($i = 1, 2$) such that the rotated coordinates are

$$\underline{z}_i = (0, 0, \frac{s_i}{2}, \frac{s_i}{2}, \frac{s_i}{2}, \frac{s_i}{2})^T,$$

where $s_1 = \sqrt{0.01}$, $s_2 = \sqrt{0.02}$ in model (A), and $s_1 = \sqrt{0.0001}$, $s_2 = \sqrt{0.0002}$ in model (B). With the \underline{y}_i , this value of $\hat{\underline{\theta}}$ is a least squares estimate of $\underline{\theta}$ and the residual sums of squares are s_i^2 . We then set $\alpha = 0.05$ and apply the various methods to obtain region estimates of $\underline{\theta}$. The boundaries of these regions are shown in Fig. (5.3.1)-(5.3.12).

FIGURE (5.3.1)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\theta_1 / (\theta_1 - \theta_2)) \times (\exp(-\theta_2 \cdot X) - \exp(-\theta_1 \cdot X))$

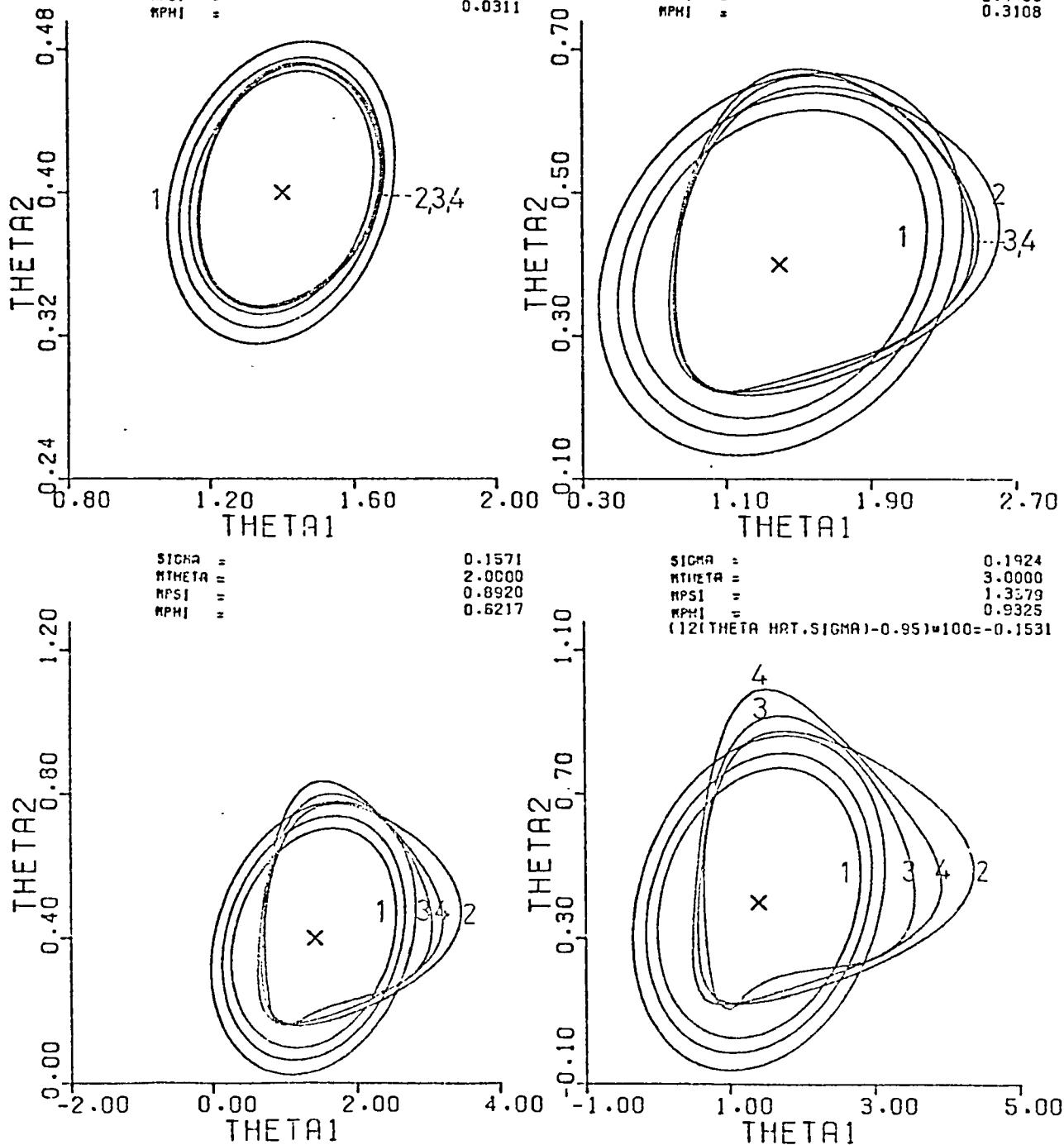
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\theta}_1 = 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.0100

SIGMA =
 $\hat{\theta}_1$ =
 $\hat{\theta}_2$ =
 $\hat{\rho}_{11}$ =

0.0351
 0.1000
 0.0446
 0.0311

SIGMA =
 $\hat{\theta}_1$ =
 $\hat{\theta}_2$ =
 $\hat{\rho}_{11}$ =

0.1111
 1.0000
 0.4460
 0.3108

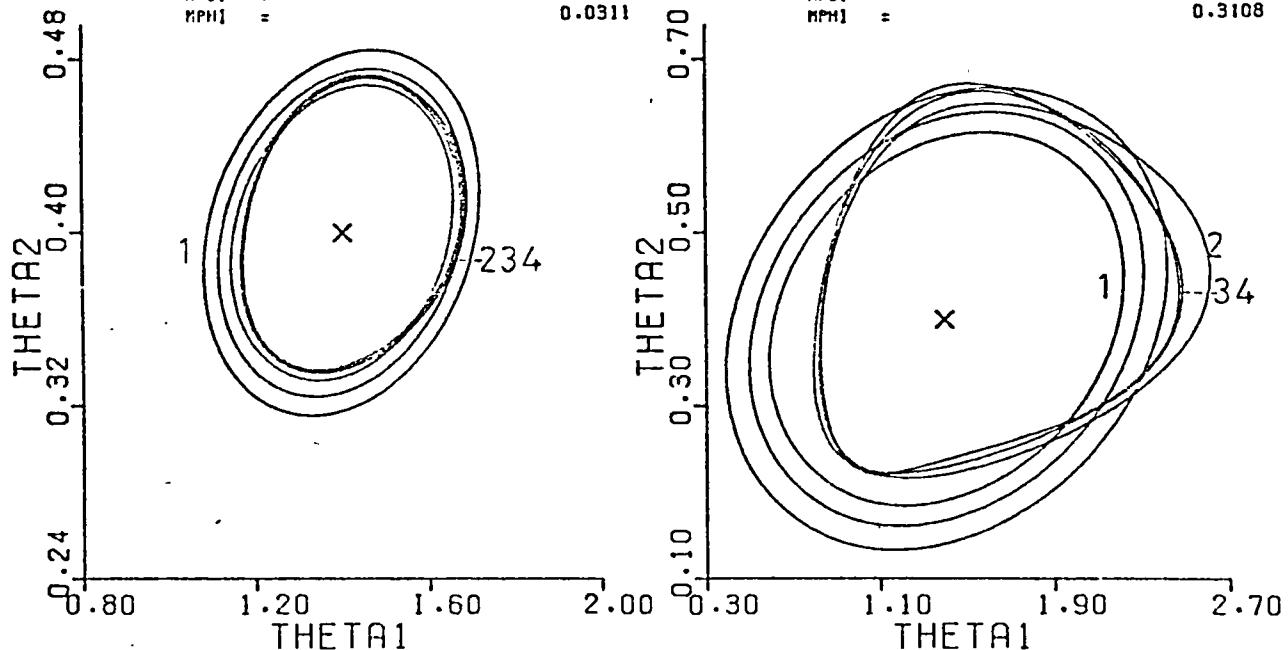


{ } : REGION ESTIMATES (NOMINAL 95.5.99 PERCENT) GIVEN BY METHOD I
 { } : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I (i=2,3,4)

FIGURE (5.3.2)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

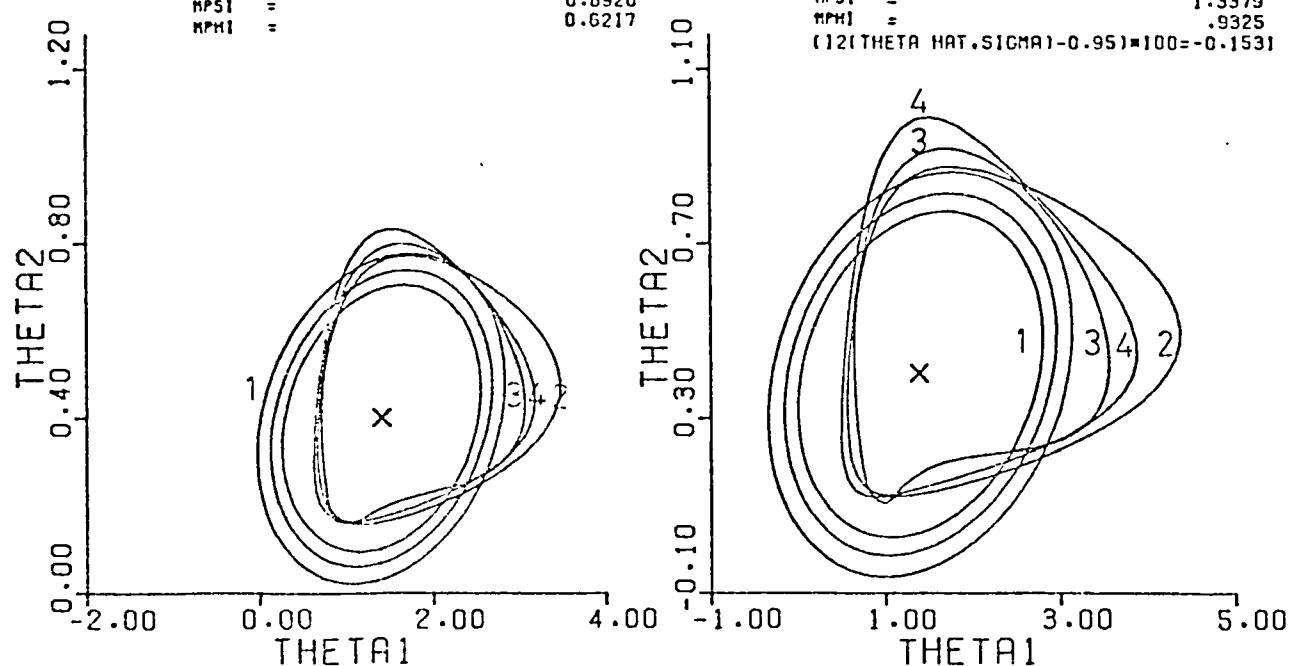
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1$ ARE 1.4000 0.4000
 RESIDUAL SUM OF SQUARES = 0.0200

SIGMA =	0.0351	SIGMA =	0.1111
MTHETA =	0.1000	MTHETA =	1.0000
MPSI =	0.0446	MPSI =	0.4460
MPHI =	0.0311	MPHI =	0.3108



SIGMA =	0.1571	SIGMA =	0.1924
MTHETA =	2.0000	MTHETA =	3.0000
MPSI =	0.8920	MPSI =	1.3379
MPHI =	0.6217	MPHI =	.9325

(12(THETA HAT, SIGMA)-0.95)*100=-0.1531



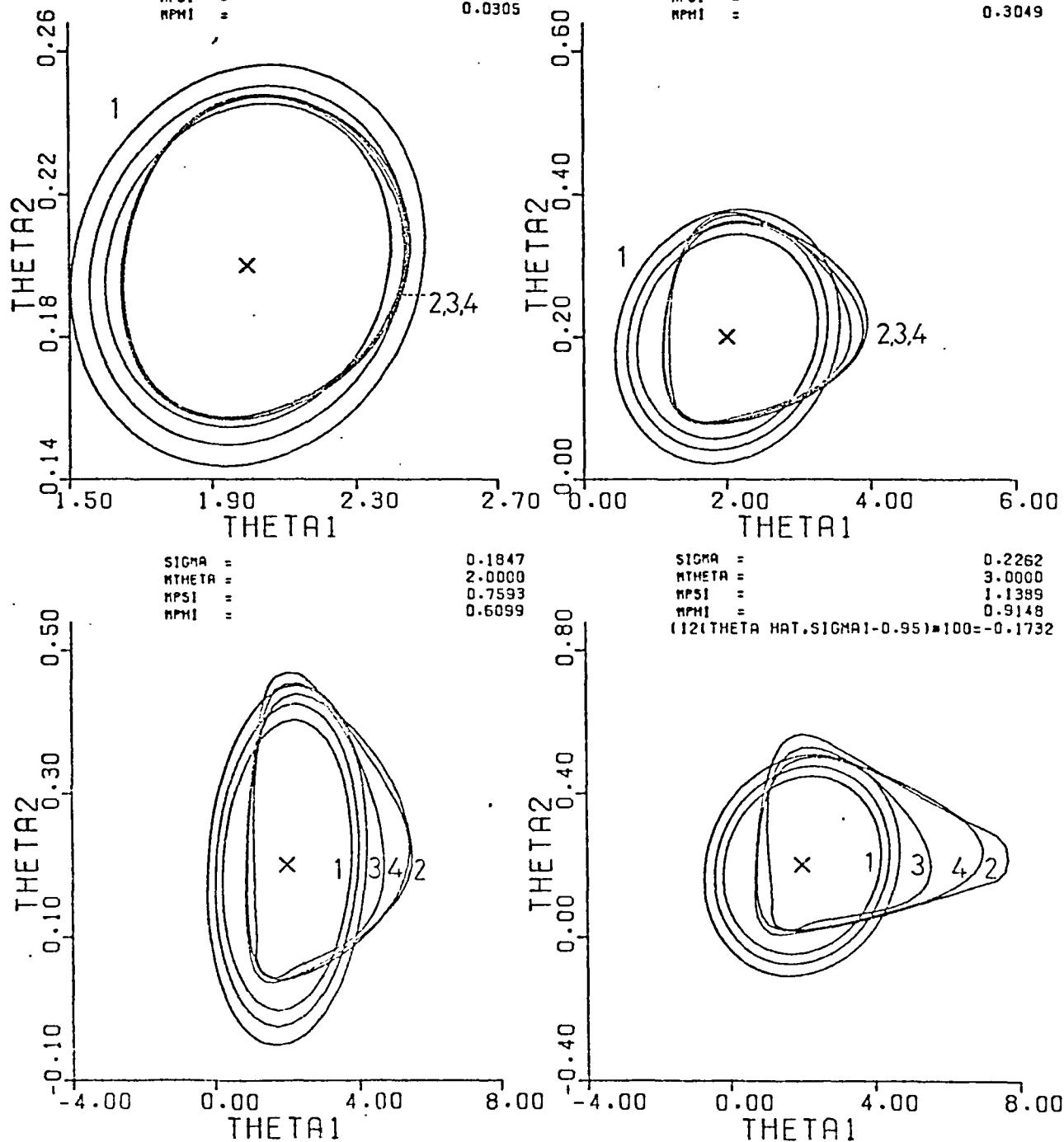
{1} : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD 1
 {1} : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD 1 (I=2.3.4)

FIGURE (5.3.3)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X)) - \exp(-\Theta_1 \cdot X))$

$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1$ ARE 2.0000 0.2000
 RESIDUAL SUM OF SQUARES = 0.0100

SIGMA = 0.0413
 MTHETA = 0.1000
 MPSI = 0.0380
 MPHI = 0.0305

SIGMA = 0.1306
 MTHETA = 1.0000
 MPSI = 0.3796
 MPHI = 0.3049



(1) : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD I
 (2) : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=2.3.4)

FIGURE (5.3.4)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$

$\hat{\Theta}_1$ ARE 2.0000 0.2000

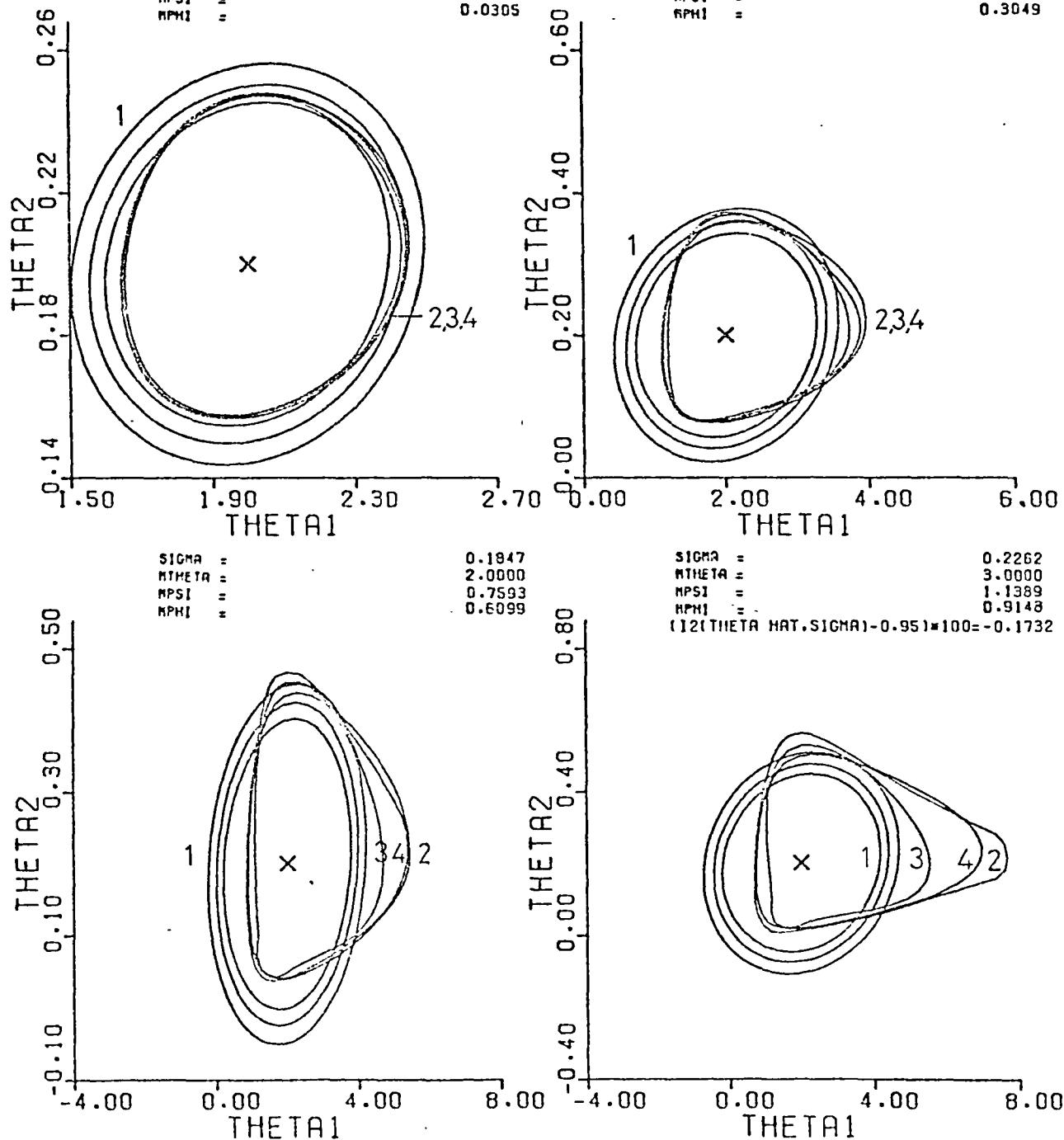
RESIDUAL SUM OF SQUARES = 0.0200

SIGMA =
 $\hat{\Theta}_{\text{THETA}}$ =
 $\hat{\Theta}_{\text{PSI}}$ =
 $\hat{\Theta}_{\text{PHI}}$ =

0.0413
 0.1000
 0.0380
 0.0305

SIGMA =
 $\hat{\Theta}_{\text{THETA}}$ =
 $\hat{\Theta}_{\text{PSI}}$ =
 $\hat{\Theta}_{\text{PHI}}$ =

0.1306
 1.0000
 0.3796
 0.3049



{1} : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD 1
 {1} : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD 1 ({1}=2,3,4)

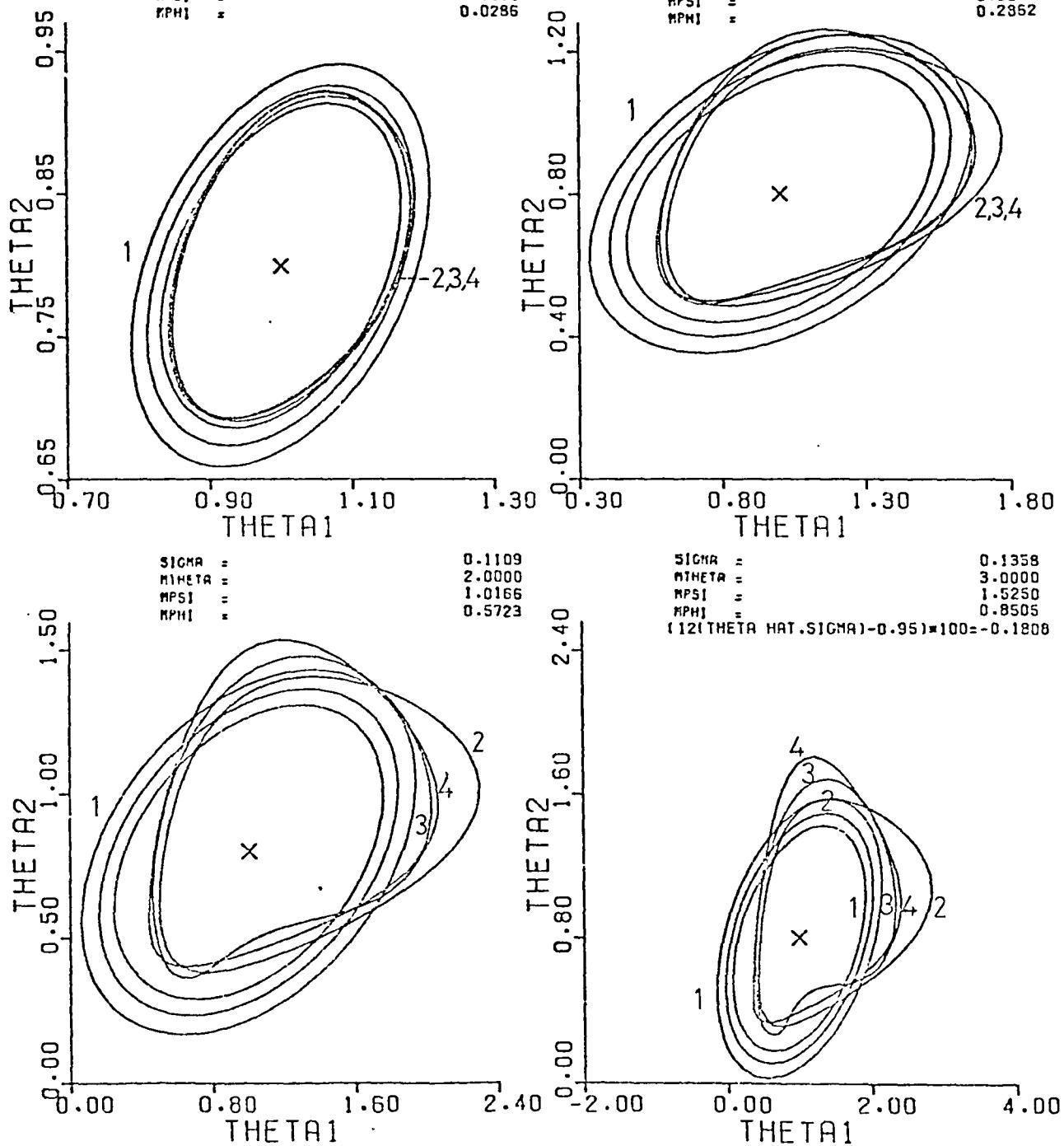
FIGURE (5.3.5)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

$X = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\Theta_1 \text{ HAT ARE } 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.0100

SIGMA = 0.0248
 $\Theta_1 \text{ HAT} = 1.0000$
 $\Theta_2 \text{ HAT} = 0.5083$
 $\Theta_3 \text{ HAT} = 0.2866$

SIGMA = 0.0784
 $\Theta_1 \text{ HAT} = 1.0000$
 $\Theta_2 \text{ HAT} = 0.5083$
 $\Theta_3 \text{ HAT} = 0.2866$

SIGMA = 0.1358
 $\Theta_1 \text{ HAT} = 3.0000$
 $\Theta_2 \text{ HAT} = 1.5250$
 $\Theta_3 \text{ HAT} = 0.8505$



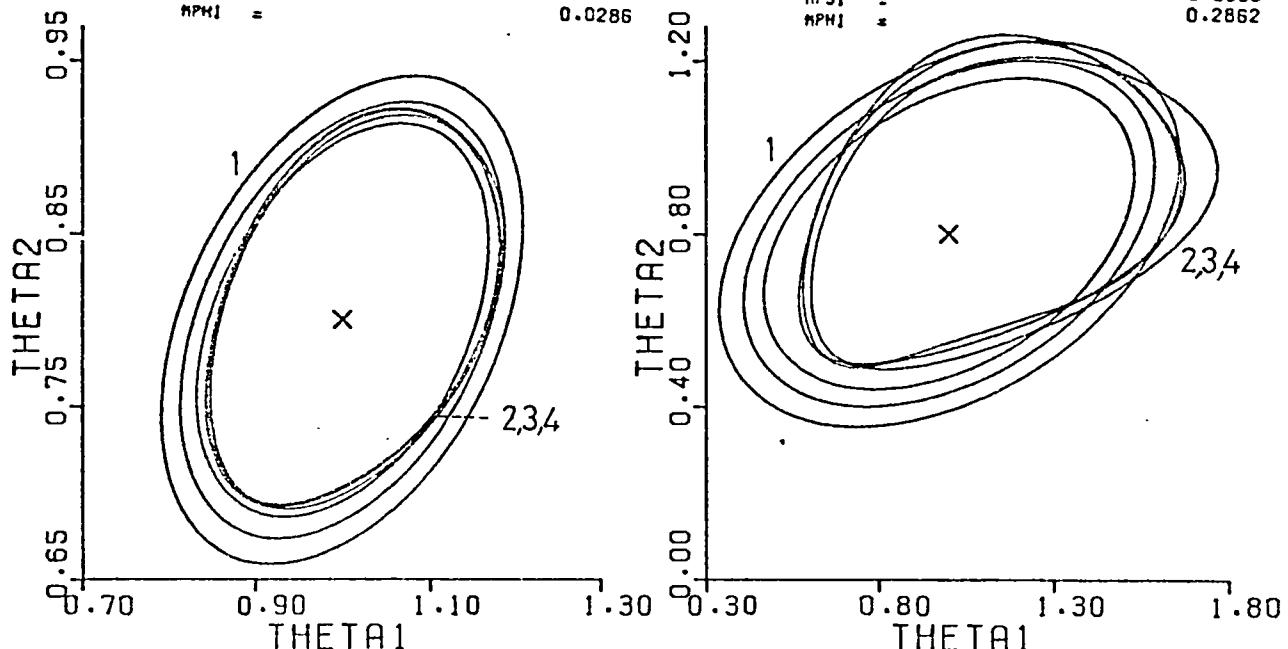
(1) : REGION ESTIMATES (NOMINAL 95.99 PERCENT) GIVEN BY METHOD I
 (1) : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I (1=2.3.4)

FIGURE (5.3.6)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

$X_i = 0.25, 0.5, 1, 0.1, 1.5, 2, 0.4, 0$
 $\Theta_1 \text{ HAT ARE } 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.0200

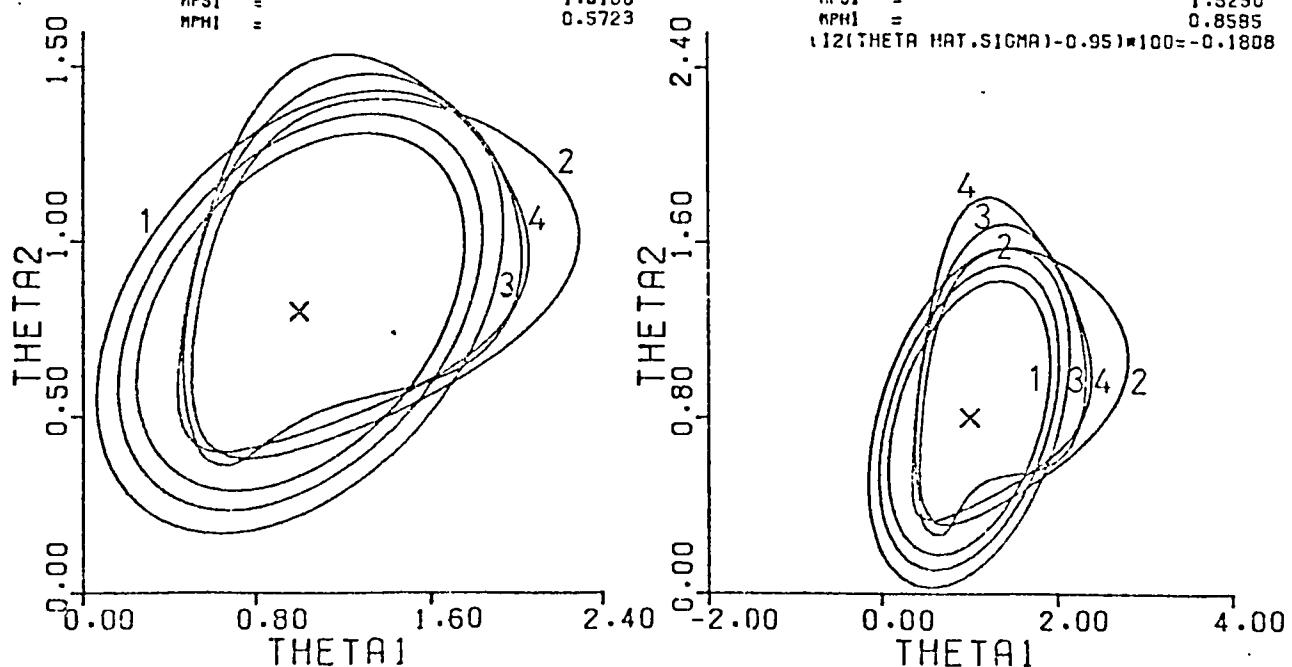
SIGMA = 0.0248
 MTTHETA = 0.1000
 MPSI = 0.0506
 MPH1 = 0.0286

SIGMA = 0.0784
 MTTHETA = 1.0000
 MPSI = 0.5083
 MPH1 = 0.2862



SIGMA = 0.1103
 MTTHETA = 2.0000
 MPSI = 1.0166
 MPH1 = 0.5723

SIGMA = 0.1358
 MTTHETA = 3.0000
 MPSI = 1.5250
 MPH1 = 0.8585
 $(12(\Theta_1 \text{ HAT}, \text{SIGMA}) - 0.95) * 100 = -0.1808$



(1) : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD 1
 (1) : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD 1 (1=2,3,4)

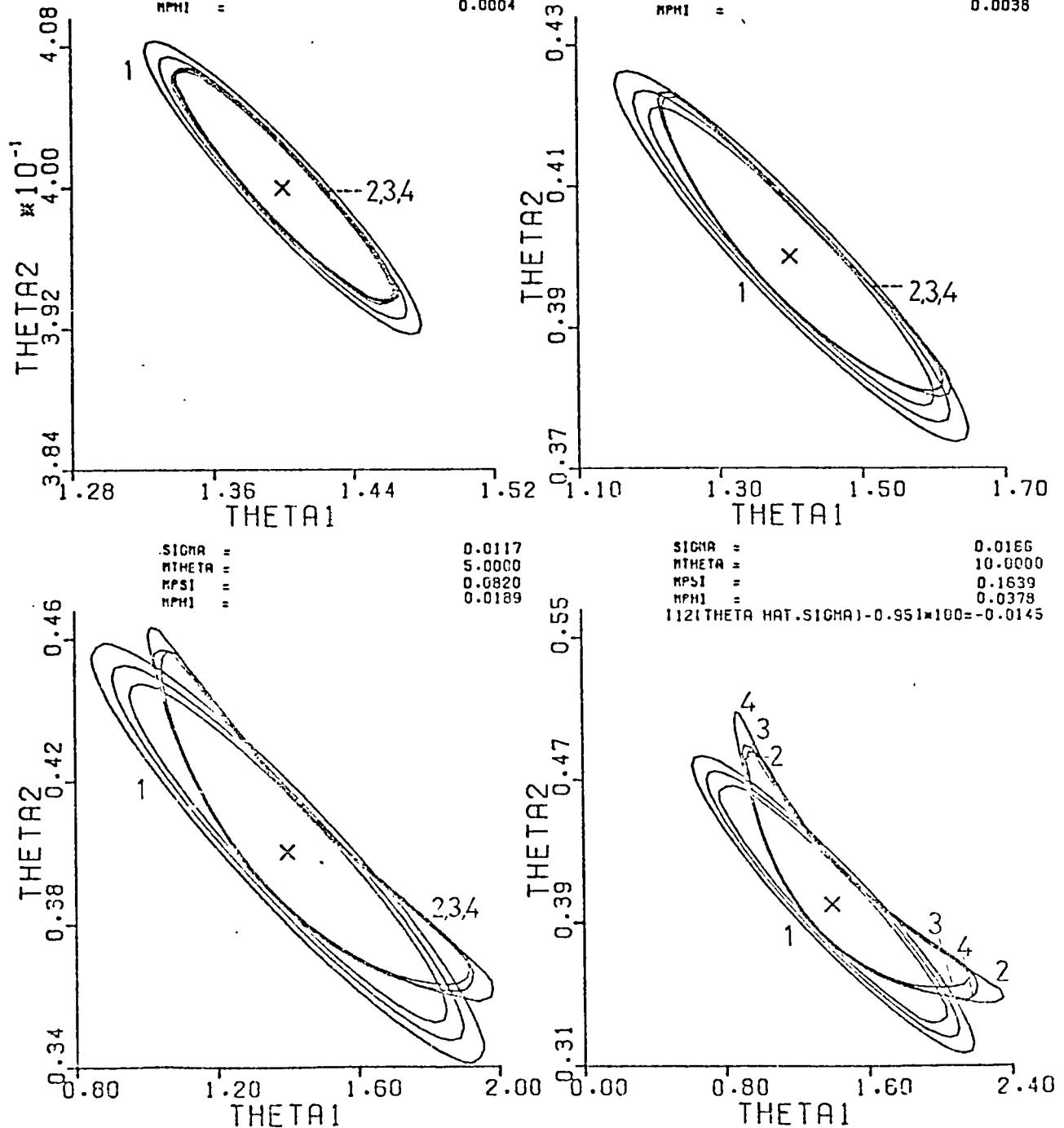
FIGURE (5.3.7)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = 1 - [\Theta_1 \exp(-\Theta_2 \cdot X_i) - \Theta_2 \exp(-\Theta_1 \cdot X_i)] / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\hat{\Theta}_1 \text{ HAT ARE } 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.0001

SIGMA =
 $\hat{\Theta}_1$ =
 $\hat{\Theta}_2$ =
 $\hat{\Theta}_{12}$ =

0.0017
 0.1000
 0.0016
 0.0004

SIGMA =
 $\hat{\Theta}_1$ =
 $\hat{\Theta}_2$ =
 $\hat{\Theta}_{12}$ =

0.0053
 1.0000
 0.0164
 0.0038



111 : REGION ESTIMATES (INCIMAL 95.97.5.99 PERCENT) GIVEN BY METHOD I
 111 : REGION ESTIMATES (INCIMAL 95 PERCENT) GIVEN BY METHOD I (I=2,3,4)

FIGURE (5.3.8)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X_1) - \Theta_2 \cdot \exp(-\Theta_1 \cdot X_1)) / (\Theta_1 - \Theta_2)$
 $X_1 = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \cdot \hat{\Theta} \text{ ARE } 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.0002

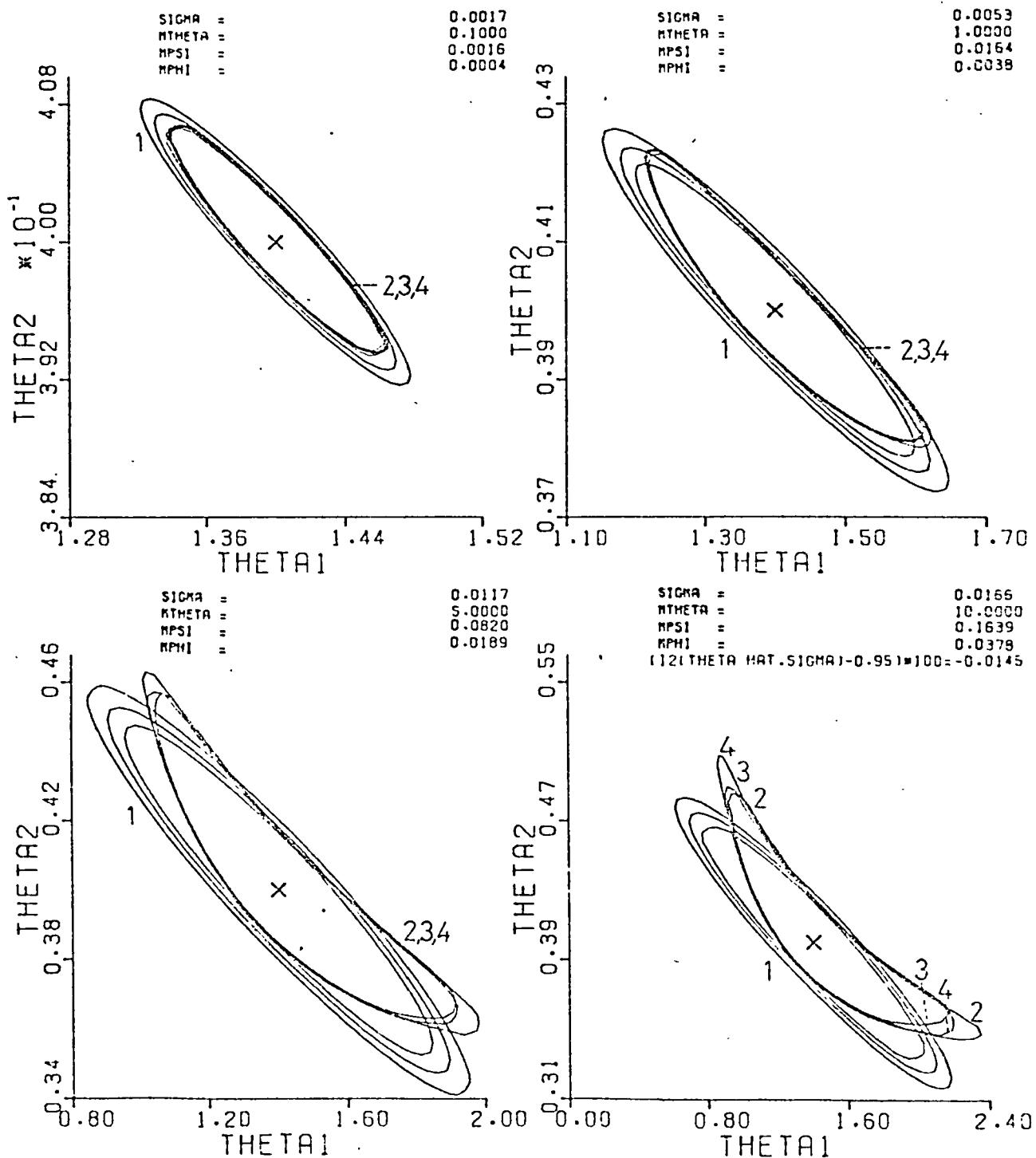


FIGURE (5.3.9)

REGION ESTIMATES IN THE MODEL

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) / (\Theta_1 - \Theta_2)$$

 $X_i = 1, 2, 3, 4, 5, 6$

THETA1 HAT ARE 2.0000 0.4000

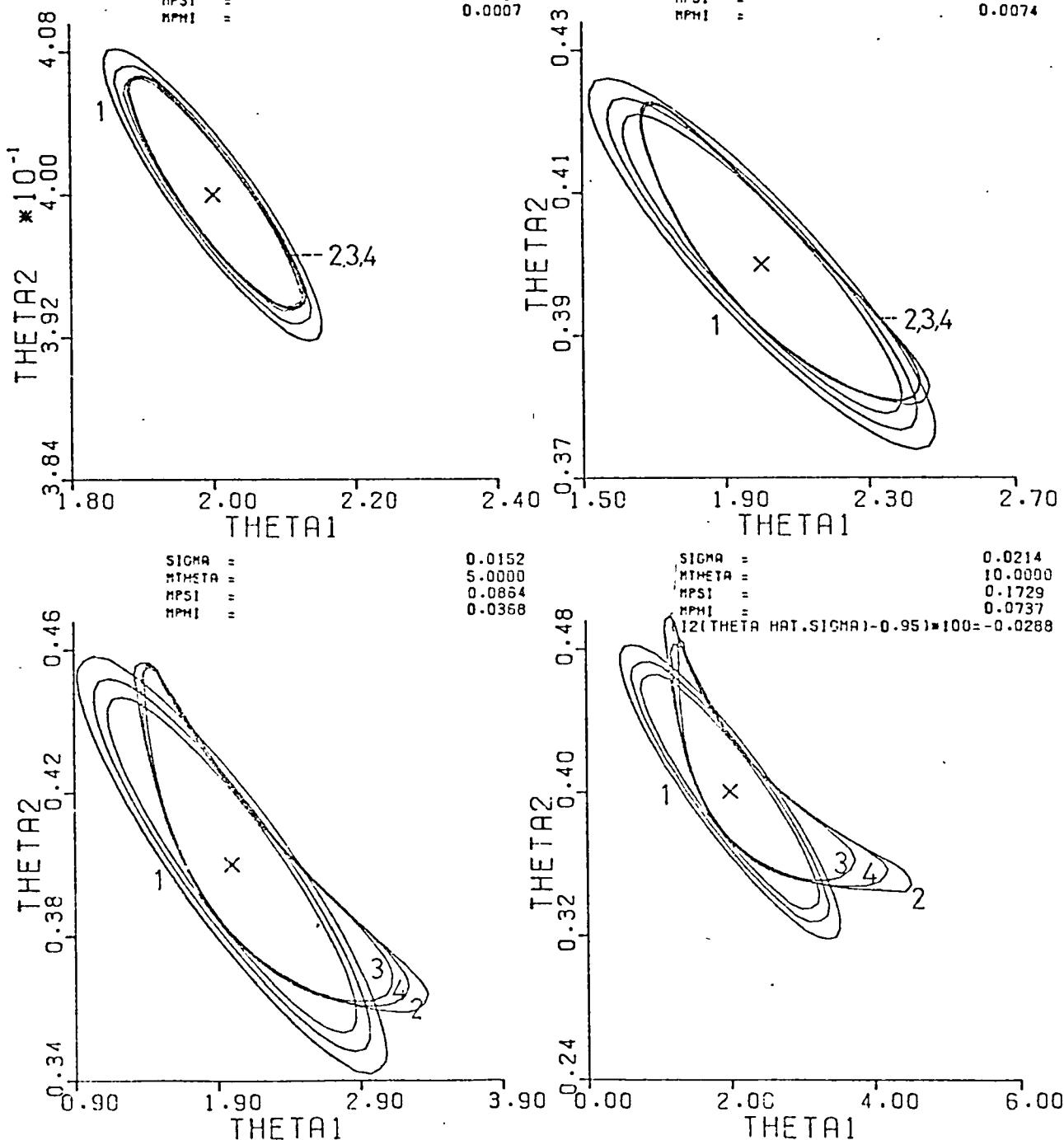
RESIDUAL SUM OF SQUARES = 0.0001

SIGMA =
 MTHETA =
 MPSI =
 MPH1 =

0.0021
 0.1000
 0.0017
 0.0007

SIGMA =
 MTHETA =
 MPSI =
 MPH1 =

0.0068
 1.0000
 0.0173
 0.0074



111 : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD I
 111 : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I (1)=2.3.41

FIGURE (5.3.10)

REGION ESTIMATES IN THE MODEL

$$E(Y) = 1 - (\theta_1 \exp(-\theta_2 x_i) - \theta_2 \exp(-\theta_1 x_i)) / (\theta_1 - \theta_2)$$

 $x_i = 1, 2, 3, 4, 5, 6$ $\theta_1 \text{ HAT ARE } 2.0000 \quad 0.4000$

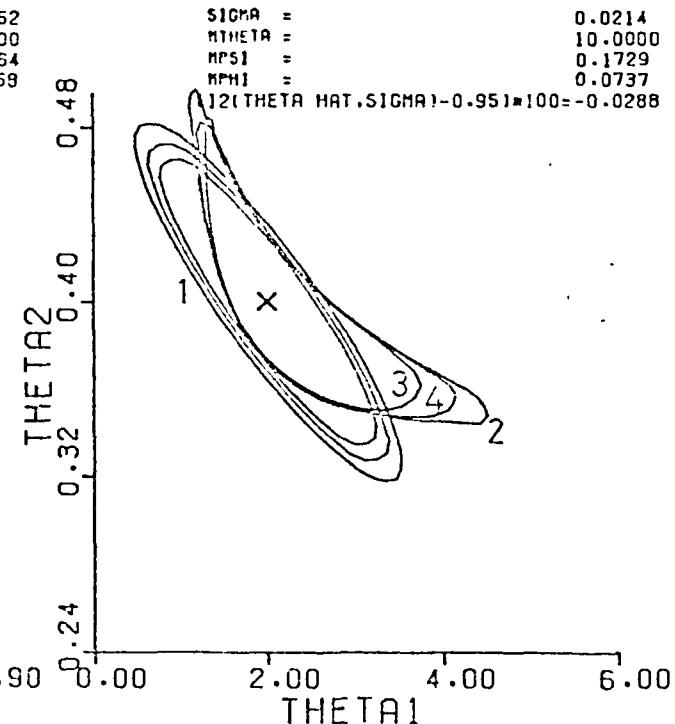
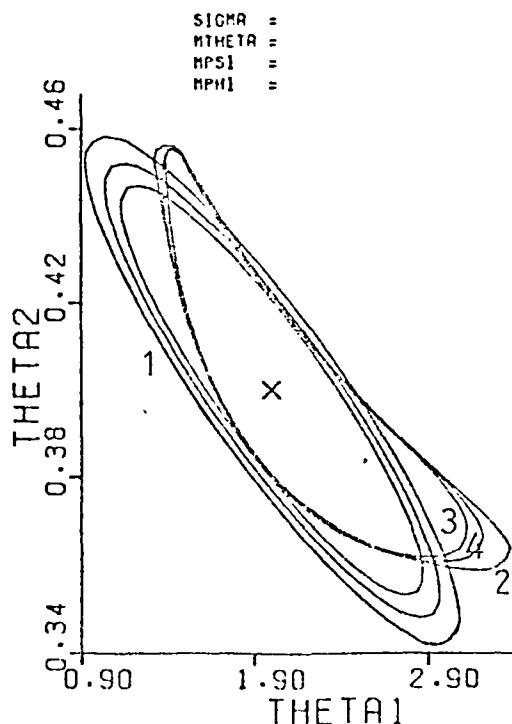
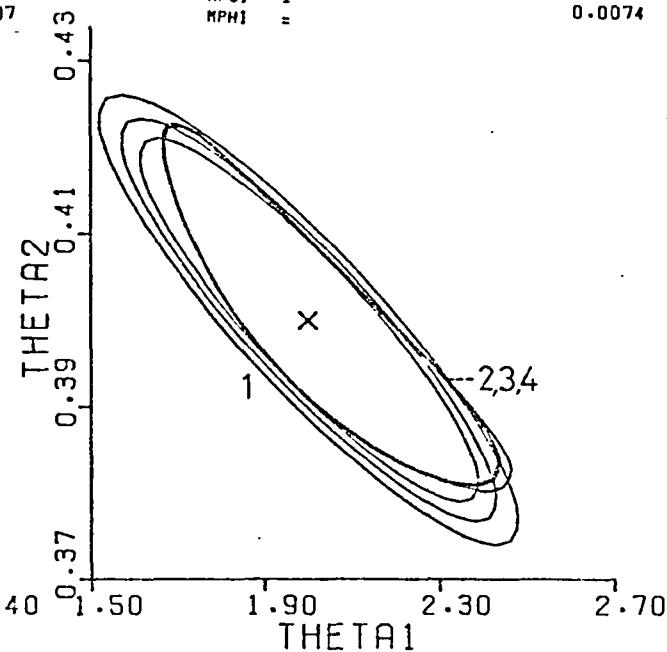
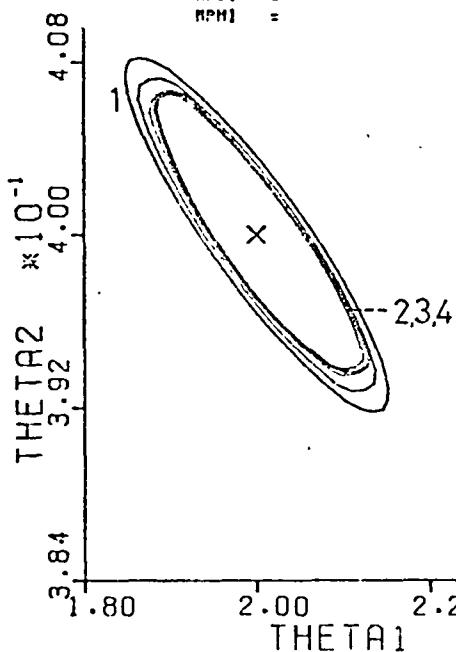
RESIDUAL SUM OF SQUARES = 0.0002

SIGMA =
 θ_1 =
 θ_2 =
 θ_{12} =

0.0021
0.1000
0.0017
0.0007

SIGMA =
 θ_1 =
 θ_2 =
 θ_{12} =

0.0068
1.0000
0.0173
0.0074



(1) : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD I
(1) : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=2,3,4$)

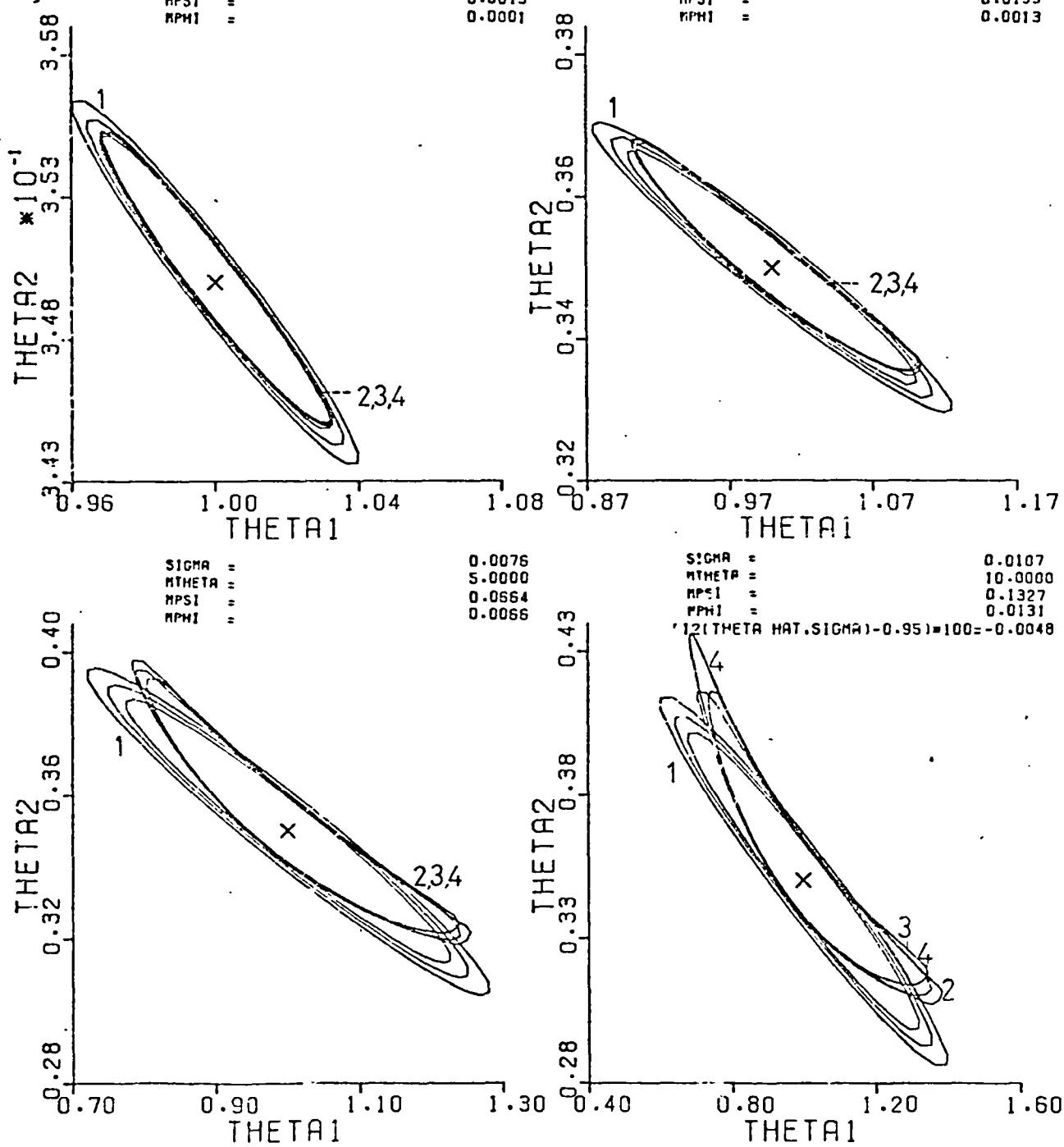
FIGURE (5.3.11)
 REGION ESTIMATES IN THE MODEL
 $E(Y) = 1 - \theta_1 \exp(-\theta_2 x_i) - \theta_2 \exp(-\theta_1 x_i) / (\theta_1 - \theta_2)$
 $x_i = 1, 2, 3, 4, 5, 6$
 $\theta_1 \text{ HAT ARE } 1.0000 \quad 0.3500$
 $\text{RESIQUAL SUM OF SQUARES} = 0.0001$

SIGMA =
 θ_1 =
 θ_2 =
 μ_{θ_1} =

0.0011
 0.1000
 0.0013
 0.0001

SIGMA =
 θ_1 =
 θ_2 =
 μ_{θ_1} =

0.0034
 0.0000
 0.0133
 0.0013



{1} : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD 1
 {1} : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD 1 [i=2.3.4]

FIGURE (5.3.12)

REGION ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $- (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $/(\Theta_1 - \Theta_2)$

$X_i = 1, 2, 3, 4, 5, 6$

Θ_1 HAT ARE 1.0000 0.3500

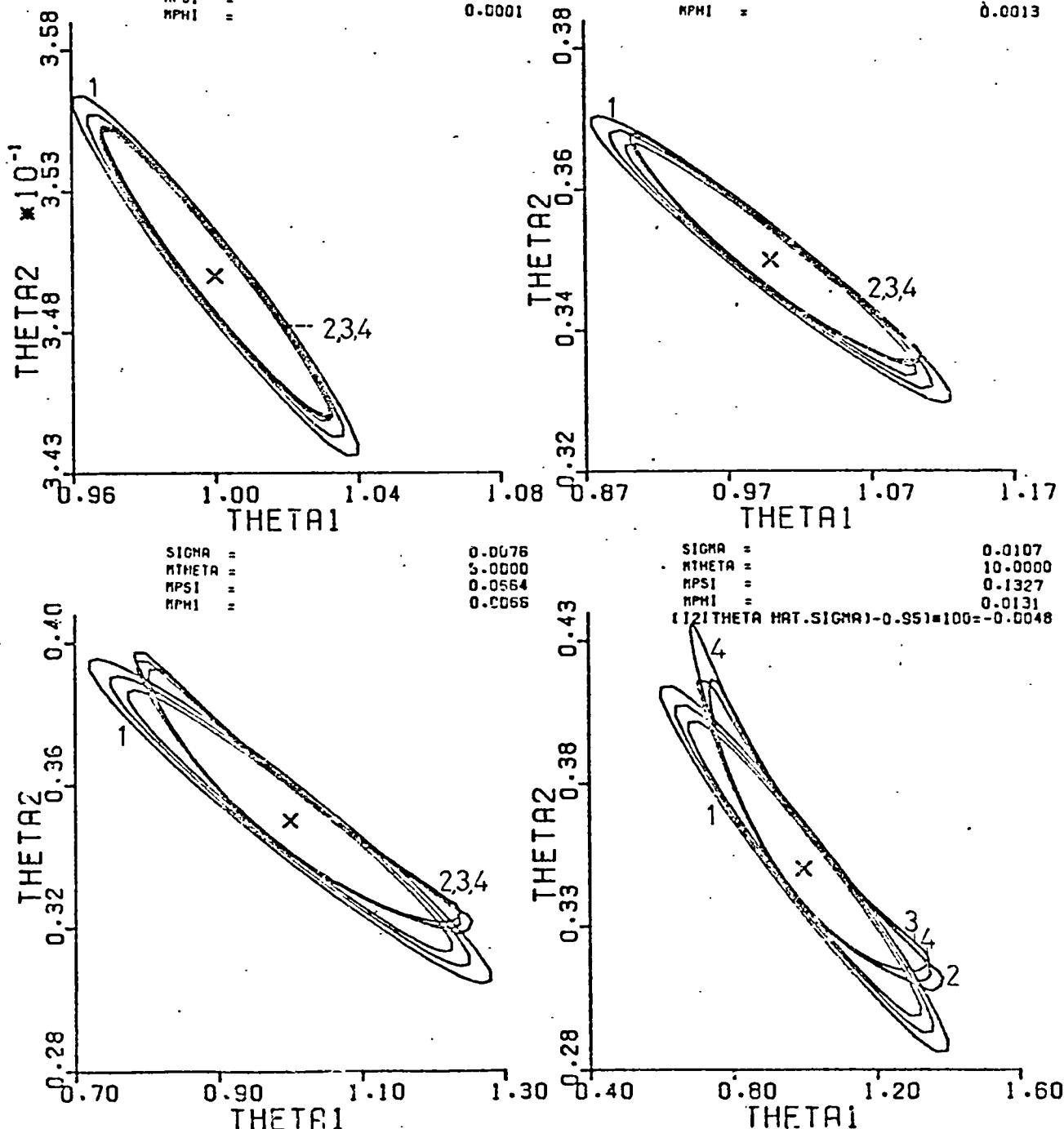
RESIDUAL SUM OF SQUARES = 0.0002

SIGMA =
 Θ_1 HAT =
 Θ_2 HAT =
 Θ_1 =
 Θ_2 =

0.0011
0.1000
0.0013
0.0001

SIGMA =
 Θ_1 HAT =
 Θ_2 HAT =
 Θ_1 =
 Θ_2 =

0.0034
1.0000
0.0133
0.0013



111 : REGION ESTIMATES (NOMINAL 95.97.5.99 PERCENT) GIVEN BY METHOD I
111 : REGION ESTIMATES (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=2,3,4)

We note that for each of the $\hat{\theta}$, M_{θ} is very much larger than $|J_2(\hat{\theta}, \sigma)|$ while M_{ψ} and M_{ϕ} are fairly close to this number. Furthermore as shown in the figures, the order given by (5.2.7) with $\theta_A = \hat{\theta}$ is fairly well preserved in the closeness of the region estimates given by methods 1, 2 and 3 to that given by method 4. We also note that the change in the residual sum of squares of an observation from s_1^2 to s_2^2 has very slight effect on the region estimate given by method 4.

We next observe that at these levels of σ , the absolute values of the differences between $J_2(\hat{\theta}, \sigma)$ and $J_2(\theta_T, \sigma)$ are small fractions of a percent. This implies that our choice of the values of σ , $\hat{\theta}$ and θ_T has resulted in situations in which we can refer to the method 4 region estimates based on these values of $\hat{\theta}$ as approximately 100 $I_2(\hat{\theta}, \sigma)\%$ region estimates.

We further observe that the four methods give almost identical region estimates when $M_{\theta} = 0.1$. As the region estimates given by method 4 are approximately 95% region estimates, the region estimates given by methods 1, 2 and 3 are also approximately 95% region estimates.

When $M_{\theta} = 1$, methods 1 and 4 give slightly different region estimates, and whenever the corresponding M_{ψ} and M_{ϕ} are less than 0.1, methods 2, 3 and 4 give almost identical region estimates. This observation is consistent with the observation in the case when $M_{\theta} = 0.1$.

When $M_{\theta} > 1$, the region estimates given by methods 1 and 4 are fairly significantly different. The region estimates given by method 2 or 3 differ slightly from those given by method 4 whenever M_{ψ} or M_{ϕ} is less than 1. This observation is consistent with the observation in the case when $M_{\theta} = 1$.

The above observations indicate that for these models with θ_T near $(1.4, 0.4)^T$ and the levels of σ similar to those considered before, a region estimate of the parameter vector θ based on method 1, 2, or 3

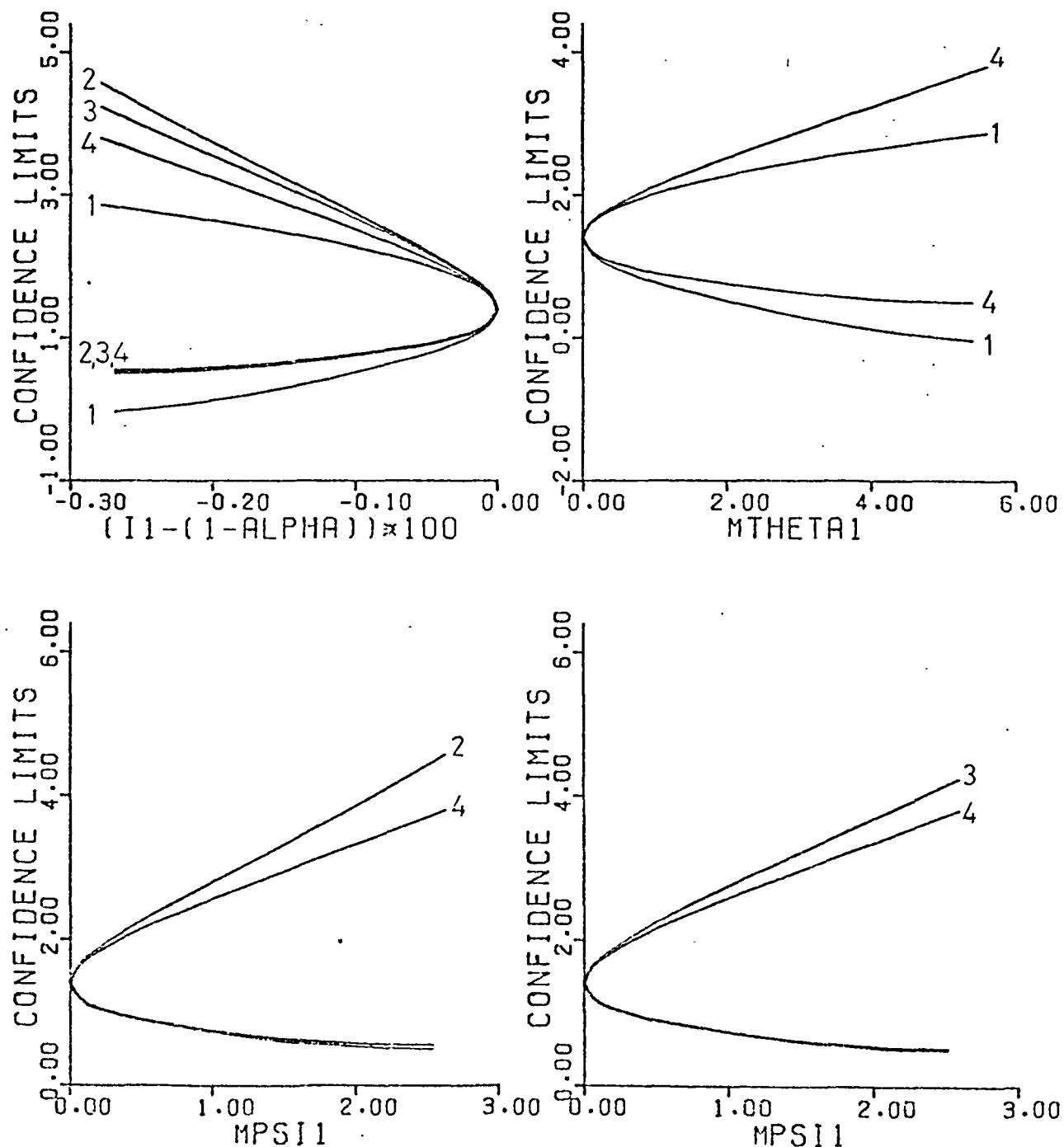
is an approximately 95% region estimate provided that the value of the corresponding nonlinearity M_β , where $\beta = \theta, \psi$ or ϕ , is less than or equal to 0.1.

Section 5.4 Interval estimates of θ_i

In this section we compare the limits of the interval estimates given by the four methods. The values of θ_T , $\hat{\theta}_i$, y_i , s_i and α are chosen to be the same as in section 5.3. The various methods are then applied to obtain interval estimates of θ_i . The limits of these intervals are shown in Fig. (5.4.1)-(5.4.24).

FIGURE (5.4.1)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\theta_1 / (\theta_1 - \theta_2))$
 $\times (\exp(-\theta_2 \cdot X) - \exp(-\theta_1 \cdot X))$

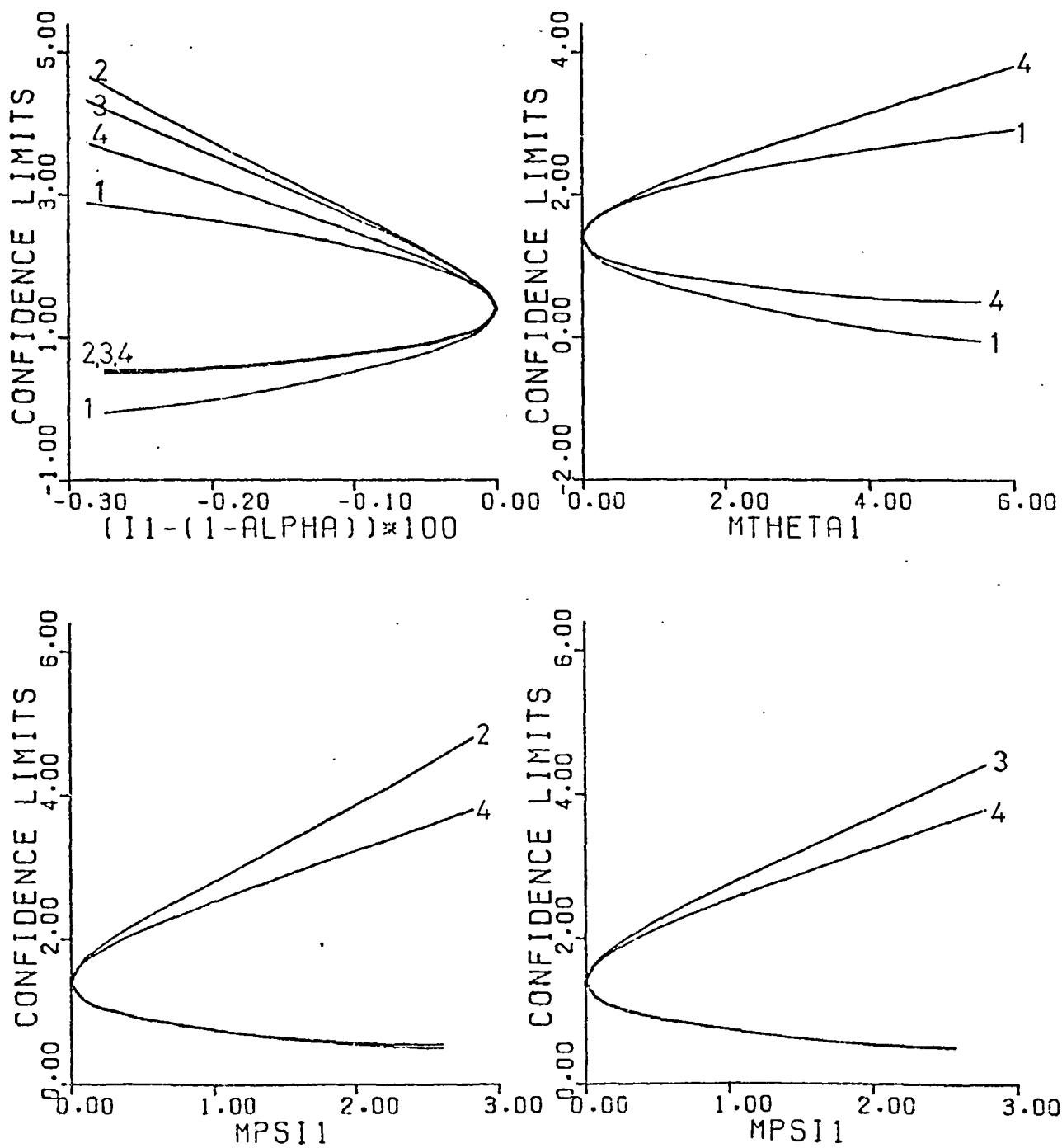
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\theta}_1 = 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.1300



$\{I_i\}$: INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1, 2, 3, 4$)

FIGURE (5.4.2)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$

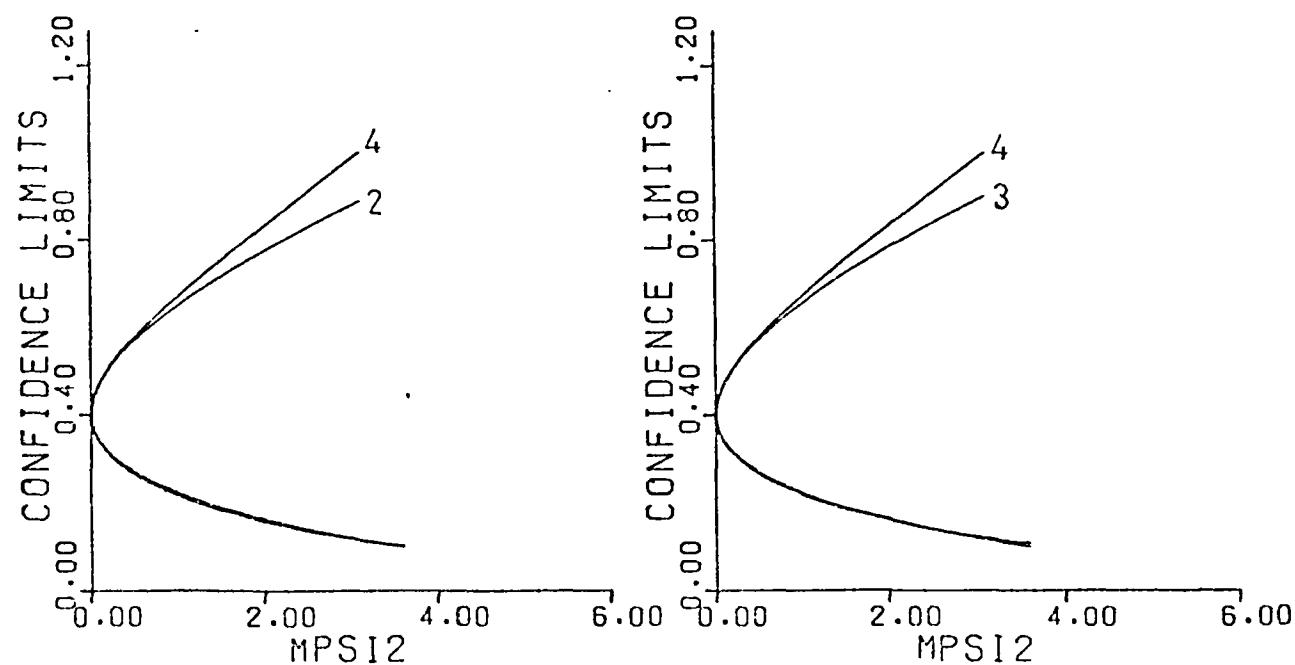
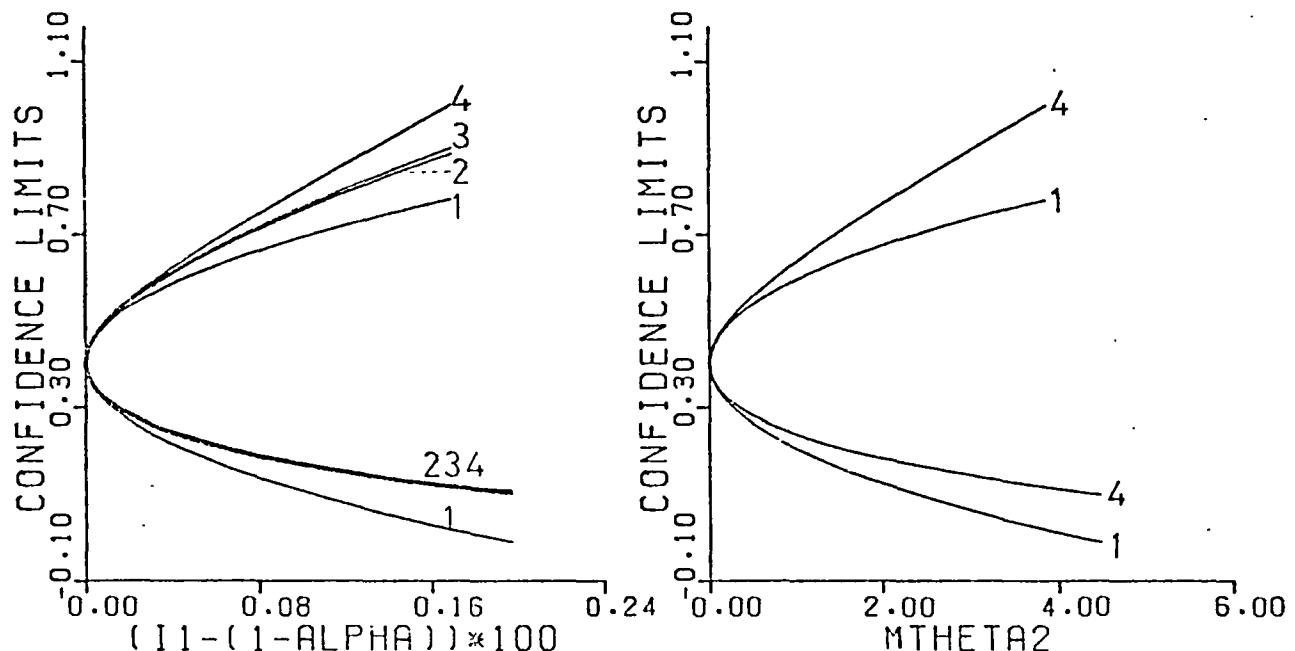
$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.2000



(11) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($11=1,2,3,4$)

FIGURE (5.4.3)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1)/(\Theta_1 - \Theta_2)$
 $\bullet (\exp(-\Theta_2 \cdot X)) - \exp(-\Theta_1 \cdot X))$

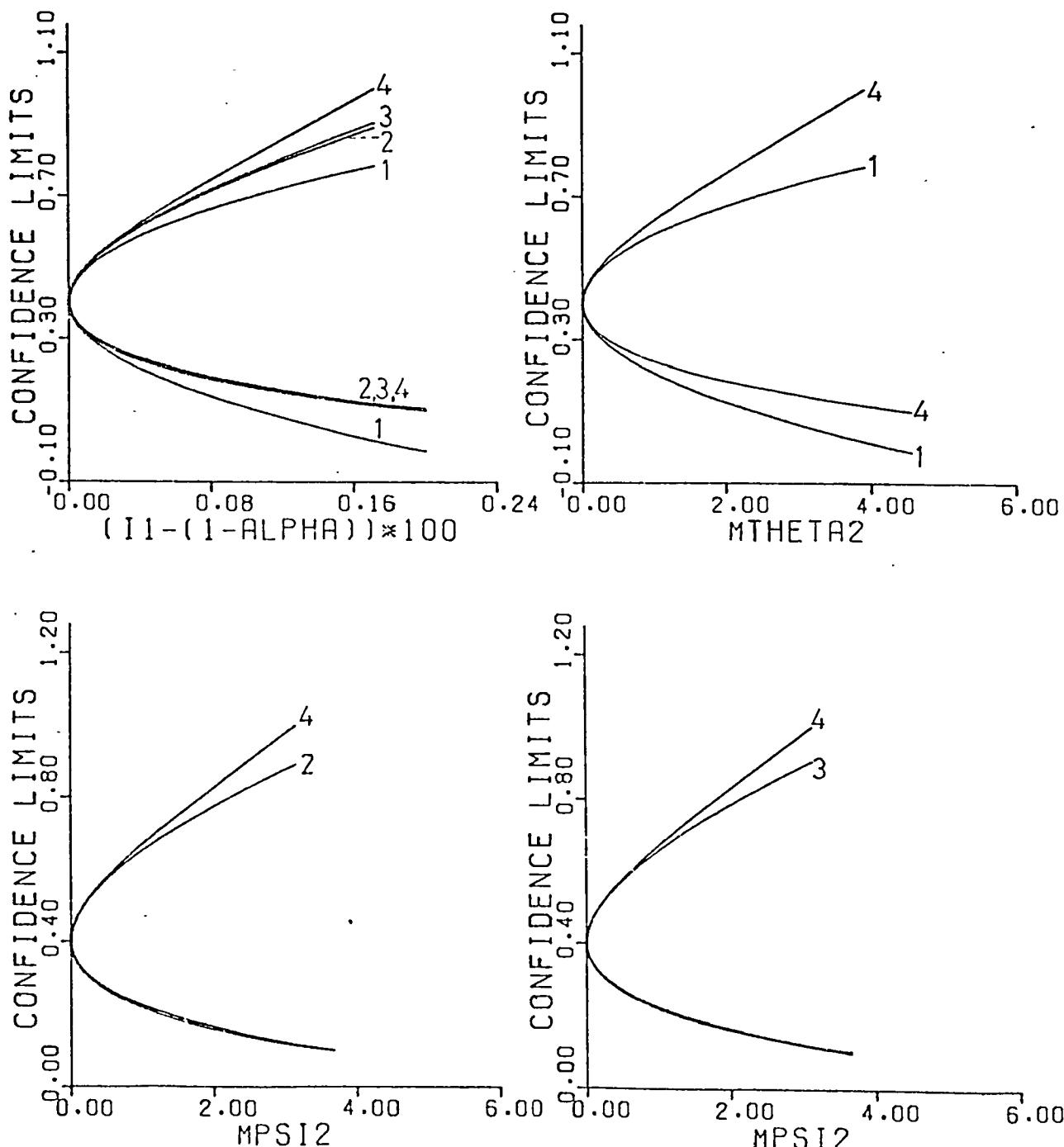
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 \text{ HAT ARE } 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.1000



111 : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.4)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\theta_1 / (\theta_1 - \theta_2))$
 $\times (\exp(-\theta_2 \cdot X_i) - \exp(-\theta_1 \cdot X_i))$

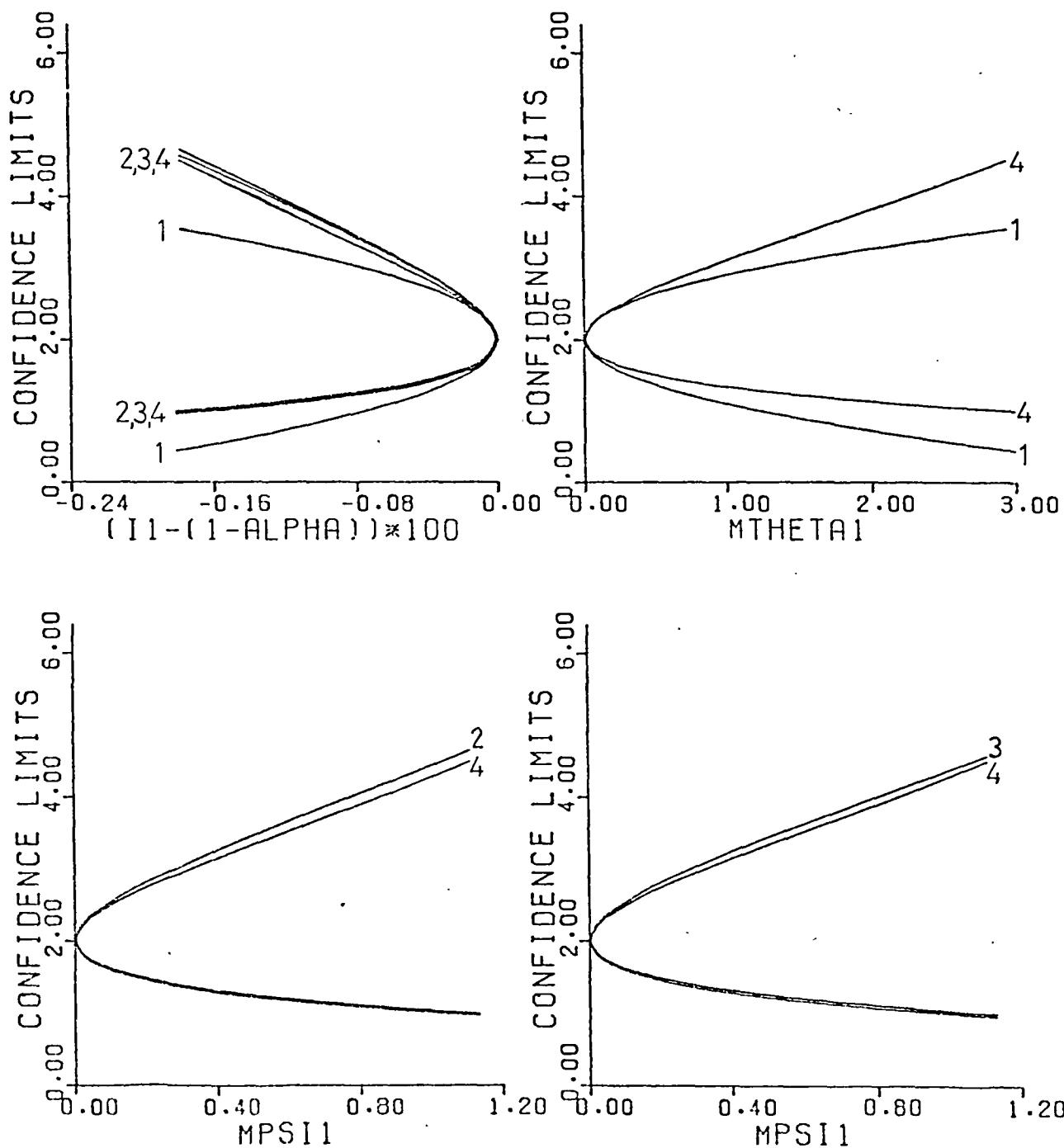
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\theta}_1 = 1.4000 \quad \hat{\theta}_2 = 0.4000$
 RESIDUAL SUM OF SQUARES = 0.2000



(\bar{Y}) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($I=1,2,3,4$)

FIGURE (5.4.5)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

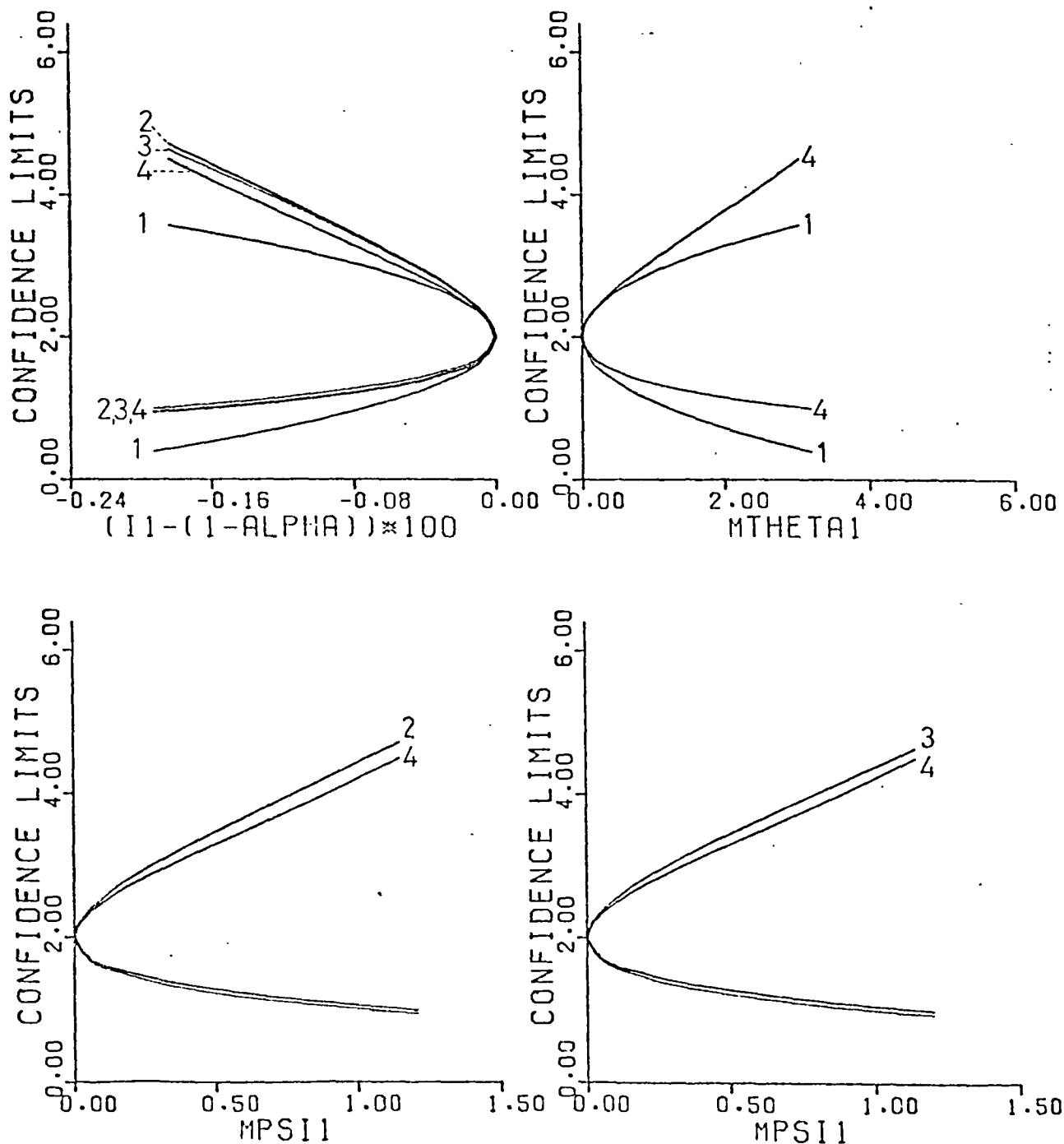
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 2.0000 \quad 0.2000$
 RESIDUAL SUM OF SQUARES = 0.1000



(I) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.6)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$

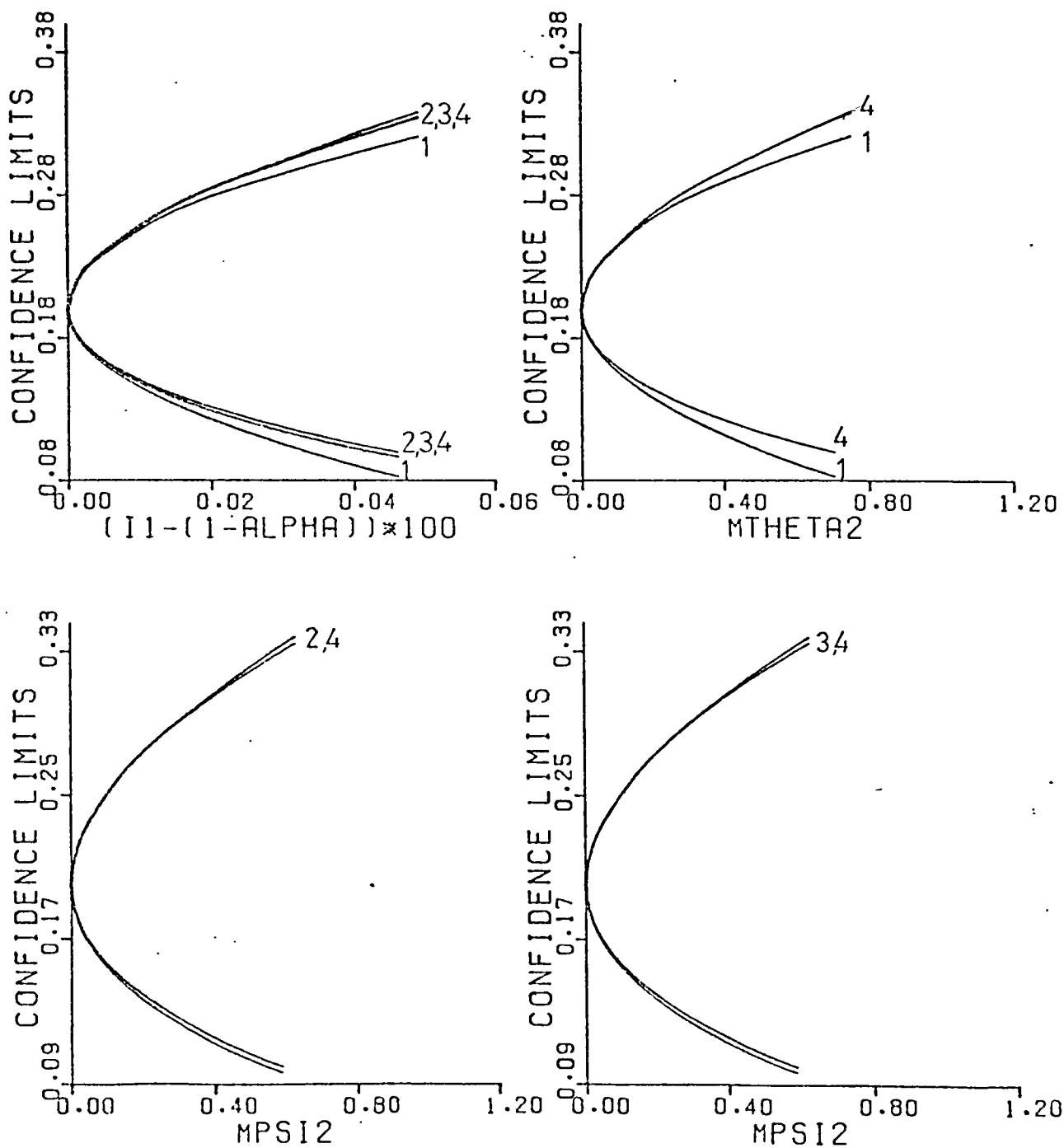
$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 2.0000 \quad 0.2000$
 RESIDUAL SUM OF SQUARES = 0.2000



(I) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.7)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = [\theta_1 / (\theta_1 - \theta_2)]$
 $\times (\exp(-\theta_2 \cdot X) - \exp(-\theta_1 \cdot X))$

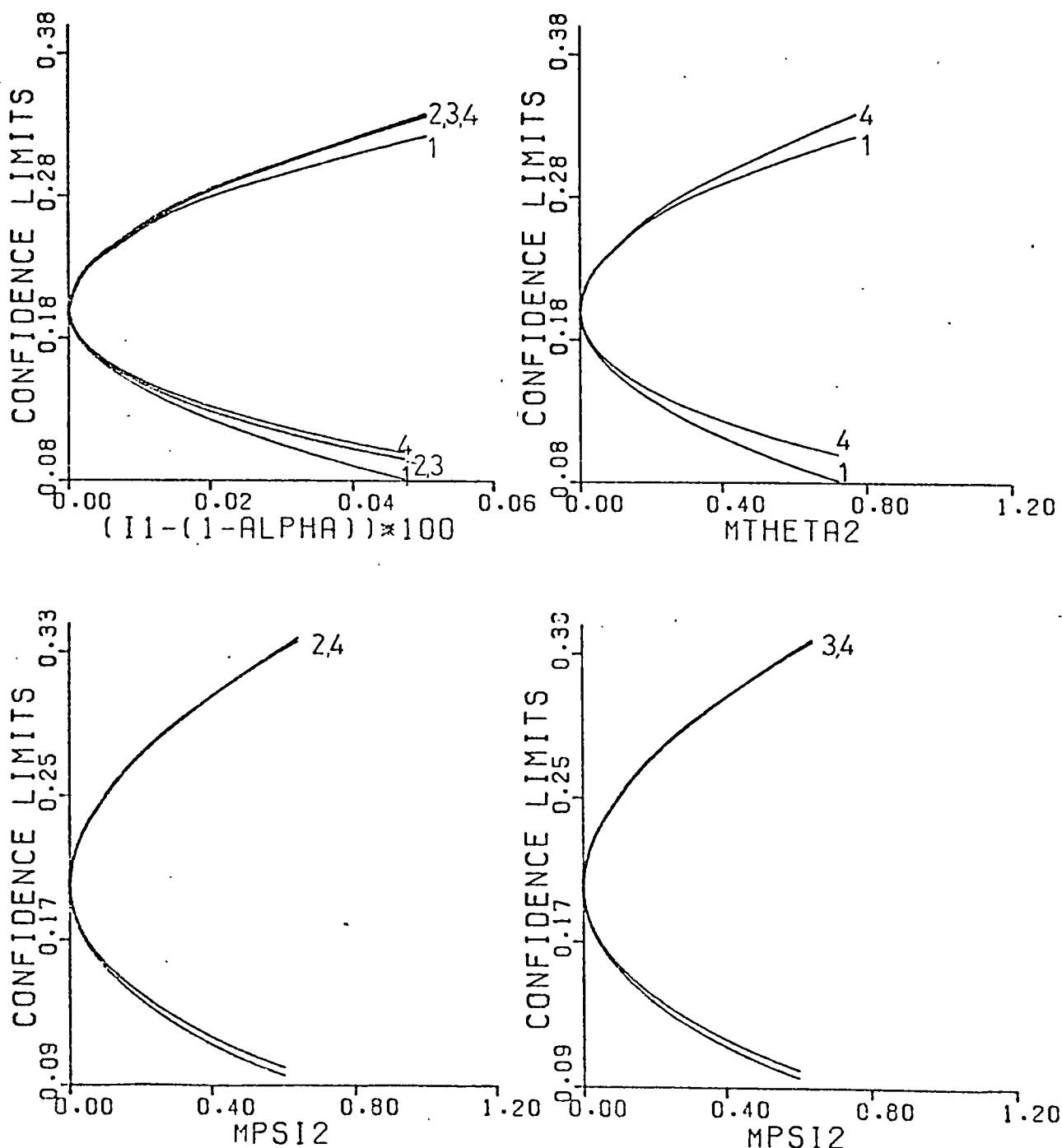
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\theta}_1 = 2.0000 \quad \hat{\theta}_2 = 0.2000$
 RESIDUAL SUM OF SQUARES = 0.1000



I_1 : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1,2,3,4$)

FIGURE (5.4.8)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

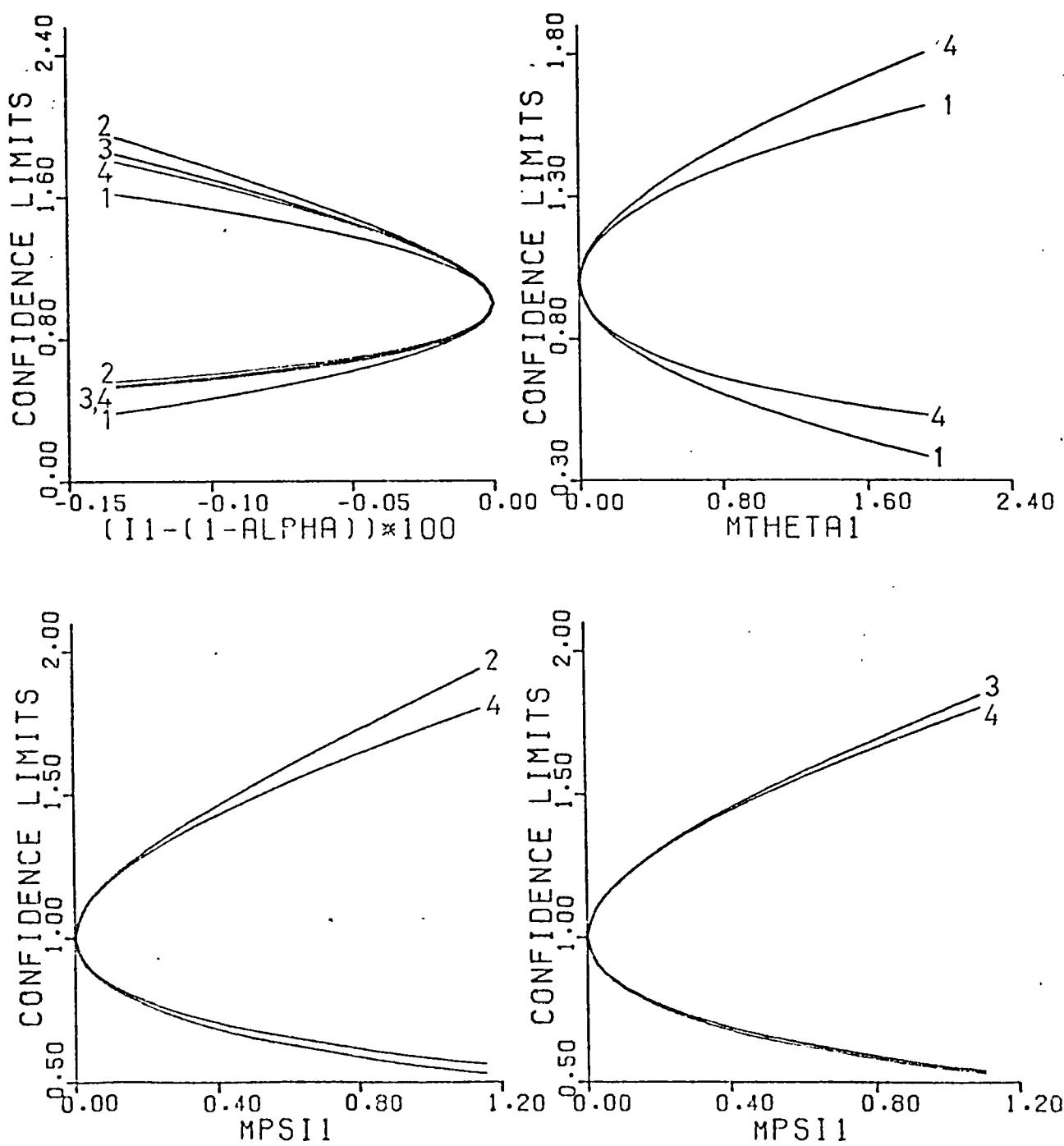
$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 2.0000 \quad \hat{\Theta}_2 = 0.2000$
 RESIDUAL SUM OF SQUARES = 0.2000



I_i : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1,2,3,4$)

FIGURE (5.4.9)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

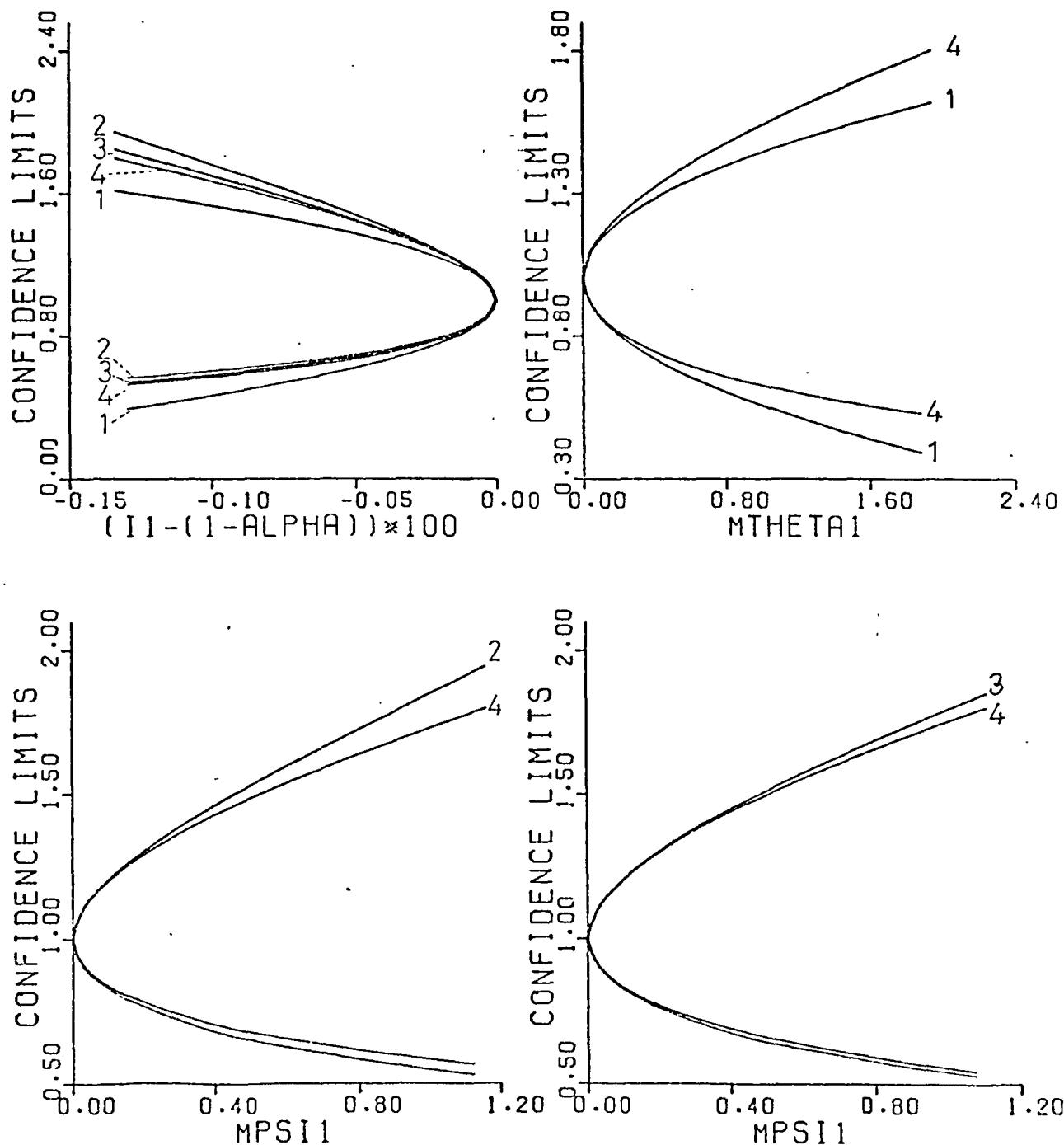
$X_1 = 0.25, 0.5, 1, 0.1, 1.5, 2, 0.4, 0$
 $\hat{\Theta}_1 = 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.1000



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.10)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$

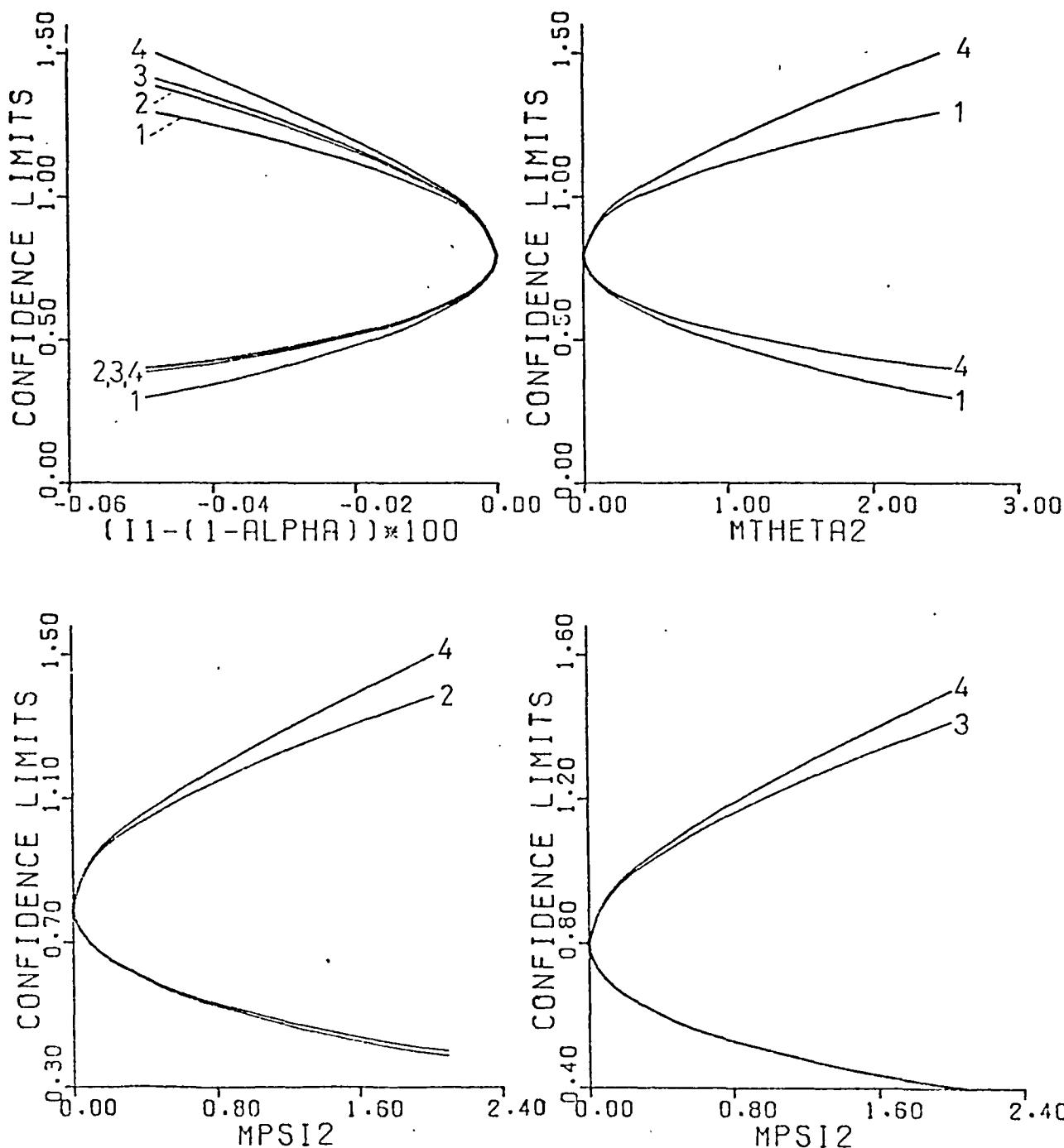
$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.2000



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (1=1,2,3,4)

FIGURE (5.4.11)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2)) \times (\exp(-\Theta_2 \cdot X_1) - \exp(-\Theta_1 \cdot X_1))$

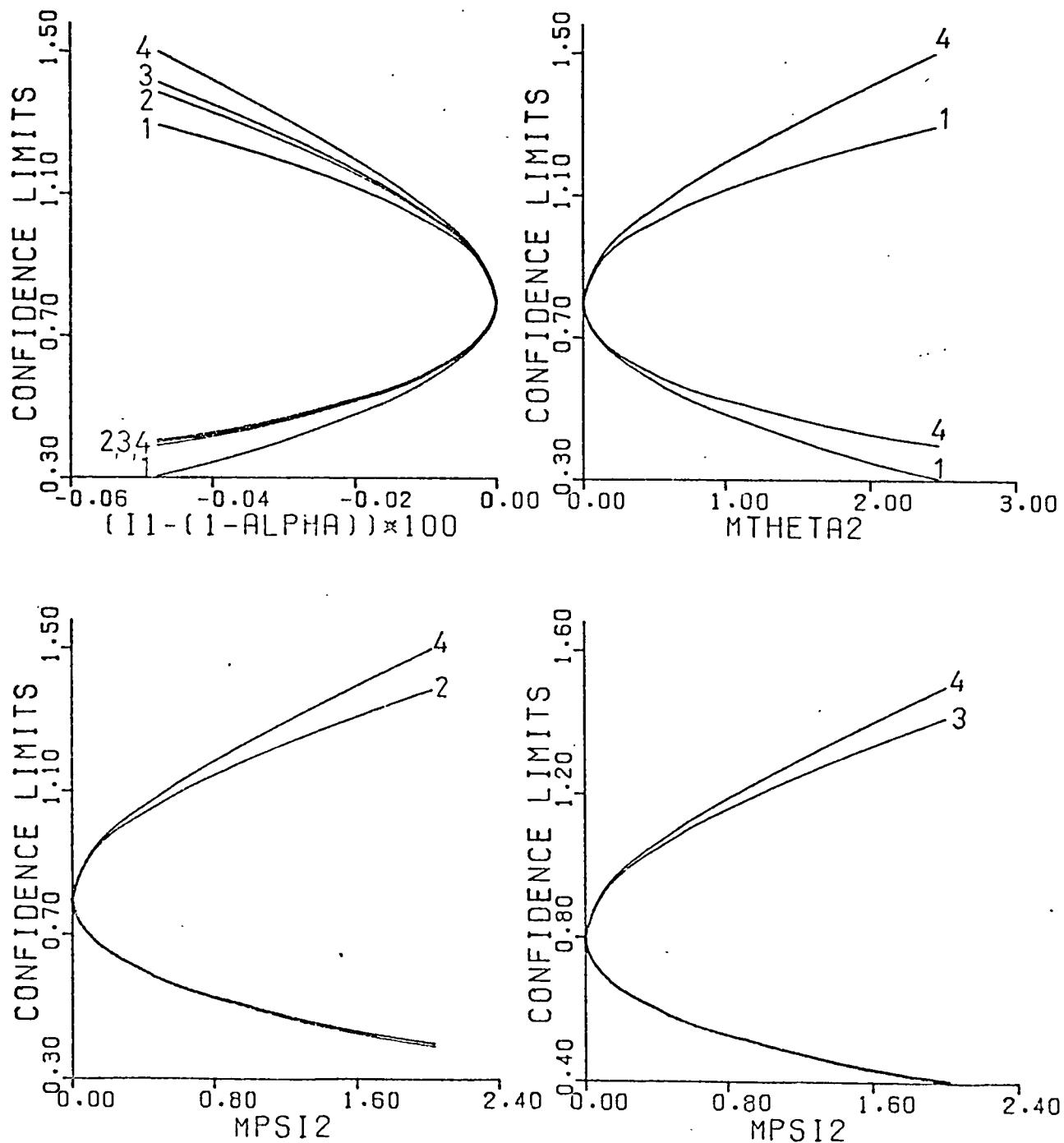
$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.1000



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (1=1,2,3,4)

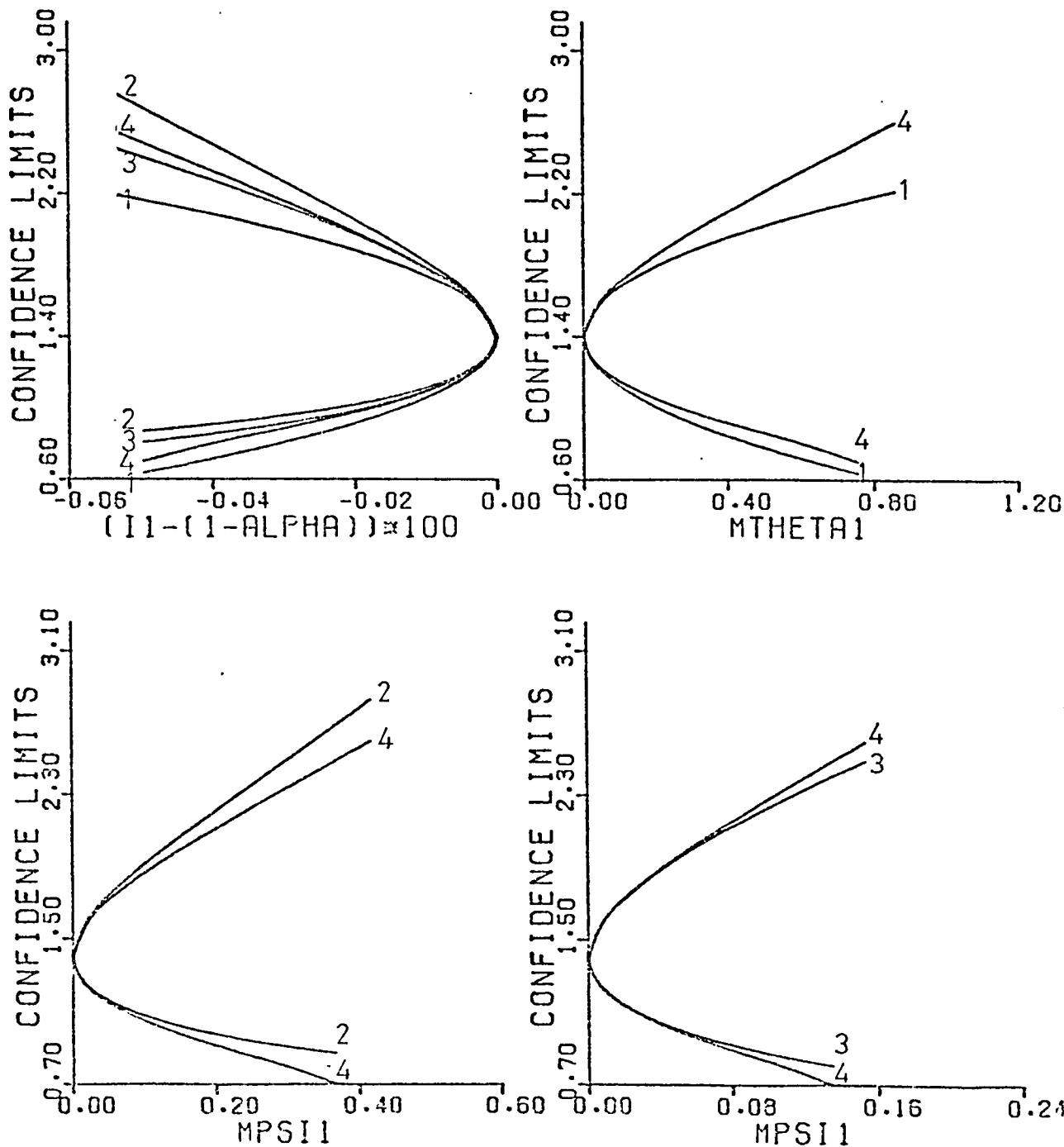
FIGURE (5.4.12)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\theta_1 / (\theta_1 - \theta_2))$
 $\times (\exp(-\theta_2 \cdot X_1) - \exp(-\theta_1 \cdot X_1))$

$X_1 = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\theta}_1 = 1.0000 \quad 0.8000$
 RESIDUAL SUM OF SQUARES = 0.2000



(I) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.13)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\theta_1 \exp(-\theta_2 \cdot x_i) - \theta_2 \exp(-\theta_1 \cdot x_i)) / (\theta_1 - \theta_2)$
 $x_i = 1, 2, 3, 4, 5, 6$
 $\hat{\theta}_1 = 1.4000 \quad 0.4000$
 RESIDUAL SUM OF SQUARES = 0.0001



$(1-\alpha)$: INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1, 2, 3, 4$)

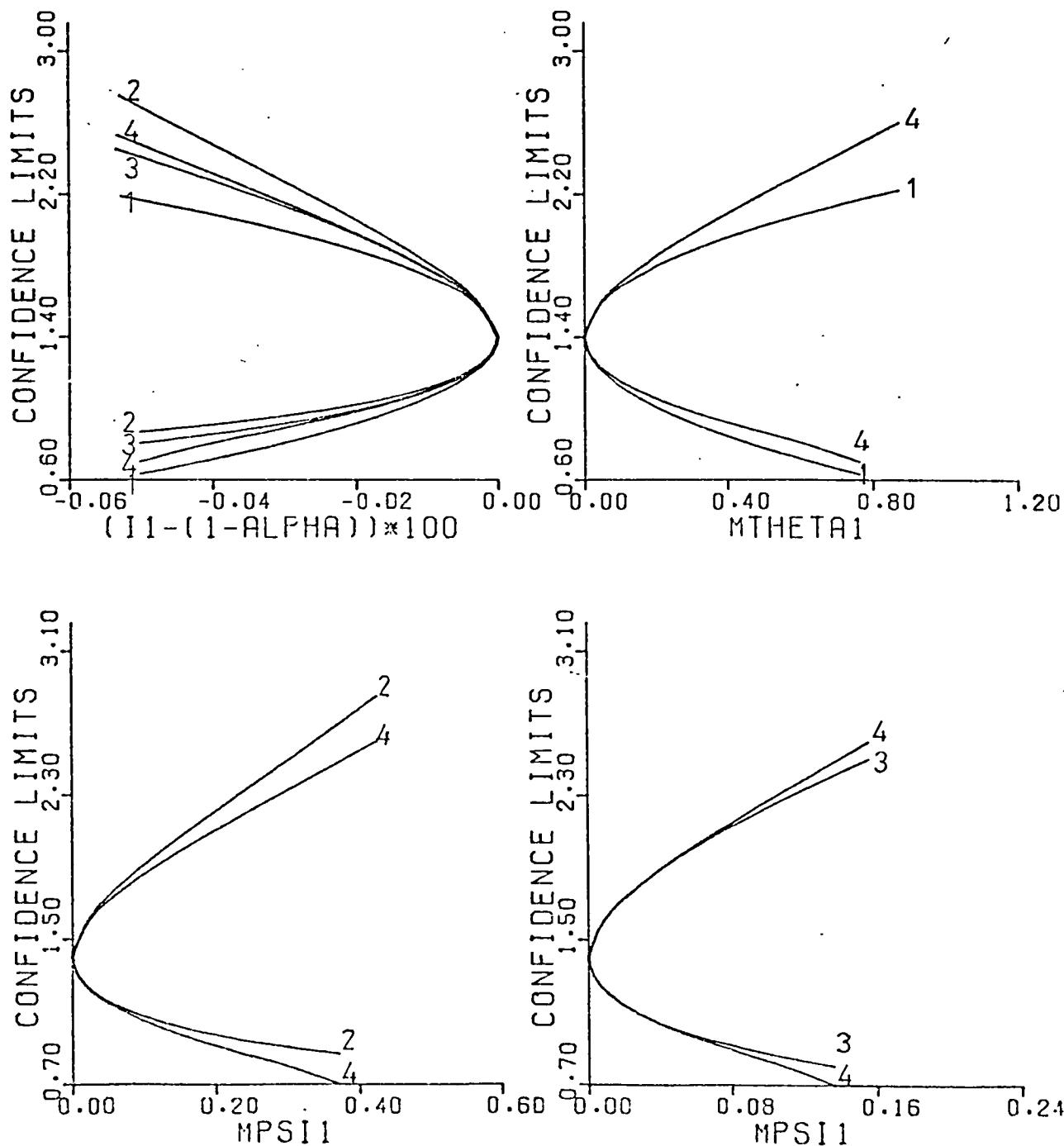
FIGURE (5.4.14)

INTERVAL ESTIMATES IN THE MODEL

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - \Theta_2 \cdot \exp(-\Theta_1 \cdot X)) / (\Theta_1 - \Theta_2)$$

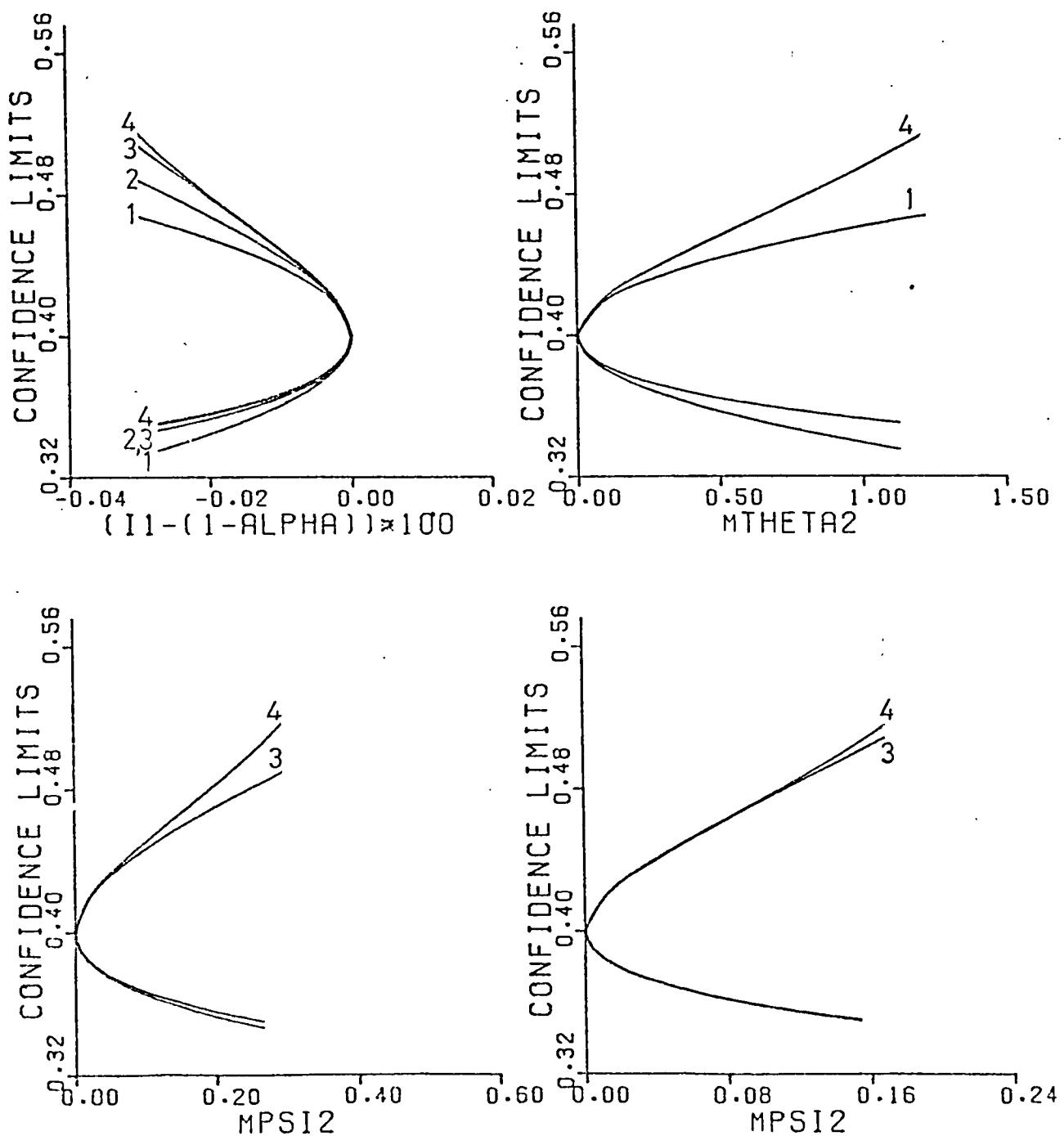
 $X_i = 1, 2, 3, 4, 5, 6$ $\hat{\Theta}_1 \text{ HAT ARE } 1.4000 \quad 0.4000$

RESIDUAL SUM OF SQUARES = 0.0002



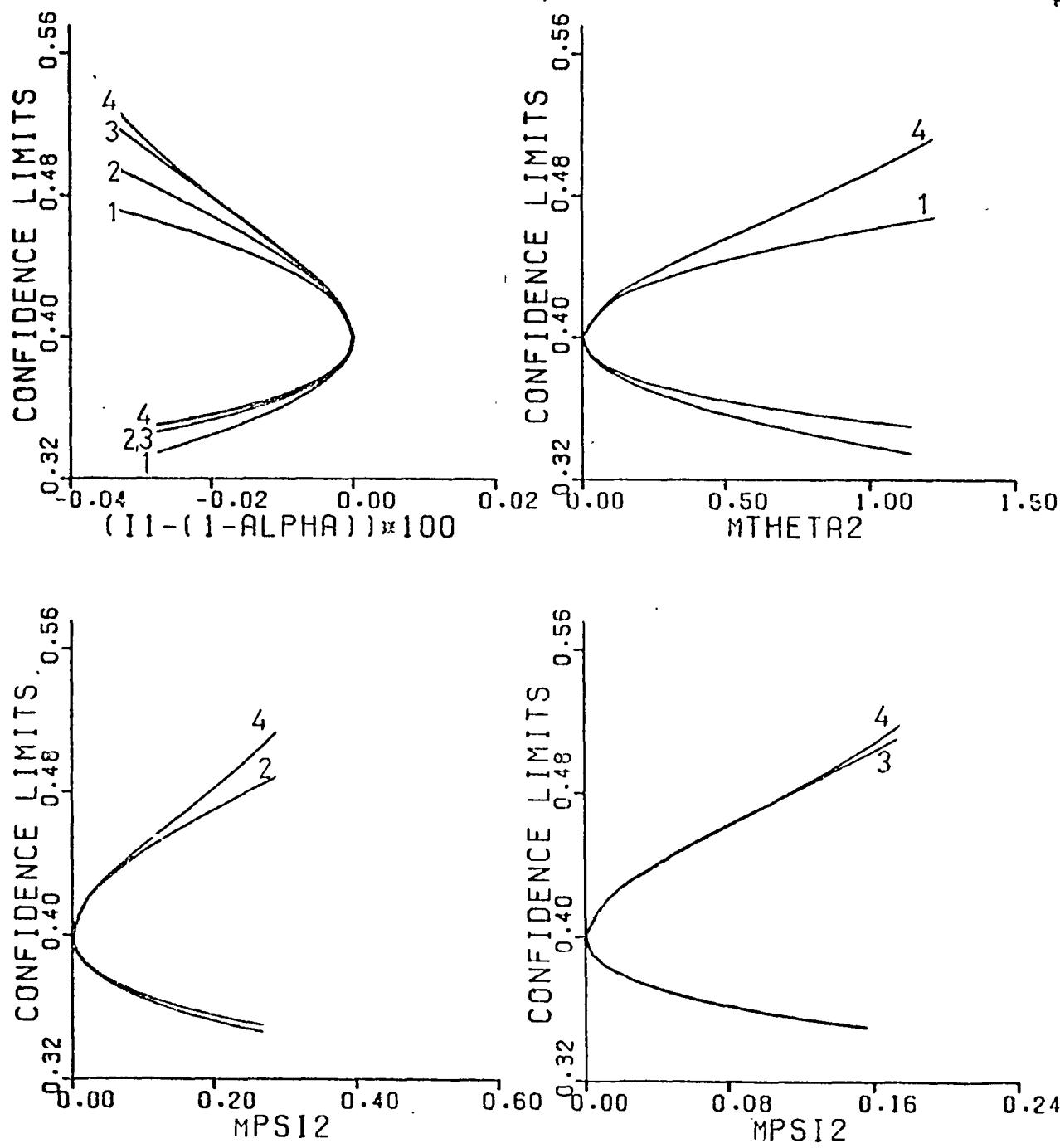
(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD 1 (I=1,2,3,4)

FIGURE (5.4.15)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $- (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $/(\Theta_1 + \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \text{ HAT ARE } 1.4000 \quad 0.4000$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0001$



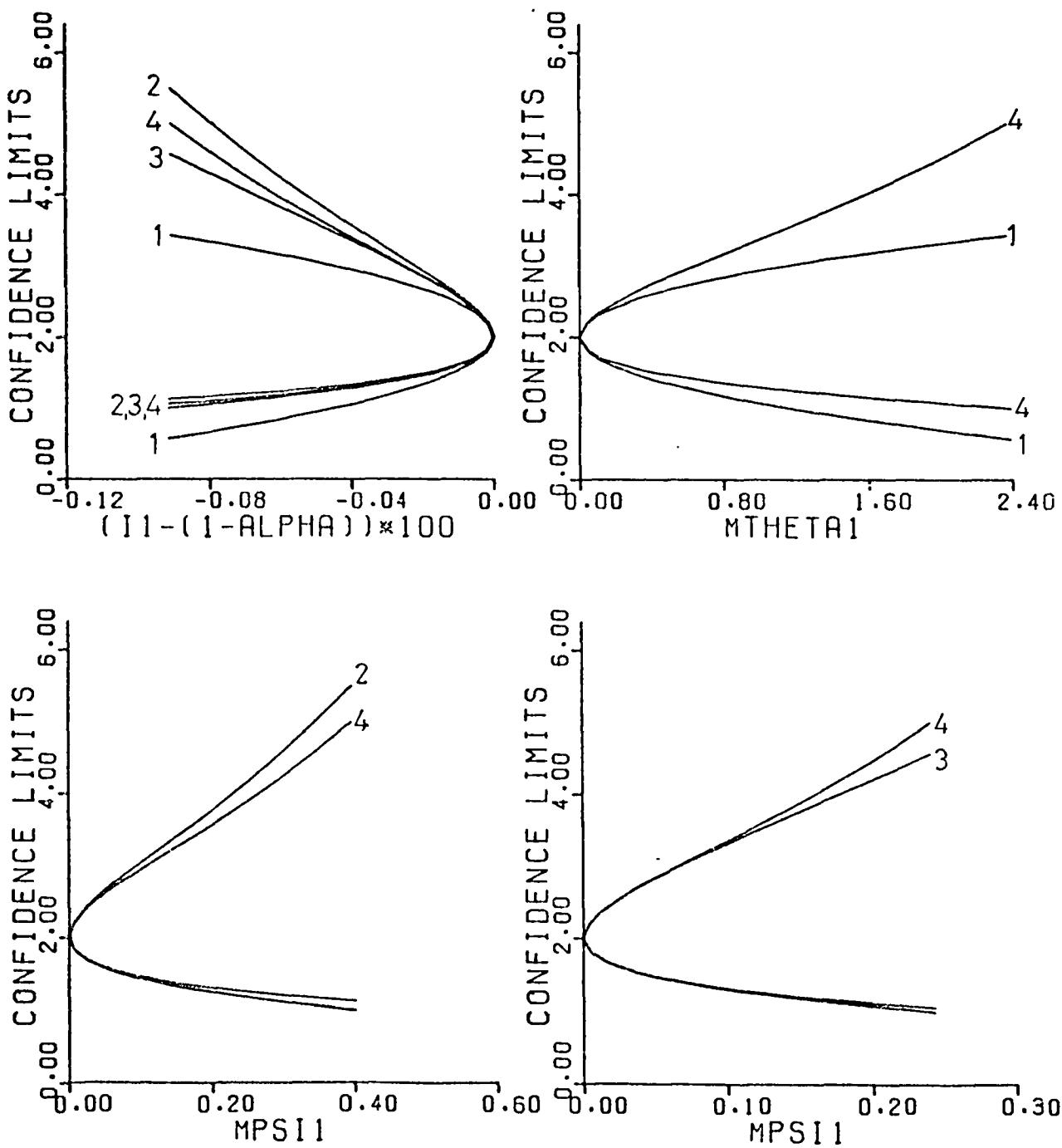
(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.16)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $\quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $\quad / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \text{ HAT ARE } 1.4000 \quad 0.4000$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0002$



$(1 - \text{INTERVAL ESTIMATE (NOMINAL 95 PERCENT)})$ GIVEN BY METHOD I ($i=1, 2, 3, 4$)

FIGURE (5.4.17)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $\quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $\quad / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \text{ HAT ARE } 2.0000 \quad 0.4000$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0001$



$(1) = \text{INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I } (i=1,2,3,4)$

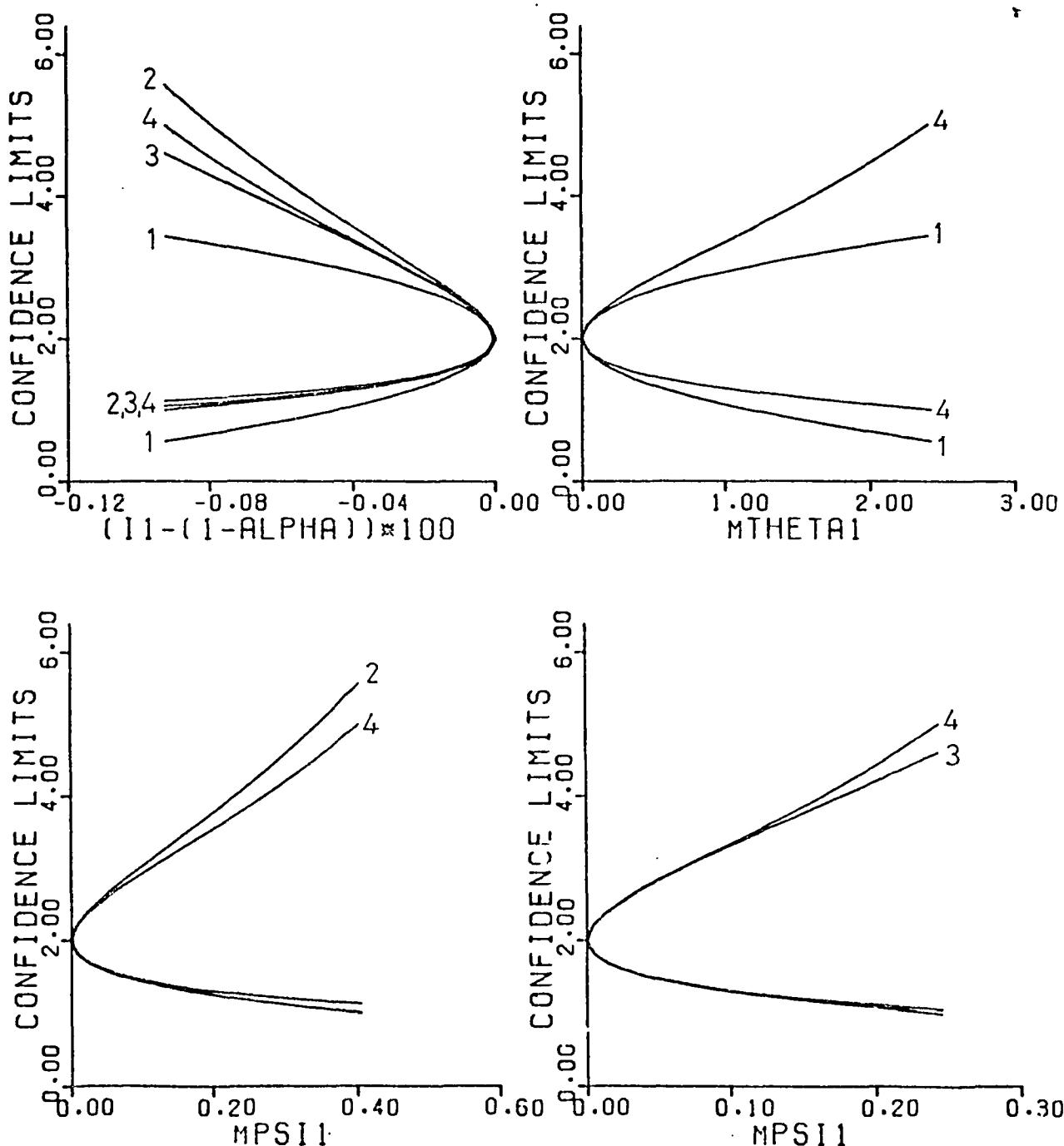
FIGURE (5.4.18)

INTERVAL ESTIMATES IN THE MODEL

$$E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X)) \\ - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)) \\ / (\Theta_1 - \Theta_2)$$

 $X_i = 1, 2, 3, 4, 5, 6$ $\hat{\Theta}_1 = 2.0000 \quad 0.4000$

RESIDUAL SUM OF SQUARES = 0.0002



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.19)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X_i))$
 $- (\Theta_2 \cdot \exp(-\Theta_1 \cdot X_i))$
 $/(\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \cdot \text{HAT} \text{ ARE } 2.0000 \quad 0.4000$
 $\text{RESIUAL SUM OF SQUARES} = 0.0001$

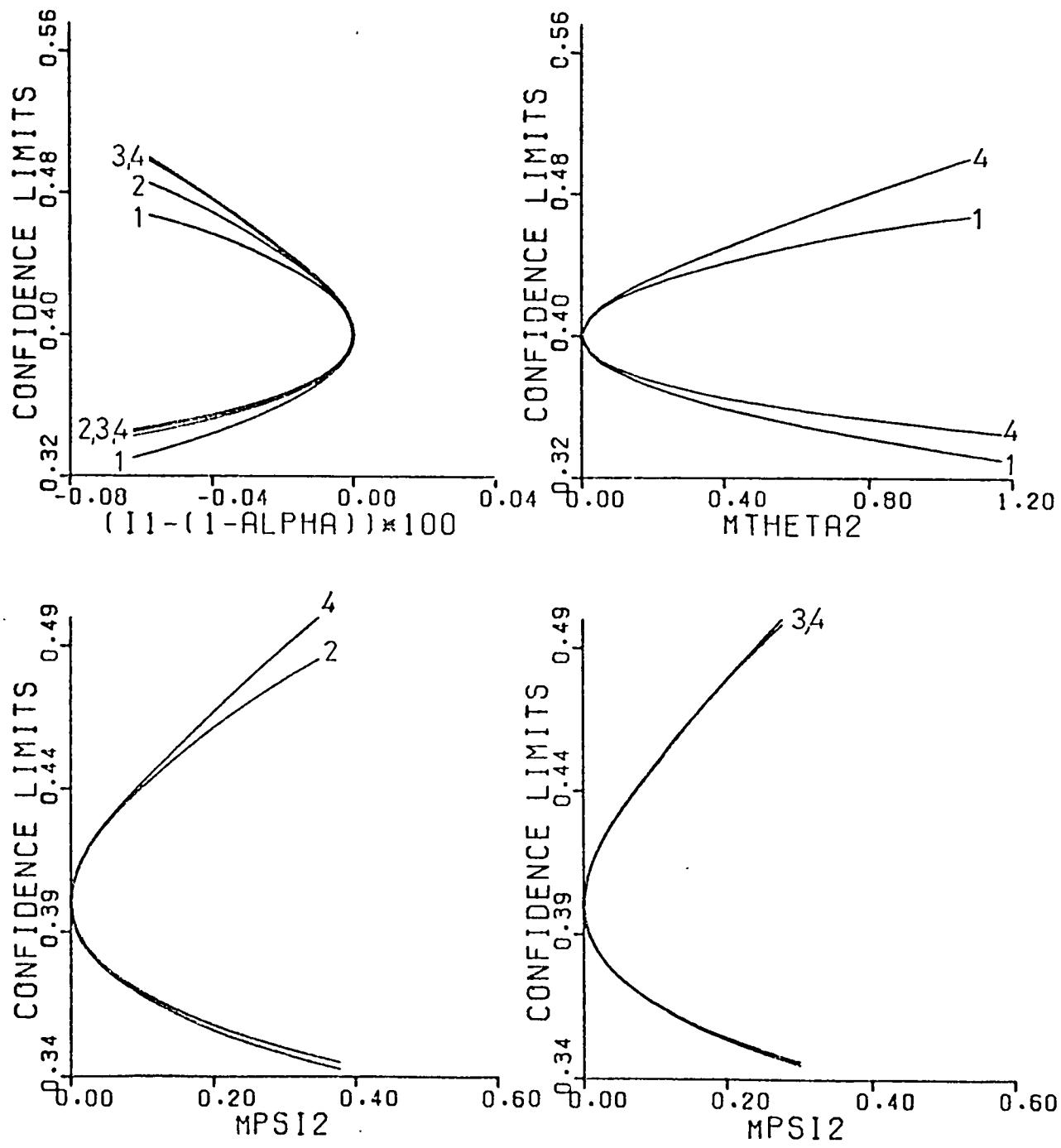
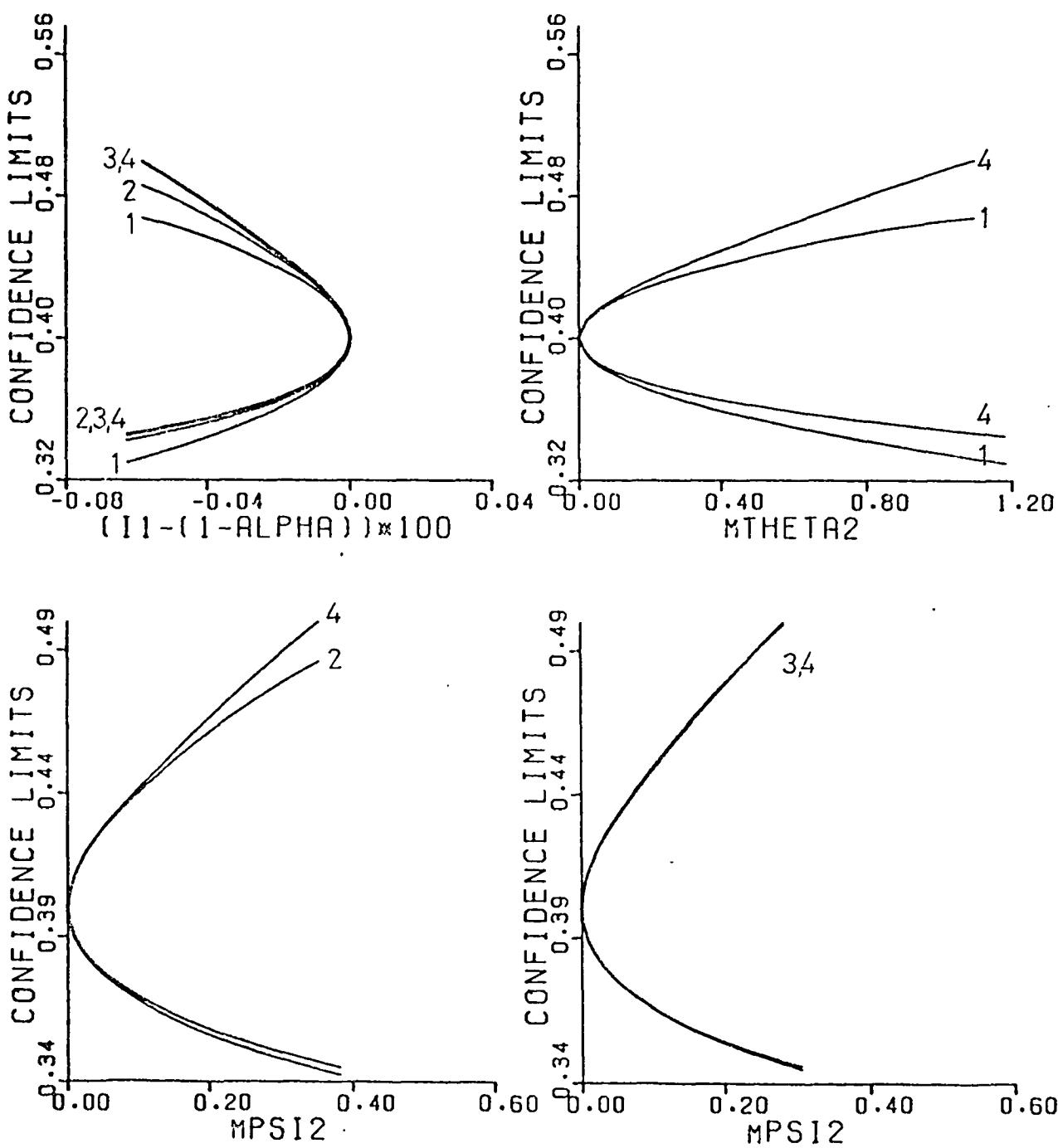
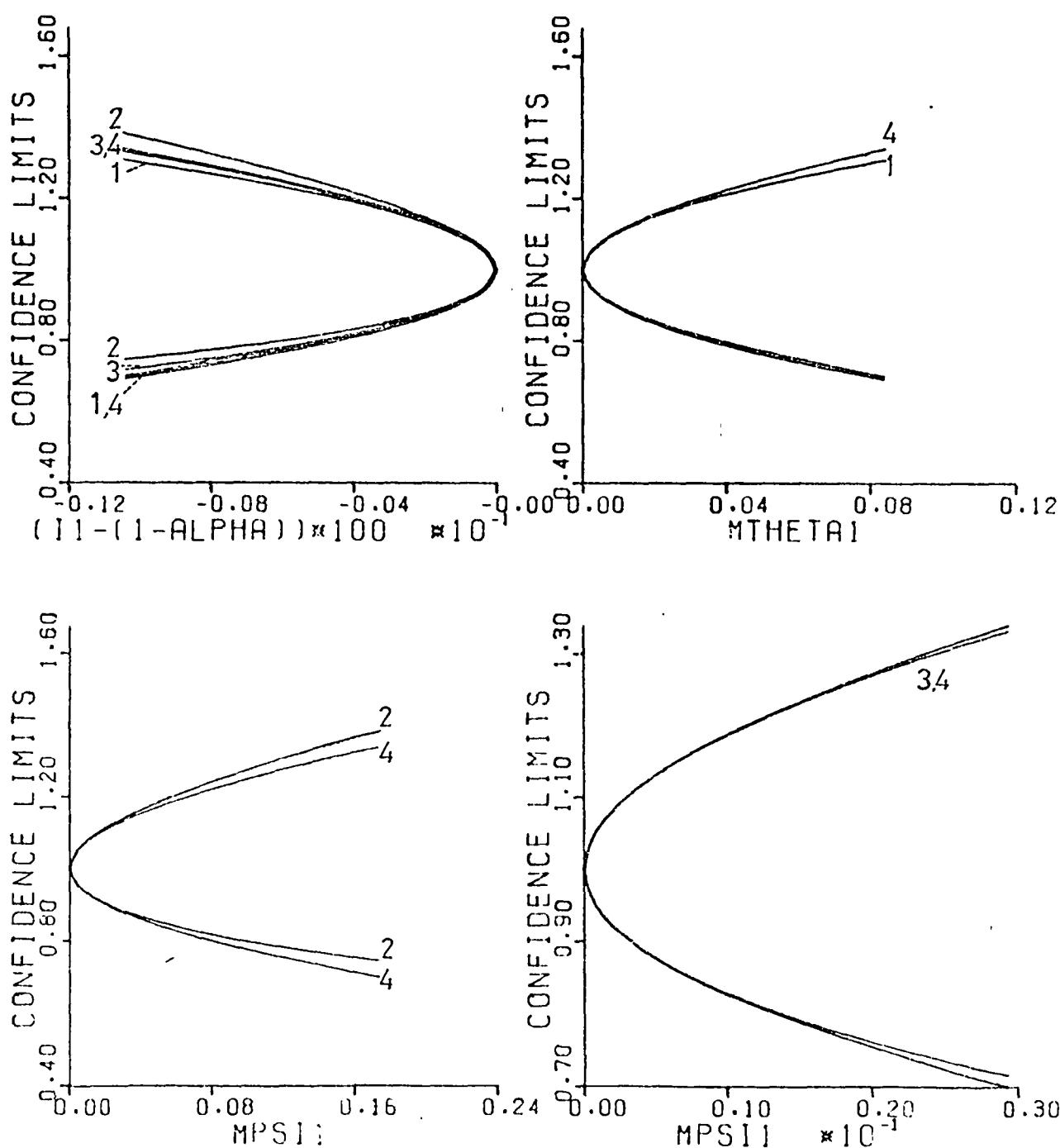


FIGURE (5.4.20)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $\quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $\quad / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\hat{\Theta}_1 \text{ HAT ARE } 2.0000 \quad 0.4000$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0002$



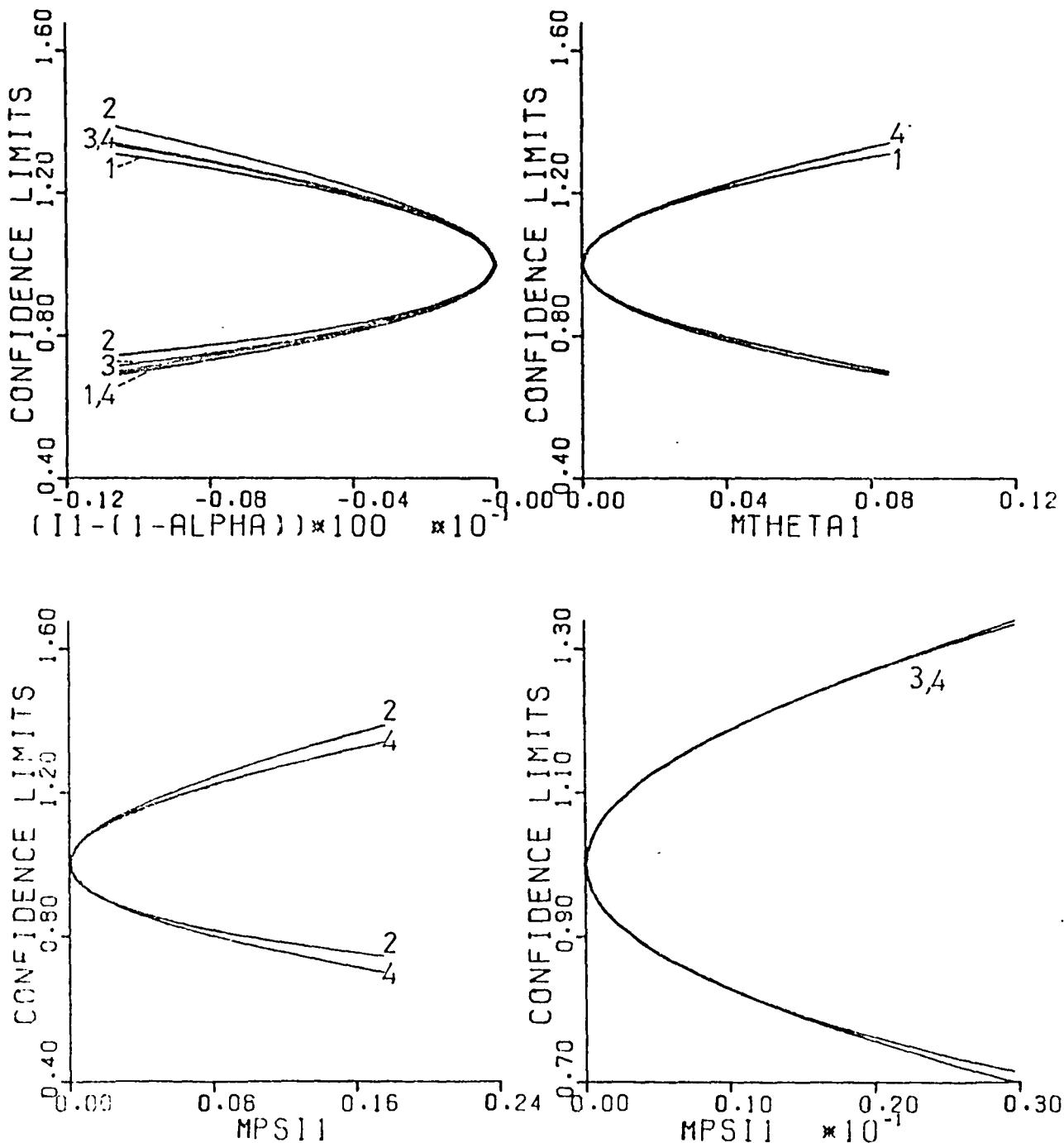
$(1 - (1 - \text{ALPHA})) \times 100$: INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1, 2, 3, 4$)

FIGURE (5.4.21)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $\quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X)))$
 $\quad / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \text{ HAT ARE } 1.0000 \quad 0.3500$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0001$



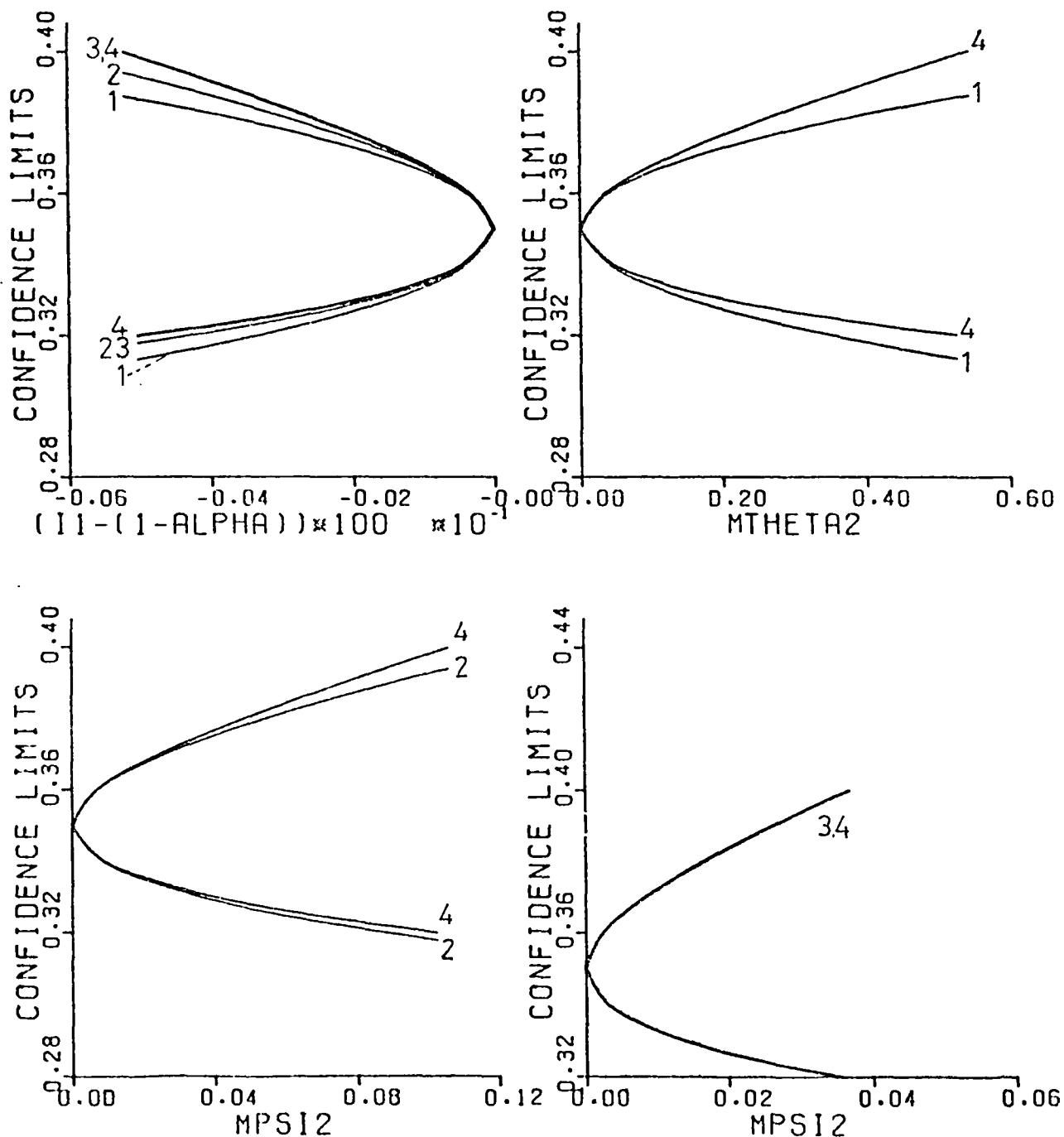
(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

FIGURE (5.4.22)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \cdot \exp(-\Theta_2 \cdot X))$
 $\quad - (\Theta_2 \cdot \exp(-\Theta_1 \cdot X))$
 $\quad / (\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\Theta_1 \text{ HAT ARE } 1.0000 \quad 0.3500$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0002$



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I ($i=1,2,3,4$)

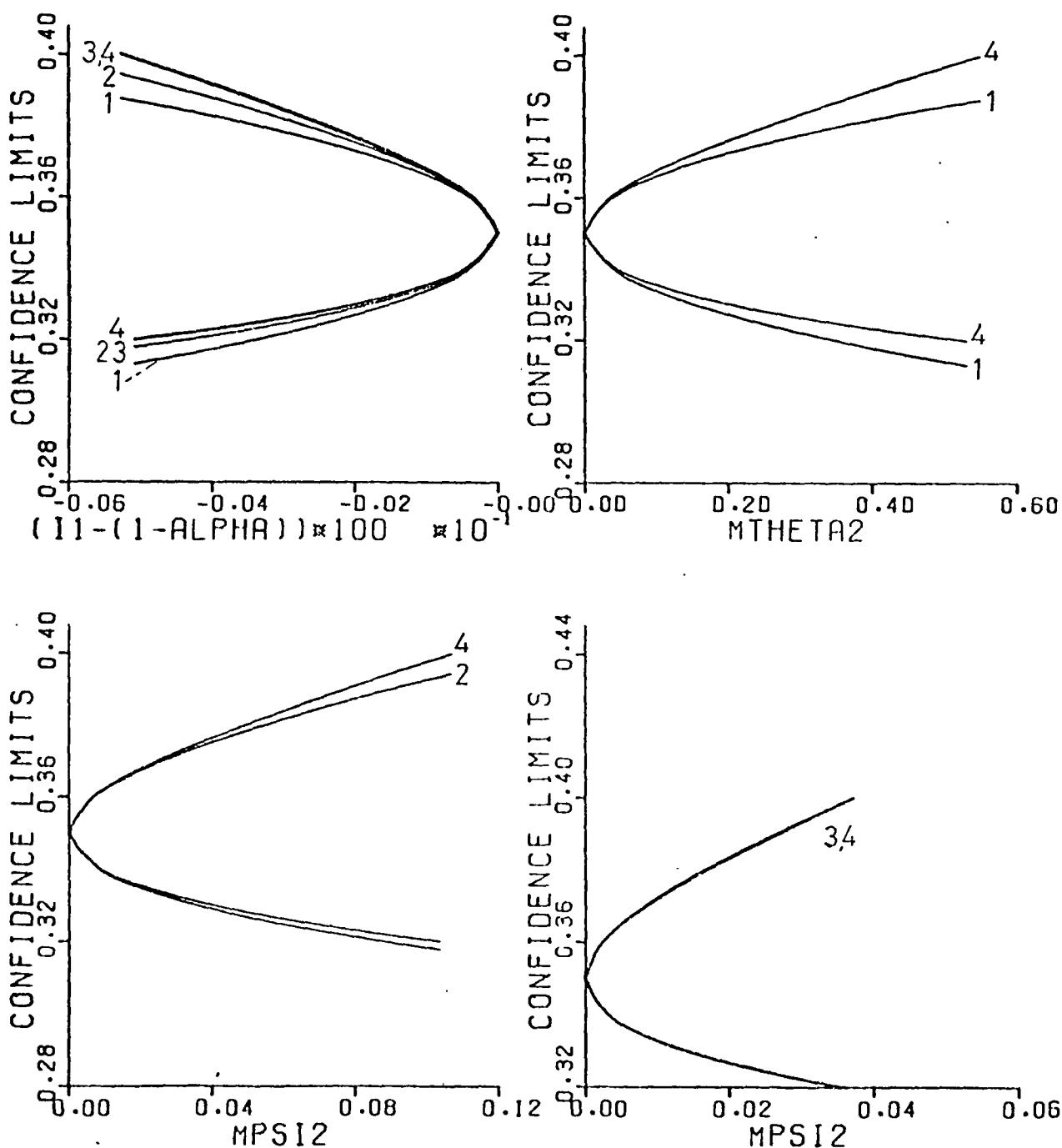
FIGURE (5.4.23)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = 1 - (\Theta_1 \exp(-\Theta_2 \cdot X))$
 $- (\Theta_2 \exp(-\Theta_1 \cdot X))$
 $/(\Theta_1 - \Theta_2)$
 $X_i = 1, 2, 3, 4, 5, 6$
 $\hat{\Theta}_1 \text{ HAT ARE } 1.0000 \quad 0.3500$
 $\text{RESIDUAL SUM OF SQUARES} = 0.0001$



11 : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (1=1,2,3,4)

FIGURE (5.4.24)
 INTERVAL ESTIMATES IN THE MODEL
 $E(Y) = (\Theta_1 / (\Theta_1 - \Theta_2))$
 $\cdot (\exp(-\Theta_2 \cdot X) - \exp(-\Theta_1 \cdot X))$

$X_i = 0.25, 0.5, 1.0, 1.5, 2.0, 4.0$
 $\hat{\Theta}_1 = 1.0000 \quad \hat{\Theta}_2 = 0.3500$
 RESIDUAL SUM OF SQUARES = 0.0002



(1) : INTERVAL ESTIMATE (NOMINAL 95 PERCENT) GIVEN BY METHOD I (I=1,2,3,4)

We note that for each of the $\hat{\theta}_i$, the order given by (5.2.8) with $\hat{\theta}_A = \hat{\theta}_i$ is fairly well preserved in the closeness of the interval estimates given by methods 1 and 3 to those given by method 4. We also note that the change in the residual sum of squares of an observation from s_1^2 to s_2^2 has very slight effect on the interval estimate given by method 4.

We next observe that the values of $J_1(\hat{\theta}_T, \sigma)$ and $J_1(\hat{\theta}_i, \sigma)$ as shown in the figures are fractions of a percent. This implies that our choice of the values of σ , $\hat{\theta}_i$ and $\hat{\theta}_T$ has resulted in situations in which we can refer to the method 4 interval estimates based on these values of $\hat{\theta}_i$ as approximately 100 $I_1(\hat{\theta}_i, \sigma)\%$ interval estimates.

We further observe that the four methods give almost identical interval estimates when $M_{\theta_i} \leq 0.1$. As the interval estimates given by method 4 are approximately 95% interval estimates, the interval estimates given by methods 1, 2 and 3 are also approximately 95% interval estimates.

When $M_{\psi_i} \leq 0.1$, an interval estimate based on the transformation ψ_i is almost identical to that given by method 4. This interval estimate based on ψ_i is therefore an approximately 95% interval estimate.

The above observations indicate that for these models with $\hat{\theta}_T$ near $(1.4, 0.4)^T$ and the levels of σ similar to those considered before, an interval estimate of the parameter θ_i based on method 1, 2, or 3 is an approximately 95% interval estimate provided that the value of the corresponding nonlinearity M_β , where $\beta = \theta_i$ or ψ_i , is less than or equal to 0.1. Note that this value of 0.1 for the bounds of M_β is the same as that for the case when $p = k^* = 2$ (c.f. section 5.3).

For other models which are unconstrained or which can be treated as unconstrained for statistical purposes, we can likewise obtain bounds for the values of nonlinearity within which it is justifiable to use linear theory to obtain the corresponding interval and region estimates. It is expected that the values of these bounds are fractions of one.

Appendix 1 Householder Transformations

Let \underline{c}_i be the i^{th} column of the $(n \times p)$ matrix \underline{C} of rank p . We wish to find the Householder transformations

$$\underline{x}^{(j)} = \underline{x} - [\underline{y}^{(j)}][\underline{y}^{(j)}]^T, \quad (j = 1, 2, \dots, p),$$

such that $\underline{H}^{(p)} \underline{H}^{(p-1)} \dots \underline{H}^{(1)} \underline{C}$ is a $(p \times p)$ upper triangular matrix with an $((n-p) \times p)$ matrix beneath it.

We first find $\underline{y}^{(1)}$ such that $\underline{H}^{(1)} \underline{c}_1$ is a column vector with only one nonzero component, and this component is at the first position. The computational procedure involved in finding the components $v_j^{(1)}$ of $\underline{y}^{(1)}$ is as follows

- (i) compute $r_1 = \sqrt{\sum_{j=1}^n c_{j1}^2}$ which is chosen to have the same sign as c_{11} ,
- (ii) compute $v_1^{(1)} = \sqrt{1 + c_{11}/r_1}$,
- (iii) compute $v_j^{(1)} = c_{j1}/(r_1 v_1^{(1)})$ for $j \neq 1$.

Now let

$$\underline{c}_T^{(j)} = \underline{H}^{(j-1)} \underline{H}^{(j-2)} \dots \underline{H}^{(1)} \underline{c}_j,$$

$$c_{Mu}^{(j)} = 0 \quad \text{for } u < j,$$

and

$$c_{Mu}^{(j)} = c_{Tu}^{(j)} \quad \text{for } u \geq j,$$

where $j = 2, 3, \dots, p$, and $c_{Tu}^{(j)}$ and $c_{Mu}^{(j)}$ are the u^{th} components of $\underline{c}_T^{(j)}$ and $\underline{c}_M^{(j)}$ respectively.

We next find $\underline{v}^{(j)}$, where $j = 2, 3, \dots, p$, such that $\underline{H}^{(j)} \underline{c}_M^{(j)}$ is a column vector with only one nonzero component, and this component is at the j^{th} position. The computational procedure involved in finding $\underline{v}^{(j)}$ is as follows

(i) compute $r_j = \sqrt{\sum_{u=j}^n (c_{Mu}^{(j)})^2}$ chosen to have the same sign as $c_{Mj}^{(j)}$,

(ii) compute $v_j^{(j)} = \sqrt{1 + c_{Mj}^{(j)} / r_j}$,

(iii) compute $v_k^{(j)} = c_{Mk}^{(j)} / (r_j v_j^{(j)})$ for $k \neq j$.

With these $\underline{v}^{(j)}$, the corresponding $\underline{H}^{(j)}$ are the required Householder transformations.

 Appendix 2 Derivation of $I_{1a_j^*}$ and $I_{1a_j^*a_k^*}$ in (3.3.43)-(3.3.48)

Let

$$ST1 = \{1, 2, \dots, p-k^*\}$$

$$ST2 = \{p-k^*+1, p-k^*+2, \dots, p\}$$

and

$$ST3 = \{p+1, p+2, \dots, n\}.$$

Let $i \in ST2$ and $j \in ST1$. We have

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijj}^*} \right]_{a^*=0} = 2d_1^* s_i \sqrt{z_i^{(s)}} z_j^2,$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^{*2}} \right]_{a^*=0} = 2[2z_i^{(s)} z_j^4 - z_j^4 + 4d_1^{*2} z_i^{(s)} z_j^2],$$

$$I_{1a_{ijj}^*} = 0,$$

and

$$I_{1a_{ijj}^* a_{ijj}^*} = 2x_{k^*+2}^2 (d_1^*)^2 .$$

Let $i \in ST_2$ and $j, k \in ST_1$ where $j \neq k$. We have

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijk}^*} \right]_{a^*=0} = 4 s_i \sqrt{z_i^{(s)}} d_1^* z_j z_k, \quad \left[\frac{\partial r_1^{(s)}}{\partial a_{ijj}^*} \right]_{a^*=0} = 2 s_i \sqrt{z_i^{(s)}} d_1^* z_j^2,$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijk}^{*2}} \right]_{a^*=0} = 16 z_i^{(s)} z_j^2 z_k^2 - 8 z_j^2 z_k^2 + 8 z_i^{(s)} d_1^{*2} (z_j^2 + z_k^2),$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^* \partial a_{ikk}^*} \right]_{a^*=0} = 4 z_i^{(s)} z_j^2 z_k^2 - 2 z_j^2 z_k^2,$$

$$I_1 a_{ijk}^* = 0$$

$$I_{1a_{ijk}^* a_{ijk}^*} = 8 x_{k^*+2}^2 (d_1^*)^2$$

and

$$I_{1a_{ijj}^* a_{ikk}^*} = -2 x_{k^*+2}^2 (d_1^*)^2.$$

Let $i \in ST_3$ and $j, k \in ST_2$ where $j \neq k$. We have

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijj}^*} \right]_{a^*=0} = -2 d_1^{*2} z_i z_j^{(s)},$$

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijk}^*} \right]_{a^*=0} = -4 d_1^{*2} z_i s_j s_k \sqrt{z_j^{(s)}} \sqrt{z_k^{(s)}}$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^{*2}} \right]_{a^*=0} = 8 z_i^2 (z_j^{(s)})^2 d_1^{*2} - 2 z_j^{(s)} d_1^{*2} (4 z_i^2 - z_j^2) d_1^{*2},$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijk}^* \partial a_{ijk}^*} \right]_{a^*=0} = 32z_i^2 z_j^2 z_k^2 d_1^{*2} - 8z_i^2 (z_j^{(s)} + z_k^{(s)}) d_1^{*2} + 8z_j^2 z_k^2 d_1^{*4},$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^* \partial a_{ikk}^*} \right]_{a^*=0} = 8z_i^2 z_j^2 z_k^2 d_1^{*2} + 2z_j^2 z_k^2 d_1^{*4},$$

$$I_{la_{ijj}^*} = 0,$$

$$I_{la_{ijk}^*} = 0,$$

$$I_{la_{ijj}^* a_{ijj}^*} = -2\chi_{k^*+2}^2 (d_1^{*2})$$

$$I_{la_{ijk}^* a_{ijk}^*} = -8\chi_{k^*+2}^2 (d_1^{*2})$$

and

$$I_{la_{ijj}^* a_{ikk}^*} = 2\chi_{k^*+2}^2 (d_1^{*2}).$$

Let $i \in ST3$, $j \in ST1$ and $k \in ST2$. We have

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijj}^*} \right]_{a^*=0} = 0,$$

$$\left[\frac{\partial r_1^{(s)}}{\partial a_{ijk}^*} \right]_{a^*=0} = -4d_1^* z_i z_j s_k \sqrt{z_k^{(s)}}$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^* \partial a_{ijj}^*} \right]_{a^*=0} = 0,$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijk}^*} \right]_{a^*=0} = 16z_i^2 z_j^2 z_k^{2-(s)} - 8(z_i^2 z_j^2 + z_i^2 z_k^{2-(s)}) d_1^{*2} - z_j^2 z_k^{2-(s)} d_1^{*2},$$

$$\left[\frac{\partial^2 r_1^{(s)}}{\partial a_{ijj}^* \partial a_{ikk}^*} \right]_{a^*=0} = 2z_j^2 z_k^{2-(s)} d_1^{*2},$$

$$I_{1a_{ijj}^* a_{ijj}^*} = 0,$$

$$I_{1a_{ijk}^* a_{ijk}^*} = -8x_{k^*+2}^2 (d_1^{*2}),$$

and

$$I_{1a_{ijj}^* a_{ikk}^*} = 2x_{k^*+2}^2 (d_1^{*2}).$$

Next, it can be shown that all the $I_{1a_{ijk}^*}$ and $I_{1a_{i_1 j_1 k_1}^* a_{i_2 j_2 k_2}^*}$ other than those already derived are zero.

APPENDIX 3 Programs PARTIT, POWPRO, and subroutines POWSUC, POWSUA,
POWSUB, POWSUC and SIGSUC

76/09/17 IMPERIAL COLLEGE FORTRAN COMPILER KRONOS 2.1.X PSR2+ 77/05/03. 21.22.28.

```

MNF(B=PARTIT)
C
C      PROGRAM PARTIT IS FOR PARTITIONING THE SET OF ALL
C      (A+I1J1K1,A+I2J2K2) INTO SUBSETS(C.F. SECTION (4.9))
C
000000B   1.      PROGRAM PARTIT(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT,TAPE7)
00606UB   2.      COMMON/MSKP/MS(6),KP(500,21)
00606UB   3.      IREAD=5
00606UB   4.      IPRIINT=6
006061B   5.      ITAPE7=7
006062B   6.      READ(IREAD,2) KCH,IN,NPAR,NOBS,KSTAR
C
C      NPAR IS TOTAL NUMBER OF COMPONENTS IN THE PARAMETER VECTOR
C      NOBS IS TOTAL NUMBER OF OBSERVATIONS
C      KSTAR IS NUMBER OF COMPONENTS OF INTEREST IN THE PARAMETER VECTOR
C
006074B   7.      2 FORMAT(A6,I4,14I5)
006074B   8.      CALL CHECIN(KCH,IN,6HNPARB,0)
C
C      ROUTINE CHECIN CHECKS THAT DATA CARD IS CORRECT
C
006076B   9.      WRITE(IPRINT,3) NPAR,NOBS,KSTAR
006105B   10.     3 FORMAT(1X,5HNPAR=.I2,1H/,5HNOBS=.I2,1H/,6HKSTAR=.I2/)
006105B   11.     - NPMK=NPAR-KSTAR
006106B   12.     - NPMKP1=NPMK+1
006107B   13.     - NPARP=NPAR+1
006110B   14.     - NPMKP2=NPMK+2
006111B   15.     - NPARP2=NPAR+2
006112B   16.     - IP=0
C
C      THESE DO LOOPS ARE FOR PARTITIONING THE SET OF ALL
C      (A+I1J1K1,A+I2J2K2) INTO SUBSETS. THE SUBSETS ARE INDEXED BY IP
C
006113B   17.     DO 160 I1=NPMKP1,NOBS
006116B   18.     DO 150 I2=NPMKP1,NOBS
006120B   19.     DO 140 J1=1,NPAR
006123B   20.     DO 130 K1=J1,NPAR
006124B   21.     DO 120 J2=1,NPAR
006126B   22.     DO 110 K2=J2,NPAR
006127B   23.     IF(.NOT.(I1.LE.I2)) GO TO 110
006132B   24.     IP=IP+1
006134B   25.     MS(1)=I1
006134B   26.     MS(2)=J1
006135B   27.     MS(3)=K1
006137B   28.     MS(4)=I2
006140B   29.     MS(5)=J2
006141B   30.     MS(6)=K2
006142B   31.     DO 10 IM=1,6
006144B   32.     I=MS(IM)
006144B   33.     IF(I.GE.1.AND.I.LE.NPMK) GO TO 7
006152B   34.     IF(I.GE.NPMKP1.AND.I.LE.NPAR) GO TO 8
006156B   35.     IF(I.GE.NPARP.AND.I.LE.NOBS) GO TO 9
006162B   36.     7 KP(IP,IM)=1
006165B   37.     GO TO 10
006166B   38.     8 KP(IP,IM)=2
006171B   39.     GO TO 10
006172B   40.     9 KP(IP,IM)=3
006175B   41.     10 CONTINUE
006200B   42.     IT=6
006201B   43.     DO 16 IM=1,5
006203B   44.     IMP1=IM+1
006203B   45.     DO 15 JM=IMP1,6
006207B   46.     IT=IT+1
006210B   47.     IF(MS(IM).EQ.MS(JM)) GO TO 12

```

```

006213B   48.    IF(MS(1M).NE.MS(JM)) GO TO 15
006214B   49.    12 KP(IP,IT)=1
006220B   50.    GO TO 15
006221B   51.    13 KP(IP,IT)=0
006224B   52.    15 CONTINUE
006226B   53.    16 CONTINUE
C          C (KP(IP,IT),IT=1,21) COMPLETELY SPECIFY THE IP-TH SUBSET
C
006231B   54.    IF(IP.EU.1) GO TO 40
006233B   55.    KSAME=0
006235B   56.    IPM1=IP-1
006234B   57.    DO 20 IPM=1,IPM1
006237B   58.    DO 18 IT=1,21
006241B   59.    IF(KP(IPM,IT).NE.KP(IP,IT)) GO TO 20
006241B   60.    18 CONTINUE
006250B   60.    KSAME=1
006254B   62.    GO TO 30
006254B   63.    20 CONTINUE
006255B   63.    30 IF(KSAME.EQ.0) GO TO 40
006261B   65.    IP=IP-1
006262B   66.    GO TO 110
006262B   67.    40 LSKIP=1
006263B   68.    IF(IP,NE.1) GO TO 54
006265B   69.    WRITE(1PRINT,50) IP,(MS(IB),IB=1,6),(KP(IP,IB),IB=1,21)
006311B   70.    50 FORMAT(1X,3HIP=15,1H/,21HI1,J1,K1,I2,J2,K2 ARE,6(I2,1X),1H/,12HKP
               1(IP,*)) ARE,21(I1,1X))
006311B   71.    GO TO 58
006311B   72.    54 LSKIP=1
006312B   73.    WRITE(1PRINT,56) IP,(MS(IB),IB=1,6),(KP(IP,IB),IB=1,21)
006336B   74.    56 FORMAT(1X,3HIP=15,1H/,21X+6(I2+1X),1H/.12X,21(I1,1X))
006336B   75.    58 LSKIP=1
006336B   76.    WRITE(1TAPE7,60) IP,(MS(IB),IB=1,6),(KP(IP,IB),IB=1,21)
006362B   77.    60 FORMAT(1X,28I4)
006362B   78.    110 CONTINUE
006364B   79.    120 CONTINUE
006367B   80.    130 CONTINUE
006372B   81.    140 CONTINUE
006375B   82.    150 CONTINUE
006400B   83.    160 CONTINUE
006403B   84.    STOP
006406B   85.    END
C          C PROGRAM POWPRO,SUBROUTINES POWSUO,POWSUA,POWSUB AND POWSUC ARE FOR
C          C REPRESENTING (1) PRODUCT OF FIRST PARTIAL DERIVATIVES OF (R+)**2
C          C W.R.T. A+I1J1K1 AND A+I2J2K2 EVALUATED AT A+=0 AND (2) SECOND
C          C PARTIAL DERIVATIVES OF (R+)**2 W.R.T. TO A+I1J1K1,A+I2J2K2
C          C EVALUATED AT A+=0(C,F.SECTIONS (4.2) TO (4.6))
C
000000B   1.    PROGRAM POWPRO(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT,TAPE9,TAPE10,
               1TAPE7,1TAPE11)
014140B   2.    COMMON/MSKP/MS16),KP(30,21)
014140B   3.    COMMON/AC1AAZ/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
               10,10)
014140B   4.    COMMON/NKPP/N(2,2,3),KPP(8,6)
014140B   5.    COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10+10),AZ1(4,
               110,10)
014140B   6.    COMMON/1GIGMX/IG(4),IGMAX(4)
014140B   7.    COMMON/ACGAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
               110,10)
014140B   8.    COMMON/AKSTAR/AKSTAR(50),DEGREE(50),GAMMA(10)
014140B   9.    COMMON/1READ/IREAD,IPRINT,1TAPE
014140B   10.   COMMON/NOBS/NOBS,NPAR,KSTAR,NPMK,NPMKP1,NPARP
014140B   11.   COMMON/KS/KS,IP,11MAX,II
014140B   12.   COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
014140B   13.   COMMON/LPRINT/LPN(50)

```

```

0141408      14.      COMMON/IWA8/IWA,IWB
0141408      15.      COMMON/KPOWERK/KPOWER
0141408      16.      IREADU=5
0141408      17.      IPHINT=6
0141418      18.      ITAPE7=7
0141428      19.      READ(IREAD,4) KCH,IN,NPAR,NOBS,KSTAR
C
C      NPAR IS TOTAL NUMBER OF COMPONENTS IN THE PARAMETER VECTOR
C      NOBS IS TOTAL NUMBER OF OBSERVATIONS
C      KSTAR IS NUMBER OF COMPONENTS OF INTEREST IN THE
C      PARAMETER VECTOR
C
0141548      20.      4 FORMAT(A6,I4,14I5)
0141548      21.      CALL CHECIN(KCH,IN,6HNPAR08,0)
C
C      ROUTINE CHECIN CHECKS THAT THE DATA CARD IS CORRECT
C
0141568      22.      WRITE(IPRINT,5) NPAR,NOBS,KSTAR
0141658      23.      5 FORMAT(/1X,4NPAR,NOBS,KSTAR AREA,3(I3,1X))
0141658      24.      NPMK=NPAR-KSTAR
0141668      25.      NPMKP1=NPMK+1
0141678      26.      NPARP=NPAR+1
C
0141708      27.      READ(IREAD,4) KCH,IN,(LPN(I),I=1,7)
0142038      28.      CALL CHECIN(KCH,IN,6HLPN(1),0)
0142058      29.      READ(IREAD,4) KCH,IN,(LPN(I),I=8,14)
0142208      30.      CALL CHECIN(KCH,IN,6HLPN(8),0)
C
C      LPN(*) DECIDE WHETHER THE INTERMEDIATE RESULTS WILL BE PRINTED
C      OUT
C
0142228      31.      KAAC=6
0142228      32.      READ(IREAD,6) KCH,IN,(AAC(I),I=1,KAAC)
0142378      33.      WRITE(IPRINT,8) (AAC(I),I=1,KAAC)
C
C      AAC(*) ARE SOME NUMBERS ASSOCIATED WITH THE EQUATION
C      SDA1(Z)=(D+)**2 (C.F. SECTION (4.4)). THEY ARE RESPECTIVELY
C      2.0,-2.0,-1.0,1.0,4.0 AND -4.0
C
0142518      34.      6 FORMAT(A6,I4,7F10.6)
0142518      35.      8 FORMAT(/1X,4AAC(*) AREA,6(F10.4,1X)/)
0142518      36.      CALL CHECIN(KCH,IN,6HAAC(1),0)
0142538      37.      READ(IREAD,4) KCH,IN,KPOWER
0142628      38.      CALL CHECIN(KCH,IN,6HKPOWER,0)
C
C      NORMALLY KPOWER=1 EXCEPT IN THE CASE WHEN WE WANT TO USE POWPRO,
C      POWSUU,POWSUA,POWSUB AND SIGSUC TO DERIVE II(THETA,SIGMA)
C
0142648      39.      READ(IREAD,4) KCH,IN,KSELEC,IS1,JS1,KS1,IS2,JS2,KS2
0143018      40.      CALL CHECIN(KCH,IN,6HKSELEC,0)
0143038      41.      READ(IREAD,4) KCH,IN,KSTART
0143128      42.      CALL CHECIN(KCH,IN,6HKSTART,0)
C
C      NORMALLY KSELEC=0,KSTART=0 AND IS1,JS1,KS1,IS2,JS2,KS2 ARE ANY
C      POSITIVE INTEGERS EXCEPT IN THE CASE WHEN WE WANT TO CHECK SOME
C      PARTICULAR PART OF THE PROGRAM
C
0143148      43.      READ(IREAD,4) KCH,IN,IMAX
C
C      IMAX IS TOTAL NUMBER OF SUBSETS GENERATED BY PARTITIONING THE SET
C      OF ALL (A+I1J1K1,A+I2J2K2) (C.F. PROGRAM PARTIT)
C
0143238      44.      CALL CHECIN(KCH,IN,5HIMAX,0)
0143258      45.      IP=0
0143258      46.      IWA=0
0143258      47.      IWB=0
0143268      48.      DD 1910 ID=1,IMAX

```

```

014331B   49.    IP=IP+1
014332B   50.    IF(KP0WERK.EQ.1) IP=1
014336B   51.    READ(ITAPE7,10) IDD,(MS(IB),IB=1,6),(KP(IP,IB),IB=1,21)
C          (KP(IP,IB),IB=1,21) COMPLETELY SPECIFY THE IP-TH SUBSET
C
014361B   52.    10 FORMAT(1X,28I4)
014361B   53.    IF(.NOT.(KSELLEC.EQ.0.OR.(I1.EQ.IS1.AND.J1.EQ.JS
014403B   54.    1.I2.EQ.IS2.AND.J2.EQ.JS2.AND.K2.EQ.KS2))) GO TO 1910
014404B   55.    IF(LPN(1).EQ.0) GO TO 35
014415B   56.    33 FORMAT(/1X,21HI1,J1,K1,I2,J2,K2 ARE,6(I2,1X))
014415B   57.    WRITE(1PRINT,33) (MS(I),I=1,6)
014432B   58.    34 FORMAT(1X,5HIP = ,I3,1H/,12HKP(IP,I) ARE,21(I1,1X))
014432B   59.    45 LSKIP=1
014432B   60.    I1=MS(1)
014433B   61.    J1=MS(2)
014435B   62.    K1=MS(3)
014436B   63.    I2=MS(4)
014440B   64.    J2=MS(5)
014441B   65.    K2=MS(6)
014443B   66.    KS=0
014444B   67.    IF(I1.EQ.I2.AND.J1.EQ.J2.AND.K1.EQ.K2.AND.J1.EQ.K1) KS=1
014455B   68.    IF(I1.EQ.I2.AND.J1.EQ.J2.AND.K1.EQ.K2.AND.J1.NE.K1) KS=2
014466B   69.    KII=30
014466B   70.    DO 38 II=1,KII
014470B   71.    DO 36 IZ=1,NOBS
014473B   72.    AZ(I1,IZ)=0.0
014473B   73.    36 CONTINUE
014501B   73.    38 CONTINUE
C          TO REPRESENT THE EQUATION SDA1(Z)=(D+)**2 IN A COMPUTER
C          (C.F. SECTION (4.4))
C          NONLINEAR TERMS IN SDA1(Z) ARE INDEXED BY II
C
014504B   75.    II=0
014504B   76.    IF(.NOT.(KP(IP,1).EQ.2.AND.KP(IP,2).EQ.1.AND.KP(IP,3).EQ.1)) GO TO
014504B   144
C          TO REPRESENT 2.0(SUM FROM I EQUAL TO P-KSTAR+1 TO P) (SUM FROM
C          J EQUAL TO 1 TO P-KSTAR) (SUM FROM K EQUAL TO 1 TO P-KSTAR)
C          OF A+IJKZIZJZK
C
014515B   77.    II=II+1
014516B   78.    IF(J1.EQ.K1) AC(II)=AAC(1)
014522B   79.    IF(J1.NE.K1) AC(II)=2.0*AAC(1)
014526B   80.    AA(II,1)=1.0
014526B   81.    AA(II,2)=0.0
014530B   82.    DO 40 IZ=1,NOBS
014533B   83.    IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
014542B   84.    IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
014550B   85.    IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
014556B   86.    40 CONTINUE
014560B   87.    IF(LPN(5).EQ.0) GO TO 43
014561B   88.    WRITE(1PRINT,41) (MS(I),I=1,6)
014572B   89.    41 FORMAT(/1X,21HI1,J1,K1,I2,J2,K2 ARE,6(I2,1X))
014572B   90.    WRITE(1PRINT,42) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
014616B   91.    42 FORMAT(/1X,5HII = ,I2,1X,9HAC(II) = ,F4.1,2X,11HAA(II,1) = ,F4.1,2
014616B   92.    1X,11HAA(II,2) = ,F4.1,2X,13HAZ(II,IZ) ARE,10(F4.1,1X))
014616B   93.    43 LSKIP=1
014616B   94.    44 IF(.NOT.(KP(IP,4).EQ.2.AND.KP(IP,5).EQ.1.AND.KP(IP,6).EQ.1)) GO TO
014627B   110
014631B   94.    IF(.NOT.KS.EQ.0) GO TO 110
014633B   95.    II=II+1
014633B   96.    IF(J2.EQ.K2) AC(II)=AAC(1)
014637B   97.    IF(J2.NE.K2) AC(II)=2.0*AAC(1)

```

```

014643B   98.      AA(II,1)=0.0
014643B   99.      AA(II,2)=1.0
014645B   100.     DO 50 IZ=1,NOBS
014650B   101.     IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
014657B   102.     IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
014665B   103.     IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
014673B   104.     50 CONTINUE
014675B   105.     IF(LPN(5).EQ.0) GO TO 60
014676B   106.     WRITE(1PRINT,41) (MS(I),I=1,6)
014707B   107.     WRITE(1PRINT,42) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
014733B   108.     60 LSKIP=1
014733B   109.     110 LOUT=1
014734B   110.     IF(.NOT.((KP(IP,1).EQ.3).AND.(KP(IP,2).EQ.2.OR.KP(IP,3).EQ.2))) GO
1 TU 124
C
C      TO REPRESENT -2.0(SUM FROM I EQUAL TO P+1 TO N) (SUM FROM J EQUAL
C      TO 1 TO P, SUM FROM K EQUAL TO 1 TO P, AND AT LEAST ONE OF J,K IS
C      EQUAL TO P-KSTAR+1 OR P-KSTAR+2,...,OR P) OF A+IJKLIZJZK
C
014746B   111.     II=II+1
014747B   112.     IF(J1.EQ.K1) AC(II)=AAC(2)
014754B   113.     IF(J1.NE.K1) AC(II)=2.0*AAC(2)
014760B   114.     AA(II,1)=1.0
014760B   115.     AA(II,2)=0.0
014762B   116.     DO 120 IZ=1,NOBS
014765B   117.     IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
014774B   118.     IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
015002B   119.     IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
015010B   120.     120 CONTINUE
015012B   121.     IF(LPN(5).EQ.0) GO TO 122
015013B   122.     WRITE(1PRINT,41) (MS(I),I=1,6)
015024B   123.     WRITE(1PRINT,42) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
015050B   124.     122 LSKIP=1
015050B   125.     124 IF(.NOT.(KS.EQ.0)) GO TO 180
015053B   126.     IF(.NOT.((KP(IP,4).EQ.3).AND.(KP(IP,5).EQ.2.OR.KP(IP,6).EQ.2))) GO
1 TU 180
015065B   127.     II=II+1
015066B   128.     IF(J2.EQ.K2) AC(II)=AAC(2)
015072B   129.     IF(J2.NE.K2) AC(II)=2.0*AAC(2)
015076B   130.     AA(II,1)=0.0
015076B   131.     AA(II,2)=1.0
015100B   132.     DO 130 IZ=1,NOBS
015103B   133.     IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
015112B   134.     IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
015120B   135.     IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
015126B   136.     130 CONTINUE
015130B   137.     IF(LPN(5).EQ.0) GO TO 132
015131B   138.     WRITE(1PRINT,41) (MS(I),I=1,6)
015142B   139.     WRITE(1PRINT,42) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
015166B   140.     132 LSKIP=1
015166B   141.     180 LOUT=1
015167B   142.     CALL PUWSUU
015172B   143.     CALL PUWSUA
015174B   144.     1910 CONTINUE
015176B   145.     STOP
015201B   146.     END

```

000000B 1. SUBROUTINE CHECIN(KA,I,KB,J)
 C
 C ROUTINE CHECIN CHECKS THAT DATA CARD IS CORRECT
 C
 000000B 2. IF(KA.NE.KB)GO TO 100
 000001B 3. IF(I.EQ.J) GO TO 9000
 000003B 4. IF(I.EQ.0.AND.J.EQ.1) GO TO 9000
 000006B 5. 100 WRITE(6,110) KA,I,KB,J
 000016B 6. 110 FORMAT(23H ERROR IN CARD LABELLED,2X,A6,I4,2X,15HPROGRAM EXPECTS,2
 2X,A6,I4)
 000016B 7. 9000 RETURN
 000020B 8. END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAT NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
(C= RELATIVE TO //)

000000B 9. SUBROUTINE POWSUO
 C
 C ROUTINE POWSUO IS CALLED IN POWPRO
 C
 000000B 10. COMMON/MSKP/MS(6),KP(30,21)
 000000B 11. COMMON/ACAAA2/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
 10,10)
 000000B 12. COMMON/NKPP/N(2,2,3),KPP(8,6)
 000000B 13. COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10,10),AZ1(4,
 110,10)
 000000B 14. COMMON/IGIGMX/IG(4),IGMAX(4)
 000000B 15. COMMON/ACGAAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
 110,10)
 000000B 16. COMMON/AKSTAR/AKSTAR(50),DEGREE(50),GAMMA(10)
 000000B 17. COMMON/IREAD/IREAD,IPRINT,1TAPE
 000000B 18. COMMON/NOUS/NOUS,NPAR,KSTAR,NPMK,NPMKP1,NPARP
 000000B 19. COMMON/KS/KS,IP,11MAX,II
 000000B 20. COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
 000000B 21. COMMON/LPRINT/LPN(50)
 000000B 22. COMMON/IUAB/IUA,1WB
 000000B 23. 41 FORMAT(/1X,21HI1,J1,K1,I2,J2,K2 ARE,6(I2,1X))
 000000B 24. 42 FORMAT(/1X,5HII = ,I2,1X,9HAC(II) = ,F4.1,2X,11HAA(II,1) = ,F4.1,2
 1X,11HAA(II,2) = ,F4.1,2X,13HAZ(II,I2) ARE,10(F4.1,1X))
 000000B 25. IF(.NOT.(KP(IP,1).EQ.2.AND.KP(IP,2).EQ.1.ANU.KP(IP,3).EQ.1.AND.KP(
 1IP,4).EQ.2.AND.KP(IP,5).EQ.1.AND.KP(IP,6).EQ.1))GO TO 304
 C
 C TO REPRESENT -1.0(SUM FROM I EQUAL TO P-KSTAR+1 TO P) OF (SQUARE OF
 C (SUM FROM J EQUAL TO 1 TO P-KSTAR,SUM FROM K EQUAL TO 1 TO P-KSTAR)
 C OF A+1JKZJZK)
 C
 000017B 26. IF(.NOT.(KS.EQ.0)) GO TO 244
 000021B 27. IF(I1.NE.I2) GO TO 304
 000023B 28. II=II+1
 000025B 29. IF(J1.EQ.K1.AND.J2.EQ.K2) AC(II)=2.0*AAC(3)
 000033B 30. IF(J1.NE.K1.AND.J2.EQ.K2) AC(II)=4.0*AAC(3)
 000042B 31. IF(J1.EQ.K1.ANU.J2.NE.K2) AC(II)=4.0*AAC(3)
 000051B 32. IF(J1.NE.K1.AND.J2,NE,K2) AC(II)=8.0*AAC(3)
 000060B 33. AA(II,1)=1.0
 000060B 34. AA(II,2)=1.0
 000062B 35. DO 220 IZ=1,NOBS
 000064B 36. IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
 000073B 37. IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
 000101B 38. IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
 000107B 39. IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
 000115B 40. 220 CONTINUE
 000117B 41. IF(LPN(5).EQ.0) GO TO 230
 000120B 42. WRITE(IPRINT,41) (MS(I),I=1,6)

```

0001318    43.      WRITE(1PRINT,42)II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
0001558    44.      230 LSKIP=1
0001558    45.      GO TO 304
0001568    46.      244 IF(.NOT.(KS.EQ.1)) GO TO 252
0001628    47.      II=II+1
0001638    48.      AC(II)=AAC(3)
0001648    49.      AA(II,1)=2.0
0001658    50.      AA(II,2)=0.0
0001668    51.      DO 248 IZ=1,NOBS
0001718    52.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
0002008    53.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
0002068    54.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
0002148    55.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
0002228    56.      248 CONTINUE
0002248    57.      IF(LPN(5).EQ.0) GO TO 250
0002258    58.      WRITE(1PRINT,41) (MS(I),I=1,6)
0002368    59.      WRITE(1PRINT,42)II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
0002628    60.      250 LSKIP=1
0002628    61.      GO TO 304
0002638    62.      252 II=II+1
0002658    63.      AC(II)=4.0*AAC(3)
0002668    64.      AA(II,1)=2.0
0002708    65.      AA(II,2)=0.0
0002708    66.      DO 256 IZ=1,NOBS
0002738    67.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
0003028    68.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
0003108    69.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
0003168    70.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
0003248    71.      256 CONTINUE
0003268    72.      IF(LPN(5).EQ.0) GO TO 260
0003278    73.      WRITE(1PRINT,41) (MS(I),I=1,6)
0003408    74.      WRITE(1PRINT,42)II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
0003648    75.      260 LSKIP=1
0003648    76.      304 LEND=1
0003658    77.      IF(.NOT.(KP(IP,1).EQ.3.AND.KP(IP,4).EQ.3.AND.(KP(IP,2).EQ.2.OR.KP(IP,3).EQ.2.OR.KP(IP,5).EQ.2.OR.KP(IP,6).EQ.2))) GO TO 410
C
C      TO REPRESENT 1.0(SUM FROM I EQUAL TO P+1 TO N) (SUM FROM J EQUAL TO
C      1 TO P, SUM FROM K EQUAL TO 1 TO P, SUM FROM L EQUAL TO 1 TO P, SUM
C      FROM M EQUAL TO 1 TO P, AND AT LEAST ONE OF J,K,L,M IS EQUAL TO
C      P-KSTAR+1 OR P-KSTAR+2....OR P) OF A+IJKA+ILMZJZKZLZM
C
0004068    78.      IF(.NOT.(KS.EQ.0)) GO TO 344
0004108    79.      IF(II.NE.IZ) GO TO 410
0004128    80.      II=II+1
0004138    81.      IF(J1.EQ.K1.AND.J2.EQ.K2) AC(II)=2.0*AAC(4)
0004228    82.      IF(J1.NE.K1.AND.J2.EQ.K2) AC(II)=4.0*AAC(4)
0004318    83.      IF(J1.EQ.K1.AND.J2.NE.K2) AC(II)=4.0*AAC(4)
0004408    84.      IF(J1.NE.K1.AND.J2.NE.K2) AC(II)=8.0*AAC(4)
0004478    85.      AA(II,1)=1.0
0004478    86.      AA(II,2)=1.0
0004518    87.      DO 320 IZ=1,NOBS
0004538    88.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
0004628    89.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
0004708    90.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
0004768    91.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
0005048    92.      320 CONTINUE
0005068    93.      IF(LPN(5).EQ.0) GO TO 330
0005078    94.      WRITE(1PRINT,41) (MS(I),I=1,6)
0005208    95.      WRITE(1PRINT,42)II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOBS)
0005448    96.      330 LSKIP=1
0005448    97.      GO TO 410
0005458    98.      344 IF(.NOT.(KS.EQ.1)) GO TO 352
0005518    99.      II=II+1
0005528   100.      AC(II)=AAC(4)
0005538   101.      AA(II,1)=2.0

```

```

0005548    102.      AA(II,2)=0.0
0005558    103.      DO 348 IZ=1,NOBS
0005608    104.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
0005678    105.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
0005758    106.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
0006038    107.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
0006118    108.      348 CONTINUE
0006138    109.      IF(LPN(5).EQ.0) GO TO 350
0006148    110.      WRITE(IPRINT,41) (MS(I),I=1,6)
0006258    111.      WRITEL(IPRINT,42) II,AC(II),AA(II,1),(AZ(II,IB),IB=1,NOBS)
0006518    112.      350 LSKIP=1
0006518    113.      GO TO 410
0006528    114.      352 II=II+1
0006548    115.      AC(II)=4.0*AC(4)
0006558    116.      AA(II,1)=2.0
0006578    117.      AA(II,2)=0.0
0006578    118.      DO 356 IZ=1,NOBS
0006628    119.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
0006718    120.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
0006778    121.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
0007058    122.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
0007138    123.      356 CONTINUE
0007158    124.      IF(LPN(5).EQ.0) GO TO 360
0007168    125.      WRITE(IPRINT,41) (MS(I),I=1,6)
0007278    126.      WRITE(IPRINT,42) II,AC(II),AA(II,1),(AZ(II,IB),IB=1,NOBS)
0007538    127.      360 LSKIP=1
0007538    128.      410 LOUT=1
0007548    129.      RETURN
0007578    130.      END

C
C   ROUTINE POWSUA IS CALLED IN POWPRO
C

0000008    1.      SUBROUTINE POWSUA
000000R    2.      COMMON/MSKP/MS(6),KP(30,21)
000000R    3.      COMMON/ACAAA/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
10,10)
000000R    4.      COMMON/NKPP/N(2,2,3),KPP(8,6)
000000R    5.      COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10,10),AZ1(4,
110,10)
000000R    6.      COMMON/IGIGMX/IG(4),IGMAX(4)
000000B    7.      COMMON/ACGAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
110,10)
000000R    8.      COMMON/AKSTAR/AKSTAR(50),DEGREE(50),GAMMA(10)
000000B    9.      COMMON/IREAD/IREAD,IPRINT,ITAPE
000000R   10.      COMMON/NOBS/NOES,NPAR,KSTAR,NPMK,NPMKP1,NPARP
000000R   11.      COMMON/KS/KS,IP,IIMAX,II
000000R   12.      COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
000000R   13.      COMMON/LPRINT/LPN(50)
000000R   14.      COMMON/IQAB/IQA,IQB
000000R   15.      COMMON/KPOWER/KPOWER
000000R   16.      N(1,1,1)=I1
000000B   17.      N(1,1,2)=J1
000002B   18.      N(1,1,3)=K1
000003R   19.      N(1,2,1)=I1
000004R   20.      N(1,2,2)=K1
000005R   21.      N(1,2,3)=J1
000006B   22.      N(2,1,1)=I2
000010R   23.      N(2,1,2)=J2
000011R   24.      N(2,1,3)=K2
000013B   25.      N(2,2,1)=I2
000014R   26.      N(2,2,2)=K2
000014R   27.      N(2,2,3)=J2
000016B   28.      IF(.NOT.((KP(IP,1).NE.1.AND.KP(IP,4).NE.1).AND.(KP(IP,1).EQ.2.OR.K
1P(IP,4).EQ.2).AND.KP(IP,2).EQ.1.AND.KP(IP,3).EQ.1.AND.KP(IP,5).EQ.
11.AND.KP(IP,6).EQ.1)) GO TO 510

C
C   TO REPRESENT 4.0 * SUM FROM H EQUAL TO P-KSTAR+1 TO N, SUM FROM
C   I EQUAL TO P-KSTAR+1 TO M, AND AT LEAST ONE OF H,I IS EQUAL TO
C   P-KSTAR+1 OR OR P-KSTAR+2,...,OR P1 (SUM FROM J EQUAL TO 1 TO
C   P-KSTAR, SUM FROM K EQUAL TO 1 TO P-KSTAR, SUM FROM L EQUAL TO 1 TO
C   P-KSTAR) OF A+HJLA+IKLHZIZJZK
C

```

```

000042R 29. IF(.NOT.(KS.EQ.0)) GO TO 460
000045R 30. IPP=0
000045R 31. DO 454 IPERM=1,2
000050R 32. IF(IPERM.EQ.1) GO TO 420
000052R 33. IF(IPERM.EQ.2) GO TO 424
000053R 34. 420 IPER1=1
000054R 35. IPER2=2
000055R 36. GO TO 426
000056R 37. 424 IPER1=2
000057R 38. IPER2=1
000060R 39. 426 LSKIP=1
000061R 40. DO 450 J=1,2
000063R 41. DO 446 K=1,2
000065R 42. IF(N(IPER1,J,3).NE.N(IPER2,K,3)) GO TO 446
000074R 43. IPP=IPP+1
000076R 44. DO 428 IM=1,3
000100R 45. KPP(IPP,IM)=N(IPER1,J,IM)
000105R 46. IN=IM+3
000106R 47. KPP(IPP,IN)=N(IPER2,K,IM)
000113R 48. 428 CONTINUE
000114R 49. IF(IPP.EQ.1) GO TO 436
000115R 50. KSAME=0
000116R 51. IPPM=IPP-1
000117R 52. DO 432 I=1,IPPM
000121R 53. DO 430 L=1,6
000123R 54. IF(KPP(IPP,L).NE.KPP(I,L)) GO TO 432
000123R 55. 430 CONTINUE
000132R 55. KSAME=1
000136R 57. GO TO 434
000136R 58. 432 CONTINUE
000137R 58. 434 IF(KSAME.EQ.0) GO TO 436
000143R 60. IPP=IPP-1
000144R 61. GO TO 446
000144B 62. 436 LSKIP=1
000145R 63. II=II+1
000147R 64. AC(II)=AAC(5)
000150R 65. AA(II,1)=1.0
000151R 66. AA(II,2)=1.0
000152R 67. DO 438 IZ=1,NORS
000155R 68. IF(IZ.EQ.KPP(IPP,1)) AZ(II,IZ)=AZ(II,IZ)+1.0
000164R 69. IF(IZ.EQ.KPP(IPP,2)) AZ(II,IZ)=AZ(II,IZ)+1.0
000172R 70. IF(IZ.EQ.KPP(IPP,4)) AZ(II,IZ)=AZ(II,IZ)+1.0
000200R 71. IF(IZ.EQ.KPP(IPP,5)) AZ(II,IZ)=AZ(II,IZ)+1.0
000206R 72. 438 CONTINUE
000210R 73. IF(LPN(5).EQ.0) GO TO 444
000211R 74. WRITE(IPRINT,440) (MS(I),I=1,6).
000222R 75. 440 FORMAT(/1X,21H1,I1,K1,I2,K2 ARE,6(I2,1X))
000222R 76. WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NORS)
000246R 77. 442 FORMAT(/1X,5HII = ,I2,1X,9HAC(II) = ,F4.1,2X,11HAA(II,1) = ,F4.1,2
               1X,11HAA(II,2) = ,F4.1,2X,13HAZ(II,IZ) ARE,10(F4.1,1X))
000246R 78. 444 LSKIP=1
000246R 79. 446 CONTINUE
000251R 80. 450 CONTINUE
000254R 81. 454 CONTINUE
000257R 82. GO TO 510
000260R 83. 460 IF(.NOT.(KS.EQ.1)) GO TO 480
000263R 84. II=II+1
000264R 85. AC(II)=AAC(5)
000265R 86. AA(II,1)=2.0
000266R 87. AA(II,2)=0.0
000267R 88. DO 470 IZ=1,NOBS
000272R 89. IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
000301R 90. IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
000307R 91. IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
000315R 92. IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
000323R 93. 470 CONTINUE
000325R 94. IF(LPN(5).EQ.0) GO TO 474

```

```

000326B      95.      WRITE(IPRINT,440) (MS(I),I=1,6)
000337B      96.      WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NORS)
000363B      97.      474 LSKIP=1
000363B      98.      GO TO 510
000364B      99.      480 II=II+1
000366B     100.      AC(II)=AAC(5)
000367B     101.      AA(II,1)=2.0
000370B     102.      AA(II,2)=0.0
000371B     103.      DO 490 IZ=1,NORS
000374B     104.      IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
000403B     105.      IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
000411B     106.      IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
000417B     107.      IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
000425B     108.      490 CONTINUE
000427B     109.      IF(LPN(5).EQ.0) GO TO 491
000430B     110.      WRITE(IPRINT,440) (MS(I),I=1,6)
000441B     111.      WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NOPS)
000465B     112.      491 LSKIP=1
000465B     113.      II=II+1
000467B     114.      AC(II)=AAC(5)
000470B     115.      AA(II,1)=2.0
000471B     116.      AA(II,2)=0.0
000472B     117.      DO 492 IZ=1,NORS
000475B     118.      IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
000504B     119.      IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
000512B     120.      IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
000520B     121.      IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0
000526B     122.      492 CONTINUE
000530B     123.      IF(LPN(5).EQ.0) GO TO 510
000531B     124.      WRITE(IPRINT,440) (MS(I),I=1,6)
000542B     125.      WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NORS)
000566B     126.      510 LEND=1
000566B     127.      IF(.NOT.((KP(IP,1).EQ.3.AND.KP(IP,4).EQ.3).AND.(KP(IP,2).EQ.2.OR.KP(IP,3).EQ.2.OR.KP(IP,5).EQ.2.OR.KP(IP,6).EQ.2))) GO TO 610
C           TO REPRESENT -4.0(SUM FROM H EQUAL TO P+1 TO N) (SUM FROM I EQUAL TO P+1 TO N) (SUM FROM J EQUAL 1 TO P,SUM FROM K EQUAL TO 1 TO P,SUM FROM L EQUAL TO 1 TO P.AND AT LEAST ONE OF J,K,L IS EQUAL TO P-KSTAR+1 OR P-KSTAR+2,...,OR P) OF A+HJLA+IKLZHIZJZK
C
000607B     128.      IF(.NOT.(KS.EQ.0)) GO TO 540
000613B     130.      IPP=0
000614B     131.      DO 536 IPERM=1,2
000616B     132.      IF(IPERM.EQ.1) GO TO 512
000620B     133.      IF(IPERM.EQ.2) GO TO 514
000621B     134.      512 IPER1=1
000622B     135.      IPER2=2
000623B     136.      GO TO 516
000624B     137.      514 IPER1=2
000625B     138.      IPER2=1
000626B     139.      516 LSKIP=1
000627B     140.      DO 534 J=1,2
000631B     141.      DO 532 K=1,2
000633B     142.      IF(N(IPER1,J,3).NE.N(IPER2,K,3)) GO TO 532
000642B     143.      IPP=IPP+1
000644B     144.      DO 518 IM=1,3
000646B     145.      KPP(IPP,IM)=N(IPER1,J,IM)
000653B     146.      IN=IM+3
000654B     147.      KPP(IPP,IN)=N(IPER2,K,IM)
000661B     148.      518 CONTINUE
000662B     149.      IF(IPP.EQ.1) GO TO 528
000663B     150.      KSAME=0
000664B     151.      IPPM=IPP-1
000665B     152.      DO 524 I=1,IPPM
000667B     153.      DO 520 L=1,6
000671B     154.      IF(KPP(IPP,L).NE.KPP(I,L)) GO TO 524

```

```

000671B      155.    520 CONTINUE
000700R      155.    KSAME=1
000704R      157.    GO TO 526
000704B      158.    524 CONTINUE
000705B      158.    526 IF(KSAME.EQ.0) GO TO 528
000711B      160.    IPP=IPP-1
000712R      161.    GO TO 532
000712B      162.    528 LSKIP=1
000713B      163.    II=II+1
000715R      164.    AC(II)=AAC(6)
000716B      165.    AA(II,1)=1.0
000717B      166.    AA(II,2)=1.0
000720R      167.    DO 530 IZ=1,N0BS
000723B      168.    IF(IZ.EQ.KPP(IPP,1)) AZ(II,IZ)=AZ(II,IZ)+1.0
000732B      169.    IF(IZ.EQ.KPP(IPP,2)) AZ(II,IZ)=AZ(II,IZ)+1.0
000740B      170.    IF(IZ.EQ.KPP(IPP,4)) AZ(II,IZ)=AZ(II,IZ)+1.0
000746B      171.    IF(IZ.EQ.KPP(IPP,5)) AZ(II,IZ)=AZ(II,IZ)+1.0
000754B      172.    530 CONTINUE
000756B      173.    IF(LPN(5).EQ.0) GO TO 531
000757B      174.    WRITE(IPRINT,440) (MS(I),I=1,6)
000770B      175.    WRITE(IPRINT,442) II,AC(II),AA(II,1),(AZ(II,IB),IB=1,N0RS)
001014B      176.    531 LSKIP=1
001014B      177.    532 CONTINUE
001017B      178.    534 CONTINUE
001022B      179.    536 CONTINUE
001025B      180.    GO TO 610
001026B      181.    540 IF(.NOT.(KS.EQ.1)) GO TO 560
001031B      182.    II=II+1
001032B      183.    AC(II)=AAC(6)
001033B      184.    AA(II,1)=2.0
001034B      185.    AA(II,2)=0.0
001035B      186.    DO 550 IZ=1,N0BS
001040B      187.    IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
001047B      188.    IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
001055B      189.    IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
001063B      190.    IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
001071B      191.
001071B      192.    550 CONTINUE
001073B      193.    IF(LPN(5).EQ.0) GO TO 554
001074B      194.    WRITE(IPRINT,440) (MS(I),I=1,6)
001105B      195.    WRITE(IPRINT,442) II,AC(II),AA(II,1),(AZ(II,IB),IB=1,N0RS)
001131B      196.    554 LSKIP=1
001131B      197.    GO TO 610
001132B      198.    560 II=II+1
001134B      199.    AC(II)=AAC(6)
001135B      200.    AA(II,1)=2.0
001136B      201.    AA(II,2)=0.0
001137B      202.    DO 570 IZ=1,N0RS
001142B      203.    IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
001151B      204.    IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
001157B      205.    IF(IZ.EQ.J1) AZ(II,IZ)=AZ(II,IZ)+1.0
001165B      206.    IF(IZ.EQ.J2) AZ(II,IZ)=AZ(II,IZ)+1.0
001173B      207.    570 CONTINUE
001175B      208.    IF(LPN(5).EQ.0) GO TO 574
001176B      209.    WRITE(IPRINT,440) (MS(I),I=1,6)
001207B      210.    WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,N0RS)
001233B      211.    574 LSKIP=1
001233B      212.    II=II+1
001235B      213.    AC(II)=AAC(6)
001236B      214.    AA(II,1)=2.0
001237B      215.    AA(II,2)=0.0
001240B      216.    DO 580 IZ=1,N0BS
001243B      217.    IF(IZ.EQ.I1) AZ(II,IZ)=AZ(II,IZ)+1.0
001252B      218.    IF(IZ.EQ.I2) AZ(II,IZ)=AZ(II,IZ)+1.0
001260B      219.    IF(IZ.EQ.K1) AZ(II,IZ)=AZ(II,IZ)+1.0
001266B      220.    IF(IZ.EQ.K2) AZ(II,IZ)=AZ(II,IZ)+1.0

```

```

001274R 221. 580 CONTINUE
001276R 222. IF(ILPN(5).EQ.0) GO TO 584-
001277R 223. WRITE(IPRINT,440) (MS(I),I=1,6)
001310R 224. WRITE(IPRINT,442) II,AC(II),AA(II,1),AA(II,2),(AZ(II,IB),IB=1,NORS)
001334R 225. 584 LSKIP=1
001334R 226. 610 LOUT=1
001335R 227. IIMAX=II
C
C      NOW WE HAVE AC(II),AA(II,1),AA(II,2) AND AZ(II,I2) WHERE II=1,IIMAX
C      AND IZ=1,NORS
C
001336R 228. CALL POWSUB
001341R 229. RETURN
001343R 230. END

C
C      ROUTINE POWSUB IS CALLED IN POWSUA
C
0000000B 1.   SUBROUTINE POWSUB
0000000B 2.   COMMON/MSKP/MS(6),KP(30,21)
0000000B 3.   COMMON/ACAAA/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
10,10)
0000000B 4.   COMMON/NKPP/N(2,2,3),KPP(8,6)
0000000B 5.   COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10,10),AZ1(4,
110,10)
0000000B 6.   COMMON/IGIGMX/IG(4),IGMAX(4)
0000000B 7.   COMMON/ACGAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
110,10)
0000000B 8.   COMMON/AKSTAR/AKSTAR(50),DEGREE(50),GAMMA(10)
0000000B 9.   COMMON/IREAD/IREAD,IPRINT,1TAPE
0000000B 10.  COMMON/NOBS/NOBS,NPAR,KSTAR,NPMK,NPMKP1,NPAPR
0000000B 11.  COMMON/KS/KS,IP,IIMAX,II
0000000B 12.  COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
0000000B 13.  COMMON/LPRINT/LPN(50)
0000000B 14.  COMMON/IQAB/IQA,IQB
0000000B 15.  COMMON/KPOWER/KPOWER
0000000B 16.  IF(.NOT.IIMAX.EQ.0) GO TO 628
000002B 17.  IP=IP-1
000005B 18.  GO TO 812
000005B 19.  628 LSKIP=1
C
C      TO REPRESENT THE EQUATION IN SECTION (4.5)
C
000006B 20.  DO 660 IV=1,IIMAX
000016B 21.  II=IV
000016B 22.  IF(NPMK.EQ.0) GO TO 633
000020B 23.  DO 632 IZ=1,NPMK
000027B 24.  SIZN(II,IZ)=1.0
000027B 25.  632 CONTINUE
000050B 25.  633 LSKIP=1
000053B 27.  DO 634 IZ=NPARP,NOBS
000066B 28.  SIZN(II,IZ)=1.0
000066B 29.  634 CONTINUE

```

```

0001078   29.      TS=0.0
0001108   31.      DO 654 IZ=NPMKP1,NPAR
0001238   32.      TS=TS+AZ(II,IZ)
0001328   33.      IF(INT(AZ(II,IZ)),EQ.0) GO TO 640
0001438   34.      NTEST=INT(AZ(II,IZ)/2.0)
0001528   35.      VTEST=NTEST
0001548   36.      VTEST=(AZ(II,IZ)/2.0)-VTEST
0001638   37.      IF(VTEST.GT.0.0) SIZN(II,IZ)=-1.0
0001758   38.      IF(.NOT.(VTEST.GT.0.0)) SIZN(II,IZ)=1.0
0002068   39.      GO TO 654
0002078   40.      SIZN(II,IZ)=1.0
0002208   41.      654 CONTINUE
0002258   42.      R(II)=IS
0002318   43.      IF(LPN(4),EQ.0) GO TO 659
0002338   44.      WRITE(IPRINT,805) (MS(I),I=1,6)
0002538   45.      IF(.NOT.II.EQ.1) GO TO 658
0002568   46.      WRITE(IPRINT,657)
0002658   47.      657 FORMAT(/1X,60HII,II,R(II),AC(II),AA(II,1),AA(II,2),SIZN(II,*),AZ(II,*) AKE)
0002658   48.      658 LSKIP=1
0002668   49.      WRITE(IPRINT,806) II,II,R(II),AC(II),AA(II,1),AA(II,2),((SIZN(II,I
1Z),AZ(II,IZ)),IZ=1,NOBS)
0003568   50.      659 LSKIP=1
0003578   51.      660 CONTINUE
C
C      TO REPRESENT FIRST PARTIAL DERIVATIVE OF (R+)**2 W.R.T. A+I1J1K1
C      EVALUATED AT A+=0,FIRST PARTIAL DERIVATIVE OF (R+)**2 W.R.T.
C      A+I2J2K2 EVALUATED AT A+=0 AND SECOND PARTIAL DERIVATIVE OF (R+)**2
C      W.R.T. A+I1J1K1,A+I2J2K2 EVALUATED AT A+=0 (C,F, SECTION (4,6))
C
0003658   52.      III=0
0003658   53.      DO 700 IV=1,IIMAX
0003748   54.      II=IV
0003748   55.      IF(.NOT.((INT(AA(II,1)),EQ.1,AND.INT(AA(II,2)),EQ.0).OR.(INT(AA(II
1,1)),EQ.0,AND.INT(AA(II,2)),EQ.1))) GO TO 700
0004208   56.      III=III+1
0004218   57.      DO 670 IRU=1,4
0004248   58.      VIRD=IRU
0004248   59.      D1(IRU,III)=R(II)+(VIRD-2.0)
0004408   60.      AC1(IRU,III)=0.5*VIRD*AC(II)
0004538   61.      AA1(IRU,III,1)=AA(II,1)
0004658   62.      AA1(IRU,III,2)=AA(II,2)
0004768   63.      DO 666 IZ=1,NOBS
0005068   64.      SIGN1(IRU,III,IZ)=SIZN(II,IZ)
0005248   65.      AZ1(IRU,III,IZ)=AZ(II,IZ)
0005418   66.      666 CONTINUE
0005448   67.      670 CONTINUE
0005478   68.      700 CONTINUE
0005548   69.      II1MAX=III
0005548   70.      IF(LPN(6),EQ.0) GO TO 709
0005568   71.      WRITE(IPRINT,803) (MS(I),I=1,6)
0005768   72.      WRITE(IPRINT,702)
0006058   73.      702 FORMAT(/1X,98HIRU,IRU,D1(IRU,III),AC1(IRU,III),AA1(IRU,III,1),AA1(
1IRU,III,2),SIGN1(IRU,III,*),AZ1(IRU,III,*)) AKE)
0006058   74.      DO 708 III=1,II1MAX
0006148   75.      DO 706 IRU=1,4
0006178   76.      WRITE(IPRINT,806) IRU,IRU,D1(IRU,III),AC1(IRU,III),AA1(IRU,III,1),
1AA1(IRU,III,2),((SIGN1(IRU,III,IB),AZ1(IRU,III,IB)),IB=1,NOBS)
0007058   77.      706 CONTINUE
0007108   78.      708 CONTINUE
0007158   79.      709 LSKIP=1
0007168   80.      III=II1MAX
0007208   81.      DO 730 IV=1,IIMAX
0007278   82.      II=IV
0007278   83.      IF(.NOT.((INT(AA(II,1)),EQ.1,AND.INT(AA(II,2)),EQ.0).OR.(INT(AA(II
1,1)),EQ.0,AND.INT(AA(II,2)),EQ.1))) GO TO 720
0007538   84.      D(II)=H(II)

```

```

000757B   85.    IF(INT(R(II)),EQ.0) GO TO 730
C          AC(II),AA(II,*),SIZN(II,IZ) AND AZ(II,IZ) ARE NOT CHANGED
C
000765B   86.    DO 714 III=1,IIIMAX
000774B   87.    III=III+1
000775B   88.    IRD=R(II)
001003B   89.    D(III)=D1(IRD,III)
001014B   90.    AC(III)=AC(II)*AC1(IRD,III)
001031B   91.    AA(III,1)=AA(II,1)+AA1(IRD,III,1)
001047B   92.    AA(III,2)=AA(II,2)+AA1(IRD,III,2)
001066B   93.    UO 710 IZ=1,NOBS
001077B   94.    SIZN(III,IZ)=SIZN(II,IZ)*SIGN1(IRD,III,IZ)
001123B   95.    AZ(III,IZ)=AZ(II,IZ)+AZ1(IRD,III,IZ)
001147B   96.    710 CONTINUE
001151B   97.    714 CONTINUE
001156B   98.    GO TO 730
001157B   99.    720 D(II)=R(II)

C          AC(II),AA(II,*),SIGN(II,IZ) AND AZ(II,IZ) ARE NOT CHANGED
C
001165B   100.   730 CONTINUE
001173B   101.   IIIMAX=III

C          PARTIAL DERIVATIVES OF (R+)**2 ARE INDEXED BY KKG
C          FIND IG(I)-TH TERM OF THE I-TH DERIVATIVES
C

001173B   102.   DO 740 J=1,4
001177B   103.   IG(J)=0
001177B   104.   740 CONTINUE
001213B   104.   DO 800 I=1,IIIMAX
001223B   106.   IF(INT(AA(I,1)),EQ.1.AND.INT(AA(I,2)),EQ.0) GO TO 750
001240B   107.   IF(INT(AA(I,1)),EQ.0.AND.INT(AA(I,2)),EQ.1) GO TO 760
001253B   108.   IF(INT(AA(I,1)),EQ.2.AND.INT(AA(I,2)),EQ.0) GO TO 770
001266B   109.   IF(INT(AA(I,1)),EQ.1.AND.INT(AA(I,2)),EQ.1) GO TO 776
001302B   110.   GO TO 800
001303B   111.   750 IG(1)=IG(1)+1
001305B   112.   IIIG=IG(1)
001305B   113.   KKG=1
001306B   114.   GO TO 780
001310B   115.   760 IG(2)=IG(2)+1
001312B   116.   IIIG=IG(2)
001312B   117.   KKG=2
001315B   118.   GO TO 780
001315B   119.   770 IG(3)=IG(3)+1
001317B   120.   IIIG=IG(3)
001317B   121.   KKG=3
001320B   122.   GO TO 780
001322B   123.   776 IG(4)=IG(4)+1
001324B   124.   IIIG=IG(4)
001324B   125.   KKG=4
001325B   126.   780 DG(KKG,IIIG)=D(I)
001342B   127.   IF(KKG,EQ.3) ACG(KKG,IIIG)=2.0*AC(I)
001356B   128.   IF(KKG,NE.3) ACG(KKG,IIIG)=AC(I)
001371B   129.   AAG(KKG,IIIG,1)=AA(I,1)
001402B   130.   AAG(KKG,IIIG,2)=AA(I,2)
001414B   131.   DO 790 IZ=1,NOBS
001423B   132.   SIGNG(KKG,IIIG,IZ)=SIZN(I,IZ)
001441B   133.   AZG(KKG,IIIG,IZ)=AZ(I,IZ)
001451B   134.   790 CONTINUE
001454B   135.   800 CONTINUE
001460B   136.   DO 802 I=1,4
001462B   137.   IGMAX(I)=IG(I)
001462B   138.   802 CONTINUE
001476B   138.   IF(ILPN(I),EQ.0) GO TO 812
001501B   140.   KPKINT=0
001501B   141.   DO 810 KKG=1,4

```

```

0015058 142. IF(IGMAX(KKG).EQ.0) GO TO 810
0015128 143. DO 808 IIIG=1,IGMAX(KKG)
0015258 144. KPRINT=KPRINT+1
0015268 145. IF(.NOT.(KPRINT.EQ.1)) GO TO 805
0015318 146. WRITE(IPRINT,803) (MS(I),I=1,6)
0015528 147. 803 FORMAT(/1X,21HII,J1,K1,I2,J2,K2 ARE,6(1Z,1X))
0015528 148. WRITE(IPRINT,804)
0015618 149. 804 FORMAT(/1X,83HKKG,IIIG,DG(KKG,IIIG),ACG(KKG,IIIG),AAG(KKG,IIIG,*),SIGN
1G(KKG,IIIG,*),AZG(KKG,IIIG,*)) ARE)
0015618 150. 805 LSKIP=1
0015628 151. WRITE(IPRINT,806) KKG,IIIG,DG(KKG,IIIG),ACG(KKG,IIIG),AAG(KKG,IIIG,1),
1 AAG(KKG,IIIG,2),((SIGNG(KKG,ITG,IZ),AZG(KKG,IIIG,IZ)),IZ=1,NOBS)
001661P 152. 806 FORMAT(/1X,2(1H/,I3),2(1H/,F4.1),1H/,2(F4.1,1X),10(1H/,F4.1,1X,F3.
11))
0016618 153. 808 CONTINUE
001665R 154. 810 CONTINUE
001670R 155. 812 LSKIP=1
001672B 156. IF(KPOWER.EQ.1) CALL PWSUC
0017018 157. IF(KPOWER.NE.1) CALL SIGSUC
001706B 158. RETURN
001710B 159. END

```

C C ROUTINE PWSUC IS CALLED IN PWSUB
C

```

000000B 1. SUBROUTINE PWSUC
000000B 2. COMMON/MSKP/MS(6),KP(30,21)
000000B 3. COMMON/ACAAA/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
10,10)
000000B 4. COMMON/NKPP/N(2,2,3),KPP(8,6)
000000B 5. COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10,10),AZ1(4,
110,10)
000000B 6. COMMON/IG1GMX/IG(4),IGMAX(4)
000000B 7. COMMON/ACGAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
110,10)
000000B 8. COMMON/AKSTAR/AKSTAR(50),DEGKEE(50),GAMMA(10)
000000B 9. COMMON/IREAD/IREAD,IPRINT,ITAPE
000000B 10. COMMON/NOBS/NOBS,NPAR,KSTAK,NPMK,NPMKP1,NPARP
000000B 11. COMMON/KS/KS,IP,IIMAX,II
000000B 12. COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
000000B 13. COMMON/LPRINT/LPN(50)
000000B 14. COMMON/IQAB/IQA,1QB
000000B 15. COMMON/MS123/MS1(4),MS2(4),MS3(4),MMS1(4),MMS2(4),MMS3(4)
000000B 16. IF(IIMAX.EQ.0) GO TO 1020

```

C C WE HAVE FIRST AND SECOND ORDER PARTIAL DERIVATIVES OF (R+)**2
C TO FIND PRODUCT OF THE TWO FIRST ORDER DERIVATIVES
C

```

0000018 17. IC=0
0000018 18. IA=1
0000028 19. IF(KS.EQ.0) JB=2
0000058 20. IF(KS.EQ.1) JB=1
0000108 21. IF(KS.EQ.2) JB=1
0000138 22. IF(.NOT.(KS.EQ.0)) GO TO 814
0000148 23. IGMAX2=IGMAX(2)
0000158 24. IF(IGMAX(1).EQ.0.OR.IGMAX(2).EQ.0) GO TO 864
0000208 25. GO TO 818
0000218 26. 814 LSKIP=1
0000228 27. IGMAX2=IGMAX(1)
0000248 28. IF(IGMAX2.EQ.0) GO TO 864
0000258 29. 818 LSKIP=1
0000278 30. DO 860 I=1,IGMAX(1)
0000408 31. DO 850 J=1,IGMAX2
0000508 32. IC=IC+1
0000518 33. D(IC)=DG(IA,I)+DG(JB,J)
0000708 34. AC(IC)=ACG(IA,I)*ACG(JB,J)
0001108 35. DO 822 IZ=1,NOBS

```

```

0001218      56.      SIZN(IC,IZ)=SIGNG(IA,I,IZ)*SIGNG(JB,J,IZ)
0001468      57.      AZ(IC,IZ)=AZG(IA,I,IZ)+AZG(JB,J,IZ)
0001748      58.      822 CONTINUE
0001768      59.      IF(IC.EQ.1) GO TO 830
0002008      60.      ICM1=IC-1
0002018      61.      ISAME=0
0002018      62.      DO 826 IAADD=1,ICM1
0002118      63.      IF(INT(U(IAADD)),NE.INT(D(IC))) GO TO 826
0002248      64.      DO 824 IZ=1,NOBS
0002358      65.      IF(INT(SIZN(IAADD,IZ)),NE.INT(SIZN(IC,IZ))) GO TO 826
0002568      66.      IF(INT(AZ(IAADD,IZ)),NE.INT(AZ(IC,IZ))) GO TO 826
0002568      67.      824 CONTINUE
0002718      68.      ISAME=1
0002768      69.      IAUDS=IAADD
0002778      70.      GO TO 828
0002778      71.      826 CONTINUE
0003018      72.      828 IF(ISAME.EQ.0) GO TO 830
0003108      73.      AC(IAUDS)=AC(IAUDS)+AC(IC)
0003218      74.      IC=IC-1
0003220      75.      830 LSKIP=1
0003248      76.      850 CONTINUE
0003328      77.      860 CONTINUE
0003368      78.      864 LSKIP=1
0003378      79.      ICMAXM=IC
C
C      NOW THE TERMS IN THE EXPRESSION OF THE PRODUCT OF THE FIRST
C      PARTIAL DERIVATIVES OF (R+)**2 W.R.T. A+I1J1K1 AND A+I2J2K2
C      EVALUATED AT A+=0 ARE REPRESENTED BY
C      ((D(JC),AC(JC),((SIZN(JC,IB),AZ(JC,IB)),IB=1,NOBS)),JC=1,ICMAXM)
C
0003418      80.      IF(ICMAXM,EQ.0) GO TO 869
0003428      81.      IF(ILPN(11).EQ.0) GO TO 869
0003438      82.      WRITE(1PRINT,865) (MS(IB),IB=1,6)
0003638      83.      DO 868 JC=1,ICMAXM
0003728      84.      WRITE(1PRINT,866) JC,D(JC),AC(JC),((SIZN(JC,IB),AZ(JC,IB)),IB=1,NO
     1BS)
0004418      85.      868 CONTINUE
0004458      86.      865 FORMAT(1X,21HI1,J1,K1,I2,J2,K2 AHE,6(I2,1X))
0004458      87.      866 FORMAT(1X,1H/,3HIC=,12,1H/,6HU(IC)=,F4.1,1H/,7HAC(IC)=,F4.1/1X +2
     13HSIZN(IC,*),AZ(IC,*) AHE,10(F4.1,1H*,F3.1,1H/))
0004458      88.      869 LSKIP=1
0004468      89.      IF(ICMAXM,EQ.0) GO TO 930
0004508      90.      DO 924 IC=1,ICMAXM
0004578      91.      DO 882 I=1,4
0004628      92.      MS1(I)=0
0004668      93.      MS2(I)=0
0004728      94.      MS3(I)=0
0004768      95.      MMS1(I)=0
0005028      96.      MMS2(I)=0
0005068      97.      MMS3(I)=0
0005128      98.      882 CONTINUE
0005158      99.      IH=0
0005158     100.      IF(NPMK.EQ.0) GO TO 887
0005178     101.      DO 886 IZ=1,NPMK
0005268     102.      IF(INT(AZ(IC,IZ)),EQ.0) GO TO 886
0005378     103.      DO 884 I=1,6
0005428     104.      IF(MS(I).NE.IZ) GO TO 884
0005508     105.      IH=IH+1
0005518     106.      MS1(IH)=I
0005568     107.      MMS1(IH)=INT(AZ(IC,IZ))
0005638     108.      GO TO 886
0005638     109.      884 CONTINUE
0005658     110.      886 CONTINUE
0005758     111.      887 LSKIP=1
0005768     112.      IH=0
0005778     113.      DO 890 IZ=NPMKP1,NPAR
0006128     114.      IF(INT(AZ(IC,IZ)),EQ.0) GO TO 890

```

```

000623B      95.    DO 888 I=1,6
000626B      96.    IF(MS(I).NE.IZ) GO TO 888
000634B      97.    IH=IH+1
000635B      98.    MS2(IH)=I
000642B      99.    MMS2(IH)=INT(AZ(IC,IZ))
000647B     100.   GO TO 890
000647B     101.   888 CONTINUE
000651B     101.   890 CONTINUE
000661B     103.   IH=0
000661B     104.   DO 894 IZ=NPARP,NOBS
000674B     105.   IF(INT(AZ(IC,IZ)).EQ.0) GO TO 894
000705B     106.   DO 892 I=1,6
000710B     107.   IF(MS(I).NE.IZ) GO TO 892
000716B     108.   IH=IH+1
000717B     109.   MS3(IH)=I
000724B     110.   MMS3(IH)=INT(AZ(IC,IZ))
000731B     111.   GO TO 894
000731B     112.   892 CONTINUE
000733B     112.   894 CONTINUE
C
C      NOW A TERM IN THE EXPRESSION OF THE PRODUCT OF THE FIRST PARTIAL
C      DERIVATIVES OF (R+)**2 W.R.T. A+I1J1K1 AND A+I2J2K2 EVALUATED AT
C      A+=0 IS REPRESENTED BY
C      D(IC),AC(IC),((MS1(I),MMS1(I)),I=1,4),((MS2(I),MMS2(I)),I=1,4)
C      AND ((MS3(I),MMS3(I)),I=1,4)
C
000743B     114.   IWA=IWA+1
000744B     115.   IF(LPN(112).EQ.0) GO TO 901
000745B     116.   WRITE(1FH,900) IWA,((MS1(IB),MMS1(IB)),IB=1,4),((MS2(IB),MMS2(IB)),IB=1,4),((MS3(IB),MMS3(IB)),IB=1,4)
001027B     117.   900 FORMAT(1X,4HIWA=,I5,22HMS1(*).HMS1(*) ETC ARE,1H/.4(I1,1H*,I1,1H/),
               1,2H//,4(I1,1H*,I1,1H/),2H//,4(I1,1H*,I1,1H/),2H//)
001027B     118.   901 LSKIP=1
001030B     119.   ITAPE9=9
001032B     120.   WRITE(11TAPE9,910) IWA,ICMAXM,(MS(IB),IB=1,6),(KP(IP,IB),IB=1,21),(
               1(MS1(IB),MMS1(IB)),IB=1,4),((MS2(IB),MMS2(IB)),IB=1,4),((MS3(IB),M
               MS3(IB)),IB=1,4),D(IC),AC(IC)
001160B     121.   910 FORMAT(1X,I5,I5,5I2,2F5.1)
001160B     122.   924 CONTINUE
001164B     123.   930 LSKIP=1
C
C      TO REPRESENT THE EXPRESSION OF THE SECOND PARTIAL DERIVATIVE OF
C      (R+)**2 W.R.T. A+I1J1K1 AND A+I2J2K2 EVALUATED AT A+=0
C
001165B     124.   IF(KS.EQ.0) IGMAX4=IGMAX(4)
001171B     125.   IF(KS.EQ.1.OR.KS.EQ.2) IGMAX4=IGMAX(3)
001176B     126.   IF(IGMAX4.EQ.0) GO TO 986
001177B     127.   IC=0
001177B     128.   DO 984 I=1,IGMAX4
001206B     129.   IC=IC+1
001207B     130.   IF(KS.EQ.0) D(IC)=DG(4,I)
001224B     131.   IF(KS.EQ.1.OR.KS.EQ.2) D(IC)=DG(3,I)
001242B     132.   IF(KS.EQ.0) AC(IC)=ACG(4,I)
001256B     133.   IF(KS.EQ.1.OR.KS.EQ.2) AC(IC)=ACG(3,I)
001274B     134.   DO 970 IZ=1,NOBS
001303B     135.   IF(KS.EQ.0) SIZN(IC,IZ)=SIGNG(4,I,IZ)
001323B     136.   IF(KS.EQ.1.OR.KS.EQ.2) SIZN(IC,IZ)=SIGNG(3,I,IZ)
001344B     137.   IF(KS.EQ.0) AZ(IC,IZ)=AZG(4,I,IZ)
001363B     138.   IF(KS.EQ.1.OR.KS.EQ.2) AZ(IC,IZ)=AZG(3,I,IZ)
001405B     139.   970 CONTINUE
001411B     140.   IF(IC.EQ.1) GO TO 984
001412B     141.   ICM1=IC-1
001413B     142.   ISAME=0
001414B     143.   DO 976 IAUD=1,ICM1
001423B     144.   IF(INT(D(IAUD)).NE.INT(D(IC))) GO TO 976
001436B     145.   DO 974 IZ=1,NOBS
001445B     146.   IF(INT(SIZN(IAUD,IZ)).NE.INT(SIZN(IC,IZ))) GO TO 976

```

```

001470B 147. IF(INT(AZ(IADD,IZ)).NE.INT(AZ(IC,IZ))) GO TO 976
001470B 148. 974 CONTINUE
001503B 148. ISAME=1
001510B 150. IAUDS=IADD
001511B 151. GO TO 978
001511B 152. 976 CONTINUE
001513B 152. 978 IF(ISAME.EQ.0) GO TO 984
001522B 154. AC(IAUDS)=AC(IAUD)+AC(IC)
001556B 155. IC=IC-1
001540B 156. 984 CONTINUE
001545B 157. LSKIP=1
001546B 158. ICMAX=IC

C
C      NOW THE TERMS IN THE EXPRESSION OF THE SECOND PARTIAL DERIVATIVE
C      OF (R+)**2 W.R.T. A+I1J1K1 AND A+I2J2K2 EVALUATED AT A+=0
C      ARE REPRESENTED BY
C      (D(IC),AC(IC),((SIZN(IC,IZ),AZ(IC,IZ)),IZ=1,NOBS))
C
001550B 159. IF(ICMAX.NE.0) GO TO 987
001551B 160. GO TO 1020
001552B 161. 987 LSKIP=1
001554B 162. IF(LPN(11).EQ.0) GO TO 996
001556B 163. WRITE(IPRINT,990)
001565B 164. 990 FORMAT(/1X,22HSECOND DERIVATIVES ARE)
001565B 165. WRITE(IPRINT,865) (MS(IB),IB=1,6)
001605B 166. DO 994 IC=1,ICMAX
001614B 167. WRITE(IPRINT,866) IC,D(IC),AC(IC),((SIZN(IC,IB),AZ(IC,IB)),IB=1,NO
               185)

001663B 168. 994 CONTINUE
001667B 169. 996 LSKIP=1
001670B 170. DO 1014 IC=1,ICMAX
001700B 171. DO 997 I=1,4
001703B 172. MS1(I)=0
001707B 173. MS2(I)=0
001713B 174. MS3(I)=0
001717B 175. MMS1(I)=0
001723B 176. MMS2(I)=0
001727B 177. MMS3(I)=0
001733B 178. 997 CONTINUE
001736B 179. IH=0
001736B 180. IF(NPMK.EQ.0) GO TO 1001
001740B 181. DO 1000 IZ=1,NPMK
001747B 182. IF(INT(AZ(IL,IZ)).EQ.0) GO TO 1000
001760B 183. DO 998 I=1,6
001763B 184. IF(MS(I).NE.IZ) GO TO 998
001771B 185. IH=IH+1
001772B 186. MS1(IH)=I
001777B 187. MMS1(IH)=INT(AZ(IC,IZ))
002004B 188. GO TO 1000
002004B 189. 998 CONTINUE
002006B 190. 1000 CONTINUE
002016B 191. 1001 LSKIP=1
002017B 192. IH=0
002020B 193. DO 1004 IZ=NPMKP1,NPAR
002033B 194. IF(INT(AZ(IC,IZ)).EQ.0) GO TO 1004
002044B 195. DO 1002 I=1,6
002047B 196. IF(MS(I).NE.IZ) GO TO 1002
002055B 197. IH=IH+1
002056B 198. MS2(IH)=I
002063B 199. MMS2(IH)=INT(AZ(IC,IZ))
002070B 200. GO TO 1004
002070B 201. 1002 CONTINUE
002072B 201. 1004 CONTINUE
002102B 203. IH=0
002102B 204. DO 1008 IZ=NPARP,NOBS
002115B 205. IF(INT(AZ(IL,IZ)).EQ.0) GO TO 1008
002126B 206. DO 1006 I=1,6

```

```

0021318 207. IF(MS(I).NE.IZ) GO TO 1006
0021378 208. IH=IH+1
0021408 209. MS3(IH)=1
0021458 210. MMS3(IH)=INT(AZ(IC,IH))
0021528 211. GO TO 1008
0021528 212. 1006 CONTINUE
0021548 212. 1008 CONTINUE
C
C      NOW A TERM IN THE EXPRESSION OF THE SECOND PARTIAL DERIVATIVE OF
C      (R*)**2 W.R.T. A*I1J1K1 AND A*I2J2K2 EVALUATED AT A+=0 IS
C      REPRESENTED BY
C      D(IC),AC(IC),((MS1(I)),MMS1(I)),I=1,4),((MS2(I)),MMS2(I)),I=1,4)
C      AND ((MS3(I)),MMS3(I)),I=1,4)
C
0021648 214. IWB=IWB+1
0021658 215. IF(LPN(112).EQ.0) GO TO 1010
0021668 216. WRITE(IPRINT,1009) IQB,((MS1(IB)),MMS1(IB)),IB=1,4),((MS2(IB)),MMS2(IB)),IB=1,4)
0022508 217. 1009 FORMAT(1X,4HI0B=,I5,22HMS1(*),MMS1(*)) ETC AKE:1H/,4(I1,1H*,I1,1H/)
1.2H//,4(I1,1H*,I1,1H/),2H//,4(I1,1H*,I1,1H/),2H//)
0022508 218. 1010 LSKIP=1
0022518 219. ITAPE1U=10
0022538 220. WRITE(11APE1U,910) IQB,ICMAX ,((MS1(IB)),IB=1,6),(KP(1P,IB),IB=1,21),(
1,(MS1(IB)),MMS1(IB)),IB=1,4),((MS2(IB)),MMS2(IB)),IB=1,4),((MS3(IB)),M
1MS3(IB)),IB=1,4),D(IC),AC(IC)
0024018 221. 1014 CONTINUE
0024058 222. 1020 LSKIP=1
0024068 223. RETURN
0024118 224. END
C
C      ROUTINE SIGSUC IS CALLED IN POWSUB AND IT IS FOR DERIVING
C      I1(THETA,SIGMA)
C
0000008 1. SUBROUTINE SIGSUC
0000008 2. COMMON/MSKP/MS(6),KP(30,21)
0000008 3. COMMON/ACAAZ/AC(30),R(30),D(30),AAC(30),AA(30,2),SIZN(30,10),AZ(3
10,10)
0000008 4. COMMON/NKPP/N(2,2,3),KPP(8,6)
0000008 5. COMMON/AC1AA1/AC1(4,10),D1(4,10),AA1(4,10,2),SIGN1(4,10,10),AZ1(4,
110,10)
0000008 6. COMMON/IGIGMX/IG(4),IGMAX(4)
0000008 7. COMMON/ACGAAG/ACG(4,10),DG(4,10),AAG(4,10,2),SIGNG(4,10,10),AZG(4,
110,10)
0000008 8. COMMON/AKSTAR/AKSTAR(50),DEGREE(50),GAMMA(10)
0000008 9. COMMON/IREAD/IREAD,IPRINT,ITAPE
0000008 10. COMMON/NOBS/NOBS,NPAR,KSTAR,NPMK,NPMKP1,NPARP
0000008 11. COMMON/KS/KS,IP,IIMAX,II
0000008 12. COMMON/I1J1K1/I1,J1,K1,I2,J2,K2
0000008 13. COMMON/LPRINT/LPN(50)
0000008 14. COMMON/IQAB/IQA,IQB
0000008 15. IF(IIMAX.EQ.0) GO TO 1010
C
C      TO REPRESENT (((KSTAR/2)-1)/((DE)**2))-0.5)*FIRST DERIVATIVE OF
C      (R*)**2 W.R.T. A*I1J1K1 EVALUATED AT A+=0*FIRST DERIVATIVE OF
C      (R*)**2 W.R.T. A*I2J2K2 EVALUATED AT A+=0
C
0000018 16. IC=0
0000018 17. IA=1
0000028 18. IF(KS.EQ.0) JB=2
0000058 19. IF(KS.EQ.1) JB=1
0000108 20. IF(KS.EQ.2) JB=1
0000138 21. IF(.NOT.(KS.EQ.0)) GO TO 814
0000148 22. IGMAX2=IGMAX(2)
0000158 23. IF(IGMAX(1).EQ.0.OR.IGMAX(2).EQ.0) GO TO 864
0000208 24. GO TO 818
0000218 25. 814 LSKIP=1

```

```

000022B   26.      IGMAX2=IGMAX(1)
000024B   27.      IF(IGMAX2.EQ.0) GO TO 864
000025B   28.      LSKIP=1
000027B   29.      DO 860 I=1,IGMAX(1)
000040B   30.      DO 850 J=1,IGMAX2
000050B   31.      IC=IC+1
000051B   32.      AKSTAR(IC)=0.0
000055B   33.      D(IC)=DG(IA,I)+DG(JB,J)
000075B   34.      AC(IC)=-0.5*ACG(IA,I)*ACG(JB,J)
000116B   35.      DO 822 IZ=1,NOBS
000127B   36.      SIZN(IC,IZ)=SIGNG(IA,I,IZ)*SIGNG(JB,J,IZ)
000154B   37.      AZ(IC,IZ)=AZG(IA,I,IZ)+AZG(JB,J,IZ)
000202B   38.      822 CONTINUE
000204B   39.      IF(IC.EQ.1) GO TO 830
000206B   40.      ICM1=IC-1
000207B   41.      ISAME=0
000207B   42.      DO 826 IADD=1,ICM1
000217B   43.      IF(INT(AKSTAR(IADD)).NE.INT(AKSTAR(IC))) GO TO 826
000232B   44.      IF(INT(D(IADD)).NE.INT(D(IC))) GO TO 826
000244B   45.      DO 824 IZ=1,NOBS
000254B   46.      IF(INT(SIZN(IADD,IZ)).NE.INT(SIZN(IC,IZ))) GO TO 826
000276B   47.      IF(INT(AZ(IADD,IZ)).NE.INT(AZ(IC,IZ))) GO TO 826
000276B   48.      824 CONTINUE
000311B   48.      ISAME=1
000316B   50.      IADDUS=IADD
000317B   51.      GO TO 828
000317B   52.      826 CONTINUE
000321B   52.      828 IF(ISAME.EQ.0) GO TO 830
000330B   54.      AC(IADDUS)=AC(IADDUS)+AC(IC)
000341B   55.      IC=IC-1
000342B   56.      830 IC=IC+1
000345B   57.      AKSTAR(IC)=0.0
000351B   58.      D(IC)=DG(IA,I)+DG(JB,J)-2.0
000373B   59.      AC(IC)=-ACG(IA,I)*ACG(JB,J)
000414B   60.      DO 832 IZ=1,NOBS
000425B   61.      SIZN(IC,IZ)=SIGNG(IA,I,IZ)*SIGNG(JB,J,IZ)
000452B   62.      AZ(IC,IZ)=AZG(IA,I,IZ)+AZG(JB,J,IZ)
000500B   63.      832 CONTINUE
000502B   64.      IF(IC.EQ.2) GO TO 840
000504B   65.      ICM1=IC-1
000505B   66.      ISAME=0
000505B   67.      DO 836 IADD=1,ICM1
000515B   68.      IF(INT(AKSTAR(IADD)).NE.INT(AKSTAR(IC))) GO TO 836
000530B   69.      IF(INT(D(IADD)).NE.INT(D(IC))) GO TO 836
000542B   70.      DO 834 IZ=1,NOBS
000551B   71.      IF(INT(SIZN(IADD,IZ)).NE.INT(SIZN(IC,IZ))) GO TO 836
000574B   72.      IF(INT(AZ(IADD,IZ)).NE.INT(AZ(IC,IZ))) GO TO 836
000574B   73.      834 CONTINUE
000607B   73.      ISAME=1
000614B   75.      IADDUS=IADD
000615B   76.      GO TO 838
000615B   77.      836 CONTINUE
000617B   77.      838 IF(ISAME.EQ.0) GO TO 840
000626B   79.      AC(IADDUS)=AC(IADDUS)+AC(IC)
000637B   80.      IC=IC-1
000640B   81.      840 IC=IC+1
000643B   82.      AKSTAR(IC)=1.0
000647B   85.      D(IC)=DG(IA,I)+DG(JB,J)-2.0
000671B   84.      AC(IC)=0.5*ACG(IA,I)*ACG(JB,J)
000713B   85.      DO 842 IZ=1,NOBS
000723B   86.      SIZN(IC,IZ)=SIGNG(IA,I,IZ)*SIGNG(JB,J,IZ)
000750B   87.      AZ(IC,IZ)=AZG(IA,I,IZ)+AZG(JB,J,IZ)
000776B   88.      842 CONTINUE
001000B   89.      IF(IC.EQ.3) GO TO 850
001002B   90.      ICM1=IC-1
001003B   91.      ISAME=0
001003B   92.      DO 846 IADD=1,ICM1

```

```

001013B      93.    IF(INT(AKSTAR(IADD)).NE.INT(AKSTAR(IC))) GO TO 846
001026B      94.    IF(INT(U(IADD)).NE.INT(D(IC))) GO TO 846
001040B      95.    DO 844 IZ=1,NOBS
001047B      96.    IF(INT(SIZN(IADD,IZ)).NE.INT(SIZN(IC,IZ))) GO TO 846
001072B      97.    IF(INT(AZ(IADD,IZ)).NE.INT(AZ(IC,IZ))) GO TO 846
001072B     98.    844 CONTINUE
001105B     98.    ISAME=1
001112B    100.    IAUDS=IADD
001113B    101.    GO TO 848
001113B    102.    846 CONTINUE
001115B    102.    848 IF(ISAME.EQ.0) GO TO 850
001124B    104.    AC(IADD)=AC(IAUDS)+AC(IC)
001135B    105.    IC=IC-1
001136B    106.    850 CONTINUE
001143B    107.    860 CONTINUE
001147B    108.    864 LSKIP=1
001150B    109.    ICMAXM=IC
001152B    110.    IF(ICMAXM.EQ.0) GO TO 868
001153B    111.    IF(LPN(7).EQ.0) GO TO 868
001154B    112.    WRITE(1PRINT,865) (MS(IB),IB=1,6)
001174B    113.    865 FORMAT(/1X,21H1.J1.K1,I2,J2,K2 ARE,6(I2,1X))
001174B    114.    DO 867 JC=1,ICMAXM
001203B    115.    WRITE(1PRINT,866) JC,AKSTAR(JC),D(JC),AC(JC),((SIZN(JC,IB),AZ(JC,I
1B)),IB=1,NOBS)
001257B    116.    866 FORMAT(/1X,1H/,3HIC=,I2,1H/,11HAKSTAR(IC)=,F4.1,1H/,6HD(IC)=,F4.1,
1H/,7HAC(IC)=,F4.1/ 1X,23HSIZN(IC,*),AZ(IC,*) ARE,10(F4.1,F3.1,1H/
1))
001257B    117.    867 CONTINUE
001263B    118.    868 LSKIP=1
C
C      TO REPRESENT SECOND PARTIAL DERIVATIVE OF (R*)**2 W.R.T.
C      A*I1J1K1+A*I2J2K2 EVALUATED AT A*=0
C
001264B    119.    IF(KS.EQ.0) IGMAX4=IGMAX(4)
001270B    120.    IF(KS.EQ.1.OR.KS.EQ.2) IGMAX4=IGMAX(3)
001275B    121.    IF(IGMAX4.EQ.0) GO TO 886
001276B    122.    DO 884 I=1,IGMAX4
001304B    123.    IC=IC+1
001305B    124.    AKSTAR(IC)=0.0
001311B    125.    IF(KS.EQ.0) D(IC)=DG(4,I)
001326B    126.    IF(KS.EQ.1.OR.KS.EQ.2) D(IC)=DG(3,I)
001344B    127.    IF(KS.EQ.0) AC(IC)=ACG(4,I)
001360B    128.    IF(KS.EQ.1.OR.KS.EQ.2) AC(IC)=ACG(3,I)
001376B    129.    DO 870 IZ=1,NOBS
001405B    130.    IF(KS.EQ.0) SIZN(IC,IZ)=SIGNG(4,I,IZ)
001425B    131.    IF(KS.EQ.1.OR.KS.EQ.2) SIZN(IC,IZ)=SIGNG(3,I,IZ)
001446B    132.    IF(KS.EQ.0) AZ(IC,IZ)=AZG(4,I,IZ)
001465B    133.    IF(KS.EQ.1.OR.KS.EQ.2) AZ(IC,IZ)=AZG(3,I,IZ)
001507B    134.    870 CONTINUE
001513B    135.    IF(IC.EQ.1) GO TO 884
001514B    136.    ICM1=IC-1
001515B    137.    ISAME=0
001516B    138.    DO 876 IADD=1,ICM1
001525B    139.    IF(INT(U(IADD)).NE.INT(D(IC))) GO TO 876
001540B    140.    IF(INT(AKSTAR(IADD)).NE.INT(AKSTAR(IC))) GO TO 876
001552B    141.    DO 874 IZ=1,NOBS
001561B    142.    IF(INT(SIZN(IADD,IZ)).NE.INT(SIZN(IC,IZ))) GO TO 876
001604B    143.    IF(INT(AZ(IADD,IZ)).NE.INT(AZ(IC,IZ))) GO TO 876
001604B    144.    874 CONTINUE
001617B    144.    ISAME=1
001624B    146.    IAUDS=IADD
001625B    147.    GO TO 878
001625B    148.    876 CONTINUE
001627B    148.    878 IF(ISAME.EQ.0) GO TO 884
001636B    150.    AC(IADD)=AC(IAUD)+AC(IC)
001652B    151.    IC=IC-1
001654B    152.    884 CONTINUE

```

```

0016618   153.  886 LSKIP=1
0016628   154.  ICMAX=IC
0016648   155.  IF(ICMAX.NE.0) GO TO 887
0016658   156.  IP=IP-1
0016668   157.  GO TO 1010
0016703   158.  887 LSKIP=1
0016718   159.  IF(ICMAX.LE.ICMAXM) GO TO 889
0016748   160.  IF(LPN(7).EQ.0) GO TO 889
0016768   161.  WRITE(IPRINT,865) (MS(IB),IB=1,6)
0017168   162.  ICMXMP=ICMAXM+1
0017178   163.  DO 888 IC=1,ICMAX
0017268   164.  WRITE(IPRINT,866) IC,AKSTAR(IC),D(IC),AC(IC),((SIZN(IC,IB),AZ(IC,I
1B)),IB=1,NOBS)
0020028   165.  888 CONTINUE
0020068   166.  889 LSKIP=1
C
C      NOW (((KSTAR/2)-1)/((D*)**2))-0.5)*FIRST PARTIAL DERIVATIVE OF
C      (R*)**2 W.R.T. A*I1J1K1 EVALUATED AT A*=0*FIRST PARTIAL DERIVATIVE
C      OF (R*)**2 W.R.T. A*I2J2K2 EVALUATED AT A*=0+SECOND PARTIAL
C      DERIVATIVE OF (R*)**2 W.R.T. A*I1J1K1,A*I2J2K2 EVALUATED AT A*=0
C      ARE REPRESENTED BY
C      (AKSTAR(IC),D(IC),AC(IC),SIZN(IC,IZ),AZ(IC,IZ),WHERE IC=1,ICMAX)
C
C      TO REPRESENT I1A*I1J1K1A*I2J2K2
C
0020078   167.  DO 960 IC=1,ICMAX
0020178   168.  TP=1.0
0020178   169.  IF(NPMK.EQ.0) GO TO 894
0020228   170.  DO 890 IZ=1,NPMK
0020308   171.  IF(INT(AZ(IC,IZ)).EQ.0) GO TO 890
0020418   172.  IF(INT(AZ(IC,IZ)).EQ.1) TP=0.0
0020458   173.  IF(INT(AZ(IC,IZ)).EQ.2) GO TO 890
0020468   174.  IF(INT(AZ(IC,IZ)).EQ.3) TP=0.0
0020528   175.  IF(INT(AZ(IC,IZ)).EQ.4) TP=3.0*TP
0020568   176.  890 CONTINUE
0020628   177.  894 LSKIP=1
0020638   178.  DO 900 IZ=NPARP,NOBS
0020768   179.  IF(INT(AZ(IC,IZ)).EQ.0) GO TO 900
0021078   180.  IF(INT(AZ(IC,IZ)).EQ.1) TP=0.0
0021138   181.  IF(INT(AZ(IC,IZ)).EQ.2) GO TO 900
0021148   182.  IF(INT(AZ(IC,IZ)).EQ.3) TP=0.0
0021208   183.  IF(INT(AZ(IC,IZ)).EQ.4) TP=3.0*TP
0021248   184.  900 CONTINUE
0021308   185.  AC(IC)=TP*AC(IC)
0021358   186.  KSIGN=0
0021358   187.  DO 910 IZ=NPMKP1,NPAR
0021508   188.  IF(INT(SIZN(IC,IZ)).NE.-1) GO TO 910
0021618   189.  KSIGN=1
0021628   190.  GO TO 920
0021628   191.  910 CONTINUE
0021648   191.  920 IF(KSIGN.EQ.0) GO TO 930
0021728   193.  AC(IC)=0.0
0021778   194.  GO TO 960
0022018   195.  930 TP=1.0
0022038   196.  IS=0.0
0022038   197.  RTPI=SQRT(3.1415926536)
0022108   198.  GAMMA(1)=RTPI
0022108   199.  GAMMA(2)=1.0
0022118   200.  GAMMA(3)=RTPI/2.0
0022138   201.  GAMMA(4)=2.0
0022158   202.  GAMMA(5)=3.0*RTPI/4.0
0022178   203.  GAMMA(6)=2.0
0022208   204.  GAMMA(7)=15.0*RTPI/8.0
0022238   205.  T1=0.3989422804
0022248   206.  DO 950 IZ=NPMKP1,NPAR
0022378   207.  T2=0.5*(AZ(IC,IZ)+1.0)

```

```

002247B 208.    IGAM=AZ(IC,I2)+1.0
002250B 209.    TP=TP*(2.0**T2)*GAMMA(IGAM)*T1
002262B 210.    TS=TS+AZ(IC,I2)
002264B 211.    950 CONTINUE
002266B 212.    AC(IC)=TP*AC(IC)
002274B 213.    DEGREE(IC)=TS
002301B 214.    IF(.NOT.(INT(DEGREE(IC)),EQ.2.AND.INT(AKSTAR(IC)),EQ.1)) GO TO 96
10
002314B 215.    AKSTAR(IC)=0.0
002320B 216.    DEGREE(IC)=0.0
002324B 217.    960 CONTINUE
002331B 218.    IF(IPN(8),EQ.0) GO TO 967
002332B 219.    WRITE(1PRINT,865) (MS(I8),I8=1,6)
002352B 220.    DO 964 IC=1,ICMAX
002361B 221.    WRITE(1PRINT,962) IC,AKSTAR(IC),D(IC),AC(IC),DEGREE(IC)
002416B 222.    962 FORMAT(1X,5HIC=,I2,1H/,11HAKSTAR(IC)=,F4.1,1H/,6HU(IC)=,F4.1,1H/,17HAC(IC)=,F4.1,1H/,11HDEGREE(IC)=,F4.1)
002416B 223.    964 CONTINUE
002422B 224.    967 LSKIP=1

C
C   THE IC-TH TERM IN I1A*I1J1K1A*I2J2K2 IS OF THE FORM
C   (KSTAR**AKSTAR(IC))*((D**2)**D(IC))*(AC(IC)*DENSITY OF CHI SQUARE
C   DISTRIBUTION ON KSTAR DEGREES OF FREEDOM EVALUATED AT (D**2)**2)*
C   (DENSITY OF CHI SQUARE DISTRIBUTION ON KSTAR + DEGREE(IC) DEGREES
C   OF FREEDOM EVALUATED AT 1)/(DENSITY OF CHI SQUARE DISTRIBUTION ON
C   KSTAR DEGREES OF FREEDOM EVALUATED AT 1)
C

002423B 225.    IIC=1
002424B 226.    IF(ICMAX,EQ.1) GO TO 978
002426B 227.    DO 976 IC=2,ICMAX
002436B 228.    DO 968 IADD=1,IIC
002446B 229.    IF(INT(AKSTAR(IADD)),NE.INT(AKSTAR(IC)),OR.INT(D(IADD)),NE.INT(D(I
1C)),OR.INT(DEGREE(IADD)),NE.INT(DEGREE(IC))) GO TO 968
      AC(IADD)=AC(IADD)+AC(IC)
      GO TO 976
002501B 230.    968 CONTINUE
002512B 231.    IIC=IIC+1
002514B 232.    AC(IIC)=AC(IC)
002520B 233.    AKSTAR(IIC)=AKSTAR(IC)
002530B 234.    D(IIC)=D(IC)
002541B 235.    DEGREE(IC)=DEGREE(IC)
002552B 236.    976 CONTINUE
002564B 237.    978 IICMAX=IIC
002572B 238.    978 KZERO=0
002574B 239.    979 FORMAT(1X,68HIP,I1,J1,K1,I2,J2,K2,KP(IP,*),AKSTAR(IC),D(IC),AC(IC
1),DEGREE(IC) ARE)
002574B 240.    DO 1000 IC=1,IICMAX
002604B 241.    IF(INT(AC(IC)),EQ.0) GO TO 1000
002612B 242.    WRITE(1PRINT,979)
002621B 243.    WRITE(1PRINT,980) IP,(MS(I),I=1,6),(KP(IP,I),I=1,21),AKSTAR(IC),D(
1C),AC(IC),DEGREE(IC)
002705B 244.    980 FORMAT(1X,13,1H/,6(I2,1X),1H/,21(I1,1X),4(1H/,F10.5))
002705B 245.    KZERO=1
002705B 246.    1000 CONTINUE
002712B 247.    IF(KZERO,EQ.1) GO TO 1010
002713B 248.    IP=IP-1
002715B 249.    1010 LSKIP=1
002717B 250.    RETURN
002722B 251.    252.    END

```

APPENDIX 4 Programs POWCAL, and subroutines COEF11, COEF2, COEF1,E1000, E2000, etc.

```

C
C   PROGRAM POWCAL,SUBROUTINES COEF11,COEF2,COEF1,E1000+E2000,ETC ARE
C   FOR CALCULATING BETA1(THETA A,SIGMA A)
C
000000R 1.    PROGRAM POWCAL(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT,TAPE9,TAPE10)
0101008 2.    COMMON/MSKPAB/MSA(23,6),MSB(73,6),KPA(23,21),KPB(73,21)
0101008 3.    COMMON/MMSA/M*SA1(23,4),MMSA2(23,4),MMSA3(23,4)
010100R 4.    COMMON/MSAR12/MSA1(23,4),MSA2(23,4),MSA3(23,4)
010100R 5.    COMMON/MSB/MSB1(73,4),MSB2(73,4),MSB3(73,4)
010100B 6.    COMMON/MMSB/MMSB1(73,4),MMSB2(73,4),MMSB3(73,4)
010100B 7.    COMMON/DAC/DA(23),ACA(23),DB(73),ACR(73)
010100B 8.    COMMON/ICMAXM/ICMAXM( 50),ICMAX(100)
010100B 9.    COMMON/MSKP/MS(6),KP(21)
010100R 10.   COMMON/NPAROB/NPAR,NOBS,KSTAR
010100B 11.   COMMON/DPLUS/DPLUS
010100B 12.   COMMON/KSTRP/KSTRP(20)
010100B 13.   COMMON/CHINON/CHINON(12)
010100R 14.   COMMON/FACGAW/FAC(150),GAMMA(150)
010100B 15.   COMMON/ZA/ZA(10)
010100B 16.   COMMON/A/A(5,3,3)
010100B 17.   COMMON/LPN/LPN(14)
010100B 18.   COMMON/IPRINT/IREAD,IPRINT,ITAPE7,ITAPE8,ITAPE9,ITAP10
010100B 19.   COMMON/SIGMA/SIGMA
010100R 20.   DIMENSION ZAS(20)
010100B 21.   IREAD=5
010124B 22.   IPRINT=6
010125R 23.   ITAPE9=9
010126B 24.   ITAP10=10
010127B 25.   READ(IREAD,4) KCH,IN,DPLUS

C
C   DPLUS**2 IS (((SIGMA ZERO)**2)/((SIGMA A)**2))*((1-ALPHA)100 UPPER
C   PERCENTAGE POINT OF A CHI SQUARE DISTRIBUTION ON KSTAR DEGREES OF
C   FREEDOM)
C
010137B 26.   4 FORMAT(A6,I4,7F10.4)
010137B 27.   6 FORMAT(A6,I4,10F6.3)
010137B 28.   CALL CHECIN(KCH,IN,5HDPLUS,0)
010141B 29.   READ(IREAD,4) KCH,IN,SIGVA
010150R 30.   CALL CHECIN(KCH,IN,5HSIGMA,0)
010152B 31.   READ(IREAD,10) KCH,IN,NPAR,NOBS,KSTAR
010163R 32.   10 FORMAT(A6,I4,14I5)
010163R 33.   CALL CHECIN(KCH,IN,6HNPAROB,0)
010165B 34.   WRITE(IPRINT,11) NPAR,NOBS,KSTAR

C
C   NPAR IS TOTAL NUMBER OF COMPONENTS IN THE PARAMETER VECTOR
C   NOBS IS TOTAL NUMBER OF OBSERVATIONS
C   KSTAR IS NUMBER OF COMPONENTS OF INTEREST IN THE PARAMETER VECTOR
C
010174B 35.   11 FORMAT(1X,ANPAR=0,I3,1X,ANOBS=0,I3,1X,AKSTAR=0,I3)
010174B 36.   NPMK=NPAR-KSTAR
010175B 37.   NPMKP1=NPMK+1
010176B 38.   NPMKP2=NPMK+2
010177B 39.   NPARP=NPAR+1
010200B 40.   NPARP2=NPAR+2
010201B 41.   READ(IREAD,4) KCH,IN,(ZAS(IB),IB=1,5)
010215B 42.   CALL CHECIN(KCH,IN,3HZAS,0)
010217B 43.   READ(IREAD,4) KCH,IN,(ZAS(IB),IB=6,10)
010232B 44.   CALL CHECIN(KCH,IN,4HZAS5,0)
010234B 45.   DO 12 IOBS=1,NOBS
010236B 46.   ZA(IOBS)=ZAS(IOBS)/SIGMA

C
C   ZA(I) ARE ZAI
C
010236B 47.   12 CONTINUE
010241B 48.   READ(IREAD,10) KCH,IN,IQAMAX,IQBMAX

```

```

C
C IQAMAX IS TOTAL NUMBER OF TERMS IN THE EXPRESSIONS OF THE
C PRODUCTS OF THE FIRST PARTIAL DERIVATIVES OF (R+)**2 W.R.T.
C A+I1J1K1 AND A+I2J2K2 EVALUATED AT A+=0
C
C IQBMAX IS TOTAL NUMBER OF TERMS IN THE EXPRESSIONS OF THE SECOND
C PARTIAL DERIVATIVES OF (R+)**2 W.R.T. A+I1J1K1,A+I2J2K2 EVALUATED
C AT A+=0
C
010252B   50. CALL CHECIN(KCH,IN,6HIQAMAX,0)
010254B   51. READ(IREAD,10) KCH,IN,KCHIMX
C
C KCHIMX IS THE NUMBER OF TERMS USED IN REPRESENTING NON-CENTRAL
C CHI SQUARE DISTRIBUTION BY A LINEAR COMBINATION OF CENTRAL CHI
C SQUARE DISTRIBUTION
C
010263B   52. CALL CHECIN(KCH,IN,6HKCHIMX,0)
010265B   53. READ(IREAD,4) KCH,IN,VSMALL
C
C VSMALL IS SOME SMALL POSITIVE NUMBER
C
010274B   54. CALL CHECIN(KCH,IN,6HVSMALL,0)
010276B   55. READ(IREAD,10) KCH,IN,(LPN(I),I=1,7)
010311B   56. CALL CHECIN(KCH,IN,6HLPN(1),0)
010313B   57. READ(IREAD,10) KCH,IN,(LPN(I),I=8,14)
C
C LPN(*) DECIDES WHETHER THE INTERMEDIATE RESULTS WILL BE PRINTED OUT
C
010326B   58. CALL CHECIN(KCH,IN,6HLPN(8),0)
010330B   59. DO 14 IOBS=1,NOBS
010332B   60. IF(.NOT.NPAR.GE.4) GO TO 13
010335B   61. READ(IREAD,6) KCH,IN,((A(IOBS,JB,KB),KB=JB,NPAR),JB=1,NPAR)
010360B   62. CALL CHECIN(KCH,IN,6HA(***),IOBS)
010362B   63. GO TO 14
010362B   64. 13 LSKIP=1
010363B   65. READ(IREAD,4) KCH,IN,((A(IOBS,JB,KB),KB=JB,NPAR),JB=1,NPAR)
C
C A(I,J,K) ARE A+IJK
C
010407B   66. CALL CHECIN(KCH,IN,6HA(***),IOBS)
010411B   67. 14 CONTINUE
010413B   68. DO 24 IQQA=1,IQAMAX
010415B   69. IQA=IQQA
010415B   70. READ(ITAPE9,20) IQA,ICMAXM(IQA),(MSA(IQA,I),I=1,6),(KPA(IQA,I),I=1
1,21),((MSA1(IQA,I),MMSA1(IQA,I)),I=1,4),((MSA2(IQA,I),MMSA2(IQA,I))
1,I=1,4),((MSA3(IQA,I),MMSA3(IQA,I)),I=1,4),DA(IQA),ACA(IQA)
010517B   71. 20 FORMAT(1X,I5,I5,5I2,2F5.1)
C
C (KPA(IQA,I),I=1,21) COMPLETELY SPECIFY A SUBSET GENERATED
C BY PARTITIONING THE SET OF ALL (A+I1J1K1,A+I2J2K2)
C (C.F. SECTION (4,9))
C (MSA(IQA,I),I=1,6) REPRESENT A TYPICAL ELEMENT OF THIS SUBSET
C ICMAXM(IQA) IS THE TOTAL NUMBER OF TERMS IN THE EXPRESSION OF
C THE PRODUCT OF THE FIRST PARTIAL DERIVATIVES OF (R+)**2 W.R.T.
C A+I1J1K1 AND A+I2J2K2 EVALUATED AT A+=0
C
C A TERM IN AN EXPRESSION IS REPRESENTED BY ACA(IQA)(C.F.COLUMN(4) IN
C SECTION (4,11)) DA(IQA)(C.F.COLUMN(3))((MSA1(IQA,I),MMSA1(IQA,I)),
C I=1,4)(C.F.COLUMNS (32) TO (39))((MSA2(IQA,I),MMSA2(IQA,I)),I=1,4)
C (C.F.COLUMNS (40) TO (47)) AND ((MSA3(IQA,I),MMSA3(IQA,I)),I=1,4)
C (C.F. COLUMNS (48) TO (55))
C
010517B   72. 24 CONTINUE
010521B   73. DO 28 IQQB=1,IQBMAX
010523B   74. IQB=IQQB

```

010523B 75. READ(ITAP10,20) IQB,ICMAX(IQB), (MSR(IQB,I),I=1,6),(KPB(IQB,I),I=1,21),((MSB1(IQB,I),MMSB1(IQB,I)),I=1,4),((MSB2(IQB,I),MMSB2(IQB,I)),I=1,4),((MSB3(IQB,I),MMSB3(IQB,I)),I=1,4),DB(IQB),ACB(IQB)

C C (KPB(IQB,I),I=1,21) COMPLETELY SPECIFY A SUBSET
C GENERATED BY PARTITIONING THE SET OF ALL (A+I1J1K1,A+I2J2K2)
C (C.F.SECTION(4,9))
C (MSB(IQB,I),I=1,6) REPRESENTS A TYPICAL ELEMENT OF THIS SUBSET
C ICMAX(IQB) IS THE TOTAL NUMBER OF TERMS IN THE EXPRESSION OF THE
C SECOND DERIVATIVE OF (H+)**2 W.R.T. A+I1J1K1,A+I2J2K2 EVALUATED AT
C A=0
C
C A TERM IN AN EXPRESSION IS REPRESENTED BY ACB(IQB) (C.F. COLUMN (4)
C IN SECTION (4,11)),DB(IQB) (C.F. COLUMN (3)),
C ((MSB1(IQB,I),MMSB1(IQB,I)),I=1,4)(C.F.COLUMNS (32) TO (39)),
C ((MSB2(IQB,I),MMSB2(IQB,I)),I=1,4)(C.F.COLUMNS (40) TO (47)) AND
C ((MSB3(IQB,I),MMSB3(IQB,I)),I=1,4)(C.F.COLUMNS (48) TO (55))

C
010630B 76. 28 CONTINUE
010632B 77. IF(LPN(13).EQ.0) GO TO 50
010633B 78. WRITE(IPRINT,30)
010637B 79. 30 FORMAT(/1X,0IQA,ICMAXM(IQA),DA(IQA),ACA(IQA),MSA(IQA,*),KPA(IQA,*),
1,MSA1(IQA,*),MMSA1(IQA,*),ETC AREA//)/*
010637B 80. WRITE(IPRINT,31)
010643B 81. 31 FORMAT(/1X,0IQB,ICMAX(IQB),DB(IQB),ACB(IQB),MSB(IQB,*),KPB(IQB,*),
1SB1(IQB,*),MMSB1(IQB,*),ETC AREA//)/*
010643B 82. DO 34 IQA=1,IQAMAX
010645B 83. WRITE(IPRINT,32) IQA,ICMAXM(IQA),DA(IQA),ACA(IQA),(MSA(IQA,I),I=1,
16),(KPA(IQA,I),I=1,21),((MSA1(IQA,I),MMSA1(IQA,I)),I=1,4),((MSA2(I
1QA,I),MMSA2(IQA,I)),I=1,4),((MSA3(IQA,I),MMSA3(IQA,I)),I=1,4)
010741B 84. 32 FORMAT(1X,I3.1H ,I2.1H ,2(F5.1,1H),6I1,1H ,21I1,1H ,3(8I1,1X))
010741B 85. 34 CONTINUE
010743B 86. WRITE(IPRINT,36)
010747B 87. 36 FORMAT(////////,1X,1H)
010747B 88. DO 42 IQB=1,IQBMAX
010751B 89. WRITE(IPRINT,32) IQB,ICMAX(IQB), DB(IQB),ACB(IQB),(MSB(IQB,I),I=1,
16),(KPB(IQB,I),I=1,21),((MSB1(IQB,I),MMSB1(IQB,I)),I=1,4),((MSB2(I
1QB,I),MMSB2(IQB,I)),I=1,4),((MSB3(IQB,I),MMSB3(IQB,I)),I=1,4)
011045B 90. 42 CONTINUE
011047B 91. WRITE(IPRINT,36)
011053B 92. 43 FORMAT(1X,0FAC(*) AREA,10(F10.5,1X))
011053B 93. 44 FORMAT(1X,0GAMMA(*) AREA,10(F10.5,1X))
011053B 94. 45 FORMAT(1X,0VNAMUA=0,F10.5,0EXP(-0.5*VNAMDA)*EXP(-0.5*DSTAR*DSTAR)
1=0,F10.5/1X,0ZA(*) AREA,10(F10.5,1X)/)
011053B 95. 50 LSKIP=1
011053B 96. KCHIMP=KCHIMX+13
011055B 97. FAC(I)=1.0
011056B 98. DO 52 I=2,KCHIMP
011061B 99. IM1=I-1
011061B 100. VIM1=IM1
011063B 101. FAC(I)=VIM1*FAC(IM1)
011064B 102. 52 CONTINUE

C C FAC(I) IS FACTORIAL OF (I-1)

C
011066B 103. IF(LPN(1),EQ.0) GO TO 53
011067B 104. WRITE(IPRINT,43) (FAC(I),I=1,10)
011100B 105. 53 LSKIP=1
011100B 106. GAMMA(1)=SQRT(3.1415926536)
011103B 107. DO 54 I=3,KCHIMP,2
011105B 108. IM2=I-2
011106B 109. VI=I
011107B 110. GAMMA(I)=0.5*(VI-2.0)*GAMMA(IM2)
011113B 111. 54 CONTINUE
011115B 112. DO 56 I=2,KCHIMP,2
011116B 113. J=I/2

```

011116B 114.      GAMMA(I)=FAC(J)
011120B 115.      56 CONTINUE
011122B 116.      IF(LPN(1).EQ.0) GO TO 57
011123B 117.      WRITE(IPRINT,44) (GAMMA(I),I=1,10)
011134B 118.      57 LSKIP=1
011134B 119.      TV=0.0
011135B 120.      DO 58 IOBS=1,NOBS
011140B 121.      TV=TV+(ZA(IOBS)*ZA(IOBS))
011140B 122.      58 CONTINUE
011144B 122.      VNAMDA=TV
011144B 124.      T1=EXP(-0.5*VNAMDA)
011151B 125.      T2=EXP(-0.5*DPLUS*DPLUS)
011156B 126.      TT12=T1*T2
011157B 127.      IF(LPN(1).EQ.0) GO TO 59
011161B 128.      WRITE(IPRINT,45) VNAMDA,TT12,(ZA(IB),IB=1,NOBS)
011175B 129.      59 LSKIP=1
011175B 130.      DO 70 ICHI=1,12
011177B 131.      VICH=ICHI
011177B 132.      TC=0.0
011200B 133.      I=-1
011201B 134.      60 I=I+1
011205B 135.      VI=I
011205B 136.      IP1=I+1
011206B 137.      T3=0.5*(VICH+2.0*VI)
011211B 138.      TCPRV=TC
011213B 139.      TC=TC+(T1*((.5*VNAMDA)**I)/FAC(IP1))*((DPLUS*DPLUS)**(T3-1.))*T2/(
    1.0**T3)*GAMMA(ICHI+2*I))
011245B 140.      IF(LPN(1).EQ.0) GO TO 61
011246B 141.      WRITE(IPRINT,67) TCPRV,TC
011254B 142.      61 LSKIP=1
011254B 143.      IF(I.LE.20) GO TO 60
011256B 144.      IF(I.GE.KCHIMX) GO TO 64
011260B 145.      IF(ABS(TC-TCPRV).GT.VSMALL) GO TO 60
011264B 146.      CHINON(ICHI)=TC
011265B 147.      WRITE(IPRINT,62) ICHI,CHINON(ICHI)
011275B 148.      62 FORMAT(1X,AICHI=0,I2,A(CHINON(ICHI))=0,F12.6)
011275B 149.      GO TO 68
011275B 150.      64 CHINON(ICHI)=TC
011276B 151.      WRITE(IPRINT,66) ICHI,KCHIMX,TCPRV,TC
011310B 152.      66 FORMAT(1X,AICHI=0,I2,1H/,AKCHIMX=0,I4,1H/,A(CHINON(ICHI))=0,2(F12.6,
    11X))
011310B 153.      67 FORMAT(1X,A(TCPRV=0,F10.5,A(=0,F10.5)
011310B 154.      68 LSKIP=1
011310B 155.      70 CONTINUE
C
C      NOW DENSITIES OF NON-CENTRAL CHI SQUARE DISTRIBUTIONS OF 1,2,...,12
C      DEGREES OF FREEDOM AND PARAMETER VNAMDA EVALUATED AT (D+)**2
C      ARE FOUND
C
C      TO CALCULATE (SUM FROM I1 EQUAL TO 1 TO N)(SUM FROM I2 EQUAL TO 1
C      TO N)(SUM FROM J1 EQUAL TO 1 TO P)(SUM FROM K1 EQUAL TO J1 TO P)
C      (SUM FROM J2 EQUAL TO 1 TO P)(SUM FROM K2 EQUAL J2 TO P) OF
C      0.5*BETA1A+I1J1K1A+I2J2K2*A+I1J1K1*A+I2J2K2
C
011313B 156.      TS=0.0
011313B 157.      DO 290 I1=NPMKP1,NOBS
011316B 158.      DO 280 I2=NPMKP1,NOBS
011320B 159.      DO 270 J1=1,NPAR
011323B 160.      DO 260 K1=J1,NPAR
011324B 161.      DO 250 J2=1,NPAR
011326B 162.      DO 240 K2=J2,NPAR
011327B 163.      MS(1)=I1
011327B 164.      MS(2)=J1
011331B 165.      MS(3)=K1
011332B 166.      MS(4)=I2

```

```

011334B   167.      MS(5)=J2
011335B   168.      MS(6)=K2
011336B   169.      IF(I1.LE.I2) GO TO 72
011341B   170.      MS(1)=I2
011342B   171.      MS(2)=J2
011343B   172.      MS(3)=K2
011344B   173.      MS(4)=I1
011345B   174.      MS(5)=J1
011346B   175.      MS(6)=K1
011347B   176.      72 LSKIP=1
011350B   177.      DO 74 IM=1,6
011352B   178.      I=MS(IM)
011352B   179.      IF(I.GE.1.AND.I.LE.NPMK) KP(IM)=1
011361B   180.      IF(I.GE.NPMKP1.AND.I.LE.NPAR) KP(IM)=2
011367B   181.      IF(I.GE.NPARP.AND.I.LE.NOBS) KP(IM)=3
011374B   182.      74 CONTINUE
011376B   183.      IT=6
011377B   184.      DO 80 IM=1,5
011401B   185.      IMP1=IM+1
011401B   186.      DO 78 JM=IMP1,6
011405B   187.      IT=IT+1
011406B   188.      IF(MS(IM).EQ.MS(JM)) KP(IT)=1
011413B   189.      IF(MS(IM).NE.MS(JM)) KP(IT)=0
011415B   190.      78 CONTINUE
011417B   191.      80 CONTINUE
011422B   192.      IQAYES=0
011422B   193.      DO 110 IQA=1,IQAMAX
011425B   194.      DO 90 I=1,21
011427B   195.      IF(KP(I).NE.KPA(IQA,I)) GO TO 110
011427B   196.      90 CONTINUE
011434B   196.      IQAYES=IQA
011440B   198.      GO TO 120
011440B   199.      110 CONTINUE
011441B   199.      120 IF(IQAYES.EQ.0) GO TO 130
011446B   201.      I=IQAYES
011446B   202.      DO 124 IC=1,ICMAXM(I)
011452B   203.      I=IQAYES-1+IC
011453B   204.      CALL COEF11(I,UNDEF)

C      C      ROUTINE COEF11 CALCULATES BETA1A+I1J1K1A+I2J2K2 ASSUMING THAT
C      C      SECOND PARTIAL DERIVATIVE OF (R+)**2 IS ZERO
C      C      TS=TS+UNDEF*A(I1,J1,K1)*A(I2,J2,K2)
011457B   205.      124 CONTINUE
011472B   206.      130 LSKIP=1
011475B   207.      IQBYES=0
011475B   208.      DO 180 IQB=1,IQBMAX
011501B   209.      DO 170 I=1,21
011503B   210.      IF(KP(I).NE.KPB(IQB,I)) GO TO 180
011503B   211.      170 CONTINUE
011510B   212.      IQBYES=IQB
011514B   214.      GO TO 190
011514B   215.      180 CONTINUE
011515B   215.      190 IF(IQBYES.EQ.0) GO TO 200
011522B   217.      I=IQBYES
011522B   218.      DO 194 IC=1,ICMAX(I)
011526B   219.      I=IQBYES-1+IC
011527B   220.      CALL COEF2(I,UNDEF)

C      C      ROUTINE COEF2 CALCULATES BETA1A+I1J1K1A+I2J2K2 ASSUMING THAT
C      C      PRODUCT OF FIRST PARTIAL DERIVATIVE OF (R+)**2 IS ZERO
C      C      TS=TS+UNDEF*A(I1,J1,K1)*A(I2,J2,K2)
011533B   221.      194 CONTINUE
011546B   222.      200 LSKIP=1
011551B   223.      224 CONTINUE

```

```

011554B   225. 250 CONTINUE
011557B   226. 260 CONTINUE
011562B   227. 270 CONTINUE
011565B   228. 280 CONTINUE
011570B   229. 290 CONTINUE
011573B   230. TS=0.5*TS*SIGMA*SIGMA
011576B   231. TS2NEG=-TS

C          TO CALCULATE (SUM FROM I EQUAL TO 1 TO N)(SUM FROM J EQUAL TO 1
C          TO P)(SUM FROM K EQUAL TO J TO P) OF BETA1A+IJK*A+IJK
C

011576B   232. DO 350 I=NPMKP1,NOBS
011601B   233. DO 340 J=1,NPAR
011604B   234. DO 330 K=J,NPAR
011605B   235. CALL COEF1(I,J,K,UNDEF)

C          ROUTINE COEF1 CALCULATES BETA1A+IJK
C

011610B   236. TS=TS+UNDEF*A(I,J,K)*SIGMA
011617B   237. 330 CONTINUE
011622B   238. 340 CONTINUE
011624B   239. 350 CONTINUE
011627B   240. TS1NEG=-(TS+TS2NEG)
011631B   241. TS12NG=-TS
011632B   242. WRITE(IPRINT,360) SIGMA,TS1NEG,TS2NEG,TS12NG
011643B   243. 360 FORMAT(1X,0SIGMA=0,F20.6,0POWER=ALPHA A +(0,F10.6,0)+(0,F10.6,0)=
1ALPHA A 0,0+(0,F10.6,0)0)
011643B   244. STOP
011645B   245. END

C          ROUTINE COEF11 CALCULATES BETA1A+I1J1K1A+I2J2K2 ASSUMING THAT
C          SECOND PARTIAL DERIVATIVES OF (R+)**2 IS ZERO
C

000000B   1. SUBROUTINE COEF11(II,UNDEF)
000000B   2. COMMON/MSKPAB/MSA(23,6),MSB(73,6),KPA(23,21),KPB(73,21)
000000B   3. COMMON/MMSA/MMSA1(23,4),MMSA2(23,4),MMSA3(23,4)
000000B   4. COMMON/MSAB12/MSA1(23,4),MSA2(23,4),MSA3(23,4)
000000B   5. COMMON/MSB/MSB1(73,4),MSB2(73,4),MSB3(73,4)
000000B   6. COMMON/MMSB/MMSB1(73,4),MMSB2(73,4),MMSB3(73,4)
000000B   7. COMMON/DAC/DA(23),ACA(23),DB(73),ACB(73)
000000B   8. COMMON/ICMAXM/ICMAXM( 50),ICMAX(100)
000000B   9. COMMON/MSKP/MS(6),KP(21)
000000B  10. COMMON/NPAROB/NPAR,NOBS,KSTAR
000000B  11. COMMON/DPLUS/DPLUS
000000B  12. COMMON/KSTRP/KSTRP(20)
000000B  13. COMMON/CHINON/CHINON(12)
000000B  14. COMMON/FACGAM/FAC(150),GAMMA(150)
000000B  15. COMMON/ZA/ZA(10)
000000B  16. COMMON/A/A(5,3,3)
000000B  17. COMMON/LPN/LPN(14)
000000B  18. COMMON/NS/NS(4)
000000B  19. COMMON/IPRINT/IREAD,IPRINT,ITAPE7,ITAPE8,ITAPE9,ITAP10
000000B  20. COMMON/TP/TP
000000B  21. I=II
000000B  22. IF(LPN(8),E0,0) GO TO 8
000002B  23. WRITE(IPRINT,2) (MS(IB),IB=1,6),(KP(IB),IB=1,21)
000034B  24. 2 FORMAT(1X,0I1,J1,K1,I2,J2,K2 ARE0,6(I1,1X),1H/,0KP(*) ARE0,21(I1,1
1X))
                  WRITE(IPRINT,4)
000043B  25. 4 FORMAT(1X,0I,ICMAXM(I),DA(I),ACA(I),MSA(*),KPA(*),MSA1(I,*),MMSA1(
1I,*)) ETC ARE0
000043B  26. WRITE(IPRINT,6) I,ICMAXM(I),DA(I),ACA(I),(MSA(I,IB),IB=1,6),(KPA(I
1,IB),IB=1,21),((MSA1(I,IR),MMSA1(I,IB)),IB=1,4),((MSA2(I,IB),MMSA2
1(I,IB)),IB=1,4),((MSA3(I,IB),MMSA3(I,IB)),IB=1,4)
000220B  27. 6 FORMAT(1X,I5,1H/,I2,1H/,2(F4.1,1H/),6I1,1H/,21I1,1H/,3(4(I1,1H*,I1

```

```

      1,1H//),2H//)
000220B   29.   8 LSKIP=1
000221B   30.   TP=(DPLUS**DA(I))*ACA(I)
000233B   31.   IF(LPN(9).EQ.0) GO TO 12
000235B   32.   WRITE(IPRINT,10) DPLUS,DA(I),ACA(I),TP
000260B   33.   10 FORMAT(1X,0(DPLUS**DA(I))*ACA(I)=(0,F10.5,2H**,F4.1,2H)*,F4.1,1H=,
1F10.5)
000260B   34.   12 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF (Z(1)**L(1))*(Z(2)**L(2))*. . .
C      (Z(P-KSTAR)**L(P-KSTAR))
C
000261B   35.   DO 14 IH=1,4
000265B   36.   IF(MMSA1(I,IH).EQ.0) GO TO 14
000276B   37.   IH1=MSA1(I,IH)
000303B   38.   IH2=MS(IH1)
000310B   39.   TZ=ZA(IH2)
000315B   40.   MMS=MMSA1(I,IH)
000324B   41.   CALL EXPZ(TZ,MMS,TP)
C
C      ROUTINE EXPZ CALCULATES EXPECTATION OF Z,Z**2,Z**3 AND Z**4
C
000331B   42.   14 CONTINUE
000334B   43.   IF(LPN(9).EQ.0) GO TO 18
000335B   44.   WRITE(IPRINT,16) TP
000345B   45.   16 FORMAT(1X,0(DPLUS**DA(I))*ACA(I)*EXPECTATION OF PRODUCT OF FIRST P
1-KSTAR R.V. =0,F10.5)
000345B   46.   18 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF (Z(P+1)**L(P+1))*(Z(P+2)*L(P+2))*. . .
C      (Z(N)*L(N))
C
000346B   47.   DO 20 IH=1,4
000352B   48.   IF(MMSA3(I,IH).EQ.0) GO TO 20
000363B   49.   IH1=MSA3(I,IH)
000370B   50.   IH2=MS(IH1)
000375B   51.   TZ=ZA(IH2)
000402B   52.   MMS=MMSA3(I,IH)
000411B   53.   CALL EXPZ(TZ,MMS,TP)
000416B   54.   20 CONTINUE
000421B   55.   IF(LPN(9).EQ.0) GO TO 23
000422B   56.   WRITE(IPRINT,22) TP
000432B   57.   22 FORMAT(1X,0(DPLUS**DA(I))*ACA(I)*EXPECTATION OF PRODUCT OF FIRST P
1-KSTAR R.V. AND LAST N-P R.V. =0,F10.5)
000432B   58.   23 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF ((S(P-KSTAR+1)SQRT(ZBARS(P-KSTAR+1)))
C      **L(P-KSTAR+1))*((S(P-KSTAR+2)SQRT(ZBARS(P-KSTAR+2)))**L(P-KSTAR+2)
C      )*. . .*((S(P)SQRT(ZBARS(P)))**L(P))
C
000433B   59.   CALL COEG11(II,UNDEF)
000442B   60.   RETURN
000444B   61.   END

```

000000B 62. SUBROUTINE COEG11(II,UNDEF)

C ROUTINE COEG11 CALCULATES EXPECTATION OF
 $((S(P-KSTAR+1)SQRT(ZBARS(P-KSTAR+1)))**L(P-KSTAR+1)*((S(P-KSTAR+2)$
 $SQRT(ZBARS(P-KSTAR+2)))**L(P-KSTAR+2))*...*((S(P)SQRT(ZBARS(P)))*$
 $L(P))$

000000B 63. COMMON/MSKPAB/MSA(23,6),MSB(73,6),KPA(23,21),KPB(73,21)
 000000B 64. COMMON/MMSA/MMSA1(23,4),MMSA2(23,4),MMSA3(23,4)
 000000B 65. COMMON/MSAB12/MSA1(23,4),MSA2(23,4),MSA3(23,4)
 000000B 66. COMMON/MSB/MSB1(73,4),MSB2(73,4),MSB3(73,4)
 000000B 67. COMMON/MMSB/MMSB1(73,4),MMSB2(73,4),MMSB3(73,4)
 000000B 68. COMMON/DAC/DA(23),ACA(23),DB(73),ACB(73)
 000000B 69. COMMON/ICMAXM/ICMAXM(50),ICMAX(100)
 000000B 70. COMMON/MSKP/MS(6),KP(21)
 000000B 71. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 72. COMMON/DPLUS/DPLUS
 000000B 73. COMMON/KSTRP/KSTRP(20)
 000000B 74. COMMON/CHINON/CHINON(12)
 000000B 75. COMMON/FACGAM/FAC(150),GAMMA(150)
 000000B 76. COMMON/ZA/ZA(10)
 000000B 77. COMMON/A/A(5,3,3)
 000000B 78. COMMON/LPN/LPN(14)
 000000B 79. COMMON/NS/NS(4)
 000000B 80. COMMON/IPRINT/IREAD,IPRINT,ITAPE7,ITAPE8,ITAPE9,ITAP10
 000000B 81. COMMON/TP/TP
 000000B 82. I=II
 000000B 83. NPMKP1=NPAR-KSTAR+1
 000003B 84. DO 24 II=1,20
 000006B 85. KSTRP(II)=KSTAR+II
 000006B 86. 24 CONTINUE
 000023B 86. VKSTAR=KSTAR
 000025B 88. DPLSA=((0.5*VKSTAR-1.0)/(DPLUS**2))-0.5
 000033B 89. DPLSB=0.5/DPLUS
 000033B 90. DO 30 II=1,4
 000037B 91. NS(II)=0
 000037B 92. 30 CONTINUE
 000053B 92. DO 50 IH=1,4
 000057B 94. DO 40 IIH=1,4
 000062B 95. IF(MMSA2(I,IH).EQ.IIH) NS(IIH)=NS(IIH)+1
 000101B 96. 40 CONTINUE
 000104B 97. 50 CONTINUE
 000110B 98. IF(NS(1).EQ.1.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 100
 000116B 99. IF(NS(1).EQ.0.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 120
 000123B 100. IF(NS(1).EQ.2.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 140
 000130B 101. IF(NS(1).EQ.0.AND.NS(2).EQ.0.AND.NS(3).EQ.1.AND.NS(4).EQ.0)
 1GO TO 160
 000135B 102. IF(NS(1).EQ.1.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 180
 000143B 103. IF(NS(1).EQ.3.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 200
 000150B 104. IF(NS(1).EQ.0.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.1)
 1GO TO 220
 000155B 105. IF(NS(1).EQ.1.AND.NS(2).EQ.0.AND.NS(3).EQ.1.AND.NS(4).EQ.0)
 1GO TO 240
 000163B 106. IF(NS(1).EQ.0.AND.NS(2).EQ.2.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 260
 000170B 107. IF(NS(1).EQ.2.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 280
 000176B 108. IF(NS(1).EQ.4.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
 1GO TO 300

C C 1000

```

000202B   109.  100 IH1=MS(MSA2(I,1))
000214B   110.  TZ=ZA(IH1)
000221B   111.  CALL E1000(TZ,TS1)
000226B   112.  TS3=0.0
000226B   113.  DO 110 IPAR=NPMKP1,NPAR
000241B   114.  IF(IPAR.NE.IH1) CALL E1100(ZA(IH1),ZA(IPAR),TS2)
000257B   115.  IF(IPAR.EQ.IH1) CALL E2000(ZA(IH1),TS2)
000271B   116.  TS3=TS3+ZA(IPAR)*TS2
000277B   117.  110 CONTINUE
000304B   118.  TP=TP*(DPLSA*TS1+DPLSB*TS3)
000310B   119.  GO TO 420
C
C      2000
C
000311B   120.  120 IH1=MS(MSA2(I,1))
000324B   121.  TZ=ZA(IH1)
000331B   122.  CALL E2000(TZ,TS1)
000336B   123.  TS3=0.0
000336B   124.  DO 130 IPAR=NPMKP1,NPAR
000351B   125.  IF(IPAR.EQ.IH1) CALL E3000(ZA(IH1),TS2)
000363B   126.  IF(IPAR.NE.IH1) CALL E2100(ZA(IH1),ZA(IPAR),TS2)
000401B   127.  TS3=TS3+ZA(IPAR)*TS2
000407B   128.  130 CONTINUE
000414B   129.  TP=TP*(DPLSA*TS1+DPLSB*TS3)
000420B   130.  GO TO 420
C
C      1100
C
000421B   131.  140 IH1=MS(MSA2(I,1))
000434B   132.  TZ1=ZA(IH1)
000441B   133.  IH2=MS(MSA2(I,2))
000452B   134.  TZ2=ZA(IH2)
000457B   135.  CALL E1100(TZ1,TZ2,TS1)
000464B   136.  TS3=0.0
000464B   137.  DO 150 IPAR=NPMKP1,NPAR
000477B   138.  IF(IPAR.EQ.IH1) CALL E2100(ZA(IH1),ZA(IH2),TS2)
000516B   139.  IF(IPAR.EQ.IH2) CALL E2100(ZA(IH2),ZA(IH1),TS2)
000534B   140.  IF(IPAR.NE.IH1.AND.IPAR.NE.IH2) CALL E1110(ZA(IH1),ZA(IH2),ZA(IPAR
1),TS2)
000557B   141.  TS3=TS3+ZA(IPAR)*TS2
000565B   142.  150 CONTINUE
000572B   143.  TP=TP*(DPLSA*TS1+DPLSB*TS3)
000576B   144.  GO TO 420
C
C      3000
C
000577B   145.  160 IH1=MS(MSA2(I,1))
000612B   146.  TZ=ZA(IH1)
000617B   147.  CALL E3000(TZ,TS1)
000624B   148.  TS3=0.0
000624B   149.  DO 170 IPAR=NPMKP1,NPAR
000637B   150.  IF(IPAR.EQ.IH1) CALL E4000(ZA(IH1),TS2)
000651B   151.  IF(IPAR.NE.IH1) CALL E3100(ZA(IH1),ZA(IPAR),TS2)
000667B   152.  TS3=TS3+ZA(IPAR)*TS2
000675B   153.  170 CONTINUE
000702B   154.  TP=TP*(DPLSA*TS1+DPLSB*TS3)
000706B   155.  GO TO 420
C
C      2100
C
000707B   156.  180 IH1=MS(MSA2(I,1))
000722B   157.  TZ1=ZA(IH1)
000727B   158.  IH2=MS(MSA2(I,2))
000740B   159.  TZ2=ZA(IH2)
000745B   160.  IF(MMSA2(I,1).EQ.2) GO TO 184
000754B   161.  IIH1=IH1
000755B   162.  IIH2=IH2

```

```

000757B 163. IH1=IIH2
000757B 164. IH2=IIH1
000761B 165. TTZ1=TZ1
000762B 166. TTZ2=TZ2
000764B 167. TZ1=TT22
000764B 168. TZ2=TTZ1
000766B 169. 184 LSKIP=1
000770B 170. CALL E2100(TZ1,TZ2,TS1)
000775B 171. TS3=0.0
000775B 172. DO 190 IPAR=NPMKP1,NPAR
001010B 173. IF(IPAR.EQ.IH1) CALL E3100(ZA(IH1),ZA(IH2),TS2)
001027B 174. IF(IPAR.EQ.IH2) CALL E2200(ZA(IH1),ZA(IH2),TS2)
001045B 175. IF(IPAR.NE.IH1.AND.IPAR.NE.IH2) CALL E2110(ZA(IH1),ZA(IH2),ZA(IPAR
1),TS2)
001070B 176. TS3=TS3+ZA(IPAR)*TS2
001076B 177. 190 CONTINUE
001103B 178. TP=TP*(DPLSA*TS1+DPLSB*TS3)
001107B 179. GO TO 420
C
C 1110
C
001110B 180. 200 IH1=MS(MSA2(I,1))
001123B 181. TZ1=ZA(IH1)
001130B 182. IH2=MS(MSA2(I,2))
001141B 183. TZ2=ZA(IH2)
001146B 184. IH3=MS(MSA2(I,3))
001160B 185. TZ3=ZA(IH3)
001165B 186. CALL E1110(TZ1,TZ2,TZ3,TS1)
001172B 187. TS3=0.0
001172B 188. DO 210 IPAR=NPMKP1,NPAR
001205B 189. IF(IPAR.EQ.IH1) CALL E2110(ZA(IH1),ZA(IH2),ZA(IH3),TS2)
001230B 190. IF(IPAR.EQ.IH2) CALL E2110(ZA(IH2),ZA(IH1),ZA(IH3),TS2)
001252B 191. IF(IPAR.EQ.IH3) CALL E2110(ZA(IH3),ZA(IH1),ZA(IH2),TS2)
001274B 192. IF(IPAR.NE.IH1.AND.IPAR.NE.IH2.AND.IPAR.NE.IH3) CALL E1111(ZA(IH1)
1,ZA(IH2),ZA(IH3),ZA(IPAR),TS2)
001325B 193. TS3=TS3+ZA(IPAR)*TS2
001333B 194. 210 CONTINUE
001340B 195. TP=TP*(DPLSA*TS1+DPLSB*TS3)
001344B 196. GO TO 420
C
C 4000
C
001345B 197. 220 IH1=MS(MSA2(I,1))
001360B 198. TZ=ZA(IH1)
001365B 199. CALL E4000(TZ,TS1)
001372B 200. TS3=0.0
001372B 201. DO 230 IPAR=NPMKP1,NPAR
001405B 202. IF(IPAR.EQ.IH1) CALL E5000(ZA(IH1),TS2)
001417B 203. IF(IPAR.NE.IH1) CALL E4100(ZA(IH1),ZA(IPAR),TS2)
001435B 204. TS3=TS3+ZA(IPAR)*TS2
001443B 205. 230 CONTINUE
001450B 206. TP=TP*(DPLSA*TS1+DPLSB*TS3)
001454B 207. GO TO 420
C
C 3100
C
001455B 208. 240 IH1=MS(MSA2(I,1))
001470B 209. TZ1=ZA(IH1)
001475B 210. IH2=MS(MSA2(I,2))
001506B 211. TZ2=ZA(IH2)
001513B 212. IF(MMSA2(I,1).EQ.3) GO TO 244
001522B 213. IIH1=IH1
001523B 214. IIH2=IH2
001525B 215. IH1=IIH2
001525B 216. IH2=IIH1
001527B 217. TTZ1=TZ1
001530B 218. TTZ2=TZ2

```

```

001532B 219.      TZ1=TTZ2
001532B 220.      TZ2=TTZ1
001534B 221.      244 LSKIP=1
001536B 222.      CALL E3100(ZA(IH1),ZA(IH2),TS1)
001543B 223.      TS3=0.0
001543B 224.      DO 250 IPAR=NPMKP1,NPAR
001556B 225.      IF(IPAR.EQ.IH1) CALL E4100(ZA(IH1),ZA(IH2),TS2)
001575B 226.      IF(IPAR.EQ.IH2) CALL E3200(ZA(IH1),ZA(IH2),TS2)
001613B 227.      IF(IPAR.NE.IH1.AND.IPAR.NE.IH2) CALL E3110(ZA(IH1),ZA(IH2),ZA(IPAR
1),TS2)
001636B 228.      TS3=TS3+ZA(IPAR)*TS2
001644B 229.      250 CONTINUE
001651B 230.      TP=TP*(DPLSA*TS1+DPLSB*TS3)
001655B 231.      GO TO 420
C
C      2200
C
001656B 232.      260 IH1=MS(MSA2(I,1))
001671B 233.      TZ1=ZA(IH1)
001676B 234.      IH2=MS(MSA2(I,2))
001707B 235.      TZ2=ZA(IH2)
001714B 236.      CALL E2200(TZ1,TZ2,TS1)
001721B 237.      TS3=0.0
001721B 238.      DO 270 IPAR=NPMKP1,NPAR
001734B 239.      IF(IPAR.EQ.IH1) CALL E3200(ZA(IH1),ZA(IH2),TS2)
001753B 240.      IF(IPAR.EQ.IH2) CALL E3200(ZA(IH2),ZA(IH1),TS2)
001771B 241.      IF(IPAR.NE.IH1.AND.IPAR.NE.IH2) CALL E2210(ZA(IH1),ZA(IH2),ZA(IPAR
1),TS2)
002014B 242.      TS3=TS3+ZA(IPAR)*TS2
002022B 243.      270 CONTINUE
002027B 244.      TP=TP*(DPLSA*TS1+DPLSB*TS3)
002033B 245.      GO TO 420
C
C      2110
C
002034B 246.      280 IH1=MS(MSA2(I,1))
002047B 247.      TZ1=ZA(IH1)
002054B 248.      IH2=MS(MSA2(I,2))
002065B 249.      TZ2=ZA(IH2)
002072B 250.      IH3=MS(MSA2(I,3))
002104B 251.      TZ3=ZA(IH3)
002111B 252.      IF(MMSA2(I,1).EQ.2) GO TO 288
002120B 253.      IF(MMSA2(I,2).EQ.2) GO TO 282
002126B 254.      IF(MMSA2(I,3).EQ.2) GO TO 284
002133B 255.      282 TTZ1=TZ1
002136B 256.      TTZ2=TZ2
002137B 257.      TZ1=TTZ2
002137B 258.      TZ2=TTZ1
002141B 259.      IIH1=IH1
002142B 260.      IIH2=IH2
002143B 261.      IH1=IIH2
002143B 262.      IH2=IIH1
002145B 263.      GO TO 288
002146B 264.      284 TTZ1=TZ1
002151B 265.      TTZ3=TZ3
002152B 266.      TZ1=TTZ3
002152B 267.      TZ3=TTZ1
002154B 268.      IIH1=IH1
002155B 269.      IIH3=IH3
002156B 270.      IH1=IIH3
002156B 271.      IH3=IIH1
002160B 272.      288 LSKIP=1
002162B 273.      CALL E2110(TZ1,TZ2,TZ3,TS1)
002167B 274.      TS3=0.0
002167B 275.      DO 290 IPAR=NPMKP1,NPAR
002202B 276.      IF(IPAR.EQ.IH1) CALL E3110(ZA(IH1),ZA(IH2),ZA(IH3),TS2)
002225B 277.      IF(IPAR.EQ.IH2) CALL E2210(ZA(IH1),ZA(IH2),ZA(IH3),TS2)

```

```

002247B   278.    IF(IPAR.EQ.IH3) CALL E2210(ZA(IH1),ZA(IH3),ZA(IH2),TS2)
002271B   279.    IF(IPAR.NE.IH1.AND.IPAR.NE.IH2.AND.IPAR.NE.IH3) CALL E2111(ZA(IH1)
                  1,ZA(IH2),ZA(IH3),ZA(IPAR),TS2)
002322B   280.    TS3=TS3+ZA(IPAR)*TS2
002330B   281.    290 CONTINUE
002335B   282.    TP=TP*(DPLSA*TS1+DPLSB*TS3)
002341B   283.    GO TO 420

C
C      1111
C

002342B   284.    300 IH1=MS(MSA2(I,1))
002355B   285.    TZ1=ZA(IH1)
002362B   286.    IH2=MS(MSA2(I,2))
002373B   287.    TZ2=ZA(IH2)
002400B   288.    IH3=MS(MSA2(I,3))
002412B   289.    TZ3=ZA(IH3)
002417B   290.    IH4=MS(MSA2(I,4))
002431B   291.    TZ4=ZA(IH4)
002436B   292.    CALL E1111(TZ1,TZ2,TZ3,TZ4,TS1)
002443B   293.    TS3=0.0
002443B   294.    DO 310 IPAR=NPMKP1,NPAR
002456B   295.    IF(IPAR.EQ.IH1) CALL E2111(ZA(IH1),ZA(IH2),ZA(IH3),ZA(IH4),TS2)
002505B   296.    IF(IPAR.EQ.IH2) CALL E2111(ZA(IH2),ZA(IH1),ZA(IH3),ZA(IH4),TS2)
002533B   297.    IF(IPAR.EQ.IH3) CALL E2111(ZA(IH3),ZA(IH1),ZA(IH2),ZA(IH4),TS2)
002561B   298.    IF(IPAR.EQ.IH4) CALL E2111(ZA(IH4),ZA(IH1),ZA(IH2),ZA(IH3),TS2)
002607B   299.    IF(IPAR.NE.IH1.AND.IPAR.NE.IH2.AND.IPAR.NE.IH3.AND.IPAR.NE.IH4) CAL
                  1L E11111(ZA(IH1),ZA(IH2),ZA(IH3),ZA(IH4),ZA(IPAR),TS2)
                  TS3=TS3+ZA(IPAR)*TS2
002646B   300.    310 CONTINUE
002654B   301.    TP=TP*(DPLSA*TS1+DPLSB*TS3)
002661B   302.    420 LSKIP=1
002665B   303.    UNDEF=TP
002667B   304.    -----
002671B   305.    -IF(LPN(9).EQ.0) GO TO -440-
002673B   306.    WRITE(IPRINT,430) TP
002703B   307.    430 FORMAT(1X,0(DSTAR**DA(I))*ACA(I)*(EXPECTATION OF PRODUCT OF FIRST
                  1P-KSTAR R.V. AND LAST N-P R.V.)0/1X,0*EXPECTATION OF PRODUCT OF MI
                  1DDE KSTAR R.V.)0,F10.5)
002703B   308.    440 LSKIP=1
002704B   309.    RETURN
002707B   310.    END

C
C      ROUTINE COEF2 CALCULATES BETA1A+I1J1K1A+I2J2K2 ASSUMING THAT
C      PRODUCT OF FIRST PARTIAL DERIVATIVES OF (R+)**2 IS ZERO
C

000000B   1.     SUBROUTINE COEF2(II,UNDEF)
000000B   2.     COMMON/MSKPAR/MSA(23,6),MSB(73,6),KPA(23,21),KPB(73,21)
000000B   3.     COMMON/MSAB12/MSA1(23,4),MSA2(23,4),MSA3(23,4)
000000B   4.     COMMON/MMSA/MMSA1(23,4),MMSA2(23,4),MMSA3(23,4)
000000B   5.     COMMON/MSB/MSB1(73,4),MSB2(73,4),MSB3(73,4)
000000B   6.     COMMON/MMSB/MMSB1(73,4),MMSB2(73,4),MMSB3(73,4)
000000B   7.     COMMON/DAC/DA(23),ACA(23),DB(73),ACB(73)
000000B   8.     COMMON/ICMAXM/ICMAXM( 50),ICMAX(100)
000000B   9.     COMMON/MSKP/MS(6),KP(21)
000000B  10.    COMMON/NPAR08/NPAR,N0BS,KSTAR
000000B  11.    COMMON/DPLUS/DPLUS
000000B  12.    COMMON/KSTRP/KSTRP(20)
000000B  13.    COMMON/CHINON/CHINON(12)
000000B  14.    COMMON/FACGAM/FAC(150),GAMMA(150)
000000B  15.    COMMON/ZA/ZA(10)
000000B  16.    COMMON/A/A(5,3,3)
000000B  17.    COMMON/LPN/LPN(14)
000000B  18.    COMMON/NS/NS(4)
000000B  19.    COMMON/IPRINT/IREAD,IPRINT,ITAPE7,ITAPE8,ITAPE9,ITAP10
000000B  20.    I=II
000000B  21.    IF(LPN(10).EQ.0) GO TO 8
000000B  22.    WRITE(IPRINT,21) (MS(IB),IB=1,6),(KP(IB),IB=1,21)
000002B  23.    2 FORMAT(1X,0I1,J1,K1,I2,J2,K2 ARE9,6(I1,1X),1H/,0KP(*) ARE0,21(I1,1
                  1X))

```

```

000021B      24.      WRITE(IPRINT,4)
000025B      25.      4 FORMAT(1X,A1,ICAMX(I),DB(I),ACB(I),MSB(*),KPB(*),MSB1(I,*),MMSB1(I
                  1*) ETC AREA)
000025B      26.      WRITE(IPRINT,6) I,ICMAX(I),DB(I),ACB(I),(MSB(I,IB),IB=1,6),(KPB(I,
                  1B),IB=1,21),((MSB1(I,IB),MMSR1(I,IB)),IR=1,4),((MSB2(I,IB),MMSB2(I,
                  1B)),IB=1,4),((MSB3(I,IB),MMSB3(I,IB)),IR=1,4)
000131B      27.      6 FORMAT(1X,I5,_H/,I2,1H/.2(F4.1,1H/),6I1,1H/,21I1,1H/,3I4(I1,1H*,I1
                  1,1H/),2H//)
000131B      28.      8 LSKIP=1
000131B      29.      TP=(DPLUS**DB(I))*ACB(I)
000136B      30.      IF(LPN(11).EQ.0) GO TO 12
000140B      31.      WRITE(IPRINT,10) DPLUS,DB(I),ACB(I),TP
000152B      32.      10 FORMAT(1X,A(DPLUS**DB(I))*ACB(I)=(A,F10.5,2H**,F4.1,2H)*,F4.1,1H=,
                  1F10.5)
000152B      33.      12 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF (Z(1)**L(1))*(Z(2)**L(2))*...*
C      (Z(P-KSTAR)**L(P-KSTAR))
C
000152B      34.      DO 14 IH=1,4
000154B      35.      IF(MMSB1(I,IH).EQ.0) GO TO 14
000161B      36.      IH1=MSB1(I,IH)
000163B      37.      IH2=MS(IH1)
000165B      38.      TZ=ZA(IH2)
000166B      39.      MMS=MMSB1(I,IH)
000171B      40.      CALL EXPZ(TZ,MMS,TP)
C
C      ROUTINE EXPZ CALCULATES EXPECTATION OF Z,Z**2,Z**3 AND Z**4
C
000174B      41.      14 CONTINUE
000176B      42.      IF(LPN(11).EQ.0) GO TO 18
000177B      43.      WRITE(IPRINT,16) TP
000204B      44.      16 FORMAT(1X,A(DSTAR**DB(I))*ACB(I)*EXPECTATION OF PRODUCT OF FIRST P
                  1-KSTAR R.V. =A,F10.5)
000204B      45.      18 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF (Z(P+1)**L(P+1))*(Z(P+2)*L(P+2))*...*
C      (Z(N)*L(N))
C
000204B      46.      DO 20 IH=1,4
000206B      47.      IF(MMSB3(I,IH).EQ.0) GO TO 20
000213B      48.      IH1=MSB3(I,IH)
000215B      49.      IH2=MS(IH1)
000217B      50.      TZ=ZA(IH2)
000220B      51.      MMS=MMSB3(I,IH)
000223B      52.      CALL EXPZ(TZ,MMS,TP)
000226B      53.      20 CONTINUE
000230B      54.      IF(LPN(11).EQ.0) GO TO 23
000231B      55.      WRITE(IPRINT,22) TP
000236B      56.      22 FORMAT(1X,A(DSTAR**DB(I))*ACB(I)*EXPECTATION OF PRODUCT OF FIRST P
                  1-KSTAR R.V. AND LAST N-P R.V. =A,F10.5)
000236B      57.      23 LSKIP=1
C
C      TO CALCULATE EXPECTATION OF ((S(P-KSTAR+1)SQRT(ZBARS(P-KSTAR+1)))
C      **L(P-KSTAR+1))*((S(P-KSTAR+2)SQRT(ZBARS(P-KSTAR+2))**L(P-KSTAR+2)
C      )*...*((S(P)SQRT(ZBARS(P)))*L(P))
C
000236B      58.      DO 30 II=1,4
000240B      59.      NS(II)=0
000240B      60.      30 CONTINUE
000242B      60.      DO 50 IH=1,4
000244B      62.      DO 40 IIH=1,4
000246B      63.      IF(MMSB2(I,IH).EQ.IIH) NS(IIH)=NS(IIH)+1
000256B      64.      40 CONTINUE
000260B      65.      50 CONTINUE

```

```

000263B   66.      IF(NS(1).EQ.0.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 80
000270B   67.      IF(NS(1).EQ.1.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 100
000275B   68.      IF(NS(1).EQ.0.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 120
000302B   69.      IF(NS(1).EQ.2.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 140
000307B   70.      IF(NS(1).EQ.0.AND.NS(2).EQ.0.AND.NS(3).EQ.1.AND.NS(4).EQ.0)
              GO TO 160
000314B   71.      IF(NS(1).EQ.1.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 180
000322B   72.      IF(NS(1).EQ.3.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 200
000327B   73.      IF(NS(1).EQ.0.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.1)
              GO TO 220
000334B   74.      IF(NS(1).EQ.1.AND.NS(2).EQ.0.AND.NS(3).EQ.1.AND.NS(4).EQ.0)
              GO TO 240
000342B   75.      IF(NS(1).EQ.0.AND.NS(2).EQ.2.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 260
000347B   76.      IF(NS(1).EQ.2.AND.NS(2).EQ.1.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 280
000355B   77.      IF(NS(1).EQ.4.AND.NS(2).EQ.0.AND.NS(3).EQ.0.AND.NS(4).EQ.0)
              GO TO 300
C
C      0000
C
000362B   78.      80 TP=TP*CHINON(KSTAR)
000363B   79.      GO TO 420
C
C      1000
C
000364B   80.      100 IH1=MS(MSB2(I,1))
000367B   81.      TZ=ZA(IH1)
000370B   82.      CALL E1000(TZ,TS)
000373B   83.      TP=TP*TS
000374B   84.      GO TO 420
C
C      2000
C
000375B   85.      120 IH1=MS(MSB2(I,1))
000377B   86.      TZ=ZA(IH1)
000400B   87.      CALL E2000(TZ,TS)
000403B   88.      TP=TP*TS
000404B   89.      GO TO 420
C
C      1100
C
000405B   90.      140 IH1=MS(MSB2(I,1))
000407B   91.      TZ1=ZA(IH1)
000410B   92.      IH2=MS(MSB2(I,2))
000412B   93.      TZ2=ZA(IH2)
000414B   94.      CALL E1100(TZ1,TZ2,TS)
000417B   95.      TP=TP*TS
000420B   96.      GO TO 420
C
C      3000
C
000421B   97.      160 IH1=MS(MSB2(I,1))
000423B   98.      TZ=ZA(IH1)
000424B   99.      CALL E3000(TZ,TS)
000427B  100.      TP=TP*TS
000430B  101.      GO TO 420
C
C      2100
C

```

```

000431B   102.    180 IH1=MS(MSB2(I,1))
000433B   103.    TZ1=ZA(IH1)
000434B   104.    IH2=MS(MSB2(I,2))
000436B   105.    TZ2=ZA(IH2)
000440B   106.    IF(MMSB2(I,1).EQ.2) GO TO 184
000442B   107.    TTZ1=TZ1
000443B   108.    TTZ2=TZ2
000444B   109.    TZ1=TTZ2
000445B   110.    TZ2=TTZ1
000446B   111.    184 LSKIP=1
000450B   112.    CALL E2100(TZ1,TZ2,TS)
000453B   113.    TP=TP*TS
000454B   114.    GO TO 420
C
C      1110
C
000455B   115.    200 IH1=MS(MSB2(I,1))
000457B   116.    TZ1=ZA(IH1)
000460B   117.    IH2=MS(MSB2(I,2))
000462B   118.    TZ2=ZA(IH2)
000464B   119.    IH3=MS(MSB2(I,3))
000466B   120.    TZ3=ZA(IH3)
000470B   121.    CALL E1110(TZ1,TZ2,TZ3,TS)
000473B   122.    TP=TP*TS
000474B   123.    GO TO 420
C
C      4000
C
000475B   124.    220 IH1=MS(MSB2(I,1))
000477B   125.    TZ=ZA(IH1)
000500B   126.    CALL E4000(TZ,TS)
000503B   127.    TP=TP*TS
000504B   128.    GO TO 420
C
C      3100
C
000505B   129.    240 IH1=MS(MSB2(I,1))
000507B   130.    TZ1=ZA(IH1)
000510B   131.    IH2=MS(MSB2(I,2))
000512B   132.    TZ2=ZA(IH2)
000514B   133.    IF(MMSB2(I,1).EQ.3) GO TO 244
000516B   134.    TTZ1=TZ1
000517B   135.    TTZ2=TZ2
000520B   136.    TZ1=TTZ2
000521B   137.    TZ2=TTZ1
000522B   138.    244 LSKIP=1
000524B   139.    CALL E3100(TZ1,TZ2,TS)
000527B   140.    TP=TP*TS
000530B   141.    GO TO 420
C
C      2200
C
000531B   142.    260 IH1=MS(MSB2(I,1))
000533B   143.    TZ1=ZA(IH1)
000534B   144.    IH2=MS(MSB2(I,2))
000536B   145.    TZ2=ZA(IH2)
000540B   146.    CALL E2200(TZ1,TZ2,TS)
000543B   147.    TP=TP*TS
000544B   148.    GO TO 420
C
C      2110
C
000545B   149.    280 IH1=MS(MSB2(I,1))
000547B   150.    TZ1=ZA(IH1)
000550B   151.    IH2=MS(MSB2(I,2))
000552B   152.    TZ2=ZA(IH2)

```

```

000554B 153. IH3=MS(MSB2(I,3))
000556B 154. TZ3=ZA(IH3)
000560B 155. IF(MMSB2(I,1).EQ.2) GO TO 288
000562B 156. IF(MMSB2(I,2).EQ.2) GO TO 282
000564B 157. IF(MMSB2(I,3).EQ.2) GO TO 284
000565B 158. 282 TTZ1=TZ1
000566B 159. TTZ2=TTZ1
000570B 160. TZ1=TTZ2
000570B 161. TZ2=TTZ1
000572B 162. GO TO 288
000572B 163. 284 TTZ1=TZ1
000573B 164. TTZ3=TTZ1
000575B 165. TZ1=TTZ3
000575B 166. TZ3=TTZ1
000577B 167. 288 LSKIP=1
000600B 168. CALL E2110(TZ1,TZ2,TZ3,TS)
000603B 169. TP=TP*TS
000604B 170. GO TO 420
C
C      1111
C
000605B 171. 300 IH1=MS(MSB2(I,1))
000607B 172. TZ1=ZA(IH1)
000610B 173. IH2=MS(MSB2(I,2))
000612B 174. TZ2=ZA(IH2)
000614B 175. IH3=MS(MSB2(I,3))
000616B 176. TZ3=ZA(IH3)
000620B 177. IH4=MS(MSB2(I,4))
000622B 178. TZ4=ZA(IH4)
000624B 179. CALL E1111(TZ1,TZ2,TZ3,TZ4,TS)
000627B 180. TP=TP*TS
000630B 181. 420 LSKIP=1
000631B 182. UNDEF=TP
000632B 183. IF(LPN(11).EQ.0) GO TO 440
000634B 184. WRITE(IPRINT,430) TP
000641B 185. 430 FORMAT(1X,a(DSTAR**DB(I))*ACB(I)*(EXPECTATION OF PRODUCT OF FIRST
          1P-KSTAR R.V. AND LAST N-P R.V.)a/1X,a*EXPECTATION OF PRODUCT OF MI
          TIDDLE KSTAR R.V.)=a,F10.5)
000641B 186. 440 LSKIP=1
000641B 187. RETURN
000644B 188. END
000000B 189. SUBROUTINE COEF1(II,JJ,KK,UNDEF)
C
C      ROUTINE COEF1 CALCULATES BETA1A+IJK
C
000000B 190. COMMON/MSKPAB/MSA1(23,6),MSB(73,6),KPA(23,21),KPB(73,21)
000000B 191. COMMON/MMSA/MMSA1(23,4),MMSA2(23,4),MMSA3(23,4)
000000B 192. COMMON/MSAB12/MSA1(23,4),MSA2(23,4),MSA3(23,4)
000000B 193. COMMON/MSB/MSB1(73,4),MSB2(73,4),MSB3(73,4)
000000B 194. COMMON/MMSB/MMSB1(73,4),MMSB2(73,4),MMSB3(73,4)
000000B 195. COMMON/DAC/DA123),ACA(23),DB(73),ACR(73)
000000B 196. COMMON/ICMAXM/ICMAXM(50),ICMAX(100)
000000B 197. COMMON/MSKP/MS(6),KP(21)
000000B 198. COMMON/NPAROB/NPAR,NOBS,KSTAR
000000B 199. COMMON/DPLUS/DPLUS
000000B 200. COMMON/KSTRP/KSTRP(20)
000000B 201. COMMON/CHINON/CHINON(12)
000000B 202. COMMON/FACGAM/FAC(150),GAMMA(150)
000000B 203. COMMON/ZA/ZA(10)
000000B 204. COMMON/A/A(5,3,3)
000000B 205. COMMON/LPN/LPN(14)
000000B 206. COMMON/IPRINT/IREAD,IPRINT,ITAPE7,ITAPE8,ITAPE9,ITAP10
000000B 207. I=II
000000B 208. J=JJ
000002B 209. K=KK
000003B 210. NPMK=NPAR-KSTAR

```

```

000005B   211.      NPMKP1=NPMK+1
000006B   212.      NPARP=NPAR+1
000007B   213.      TS=0.0
000010B   214.      IF(.NOT.(I.GE.NPMKP1.AND.I.LE.NPAR.AND.J.GE.1.AND.J.LE.NPMK.AND.K.
1GE.1.AND.K.LE.NPMK)) GO TO 30
000022B   215.      CALL E1000(ZA(I),TS1)
000026B   216.      IF(.NOT.(J.NE.K)) GO TO 10
000031B   217.      TS=4.0*ZA(J)*ZA(K)*DPLUS*TS1
000035B   218.      GO TO 20
000037B   219.      10 TS=2.0*(1.0+ZA(J)*ZA(J))*DPLUS*TS1
000043B   220.      20 LSKIP=1
000044B   221.      GO TO 100
000045B   222.      30 IF(.NOT.(I.GE.NPARP.AND.I.LE.NOBS))GO TO 100
000052B   223.      IF(.NOT.((J.GE.1.AND.J.LE.NPMK).AND.(K.GE.NPMKP1.AND.K.LE.NPAR)))G
10 TO 40
000062B   224.      CALL E1000(ZA(K),TS1)
000066B   225.      TS=-4.0*ZA(I)*ZA(J)*DPLUS*TS1
000073B   226.      GO TO 100
000074B   227.      40 IF(.NOT.((J.GE.NPMKP1.AND.J.LE.NPAR).AND.(K.GE.1.AND.K.LE.NPMK)))
1GO TO 50
000105B   228.      CALL E1000(ZA(J),TS1)
000110B   229.      TS=-4.0*ZA(I)*ZA(K)*DPLUS*TS1
000115B   230.      GO TO 100
000116B   231.      50 IF(.NOT.((J.GE.NPMKP1.AND.J.LE.NPAR).AND.(K.GE.NPMKP1.AND.K.LE.NPA
1R))) GO TO 100
000126B   232.      IF(.NOT.(J.NE.K)) GO TO 60
000131B   233.      CALL E1100(ZA(J),ZA(K),TS1)
000136B   234.      TS=-4.0*ZA(I)*DPLUS*DPLUS*TS1
000142B   235.      GO TO 70
000143B   236.      60 IF(.NOT.(J.EQ.K)) GO TO 70
000147B   237.      CALL E2000(ZA(J),TS1)
000152B   238.      TS=-2.0*ZA(I)*DPLUS*DPLUS*TS1
000155B   239.      70 LSKIP=1
000156B   240.      100 LSKIP=1
000157B   241.      UNDEF=TS
000160B   242.      RETURN
000163B   243.      END
1.          SUBROUTINE E1000(TZ,TS)
2.          COMMON/KSTRP/KSTRP(20)
3.          COMMON/DPLUS/DPLUS
4.          COMMON/CHINON/CHINON(12)
5.          COMMON/NPAROB/NPAR,NOBS,KSTAR
6.          TS=(TZ/DPLUS)*CHINON(KSTRP(2))
000003B   7.          RETURN
000005B   8.          END

```

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS	(B=RELATIVE ADDRESS)	(C= RELATIVE TO //)		
0000000C CHINON	0000000C DPLUS	0000000C KSTRP	0000000C NPAR	000001C NOBS
0000000R	9.	SUBROUTINE E2000(TZ,TS)		
0000000B	10.	COMMON/KSTRP/KSTRP(20)		
0000000B	11.	COMMON/DPLUS/DPLUS		
0000000B	12.	COMMON/CHINON/CHINON(12)		
0000000B	13.	COMMON/NPAROB/NPAR,NOBS,KSTAR		
0000000B	14.	TS=(1.0/(DPLUS**2))*(CHINON(KSTRP(2))+TZ*TZ*CHINON(KSTRP(4)))		
000007B	15.	RETURN		
000012B	16.	END		

000000B	17.	SUBROUTINE E1100(TZ1,TZ2,TS)
000000B	18.	COMMON/KSTRP/KSTRP(20)
000000B	19.	COMMON/DPLUS/DPLUS
000000B	20.	COMMON/CHINON/CHINON(12)
000000B	21.	COMMON/NPAROB/NPAR,NOBS,KSTAR
000000B	22.	TS=(TZ1*TZ2/(DPLUS**2))*CHINON(KSTRP(4))
000004B	23.	RETURN
000007B	24.	END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B	25.	SUBROUTINE E3000(TZ,TS)		
000000B	26.	COMMON/KSTRP/KSTRP(20)		
000000B	27.	COMMON/DPLUS/DPLUS		
000000B	28.	COMMON/CHINON/CHINON(12)		
000000B	29.	COMMON/NPAROB/NPAR,NOBS,KSTAR		
000000B	30.	TS=(TZ/(DPLUS**3))*(3.0*CHINON(KSTRP(4))+TZ*TZ*CHINON(KSTRP(6)))		
000010B	31.	RETURN		
000013B	32.	END		

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B	33.	SUBROUTINE E2100(TZ1,TZ2,TS)		
000000B	34.	COMMON/KSTRP/KSTRP(20)		
000000B	35.	COMMON/DPLUS/DPLUS		
000000B	36.	COMMON/CHINON/CHINON(12)		
000000B	37.	COMMON/NPAROB/NPAR,NOBS,KSTAR		
000000B	38.	TS=(TZ2/(DPLUS**3))*(CHINON(KSTRP(4))+TZ1*TZ1*CHINON(KSTRP(6)))		
000007B	39.	RETURN		
000012B	40.	END		

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B	41.	SUBROUTINE E1110(TZ1,TZ2,TZ3,TS)		
000000B	42.	COMMON/KSTRP/KSTRP(20)		
000000B	43.	COMMON/DPLUS/DPLUS		
000000B	44.	COMMON/CHINON/CHINON(12)		
000000B	45.	COMMON/NPAROB/NPAR,NOBS,KSTAR		
000000B	46.	TS=(TZ1*TZ2*TZ3/(DPLUS**3))*CHINON(KSTRP(6))		
000005B	47.	RETURN		
000010B	48.	END		

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B	49.	SUBROUTINE E4000(TZ,TS)		
000000B	50.	COMMON/KSTRP/KSTRP(20)		
000000B	51.	COMMON/DPLUS/DPLUS		
000000B	52.	COMMON/CHINON/CHINON(12)		
000000B	53.	COMMON/NPAROB/NPAR,NOBS,KSTAR		
000000B	54.	TS=(1.0/(DPLUS**4))*(3.0*CHINON(KSTRP(4))+6.0*TZ*TZ*CHINON(KSTRP(6)))+TZ*TZ*TZ*TZ*CHINON(KSTRP(8)))		
000014B	55.	RETURN		
000017B	56.	END		

000000B 57. SUBROUTINE E3100(TZ1,TZ2,TS)
 000000B 58. COMMON/KSTRP/KSTRP(20)
 000000B 59. COMMON/DPLUS/DPLUS
 000000B 60. COMMON/CHINON/CHINON(12)
 000000B 61. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 62. TS=(TZ1*TZ2/(DPLUS**4))*(3.0*CHINON(KSTRP(6))+TZ1*TZ1*CHINON(KSTRP
 1(8)))
 000011B 63. RETURN
 000014B 64. END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
 (C= RELATIVE TO //)
 000000C CHINON 000000C DPLUS 000000C KSTRP 000000C NPAR 000001C NOBS
 000000B 65. SUBROUTINE E2200(TZ1,TZ2,TS)
 000000B 66. COMMON/KSTRP/KSTRP(20)
 000000B 67. COMMON/DPLUS/DPLUS
 000000B 68. COMMON/CHINON/CHINON(12)
 000000B 69. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 70. TS=(1.0 /(DPLUS**4))*(CHINON(KSTRP(4))+(TZ1*TZ1+TZ2*TZ2)*CHINON(KST
 1RP(6))+TZ1*TZ1*TZ2*TZ2*CHINON(KSTRP(8)))
 000015B 71. RETURN
 000020B 72. END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
 (C= RELATIVE TO //)
 000000C CHINON 000000C DPLUS 000000C KSTRP 000000C NPAR 000001C NOBS
 000000B 73. SUBROUTINE E2110(TZ1,TZ2,TZ3,TS)
 000000B 74. COMMON/KSTRP/KSTRP(20)
 000000B 75. COMMON/DPLUS/DPLUS
 000000B 76. COMMON/CHINON/CHINON(12)
 000000B 77. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 78. TS=(TZ2*TZ3/(DPLUS**4))*(CHINON(KSTRP(6))+TZ1*TZ1*CHINON(KSTRP(8))
 1)
 000011B 79. RETURN
 000014B 80. END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
 (C= RELATIVE TO //)
 000000C CHINON 000000C DPLUS 000000C KSTRP 000000C NPAR 000001C NOBS
 000000B 81. SUBROUTINE E1111(TZ1,TZ2,TZ3,TZ4,TS)
 000000B 82. COMMON/KSTRP/KSTRP(20)
 000000B 83. COMMON/DPLUS/DPLUS
 000000B 84. COMMON/CHINON/CHINON(12)
 000000B 85. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 86. TS=(TZ1*TZ2*TZ3*TZ4/(DPLUS**4))*CHINON(KSTRP(8))
 000006B 87. RETURN
 000011B 88. END

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
 (C= RELATIVE TO //)
 000000C CHINON 000000C DPLUS 000000C KSTRP 000000C NPAR 000001C NOBS
 000000B 89. SUBROUTINE E5000(TZ,TS)
 000000B 90. COMMON/KSTRP/KSTRP(20)
 000000B 91. COMMON/DPLUS/DPLUS
 000000B 92. COMMON/CHINON/CHINON(12)
 000000B 93. COMMON/NPAROB/NPAR,NOBS,KSTAR
 000000B 94. TS=(TZ/(DPLUS**5))*(15.0*CHINON(KSTRP(6))+10.0*TZ*TZ*CHINON(KSTRP
 18))+TZ*TZ*TZ*TZ*CHINON(KSTRP(10)))
 000014B 95. RETURN
 000017B 96. END

```

000000B   97.      SUBROUTINE E4100(TZ1,TZ2,TS)
000000B   98.      COMMON/KSTRP/KSTRP(20)
000000B   99.      COMMON/DPLUS/DPLUS
000000B  100.      COMMON/CHINON/CHINON(12)
000000B  101.      COMMON/NPAR08/NPAR,NOBS,KSTAR
000000B  102.      TS=(TZ2/(DPLUS**5))*(TZ1*TZ1*TZ1*CHINON(KSTRP(10))+6.0*TZ1*TZ1
                   *CHINON(KSTRP(8))+3.0*CHINON(KSTRP(6)))
000015B   103.      RETURN
000020B   104.      END

```

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
(C= RELATIVE TO //)

000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B 105.	SUBROUTINE E3200(TZ1,TZ2,TS)			
000000B 106.	COMMON/KSTRP/KSTRP(20)			
000000B 107.	COMMON/DPLUS/DPLUS			
000000B 108.	COMMON/CHINON/CHINON(12)			
000000B 109.	COMMON/NPAR08/NPAR,NOBS,KSTAR			
000000B 110.	TS=(TZ1/(DPLUS**5))*(3.0*CHINON(KSTRP(6))+TZ1*TZ1*CHINON(KSTRP(8)) +3.0*TZ2*TZ2*CHINON(KSTRP(8))+TZ1*TZ1*TZ2*TZ2*CHINON(KSTRP(10)))			
000016B 111.	RETURN			
000021B 112.	END			

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
(C= RELATIVE TO //)

000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B 113.	SUBROUTINE E3110(TZ1,TZ2,TZ3,TS)			
000000B 114.	COMMON/KSTRP/KSTRP(20)			
000000B 115.	COMMON/DPLUS/DPLUS			
000000B 116.	COMMON/CHINON/CHINON(12)			
000000B 117.	COMMON/NPAR08/NPAR,NOBS,KSTAR			
000000B 118.	TS=(TZ1*TZ2*TZ3/(DPLUS**5))*(3.0*CHINON(KSTRP(8))+TZ1*TZ1*CHINON(K STRP(10)))			
000013B 119.	RETURN			
000015B 120.	END			

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS (B=RELATIVE ADDRESS)
(C= RELATIVE TO //)

000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B 121.	SUBROUTINE E2210(TZ1,TZ2,TZ3,TS)			
000000B 122.	COMMON/KSTRP/KSTRP(20)			
000000B 123.	COMMON/DPLUS/DPLUS			
000000B 124.	COMMON/CHINON/CHINON(12)			
000000B 125.	COMMON/NPAR08/NPAR,NOBS,KSTAR			
000000B 126.	TS=(TZ3/(DPLUS**5))*(CHINON(KSTRP(6))+(TZ1*TZ1+TZ2*TZ2)*CHINON(KST RP(8))+TZ1*TZ1*TZ2*TZ2*CHINON(KSTRP(10)))			
000015B 127.	RETURN			
000020B 128.	END			

```

000000B 129.      SUBROUTINE E2111(TZ1,TZ2,TZ3,TZ4,TS)
000000B 130.      COMMON/KSTRP/KSTRP(20)
000000B 131.      COMMON/DPLUS/DPLUS
000000B 132.      COMMON/CHINON/CHINON(12)
000000B 133.      COMMON/NPAROB/NPAR,NOBS,KSTAR
000000B 134.      TS=(TZ2*TZ3*TZ4/(DPLUS**5))*(CHINON(KSTRP(8))+TZ1*TZ1*CHINON(KSTRP
134.))
000012B 135.      RETURN
000015B 136.      END

```

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B 137.	SUBROUTINE E11111(TZ1,TZ2,TZ3,TZ4,TZ5,TS)			
000000B 138.	COMMON/KSTRP/KSTRP(20)			
000000B 139.	COMMON/DPLUS/DPLUS			
000000B 140.	COMMON/CHINON/CHINON(12)			
000000B 141.	COMMON/NPAROB/NPAR,NOBS,KSTAR			
000000B 142.	TS=(TZ1*TZ2*TZ3*TZ4*TZ5/(DPLUS**5))*CHINON(KSTRP(10))			
000007B 143.	RETURN			
000012B 144.	END			

NUMBER AND NAME CROSS REFERENCE MAP (/N= N REFERENCES IN STATEMENT)

VARIABLE AND ARRAY NAMES SORTED BY ADDRESS		(B=RELATIVE ADDRESS) (C= RELATIVE TO //)		
000000C CHINON	000000C DPLUS	000000C KSTRP	000000C NPAR	000001C NOBS
000000B 145.	SUBROUTINE EXPZ(TZZ,MMS,TPP)			
000000B 146.	TZ=TZZ			
000000B 147.	TP=TPP			
000002B 148.	IF(MMS.EQ.1) TP=TP*TZ			
000006B 149.	IF(MMS.EQ.2) TP=TP*(1.0+TZ*TZ)			
000013B 150.	IF(MMS.EQ.3) TP=TP*(3.0*TZ+TZ*TZ*TZ)			
000020B 151.	IF(MMS.EQ.4) TP=TP*(3.0+6.0*TZ*TZ+TZ*TZ*TZ)			
000027B 152.	TPP=TP			
000027B 153.	RETURN			
000032B 154.	END			

References

- Bard, Y. (1974). Nonlinear parameter estimation. Academic Press.
New York and London.
- Barnard, G.A. (1960). Discussion on Mr. Beale's paper, J. Roy. Statist. Soc. Ser. B, 22, 78-79.
- Bartholomew, D.J. (1970). A comparison of frequentist and Bayesian approaches to inference with prior knowledge. Foundations of statistical inference - Proceedings of a symposium held at University of Waterloo.
- Bartholomew, D.J. (1961). A test of homogeneity of means under restricted alternatives. J. Roy. Statist. Soc. Ser. B. 23, 239-281.
- Beale, E.M.L. (1960). Confidence regions in nonlinear estimation. J. Roy. Statist. Soc. Ser. B. 22, 41-76.
- Box, G.E.P. and Cox, D.R. (1964). An analysis of transformations. J. Roy. Statist. Soc., Series B, 26, 211-252.
- Box, M.J. (1971). Bias in nonlinear estimation. J. Roy. Statist. Soc. Ser. B. 171-190.
- Graybill, F.A. (1961). An introduction to linear statistical models, Vol.I, McGraw-Hill: New York.
- Guttman, I. and Meeter, D.A. (1965). On Beale's measures of nonlinearity. Technometric, vol.7, No.4, November, 1965.
- Guttman, I., Pereyra, V. and Scolnik, H.D. (1973). Least squares estimation for a class of nonlinear models. Technometric. Vol.15, No.2, May, 1973.
- Halperin, M. (1963). Confidence interval estimation in nonlinear regression. J.R. Statist. Soc. B, 25, 330-333.
- Halperin, M. (1964). Note on interval estimation in nonlinear regression when responses are correlated. J. Roy. Statist. Soc. Series B, 26, 267-269.

- Hartley, H.O. (1964). Exact confidence regions for the parameters in nonlinear regression laws. *Biometrika*. 51, 347-353.
- Lehmann, E.L. (1959). Testing statistical hypotheses. Wiley: New York.
- Linssen, H.N. (1975). Nonlinearity measures: a case study. *Statistica Neerlandica* Vol.29, no.3, Rotterdam.
- Ross, G.J.S. (1970). The efficient use of function minimization in nonlinear maximum likelihood estimation. *J.R. Statist. Soc., C*. 19, 205-221.
- Ross, G.J.S. (1975). Simple nonlinear modelling for the general user. Warszawa 1-9 IX. ISI/BS. Invited paper 81.
- Seber, G.A.F. (1963). The non-central chi-squared and beta distributions. *Biometrika*, 50, 542-544.
- Seber, G.A.F. (1964). The linear hypothesis and large sample theory. *Annals Math. Statist.*, 35, 773-779.
- Wald, A. (1943). Tests of statistical hypotheses concerning several parameters when the number of observations is large. *Trans. Amer. Math. Soc.*, 54, 426-482.
- Wilks, S.S. and Daly, J.F. (1939). An optimum property of confidence region associated with the likelihood function. *Ann. Math. Statist.*, 10, 225-235.
- Williams, E.J. (1962). Exact fiducial limits in non-linear estimation, *J.R. Statist. Soc. B*, 24, 125-139.