Linear Statistical Inference and its Applications
WILEY SERIES IN PROBABILITY AND STATISTICS

Established by WALTER A. SHEWHART and SAMUEL S. WILKS


A complete list of the titles in this series appears at the end of this volume.
To Bhargavi
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The purpose of this book is to present up-to-date theory and techniques of statistical inference in a logically integrated and practical form. Essentially, it incorporates the important developments in the subject that have taken place in the last three decades. It is written for readers with a background knowledge of mathematics and statistics at the undergraduate level.

Quantitative inference, if it were to retain its scientific character, could not be divested of its logical, mathematical, and probabilistic aspects. The main approach to statistical inference is inductive reasoning, by which we arrive at "statements of uncertainty." The rigorous expression that degrees of uncertainty require are furnished by the mathematical methods and probability concepts which form the foundations of modern statistical theory. It was my awareness that advanced mathematical methods and probability theory are indispensable accompaniments in a self-contained treatment of statistical inference that prompted me to devote the first chapter of this book to a detailed discussion of vector spaces and matrix methods and the second chapter to a measure-theoretic exposition of probability and development of probability tools and techniques.

Statistical inference techniques, if not applied to the real world, will lose their import and appear to be deductive exercises. Furthermore, it is my belief that in a statistical course emphasis should be given to both mathematical theory of statistics and to the application of the theory to practical problems. A detailed discussion on the application of a statistical technique facilitates better understanding of the theory behind the technique. To this end, in this book, live examples have been interwoven with mathematical results. In addition, a large number of problems are given at the end of each chapter. Some are intended to complement main results derived in the body of the chapter, whereas others are meant to serve as exercises for the reader to test his understanding of theory.

The selection and presentation of material to cover the wide field of
statistical inference have not been easy. I have been guided by my own experience in teaching undergraduate and graduate students, and in conducting and guiding research in statistics during the last twenty years. I have selected and presented the essential tools of statistics and discussed in detail their theoretical bases to enable the readers to equip themselves for consultation work or for pursuing specialized studies and research in statistics.

Why Chapter 1 provides a rather lengthy treatment of the algebra of vectors and matrices needs some explanation. First, the mathematical treatment of statistical techniques in subsequent chapters depends heavily on vector spaces and matrix methods; and second, vector and matrix algebra constitute a branch of mathematics widely used in modern treatises on natural, biological, and social sciences. The subject matter of the chapter is given a logical and rigorous treatment and is developed gradually to an advanced level. All the important theorems and derived results are presented in a form readily adaptable for use by research workers in different branches of science.

Chapter 2 contains a systematic development of the probability tools and techniques needed for dealing with statistical inference. Starting with the axioms of probability, the chapter proceeds to formulate the concepts of a random variable, distribution function, and conditional expectation and distributions. These are followed by a study of characteristic functions, probability distributions in infinite dimensional product spaces, and all the important limit theorems. Chapter 2 also provides numerous propositions, which find frequent use in some of the other chapters and also serve as good equipment for those who want to specialize in advanced probability theory.

Chapter 3 deals with continuous probability models and the sampling distributions needed for statistical inference. Some of the important distributions frequently used in practice, such as the normal, Gamma, Cauchy, and other distributions, are introduced through appropriate probability models on physical mechanisms generating the observations. A special feature of this chapter is a discussion of problems in statistical mechanics relating to the equilibrium distribution of particles.

Chapter 4 is devoted to inference through the technique of analysis of variance. The Gauss-Markoff linear model and the associated problems of estimation and testing are treated in their wide generality. The problem of variance-components is considered as a special case of the more general problem of estimating intraclass correlation coefficients. A unified treatment is provided of multiclassified data under different sampling schemes for classes within categories.

The different theories and methods of estimation form the subject matter of Chapter 5. Some of the controversies on the topic of estimation are examined; and to remove some of the existing inconsistencies, certain modifications are introduced in the criteria of estimation in large samples.

Problems of specification, and associated tests of homogeneity of parallel
samples and estimates, are dealt with in Chapter 6. The choice of a mathematical model from which the observations could be deemed to have arisen is of fundamental importance because subsequent statistical computations will be made on the framework of the chosen model. Appropriate tests have been developed to check the adequacy of proposed models on the basis of available facts.

Chapter 7 provides the theoretical background for the different aspects of statistical inference, such as testing of hypotheses, interval estimation, experimentation, the problem of identification, nonparametric inference, and so on.

Chapter 8, the last chapter, is concerned with inference from multivariate data. A special feature of this chapter is a study of the multivariate normal distribution through a simple characterization, instead of through the density function. The characterization simplifies the multivariate theory and enables suitable generalizations to be made from the univariate theory without further analysis. It also provides the necessary background for studying multivariate normal distributions in more general situations, such as distributions on Hilbert space.

Certain notations have been used throughout the book to indicate sections and other references. The following examples will help in their interpretation. A subsection such as 4f.3 means subsection 3 in section f of Chapter 4. Equation (4f.3.6) is the equation numbered 6 in subsection 4f.3 and Table 4f.3p is the table numbered second in subsection 4f.3. The main propositions (or theorems) in each subsection are numbered: (i), (ii), etc. A back reference such as [(iii). 5d.2] indicates proposition (iii) in subsection 5d.2.

A substantial part of the book was written while I was a visiting professor at the Johns Hopkins University, Baltimore, in 1963–1964, under a Senior Scientist Fellowship scheme of the National Science Foundation, U.S.A. At the Johns Hopkins University, I had the constant advice of G. S. Watson, Professor of Statistics, who read the manuscript at the various stages of its preparation. Comments by Herman Chernoff on Chapters 7 and 8, by Rupert Miller and S. W. Dharmadhikari on Chapter 2, and by Ralph Bradley on Chapters 1 and 3, have been extremely helpful in the preparation of the final manuscript. I wish to express my thanks to all of them. The preparation and revision of the manuscript would not have been an easy task without the help of G. M. Das, who undertook the heavy burden of typing and organizing the manuscript for the press with great care and diligence.

Finally, I wish to express my gratitude to the late Sir Ronald A. Fisher and to Professor P. C. Mahalanobis under whose influence I have come to appreciate statistics as the new technology of the present century.

Calcutta, India
June, 1965

C. R. Rao
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Preface to the
Second Edition

As in the first edition, the aim has been to provide in a single volume a full discussion of the wide range of statistical methods useful for consulting statisticians and, at same time, to present in a rigorous manner the mathematical and logical tools employed in deriving statistical procedures, with which a research worker should be familiar.

A good deal of new material is added, and the book is brought up to date in several respects, both in theory and applications.

Some of the important additions are different types of generalized inverses, concepts of statistics and subfields, MINQUE theory of variance components, the law of iterated logarithms and sequential tests with power one, analysis of dispersion with structural parameters, discrimination between composite hypotheses, growth models, theorems on characteristic functions, etc.

Special mention may be made of the new material on estimation of parameters in a linear model when the observations have a possibly singular covariance matrix. The existing theories and methods due to Gauss (1809) and Aitken (1935) are applicable only when the covariance matrix is known to be nonsingular. The new unified approaches discussed in the book (Section 4i) are valid for all situations whether the covariance matrix is singular or not.

A large number of new exercises and complements have been added.

I wish to thank Dr. M. S. Avadhani, Dr. J. K. Ghosh, Dr. A. Maitra, Dr. P. E. Nüesch, Dr. Y. R. Sarma, Dr. H. Toutenberg, and Dr. E. J. Williams for their suggestions while preparing the second edition.

New Delhi

C. R. Rao

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Chapter 1

ALGEBRA OF VECTORS
AND MATRICES

Introduction. The use of matrix theory is now widespread in both pure mathematics and the physical and the social sciences. The theory of vector spaces and transformations (of which matrices are a special case) have not, however, found a prominent place, although they are more fundamental and offer a better understanding of problems. The vector space concepts are essential in the discussion of topics such as the theory of games, economic behavior, prediction in time series, and the modern treatment of univariate and multivariate statistical methods.

The aim of the first chapter is to introduce the reader to the concepts of vector spaces and the basic results. All important theorems are discussed in great detail to enable the beginner to work through the chapter. Numerous illustrations and problems for solution are given as an aid to further understanding of the subject. To introduce wide generality (this is important and should not cause any difficulty in understanding the theory) the elements used in the operations with vectors are considered as belonging to any Field in which addition and multiplication are defined in a consistent way (as in the ordinary number system). Thus, the elements $e_1, e_2, \ldots$ (finite or infinite in number) are said to belong to a field $F$, if they are closed under the operations of addition ($e_i + e_j$) and multiplication ($e_i e_j$), that is, sums and products of elements of $F$ also belong to $F$, and satisfy the following conditions:

\begin{align*}
(A_1) & \quad e_i + e_j = e_j + e_i \quad \text{(commutative law)} \\
(A_2) & \quad e_i + (e_j + e_k) = (e_i + e_j) + e_k \quad \text{(associative law)} \\
(A_3) & \quad \text{For any two elements } e_i, e_j, \text{ there exists an element } e_k \text{ such that } e_i + e_k = e_j.
\end{align*}

The condition $(A_3)$ implies that there exists an element $e_0$ such that $e_i + e_0 = e_i$ for all $i$. The element $e_0$ is like 0 (zero) of the number system.
2 ALGEBRA OF VECTORS AND MATRICES  

(M_1) \quad e_i e_j = e_j e_i \quad \text{(commutative law)}

(M_2) \quad e_i (e_j e_k) = (e_i e_j) e_k \quad \text{(associative law)}

(M_3) \quad e_i (e_j + e_k) = e_i e_j + e_i e_k \quad \text{(distributive law)}

(M_4) \quad \text{For any two elements } e_i \text{ and } e_j \text{ such that } e_i \neq e_0, \text{ the zero element, there exists an element } e_k \text{ such that } e_i e_k = e_j.

(M_4) \text{ implies that there is an element } e_i \text{ such that } e_i e_i = e_i \text{ for all } i. \text{ The element } e_i \text{ is like 1 (unity) of the number system.}

The study of vector spaces is followed by a discussion of the modern matrix theory and quadratic forms. Besides the basic propositions, a number of results used in mathematical physics, economics, biology and statistics, and numerical computations are brought together and presented in a unified way. This would be useful for those interested in applications of the matrix theory in the physical and the social sciences.

1a VECTOR SPACES

1a.1 Definition of Vector Spaces and Subspaces

Concepts such as force, size of an organism, an individual’s health or mental abilities, and price level of commodities cannot be fully represented by a single number. They have to be understood by their manifestations in different directions, each of which may be expressed by a single number. The mental abilities of an individual may be judged by his scores \((x_1, x_2, \ldots, x_k)\) in \(k\) specified tests. Such an ordered set of measurements may be simply represented by \(x\), called a row vector. If \(y = (y_1, y_2, \ldots, y_k)\) is the vector of scores for another individual, the total scores for the two individuals in the various tests may be represented by \(x + y\) with the definition \(x + y = (x_1 + y_1, x_2 + y_2, \ldots, x_k + y_k)\). (1a.1.1)

This rule of combining or adding two vectors is the same as that for obtaining the resultant of two forces in two or three dimensions, known as the parallelogram law of forces. Algebraically this law is equivalent to finding a force whose components are the sum of the corresponding components of the individual forces.

Given a force \(f = (f_1, f_2, f_3)\), it is natural to define \(cf = (cf_1, cf_2, cf_3)\) (1a.1.2) as a force \(c\) times the first, which introduces a new operation of multiplying a vector by a scalar number such as \(c\). Further, given a force \(f\), we can counterbalance it by adding a force \(g = (-f_1, -f_2, -f_3) = (-1)f\), or by applying \(f\) in the opposite direction, we have the resulting force \(f + g = (0, 0, 0)\). Thus we have the concepts of a negative vector such as ‘\(-f\)’ and a null vector \(0 = (0, 0, 0)\), the latter having the property \(x + 0 = x\) for all \(x\).
It is seen that we are able to work with the new quantities, to the extent permitted by the operations defined, in the same way as we do with numbers. As a matter of fact, we need not restrict our algebra to the particular type of new quantities, viz., ordered sets of scalars, but consider a collection of elements \( x, y, z, \ldots \), finite or infinite, which we choose to call vectors and \( c_1, c_2, \ldots \), scalars constituting a field (like the ordinary numbers with the operations of addition, subtraction, multiplication, and division suitably defined) and lay down certain rules of combining them.

**Vector Addition.** The operation of addition indicated by \(+\) is defined for any two vectors leading to a vector in the set and is subject to the following rules:

(i) \[ x + y = y + x \] (commutative law)
(ii) \[ x + (y + z) = (x + y) + z \] (associative law)

**Null Element.** There exists an element in the set denoted by \( 0 \) such that

(iii) \[ x + 0 = x, \quad \text{for all } x. \]

**Inverse (Negative) Element.** For any given element \( x \), there exists a corresponding element \( \xi \) such that

(iv) \[ x + \xi = 0. \]

**Scalar Multiplication.** The multiplication of a vector \( x \) by a scalar \( c \) leads to a vector in the set, represented by \( cx \), and is subject to the following rules.

(v) \[ c(x + y) = cx + cy \] (distributive law for vectors)
(vi) \[ (c_1 + c_2)x = c_1x + c_2x \] (distributive law for scalars)
(vii) \[ c_1(c_2x) = (c_1c_2)x \] (associative law)
(viii) \[ ex = x \] (where \( e \) is the unit element in the field of scalars).

A collection of elements (with the associated field of scalars \( F \)) satisfying the axioms (i) to (viii) is called a *linear vector space* \( \mathcal{V} \) or more explicitly \( \mathcal{V}(F) \) or \( \mathcal{V}_F \). Note that the conditions (iii) and (iv) can be combined into the single condition that for any two elements \( x \) and \( y \) there is a unique element \( z \) such that \( x + z = y \).

The reader may satisfy himself that, for a given \( k \), the collection of all ordered sets \( (x_1, \ldots, x_k) \) of real numbers with the addition and the scalar multiplication as defined in (1a.1.1) and (1a.1.2) form a vector space. The same is true of all polynomials of degree not greater than \( k \), with coefficients belonging to any field and addition and multiplication by a constant being defined in the usual way. Although we will be mainly concerned with vectors
which are ordered sets of real numbers in the treatment of statistical methods covered in this book, the axiomatic set up will give the reader proper insight into the new algebra and also prepare the ground for a study of more complicated vector spaces, like Hilbert space (see Halmos, 1951), which are being increasingly used in the study of advanced statistical methods.

A linear subspace, subspace, or linear manifold in a vector space \( \mathcal{V} \) is any subset of vectors \( \mathcal{M} \) closed under addition and scalar multiplication, that is, if \( x \) and \( y \in \mathcal{M} \), then \( (cx + dy) \in \mathcal{M} \) for any pair of scalars \( c \) and \( d \). Any such subset \( \mathcal{M} \) is itself a vector space with respect to the same definition of addition and scalar multiplication as in \( \mathcal{V} \). The subset containing the null vector alone, as well as that consisting of all the elements in \( \mathcal{V} \), are extreme examples of subspaces. They are called improper subspaces whereas others are proper subspaces.

As an example, all linear combinations of a given fixed set \( S \) of vectors \( \alpha_1, \ldots, \alpha_k \) is a subspace called the linear manifold \( \mathcal{M}(S) \) spanned by \( S \). This is the smallest subspace containing \( S \).

Consider \( k \) linear equations in \( n \) variables \( x_1, \ldots, x_n \),

\[
a_{i1} x_1 + \cdots + a_{in} x_n = 0, \quad i = 1, \ldots, k,
\]

where \( a_{ij} \) belongs to any field \( F \). The reader may verify that the totality of solutions \( (x_1, \ldots, x_n) \) considered as vectors constitutes a subspace with the addition and scalar multiplication as defined in (1a.1.1) and (1a.1.2).

1a.2 Basis of a Vector Space

A set of vectors \( u_1, \ldots, u_k \) is said to be linearly dependent if there exist scalars \( c_1, \ldots, c_k \), not all simultaneously zero, such that \( c_1 u_1 + \cdots + c_k u_k = 0 \), otherwise it is independent. With such a definition the following are true:

1. The null vector by itself is a dependent set.
2. Any set of vectors containing the null vector is a dependent set.
3. A set of non-zero vectors \( u_1, \ldots, u_k \) is dependent when and only when a member in the set is a linear combination of its predecessors.

A linearly independent subset of vectors in a vector space \( \mathcal{V} \), generating or spanning \( \mathcal{V} \) is called a basis (Hamel basis) of \( \mathcal{V} \).

(i) Every vector space \( \mathcal{V} \) has a basis.

To demonstrate this, let us choose sequentially non-null vectors \( \alpha_1, \alpha_2, \ldots \), in \( \mathcal{V} \) such that no \( \alpha_i \) is dependent on its predecessors. In this process it may so happen that after the \( k \)th stage no independent vector is left in \( \mathcal{V} \), in which case \( \alpha_1, \ldots, \alpha_k \) constitute a basis and \( \mathcal{V} \) is said to be a finite \((k)\) dimensional vector space. On the other hand, there may be no limit to the process of
choosing $\alpha_i$, in which case $\mathcal{V}$ is said to have infinite dimensions. Further argument is needed to show the actual existence of an infinite set of independent vectors which generate all the vectors. This is omitted as our field of study will be limited to finite dimensional vector spaces. The following results concerning finite dimensional spaces are important.

(ii) If $\alpha_1, \ldots, \alpha_k$ and $\beta_1, \ldots, \beta_s$ are two alternative choices for a basis, then $s = k$.

Let, if possible, $s > k$. Consider the dependent set $\beta_1, \alpha_1, \ldots, \alpha_k$. If $\alpha_i$ depends on the predecessors, then $\mathcal{V}$ can also be generated by $\beta_1, \alpha_1, \ldots, \alpha_{i-1}, \alpha_{i+1}, \ldots, \alpha_k$, in which case the set $\beta_2, \beta_1, \alpha_1, \ldots, \alpha_{i-1}, \alpha_{i+1}, \ldots, \alpha_k$ is dependent. One more $\alpha$ can now be omitted. The process of adding a $\beta$ and omitting an $\alpha$ can be continued (observe that no $\beta$ can be eliminated at any stage) till the set $\beta_1, \ldots, \beta_k$ is left, which itself spans $\mathcal{V}$ and hence $(s-k)$ of the $\beta$ vectors are redundant. The cardinal number $k$ common to all bases represents the minimal number of vectors needed to span the space or the maximum number of independent vectors in the space. We call this number as the dimension or rank of $\mathcal{V}$ and denote it by $d(\mathcal{V})$.

(iii) Every vector in $\mathcal{V}$ has a unique representation in terms of a given basis.

If $\sum a_i \alpha_i$ and $\sum b_i \alpha_i$ represent the same vector, then $\sum (a_i - b_i) \alpha_i = 0$, which is not possible unless $a_i - b_i = 0$ for all $i$, since $\alpha_i$ are independent.

The Euclidean space $E_k(F)$ of all ordered sets of $k$ elements $(x_1, \ldots, x_k)$, $x_i \in F$ (a field of elements) is of special interest. The vectors may be considered as points of $k$ dimensional Euclidean space. The vectors $e_1 = (1, 0, \ldots, 0)$, $e_2 = (0, 1, 0, \ldots, 0)$, $\ldots$, $e_k = (0, 0, \ldots, 1)$ are in $E_k$ and are independent, and any vector $x = (x_1, \ldots, x_k) = x_1 e_1 + \cdots + x_k e_k$. Therefore $d(E_k) = k$, and the vectors $e_1, \ldots, e_k$ constitute a basis and thus any other independent set of $k$ vectors. Any vector in $E_k$ can, therefore, be represented as a unique linear combination of $k$ independent vectors, and, naturally, as a combination (not necessarily unique) of any set of vectors containing $k$ independent vectors.

When $F$ is the field of real numbers the vector space $E_k(F)$ is denoted simply by $R^k$. In the study of design of experiments we use Galois fields ($GF$) with a finite number of elements, and consequently the vector space has only a finite number of vectors. The notation $E_k(GF)$ may be used for such spaces. The vector space $E_k(F)$ when $F$ is the field of complex numbers is represented by $U^k$ and is called a $k$ dimensional unitary space. The treatment of Section 1a.3 is valid for any $F$. Later we shall confine our attention to $R^k$ only. We prove an important result in (iv) which shows that study of finite dimensional vector spaces is equivalent to that of $E_k$.

(iv) Any vector space $\mathcal{V}_F$ for which $d(\mathcal{V}_F) = k$ is isomorphic to $E_k$. 
If \( \alpha_1, \ldots, \alpha_k \) is a basis of \( \mathcal{V}_F \), then an element \( u \in \mathcal{V}_F \) has the representation
\[
u = a_1\alpha_1 + \cdots + a_k\alpha_k \quad \text{where} \quad a_i \in F, \quad i = 1, \ldots, k.
\]
The correspondence
\[
u \rightarrow \nu^* = (a_1, \ldots, a_k), \quad \nu^* \in F_k(F)
\]
establishes the isomorphism. For if \( u \rightarrow \nu^*, \quad v \rightarrow \nu^* \), then \( u + v \rightarrow \nu^* + \nu^* \) and \( cu \rightarrow cu^* \). This result also shows that any two vector spaces with the same dimensions are isomorphic.

### 1a.3 Linear Equations

Let \( \alpha_1, \ldots, \alpha_m \) be \( m \) fixed vectors in an arbitrary vector space \( \mathcal{V}_F \) and consider the linear equation in the scalars \( x_1, \ldots, x_m \in F \) (associated Field)
\[
x_1\alpha_1 + \cdots + x_m\alpha_m = 0. \tag{1a.3.1}
\]

(i) A necessary and sufficient condition that (1a.3.1) has a nontrivial solution, that is, not all \( x_i \) simultaneously zero, is that \( \alpha_1, \ldots, \alpha_m \) should be dependent.

(ii) The solutions considered as (row) vectors \( \mathbf{x} = (x_1, \ldots, x_m) \) in \( F_m(F) \)

This is true for if \( \mathbf{x} \) and \( \mathbf{y} \) are solutions then \( a\mathbf{x} + b\mathbf{y} \) is also a solution. Note that \( \alpha_i \), themselves may belong to any vector space \( \mathcal{V}_F \).

(iii) Let \( \mathcal{S} \) be the linear manifold or the subspace of solutions and \( \mathcal{M} \), that of the vectors \( \alpha_1, \ldots, \alpha_m \). Then \( d[\mathcal{S}] = m - d[\mathcal{M}] \) where the symbol \( d \) denotes the dimensions of the space.

Without loss of generality let \( \alpha_1, \ldots, \alpha_k \) be independent, that is, \( d[\mathcal{M}] = k \), in which case,
\[
\alpha_j = a_{j1}\alpha_1 + \cdots + a_{jk}\alpha_k, \quad j = k + 1, \ldots, m.
\]

We observe that the vectors
\[
\beta_1 = (a_{k+1,1}, \ldots, a_{k+1,k}, -1, 0, \ldots, 0) \\
\cdots \\
\beta_{m-k} = (a_{m,1}, \ldots, a_{m,k}, 0, 0, \ldots, -1) \tag{1a.3.2}
\]
are independent and satisfy the equation (1a.3.1). The set (1a.3.2) will be a basis of the solution space \( \mathcal{S} \) if it spans all the solutions. Let \( \mathbf{y} = (y_1, \ldots, y_m) \) be any solution and consider the vector
\[
y + y_{k+1}\beta_1 + \cdots + y_m\beta_{m-k} \tag{1a.3.3}
\]
which is also a solution. But this is of the form \( (z_1, \ldots, z_k, 0, \ldots, 0) \) which means \( z_1\alpha_1 + \cdots + z_k\alpha_k = 0 \). This is possible only when \( z_1 = \cdots = z_k = 0 \).