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MONOGRAPHS**

Volume 53

**Statistical Decision
Rules and Optimal
Inference**

N. N. Čencov



American Mathematical Society

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Providence, Rhode Island

СТАТИСТИЧЕСКИЕ
РЕШАЮЩИЕ ПРАВИЛА
И ОПТИМАЛЬНЫЕ
ВЫВОДЫ

Н. Н. ЧЕНЦОВ

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ABSTRACT. This monograph is devoted to the general theory of statistical inference. The approaches developed here permit the author to consider, from a single point of view, the main concepts and laws of mathematical statistics, methods of constructing optimal statistical estimates, etc. The book is intended for those working in mathematical statistics, information theory, game theory, and also applications of probabilistic and statistical methods. Individual sections may be of interest to specialists in measure theory, differential geometry and nonlinear functional analysis.

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PREFACE

The general concepts of a statistical decision and a statistical decision rule are basic for all of modern statistical theory. According to Wald, every particular statistical problem is a problem of decision-making: the statistician, having processed certain observational material, must draw conclusions as to the observed phenomenon. Since the outcome of each observation is random, one cannot usually expect these conclusions to be absolutely accurate. It is a job for the theory to ascertain the minimal unavoidable uncertainty of the conclusions in the problem and to indicate an optimal decision rule.

In classical problems of mathematical statistics, one is required to determine the (unknown) probability distribution of the outcomes on the basis of independent observations and certain additional information. When the number of observations used in such cases increases, one can establish various simple and quite general asymptotic relationships.

In any theory, a general law should be amenable to equivalent formulations; that is to say, the statement of the law should not vary when a situation is replaced by another one, equivalent to the former (within the framework of that theory); otherwise it would not be a general law. In classical geometry, such "changes of situation" form a group. In mathematical statistics the description of the set of equivalent situations is more complicated.

The system of all statistical decision rules for all conceivable statistical problems, together with the natural operation of composition, forms an algebraic category. This category generates a uniform geometry of families of probability laws, in which the "figures" are the families and the "motions" are the decision rules. Two families are "congruent" if and only if they possess equivalent statistical properties.

An apt name for the subject of this monograph might be "geometrical statistics". The algebra of decision rules and the natural geometry that it

generates are studied here from a statistical standpoint. The geometrical methods and language that we develop are then applied to the equivariant theory of optimal estimates.

The book is intended for specialists in mathematical statistics, information theory and game theory, and also for those interested in applications of probability-theoretical methods. The reader is expected to be familiar with probability theory and measure theory, to the extent of Kolmogorov's *Grundbegriffe der Wahrscheinlichkeitsrechnung* and Halmos' *Measure Theory*, or Neveu's *Bases Mathématiques du Calcul des Probabilités*. Formally speaking, a knowledge of statistics is not assumed; all the necessary concepts are introduced and explained in §1 of the Introduction. Also included in the Introduction is the requisite material from the theory of smooth manifolds and category theory.

My interest in the "uncertainty" of statistical estimates arose when I was engaged in the development of computer-oriented methods for the estimation of unknown densities; further encouragement came from my teacher N. V. Smirnov. Without his approval, I would probably not have risked taking my research so far from orthodox statistics.

My discussions with Ju. V. Linnik had considerable influence on my work. Many of the theorems proved here are answers to his clearly phrased questions. I am indebted to him for his unflagging interest in my research and his kind attention.

I had many useful discussions with experts concerning both the basic ideas of the theory and individual results; thanks are due in this respect to L. N. Bol'shev, B. N. Delone, I. M. Gel'fand, B. V. Gnedenko, A. M. Kagan, A. N. Kolmogorov, A. A. Ljapunov, Ju. V. Prohorov, Richard Sacksteder, Ju. M. Smirnov, Charles Stein and V. S. Vladimirov.

It is my pleasure to thank Elena Aleksandrovna Morozova for her great help and valuable advice on geometrical matters, and to Fridrih Izrailevič Karpelevič for numerous profound and important remarks.

Finally, I wish to thank the book's editor, A. V. Černavskiĭ, who read the manuscript attentively and helped to eliminate various shortcomings.

N. N. Čencov

NOTES AND COMMENTS

§1

A general description of equivalent statistical problems was given by Blackwell [2, 1951], [3, 1953]. At the same time, similar ideas were developed by Stein [1951, oral communication; see 3] and Sherman [1, 1951]. Earlier work had considered only equivalent reduction of statistical problems connected with the transition to sufficient statistics; see Fisher [3, 1925], and also Halmos and Savage [1, 1949], Bahadur [1, 1954], Burkholder [1], and LeCam [2, 1964], [3, 1969]. Cencov [5, 1965] and Morse and Sacksteder [1, 1966] noticed that this equivalence relation was, roughly speaking, generated by the category of statistical decision rules. This enabled them to “algebraicize” certain concepts of statistics. Further steps in this direction were taken by Romier and coworkers in their “*Introduction à la statistique mathématique*” (Romier [2], [3], Littaye-Petit et al. [1], Martin and Vaguelsy [1], Laurant et al. [1], Martin et al. [1]). The idea of considering families of probability distributions not as objects on their own but as “figures” in a Kleinian geometry with a category of transformations is due to the author [5], [7].

The systematic investigation of statistical properties which are invariant (equivariant) in the category of statistical decision rules was initiated by Sacksteder and Cencov. Attention was paid previously only to invariance determined by the (group) symmetry of a specific problem; see Lehmann [1]; also Berk [1] and Brillinger [1].

(1) The history of statistics as the science of statistical inference usually begins with the amusing episode recounted by Bertrand in the preface to his course “*Calcul des probabilités*” [1]:

“One day in Naples the reverend Galiani saw a man from the Basilicata who, shaking three dice in a cup, wagered to throw three sixes . . . Such luck is possible, you say. Yet the man succeeded a second time, and the bet was repeated. He put back the dice in the cup, three, four, five times, and each time he produced three sixes. “Sangue di Bacco”, exclaimed the reverend, ‘the dice are loaded!’ And they were . . .” [Quoted from G. Pólya, *Patterns of Plausible Inference*.]

(2) This approach was first put forth explicitly by Neyman and Pearson [1, 1928] in their theory of hypothesis testing.

(3) Essentially, this is a special case of the game-theoretical approach. Long ago Laplace [1, 1820] likened the derivation of an estimate to a game of chance in which the statistician suffers defeat if his estimates are inferior.

A detailed description of the concept of “statistical problem” according to Wald [1], [3] was given by Lehmann [1]. See also Birnbaum [1], De Groot and Rao [1], and Hoeffding [1].

(4) Choice of action based on results of an auxiliary experiment was first systematically considered by von Neumann in game theory. See von Neumann [1, 1928], and also von Neumann and Morgenstern [1] and Blackwell and Girshick [1] (cf. footnote (3), above).

(5) The statistician deals directly with the outcomes of the experiment only in the simplest problems, involving a discrete sample space. The slightest additional complication in the problem leads to distortion of the experimental result, due to measurement errors and grouping (when of necessity the measurement is rounded off). In such situations the decision is described by a compound decision rule $\Pi_{\text{instrument}} \circ \Pi_{\text{processing}}$, where the first factor is independent of the statistician. In this book we shall consider only the ideal situation, in which the distortions due to $\Pi_{\text{instrument}}$ may be neglected.

§2

(1) Cap = Collection of All Probability distributions.

(2) In all topological concepts we follow Kelley [1].

(3) In §13 we shall introduce the Hellinger integrals of the Radon-Nikodým derivative, such as $\int [(dQ/dP)(\omega)]^2 P\{d\omega\}$, as the limits of integral sums of a special form, with specific conventions for resolution of the indeterminacy $\infty \cdot 0$, different from the usual.

(4) See the work of C. Ionescu Tulcea, under whose "lifting" $P\{\cdot\} \rightarrow p(\cdot)$, to a (finite) linear combination of laws corresponds the linear combination of the densities, and to a geodesic mean (see Definition 18.2) corresponds the normalized geometric mean of the densities. The correspondence may fail for countable linear combinations (see end of §9.4).

(5) The possibility of modifying almost proper conditional distributions to obtain proper distributions depends only on the properties of the mapping f . If there exists a proper conditional distribution for at least one Q , it will serve to modify any $P\{\cdot|f\}$.

Blackwell and Ryll-Nardzewski [1] proved that not every almost proper Borel conditional distribution can be modified by a Borel procedure to obtain a proper conditional distribution. Their arguments rely on the fact, established by Novikov [1], that it is not possible to B-uniformize an arbitrary Borel function (see subsection 14).

(6) This approach to the theory of stochastic processes was developed by Lévy [1]; see also Wiener [1]. For the connection with the theory of Monte Carlo methods, see Gel'fand, Frolov and Cencov [1], Rankin [1] and Cencov [11]. A similar approach was suggested by Blackwell [4] (see also Sazonov [1]).

(7) Provided that the initial function is not many-valued. Nevertheless, any B-function may be uniformized by an A(B)-function, where A(B) is the σ -algebra generated by the analytic sets (see Saks [1]) $B \subset A(B) \subset B^*$. See also Luzin [1], Luzin and Novikov [1], and Luzin and Sierpiński [1].

(8) The proofs of Lemma 5.11 and all other assertions of §5 rely only on general facts of measure theory and make no use of the concept of a constructive measure. Therefore the proof given here of Lemma 2.15 contains no vicious circle.

§3

(1) Applying the Taylor formula with integral remainder to the function $f(x) = f(\varphi_\alpha^{-1}(x))$ of the local coordinates $(x^{(1)}, \dots, x^{(n)})$, we can write (see Helgason [1])

$$f(x) = f(x_0) + \sum_{i=1}^n (x^{(i)} - x_0^{(i)})g_i(x),$$

where $g_i(x)$ is a differentiable function, $g_i(x) \approx X_i f(x)$, $g_i(x_0) = (X_i f)_p$, and $x_0 = \varphi_\alpha(p)$. It follows at once from conditions (3.3) and (3.4) that

$$(Xf)_p = \sum_i (X_i f)_p \cdot (Xx^{(i)})(p).$$

This gives (3.10).

(2) The concept of affine (linear) connection is due to Levi-Civita [1]; see also Weyl [1]. Originally the definition of linear connection referred to surfaces in euclidean space, and the

concept was then carried over to Riemannian spaces. Any Riemannian differential metric defines a torsion-free linear connection (see, for example, Favard [1]).

§4

The invariants of a family of probability laws were first considered as invariants of an object of a category by Morse and Sacksteder [1]; see also Sacksteder [1], [2]; monotone invariants were first studied by the author in [7]. Covariants in geometry (with the group of motions) were investigated by Rozenfel'd [1].

(1) Yet another example of a category is the category of all measurable spaces (Ω, S) with all measurable mappings of one into the other. There is a closure operation $(\Omega, S) \rightarrow (\Omega, S^*)$ for the objects of this category. By Lemma 2.1 this operation is a functor.

§5

(1) A transition probability distribution describes not only a decision rule but also, for example, a communication channel with random noise, with "input alphabet" Ω and "output alphabet" \mathcal{C} . Thus, as remarked by Dobrušin, the category of Markov morphisms is at the same time a category of statistical communication channels without memory etc. (see also Csizsár [5]). Note that Theorem 9.1 (on sufficient statistics) has a very graphic interpretation in terms of a two-way communication channel. In fact, it was shown by Wrighton [1] that in a certain sense the problem of statistical inference is a degenerate problem of communication in the presence of noise.

(2) It can be shown that the integrals (5.5) are equal to the double integral

$$\iint_{\Omega' \times \Omega''} P\{d\omega'\} \Pi(\omega'; d\omega'') f(\omega'')$$

understood as a limit of Darboux-Young integral sums

$$\sum_{(B \times A)} P\{B\} \Pi(\omega'_B, A) f(\omega_A)$$

with respect to the filter of finite partitions of the space $\Omega' \times \Omega''$ into measurable rectangles $B \times A$, $B \in S'$, $A \in S''$. To prove the existence of the limit, one must consider the lower and upper Lebesgue integral sums for the partition with

$$A_k = \{\omega'' : kn^{-1} < f(\omega'') < (k+1)n^{-1}\}, \quad k = 0, 1, \dots, n;$$

$$B_j^{(k)} = \{\omega' : jn^{-2} < \Pi(\omega'; A_k) < (j+1)n^{-2}\}, \quad j = 0, 1, \dots, n^2,$$

and then let $n \rightarrow \infty$.

(3) Incidentally, the usual δ -function defines a functional on continuous functions, while any measurable function can be integrated with respect to a δ -measure. Hence they define distinct functionals.

§6

(1) For example, let $\{P_1, Q_1\} \geq \{P_2, Q_2\}$ (Π_{12}) and, conversely, $\{P_1, Q_1\} \leq \{Q_2, P_2\}$ (Π_{21}). It is readily checked that then $\{P_1, Q_1\} \sim \{P_2, Q_2\}$ ($\Pi_{12}, \Pi_{21} \circ \Pi_{12} \circ \Pi_{21}$), and $\{P_1, Q_1\} \sim \{Q_1, P_1\}$ ($\Pi_{12} \circ \Pi_{21}, \Pi_{12} \circ \Pi_{21}$).

§7

(1) All the auxiliary lemmas of §5 were proved for bounded functions. They carry over trivially to nonnegative functions which may take the value $+\infty$.

(2) In Morse and Sacksteder [1] the integral invariants were not introduced, and their arguments were therefore a little more complicated.

(3) Or weakly continuous Markov chains with compact state space (see Bebutov [1]).

§8

(1) In contrast to the term "divergence" (see Kullback [3]), "deviation" reflects the asymmetric way in which the probability measures enter.

(2) A similar assertion was made (without proof) by Rosenblatt-Roth in the text of [1]. A more general statement, using slightly different terminology, was however proved earlier by Csiszár [5].

(3) Of course, for arbitrary pairs the equality $I[Q|P] = I[Q'|P']$ does not imply equivalence. We published (8.6) in [7].

(4) The functional $\chi^2(Q, P)$ was studied by Kagan [1] as a measure of the difference between Q and P (and called the *divergence* of P and Q). The functional (8.10) was studied by Perez [1] and Onicescu [1], under the name of information energy.

(5) In the same way as Shannon's entropy i_{Sh} describes the exponential growth $\exp[Ni_{\text{Sh}}]$ of the number of highly probable messages when the length N of the message is increased. Recall that for a discrete space $\Omega = \{\omega_1, \dots, \omega_m\}$, uniform distribution $Q \leftrightarrow (1/m, \dots, 1/m)$ of outcomes and distribution $P \leftrightarrow (p_1, \dots, p_m)$ we have

$$i_{\text{Sh}}[P] = - \sum_{k=1}^m p_k \ln p_k = \ln m - I[Q|P].$$

(6) The proof of (8.21) goes back to an unpublished paper of Stein [3] and the dissertation of Joshi [1] (in this connection Chernoff [2] and Kullback [3]). The relation (8.22) is due to Chernoff [1], whose proof relies on a rather delicate limit theorem of Cramer [3]. Formula (8.23) is due to N. P. Salihov. Salihov has recently proved that in testing several hypotheses P_1, \dots, P_m the rate of exponential decrease of the maximum probability of error is given by

$$I = \min_{j,k} \inf_{R \in \text{Caph}} \max\{I[P_j|R], I[P_k|R]\}.$$

(7) If the quantile falls on an atom of the distribution of $f_N(\epsilon)$, $P_1^N\{\epsilon: f_N(\epsilon) = k_N\} > 0$, then, as usual, randomization is carried out (see Lehmann [1], Chapter 3), so that the hypothesis P_1 is rejected invariably if $f_N(\epsilon) < k_N$ and with a certain probability q if $f_N(\epsilon) = k_N$, where

$$P_1^N\{\epsilon: f_N(\epsilon) < k_N\} + qP_1^N\{\epsilon: f_N(\epsilon) = k_N\} = b.$$

(8) The convergence $k_N \rightarrow -I[P_0|P_1]$ also occurs if $-I[P_0|P_1] = -\infty$, since the difference $\ln p_1(\omega) - \ln p_0(\omega)$ is P_1 -quasi-integrable. A rigorous proof of this version of the law of large numbers for the likelihood function is contained in the proof of Lemma 27.18.

(9) If $I = +\infty$, the estimation from below is trivial.

(10) The asymptotic behavior described by Mourier and Sakaguchi (see Kullback [3]) is apparently erroneous.

(11) Thus, Rényi [1] considered the invariants $\ln J_u$ and $d \ln J_u / du$, related to the information deviations by (8.25) and (8.26). The integral $J_{1/2}(P, Q)$ was, however, investigated by Bhat-tacharyya [1, 1943].

A concept closely related to information deviation is that of information (contained in one random variable relative to another); see Kolmogorov [5] and Gelfand, Kolmogorov and Jaglom [1]. If F_{12} is the joint distribution law of two random variables $\xi = x(\omega)$ and $\eta = y(\omega)$, and F_1 and F_2 are the marginal laws, then $I = I[F_1 \times F_2 | F_{12}]$.

It is worth mentioning that $2 \arccos J_{1/2}(P, Q)$ is the distance between P and Q in the natural Riemannian metric defined by the Fisher information tensor (see §11 and §12.9).

§9

(1) Here lies the essential difference between the semigroup of Markov transformations and the semigroup of all linear transformations; in the latter almost all transformations are invertible (within the semigroup).

(2) Lemma 9.4 has been known for a long time, and not only to specialists in Markov chains. See, for example, Krein and Rutman [1] or Bebutov [1]. The sets ϵ of outcomes ω described in the

condition are (except for ϵ_0) sets of communicating recurrent states of the Markov chain Π . We have omitted the statement and proof of the fact that every such set has its own stationary distribution.

§10

(1) Let $C_y^+ = \{\omega: f(\omega) > y\} \cap A_j$, $C_y^- = A_j - C_y^+$. Then, since A_j is an atom, either C_y^+ or C_y^- is a \mathbf{Z} -set. By induction, it follows that for any finite sequence $-\infty = y_0 < y_1 < \dots < y_N = +\infty$ only one of the sets $\{\omega: y_{k-1} < f(\omega) < y_k\} \cap A_j$ is not a \mathbf{Z} -set. Consider a countable sequence of numbers, dense in the real line, with $x_0 = -\infty$ and $x_1 = +\infty$. Subject to reordering, every finite subsequence x_0, x_1, \dots, x_N defines a partition of the real line into half-closed intervals. Applying the previous reasoning, we see that only one of the intervals does not correspond to a \mathbf{Z} -set, and with increasing N this interval becomes smaller. As $N \rightarrow \infty$ these half-closed intervals $(t_N', t_N'']$ shrink to a point, which we denote by τ . Denoting $B_N = \{\omega: t_N' < f(\omega) < t_N''\} \cap A_j$, we have $A_j - B_N = C_N \in \mathbf{Z}$. Since \mathbf{Z} is a σ -ideal, it follows that $\bigcup_1^\infty C_N = C \subseteq A_j$ and $C \in \mathbf{Z}$, whence $A_j \neq C$. Thus $\bigcap_1^\infty B_N = B = A_j - C$ is not empty. The function $f(\omega)$ takes the value τ on the set B , and $A_j - \dot{B} = C \in \mathbf{Z}$.

(2) For any (f^1, \dots, f^m) there is a corresponding class of equivalent \mathbf{S} -measurable functions, a suitable representative of which is the function $f(\omega) = f^j$ for $\omega \in A_j$, where A_1, \dots, A_m is a canonical partition of Ω into atoms. The function $\nu\{\cdot\}$ defined on the σ -algebra \mathbf{S} by

$$\nu\{H\} = \int_H f(\omega)\mu\{d\omega\},$$

is a μ -dominated measure, independent of the specific choice of f from its equivalence class. Obviously,

$$\nu_j = \nu\{A_j\} = \int_{A_j} f(\omega)\mu\{d\omega\} = f^j \times \mu_j.$$

Thus any \mathbf{S} -measurable function $f(\cdot)$ determines a linear transformation of the cone of measures, \mathbf{Z} -equivalent functions determining the same transformation.

(3) In a finite-dimensional space all norms are equivalent (see (14.3)). Here, therefore, the expression $\alpha(\tau)$ has an absolute meaning. The fact that this is not so in infinite-dimensional spaces makes it difficult to define smooth families (see §§13 and 14).

(4) A "function-measure" in the terminology of Bogoljubov [1].

(5) Computational formulas of this type are used, for example, in Gel'fand, Frolov and Čencov [1].

(6) The category π CAP of collections Capf with the system of positive centrally-projective homomorphisms was mentioned in Čencov [5] in connection with the study of natural equivalences of families. The connected group of invertible positive centrally-projective transformations is the translation group.

§11

Various authors have repeatedly expressed the opinion that the Fisher information tensor defines a natural Riemannian metric in the manifold of mutually absolutely continuous probability distributions (see, for example, Kullback [2], [3]). However, serious results in this area were obtained only by Kozlov [1], whose work stimulated our own investigations.

(1) Congruent embeddings of the simplexes Capf are described by linear mappings in both natural and canonical coordinates. For this reason the statement of the lemma is true for any index values which vary according to a tensor law under (linear) transformation from one canonical (or natural) coordinate system to another. This argument is applied in Lemma 11.4 to the matrix of second derivatives in canonical coordinates, and it may be applied to fields of differential operators.

Note that since the geometry is almost homogeneous, equivariant scalar fields, (contra)vector fields and tensor contravariant fields vanish identically if they vanish at some point. This is not

the case for covalent tensor fields (since they are "carried backward" by the mappings, not "taken out").

(2) The geometries of the simplex C_{ph} in natural and canonical coordinates are very reminiscent of intrinsic geometries of the first and second kind of the simplex as hypersurface in the enveloping m -dimensional space $\text{Var}(\Omega, S_m)$ equipped (see Norden [1]) with a unique equivariant field of normals $\mathbf{n}(P) = p \leftrightarrow P$.

It is highly important that the equivariant tensor \mathbf{g} coincides for "tangent" measures with the Radon-Nikodým tensor, which converts a measure on S into an S -measurable function, i.e. into an element of the dual space.

(3) The gradients of an invariant function on different objects are represented by covectors of different dimensions. In the case of congruent embedding, however, the scalar product of any embedded tangent vector with the gradient remains unchanged, since it is equal to the derivative in the appropriate direction. This implies that the gradient is equivariant.

§12

I am indebted to E. A. Morozova for advising me to seek the natural geometry of a linear connection. Theorem 12.3 provides the answer to a question put to the author by Ju. V. Linnik. I am indebted to F. I. Karpelevič for sharpening its formulation.

(1) G. and G. G. Vranceanu [1] associate a certain generalized affine connection with every time-continuous Markov process having finitely many states. Their formulation of the problem differs from ours.

(2) This follows from the uniqueness theorem for the solution of system (12.18) or of the system of second-order equations obtained by eliminating the parameters.

(3) A geodesic is a trajectory of a one-parameter subgroup of the translation group. According to the supplement to Lemma 10.2, this group is simply transitive, while the correspondence between the coordinate s and the law P is unique in view of Lemma 10.3.

§13

The monotonicity of the approximating sums for the information deviation was apparently first pointed out by Sanov [1] (see also the proof in Kallianpur [1]). A general approximation theory for functionals of the form (13.18) has been developed by Csiszár [2], [3] and independently by Ghurye [1] and Cencov [12], [13]. Special subclasses of such functionals have also been investigated by other authors; see Martin and Oheix [1] and their subsequent publications.

(1) Translations, as projective transformations of a simplex, are also fractional-linear. Passage to a conditional distribution may be described as a limit of translations when the canonical parameter becomes infinite (see §21.8).

(2) For the definition of projective and inductive limits, see Palamodov [1] and also Scheffer [1].

(3) Recall that in view of the conventions (13.23) the second integral is a Lebesgue integral only when $Q \gg P$. But if

$$P\{\omega: (dQ/dP)(\omega) = 0\} > 0,$$

then the Q -integral of an infinite function over this set is put equal to $+\infty$.

§14

(1) Since the function f may not be one-to-one, the surface defined by its image may be self-intersecting. The tangent plane is a local concept. If $f(x_1) = f(x_2) = y_0$, then there are two tangents at the same point y_0 of the space, corresponding to different points (x_1, y_0) and (x_2, y_0) of the surface.

It should be noted that condition (14.5) is quite restrictive. It is certainly not satisfied when $\dim X > \dim Y$ (i.e. the dimension of a smooth surface cannot exceed the dimension of the enveloping space).

(2) However, this situation is typical even in linear topological spaces (see Averbuh and Smoljanov [1]). One then has to be content with functionals which are differentiable with respect to the subspace of increments of a finite "norm".

Note that the weakest $\mathcal{L}^1(P)$ -metric (which coincides with variation for dominated measures) is too weak for us (see above, subsection 3), while the strongest $\mathcal{L}^\infty(P)$ -metric defined by the essential supremum of the Radon-Nikodým derivative is far too strong—even the family of normal laws is not continuous in it.

(3) A definition similar to $\mathcal{L}^2(P)$ -differentiability, for families of one real parameter, was given by Schmetterer [1].

(4) This follows from well-known theorems of classical analysis. We shall not present the proofs, since the corresponding proposition is easier to state and to prove for continuously differentiable surfaces (see Lemma 16.3).

§15

(1) It is natural to confine the study of single-fold differentiability to differentiability in the field of $\mathcal{L}^2(P)$ -metrics. However, n -fold differentiability should be considered in the stronger $\mathcal{L}^\infty(P)$ -metrics, or, more precisely, in a hierarchy of metrics (see §26.5); otherwise such “analytic functions” as $I[Q|P]$ turn out to be nondifferentiable.

(2) That is, we obtain an estimate for the squared norm in the quotient space of $\mathcal{L}^2(P)$ by a line of constants (cf. Schmetterer [1] and Čencov [9]). The corresponding induced metric was introduced by Gel'fand (see Hille [1]).

(3) (15.12) and (15.14) are known as the (one-dimensional) information inequalities. Associated with their rigorous proofs are such names as Cramér [1], [2], C. R. Rao [1], Darmois [1], Fréchet [1], as well as Dugué [1] and many others (see van der Waerden [1]).

(4) See Cramér [1], and C. R. Rao [2].

(5) For example, P''' etc. (see Bhattacharyya [2], Bol'šev [1] or Seth [1]). The more the law P admits unbiased estimators $f(\omega)$ for zero, $\mathbb{M}_P f(\omega) = 0$, the larger the lower bound for the variance of the unbiased estimator (see Kagan [4] and C. R. Rao [3]).

(6) The definition of an efficient estimator goes back to Fisher [1], [2]. Regarding estimators for which the risk coincides with the bound (15.23), see De Groot and M. Rao [1].

(7) This theorem was considered with no clearly formulated smoothness conditions by Bhattacharyya [2]. A logical error has slipped into the proof given by Kullback [3, Chapter 3]. Fraser [1] gives the proof for one parameter only. In contradistinction to the authors just listed, we have not presumed the existence of smooth densities, and we allow the appearance of formal estimators. A stronger version of this theorem is Theorem 23.6.

§16

(1) But convergence (16.2) in the \mathcal{L}^2 -metric follows from the truth of condition (16.1) in the \mathcal{L}^4 -metric (cf. §26.5 and footnote 4 to §26).

(2) Or as the limit of Riemann sums in the local \mathcal{L}^u -metric.

(3) A related fact was actually used in the proof of Lemma 16.1.

(4) A fortiori, according to Corollary 1 to Lemma 16.1, differentiable with u th moment in the sense of Definition 14.3.

(5) Continuity of $p'_i(\omega; t)$ is not required here. Neither is measurability of $\tau(\omega)$, since the measurability of

$$(p'_{(e)}(\omega; \tau(\omega))) = |t - \theta|^{-1}[p(\omega; t) - p(\omega; \theta)]$$

follows from the definition.

(6) Integral corollaries from the information inequality for one-parameter families were first obtained by Blyth [1], Girshick and Savage [1], and Hodges and Lehmann [1]; see also Karlin [1]. The examples of Stein [1], [2] show that these corollaries do not carry over to $n > 3$ parameters. This is why Kiefer [1], in his fundamental review of the theory of optimal multivariate estimators, expressed doubt as to the applicability of the information inequality as a tool for construction of optimal estimators. The theorem stated here was proved by the present author in [9].

Integral corollaries for the variance of an asymptotically normal law were obtained by LeCam [1] and Schmetterer [1], in the case $n = 1$.

§17

(1) This definition was apparently first given by Minty [1]. For further research on the subject, see Rockafellar [1].

(2) Any one-to-one continuous mapping of the real line into itself, with continuous inverse, is automatically monotone and takes monotone functions to monotone functions. This is not the case for multidimensional spaces.

§18

Exponent families are perhaps the most important and well-studied class of families. Systematic investigation of their general theory began with Koopman [1, 1936], who characterized them as families having finitely many sufficient statistics (see also Dynkin [1, 1951]). They appeared independently in problems of statistical physics (see Khintchine [1]).

An account of the general theory of exponent families was given in Kullback [3, 1959] and Lehmann [1, 1960]. General analytical approaches to the theory of parameter estimation and hypothesis discrimination for these families were proposed in monographs by Linnik [2], [3]. A number of fundamental problems of the theory were also touched upon in Blackwell and Girschick [1] and Robbins [1]. Our own research has dealt mainly with the "geometrical" aspect of the theory (see Cencov [4], [6]).

(1) The component of the identity of the group of centrally-projective transformations of $\text{Caph}(\Omega, \mathbf{S}_m, \mathbb{D})$ into itself, see §10.

(2) More precisely, the statistics $q_1(\omega), \dots, q_n(\omega)$ and $q_0(\omega) = 1(\omega)$ are linearly independent.

(3) Though this density may determine a random functional on smooth functions, i.e. define a generalized random variable in Gel'fand's sense [2].

(4) Though the dimension of the family of generalized random variables is n .

(5) These limits form the Vorob'ev-Faddeev fine boundary of the simplex $\text{Caph}(\Omega, \mathbf{S}_m, \mathbb{D})$. Under these conditions, to each strictly dominated law corresponds a whole manifold of limits. The point is that the simplex $\text{Caph}(\Omega, \mathbf{S}_m, \mathbb{D})$, as a homogeneous manifold of zero curvature, has no natural boundary (see Gel'fand and Graev [1]). The boundary, obtained by Vorob'ev and Faddeev [1] from other considerations, may be obtained by the geometrical methods of Karpelevič [1].

(6) This means that the correspondence $s \rightarrow P_s$ determined by (18.1) and (18.2) defines a chart of the family.

(7) Recall that, in any semioordered space, first-order homogeneous functions are defined in an invariant manner (see Kantorovič, Vulih and Pinsker [1]).

(8) The desirability of presenting a direct proof for this theorem was pointed out to the author by Ju. V. Linnik.

The theorem may also be derived from the general results of §7.

(9) This equation is transformed to a form more convenient for our purposes (cf. Bernstein [1]). Regarding the equation $p'(x) = sg'(x)p(x)$, see also Mathai [1].

§19

(1) The canonical affine parameter of a geodesic (see Favard [1]), or the canonical variable (see Khintchine [1]). This terminology is more "canonical" than that adopted by Linnik [2].

(2) The transformation from canonical to natural parameter in statistical physics is associated with the introduction of the notion of temperature (Khintchine [1]). A natural parameter was considered as an ancillary tool by Bhattacharyya [2] and Kullback [3]. The term "natural parameter" itself is due to the author [6]. The main result of the section were published in the indicated papers.

(3) See, for example, Dieudonné [1].

(4) Since for $s^{(2)} > 0$ the integral of the positive function $\exp[s^{(2)}x^2 + s^{(1)}x]$ from $-\infty$ to $+\infty$ is identically equal to $+\infty$.

(5) Unique with probability one, since both $q(\omega)$ and $p(\omega; s)$ are defined up to values on any Z -set.

§20

(1) Formula (20.8) is due to Huzurbazar [1]. (20.9) was proved by Kullback [3], who also provides references to the work of other authors. In the multidimensional case the Legendre conjugacy of the parameters, and, accordingly, (20.10), was first explicitly considered in Cencov [6]. The Young inequality (20.12) was proved much earlier by Sanov, by minimization of a suitable expression.

(2) Since $I[P|R] = +\infty$ when P does not dominate R , formulas (20.15) and (20.16) permit a description of the geodesic hull spanned by mutually absolutely continuous laws P_i , as the family of probability distributions $\{P_A\}$ where each $P_A = R$ minimizes a sum $\sum_i \alpha_i I[P_i|R]$. Formula (20.15) recalls the expression for the moment of inertia about an arbitrary point in terms of the moment of inertia about the center of gravity. (And when the "masses" of points vary, the center of gravity runs over the entire linear span.)

(3) For nonconstructive families, all that one can assert is weak equivalence in the sense of Morse and Sacksteder (see above, Definition 6.4).

§21

(1) This simple example shows that the natural chart of an infinite-dimensional exponent family has a more intricate structure than one might expect at first glance. Example 6 (below) is due to the author [12].

(2) See also Basu [1], Doss [1] and Kale [1].

(3) This boundary is the most economical. The finer and therefore more massive boundary of Vorob'ev and Faddeev [1] converts the simplex $\text{Caph}(\Omega, S_m, \mathbb{D})$ into a compact set on which the conditional probabilities of each event relative to any nonempty hypothesis are everywhere continuous.

§22

(1) The theorem may be slightly generalized by allowing the law to be a point of the boundary at infinity.

(2) The fact that the maximum likelihood method coincides with the method of minimum information deviation has been noticed by many authors, among them Kullback and Cencov; see also Kriz and Talacko [1], and Hartigan [1].

§23

(1) It is convex by Lemma 19.1. Thus, for exponent families γ with domain G^γ satisfying (23.4), the domain of the natural parameter is convex.

(2) Definition 27.4 of a smooth family is a substitute, since it demands that the densities, not the measures themselves, be smooth.

(3) By Lemma 28.5, it is smooth in the sense of Definition 27.5. The definitions themselves were sought so that geodesic (exponent) families satisfy them (and for differentiable families, by analogy with finite-dimensional manifolds Caph , the information inequality be automatically valid).

(4) The inequality between the first and last members in (23.12) follows at once from (26.20). The direct proof is also quite easy.

(5) If the functional $\int \varphi[(dQ/dP)(\omega)]P\{d\omega\}$, where φ is a convex function, is not continuous on $\text{Caph}(\Omega, S, Z)$ in variation $|\mu|$, then according to Csiszár [2], [4] it does not define a uniform topology (evidently, a topology can be defined only by a hierarchy of such functionals; see §26.5). The terms used here are therefore conditional.

§24

(1) By §27.5, estimators of the form P_α , where α is a parameter estimator, for an exponent family \mathcal{U} with Gaussian loss function, form a complete class. In regard to this problem our formulation is close to that of De Groot and M. M. Rao [1] and M. M. Rao [1]; see also De Groot [1].

(2) Lemma 24.1 and its corollary are essentially rephrased versions of a well-known inequality of Blackwell [1], Kolmogorov [4] and C. R. Rao [1]; see also M. M Rao [1].

(3) The proof given in our paper [9] is based not on Theorem 16.1 but on a simpler version, dealing specifically with the family \mathcal{N} .

(4) The author arrived at the "nonsymmetric Pythagorean geometry" of §§22 and 23 in [10] when he tried to generalize the geometry of the Gaussian method of least squares (see Gauss [1] and Linnik [1]) to arbitrary geodesic (exponent) families. The actual statement of Lemma 24.4, in some form or another, has frequently been used by statisticians to improve decision rules (see Thompson [1]).

(5) We put aside the very important and interesting (but much more complicated) Bayesian formulation of Robbins [1], [2] (see also Neyman [2]).

(6) Note that if $L(P, Q)$ is monotone and admits both expansions (24.24) and (26.10), then the constants $c_0[L]$ in both formulas are equal, but without the additional assumptions of uniform differentiability one cannot assert that the constants $c[L]$ are equal.

§25

(1) See also Frolov and Čencov [1], and the following publications: Van Ryzin [1], Schwartz [1], Kronmal and Tarter [1], Bosq [1], [2], [3], Sesan et al. [1], and Sizova [1]. Estimators of type (25.1) have been discussed earlier by Rosenblatt [1]; see also Schuster [1].

(2) It is easy to see that $\sum_j \mathbf{M}[\alpha_j(\xi)]^2$ is independent of the choice of the orthonormal basis $\varphi_1(x), \dots, \varphi_n(x)$ in E_n , since for each x the sum $\sum_k [\varphi_k(x)]^2$ is invariant under orthonormal changes of basis.

(3) The integrand in (25.31) is also independent of the choice of basis (cf. footnote 2, above).

(4) The norm of the deviation of the histogram from the graph of the density, in the metric of the space C , has been estimated by Smirnov [1], [2] (see also Tumanjan [1]). If the grouping is optimal, it decreases almost like $N^{-1/3}$ (up to a logarithmic factor).

(5) See also Rosenblatt [1].

(6) These constraints may be relaxed somewhat by making them closer to quasi-homogeneity conditions (see Definition 28.1).

(7) We used this argument in [3]. It leads to a less precise accuracy bound than the inequality of Lemma 16.6, and is applicable only when the quality of the estimator is measured by maximum risk. On the other hand, it works under weaker constraints, and this is essential in the case of infinite-dimensional families.

(8) We are assuming here that all the moments $\int \varphi_k(x) p^*(x) \mu\{dx\}$ of the initial estimator p^* are measurable, which implies that p_n^* is measurable and hence (by Lemma 17.14) that π_n^* is measurable.

(9) Under very weak restrictions (cf. Tumanjan [1]) the squared norm (25.10) of the deviation of the estimator is asymptotically normal (see Bosq [2], also [1]). Hence one can obtain sharper results for asymptotic confidence limits than we obtained in Čencov [3], [12].

(10) The question of statistical estimation of a smooth curve is very timely (see Tukey [2] and Whittle [1]).

§26

(1) Random measures have been introduced in many publications as random functionals (i.e. random generalized functions); see Prohorov [1] and Gel'fand [2]. Discrete empirical random measures (25.9) have also been considered. Both concepts admit an effective definition.

(2) At first sight it might seem natural, following Laplace [1], to adopt as loss function some invariant metric or a substitute thereof (see the review of Adhikari and Joshi [1], where more than ten such functions are listed). However, as Gauss noted [1], the theory is much simplified if one takes a quadratic loss function (see LeCam [1]). This is why our loss function is a nonsymmetric analog $2I[P|P^*]$ of the squared euclidean distance—an analytic, "approximately quadratic" functional for which the nonsymmetric Pythagorean theorem and various other geometrical theorems are valid (see subsection 5).

An interesting unsolved problem is to describe all natural matrix loss functions (see Kagan [4]).

(3) This follows from a result of Csiszár [2], [4] (see footnote 5 to §23).

(4) For example,

$$\begin{aligned} \|R - P\|_{(P,2)} &< \|R - Q\|_{(P,2)} + \|Q - P\|_{(P,2)} \\ &< [\|Q\|_{(P,3)}]^{3/4} \|R - Q\|_{(Q,4)} + \|Q - P\|_{(P,2)}. \end{aligned}$$

(5) Moreover, $P\{A'\}I[P'R'] \succ I[P|R]$ as $A' \succ \Omega$.

§27

(1) This is a way of using an argument of Blackwell, Kolmogorov and Rao (see footnote 2 to §24).

(2) Under our assumptions, we can no longer assert that the majorant

$$g_1(\omega) \{ h(\omega) + 3[h(\omega)]^2 + [h(\omega)]^3 \}$$

for the third derivatives is integrable.

(3) The statement of the corollary follows from the theory of §16, since the smooth families are contained in the class of continuously differentiable families.

(4) This class contains all compact exponent families corresponding to a compact subdomain of the canonical parameter (see Lemma 28.5).

(5) If the dimension n is large, these estimators become nontrivial only for large N . If one assumes that the higher moments of the majorant are bounded, the estimators become efficient at values as low as $n \sim N^{2+\delta}$.

(6) Another possible procedure for localizing the root is used in the proof of Theorem 27.3.

(7) Unless restrictions are imposed on the compactness of the family and of the maximization domain, the optimality and even consistency of the maximum likelihood estimator become problematic. Linnik and Mitrofanova [1] give a very delicate proof that the maximum likelihood estimator is efficient for a certain subclass of the exponent families.

§28

(1) A brief account of the contents of this section was given in our note [8] (see also [12]).

(2) Of course, only the rate of decrease of the quantity $n \asymp \Gamma(N)$ is determined; n itself may be fixed arbitrarily within certain bounds.

(3) Recall that the indices for the information matrix in canonical coordinates s^j are written as subscripts: v_{jk} . The notation w^{jk} is reserved for the information matrix in natural coordinates t_j .

§29

The definition of a random probability measure (on the unit interval or on the whole real line) by its random distribution function goes back to Kolmogorov [1]. Dubins and Freedman [1] systematically considered the definition of a random distribution function $F(t)$ as a monotone stochastic process on the real line, i.e. in terms of probabilities of quasi-intervals:

$$\mathcal{P} \left\{ a'_i < F(t_i) < a''_i, i = 1, \dots, n \right\}.$$

These probabilities define a Baire distribution and, according to Kakutani [1] (see Halmos [1]), a unique regular Borel distribution, which is easily shown to be concentrated on monotone functions $F(t)$. However, as follows from a well-known result of the author [1], this distribution is generally not concentrated on distribution functions, since the set of all distribution functions is neither a Borel set nor absolutely measurable (although the set of continuous distribution functions has these properties). We therefore have to prove the existence of a non-Borel extension \mathcal{P} , concentrated on the distribution functions, just as in the theory of Markov processes one has to prove the existence of a canonical modification of the process (see Itô [1], and also Doob [2]).

(1) The class of $\mathbf{B}_0(\mathbf{S})$ -sets is fairly small. If the algebra \mathbf{S} is uncountable, it does not even contain "singleton subfamilies". Since every family consisting of a single set function is closed in the product topology, they are all $\mathbf{B}(\mathbf{S})$ -sets.

(2) If S has countably many generators, there is a substantial difference between the algebras $\mathbf{B}_0(S)$ and $\mathbf{B}(S)$. For example, all "singleton" families are measurable in $\mathbf{B}_0(S)$, since every σ -additive measure is completely determined by its values on countably many generators, while an arbitrary set function is determined by its values on the whole algebra S .

(3) Each of conditions 1°–3° of system (2.5) (for fixed sets H, H_1 and H_2) describes a closed $\mathbf{B}_0(S)$ -subset of the space X^S . Hence axioms 1°–3°, which define the collection of all normalized finitely-additive set functions, describe a closed $\mathbf{B}(S)$ -set as the intersection of the above-mentioned closed $\mathbf{B}_0(S)$ -sets. The descriptive nature of the set of all normalized countably-additive set functions (i.e. probability measures) is as yet unknown, notwithstanding several interesting studies (see Freedman [1]).

By virtue of the well-known continuity property of any finite measure relative to a dominating measure, and the σ -additivity of any finitely-additive measure which is continuous relative to a σ -additive measure (cf. the proof of Lemma 14.1), the collection $\text{Capd}(\Omega, S, Z)$ is a family of the type $F_{\sigma\delta}$ in X^S , and hence $\mathbf{B}(S)$ -measurable, although it is not $\mathbf{B}_0(S)$ -measurable.

(4) In view of conditions 1°–3°, condition 4° (convergence with probability one) may be replaced by convergence in probability:

For any $\varepsilon > 0$ and every fixed sequence $H_k \searrow \emptyset, H_k \in S$,

$$\mathbb{P} \left\{ \Phi \{ H_k \} > \varepsilon \right\} \rightarrow 0.$$

(5) Note that Lebesgue collections $\text{Capd}(\Omega, S, Z)$ are not only $\mathbf{B}(S)$ -measurable in X^S (see footnote 3, above) but also \mathbf{K} -measurable, also corresponding to an $F_{\sigma\delta}$ -set in the sector W .

(6) An analogous statement is readily proved for other functionals (13.18) satisfying the conditions of Theorem 13.1.

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