

## ON ASYMPTOTICALLY OPTIMAL TESTS

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Sequences of tests with error  $\exp(-nA)$  of the first type are investigated. It is shown that the error of the second type of such a sequence of tests is bounded by  $\exp(-nB)$  where  $B$  is determined by the Kullback-Leibler information distance of the hypotheses tested. The information distance between the empirical measure and the null-hypothesis on a finite partition of the sample space is proposed to use as a test statistic. A sufficient condition is given which ensures that this test has error of the second type about  $\exp(-nB)$  with the best possible  $B$ . The exact Bahadur slope of the proposed statistic is investigated.

**1. Introduction.** Let  $S = (X, \mathcal{A}, \mathcal{P})$  be a statistical space, and  $S_n = (X^{(n)}, \mathcal{A}^{(n)}, \mathcal{P}^{(n)})$  its  $n$ th power. Suppose that  $\mathcal{P}$  consists of all probability measures on  $(X, \mathcal{A})$ , and it is desired to test the null-hypothesis  $H_0: P \in \mathcal{P}_0$  on  $S_n$ , where  $\mathcal{P}_0$  is an arbitrary nonempty subset of  $\mathcal{P}$ . Let  $\varphi_n = \varphi_n(x_1, \dots, x_n)$  denote a randomized test function ( $\varphi_n$  is the probability of the rejection of  $H_0$ ), and let

$$(1.1) \quad \alpha_n(P) = \int_{X^{(n)}} \varphi_n dP^{(n)}, \quad \beta_n(P) = 1 - \alpha_n(P), \quad \alpha_n = \sup_{P \in \mathcal{P}_0} \alpha_n(P).$$

Several different ways have been proposed for defining the asymptotic optimality of a sequence of tests  $\varphi_n$ . In some of them a sequence of alternatives  $Q_n$  is chosen in such a way that the probabilities  $\beta_n(Q_n)$  are bounded away from 0 and 1. The speed of  $Q_n$  approaching  $\mathcal{P}_0$  is measured somehow and this speed is taken as a criterion of optimality. Another possibility is to fix an alternative  $Q$  and consider the rate of  $\beta_n(Q)$  tending to 0. The first method was investigated by Neyman (1937), the second by Bahadur (1960), Hoeffding (1965) and Brown (1971). The following definition is motivated by their results.

**DEFINITION 1.** A sequence of tests is of rate  $A$ ,  $0 \leq A \leq \infty$ , if

$$(1.2) \quad \limsup_{n \rightarrow \infty} \alpha_n < 1, \quad \text{in case } A = 0,$$

$$(1.3) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \log \alpha_n \leq -A, \quad \text{in case } A > 0.$$

Let  $\Phi_A$  be the set of sequences of tests of rate  $A$ , and let

$$(1.4) \quad B(A, Q, \mathcal{P}_0) = -\inf_{\varphi_n \in \Phi_A} \liminf_{n \rightarrow \infty} \frac{1}{n} \log \beta_n(Q).$$

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A sequence of tests  $\varphi_n \in \Phi_A$  is called exponential rate optimal (ERO) at  $Q$ , if

$$(1.5) \quad \lim_{n \rightarrow \infty} \frac{1}{n} \log \beta_n(Q) = -B(A, Q, \mathcal{P}_0).$$

A sequence of statistics  $T_n$  is ERO at  $Q$ , if the sequence of tests

$$(1.6) \quad \begin{aligned} \varphi_n &= 0 && \text{if } T_n < A \\ &= 1 && \text{if } T_n \geq A \end{aligned}$$

is of rate  $A$  and ERO at  $Q$  for any continuity point  $A > 0$  of  $B(A, Q, \mathcal{P}_0)$ .  $\square$

Although the above definition has a local character, our basic aim is to give statistics which depend only on  $\mathcal{P}_0$  and are ERO at as many alternatives  $Q$  as possible. Brown (1971) gives sufficient conditions ensuring that the likelihood ratio is an ERO statistic. His conditions are of topological character, and are not fulfilled if  $\mathcal{P}$  consists of all probability measures on  $(X, \mathcal{A})$ -except for the multinomial case, investigated by Hoeffding (1965). We shall extend Hoeffding's result using finite partitions of  $(X, \mathcal{A})$ .

DEFINITION 2. The empirical measure  $p_n$  is defined by

$$(1.7) \quad p_n = p_n(Y; x_1, \dots, x_n) = \frac{1}{n} \sum_{i=1}^n X_Y(x_i),$$

where  $X_Y(x) = 1$ , if  $x \in Y$ , and  $X_Y(x) = 0$  otherwise. For measuring the distance between  $p_n$  and  $\mathcal{P}_0$  we shall use a partition  $\mathcal{B}_n$  of  $(X, \mathcal{A})$ . A sub- $\sigma$ -algebra  $\mathcal{B}_n$  is called a partition of  $(X, \mathcal{A})$  if it is generated by disjoint sets  $Y_1, \dots, Y_m \in \mathcal{A}$  whose union is the whole space  $X$ . Throughout this paper, the sequence  $m = m(n)$  denotes the number of atoms in the partition  $\mathcal{B}_n$ . The Kullback-Leibler information number for  $P, Q \in \mathcal{P}$  on  $\mathcal{B}_n$  is

$$(1.8) \quad K_n(P, Q) = \sum_{i=1}^m P(Y_i) \log \frac{P(Y_i)}{Q(Y_i)},$$

where  $0/0 = 0, 0 \log 0 = 0$ ; and for any  $\mathcal{Q} \subset \mathcal{P}$

$$(1.9) \quad K_n(P, \mathcal{Q}) = \inf_{Q \in \mathcal{Q}} K_n(P, Q); \quad K_n(\mathcal{Q}, P) = \inf_{Q \in \mathcal{Q}} K_n(Q, P).$$

Let  $L_n$  denote the statistic

$$(1.10) \quad L_n = K_n(p_n, \mathcal{P}_0). \quad \square$$

DEFINITION 3. Given a family  $\mathcal{B} \subset \mathcal{A}$  of sets in  $X$ , we say that the sequence  $P_n \in \mathcal{P}$  converges to the set function  $\nu$  on  $\mathcal{B}$  if  $\lim_{n \rightarrow \infty} P_n(Y) = \nu(Y)$  for any  $Y \in \mathcal{B}$ . A monotone sequence of partitions  $\mathcal{B}_n$  is said to be adequate (wrt  $\mathcal{P}_0$ ), if  $\mathcal{P}_0$  is closed under convergence on  $\mathcal{B} = \bigcup_{n=1}^{\infty} \mathcal{B}_n$ , and  $\mathcal{B}$  generates  $\mathcal{A}$ .  $\square$

In Section 2 we investigate the function  $B(A, Q, \mathcal{P}_0)$ . Section 3 contains our main result (cf. Corollary 2): if the sequence  $\mathcal{B}_n$  is adequate, and the sequence  $m(n)$  satisfies

$$(1.11) \quad \lim_{n \rightarrow \infty} m(n) \frac{\log n}{n} = 0,$$

then the sequence of statistics  $L_n$  is ERO. (It is easy to see that any monotone  $\mathcal{B}_n$  generating  $\mathcal{A}$  is adequate, if  $\mathcal{P}_0$  is dominated and the set of densities is compact in  $L_1$ -norm. This condition fulfills trivially if  $\mathcal{P}_0$  is simple, or at least finite.) The relation of ERO statistics and the exact slope introduced by Bahadur is investigated in Section 4. Some concluding remarks, examples and problems are collected in Section 5.

This work was initiated by a survey paper of Bahadur (1971). We shall mostly use the notation introduced there, and refer to known results through that paper.

**2. The best exponent.** The investigation of the function  $B(A, Q, \mathcal{P}_0)$  (cf. Definition 1) will be based on a lemma of Stein (cf. Lemma 6.1 in Bahadur (1971)). This lemma states that

$$(2.1) \quad B(0, Q, \{P\}) = K(P, Q),$$

for any  $P, Q \in \mathcal{P}$ , where  $K$  is defined by

$$(2.2) \quad K(P, Q) = \int_x \log \frac{dP}{dQ} dP \quad \text{if } P \ll Q, \\ = \infty \quad \text{otherwise.}$$

For any  $\mathcal{Q} \subset \mathcal{P}$  let

$$(2.3) \quad K(P, \mathcal{Q}) = \inf_{Q \in \mathcal{Q}} K(P, Q); \quad K(\mathcal{Q}, P) = \inf_{Q \in \mathcal{Q}} K(Q, P).$$

For any  $0 < A \leq \infty$ , let

$$(2.4) \quad \mathcal{P}_A = \{Q: K(Q, \mathcal{P}_0) < A\},$$

$$(2.5) \quad \mathcal{P}_{A,n} = \{Q: K_n(Q, \mathcal{P}_0) < A\}.$$

**THEOREM 1.** For any  $\mathcal{P}_0 \subset \mathcal{P}$ ,  $Q \in \mathcal{P}$ ,  $0 \leq A \leq \infty$

$$(2.6) \quad B(A, Q, \mathcal{P}_0) \leq K(\mathcal{P}_A, Q).$$

**PROOF.** First we prove that for any sequence  $\varphi_n \in \Phi_A$ ,  $0 < A \leq \infty$

$$(2.7) \quad \lim_{n \rightarrow \infty} \alpha_n(R) = 0$$

holds true for any  $R \in \mathcal{P}_A$ . If we had a  $R \in \mathcal{P}_A$  such that

$$\limsup_{n \rightarrow \infty} \alpha_n(R) > 0, \quad \text{i.e.,} \quad \lim_{k \rightarrow \infty} \beta_{n_k}(R) < 1$$

for some subsequence  $n_k$ , then the sequence

$$\tilde{\varphi}_n = 1 - \varphi_{n_k} \quad \text{for } n_k \leq n < n_{k+1}$$

would be of rate 0 for testing the simple hypothesis that the actual measure is  $R$ . Hence Stein's lemma would imply that

$$(2.8) \quad \liminf_{k \rightarrow \infty} \frac{1}{n_k} \log \alpha_{n_k}(P_0) = \liminf_{n \rightarrow \infty} \frac{1}{n} \log \tilde{\beta}_n(P_0) \\ \geq -B(0, P_0, \{R\}) = -K(R, P_0)$$

for any  $P_0 \in \mathcal{P}$ . ( $\beta_n$  denotes the error of second type of  $\phi_n$ .) But there is a  $P_0 \in \mathcal{P}_0$  such that  $K(R, P_0) < A$  since  $R \in \mathcal{P}_A$ ; consequently (2.8) contradicts (1.3).

The obtained (2.7) implies that given  $R \in \mathcal{P}_A$ , any sequence  $\varphi_n \in \Phi_A$  is of rate 0 for testing the simple hypothesis that the actual measure is  $R$ . Hence  $B(A, Q, \mathcal{P}_0) \leq B(0, Q, \{R\}) = K(R, Q)$  for any  $R \in \mathcal{P}_A$ , which proves (2.6).  $\square$

**THEOREM 2.** *The sequence of tests*

$$(2.9) \quad \begin{aligned} \varphi_n &= 0 && \text{if } L_n < A \\ &= 1 && \text{if } L_n \geq A \end{aligned}$$

is of rate  $A$  for any  $0 < A \leq \infty$ , where  $L_n$  is defined by Definition 2, and  $\mathcal{B}_n$  is any sequence of partitions for which  $m(n)$  satisfies (1.11). For this sequence of tests

$$(2.10) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \log \beta_n(Q) \leq -\liminf_{n \rightarrow \infty} K_n(\mathcal{P}_{A,n}, Q)$$

holds true for any  $Q \in \mathcal{P}$ , where  $\mathcal{P}_{A,n}$  is defined by (2.5).

**COROLLARY 1.**  $B(A, Q, \mathcal{P}_0) \geq \liminf_{n \rightarrow \infty} K_n(\mathcal{P}_{A,n}, Q)$ .

**PROOF.** It is easy to see (cf. (5.33) in Bahadur (1971)) that

$$(2.11) \quad P_0(K_n(p_n, \mathcal{P}_0) \geq A) \leq (n + 1)^m e^{-nA}$$

holds true for any  $P_0 \in \mathcal{P}_0$ ; hence (1.11) implies  $\varphi_n \in \Phi_A$ . Let the numbers  $K_n(\mathcal{P}_{A,n}, Q)$  be denoted by  $B_n$ . The set  $\mathcal{P}_{A,n}$  of measures is contained in the set  $\{R : K_n(R, Q) \geq B_n\}$ , hence

$$Q(K_n(p_n, \mathcal{P}_0) < A) \leq Q(K_n(p_n, Q) \geq B_n).$$

On applying the same argument as in (2.11) we get

$$Q(K_n(p_n, Q) \geq B_n) \leq (n + 1)^m e^{-nB_n};$$

consequently (1.11) implies (2.10), too.  $\square$

**3. Adequate sequences of partitions.**

**THEOREM 3.** *If the sequence of partitions  $\mathcal{B}_n$  is adequate (cf. Definition 3), then*

$$(3.1) \quad \lim_{n \rightarrow \infty} K_n(\mathcal{P}_{a,n}, Q) \geq K(\mathcal{P}_A, Q)$$

holds true for any  $0 \leq a < A \leq \infty$ ,  $Q \in \mathcal{P}$ ,  $\mathcal{P}_0 \subset \mathcal{P}$  (cf. Definition 2, (2.4) and (2.5)).

**COROLLARY 2.** *If  $\mathcal{B}_n$  is adequate and the sequence  $m(n)$  satisfies (1.11), then the sequence of statistics  $L_n$  is ERO (cf. Definition 1).*

**COROLLARY 3.** *If  $\mathcal{P}_0$  is such that there exists an adequate sequence of partitions, then for any  $0 \leq a < A \leq \infty$ ,  $Q \in \mathcal{P}$ ,  $\mathcal{P}_0 \subset \mathcal{P}$*

$$(3.2) \quad K(\mathcal{P}_A, Q) \leq B(a, Q, \mathcal{P}_0) \leq K(\mathcal{P}_a, Q).$$

PROOF. Let the numbers  $K_n(\mathcal{P}_{a,n}, Q)$  be denoted by  $B_n$ . The sequence  $B_n$  is monotone increasing, for  $\mathcal{B}_n$  is monotone. Hence the limit on the left-hand side of (3.1) exists. (It is also easy to see that  $B_n \leq K(\mathcal{P}_a, Q)$ .) It is enough to prove that for any  $B$ , such that  $\lim_{n \rightarrow \infty} B_n < B < \infty$ , there is an  $R \in \mathcal{P}_A$  such that  $K(R, Q) \leq B$ .

There are  $P_n \in \mathcal{P}_0, R_n \in \mathcal{P}_a$  such that  $K_n(R_n, Q) < B, K_n(R_n, P_n) < a$ . For any  $j$  there is a convergent subsequence of  $P_n, R_n$  on  $\mathcal{B}_j$ , because  $\mathcal{B}_j$  has finitely many elements. By using the diagonal method we can choose convergent subsequences of  $P_n, R_n$  on  $\mathcal{B} = \bigcup_{j=1}^{\infty} \mathcal{B}_j$ , too. Let  $P_{n_k}, R_{n_k}$  be such subsequences; then there is a  $P_0 \in \mathcal{P}_0$  such that  $P_{n_k}$  tends to  $P_0$  on  $\mathcal{B}$  for  $\mathcal{B}_n$  is adequate. Let the limit of  $R_{n_k}$  be denoted by  $\nu$ :

$$(3.3) \quad \nu(Y) = \lim_{k \rightarrow \infty} R_{n_k}(Y) \quad \text{for } Y \in \mathcal{B}.$$

Now we prove that there is a probability measure  $R$  on  $(X, \mathcal{A})$  such that its restriction to  $(X, \mathcal{B})$  is  $\nu$ .

Let  $Y_1, \dots, Y_m$  be the atoms of  $\mathcal{B}_n$ . It is easy to see that the sequence

$$(3.4) \quad r_n(x) = \frac{\nu(Y_i)}{Q(Y_i)}$$

if  $x \in Y_i$  is a martingale with respect to  $\mathcal{B}_n$  on  $(X, \mathcal{A}, Q)$ . The function  $f(u, v) = u \log(u/v); f(0, v) = 0; f(u, 0) = \infty; f(0, 0) = 0$  ( $0 < u, v < 1$ ) is lower semi-continuous, i.e., for any convergent sequences  $0 \leq u_i \leq 1; 0 \leq v_i \leq 1$

$$(3.5) \quad f(\lim_{i \rightarrow \infty} u_i, \lim_{i \rightarrow \infty} v_i) \leq \liminf_{i \rightarrow \infty} f(u_i, v_i),$$

consequently  $\sum_{i=1}^m \nu(Y_i) \log(\nu(Y_i)/Q(Y_i)) \leq B$ . This implies that  $r_n$  is uniformly integrable on  $(X, \mathcal{A}, Q)$ , since

$$\int_{r_n > C} r_n dQ \leq \frac{1}{\log C} \int_{r_n > C} r_n \log r_n dQ < \frac{B + 1}{\log C}$$

for any  $C > 1$  (note that  $t \log t > -1$ ). Thus  $r_n$  is convergent  $Q$  a.s. and in  $L_1$ -norm on  $(X, \mathcal{A}, Q)$  (cf. Theorem VII.4.1 in Doob (1953)), and the measure defined by

$$(3.6) \quad R(Y) = \int_Y \lim_{n \rightarrow \infty} r_n(x) dQ$$

for any  $Y \in \mathcal{A}$  is the desired limit of  $R_n$ .

We have seen that  $K_n(R, Q) \leq B$ . Similarly  $K_n(R, P_0) \leq a < A$ , and the proof is completed by the well-known fact that  $K_n$  tends to  $K$  if  $\mathcal{B}_n$  is monotone and generates  $\mathcal{A}$  (cf. Theorem 2.4.2 in Pinsker (1964)).  $\square$

**4. The exact slope of ERO-tests.** Let  $T_n$  be a sequence of statistics for testing  $H_0: P \in \mathcal{P}_0$ , large values of  $T_n$  being significant, and let the function  $G_n(t)$  be defined by

$$(4.1) \quad G_n(t) = \sup_{P \in \mathcal{P}_0} P(T_n \geq t).$$

The sequence  $T_n$  has exact slope  $c$  when  $Q$  obtains if

$$(4.2) \quad \lim_{n \rightarrow \infty} \frac{1}{n} \log G_n(T_n) = -\frac{1}{2}c(Q) \quad Q \text{ a.s.}$$

This definition is due to Bahadur (1960). He proved that

$$(4.3) \quad \liminf_{n \rightarrow \infty} \frac{1}{n} \log G_n(T_n) \geq -K(Q, \mathcal{P}_0) \quad Q \text{ a.s.}$$

(cf. Theorem 7.5 in Bahadur (1971)). He also proved that under suitable compactification conditions the exact slope of the likelihood ratio statistic is  $2K(Q, \mathcal{P}_0)$ .

**THEOREM 4.** *If the sequence of partitions  $\mathcal{B}_n$  is adequate (cf. Definition 3), and the sequence  $m(n)$  satisfies (1.11), then the exact slope of the statistic  $L_n$  is  $2K(Q, \mathcal{P}_0)$  (cf. Definition 2).*

**PROOF.** It is enough to prove that

$$(4.4) \quad \limsup_{n \rightarrow \infty} \frac{1}{n} \log H_n(L_n) \leq -K(Q, \mathcal{P}_0) \quad Q \text{ a.s.,}$$

where

$$H_n(t) = \sup_{P \in \mathcal{P}_0} P(L_n \geq t).$$

Let  $0 < a < A < K(Q, \mathcal{P}_0)$  be arbitrary. Theorem 2 implies

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log H_n(a) \leq -a;$$

thus it is enough to prove that

$$\liminf_{n \rightarrow \infty} L_n \geq a \quad Q \text{ a.s.}$$

Theorems 2 and 3 imply that

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log Q(L_n < a) \leq -K(\mathcal{P}_A, Q);$$

hence, by the Borel–Cantelli lemma, it is enough to prove that  $K(\mathcal{P}_A, Q) > 0$ . We have seen in the proof of Theorem 3 that

$$(4.5) \quad \lim_{n \rightarrow \infty} K_n(Q, \mathcal{P}_0) = K(Q, \mathcal{P}_0),$$

thus  $A < K_n(Q, \mathcal{P}_0)$  for some  $n$ . The  $m$ -dimensional set of probabilities  $\{R(Y_i); i = 1, \dots, m\}$  of those  $R$ -s for which  $K_n(R, \mathcal{P}_0) \leq A$ , is closed, and does not contain the vector  $\{Q(Y_i), i = 1, \dots, m\}$ , hence  $K_n(\mathcal{P}_{A,n}, Q) > 0$ . This completes the proof, because  $K_n(\mathcal{P}_{A,n}, Q) \leq K(\mathcal{P}_A, Q)$ .  $\square$

**5. Remarks, examples, problems.** First of all it should be noted that the adequateness of  $\mathcal{B}_n$  actually is a property of the system  $\{\mathcal{P}_0, \mathcal{B}_n, n = 1, 2, \dots\}$ . The convergence on  $\mathcal{B} \subset \mathcal{A}$  is a weaker version of the weak convergence of

probability measures defined by

$$(5.1) \quad P(Y) = \lim_{n \rightarrow \infty} P_n(Y) \quad \text{for all } Y \in \mathcal{A};$$

consequently the weak closedness is a necessary condition for the existence of an adequate sequence of partitions. Given any convergent sequence  $R_{n_k}$  on  $\mathcal{B}$  we can define  $\nu$  by (3.3) and  $r_n$  by (3.4) (by choosing any dominating  $Q$ ). Moreover  $r_n$  will still be a martingale with respect to  $\mathcal{B}_n$ . Hence it is convergent  $Q$  a.s.; but it is no longer convergent in  $L_1$  norm, unless  $r_n$  is uniformly integrable. In any case, if  $\mathcal{P}_0$  is dominated, and its densities form a compact set with respect to  $L_1$ -norm, then the limit measure (3.6) belongs to  $\mathcal{P}_0$ .

EXAMPLE 1. Let  $S$  and  $Q \in \mathcal{P}$  be arbitrary, and let

$$(5.2) \quad \mathcal{P}_0 = \{P: K(Q, P) \geq A\}.$$

Then  $Q \notin \mathcal{P}_A$ , but in most cases the closure of  $\mathcal{P}_0$  is equal to  $\mathcal{P}$  for any  $\mathcal{B}_n$ , hence there is no adequate sequence of partitions. On the other hand, one can prove that if a sequence of partitions is adequate with respect to  $\mathcal{P}_0$ , then it is adequate with respect to  $\mathcal{P}_A$  too. Consequently if  $\mathcal{A}$  is countably generated, then any monotone sequence of partitions is adequate with respect to  $\mathcal{P}_0 = \{P: K(P, Q) < A\}$ .

EXAMPLE 2. Let  $(X, \mathcal{A})$  be the real line with the  $\sigma$ -algebra of Borel-measurable sets, and let  $\mathcal{P}_0$  be the collection of all probability measures concentrated on the dyadic rationals in  $[0, 1)$ . Let  $Q$  be the uniform distribution on  $(0, 1)$ . Let  $\mathcal{B}_n$  be the partition of  $[0, 1)$  consisting of the half-open intervals  $[(i - 1)/2^n, i/2^n)$ . Then, for any  $A < \infty$ ,  $K_n(\mathcal{P}_{A,n}, Q) = 0$  and  $K(\mathcal{P}_A, Q) = \infty$ . Hence (3.1) does not hold, and  $\mathcal{B}_n$  is not adequate.

EXAMPLE 3. Let  $X$  be arbitrary, and let  $\mathcal{A}$  be a countably generated  $\sigma$ -algebra (e.g.,  $X$  is the Euclidean space and  $\mathcal{A}$  is the  $\sigma$ -algebra of Borel-measurable sets). Let  $\mathcal{P}_0 = \{P_0\}$ , where  $P_0$  is arbitrary. Then any monotone sequence of partitions generating  $\mathcal{A}$  is adequate. Consequently

$$(5.3) \quad B(A, Q, \{P_0\}) = K(\mathcal{P}_A, Q)$$

for any continuity point  $A$  of the function  $B(A) = B(A, Q, \{P_0\})$ . This function is convex, because  $K(R, P)$  is convex in  $R$ : for any  $R_1, R_2, 0 < t < 1$  and  $R = tR_1 + (1 - t)R_2$  we have

$$K(R, P) \leq tK(R_1, P) + (1 - t)K(R_2, P),$$

(which in turn is a consequence of the convexity of  $u \log u$ ). Hence the only discontinuity that may occur is at

$$(5.4) \quad A_\infty = \sup \{A: K(\mathcal{P}_A, Q) = \infty\}.$$

(5.3) has been obtained by Blahut (1972) in the case when  $\mathcal{A}$  is finitely generated. The majority of tests of rate  $A = 0$  for this problem are ERO at any  $Q$  (e.g., the Kolmogorov-Smirnov test in a Euclidean space). This however is true only

for  $A = 0$ , and one can expect that any ERO statistic for testing goodness-of-fit is asymptotically equal to  $L_n$ .

EXAMPLE 4. Let  $(X, \mathcal{A})$  be the  $k$ -dimensional Euclidean space with the  $\sigma$ -algebra of Borel-measurable sets, and let  $\mathcal{P}_0$  be the set of normal distributions. It is easy to see that there is no adequate sequence of partitions in this case. This is partly the consequence of using only finite partitions. The finiteness of the partitions was used in an essential manner in (2.11), but there is no evidence that it would really be needed for the validity of Theorem 2. Vasicek (1976) introduced a test for normality based on sample entropy. It would be worthwhile to determine whether his statistic is ERO or not.

Another limitation of our results is the lack of error terms. The statistic  $L_n$  is ERO, but for a reasonable sample size it may be very poor. This problem consists of two parts. One is the error term in Stein's lemma and in Theorem 1. The question is the following: are there any sequences  $\varepsilon_n \rightarrow 0$  depending on  $\mathcal{P}_0$ ,  $Q$  only through the function  $B(A, Q, \mathcal{P}_0)$  such that for any sequences  $\varphi_n$  for which

$$(5.5) \quad \frac{1}{n} \log \alpha_n \leq -A - \varepsilon_n$$

we have

$$(5.6) \quad \frac{1}{n} \log \beta_n(Q) \geq -B(A, Q, \mathcal{P}_0) - \varepsilon_n.$$

The other side of this problem is the speed of convergence of the sequence  $B_n$ . It is very likely that its investigation needs different methods from those used in the proof of Theorem 3.

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