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## On the Amount of Information Contained in a Sequence of Independent Observations

IGOR VAJDA

In the present paper basic properties of a Chernoff bound established previously are summarized and new ones are derived. The Chernoff bound is figuring as an asymptotic parameter in a formula for Shannon's information contained in a sequence of independent observations concerning a discrete parameter.

By  $\theta$  we denote a random variable taking on a finite number of values  $1, 2, \dots$  and by  $\xi$  another random variable with a sample measurable space  $(X, \mathcal{X})$ . By  $\xi_1, \xi_2, \dots$  subsequent realizations of  $\xi$  will be denoted; they are supposed to be mutually independent for any given value of  $\theta$ . Finally,  $I(\theta, \xi_1, \dots, \xi_n)$  will denote the Shannon's information contained in  $(\xi_1, \xi_2, \dots, \xi_n)$  concerning  $\theta$ .

The information  $I(\theta, \xi_1, \dots, \xi_n)$  can serve as an important numerical characteristic of the following statistical problem: the statistician is interested in the value of  $\theta$  which is not directly observable but he can observe the values of  $\xi_1, \xi_2, \dots, \xi_n$ . It holds  $I(\theta, \xi_1, \dots, \xi_n) = 0$  iff (if and only if) the sample  $(\xi_1, \xi_2, \dots, \xi_n)$  is independent of  $\theta$ . In general  $I(\theta, \xi_1, \dots, \xi_n) \in [0, H(\theta)]$ , where  $H(\theta)$  is the Shannon's entropy of the random variable  $\theta$ ; relation  $I(\theta, \xi_1, \dots, \xi_n) = H(\theta)$  holds iff for any realization of  $(\xi_1, \xi_2, \dots, \xi_n)$  the value of  $\theta$  can be uniquely determined with probability 1. (Remark that the first equality holds iff  $\theta$  and  $\xi$  are independent whereas the second equality holds iff there exists a deterministic relation between  $\theta$  and  $\xi$ .)

It can be relatively very easily shown (cf. Th. 1 in [1]) that\*

$$(1) \quad I(\theta, \xi_1, \dots, \xi_n) \approx H(\theta) - \exp(-nD),$$

where  $D \in [0, +\infty]$  depends on a conditional distribution  $P_{\xi|\theta}$  of  $\xi$  only. The parameter  $D$  has been independently evaluated by A. Rényi [2] and by the author (cf. Th. 2 in [1]); it was shown that  $D$  is the Chernoff bound [3] corresponding to a Bayes testing of the simple hypotheses  $H_i: \theta = i, i = 1, 2, \dots$ , on the basis of  $(\xi_1, \xi_2, \dots$

\* We write  $a_n \approx a - \lambda^n$  instead of  $a_n = a - \lambda^{n+o(n)}, n = 1, 2, \dots$

...,  $\xi_n$ ). In [3],  $D$  has been interpreted as an asymptotic efficiency of the Bayes test suggested above.

Some basic properties of the parameter  $D$  were presented in [3], another ones were stated in [1], however, without explicit proofs. Moreover, it is to be noted that assertions (d), (e), and (g) in [1] hold only if the probability measures considered there are absolutely continuous (this supposition was not explicitly emphasized in [1]). Consequently, an analogical investigation of the "discontinuous" case which is very interesting too is advisable. Therefore, by the present rather review paper we are resuming the subject of [1].

In Theorems 5–7 and 9, 10 below the assertions of [1] are summarized (including the case where the probability measures mentioned above are not absolutely continuous). In Theorems 1–4 basic properties of a modified  $\alpha$ -entropy and a modified relative Shannon's entropy are established. The modified concepts differ from non-modified ones in the "discontinuous" case mentioned above only; it seems however that they are not only more suitable than the non-modified ones when asymptotic problems of the present type are solved, but also provide tools for a more accurate analysis of such problems. (In this respect, compare, for example, (b) and (d) in [1] or (2.8) in [2] with Th. 5 below.) Finally, in Th. 8 a convergence property of  $D$ 's corresponding to a sequence of sub- $\sigma$ -algebras of  $\mathcal{X}$  is established. Though  $D$  is a special version of the  $\alpha$ -entropy, this property cannot be deduced directly from the semimartingale convergence theorem.

## 1. MODIFIED CONCEPTS OF $\alpha$ - AND SHANNON'S ENTROPY

Already in [3] a functional of the following form

$$\int_X p^\alpha q^{1-\alpha} d\mu, \quad \alpha \in (0, 1),$$

(cf. also [4]) has been investigated, where  $p, q$  are the Radon-Nikodym densities of probability distributions  $P, Q$  on  $(X, \mathcal{X})$  with respect to another (dominating) probability distribution  $\mu$  on  $(X, \mathcal{X})$ . In accordance with [4], the functional will be denoted by  $H_\alpha(P, Q)$  and called, simply,  $\alpha$ -entropy. Some basic properties of this functional can be deduced from Theorem 4.1.s in Chap. VII of [5].

Before going into a more detailed discussion of  $\alpha$ -entropies, let us note that in the statistical model introduced above we shall suppose that  $\theta$  takes on two values 1 and 2 only and that  $P[\theta = 1] = P[\theta = 2] = \frac{1}{2}$ . It follows from what was said in [1, 2] that the general case where the number of possible values of  $\theta$  is arbitrary finite does not present any essential new difficulty. (In the general case  $D$  is defined as the minimum of the Chernoff bounds corresponding to the pairs of hypotheses  $\theta = i, \theta = j$  such that  $P[\theta = i] > 0, P[\theta = j] > 0$ , taken over all such pairs.) In the sequel,  $P$  or  $Q$  will be interpreted as the conditional distribution  $P_{\xi|\theta=1}$  or  $P_{\xi|\theta=2}$  respectively.

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$$(2) \quad \begin{aligned} P_{\xi_1, \xi_2, \dots, \xi_n | \theta=1} &= P \times P \times \dots \times P (n \text{ times}), \\ P_{\xi_1, \xi_2, \dots, \xi_n | \theta=2} &= Q \times Q \times \dots \times Q (n \text{ times}). \end{aligned}$$

In a connection with an evaluation of the parameter  $D$ , the following slightly modified concept of the  $\alpha$ -entropy will be useful

$$(3) \quad H'_\alpha(P, Q) = \int_{C(P, Q)} p^\alpha q^{1-\alpha} d\mu,$$

where  $C(P, Q) = \{pq > 0\} \in \mathcal{X}$  is a set of absolute continuity of  $P, Q$ .

It is to see at the first sight that  $H'_\alpha = H_\alpha$  if  $P, Q$  are mutually absolutely continuous and

$$(4) \quad H'_\alpha(P, Q) = H_\alpha(P, Q) \quad \alpha \neq 0, 1,$$

for every  $P, Q$ . (Let us note that, unless the contrary will be explicitly stated, we shall consider the  $\alpha$ -entropies for  $\alpha \in [0, 1]$  only.) Further, it is fruitful to notice (cf. [4, 6]) that  $H'_\alpha(P, Q)$  is the real restriction of

$$H(z) = \int_{-\infty}^{+\infty} e^{zu} dF(u),$$

where  $z = \alpha + i\beta$  is a complex number and  $F(u) = Q(\{p \leq q \exp(u)\})$  is the distribution function of the likelihood ratio corresponding to the simple hypotheses  $P$  and  $Q$ . It follows from the theory of bilateral Laplace transform (cf. [7]) that  $H'_\alpha(P, Q)$  is finite for  $\alpha \in [0, 1]$  and that  $H'_\alpha(P, Q)$  is an analytic function of  $\alpha$  on  $(0, 1)$  with derivatives (cf. [4])

$$(5) \quad \frac{d^k}{d\alpha^k} H'_\alpha(P, Q) = \int_{\mathcal{X}} p^\alpha q^{1-\alpha} \left(\log \frac{p}{q}\right)^k d\mu \quad \text{for every } \alpha \in (0, 1), k = 1, 2, \dots$$

which, however, need not be always continuous at  $\alpha = 0, 1$ .

Using (4) these results may be immediately applied to  $H'_\alpha$ ; as we shall prove below (cf. Th. 2, where properties of  $H'_\alpha$  as a function of  $\alpha$  are summarized),  $H'_\alpha(P, Q)$  is continuous on  $[0, 1]$  (or, more generally, on the set  $J$  of all  $\alpha$  for which  $H'_\alpha(P, Q) < +\infty$ ; it follows from the theory of bilateral Laplace transform that  $J$  is always an interval on the real line).

Since

$$(6) \quad H'_\alpha(P, Q) = E_Q \chi_{C(P, Q)} \left(\frac{P}{Q}\right)^\alpha,$$

the semimartingale theorem cannot be applied to  $H'_\alpha$  unless  $\chi_{C(P, Q)} = 1 [Q]$ . Nevertheless we shall see that  $H'_\alpha$  possesses all the convergence properties, which can

be derived for  $H_\alpha$  from the semimartingale convergence theorem. Of course, in view of (4), we may restrict ourselves to the case  $\alpha = 0$  or  $1$ .

Let  $\mathcal{X}_1 \subset \mathcal{X}_2 \subset \dots$  be sub- $\sigma$ -algebras of  $\mathcal{X}$  and let  $P_n, Q_n$  be restrictions of  $P, Q$  on  $\mathcal{X}_n, n = 1, 2, \dots$ . Clearly,  $C(P_n, Q_n) \in \mathcal{X}_n$ .

We shall say that a sub- $\sigma$ -algebra  $\mathcal{X}_n \subset \mathcal{X}$  is  $C(P, Q)$ -sufficient if  $P(C(P_n, Q_n)) = P(C(P, Q))$ . Obviously, if  $\mathcal{X}_n$  is  $C(P, Q)$ -sufficient, it need not be also  $C(Q, P)$ -sufficient, but if it is sufficient with respect to  $P$  and  $Q$ , then it is  $C(P, Q)$ - as well as  $C(Q, P)$ -sufficient.

**Theorem 1.** For every  $\alpha \in [0, 1]$

$$(7) \quad H'_\alpha(P_1, Q_1) \geq H'_\alpha(P_2, Q_2) \geq \dots$$

and, if  $\mathcal{X}$  is generated by the algebra

$$\mathcal{X}_0 = \bigcup_{n=1}^{\infty} \mathcal{X}_n,$$

then

$$(8) \quad \lim_n H'_\alpha(P_n, Q_n) = H'_\alpha(P, Q).$$

If  $\alpha \in (0, 1)$  then  $H'_\alpha(P_n, Q_n) = H'_\alpha(P, Q)$  iff (if and only if)  $\mathcal{X}_n$  is sufficient with respect to  $P$  and  $Q$ . If  $\alpha = 0$  or  $1$  then this equality holds iff  $\mathcal{X}_n$  is  $C(Q, P)$ -sufficient or  $C(P, Q)$ -sufficient respectively.

*Proof.* By (4), the assertion stated here for  $\alpha \in (0, 1)$  has been proved in [5]. If  $\alpha = 0, 1$  then, it may be easily deduced from the inclusion

$$(9) \quad C(P_n, Q_n) \supset C(P, Q)$$

and from the fact that  $\{1 - \chi_{C(P_n, Q_n)}\}, n = 1, 2, \dots$  is a semimartingale with respect to both  $P$  and  $Q$ .

To prove (9) it will suffice to prove that the conditional densities

$$(10) \quad p_n = E(p \mid \mathcal{X}_n), \quad q_n = E(q \mid \mathcal{X}_n)$$

may be defined in such a way that  $p(x)q(x) > 0$  implies  $p_n(x)q_n(x) > 0, x \in X$ . If  $E = \{p_n = 0\} \in \mathcal{X}_n$ , then the equality defining  $p_n$  implies that the set  $F \subset E$  of all  $x \in X$  for which  $p(x) > 0$  is of  $P$ -measure zero, i.e. we may put  $p_n = 1$  on  $F$ . Thus  $p_n(x) = 0$  implies  $p(x) = 0$  for every  $x \in X$ . Since we may analogically proceed with  $q, q_n$ , the implication requested above is true. (9) implies that  $\{1 - \chi_{C(P_n, Q_n)}\}, n = 1, 2, \dots, \infty$  is a semimartingale, Q.E.D.

**Theorem 2.**  $H'_\alpha(P, Q)$  is continuous convex function on  $[0, 1]$  with

$$(11) \quad \frac{d^k}{d\alpha^k} H'_\alpha(P, Q) = \int_{C(P, Q)} p^\alpha q^{1-\alpha} \left( \log \frac{p}{q} \right)^k d\mu \quad \text{for every } k = 1, 2, \dots \text{ and } \alpha \in [0, 1],$$

310 where the integrals in (11) are finite for  $\alpha \in (0, 1)$  and well-defined for  $\alpha = 0, 1$ , and

$$0 \leq H'_\alpha(P, Q) \leq 1,$$

where  $H'_\alpha(P, Q) = 0$  for some  $\alpha \in [0, 1]$  (and, consequently, for all  $\alpha \in [0, 1]$ ) iff  $P \perp Q$  and  $H'_\alpha(P, Q) = 1$  for some  $\alpha \in (0, 1)$  (and, consequently, for all  $\alpha \in [0, 1]$ ) iff  $P = Q$ . For  $\alpha = 0$  or  $1$ ,  $H'_\alpha(P, Q) = 1$  iff  $Q \ll P$  or  $P \ll Q$  respectively.  $H'_\alpha$  is strictly convex if neither  $P \perp Q$  nor  $P = Q$ .

*Remark.* The derivatives in (11) for  $\alpha = 0$  or  $1$  are to be considered as those on the right or left respectively.

*Proof.* We shall prove firstly that  $H'_\alpha$  is continuous on  $[0, 1]$ . One of the methods to prove this is to form a sequence of sub- $\sigma$ -algebras  $\mathcal{X}_1 \subset \mathcal{X}_2 \subset \dots$  of  $\mathcal{X}$  generated by finite measurable decompositions of  $X$ . As it was shown in [8], the decompositions may be defined in such a manner that the  $\sigma$ -algebra  $\mathcal{X}' \subset \mathcal{X}$  generated by the corresponding algebra  $\mathcal{X}_0$  (cf. Th. 1) is sufficient with respect to  $P$  and  $Q$ . Since, evidently, every  $H'_\alpha(P_n, Q_n)$  is continuous and convex on  $[0, 1]$ , it follows from Th. 1 that  $H'_\alpha(P, Q)$  is a limit of continuous and uniformly converging (on  $[0, 1]$ ) functions, i.e. it is continuous as well. The convexity will follow from (11) for  $k = 2$  and the assertions following (11) can be deduced from (3) and (11).

Thus it remains to prove that the integrals in (11) are finite or well-defined respectively and that (11) holds. But, according to (5) (see also [4]), the integrals (11) are finite for every  $\alpha \in (0, 1)$  and  $k = 1, 2, \dots$ . Since the functions  $u(\log u)^k$  are bounded from below for every  $u \in (0, +\infty)$  and  $k = 1, 2, \dots$ , the integrals in (11) are well-defined for  $\alpha = 1$  as well. The same is true also for  $\alpha = 0$  and  $k = 2, 4, 6, \dots$ . If  $k$  is odd, then we can write

$$\int_{C(P, Q)} q \left( \log \frac{p}{q} \right)^k d\mu = - \int_{C(P, Q)} q \left( \log \frac{q}{p} \right)^k d\mu$$

so that, interchanging the role of  $P$  and  $Q$  in the case  $\alpha = 1$  above we obtain the desired assertion.

Relation (11) holds for every  $\alpha \in (0, 1)$  and  $k = 1, 2, \dots$  by (4) and (5). If  $\alpha = 0$  and  $k = 1$  (for  $\alpha = 1$  as well as  $k = 2, 3, \dots$  a similar argument can be used), we can write

$$H'_0(P, Q) - H'_\alpha(P, Q) = \alpha \int_{C(P, Q)} u^{\xi(u)} \log u dQ \quad \text{for every } \alpha \in (0, 1),$$

where  $\xi(u) \in [0, \alpha]$  is a Borel function of  $u \in [0, +\infty]$  and  $u = p/q$  on  $C(P, Q)$ . If

$$\int_{C(P, Q)} \log u dQ$$

is finite the proof is obvious. Now, since  $\log u \leq u - 1$  for every real  $u$ , the following inequality holds 311

$$\int_{C(P, Q)} \log u \, dQ \leq P(C(P, Q)) - Q(C(P, Q)) < +\infty$$

and it remains to investigate the case

$$\int_{C_*} \log u \, dQ = -\infty,$$

where the set  $C_* \in \mathcal{X}$  is defined by  $C_* = \{u \leq 1\} \cap C(P, Q)$ . Since  $\xi(u) \in [0, \bar{x}]$ , it holds

$$u^\alpha \log u \leq u^\alpha \log u \quad \text{on } C_*$$

and it remains to prove that for every  $A > 0$  there exists  $\alpha \in (0, 1)$  such that

$$\int_{C_*} u^\alpha \log u \, dQ < -A.$$

If we  $F_n = \{u \geq 1/n\} \cap C_* \in \mathcal{X}$ , then

$$\lim_n \int_{F_n} \log u \, dQ = -\infty$$

so that, for some  $n$ ,

$$\int_{F_n} \log u \, dQ \leq -2A.$$

If now  $0 \leq \alpha < \log 2 / \log n$ , then

$$u^\alpha \log u < \frac{1}{2} \log u \quad \text{on } F_n,$$

and we can successively write

$$\int_{C_*} u^\alpha \log u \, dQ \leq \int_{F_n} u^\alpha \log u \, dQ < \frac{1}{2} \int_{F_n} \log u \, dQ \leq -A.$$

The same modification as above we shall also consider in connection with the generalized entropy of Shannon (or discrimination information) of  $P, Q$  introduced into the literature by S. Kullback and A. Perez, i.e. instead of

$$H(P, Q) = \int_X p \log \frac{p}{q} \, d\mu$$

we shall consider

$$(12) \quad H'(P, Q) = \int_{C(P, Q)} p \log \frac{p}{q} \, d\mu \geq P(C) \log \frac{P(C)}{Q(C)} \geq -\frac{1}{e},$$

312 where  $C$  stands for  $C(P, Q)$ . Let us notice that  $H'(P, Q)$  may take on negative values as well and that  $H'(P, Q) < 0$  implies  $H(P, Q) = +\infty$ .

**Theorem 3.** *If  $H'(P, Q) \leq 0$  then  $H'(Q, P) \geq 0$  where the strict inequality holds unless either  $P \perp Q$  on  $\mathcal{X}$  or  $P = Q$  on  $C(P, Q) \cap \mathcal{X}$ .*

*Proof.* From (12) we obtain

$$(13) \quad H'(P, Q) + H'(Q, P) \geq (P(C) - Q(C)) \log \frac{P(C)}{Q(C)} \geq 0$$

so that  $H'(P, Q) \leq 0$  or  $< 0$  implies  $H'(Q, P) \geq 0$  or  $> 0$  respectively. If  $H'(P, Q) = H'(Q, P) = 0$  then, by (13),  $P(C) = Q(C)$  so that, according to Lemma 1.1 in [9], either  $P(C) = Q(C) = 0$  (i.e.  $P \perp Q$  on  $\mathcal{X}$ ) or  $P = Q$  on  $C \cap \mathcal{X}$ .

The following identity (14) was found for discrete distributions by A. Rényi [10] (cf. also [11]).

**Theorem 4.** *For every  $P$  and  $Q$ ,*

$$(14) \quad \lim_{\alpha \rightarrow 1^-} \frac{1}{\alpha - 1} \log H'_\alpha(P, Q) = H(P, Q),$$

$$(15) \quad \lim_{\alpha \rightarrow 0^+} -\frac{1}{\alpha} \log H'_\alpha(P, Q) = H(Q, P).$$

*Proof.* We shall prove (14) only; (15) may be proved analogically. If  $P \not\ll Q$ , then  $H(P, Q) = +\infty$  and  $H'_1(P, Q) < 1$  (see Th. 2) so that (14) holds. If  $P \ll Q$ , then, by Th. 2,  $H'_1(P, Q) = 1$  so that we can successively write

$$\lim_{\alpha \rightarrow 1^-} \frac{1}{\alpha - 1} \log H'_\alpha(P, Q) = \frac{\lim_{\alpha \rightarrow 1^-} \frac{d}{d\alpha} H'_\alpha(P, Q)}{\lim_{\alpha \rightarrow 1^-} H'_\alpha(P, Q)} = \frac{H(P, Q)}{H'_1(P, Q)} = H(P, Q)$$

(cf. (11), (12)).

It is to see that (14) and (15) may be replaced by

$$\frac{d}{d\alpha} H_\alpha(P, Q)|_{\alpha=0} = -H(Q, P),$$

$$\frac{d}{d\alpha} H_\alpha(P, Q)|_{\alpha=1} = H(P, Q),$$



where the derivatives are to be considered as those on the right or left respectively. 313  
 Analogical relations

$$\frac{d}{dz} H'_z(P, Q)|_{z=0} = -H'(Q, P)$$

$$\frac{d}{dz} H'_z(P, Q)|_{z=1} = H'(P, Q)$$

follow immediately from (11) and (12).

## 2. D-DIVERGENCE

Now our attention will be paid to  $D(P, Q)$  which is a parameter of convergent in (1). The fact that  $D(P, Q)$ , as it will be defined in this section, is identical with that of the formula (1) will be proved, for the sake of completeness, in the following section.

Let us put\* (cf. (e) in [1])

$$(16) \quad D(P, Q) = \sup_{\alpha \in [0,1]} -\log H_\alpha(P, Q) = -\log \min_{\alpha \in [0,1]} H'_\alpha(P, Q).$$

According to (4) and Th. 2, the minimum in (16) exists and the second equality holds. Th. 2 also implies the following two theorems (cf. (b) and (d) in [1]).

**Theorem 5.**  $H'(P, Q), H'(Q, P) > 0$  iff

$$(17) \quad D(P, Q) = -\log H'_\alpha(P, Q),$$

for  $\alpha \in (0, 1)$  which is a unique solution of the equation

$$(18) \quad \int_{C(P, Q)} p^\alpha q^{1-\alpha} \log \frac{p}{q} d\mu = 0,$$

$H'(P, Q) \leq 0$  iff

$$(19) \quad D(P, Q) = -\log H'_1(P, Q),$$

and  $H'(Q, P) \leq 0$  iff

$$D(P, Q) = -\log H'_0(P, Q).$$

According to Th. 3,  $H'(P, Q), H'(Q, P) \leq 0$  iff  $H'(P, Q) = H'(Q, P) = 0$  which is equivalent to  $P \perp Q$  or  $P = Q$ . By Th. 2, both later conditions imply  $H'_0(P, Q) = H'_1(P, Q)$  so that Th. 5 is self-consistent. Let us recall that  $H'(P, Q) \neq H(P, Q)$ ,

\* By log we denote in this paper the natural logarithm.

314 i.e., particularly,  $H'(P, Q) < 0$ , may appear only if  $P \not\equiv Q$ , so that if  $P \equiv Q$ ,  $P \neq Q$ , then (17) is true.

**Theorem 6.**  $D(P, Q)$  is symmetric non-negative extended real valued function of  $P, Q$ .  $D(P, Q) = 0$  iff  $P = Q$  and  $D(P, Q) = +\infty$  iff  $P \perp Q$ .

The symmetry stated in this theorem follows from (16) and from the identity  $H'_\alpha(Q, P) = H'_{1-\alpha}(P, Q)$  which is true for every  $\alpha \in [0, 1]$ .

In [3] it was proved that (cf. (f) in [1])

$$D\left(\prod_{i=1}^n P_i, \prod_{i=1}^n Q_i\right) \leq \prod_{i=1}^n D(P_i, Q_i)$$

and

$$D\left(\prod_{i=1}^n P_i, \prod_{i=1}^n Q_i\right) = n D(P, Q) \quad \text{if } P_i = P, Q_i = Q, i = 1, 2, \dots$$

Th. 1 together with Th. 2 (cf. (11), (12)) yields the following result (cf. (g) in [1]).

**Theorem 7.** If  $P', Q'$  are restrictions of  $P, Q$  on a sub- $\sigma$ -algebra  $\mathcal{X}'$  of  $\mathcal{X}$ , then

$$(21) \quad D(P', Q') \leq D(P, Q)$$

where the sign of equality holds iff  $\mathcal{X}'$  is sufficient with respect to  $P, Q$  or  $C(P, Q)$ -sufficient or  $C(Q, P)$ -sufficient depending on whether  $H'(P, Q), H'(Q, P) > 0$  or  $H'(P, Q) \leq 0$  or  $H'(Q, P) \leq 0$  respectively.

The following assertion is new.

**Theorem 8.** If  $\mathcal{X}_n, P_n, Q_n$  are defined as in Th. 1, then

$$(22) \quad D(P_1, Q_1) \leq D(P_2, Q_2) \leq \dots$$

and if  $\mathcal{X}$  is generated by the algebra  $\mathcal{X}_0$  (cf. Th. 1), then

$$(23) \quad \lim_n D(P_n, Q_n) = D(P, Q).$$

*Proof.* According to (16) and Th. 7,

$$-\log H'_\alpha(P_n, Q_n) \leq D(P_n, Q_n) \leq D(P, Q)$$

where  $\alpha$  is defined by  $D(P, Q) = -\log H'_\alpha(P, Q)$ . Now it remains to apply Th. 1.

Next we shall prove that  $D(P, Q)$  as defined by (16) is identical with that defined by a different manner in (3.2) of [1]. The definition (3.2) was merely based on the concept of generalized Shannon's entropy. As a by-product the inequality  $D(P, Q) \leq \min [H(P, Q), H(Q, P)]$  will be obtained. This result becomes evident if compared with the Chernoff-Stein asymptotical formulas for the power of Neyman-Pearson tests of  $\theta = 1(2)$  against  $\theta = 2(1)$  based on  $\xi_1, \xi_2, \dots, \xi_n$ . In these formulas  $H(P, Q)$

(or  $H(Q, P)$ ) is figuring in the exponent of convergence analogically as  $D(P, Q)$  in (1). For a deeper insight into these questions we refer to [13, 11] (cf. also the following formula (36)).

Let  $P$  and  $Q$  be arbitrary fixed probability measures and denote by  $\mathcal{P}$  or  $\mathcal{Q}$  the set of all probability measures  $R$  on  $(X, \mathcal{X})$  such that

$$(24) \quad H(R, P) \geq H(R, Q)$$

or

$$(25) \quad H(R, P) \leq H(R, Q)$$

respectively. The definition we beared in mind above is as follows:

$$(26) \quad D(P, Q) = \min_{\mathcal{P}} [\inf H(R, P), \inf_{\mathcal{Q}} H(R, Q)].$$

The next our aim will be to prove and precise (26)\*.

Let  $\mathcal{P}_0 \subset \mathcal{P}$  or  $\mathcal{Q}_0 \subset \mathcal{Q}$  denote the subclasses of all  $R$  such that  $H(R, P) < +\infty$  or  $H(R, Q) < +\infty$  respectively and let  $\mathcal{A}$  stands for the set of all measures  $R$  dominated by  $\mu$  and concentrated on  $C(P, Q)$ , i.e.  $R(C(P, Q)) = 1$ .

**Lemma 1.**  $\mathcal{P}_0 \cup \mathcal{Q}_0 \subset \mathcal{A}$ , i.e. if  $R \in \mathcal{P} \cup \mathcal{Q} - \mathcal{A}$ , then  $H(R, P) \searrow H(R, Q) = +\infty$ .

*Proof.* Let  $R \in \mathcal{P} - \mathcal{A}$  and let us distinguish two alternate cases:  $R \ll \mu$ ,  $R(C) = 1$  and  $R \not\ll \mu$ ,  $R(X - C) > 0$ , where, here and in the sequel,  $C$  denotes  $C(P, Q)$ . If  $R \ll \mu$ , then also  $R \ll P$  so that, by the definition of the generalized Shannon's entropy,  $H(R, P) = +\infty$ . If  $R(X - C) > 0$ , then either  $R \not\ll P$  or  $R \not\ll Q$ . The first case we have just investigated above and if  $R \not\ll Q$ , then  $H(R, Q) = +\infty$ . This together with the condition (24) for  $R \in \mathcal{P}$  implies  $H(R, P) = +\infty$ , Q.E.D.

**Lemma 2.** If  $R \in \mathcal{A}$ , then  $R \in \mathcal{P}_0$  or  $\mathcal{Q}_0$  iff

$$(27) \quad \int_C r \log \frac{q}{p} d\mu \geq 0 \quad \text{or} \quad \leq 0$$

and  $R \in \mathcal{P}_0 \cap \mathcal{Q}_0$  iff

$$\int_C r \log \frac{q}{p} d\mu = 0,$$

where  $r = dR/d\mu$ .

*Proof.* If  $R \in \mathcal{P}_0 \subset \mathcal{A}$ , then  $r = dR/d\mu$  exists by Lemma 1. It follows from the definition of  $C$  that the integral in (27) or (28) exists. The remainder is clear.

\* During a preparation of this manuscript for printing, A. Rényi has published analogical definition of  $D$  in the printed version [2] of his lecture.

316 **Lemma 3.** For every  $R \in \mathcal{R}$  and  $\alpha \in [0, 1]$

$$(29) \quad H(R, P) \geq (1 - \alpha) \int_C r \log \frac{q}{p} d\mu - \log H_\alpha(P, Q)$$

with equality iff

$$(30) \quad r = \begin{cases} (H_\alpha(P, Q))^{-1} p^\alpha q^{1-\alpha} & \text{on } C \\ 0 & \text{out of } C. \end{cases}$$

Proof. S. Kullback proved in Chap. 3 of his book [12] that for every extended real-valued measurable statistic  $T$  defined on  $(X, \mathcal{X})$ , for every real  $\tau$  and non-negative  $\beta$ , and for every  $R \ll \mu$  the following inequality holds

$$H(R, P) \geq \tau \int_X r T d\mu + \log \beta + 1 - \beta \int_X \exp(\tau T) d\mu$$

if only the corresponding integrals exist and that the equality takes place iff  $r = \exp(\tau T)$ . Putting  $\tau = \alpha - 1$ ,

$$T = \begin{cases} \log \frac{p}{q} & \text{on } C, \\ -\infty & \text{out of } C, \end{cases}$$

and  $\beta = (H_\alpha(P, Q))^{-1}$  we obtain (29). The rest of the proof is now clear.

On the basis given by these lemmas, (26) can be easily proved. Let us consider firstly the case where  $H'(P, Q), H'(Q, P) > 0$ . Here  $P(E \cap C), Q(C - E) > 0$  and, since  $P, Q$  are absolutely continuous on  $C \cap \mathcal{X}, P(C - E), Q(E \cap C) > 0$ , where  $E = \{\log p/q \geq 0\} \in \mathcal{X}$ . It is easy to see that these facts enable us to argue that  $\mathcal{P}_0 \cap \mathcal{Q}_0 \neq \emptyset$ . Further, the definition of  $\mathcal{P}_0, \mathcal{Q}_0$  yields

$$(31) \quad \inf_{\mathcal{P}} H(R, P) = \inf_{\mathcal{P}_0} H(R, P), \quad \inf_{\mathcal{Q}} H(R, Q) = \inf_{\mathcal{Q}_0} H(R, Q).$$

However, we shall prove more, namely,

$$(32) \quad \inf_{\mathcal{P}_0} H(R, P) = \inf_{\mathcal{P}_0 \cap \mathcal{Q}_0} H(R, P) = \inf_{\mathcal{P}_0 \cap \mathcal{Q}_0} H(R, Q) = \inf_{\mathcal{Q}_0} H(R, Q).$$

**Theorem 9.** If  $H'(P, Q), H'(Q, P) > 0$ , then  $\mathcal{P}_0 \cap \mathcal{Q}_0 \neq \emptyset$ , (32) holds and

$$D(P, Q) = \inf_{\mathcal{P}_0 \cap \mathcal{Q}_0} H(R, P) = H(R, P),$$

where  $R \in \mathcal{P}_0 \cap \mathcal{Q}_0$  is uniquely  $[\mu]$  defined by (30) for  $\alpha \in (0, 1)$  given by (18).

Proof. Let  $\alpha$  in Lemma 3 be defined by (18) and let  $R \in \mathcal{P}_0 \cap \mathcal{Q}_0$  be arbitrary. Then, by Lemmas 3, 2 and Theorem 5,  $H(R, P) \geq D(P, Q)$  with equality iff (30) holds. Q.E.D.

**Theorem 10.** If  $H'(P, Q) \leq 0$  and  $P, Q$  are not mutually singular, then  $\mathcal{P}_0 \neq 0$  and

$$D(P, Q) = \inf_{\mathcal{P}_0} H(R, P) = H(R, P) \leq \inf_{\mathcal{Q}} H(R, Q),$$

where  $R \in \mathcal{P}_0$  is defined uniquely  $[\mu]$  by (30) for  $\alpha = 1$ .

*Proof.* If  $P$  and  $Q$  are not singular, then  $P(C) > 0$  and  $r$  defined by (30) for  $\alpha = 1$  is a probability density function. By Lemma 2,  $R \in \mathcal{R}$  given by  $r$  belongs to  $\mathcal{P}$  (and, consequently, to  $\mathcal{P}_0$ ) iff  $H'(P, Q) \leq 0$ . The equalities in Th. 10 now follow from Lemma 3 (for  $\alpha = 1$ ) and from (19). As to the inequality, let us notice that, replacing  $P$  and  $Q$  in Lemma 3, we may write

$$H(R, Q) \geq \int_c r \log \frac{p}{q} d\mu + D(P, Q)$$

for every  $R \in \mathcal{R}$ , where  $D(P, Q)$  is defined by (19) again. But (27) and Lemma 2 imply that the integral is non-negative for any  $R \in \mathcal{R}$ , i.e. the inequality is true.

Th. 9 and Th. 10 imply the following

**Corollary.** The relation (26) holds. If  $D(P, Q) < +\infty$ , then the minimum in (26) is attained on  $R \in \mathcal{R}$  defined by (30) for appropriately defined  $\alpha \in [0, 1]$ .

Since  $P \in \mathcal{Q}, Q \in \mathcal{P}$ , (26) implies the following inequality:

$$(33) \quad D(P, Q) \leq \min [H(P, Q), H(Q, P)].$$

### 3. TOTAL VARIATION

In [1] an estimate of  $D(P, Q)$  in terms of a more simple functional  $V(P, Q)$  was given.  $V(P, Q)$  was denoting the total variation of  $P$  and  $Q$  (cf. (h) in [1]). The total variation is defined by

$$(34) \quad V(P, Q) = \int_x |p - q| d\mu = 2 \sup_{E \in \mathcal{X}} [P(E) - Q(E)] = 2[P(F) - Q(F)],$$

where  $F = \{p \geq q\} \in \mathcal{X}$ . The estimate was of the following form\*

$$(35) \quad -\frac{1}{2} \log \left( 1 - \frac{V^2(P, Q)}{4} \right) \leq D(P, Q) \leq -\log \left( 1 - \frac{V(P, Q)}{2} \right).$$

The right hand inequality follows directly from the following formula (36) and from the inequality

$$1 - \frac{1}{2} V(P^n, Q^n) \geq (1 - \frac{1}{2} V(P, Q))^n \quad n = 1, 2, \dots,$$

\* My thanks are due to Prof. O. Kraft for calling my attention to the fact that this estimate occurs also in Ch. Kraft, Univ. California Publ. Statist. 2 (1955), 125–142 (added in proof).

318 (cf. (37)) which is the proof of Th. 1 in [1] based on. The left hand inequality may be proved by a method indicated in [14] (cf. the proof of the inequality (15) in [14]; in this proof it is indifferent whether the measures  $P, Q$  are discrete or not), but here another idea will be used.

Let  $\mathcal{X}'$  be the sub- $\sigma$ -algebra of  $\mathcal{X}$  consisting of two elements  $F, X - F \in \mathcal{X}$ , where  $F$  is defined as in (34) and let  $P', Q'$  be reductions of  $P, Q$  on  $\mathcal{X}'$ . Then, by Th. 7,  $D(P', Q') \leq D(P, Q)$ , where

$$D(P', Q') = -\log \inf_{\alpha \in (0,1)} \psi_{\alpha}(U, V) \quad \text{for } U = Q(F), \quad V = V(P, Q),$$

and where

$$\psi_{\alpha}(U, V) = \left(\frac{V}{2} + U\right)^{\alpha} U^{1-\alpha} + \left(1 - \frac{V}{2} - U\right)^{\alpha} (1-U)^{1-\alpha},$$

$$0 \leq U \leq 1 - \frac{V}{2}, \quad 0 \leq V \leq 2.$$

Thus it remains to prove that

$$\sup_{U \in [0, 1 - V/2]} \inf_{\alpha \in (0,1)} \psi_{\alpha}(U, V) \leq \sqrt{\left(1 - \frac{V^2}{4}\right)} \quad \text{for every } V \in (0, 2),$$

or

$$\sup_{U \in [0, 1 - V/2]} \psi_{1/2}(U, V) \leq \sqrt{\left(1 - \frac{V^2}{4}\right)}.$$

But, however,  $\psi_{1/2}(U, V)$  is strictly concave function of  $U$  on the interval  $[0, 1 - V/2]$  with maximum on  $U_0 = \frac{1}{2}(1 - V/2)$ , for any  $V \in (0, 2)$  so that the desired result follows from this identity:

$$\psi_{1/2}(U_0, V) = \sqrt{\left(1 - \frac{V^2}{4}\right)}.$$

The main idea of [1, 2] was based on the fact that a relation between the variation  $V(P^n, Q^n)$  and the quantity  $H(\theta) - I(\theta, \xi_1, \dots, \xi_n)$  (cf. (3) and the assumption following it) exists. This relation is represented by a both-sides estimate which is "best possible", i.e. for any value  $V$  of  $V(P^n, Q^n)$ ,  $V \in [0, 2]$ , one can find two nonnegative numbers  $L_n(V), U_n(V)$  such that

$$L_n(V) \leq H(\theta) - I(\theta, \xi_1, \dots, \xi_n) \leq U_n(V), \quad n = 1, 2, \dots$$

provided that (3) and other related assumptions hold and, moreover, both the bounds considered here are attainable, for any  $n = 1, 2, \dots$

We do not aim to discuss this relation explicitly here; it will be important for us only that on the base of such an estimate one can argue that (1) holds iff

$$(36) \quad 2 - V(P^n, Q^n) \approx \exp(-n D(P, Q))$$

for  $D = D(P, Q)$ , where

$$(37) \quad P^n = P \times P \dots \times P(n \text{ times}), \quad Q^n = Q \times Q \times \dots \times Q(n \text{ times}).$$

But, as it was shown in Th. 1 of [1], one can very easily show that (36) always holds for some  $D(P, Q)$ .

Unfortunately, these considerations do not yield that  $D(P, Q)$  figuring here satisfies (16) for every  $P, Q$ . However, this statement together with (36) has been proved firstly by H. Chernoff [3]. For the sake of completeness we next reproduce the proof of Chernoff in a slightly modified way using the definition (26) instead of (16) (cf. also Sanov [15]).

Let us suppose, firstly, that  $P = (p_1, p_2, \dots, p_s)$ ,  $Q = (q_1, q_2, \dots, q_s)$  are two discrete distributions, i.e. that

$$P[\xi = i \mid \theta = 1] = p_i, \quad P[\xi = i \mid \theta = 2] = q_i, \quad i = 1, 2, \dots, s,$$

(cf. (2)), where

$$\sum_{i=1}^s p_i = \sum_{i=1}^s q_i = 1.$$

It follows from (34) that

$$1 - \frac{1}{2}V(P^n, Q^n) = \sum_{j_1, j_2, \dots, j_s} \min [p(j_1, \dots, j_s), q(j_1, \dots, j_s)] = \\ = \sum_{A_n} p(j_1, \dots, j_s) + \sum_{B_n} q(j_1, \dots, j_s),$$

where

$$p(j_1, \dots, j_s) = \frac{n!}{\prod_{i=1}^s j_i!} \prod_{i=1}^s p_i^{j_i}, \quad q(j_1, \dots, j_s) = \frac{n!}{\prod_{i=1}^s j_i!} \prod_{i=1}^s q_i^{j_i}$$

and

$$A_n = \{j_1, j_2, \dots, j_s : j_i \geq 0, \sum_{i=1}^s j_i = n, \prod_{i=1}^s p_i^{j_i} \leq \prod_{i=1}^s q_i^{j_i}\}, \\ B_n = \{j_1, j_2, \dots, j_s : j_i \geq 0, \sum_{i=1}^s j_i = n, \prod_{i=1}^s p_i^{j_i} > \prod_{i=1}^s q_i^{j_i}\}.$$

Let us now denote by  $\mathcal{P} \cup \mathcal{Q}$  the set of all discrete probability distributions  $R = (r_1, r_2, \dots, r_s)$ , where  $\mathcal{P}, \mathcal{Q}$  are defined by (24), (25), and let  $\mathcal{R}' \subset \mathcal{P} \cup \mathcal{Q}$  be the set of all  $R$  such that for every  $r_i$  there exists an integer  $k$  such that  $r_i = k/n$ . Let us put  $\mathcal{P}' = \mathcal{P} \cap \mathcal{R}'$ ,  $\mathcal{Q}' = \mathcal{Q} \cap \mathcal{R}'$ . Clearly,

$$p(n\bar{r}_1, \dots, n\bar{r}_s) + q(n\tilde{r}_1, \dots, n\tilde{r}_s) \leq 1 - \frac{1}{2}V(P^n, Q^n) \leq \\ \leq \text{card}(A_n \cup B_n) [p(n\bar{r}_1, \dots, n\bar{r}_s) + q(n\tilde{r}_1, \dots, n\tilde{r}_s)],$$

320 where  $\bar{R} = (\bar{r}_1, \bar{r}_2, \dots, \bar{r}_s) \in \mathcal{P}'$ ,  $\bar{R} = (\bar{r}_1, \bar{r}_2, \dots, \bar{r}_s) \in \mathcal{Q}'$  are chosen according to

$$p(n\bar{r}_1, \dots, n\bar{r}_s) = \max_{R \in \mathcal{P}'} p(nr_1, \dots, nr_s),$$

$$q(n\bar{r}_1, \dots, n\bar{r}_s) = \max_{R \in \mathcal{Q}'} q(nr_1, \dots, nr_s),$$

and  $\tilde{R} = (\tilde{r}_1, \tilde{r}_2, \dots, \tilde{r}_s) \in \mathcal{P}'$ ,  $\tilde{R} = (\tilde{r}_1, \tilde{r}_2, \dots, \tilde{r}_s) \in \mathcal{Q}' - \mathcal{P}'$  are arbitrary. Since card  $(A_n \cup B_n) \leq n^s$  and since, according to the formula of Stirling,

$$p(nr_1, \dots, nr_s) = \exp(-n H(R, P) + o(n)),$$

$$q(nr_1, \dots, nr_s) = \exp(-n H(R, Q) + o(n)),$$

we can write

$$(39) \quad 1 - \frac{1}{2}V(P^n, Q^n) \leq \exp(-n D(P, Q) + o(n))$$

(cf. the inclusions  $\mathcal{P}' \subset \mathcal{P}$ ,  $\mathcal{Q}' \subset \mathcal{Q}$  and (26)). If  $R = (r_1, r_2, \dots, r_s) \in \mathcal{P} \cup \mathcal{Q}$  is such that

$$(40) \quad D(P, Q) = \min(H(R, P), H(R, Q))$$

then, in view of that one can always choose  $\bar{R} \in \mathcal{P}'$ ,  $\tilde{R} \in \mathcal{Q}' - \mathcal{P}'$  such that  $|\bar{r}_i - r_i| \leq 1/n$ ,  $|\tilde{r}_i - r_i| \leq 1/n$ ,  $i = 1, 2, \dots, s$ , i.e.

$$H(\bar{R}, P) \leq H(R, P) + o(n),$$

$$H(\tilde{R}, Q) \leq H(R, Q) + o(n),$$

the following inequality can be written

$$1 - \frac{1}{2}V(P^n, Q^n) \geq \exp(-n H(R, P) + o(n)) + \exp(-n H(R, Q) + o(n)).$$

Thus, by (40),

$$1 - \frac{1}{2}V(P^n, Q^n) \geq \exp(-n D(P, Q) + o(n)).$$

This together with (39) implies that (36) holds for  $D(P, Q)$  defined by (26) (and, consequently, by (16)) provided that  $P, Q$  are discrete distributions with a finite number of atoms.

To prove that (36) holds for  $D(P, Q)$  given by (16) (or (26)) for every  $P, Q$  let us denote by  $D^*(P, Q)$  the quantity figuring in the exponent of (36) in order to distinguish it from that given by (16) and denote, further, by  $P_s, Q_s$  restrictions of  $P, Q$  on a sub- $\sigma$ -algebra  $\mathcal{X}_s \subset \mathcal{X}$  generated by a decomposition of  $X$  consisting of  $s$  sets from  $\mathcal{X}$ . Since, evidently,  $V(P_s^n, Q_s^n) \leq V(P^n, Q^n)$  for every  $n = 1, 2, \dots$ , on the base of (36) and on the base of the result we have proved above one can argue that

$$(41) \quad D(P_s, Q_s) \leq D^*(P, Q).$$



As it is proved in [8], for every  $P, Q$  there exists a sequence  $\mathcal{X}_1 \subset \mathcal{X}_2 \subset \dots$  of sub- $\sigma$ -algebras such that the  $\sigma$ -algebra  $\mathcal{X}' \subset \mathcal{X}$  generated by

$$\bigcup_{s=1}^{\infty} \mathcal{X}_s$$

is sufficient with respect to  $P$  and  $Q$ , so that, by Th. 7 and 8,

$$\lim_s D(P_s, Q_s) = D(P, Q)$$

and, by (51),

$$(42) \quad D(P, Q) \leq D^*(P, Q).$$

To prove that the strict inequality cannot appear here we shall need the following fact, which follows from what we have already proved for discrete distributions and from results of Sec. 2: If  $P, Q$  are two discrete distributions, then

$$(43) \quad P^n(F_n) \approx \exp(-n D(P, Q))$$

if  $H'(Q, P) > 0$  (cf. (36)) and

$$(44) \quad 1 - \frac{1}{2}V(P^n, Q^n) = P^n(F_n) + Q^n(X^n - F_n)$$

(cf. (34)), where

$$(45) \quad F_n = \left\{ \prod_{i=1}^n p(x_i) \leq \prod_{i=1}^n q(x_i) \right\} \in \mathcal{X}^n.$$

Finally, we shall use the fact that  $1 - \frac{1}{2}V(P^n, Q^n) \approx \exp(-n D(P, Q))$  holds for  $D(P, Q)$  evaluated by (16) if we replace  $P, Q$  by arbitrary totally finite discrete measures  $(p_1, p_2, \dots, p_i), (q_1, q_2, \dots, q_i)$ . Indeed, in what precedes the norming conditions  $P(X) = Q(X) = 1$  never have been used.

Put  $p_i = P(E_i), q_i = p_i \exp(-\varepsilon i)$  for  $E_i = \{q \exp \varepsilon(i-1) < p < q \exp \varepsilon i\}$  for every  $i = 0, \pm 1, \pm 2, \dots$  and  $\varepsilon > 0$ . It is easy to see that  $q^2 < p^2 \exp \alpha \varepsilon(1-i)$  for  $x \in E_i, i = 0, \pm 1, \dots$  so that

$$H'_{1-\alpha}(P, Q) < \exp \varepsilon \alpha \sum_{i=-\infty}^{+\infty} p_i \exp(-\alpha \varepsilon i) \quad \text{for every } \alpha \in [0, 1].$$

Thus, for appropriately chosen  $\varepsilon$  and  $r$  we can write (cf. (16))

$$\sum_{i=-r}^r p_i \exp(-\alpha_* \varepsilon i) > \exp(-D(P, Q) - \delta),$$

where  $\delta > 0$  is an arbitrary number given in advance and  $\alpha_* \in [0, 1]$  is minimizing the sum

$$\sum_{i=-r}^r p_i \exp(-\alpha \varepsilon i) \quad \text{on } \alpha \in [0, 1],$$

322 i.e.

$$(46) \quad D(\bar{P}, \bar{Q}) \leq D(P, Q) + \delta,$$

where  $\bar{P}, \bar{Q}$  are totally finite (discrete) measures defined on by the following Radon-Nikodym densities (with respect to the dominating measure  $\mu$ ):

$$\bar{p}(x) = \frac{p_i}{\mu(E_i)}, \quad q_i(x) = \frac{q_i}{\mu(E_i)} \quad \text{for } x \in E_i, i = 0, \pm 1, \dots, \pm r,$$

and

$$\bar{p}(x) = 1, \quad \bar{q}(x) = 0 \quad \text{otherwise.}$$

It follows from the definition of  $E_i$  that  $p/q \leq p_i/q_i$  on  $E_i$  (if these ratios exist),  $i = 0, \pm 1, \dots$ , so that  $\bar{F}_n \subset F_n$  for

$$\bar{F}_n = \left\{ \prod_{i=1}^n \bar{p}(x_i) \leq \prod_{i=1}^n \bar{q}(x_i) \right\} \in \mathcal{X}^n$$

and for  $F_n$  defined by (45) and, consequently,

$$(47) \quad \bar{P}^n(F_n) = P^n(\bar{F}_n) \leq P^n(F_n).$$

In the case we have considered  $H(\bar{Q}, \bar{P}) > 0$ , so that, according to what was said in a remark above,

$$\bar{P}^n(\bar{F}_n) \approx \exp(-n D(\bar{P}, \bar{Q}))$$

(cf. (43)), i.e.

$$(48) \quad -\frac{1}{n} \log \bar{P}^n(\bar{F}_n) = D(\bar{P}, \bar{Q}) \leq D(P, Q) + \delta$$

(cf. (46)). On the other hand, taking into account (44) and (36), we can write

$$-\frac{1}{n} \log P^n(F_n) \geq D^*(P, Q).$$

This together with (47) and (48) yields the inequality

$$D^*(P, Q) \leq D(P, Q) + \delta.$$

Since  $\delta$  may be chosen arbitrarily small, the desired equality between  $D^*(P, Q)$  and  $D(P, Q)$  is proved.

We remark that A. Rényi, using a more accurate relation

$$2 - V(P^n, Q^n) = O\left(\frac{1}{\sqrt{n}} \exp[-n D(P, Q)]\right)$$

following from a more general result of R. R. Bahadur and R. Ranga Rao [16], stated in [2] the following sharpening of (1):

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$$I(\theta, \xi_1, \dots, \xi_n) = H(\theta) - O\left(\frac{1}{\sqrt{n}} \exp[-n D(P, Q)]\right).$$

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## O množství informace obsažené v posloupnosti nezávislých pozorování

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Nechť  $\theta$  značí náhodnou veličinu nabývající hodnot  $1, 2, \dots$  a  $\xi$  jinou náhodnou veličinu s měřitelným výběrovým prostorem  $(X, \mathcal{X})$ . Nechť dále  $\xi_1, \xi_2, \dots$  jsou postupné realizace veličiny  $\xi$ , o kterých budeme předpokládat, že jsou navzájem nezávislé pro každou hodnotu  $\theta$ . Nechť nakonec  $I(\theta, \xi_1, \dots, \xi_n)$  je množství Shannonovy informace o veličině  $\theta$  obsažené v  $(\xi_1, \xi_2, \dots, \xi_n)$ .

Je známo, že  $I(\theta, \xi_1, \dots, \xi_n) \in [0, H(\theta)]$ , kde  $H(\theta)$  je entropie veličiny  $\theta$ , a že informace nabývá hodnot  $0$  resp.  $H(\theta)$  právě když  $\theta$  a  $(\xi_1, \xi_2, \dots, \xi_n)$  jsou nezávislé resp. deterministicky závislé. Je tedy informace jakožto míra statistické závislosti mezi  $\theta$  a  $(\xi_1, \xi_2, \dots, \xi_n)$  důležitou číselnou charakteristikou statistického problému, který spočívá ve stanovení neznámé hodnoty parametru  $\theta$  pouze na základě znalosti hodnoty náhodného výběru  $(\xi_1, \xi_2, \dots, \xi_n)$ .

Poměrně velmi snadno (viz věta 1 v [1]) lze dokázat, že existuje parametr  $D \in [0, +\infty]$  závislý toliko na podmíněné distribuci  $P_{\xi|\theta}$  pro který platí vztah (1). Explicitní analytický výraz pro  $D$  byl nezávisle nalezen a současně publikován v referátech A. Rényiho [2] a autora [1]. Jak bylo možné intuitivně očekávat,  $D$  je totožné s tzv. Chernoffovou mezí [3], příslušnou Bayesovu testu ke stanovení správné hypotézy  $\theta = i$ ,  $i = 1, 2, \dots$  na základě  $(\xi_1, \xi_2, \dots, \xi_n)$ .

Předložená práce shrnuje vlastnosti parametru  $D$  odvozené v pracích [1, 2, 3] a dále je prohlubuje. Ve větách 5 až 7 a 9 a 10 jsou v poněkud zobecněné podobě shrnuty a dokázány ty vlastnosti parametru  $D$ , které v [1] byly vysloveny bez důkazu. Věty 1 až 4 stanoví vlastnosti modifikované  $\alpha$ -entropie a modifikované relativní Shannonovy entropie (nazývané též diskriminační informace). Obě modifikované entropie jsou ve většině případů totožné s nemoifikovanými, avšak v jistých rovněž velmi významných případech se tyto pojmy liší. Jejich zavedení umožňuje nejen formální zjednodušení úvah, ale poskytuje též možnost přesněji popsat a jemněji klasifikovat statistické problémy uvažovaného typu. Nakonec, ve větě 8 je stanovena jistá konvergenční vlastnost funkcionálu  $D$ , která neplyne přímo z konvergenčních vlastností  $\alpha$ -entropií.

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