

**AN INEQUALITY CONCERNING TESTS OF FIT OF THE
KOLMOGOROV-SMIRNOV TYPE**

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1. Introduction. Let $F_n(x)$ be the sumpolygon (empirical distribution function) of a sample of size n from a continuous distribution function $F(x)$. Let $K(x)$, $G_1(x)$, $G_2(x)$, $H_1(x)$, $H_2(x)$ be functions of x , such that for all x ,

$$G_1(x) \geq G_2(x); \quad H_2(x) \geq H_1(x).$$

The object of the present paper is to prove the following inequalities

- (1) $P[\inf_x (F_n - K) \geq 0 \mid \inf_x (G_1 - F_n) \geq 0, \inf_x (F_n - H_2) \geq 0] \\ \geq P[\inf_x (F_n - K) \geq 0 \mid \inf_x (G_2 - F_n) \geq 0, \inf_x (F_n - H_1) \geq 0],$
- (2) $P[\inf_x (K - F_n) \geq 0 \mid \inf_x (G_1 - F_n) \geq 0, \inf_x (F_n - H_2) \geq 0] \\ \leq P[\inf_x (K - F_n) \geq 0 \mid \inf_x (G_2 - F_n) \geq 0, \inf_x (F_n - H_1) \geq 0],$

where all probabilities are supposed to exist. Since these inequalities are symmetrical, it suffices to prove one of them.

These inequalities provide an approximation for the distribution of two-sided statistics of the Kolmogorov-Smirnov type. Such a distribution is written

$$P\{\sup_x n^{\frac{1}{2}}|F_n(x) - F(x)|\psi[F(x)] \leq \lambda\}$$

or more generally

$$(3) \quad P_n = P[\inf_x (G_2 - F_n) \geq 0, \inf_x (F_n - H_2) \geq 0].$$

In order to approximate P_n , take $H_1(x)$ and $H_2(x)$ in (1) smaller than zero for all x and replace $K(x)$ in (1) by $H_2(x)$; similarly take $G_1(x)$ and $G_2(x)$ in (2) larger than 1 for all x and replace $K(x)$ in (2) by $G_1(x)$. One then easily obtains the upper bound

$$(4) \quad P_n \leq P_n' P[\inf_x (G_2 - F_n) \geq 0] \cdot P[\inf_x (F_n - H_2) \geq 0] \\ \cdot \{P[\inf_x (G_1 - F_n) \geq 0] \cdot P[\inf_x (F_n - H_1) \geq 0]\}^{-1}$$

where $P_n' = P[\inf_x (G_1 - F_n) \geq 0, \inf_x (F_n - H_1) \geq 0]$. If now G_1 and H_1 are chosen close to G_2 resp. H_2 , but such that P_n' is more easily calculable than P_n , then (4) provides an interesting approximation of (3). A lower bound can be found in a similar way.

Wald and Wolfowitz [3] and [4] have given the following two bounds for P_n

$$(5) \quad P_n \leq P[\inf_x (G_2 - F_n) \geq 0] \cdot P[\inf_x (F_n - H_2) \geq 0], \\ P_n \geq P[\inf_x (G_2 - F_n) \geq 0] + P[\inf_x (F_n - H_2) \geq 0] - 1.$$

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However they did not prove (5) but only conjectured it. As a matter of fact (5) constitutes the particular form of (4) corresponding to $G_1(x) \geq 1$ and $H_1(x) \leq 0$ for all x . These bounds constitute good approximations for large values of P_n , which are the more interesting for purposes of testing. Furthermore they have the advantage of reducing the two-sided case to the one-sided case. On the other hand any other choice of $G_1(x)$ and $H_1(x)$ in (4) provides a closer upper bound than (5).

As Professor W. Hoeffding pointed out to the authors, (5) also is a consequence of a theorem of Lehmann [2]. Moreover he observed that the same theorem implies an analogous inequality for the distribution of the two-sample statistics of the Kolmogorov-Smirnov type. More details will be given in Section 4. Sections 2 and 3 are devoted to the proof of (1).

2. Lemma.

LEMMA. Let X be a many-dimensional random variable and S and T two measurable subsets of the range space of X . Let Y be a one-dimensional random variable. If for all y such that $a \leq y \leq b$

$$P[X \in S \mid (X \in T) \cap (Y = y)]$$

is a non-increasing function of y , then

$$P[X \in S \mid (X \in T) \cap (y_0 \leq Y \leq y_1)]$$

is a non-increasing function of y_0 and y_1 when $a \leq y_0 \leq y_1 \leq b$.

PROOF. Let (X', Y') have the conditional distribution of (X, Y) under the condition $X \in T$ and let

$$h(y) = P[X \in S \mid (X \in T) \cap (Y = y)] = P[X' \in S \mid Y' = y].$$

Then

$$\begin{aligned} (6) \quad & P[X \in S \mid (X \in T) \cap (y_0 \leq Y \leq y_1)] \\ &= P[X' \in S \mid y_0 \leq Y' \leq y_1] \\ &= \int_{y_0 \leq y \leq y_1} h(y) \cdot dF_{Y'}(y) / \int_{y_0 \leq y \leq y_1} dF_{Y'}(y) \end{aligned}$$

where $F_{Y'}$ denotes the distribution function of Y' . Under the assumption that $h(y)$ is non-increasing in $[a, b]$, it has to be shown that (6) is a non-increasing function of y_0 and y_1 for $a \leq y_0 \leq y_1 \leq b$, which is obviously true.

This proof has been suggested by Professor W. Hoeffding who pointed out that this lemma is closely related to Lemma 4 of Lehmann [2].

3. Proof of the inequality (1). Without loss of generality it will be supposed that $F(x) \equiv x$ for $0 \leq x \leq 1$. Let $X_1 \leq X_2 \leq \dots \leq X_n$ be the order statistics of a size n sample from $F(x)$. The requirement that

$$\inf_x [F_n(x) - K(x)] \geq 0$$

is equivalent to the requirement that every X_j should not be larger than some well defined number k_j , i.e. $X_j \leq k_j, j = 1, \dots, n$, with $0 \leq k_1 \leq k_2 \leq \dots \leq$

$k_n \leq 1$. Similarly $\inf_x [F_n(x) - H(x)] \geq 0$ is equivalent to $X_j \leq h_j, j = 1, \dots, n$, with $0 \leq h_1 \leq h_2 \leq \dots \leq h_n \leq 1$. On the other hand $\inf_x [G(x) - F_n(x)] \geq 0$ is equivalent to $X_j \geq g_j, j = 1, \dots, n$, with $0 \leq g_1 \leq g_2 \leq \dots \leq g_n \leq 1$. Proving the inequality (1) then is equivalent to proving the following theorem.

THEOREM. *If $g_j \leq h_j$ for all j , then*

$$P[\bigcap_{j=1}^n (X_j \leq k_j) \mid \bigcap_{j=1}^n (g_j \leq X_j \leq h_j)]$$

is a non-increasing function of $g_1, \dots, g_n, h_1, \dots, h_n$.

PROOF. The theorem holds if, for some $j, g_j > k_j$; in the sequel we suppose $g_j \leq k_j$ for all j .

The theorem is evidently true for sample size 1. For proving it in general we will proceed by complete induction, supposing it is true for sample sizes 1, 2, $\dots, n - 1$.

Let i be any integer in $[1, n]$. We introduce the following vectors;

$$\begin{aligned} X &= (X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n); \\ X' &= (X_1, \dots, X_{i-1}); \\ X'' &= (X_{i+1}, \dots, X_n); \\ k &= (k_1, \dots, k_{i-1}, k_{i+1}, \dots, k_n); \\ k' &= (k_1, \dots, k_{i-1}); \\ k'' &= (k_{i+1}, \dots, k_n); \\ c' &= (c, \dots, c) \quad (i - 1 \text{ components}); \\ c'' &= (c, \dots, c) \quad (n - i \text{ components}). \end{aligned}$$

In a similar way we define g, g', g'', h, h' and h'' . An inequality of two vectors will mean the simultaneous inequality of all corresponding components, e.g.: $X' \leq h'$ means

$$(X_1 \leq h_1) \cap (X_2 \leq h_2) \cap \dots \cap (X_{i-1} \leq h_{i-1}).$$

Let $g_{i-1} \leq c \leq h_{i+1}$ (it is understood that $g_0 = 0$ and $h_{n+1} = 1$) and

$$Q = P[X \leq k \mid (g \leq X \leq h) \cap (X_i = c)].$$

If $i \neq 1$ and $i \neq n$,

$$\begin{aligned} Q &= P[(X' \leq k') \cap (X'' \leq k'') \mid (g' \leq X' \leq h') \cap (g'' \leq X'' \leq h'') \cap (X_i = c)] \\ &= Q' \cdot Q'', \end{aligned}$$

with

$$\begin{aligned} Q' &= P[X' \leq k' \mid (g' \leq X' \leq h') \cap (g'' \leq X'' \leq h'') \cap (X_i = c)], \\ Q'' &= P[X'' \leq k'' \mid (X' \leq k') \cap (g' \leq X' \leq h') \cap (g'' \leq X'' \leq h'') \cap (X_i = c)]. \end{aligned}$$

Since $g_j \leq k_j$ for all j , Q'' is defined. Since X' and X'' are independent under the hypothesis $X_i = c$ (see e.g. Hajos and Rényi [1]), one has

$$Q' = P[X' \leq k' \mid (g' \leq X' \leq h') \cap (X_i = c)],$$

$$Q'' = P[X'' \leq k'' \mid (g'' \leq X'' \leq h'') \cap (X_i = c)].$$

Let $Y' = (Y_1, Y_2, \dots, Y_{i-1})$ be the order statistic of a size $(i - 1)$ sample from a rectangular distribution within $[0, 1]$. Under the hypothesis $Y' \leq c'$, Y' is distributed as the order statistic from a rectangular distribution within $[0, c]$. The same may be said of X' under the hypothesis $X_i = c$ (see e.g. Hajos and Rényi [1]). We thus have

$$Q' = P(Y' \leq k' \mid (g' \leq Y' \leq h') \cap (Y' \leq c')).$$

Similarly let Y'' be the order statistic of a size $(n - i)$ sample from a rectangular distribution within $[0, 1]$. We then obtain

$$Q'' = P[Y'' \leq k'' \mid (g'' \leq Y'' \leq h'') \cap (Y'' \geq c'')].$$

According to the present theorem applied to sample sizes smaller than n , Q' and Q'' are non-increasing functions of c when $g_{i-1} \leq c \leq h_{i+1}$. If $i = 1$ or $i = n$, we arrive easily at the same conclusion. Thus the same holds for Q .

Let

$$R = P[X \leq k \mid (g \leq X \leq h) \cap (g_i \leq X_i \leq h_i)].$$

According to the lemma, R is a non-increasing function of g_i and h_i when $g_{i-1} \leq g_i \leq h_i \leq h_{i+1}$. Let now

$$S = P[X_i \leq k_i \mid (X \leq k) \cap (g \leq X \leq h) \cap (g_i \leq X_i \leq h_i)],$$

$$T = P[(X \leq k) \cap (X_i \leq k_i) \mid (g \leq X \leq h) \cap (g_i \leq X_i \leq h_i)].$$

The probability S evidently is a non-increasing function of g_i and h_i . Since $T = R \cdot S$, T has the same property. As i is any integer in $[1, n]$, the theorem is proved.

4. Two corollaries of a theorem of Lehmann (Hoeffding). Two real-valued functions r and s of n arguments are called by Lehmann discordant for the i th coordinate if, considered as functions of the i th coordinate (with all other coordinates held fixed), they are monotone in opposite directions, i.e. either r non-decreasing and s non-increasing or the inverse.

Let $(T_1, U_1), \dots, (T_n, U_n)$ be independent pairs of random variables such that

$$(7) \quad P(T_i \leq t, U_i \leq u) \geq P(T_i \leq t) \cdot P(U_i \leq u)$$

for all u and t , and let

$$T = r(T_1, \dots, T_n), \quad U = s(U_1, \dots, U_n).$$

Lehmann proves that, if r and s are discordant for all i , then

$$P(T \geq t, U \geq u) \leq P(T \geq t) \cdot P(U \geq u)$$

for all u and t .

We now show that this theorem implies inequality (5). Let T_1, \dots, T_n be n independent random variables and let $F_n(x)$ be their sumpolygon. Let $U_i = T_i$. Define

$$T = r(T_1, \dots, T_n) = \inf_x [F_n(x) - H(x)],$$

$$U = s(U_1, \dots, U_n) = \inf_x [G(x) - F_n(x)],$$

where $G(x)$ and $H(x)$ are arbitrary functions of x . Then condition (7) is fulfilled and r and s are discordant for all i . This proves inequality (5) even in the case of discontinuous variables.

The same theorem of Lehmann implies an analogous inequality for the two-sample case. Let T_1, \dots, T_k be independent random variables with sumpolygon $F_1(x)$ and let T_{k+1}, \dots, T_n be independent random variables with sumpolygon $F_2(x)$. Let $U_i = T_i$ and define

$$D(x) = F_1(x) - F_2(x),$$

$$T = r(T_1, \dots, T_n) = \inf_x [D(x) - H(x)],$$

$$U = s(U_1, \dots, U_n) = \inf_x [G(x) - D(x)],$$

where $G(x)$ and $H(x)$ are arbitrary functions of x . Again condition (7) is fulfilled and r and s are discordant for all i . In particular when $G(x) \equiv -H(x) = a > 0$, one has

$$P[\sup_x |D(x)| \leq a] \leq \{P[\sup_x D(x) \leq a]\}^2.$$

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REFERENCES

- [1] HAJOS, G. and RÉNYI, A. (1954). Elementary proofs of some basic facts concerning order statistics. *Acta Math. Acad. Sci. Hungary* **5** 1-6.
- [2] LEHMANN, E. L. (1966). Some concepts of dependence. *Ann. Math. Statist.* **37** 1137-1153.
- [3] WALD, A. and WOLFOWITZ, J. (1939). Confidence limits for continuous distribution functions. *Ann. Math. Statist.* **10** 105-118.
- [4] WALD, A. and WOLFOWITZ, J. (1941). Note on confidence limits for continuous distribution functions. *Ann. Math. Statist.* **12** 118-119.