

Stochastically Ordered Distributions and Monotonicity of the OC-Function of Sequential Probability Ratio Tests

LUDGER RÜSCHENDORF¹

Summary. Some characterizations for the stochastic ordering of probability distributions are given. Especially a general sufficient condition for stochastic ordering is proved and the question of existence of upper and lower bounds in the class of all distributions with given marginals is considered. Stochastic ordering is applied to prove a general theorem on monotonicity of the OC-function of sequential probability ratio tests in stochastic processes.

Key words: Stochastic ordering, monotone functions, Rosenblatt transformation, OC-function, sequential probability ratio tests.

1. Introduction

Let (E, \mathfrak{A}) be a polish space supplied with the Borel σ -Algebra \mathfrak{A} and let \leq be a closed partial order on E . The set

$$M := \{f: (E, \mathfrak{A}) \rightarrow (\mathbb{R}^1, \mathfrak{B}^1); \quad f \text{ isotone and bounded}\}$$

defines a partial order \leq_M on the set of all probability measures on (E, \mathfrak{A}) ([14]) by $P_1 \leq_M P_2$ if and only if,

$$\int f dP_1 \leq \int f dP_2 \quad \text{for all } f \in M. \quad (1)$$

In section 2 of this paper we prove some characterizations of \leq_M . Especially we give a sufficient condition for \leq_M in the case of $(E, \mathfrak{A}) = (\mathbb{R}^n, \mathfrak{B}^n)$ which weakens the known sufficient conditions due to VEINOTT [21], FRANKEN, KIRSTEIN [7], KAMAE, KRENGEL and O'BRIEN [14], KALMYKOV [13], STOYAN [20] and FRANKEN, STOYAN [6]. There exist no lower and upper bounds w.r.t. \leq_M in the set of probability measures with given marginals. In contrast to this probability measures with ordered marginals are stochastically ordered.

Stochastic ordering is applied in the second part of this paper in order to prove monotonicity of the OC-function of sequential probability ratio tests. This connection was considered the first time by LEHMANN [15]. We get a general result, which includes the results known for this question. The methods are general enough in order to be applied to sequential probability ratio tests in stochastic processes. A connection to GIRSANOV's [11] Theorem on the representation of measures in function spaces continuous w.r.t. a given measure is indicated.

¹ Institut für Statistik und Wirtschaftsmathematik der RWTH, Pontstr. 51, 5100 Aachen, BRD.

2. Stochastic ordering

Let P_1, P_2 be probability measures on (E, \mathfrak{A}) . An element $A \in \mathfrak{A}$ is called isotone if $1_A \in M$. A characterization of \leq_M due to STRASSEN (cp. [14]) and KAMAE, KRENGEL and O'BRIEN [14], Th. 1, says:

$$P_1 \leq_M P_2 \quad (2)$$

\Leftrightarrow There exist E -valued random variables X_1, X_2 with distributions P_1, P_2 such that $X_1 \leq X_2$ a.s.

\Leftrightarrow (1) holds for the indicator functions of all isotone, closed sets $A \in \mathfrak{A}$.

Remark 1. a) The first equivalence implies for $(E, \mathfrak{A}) = (\mathbb{R}^n, \mathfrak{B}^n)$, that $F_1 \geq F_2$ and $h_{F_1} \leq h_{F_2}$, whenever $P_1 \leq_M P_2$, where F_i are the distribution functions of P_i and $h_{F_i}(x) := P_i([x, \infty))$, $x \in \mathbb{R}^n$, are the survival functions of P_i , $i=1, 2$. For $n=1$ also the converse is true; for $n \geq 2$ simple examples show, that the converse is false.

b) The counterexample given by LEHMANN [15], Ex. 2.2 for the first equivalence in (2) is false, since for the set $S := (B \cup D \cup A)_+$ from LEHMANN'S example (with $X_+ := \{y; \exists x \in X, x \leq y\}$) it holds true that $P_\theta(S) = \frac{15}{16}$ while $P_{\theta'}(S) = \frac{14}{16}$ and, therefore, it is not true that $P_\theta \leq_M P_{\theta'}$ in contrast to LEHMANN'S statement. Example 2.2 of LEHMANN [15] really shows that a multivariate monotone likelihood ratio does not imply stochastic ordering.

The following characterization of \leq_M allows to reduce or to enlarge the class M in order to prove (1). Especially it gives a connection to weak convergence. We assume that for isotone sets B and $x \leq y$, $d(x, B) \leq d(y, B)$, where d denotes a metric on E inducing the topology. This is fulfilled if e.g. E is a vector space, \leq is consistent with addition and d is invariant under addition.

Theorem 1. *The following statements are equivalent.*

- a) $P_1 \leq_M P_2$.
- b) (1) holds for isotone functions f with $f(x) \in \{0, 1\}$, $\forall x \in E$.
- c) (1) holds for all isotone functions f , such that the integrals exist.
- d) (1) holds for all continuous, bounded, isotone functions $f \geq 0$.

Proof. The equivalence of a) and b) follows from (2). We show that

$$b) \Rightarrow c). \text{ Let } f \text{ be integrable w.r.t. } P_1, P_2.$$

By means of a well known integration formula we get for E -valued random variables X, Y on (X, \mathfrak{B}, P) with $P^X = P_1, P^Y = P_2$

$$Ef(X) = \int_0^\infty P(f(X) \geq t) dt - \int_{-\infty}^0 P(f(X) \leq t) dt.$$

But $\{x; f(x) \geq t\}$ and $\{x; f(x) > t\}$ are isotone sets since f is isotone. Therefore b) implies

$$P(f(X) \geq t) \leq P(f(Y) \geq t)$$

and

$$P(f(X) > t) \leq P(f(Y) > t), \quad \forall t \in \mathbb{R}^1.$$

This implies

$$\int f dP_1 = Ef(X) \leq Ef(Y) = \int f dP_2.$$

It remains to show:

d) \Rightarrow b). We proof at first, that d) implies (1) for isotone functions f , continuous from the right with $f(x) \in \{0, 1\}$, $\forall x$.

Let $A := \{x; f(x) = 1\}$ and let $d(t, A) := \inf \{d(t, y); y \in A\}$, $t \in E$, where d denotes a metric on E which induces the topology on E . The functions

$$u^{(k)}(t) := \begin{cases} 0 & d(t, A) \geq \frac{1}{k} \\ 1 - kd(t, A) & d(t, A) < \frac{1}{k} \end{cases}$$

$k \in \mathbb{N}$, are bounded, continuous, isotone (since A is isotone) and ≥ 0 . Right continuity of f implies, that $u^{(k)}$ converges antitone to f . The theorem of monotone convergence implies

$$\int u^{(k)} dP_i \rightarrow \int f dP_i \quad i = 1, 2$$

and, therefore, $\int f dP_1 \leq \int f dP_2$ for all binary, right continuous functions f .

Let now in the second step f be an arbitrary isotone function and define again $A := \{x; f(x) = 1\}$. Then there exists a compact subset $K \subset A$ with

$$P_1(K) + \varepsilon \geq P_1(A),$$

since by Th. 1.4. of BILLINGSLEY [1] each probability measure on a complete separable metric space is tight. Let now K_+ be the isotonic closure of K , then I_{K_+} is isotone, right continuous and $K \subset K_+ \subset A$. The first step of this proof implies $P_1(K_+) \leq P_2(K_+)$ and, therefore,

$$P_1(A) \leq P_1(K) + \varepsilon \leq P_1(K_+) + \varepsilon \leq P_2(K_+) + \varepsilon \leq P_2(A) + \varepsilon,$$

which implies the conclusion since ε is arbitrary.

Remark 2. Theorem 1 implies Proposition 3 of KAMAE, KRENGEL and O'BRIEN [14] which states that \leq_M is a closed partial order on the set of all probability measures with the weak convergence topology. To show this let $(P_n), (Q_n)$ be sequences of probability measures on a polish space with closed order such that

$$P_n \xrightarrow{\mathfrak{D}} P, \quad Q_n \xrightarrow{\mathfrak{D}} Q \quad (\xrightarrow{\mathfrak{D}} \text{ weak convergence})$$

and assume $P_n \leq_M Q_n$, $\forall n \in \mathbb{N}$. Then by definition of weak convergence $\int f dP \leq \int f dQ$ for all continuous, bounded, isotone functions $f \geq 0$ which implies by Theorem 1 that $P \leq_M Q$ holds.

Theorem 1 also implies Proposition 2 of [14]. To show this let P, Q be probability measures on E^∞ and let $P^{(i)}, Q^{(i)}$ be the marginals of P, Q on E^i the i -fold product space. Assume that $P^{(i)} \leq_M Q^{(i)}, i \in \mathbb{N}$, and let $z \in E$ and $z_\infty := (z, z, \dots) \in E^\infty$. Then

$$P_{(i)} := P^{(i)} \otimes \varepsilon_{z_\infty} \leq_M Q^{(i)} \otimes \varepsilon_{z_\infty} =: Q_{(i)}$$

for all $i \in \mathbb{N}$, where ε_{z_∞} is the one point measure in z_∞ and, furthermore, $P_{(i)} \xrightarrow{\otimes} P, Q_{(i)} \xrightarrow{\otimes} Q$. So by the first part of this remark $P \leq_M Q$.

In this way we get a proof of Propositions 2, 3 of [14] without referring to the involved pointwise equivalence theorem of STRASSEN (cf. (2)) as is done in [14]. A further useful aspect of Theorem 1 is that it allows to apply results of the theory of balayage — which are generally proven for compact cones of continuous functions — to stochastic ordering. So e.g. the pointwise equivalence theorem of STRASSEN (cf. (2)) for probabilities on compact sets can be implied in this way by a general theorem of the theory of balayage.

Some further statements in [14] show, that in order to prove \leq_M in function spaces in many cases (e.g. SKOROHOD space, space of continuous functions) it is enough to show \leq_M for the finite dimensional distributions. Therefore, we want to consider in the following the special case $E = [0, 1]^n$ (the case $E = \mathbb{R}^n$ is implied by application of isotone transformations).

The proof of the following lemma follows by approximation on the lattice

$$G_m := \{(i_1/2^m, \dots, i_n/2^m); \quad 0 \leq i_j \leq 2^m, 1 \leq j \leq n\}.$$

Lemma 2. *Let $f: E \rightarrow \mathbb{R}^1$ be isotone and right continuous, then there exist sequences $t_{i,m} \in [0, 1]^n, \alpha, \alpha_{i,m} \in \mathbb{R}^1, \alpha_{i,m} \geq 0, m', k_{i,m} \in \mathbb{N}, 1 \leq i \leq m, m \in \mathbb{N}$, such that*

$$f_m(x) = \alpha + \sum_{i=1}^{m'} \alpha_{i,m} I_A(x) \quad (3)$$

where

$$A = \bigcup_{j=1}^{k_{i,m}} [t_{i_j,m}, 1]$$

converges monotonically to f .

Let now F, G be the distribution functions (df's) of P_1, P_2 and let X, Y be E -valued random variables with df's F, G then we define $F \leq_M G$ if $P_1 \leq_M P_2$. Furthermore, for $\lambda_i \in [0, 1], 1 \leq i \leq n$, and $x = (x_1, \dots, x_n) \in E$ we define

$$\begin{aligned} F(x_1, \lambda_1) &:= P(X_1 < x_1) + \lambda_1 P(X_1 = x_1) \\ F(x_i, \lambda_i | x_1, \dots, x_{i-1}) &:= P(X_i < x_i | X_j = x_j, 1 \leq j \leq i-1) \\ &+ \lambda_i P(X_i = x_i | X_j = x_j, j \leq i-1), \quad 2 \leq i \leq n, \end{aligned}$$

and $\tau_F: [0, 1]^n \times [0, 1]^n \rightarrow [0, 1]^n$ by

$$\tau_F(x, \lambda) := (F(x_1, \lambda_1), F(x_2, \lambda_2 | x_2), \dots, F(x_n, \lambda_n | x_1, \dots, x_{n-1})). \quad (4)$$

For $\tau_F(x, 1, \dots, 1)$ we use the abbreviation $\tau_F(x)$. τ_F is a modification of a transformation introduced by ROSENBLATT [18]. The inverse of τ_F is defined inductively

by

$$\begin{aligned}\tau_F^{-1}(x) &= (z_1, \dots, z_n) \quad \text{with} \\ z_1 &:= F^{-1}(x_1) = \inf \{y \in \mathbb{R}^1; F(y) \cong x_1\}\end{aligned}$$

(where $F(y)$ denotes the first marginal df. of F applied to y .)

$$\begin{aligned}z_2 &:= \inf \{y; F(y | z_1) \cong x_2\} = F^{-1}(x_2 | z_1) \\ z_n &:= F^{-1}(x_n | z_1, \dots, z_{n-1}).\end{aligned}$$

Similarly we define τ_G, τ_G^{-1} . Let U_1, \dots, U_n be $R(0, 1)$ -distributed random variables such that $\{U_1, \dots, U_n, X, Y\}$ are independent, then with $U = (U_1, \dots, U_n)$ we get the following

Lemma 3. a) $\tau_F(X, U)$ is a random variable whose distribution is the product of n independent $R(0, 1)$ -distributions.

b) $\tau_F^{-1}(U)$ is a random variable with df. F .

Remark (to the proof). In the special case of absolutely continuous df's part a) has been shown by ROSENBLATT [18]. The general case can be proved by successive application of the idea of Lemma 1, p. 216 of FERGUSON [5] to all conditional df's. b) follows from an integral transformation. In a slight modified form b) is stated in Theorem 1 of O'BRIEN [17].

Let P_1, P_2, F, G be defined as above, then we get

Theorem 4. a) $P_1 \leq_M P_2$ if and only if

$$\begin{aligned}\lambda_n \left\{ x \in [0, 1]^n; \tau_F^{-1} \in \bigcup_{j=1}^k [t_j, 1] \right\} &\leq \lambda_n \left\{ x \in [0, 1]^n; \tau_G^{-1} \in \bigcup_{j=1}^k [t_j, 1] \right\}, \\ \forall t_j \in [0, 1]^n, \forall k \in \mathbb{N}.\end{aligned}$$

$$\text{b) } \tau_F^{-1} \leq \tau_G^{-1} [\lambda_n]. \quad (5)$$

implies $P_1 \leq_M P_2$.

Proof. a) By Theorem 1 it is sufficient to consider bounded continuous functions $f \geq 0$ in \mathcal{M} . These functions can be approximated isototonically by Lemma 2 by functions of the type

$$\alpha + \sum_{i=1}^{m'} \alpha_{i,m} I_A(x), \quad \text{where } A = \bigcup_{j=1}^{k_{i,m}} [t_{i,j,m}, 1]$$

From Lemma 3, b) follows statement a).

b) Let U_1, \dots, U_n be stochastically independent, $R(0, 1)$ -distributed, $U := (U_1, \dots, U_n)$. Then $X := \tau_F^{-1}(U) \leq Y := \tau_G^{-1}(U) [\lambda_n]$. So b) follows from Lemma 3, b). b) is also immediate from a).

The following lemma gives 2 conditions which imply condition (b).

Lemma 5. *Let F, G be n dim. df's.*

a) *From $G(x_1) \leq F(x_1)$, $\forall x_1$ and $G(x_{i+1} | x_1, \dots, x_i) \leq F(x_{i+1} | y_1, \dots, y_i)$, $\forall (x_1, \dots, x_i) \equiv (y_1, \dots, y_i)$ follows*

$$\tau_F^{-1} \leq \tau_G^{-1}.$$

b) *If $\tau_G \leq \tau_F$ and if $G(x_{i+1} | x_1, \dots, x_i)$ is antitone in (x_1, \dots, x_i) , $i \leq n-1$, then*

$$\tau_F^{-1} \leq \tau_G^{-1}.$$

Proof. Define $\tau_F^{-1}(x_1, \dots, x_n) = : (z_1, \dots, z_n)$ and $\tau_G^{-1}(x_1, \dots, x_n) = : (w_1, \dots, w_n)$.

a) The first assumption in a) implies $z_1 \leq w_1$. If $(z_1, \dots, z_i) \leq (w_1, \dots, w_i)$, then

$$\begin{aligned} G^{-1}(x_{i+1} | w_1, \dots, w_i) &= \inf \{y \in \mathbb{R}^1; G(y | w_1, \dots, w_i) \geq x_{i+1}\} \\ &\geq \inf \{y; F(y | z_1, \dots, z_i) \geq x_{i+1}\} = F^{-1}(x_{i+1} | z_1, \dots, z_i). \end{aligned}$$

This implies $\tau_F^{-1} \leq \tau_G^{-1}$.

b) If $G(x_{i+1} | x_1, \dots, x_i)$ is antitone in (x_1, \dots, x_i) and $\tau_G \leq \tau_F$, then

$$\begin{aligned} G(x_{i+1} | x_1, \dots, x_i) &\leq G(x_{i+1} | y_1, \dots, y_i) \leq F(x_{i+1} | y_1, \dots, y_i), \\ \forall (y_1, \dots, y_i) &\equiv (x_1, \dots, x_i). \end{aligned}$$

So b) is implied by a).

Remark 3. a) From monotonic transformation it follows that Theorem 4 is true also for $E = \mathbb{R}^n$.

b) KAMAE, KRENGEL and O'BRIEN [14], Theorem 2, FRANKEN, KIRSTEIN [7], Satz 4.1, and VEINOTT [21], Theorem 4 prove stochastic ordering under the conditions a) resp. b) of Lemma 5. So these results are included in Theorem 4.

By Remark 1, a) stochastic ordering on \mathbb{R}^n implies ordering of the df's. KALMYKOV [13], STOYAN [6], Satz 5 (in the case of Markov processes) and O'BRIEN [17], Theorem 4 (in the non markovian case) prove ordering of the df's under the condition b) of Lemma 5. So their results are also included in Theorem 4.

For some different types of comparison of Markov processes cf. FRANKEN, STOYAN [6], STOYAN [20] and FRANKEN, KIRSTEIN [7].

Let F_1, \dots, F_n be n one-dimensional df's and let $\mathcal{H}(F_1, \dots, F_n)$ denote the set of all n -dim. df's with F_i as i -th marginal df. $\mathcal{H}(F_1, \dots, F_n)$ is of statistical interest since it describes the influence of dependence on a statistical problem. A characterization of $\mathcal{H}(F_1, \dots, F_n)$ by means of ordering the df's gives the Theorem on the Fréchet bounds from DALL'AGLIO [2].

A df. H is an element of $\mathcal{H}(F_1, \dots, F_n)$ if and only if

$$H_1(x) \leq H(x) \leq H_2(x), \quad \forall x \in \mathbb{R}^n,$$

where

$$H_2(x) := \min_{1 \leq i \leq n} F_i(x_i)$$

and

$$H_1(x) := \left(\sum_{i=1}^n F_i(x_i) - (n-1) \right)_+$$

for $x = (x_1, \dots, x_n)$ with $a_+ = \max\{a, 0\}$, $a \in \mathbb{R}^1$. H_2 is an element of $\mathcal{H}(F_1, \dots, F_n)$ while $H_1 \in \mathcal{H}(F_1, \dots, F_n)$ only for $n=2$ and in few special cases for $n \geq 3$. In all other cases there does not exist a lower bound in the set $\mathcal{H}(F_1, \dots, F_n)$ with respect to the pointwise ordering.

This implies by Remark 1. a) that generally for $n \geq 3$ there does not exist a largest element in $\mathcal{H}(F_1, \dots, F_n)$ w.r.t. stochastic ordering. The following example considers the question of stochastic ordering between two elements of $\mathcal{H}(F_1, \dots, F_n)$.

Example 1. a) Let $F_1 = F_2$ be the df. of the 2-point distribution $\frac{1}{2}(\varepsilon_{\{0\}} + \varepsilon_{\{1\}})$. By a simple calculation we get:

$$\mathcal{H}(F_1, F_2) = \left\{ F^\alpha; F^\alpha \text{ is df. of} \right. \\ \left. P^\alpha := \left(\frac{1}{2} - \alpha \right) (\varepsilon_{(0,1)} + \varepsilon_{(1,0)}) + \alpha (\varepsilon_{(0,0)} + \varepsilon_{(1,1)}), 0 \leq \alpha \leq \frac{1}{2} \right\}.$$

For $0 \leq \alpha_1 \leq \alpha_2 \leq \frac{1}{2}$ we have $H^{\alpha_1} \leq H^{\alpha_2}$ and $h_{H^{\alpha_1}} \leq h_{H^{\alpha_2}}$ and $H^0 = H_1$, $H^{\frac{1}{2}} = H_2$ (H_1, H_2 from DALL'AGLIO's Theorem). Let f_1, f_2 be isotone functions with $f_1(0, 0) = 0$, $f_1(1, 0) = f_1(0, 1) = f_1(1, 1) = 1$ and $f_2(0, 0) = f_2(0, 1) = f_2(1, 0) = 0$, $f_2(1, 1) = 1$. Then $\int f_1 dP^\alpha = \frac{1}{2} - \alpha$ and $\int f_2 dP^\alpha = \alpha$, so that for $0 \leq \alpha_1 < \alpha_2 \leq \frac{1}{2}$

$$\text{neither } H^{\alpha_1} \leq_M H^{\alpha_2} \text{ nor } H^{\alpha_2} \leq_M H^{\alpha_1}.$$

So $\mathcal{H}(F_1, F_2)$ is totally disordered w.r.t. \leq_M .

b) Let F_1, \dots, F_n be df's and define

$$g(x) := (g_1(x), \dots, g_n(x)), h(x) := (h_1(x), \dots, h_n(x))$$

by $h_1(x) = g_1(x) = F_1(x_1)$ and

$$h_i(x) := I_{\left[\min_{j \leq i-1} F_j(x_j), 1 \right]}(F_i(x_i)) \quad (6)$$

$$g_i(x) := I_{\left[\left(i-1 - \sum_{j=1}^{i-1} F_j(x_j) \right) \wedge 1, 1 \right]}(F_i(x_i)), \quad 2 \leq i \leq n.$$

Then $h = \tau_{H_2}$ and $g = \tau_{H_1}$ (resp. solution of the corresponding integral equation if H_1 is no df.). $h \leq g$ or $g \leq h$ holds true only in the case of one point distributions. This example leads to the following result.

Proposition 6. *There do not exist two elements in $\mathcal{H}(F_1, \dots, F_n)$ which are comparable w.r.t. \leq_M .*

Proof. Let $F, H \in \mathcal{H}(F_1, \dots, F_n)$ and assume $F \leq_M H$. Then by the pointwise equivalence theorem (2) there exist random variables X, Y on a common pro-

bability space with df's F, H such that $X \leq Y$ a.s. Therefore, $X_i \leq Y_i$ a.s. where X_i, Y_i are the i -th components of X, Y . Assume that $P(X_i < Y_i) > 0$. Then since $\{X_i < Y_i\} = \bigcup_{a \in Q} \{X_i < a < Y_i\}$ where Q is the set of all rational numbers there exists an element $a \in Q$ with $P(X_i < a < Y_i) > 0$ and, therefore,

$$\begin{aligned} P(X_i < a) &= P(X_i < a, X_i \leq Y_i) = P(X_i \leq Y_i < a) + P(X_i < a \leq Y_i) \\ &> P(X_i \leq Y_i < a) = P(Y_i < a) \end{aligned}$$

contradicting the assumption that both X_i and Y_i have the same df. F_i . Therefore, $X_i = Y_i$ a.s. This implies $X = Y$ a.s. and, therefore, $F = H$.

Distributions whose marginals are ordered are comparable w.r.t. \leq_M in the following sense:

Proposition 7. Let $F_i, G_i, 1 \leq i \leq n$ be one-dim. df's with $F_i \leq G_i$. Then it holds:

- a) To $F \in \mathcal{H}(F_1, \dots, F_n)$ exists a $G \in \mathcal{H}(G_1, \dots, G_n)$ with $G \leq_M F$.
- b) To $G \in \mathcal{H}(G_1, \dots, G_n)$ there exists a $F \in \mathcal{H}(F_1, \dots, F_n)$ with $G \leq_M F$.

Proof. a) Let $X = (X_1, \dots, X_n)$ be a random variable with df. F and let U_1, \dots, U_n be stochastically independent $R(0, 1)$ -distributed random variables independent of X . Define

$$Y := (G_1^{-1}(F_1(X_1, U_1)), \dots, G_n^{-1}(F_n(X_n, U_n))). \quad (7)$$

Lemma 3 (for $n=1$) implies $F_Y \in \mathcal{H}(G_1, \dots, G_n)$. $F_i(X_i, U_i) \leq F_i(X_i)$ implies

$$G_i^{-1} \circ F_i(X_i, U_i) \leq G_i^{-1} \circ F_i(X_i) \leq F_i^{-1} \circ F_i(X_i) \leq X_i$$

and, therefore, $Y \leq X$. So with $G := F_Y$ we have $G \leq_M F$.

b) Let X be a random variable with df. $G \in \mathcal{H}(G_1, \dots, G_n)$, then $-X = (-X_1, \dots, -X_n)$ is a random variable with df. in $\mathcal{H}(1 - G_1((-\cdot) -), \dots, 1 - G_n((-\cdot) -))$ and $1 - G_i((-\cdot) -) \leq 1 - F_i((-\cdot) -)$. Therefore, by a) there is a random variable $Y = (Y_1, \dots, Y_n)$ with df. in $\mathcal{H}(1 - F_1((-\cdot) -), \dots, 1 - F_n((-\cdot) -))$ such that $Y \leq -X$ so that $-Y \geq X$. The df. F of $-Y$ is an element of $\mathcal{H}(F_1, \dots, F_n)$ and it holds that $G \leq_M F$.

3. Monotonicity of the OC-functions of SPRT's

Let $(X_i)_{i \in \mathbb{N}}$ be a sequence of real random variables, let $P_\vartheta, \vartheta \in \Theta \subset \mathbb{R}^k$ be probability measures such that $X_{(n)} := (X_1, \dots, X_n)$ have under P_ϑ densities $f_\vartheta^n(x)$ w.r.t. a sequence μ_n of σ -finite measures on $(\mathbb{R}^n, \mathfrak{B}^n)$. Let further $\vartheta_0, \vartheta_1 \in \Theta$, $\vartheta_0 < \vartheta_1$ and define the likelihood ratio

$$Z_n := f_{\vartheta_1}^n(X_{(n)}) / f_{\vartheta_0}^n(X_{(n)}). \quad (8)$$

For boundary sequences $a_i, b_i \in \mathbb{R}^1, a_i < b_i$ we define the stopping time

$$N(w) := \begin{cases} \inf \{n \in \mathbb{N}; Z_n(w) \notin (a_n, b_n)\} \\ \infty \text{ if } \{n \in \mathbb{N}; Z_n(w) \notin (a_n, b_n)\} = \emptyset \end{cases}$$

$(Z_n, n \in \mathbb{N}, N)$ is a SPRT. The OC-function is determined by

$$Q_n(\vartheta) := P_\vartheta(N \leq n, Z_N \leq a_N)$$

and

$$P_n(\vartheta) := P_\vartheta(N \leq n, Z_N \geq b_N).$$

We are interested in conditions on $P_\vartheta, \vartheta \in \Theta$ such that for $\vartheta \leq \vartheta'$:

$$Q_n(\vartheta) \geq Q_n(\vartheta') \quad \text{and} \quad P_n(\vartheta) \leq P_n(\vartheta'), \quad \forall n \in \mathbb{N}. \quad (9)$$

Theorem 8. *The condition $P_\vartheta^{(Z_1, \dots, Z_n)} \leq_M P_{\vartheta'}^{(Z_1, \dots, Z_n)}$ for $\vartheta \leq \vartheta'$ and all $n \in \mathbb{N}$ implies (9).*

Proof. The set

$$M_n := \{(z_1, \dots, z_n) \in \mathbb{R}^n; \exists i \leq n, z_i \notin (a_i, b_i),$$

and $z_{i_0} \geq b_{i_0}$ for $i_0 := \min \{i; z_i \notin (a_i, b_i)\}\}$

is monotone. So the condition on (Z_1, \dots, Z_n) implies

$$P_n(\vartheta) = P_\vartheta^{(Z_1, \dots, Z_n)}(M_n) \leq P_{\vartheta'}^{(Z_1, \dots, Z_n)}(M_n) = P_n(\vartheta').$$

The proof of $Q_n(\vartheta) \geq Q_n(\vartheta')$ is analogous.

Theorem 8 together with Theorem 4 and Lemma 5 weaken a condition given by HOEL [12]. Furthermore, our proof seems to be much simpler than the proof given by HOEL [12]. As a corollary we get the following theorem.

Theorem 9. *If $f_{\vartheta_1}^n, f_{\vartheta_2}^n$ have a monotone likelihood-ratio in $x_{(n)} \in \mathbb{R}^n$ and if $P_\vartheta^{X(n)} \leq_M P_{\vartheta'}^{X(n)}$ for $\vartheta \leq \vartheta'$ then condition (9) holds.*

Proof. By assumption $\frac{f_{\vartheta_1}^n(x_{(n)})}{f_{\vartheta_0}^n(x_{(n)})} = h_{\vartheta_1, \vartheta_0}(x_{(n)})$ where $h_{\vartheta_1, \vartheta_0}$ is isotone. This implies that (Z_1, \dots, Z_n) is an isotonic function of (X_1, \dots, X_n) and, therefore, $P_\vartheta^{(Z_1, \dots, Z_n)} \leq_M P_{\vartheta'}^{(Z_1, \dots, Z_n)}$. So Theorem 9 follows from Theorem 8.

Remark 4. LEHMANN [16], Lemma 4, p. 101 showed that Theorem 9 holds for independent random variables with a monotone likelihood ratio. GHOSH [9] proved that (9) holds if the densities $\{f_\vartheta^n, \vartheta \in \Theta\}$ have a monotone likelihood ratio in a real valued statistic $T(x_{(n)})$. This assumption implies $P_\vartheta^{(Z_1, \dots, Z_n)} \leq_M P_{\vartheta'}^{(Z_1, \dots, Z_n)}$ for $\vartheta \leq \vartheta'$; so GHOSH's result is contained in Theorem 8.

Theorem 9 was formulated by LEHMANN [15], Theorem 2. But LEHMANN's proof is not correct. LEHMANN concludes from the assumption of the monotone likelihood ratio that $P_\vartheta^{Z_n} \leq_M P_{\vartheta'}^{Z_n}$ holds for $\vartheta \leq \vartheta'$. By (2) there exist mappings $f_n(z) \geq z$ such

that $P_{\vartheta'}^{Z_n} = P_{\vartheta}^{f_n(Z_n)}$. LEHMANN uses this relation to compare the paths $(i, Z_i), i \geq 1$ with the paths $(i, f_i(Z_i)), i \geq 1$ concluding

$$P_{\vartheta'} \{((1, Z_1), \dots, (n, Z_n)) \in B\} = P_{\vartheta} \{((1, f_1(Z_1)), \dots, (n, f_n(Z_n))) \in B\},$$

which generally is not true.

By means of Theorem 4 of KAMAE, KRENGEL and O'BRIEN [14] similar results can be given for SPRT's in stochastic processes with continuous time parameter.

Let $(X_t)_{t \geq 0}$ be a stochastic process on $(\Omega, \mathfrak{A}, P_{\vartheta})$, $\vartheta \in \Theta \subset \mathbb{E}$, \mathbb{E} partially ordered, and let for $\vartheta_0 < \vartheta_1$ fixed, P_{ϑ_1} be continuous w.r.t. P_{ϑ_0} restricted on $\mathfrak{A}_t := \mathfrak{A}(X_s; s \leq t)$, $t \geq 0$. Let $a, b: \mathbb{R}_+ \rightarrow \mathbb{R}^1$ be boundary functions $a \leq b$ and define

$$z_t := E \left(\frac{dP_{\vartheta_1}}{dP_{\vartheta_0}} \middle| \mathfrak{A}_t \right),$$

$$N := \inf \{t \in \mathbb{R}_+; Z_t \notin (a_t, b_t)\} \quad (\inf \{t; t \in \emptyset\} := \infty)$$

$$Q_t(\vartheta) := P_{\vartheta}(N \leq t, Z_N \leq a_N)$$

and

$$P_t(\vartheta) := P_{\vartheta}(N \leq t, Z_N \leq b_N).$$

We assume that $(Z_t)_{t \geq 0}$ defines a stochastic process on the Skorohod space $D[0, \infty)$ and that the sets defined above are measurable.

$(Z_t, t \geq 0, N)$ can be looked upon as SPRT. In the literature (cf. GHOSH [10], ZACKS [22], DWORETZKY, KIEFER, WOLFOWITZ [4], FRANZ, WINKLER [8]) usually the case of exponential classes for which X_t is a sufficient statistic and $Z_t = f(t, X_t)$ is considered. Examples are the Wiener process, the Poisson process and the Gammaprocess. We have the following result.

Theorem 10. $P_{\vartheta}^{(Z_{t_1}, \dots, Z_{t_n})} \leq P_{\vartheta'}^{(Z_{t_1}, \dots, Z_{t_n})}$ for all $t_i \in \mathbb{R}_+$, $1 \leq i \leq n$, $\forall n \in \mathbb{N}$ implies $P_t(\vartheta) \leq P_t(\vartheta')$ and $Q_t(\vartheta) \geq Q_t(\vartheta')$, $\forall t \geq 0$.

Remark 5. Stochastic ordering of stochastic processes can be managed in some cases by means of GIRSANOV's [11] theorem or its generalizations. Let $X = (X_t)_{t \geq 0}$ be a stochastic process such that (X, P) is a Wiener-process and let Q be a probability measure continuous w.r.t. P on

$\mathfrak{A} = \bigvee_{t \geq 0} \mathfrak{A}_t$. By GIRSANOV's theorem X has w.r.t. Q the representation

$$X_t = \int_0^t \Phi_s ds + w_t,$$

where Φ_t is the predictable projection on \mathfrak{A}_t , $\Phi_s \in L^2(\langle x \rangle, P)$, where $\langle x \rangle_t = t$ is the associated increasing process and where $(w_t, t \geq 0, Q)$ is a Wiener process. Furthermore, the exponentiation formula gives

$$E \left(\frac{dQ}{dP} \middle| \mathfrak{A}_t \right) = \exp \left\{ \int_0^t \Phi_s dX_s - \frac{1}{2} \int_0^t \Phi_s^2 ds \right\}$$

where the first integral is a stochastic integral. So Q is determined by Φ and P on \mathfrak{X} and Φ can be determined from the density. If $\int_0^t \Phi_s ds \geq 0$ a.s. then X is under Q stochastically larger than under P . This implies immediately statements on the monotonicity of SPRT's for measures which are continuous w.r.t. Wiener measure.

By means of the generalization of GIRSANOV's Theorem given by VAN SCHUPPEN, WONG [19] to general local martingales and by means of the generalized exponentiation formula given by DOLÉANS-DADE [3] similar results are possible also for jump processes.

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Résumé

On donne des caractérisations de l'ordre stochastique de probabilités. Un des buts de cette recherche est de donner une condition suffisante pour l'ordre stochastique. La question d'existence de bornes supérieures et inférieures dans la classe de toutes les probabilités avec des marges prescrites est considérée. À l'aide de l'ordre stochastique on prouve un théorème général sur la monotonie de l'OC-fonction des SPRT's en processus stochastiques.

Zusammenfassung

Es werden einige Charakterisierungen der stochastischen Ordnung von Wahrscheinlichkeitsmaßen bewiesen. Insbesondere wird eine allgemeine hinreichende Bedingung für die stochastische Ordnung angegeben und die Frage nach der Existenz von oberen und unteren Schranken in der Klasse aller Wahrscheinlichkeitsmaße mit vorgegebenen Randverteilungen untersucht. Mit Hilfe der stochastischen Ordnung wird ein allgemeiner Satz über die Monotonie der OC-funktion von sequentiellen Dichtequotiententests in stochastischen Prozessen bewiesen.

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