On mixing sequences of σ -algebras

By J. MOGYORÓDI in Budapest

1. Let $\{\Omega, \mathcal{A}, P\}$ be a probability space, i.e. let \mathcal{A} be a σ -algebra of some subsets of the basic set Ω and for $A \in \mathcal{A}$ let P(A) denote the probability measure of A. Random variables are defined as functions on Ω , which are measurable with respect to the σ -algebra \mathcal{A} . The elements of \mathcal{A} will be called events. \overline{A} denotes the event consisting in the non-occurrence of the event A. The elements of Ω will be denoted by ω . If A is an event then χ_A denotes the indicator of A, i.e. $\chi_A = 1$, if $\omega \in A$ and $\chi_A = 0$, if $\omega \notin A$. Let $\mathcal{A}' \subset \mathcal{A}$ be another σ -algebra of some subsets of Ω . We denote by $M(\xi | \mathcal{A}')$ the conditional expectation of the random variable ξ , i.e. such a random variable, which is measurable with respect to \mathcal{A}' and for which

$$\int_{A} M(\xi | \mathscr{A}') \, dP = \int_{A} \xi \, dP$$

holds, if $A \in \mathscr{A}'$. If, especially, $\xi = \chi_A$ then $M(\chi_A | \mathscr{A}')$ ($A \in \mathscr{A}$) will be called the conditional probability of the event A with respect to \mathscr{A}' and will be denoted by $P(A | \mathscr{A}')$. The simplest properties of the conditional expectation will be used in this paper.

A. RÉNYI ([1]) calls the sequence $\{B_n\}$ of random events mixing with density d if for any fixed random event E

(1)

$\lim_{n \to +\infty} P(B_n E) = dP(E)$

holds, where d is a fixed number, 0 < d < 1. It follows that $\lim_{n \to +\infty} P(B_n) = d$. For mixing sequences $\{B_n\}$ with density d the following limit relation also holds: for every fixed random event E

(1')
$$\lim_{n \to +\infty} P(E|B_n) = P(E).$$

A sequence $\{B_n\}$, satisfying the relation (1'), is *mixing* in the sense of the definition of A. RÉNYI if the limit $\lim_{n \to \infty} P(B_n) = d$ exists and 0 < d < 1.

It is obvious that, if the sequence $\{B_n\}$ is mixing with density d, then the sequence $\{\overline{B}_n\}$ is also mixing with density 1-d. Thus we have for every fixed E

(1")
$$\lim_{n \to +\infty} P(E|\overline{B}_n) = P(E).$$

The facts expressed by (1') and (1'') can be unified in the following manner: let \mathscr{F}_n denote the class $\{\Omega, B_n, \overline{B}_n, O\}$ of sets, where O is the empty set and Ω is the basic set. It is easily seen that \mathscr{F}_n is a σ -algebra. For every fixed event E we have with probability 1

$$P(E|\mathscr{F}_n) = P(E|B_n)\chi_{B_n} + P(E|\overline{B}_n)(1-\chi_{B_n}),$$

where $P(E|\mathscr{F}_n)$ denotes the conditional probability of E with respect to the σ algebra \mathscr{F}_n , χ_{B_n} and $1 - \chi_{B_n}$ are the indicators of the events B_n and \overline{B}_n , respectively. From the fact that $\{B_n\}$ is a mixing sequence with density d, it follows that we have with probability 1

(2)

 $\lim_{n\to+\infty} P(E|\mathscr{F}_n) = P(E)$

and so this limit relation holds in probability measure, too.

However, relation (2) implies the mixing property (1) of the sequences $\{B_n\}$ and $\{\overline{B}_n\}$ only if $\lim P(B_n) = d$ exists and 0 < d < 1.

Relation (2) suggests a more general formulation of the notion of mixing sequences of events.

Let $\{\mathscr{G}_n\}$ $(n=1, 2, \dots; \mathscr{G}_n \subset \mathscr{A})$ be a sequence of σ -algebras.

Definition. The sequence $\{\mathscr{G}_n\}$ of σ -algebras is called mixing if for every fixed event E the sequence

 $(3) P(E|\mathscr{G}_n) (n=1, 2, \cdots)$

converges in probability to P(E), as $n \rightarrow +\infty$.

Examples, showing that the class of the mixing sequences of σ -algebras is not empty, are given e.g. in papers [1], [2], where concrete mixing sequences of events in the sense of the definition of A. RÉNYI have been studied.

ROSENBLATT ([3]) introduced the notion of the strongly mixing and KOLMOGOROV ([6]) used the notion of the completely regular sequences of σ -algebras. Both notions require more than that of the regular sequences of σ -algebras. A sequence $\mathscr{G}_n \supset \mathscr{G}_{n+1}$ $(n=1, 2, \cdots)$ of σ -algebras is called regular if the σ -algebra] $\bigcap_{n=1}^{\infty} \mathscr{G}_n$ is a trivial one. (The trivial σ -algebra is that σ -algebra which contains only sets with measure 0 or 1.) In this note we shall show that the properties of mixing sequences of σ -algebras are close to those of regular sequences.

2. First we shall show the following

Theorem 1. Let $E \in \mathcal{A}$ be an arbitrary fixed event and let $\{\mathcal{G}_n\}$ be a mixing sequence of σ -algebras. Then

 $\lim_{n\to\infty}\sup_{B\in\mathscr{G}_n}|P(EB)-P(E)P(B)|=0.$

(4)

Proof. It can be easily seen that

$$\sup_{B \in \mathscr{G}_n} |P(EB) - P(E)P(B)| = \sup_{B \in \mathscr{G}_n} \left| \int_{B} (P(E|\mathscr{G}_n) - P(E)) dP \right| \leq \\ \leq \sup_{B \in \mathscr{G}_n} \int_{B} |P(E|\mathscr{G}_n) - P(E)| dP \leq \int_{\Omega} |P(E|\mathscr{G}_n) - P(E)| dP.$$

We have $0 \le P(E|\mathscr{G}_n) \le 1$ with probability 1. Since the sequence $\{\mathscr{G}_n\}$ is mixing we obtain (4) from (3) by LEBESGUE's convergence theorem.

Remark. If, for $n=1, 2, \dots, \mathcal{G}_n \supset \mathcal{G}_{n+1}$, then relation (4) is the necessary and sufficient condition for $\{\mathcal{G}_n\}$ to be a regular sequence of σ -algebras. (See e.g. [5].)

It is known that the set-theoretical intersection of σ -algebras is also σ -algebra. By the union of σ -algebras we mean that σ -algebra which is generated by the set-theoretical union of these σ -algebras. It is obvious that the union of trivial σ -algebras is their set-theoretical union. We mean by the "limes inferior" of a sequence of σ -algebras the σ -algebra

$$\bigcup_{n=1}^{\infty} \bigcap_{k=n}^{\infty} \mathscr{G}_k$$

and we denote it by $\liminf \mathscr{G}_n$.

Theorem 2. Let $\{\mathscr{G}_n\}$ be a mixing sequence of σ -algebras. Then $\liminf_{n \to +\infty} \mathscr{G}_n$ is a trivial σ -algebra.

Proof. It is enough to show that for $n=1, 2, \cdots$ the σ -algebras $\bigcap_{k=n}^{\infty} \mathscr{G}_k$ are trivial. Applying assertion (4) of Theorem 1 to the event $E \in \bigcap_{k=n}^{\infty} \mathscr{G}_k$ (and writing *E* instead of *B* in (4)) we obtain $P(E) = (P(E))^2$. From this our assertion follows.

The following theorem asserts that regular sequences of σ -algebras are mixing.

Theorem 3. Let $\{\mathscr{G}_n\}$ be a monotonically decreasing sequence of σ -algebras, i.e. for $n=1, 2, \cdots$ let $\mathscr{G}_n \supset \mathscr{G}_{n+1}$. In order that the sequence $\{\mathscr{G}_n\}$ be mixing, it is necessary and sufficient that the sequence $\{\mathscr{G}_n\}$ be regular.

Proof. If the sequence $\{\mathscr{G}_n\}$ is mixing then, by Theorem 2, $\liminf_{n \to \infty} \mathscr{G}_n = \bigcap_{n=1} \mathscr{G}_n$ is a trivial σ -algebra.

Conversely, if for $n=1, 2, \cdots$ we have $\mathscr{G}_n \supset \mathscr{G}_{n+1}$, then for every fixed event E the sequence

$$P(E|\mathscr{G}_n)$$
 $(n=1, 2, \cdots)$

is a martingale which converges with probability 1 (and, consequently, in probability measure, too) to the random variable

(See e.g. DOOB [4], Section VII, Theorem 4.3.) Since in our case $\bigcap_{n=1} \mathscr{G}_n$ is a trivial σ -algebra, the last conditional probability is equal to P(E) with probability 1. This proves our assertion.

3. Let $\{\mathscr{G}_n\}$ be a sequence of σ -algebras. Further consider all the random variables defined on Ω and measurable with respect to the σ -algebra \mathscr{A} which are square integrable. If \mathscr{H} denotes the set of these random variables, then for $\xi(\omega) \in \mathscr{H}$ $\eta(\omega) \in \mathscr{H}$ we define the inner product of ξ and η by

$$\int_{\Omega} \xi \cdot \eta \, dP,$$

and we denote it by (ξ, η) . The norm of a random variable ξ is defined as $(\xi, \xi)^{\pm}$ and it is denoted by $\|\xi\|$. Then \mathscr{H} is a Hilbert space with the inner product (ξ, η) Let $E \in \mathscr{G}_n$ $(n=1, 2, \cdots)$ and consider the linear combinations of the random variables $\chi_E - P(E)$, further consider also the limits in the mean of these linear combinations. The set of these random variables will be denoted by \mathscr{H}_1 . Clearly, \mathscr{H}_1 is a subspace of the Hilbert space \mathscr{H} . Let us also consider that subspace \mathscr{H}_2 of \mathscr{H} which is orthogonal to \mathscr{H}_1 .

We prove now the following assertion which is well known. We prove it only for the sake of completeness.

Lemma If $\xi \in \mathscr{H}_2$ is a random variable, then we have for every n

$$P(M(\xi|\mathscr{G}_n) = M(\xi)) = 1.$$

Proof. It is enough to prove this assertion in case that $M(\xi) = 0$. Let A be the event $\{\omega: M(\xi|\mathscr{G}_n) > \varepsilon\}$, where ε is an arbitrary fixed positive number. Then, since $M(\xi|\mathscr{G}_n)$ is \mathscr{G}_n -measurable, we have $A \in \mathscr{G}_n$. Further

$$\int_{A} M(\xi|\mathscr{G}_n) dP = \int_{\Omega} M(\xi \chi_A|\mathscr{G}_n) dP = \int_{\Omega} \xi \chi_A dP.$$

Since, by our supposition, ξ and $\chi_A - P(A)$ are orthogonal, we obtain that

$$\int_{A} M(\xi|\mathscr{G}_n) dP = P(A)M(\xi) = 0.$$

On the other hand

$$0=\int\limits_{A}M(\xi|\mathscr{G}_n)\,dP\geq \varepsilon P(A)\geq 0.$$

This results that P(A) = 0. Let further B be the event $\{\omega: M(\xi | \mathscr{G}_n) < -\varepsilon\}$ ($\varepsilon > 0$). Then $B \in \mathscr{G}_n$ and we obtain in such a way as above that

$$\int_{B} M(\xi|\mathscr{G}_n) \, dP = 0.$$

190

On mixing sequences

On the other hand

$$0 = \int_{B} M(\xi | \mathscr{G}_n) \, dP = -\varepsilon P(B) \leq 0.$$

So we have P(B) = 0. Since

$$P(|M(\xi|\mathscr{G}_n)| > \varepsilon) = P(A) + P(B) = 0,$$

and $\varepsilon > 0$ was chosen arbitrarily, we proved that

$$M(\xi|\mathscr{G}_n) = 0, \qquad (n = 1, 2, \cdots)$$

with probability 1.

We shall use this Lemma in the following assertion which facilitates to decide whether a sequence $\{\mathscr{G}_n\}$ of σ -algebras is mixing or not.

Theorem 4. Let $\{\mathscr{G}_n\}$ be a sequence of σ -algebras. A necessary and sufficient condition for $\{\mathscr{G}_n\}$ to be mixing is that for every fixed $E \in \mathscr{G}_k$ $(k = 1, 2, \dots)$ the sequence

$$P(E|\mathscr{G}_n)$$
 $(n = 1, 2, \cdots)$

of random variables converge in probability to P(E).

Proof. The necessity part of the assertion is obvious. The sufficiency part of the proof can be performed as follows. Let $\varepsilon > 0$ be an arbitrary fixed number and let *E* be an arbitrary event. Then

(5)
$$\varepsilon P(|P(E|\mathscr{G}_n) - P(E)| > \varepsilon) \leq M(|M((\chi_E - P(E))|\mathscr{G}_n)|).$$

Let us decompose $\chi_E - P(E)$ in the form

$$\xi_1 + \xi_2$$

where $\xi_1 \in \mathcal{H}_1$ and $\xi_2 \in \mathcal{H}_2$, \mathcal{H}_1 and \mathcal{H}_2 being defined as above. Since $M(\chi_E - P(E)) = 0$, further $M(\xi_1) = 0$, one has $M(\xi_2) = 0$. So by our Lemma we have for every n

$$M(\xi_2|\mathscr{G}_n)=0$$

with probability 1. On the other hand

$$M((\chi_E - P(E))|\mathscr{G}_n) = M(\xi_1|\mathscr{G}_n) + M(\xi_2|\mathscr{G}_n) = M(\xi_1|\mathscr{G}_n)$$

with probability 1. So we have

(6)
$$M(|M((\chi_E - P(E))|\mathscr{G}_n)|) = M(|M(\xi_1|\mathscr{G}_n)|).$$

 ξ_1 , being element of \mathscr{H}_1 , can be approximated in the mean by finite linear combinations of the elements $\chi_A - P(A)$, $(A \in \mathscr{G}_k, k = 1, 2, \cdots)$. Denote the sequence,

approximating ξ_1 in the mean, by $\eta_1, \eta_2, \cdots (\eta_j \in \mathcal{H}_1)$. For every fixed k the sequence

$$M(\eta_k | \mathscr{G}_n)$$
 $(n = 1, 2, \cdots)$

converges, obviously, in probability to 0. Let $\delta > 0$ be arbitrary and let us put k such that $\|\xi_1 - \eta_k\| < \delta$ be satisfied. Then fix k. It is easily seen that η_k is bounded and so is $M(\eta_k | \mathscr{G}_n)$ with probability 1. Now we have

(7)
$$M(|M(\xi_1|\mathscr{G}_n)|) \leq M(|M((\xi_1 - \eta_k)|\mathscr{G}_n)|) + M(|M(\eta_k|\mathscr{G}_n)|).$$

The second member on the right-hand side of (7) by LEBESGUE's theorem converges to 0, while the first is smaller than

$$M(|M((\xi_1-\eta_k)|\mathscr{G}_n)|) \leq M(M(|\xi_1-\eta_k||\mathscr{G}_n)) \leq ||\xi_1-\eta_k||.$$

Conferring (7), (6) and (5) we obtain that

(8)
$$\limsup_{n \to \infty} P(|P(E|\mathscr{G}_n) - P(E)| > \varepsilon) \leq \frac{\delta}{\varepsilon}.$$

Since $\varepsilon > 0$ and $\delta > 0$ vary independently each of other, (8) means the assertion of the theorem.

Theorem 4 gives similar conditions for $\{\mathscr{G}_n\}$ to be mixing as the conditions of the theorem of A. RÉNYI ([1]) for the sequence of events $\{B_n\}$ to be mixing with density d(0 < d < 1).

Theorem 5. Let $\{\mathscr{G}_n\}$ be a mixing sequence of σ -algebras and let z be a random variable having finite mean-value. Then the sequence

$$M(z \mid \mathscr{G}_n) \qquad (n = 1, 2, \cdots)$$

of random variables converges in probability to M(z).

Proof. The assertion of the theorem is true if z is of the form:

$$\sum_{k=1}^{j} c_k \chi_{E_k}$$

where c_k (k = 1, 2, ..., j) is a real number, $E_k \in \mathscr{A}$ (k = 1, 2, ..., j) are events such that $E_k \cap E_l = \emptyset$, and $\bigcup_{k=2}^{j} E_k = \Omega$, further χ_{E_k} denotes the indicator of E_k and j is finite positive integer. Since M(z) is finite, the random variable z can be approximated in L^1 norm by the random variables of the mentioned form as close as we please. Let z^* be such a random variable for which

$$M(|z^*-z|) < \varepsilon$$

· 192

holds. Then

$$\int_{\Omega} |M(z|\mathscr{G}_n) - M(z)| dP \leq \int_{\Omega} |M(z|\mathscr{G}_n) - M(z^*|\mathscr{G}_n)| dP +$$
$$+ \int_{\Omega} |M(z^*|\mathscr{G}_n) - M(z^*)| dP + \int_{\Omega} |z^* - z| dP \leq$$
$$\leq \int_{\Omega} M(|z - z^*||\mathscr{G}_n) dP + \int_{\Omega} |M(z^*|\mathscr{G}_n) - M(z^*)| dP + M(|z - z^*|).$$

The first and the third terms on the right hand side of this inequality are smaller than ε and the second converges to zero. This proves the theorem.

4. It is interesting to investigate the analogon of Theorem 4 in case of the almost everywhere convergence. Theorem 6 makes this for martingales.

Theorem 6. Let $\{\mathscr{G}_n\}$ be a sequence of σ -algebras and suppose that for every event E the conditional probabilities $P(E|\mathscr{G}_n)$ $(n=1, 2, \cdots)$ form a martingale. If for every fixed $E \in \mathscr{G}_k$ $(k=1, 2, \cdots)$ we have

$$P\left(\lim_{n \to +\infty} P(E|\mathscr{G}_n) = P(E)\right) = 1,$$

then the same holds for every event E.

Proof. We have for arbitrary fixed E

 $0 \leq \dot{P}(E|\mathscr{G}_n) \leq 1.$

Thus by the convergence theorem of the martingales (cf. DOOB [4], Section VII, Theorem 4. 1.) the limit

$$\lim_{n\to+\infty}P(E|\mathscr{G}_n)=\zeta_E(\omega),$$

exists with probability 1, where $\xi_E(\omega)$ is a random variable. We have further $M(\xi_E(\omega)) = P(E)$. So it remains to prove that $P(\xi_E(\omega) = P(E)) = 1$. Let us consider for this purpose $M(|\xi_E(\omega) - P(E)|)$. We have

(9)
$$M(|\xi_E(\omega) - P(E)|) \leq M(|\xi_E(\omega) - P(E|\mathscr{G}_n)|) + M(|M((\chi_E - P(E))|\mathscr{G}_n)|).$$

By LEBESGUE's theorem, the first term on the right hand side converges to zero as $n \to \infty$. For dealing with the second member, let us decompose the random variable $\chi_E - P(E)$ into the form

$$\xi_1 + \xi_2$$
,

where $\xi_1 \in \mathcal{H}_1$ and $\xi_2 \in \mathcal{H}_2$; \mathcal{H}_1 and \mathcal{H}_2 being defined as above. Since $\xi_1 \in \mathcal{H}_1$, one has $M(\xi_1) = 0$ and so $M(\xi_2) = 0$. By our Lemma we have for every n

$$M(\xi_2|\mathscr{G}_n)=0.$$

13 _.A

with probability 1. On the other hand

$$M((\chi_E - P(E))|\mathscr{G}_n) = M(\xi_1|\mathscr{G}_n) + M(\xi_2|\mathscr{G}_n) = M(\xi_1|\mathscr{G}_n)$$

with probability 1. So

(10)
$$M(|M((\chi_E - P(E))|\mathscr{G}_n)|) = M(|M(\xi_1|\mathscr{G}_n)|).$$

 ξ_1 , being element of \mathscr{H}_1 , can be approximated in the mean by finite linear combinations of the elements $\chi_A - P(A)$ $(A \in \mathscr{G}_k, k = 1, 2, \cdots)$. Denote this sequence, approximating ξ_1 in the mean, by η_1, η_2, \cdots . For every fixed k we have with probability 1

$$\lim_{n\to+\infty}M(\eta_k|\mathscr{G}_n)=0.$$

Let $\varepsilon > 0$ be arbitrary and let us put k such that $\|\xi_1 - \eta_k\| < \varepsilon$ be satisfied. Then fix k. It is obvious that η_k is bounded and so is $M(\eta_k | \mathcal{G}_n)$. Now we have

(11)
$$M(|M(\xi_1|\mathscr{G}_n)|) \leq M(|M((\xi_1 - \eta_k|\mathscr{G}_n)|) + M(|M(\eta_k|\mathscr{G}_n)|).$$

The second member on the right hand side of (11) converges to 0, while the first is smaller than $\|\xi_1 - \eta_k\|$. Conferring (11), (10), and (9) we see that

$$M(|\xi_E(\omega) - P(E)|) \leq \liminf_{n \to \infty} M(|\xi_E(\omega) - P(E|\mathscr{G}_n)|) +$$

$$\liminf_{n\to\infty} M(|M((\xi_1-\eta_k)|\mathscr{G}_n)|) + \liminf_{n\to\infty} M(|M(\eta_k|\mathscr{G}_n)|) \leq \varepsilon.$$

Since $\varepsilon > 0$ was chosen arbitrarily, the inequality (12) means our assertion.

5. By the aid of the mixing sequences of σ -algebras sequences of random events, which are mixing with density d(0 < d < 1) can be constructed as follows:

Theorem 7. Let $\{\mathscr{G}_n\}$ be a mixing sequence of σ -algebras and $\{B_n\}$ a sequence of random events, for which $B_n \in \mathscr{G}_n$, further $\lim_{n \to +\infty} P(B_n) = d$ exists. Then $\{B_n\}$ is a mixing sequence of events with density d.

Proof. Let E be an arbitrary event. By our supposition the condition of Theorem 1 is satisfied. So we have

$$|P(EB_n) - dP(E)| \leq \sup_{B \in \mathscr{G}_n} |P(EB) - P(E)P(B)| + P(E)|P(B_n) - d|.$$

Letting $n \rightarrow \infty$, the limit of the right hand side will be 0. This proves the theorem.

6. Consider now some consequences of the above results. We say that a sequence $\{\zeta_n\}$ $(n=1, 2, \cdots)$ of random variables is mixing if the sequence of the corresponding σ -algebras \mathscr{G}_n $(n=1, 2, \cdots)$ generated by the random variable ζ_n is mixing.

Theorem 8. If the sequence ζ_n $(n=1, 2, \cdots)$ of random variables is mixing and η is an arbitrary random variable the ζ_n is asymptotically independent of η . If,

On mixing sequences

in addition, ζ_n converges in probability to a random variable ζ , then ζ is constant with probability 1.

Proof. The first assertion follows immediately from Theorem 1. In fact, x and y being arbitrary real numbers, by (4) we obtain if $n \to \infty$

$$|P(\zeta_n < x, \eta < y) - P(\zeta_n < x)P(\eta < y)| \le$$

$$\le \sup_{B \in \mathscr{G}_n} |P(B, \eta < y) - P(B)P(\eta < y)| \to 0.$$

From this it follows especially that if ζ_n converges in probability to ζ , then for every real x

$$\lim_{n \to +\infty} P(\zeta_n < x, \zeta < x) = (P(\zeta < x))^2.$$

On the other hand, if $\varepsilon > 0$ is an arbitrary number

$$P(\zeta_n < x, \zeta < x) = P(\zeta_n < x, \zeta < x, |\zeta_n - \zeta| < \varepsilon) + P(\zeta_n < x, \zeta < x, |\zeta_n - \zeta| \ge \varepsilon).$$

The second member on the right hand side converges to 0, while the first satisfies the inequality

$$P(\zeta_n < x - \varepsilon, |\zeta_n - \zeta| < \varepsilon) \leq P(\zeta_n < x, \zeta < x |\zeta_n - \zeta| < \varepsilon) \leq P(\zeta_n < x).$$

If x and $x-\varepsilon$ are continuity points of the distribution function of ζ , then the right hand side converges to $P(\zeta < x)$ and the limit of the left hand side of the inequality is greater than $P(\zeta < x-\varepsilon)$. Since $\varepsilon > 0$ was chosen arbitrarily, we see that

$$\lim_{n \to \infty} P(\zeta_n < x, \zeta < x) = P(\zeta < x).$$

So we have

$$P(\zeta < x) = (P(\zeta < x))^2,$$

which means that for every real x

$$P(\zeta < x) = 0, \quad \text{or} \quad 1.$$

This proves our assertion.

Theorem 9. Let $\{\mathscr{G}_n\}$ be a mixing sequence of σ -algebras in the probability space $\{\Omega, \mathcal{A}, P\}$. If Q is another probability measure, defined on \mathcal{A} , and it is absolutely continuous with respect to P, then for every event E the sequence

$$Q(E|\mathscr{G}_n)$$
 $(n = 1, 2, \cdots)$

of conditional Q-probabilities converges in P (and, consequently, in Q)-probability to Q(E), as $n \to \infty$.

Proof. $Q(E|\mathcal{G}_n)$, being conditional Q-probability, is a random variable, which is measurable with respect to \mathcal{G}_n and for every $A \in \mathcal{G}_n$ we have

$$Q(EA) = \int_{A} Q(E|\mathscr{G}_n) \, dQ = \int_{A} Q(E|\mathscr{G}_n) \lambda(\omega) \, dP,$$

where $\lambda(\omega)$ is the Radon-Nikodým derivative of Q with respect to P. Now we have

$$Q(EA) = \int_{A} Q(E|\mathscr{G}_n) \lambda \, dP = \int_{A} M(Q(E|\mathscr{G}_n) \lambda | \mathscr{G}_n) \, dP = \int_{A} Q(E|\mathscr{G}_n) M(\lambda | \mathscr{G}_n) \, dP.$$

On the other hand for every $A \in \mathscr{G}_n$

$$Q(EA) = \int_{A} \chi_E \lambda \, dP = \int_{A} M(\chi_E \lambda | \mathscr{G}_n) \, dP.$$

Since the conditional expectation is uniquely determined mod P, we have with probability 1

$$Q(E|\mathscr{G}_n)M(\lambda|\mathscr{G}_n) = M(\chi_E\lambda|\mathscr{G}_n).$$

By Theorem 5 the random variables.

$$M(\lambda|\mathscr{G}_n)$$
 and $M(\chi_E \lambda|\mathscr{G}_n)$

converge in probability to 1 and to

$$M(\chi_E \lambda) = \int_E \lambda \, dP = Q(E),$$

respectively. From this and from the preceding equality our theorem follows.

Corollary. If $A_n \in \mathscr{G}_n$, $\lim_{n \to +\infty} P(A_n) = d$, then under the conditions of Theorem 9 we have for every event E

$$\lim_{n\to\infty} Q(A_n E) = dQ(E)$$

i.e., if a sequence $\{A_n\}$ is mixing with density d in the probability space $\{\Omega, \mathcal{A}, P\}$, then it is mixing with the same density in $\{\Omega, \mathcal{A}, Q\}$ provided that Q is absolutely continuous probability measure with respect to P.

Proof. By Theorem 9, $Q(E|\mathscr{G}_n)$ converges in probability to Q(E) and so

$$Q(A_n E) = \int_{A_n} Q(E|\mathscr{G}_n) \, dQ \to Q(E) \lim_{n \to \infty} P(A_n) = dQ(E),$$

as $n \to \infty$, because $\lim_{n \to \infty} Q(A_n) = \lim_{n \to \infty} \int_{A_n} M(\lambda | \mathscr{G}_n) dP = \lim_{n \to \infty} P(A_n) = d.$

As another consequence of Theorem 9 we prove now

Theorem 10. Let $\{\mathscr{G}_n\}$ be a mixing sequence of σ -algebras in the probability space $\{\Omega, \mathscr{A}, P\}$ and $\{\zeta_n\}$ a sequence of random variables such that ζ_n is \mathscr{G}_n -measurable $(n=1, 2, \cdots)$. Let further Q be a probability measure which is absolutely continuous with respect to P. If

$$\lim_{n \to +\infty} P(\zeta_n < x) = F(x),$$

where F(x) is a distribution function and the limit relation holds for every fixed x which is a continuity point of F(x), then we have at every continuity point of F(x)

$$\lim Q(\zeta_n < x) = F(x).$$

Remark. Theorem 10 is a generalization of Theorem 3.1 of [5] and of the corresponding theorem of [1], where a similar assertion has been proved for regular sequences of σ -algebras.

Proof. Let x be an arbitrary fixed continuity point of F(x). Then the event $A_n = \{\omega: \zeta_n(\omega) < x\}$ belongs to \mathscr{G}_n . So by the Corollary to Theorem 9 (putting Ω instead of E) we obtain our assertion.

References

- [1] A. RÉNYI, On mixing sequences of sets, Acta Math. Acad. Sci. Hung., 9 (1958), 215-228.
- [2] A. RÉNYI—P. RÉVÉSZ, On mixing sequences of random variables, Acta Math. Acad. Sci. Hung., 9 (1958), 338-343.
- [3] M. ROSENBLATT, A central limit theorem and a strong mixing condition, *Proc. Nat. Acad. Sci.* USA, 42 (1956), 43.

[4] J. L. DOOB, Stochastic processes (New York, 1953).

[5] T. NEMETZ-GY. VARGA, Határeloszlástételek kiterjesztéséről, Magyar Tudományos Akadémia III. Osztály Közleményei, 14 (1964), 416-421.

[6] В. А. Волконский—Ю. А. Розанов Некоторые предельные теоремы для случайных функций, *Теория вероятностей и её применения*, 4 (1959), 186—207.

[7] И. А. Ибрагимов, Некоторые предельные теоремы для стационарных процессов, *Теория вероятностей и её применения*, 7 (1962), 361—392.

(Received September 16, 1967)