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ON THE APPROXIMATION OF A RANDOM VARIABLE BY A CONDITIONING OF A GIVEN SEQUENCE

BY

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Abstract. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space. If (X_n) is a sequence of r.v.'s satisfying $X_n \to 0$ a.s. (respectively, in probability) as $n \to \infty$ and $EX_n^+ \to \infty$, $EX_n^- \to \infty$ as $n \to \infty$, then for any r.v. Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that $E(X_n | \mathfrak{A}_n |) \to Y$ a.s. (respectively, in probability) as $n \to \infty$.

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The main result of the paper is the following

THEOREM 1. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space. If (X_n) is a sequence of random variables satisfying the conditions

(1)
$$\lim_{n\to\infty} X_n = 0 \ a.s.,$$

(2)
$$\lim_{n\to\infty} EX_n^+ = \lim_{n\to\infty} EX_n^- = \infty,$$

then for any random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = Y \text{ a.s.}$$

An analogous theorem for stochastic convergence is also proved.

In [2] and [3] Paszkiewicz describes, in particular, the construction of a σ -field \mathfrak{A} such that $E(X_n | \mathfrak{A})$ does not converge to 0 for a sequence of random vectors X_n tending to 0 in some sense. We give a more precise description of $E(X | \mathfrak{A})$ when X is close to 0 in measure topology and EX^+ , EX^- are large enough.

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The idea of proof of Theorem 1 is based on the following observation. Let X be a random variable and Y be a simple random variable. If EX^+ and EX^- are large enough, then we can construct a σ -field \mathfrak{A} such that $E(X | \mathfrak{A})$ equals Y on the set where X is small enough. For some sequence (Y_n) of simple random variables converging to Y we construct a sequence (\mathfrak{A}_n) of σ -fields such that $E(X_n | \mathfrak{A}_n)$ equals Y_n on some large enough set.

All necessary properties of conditional expectation can be found in [1].

Proof of Theorem 1. From (1) we get immediately

(3)
$$\lim_{i\to\infty} P(\sup_{n\geq i} |X_n| > \varepsilon) = 0 \quad \text{for every } \varepsilon > 0.$$

Let (ε_j) be a sequence of real numbers such that $0 < \varepsilon_j < 1$ and $\varepsilon_j > 0$ as $j \to \infty$. The equality (3) implies now the existence of a strictly monotone sequence (n_j) of integers such that

$$P(\sup_{n \ge n_j} |X_n| > \varepsilon_j) \le 2^{-j} \quad \text{for } j \ge 1.$$

We put

(4)
$$A_{n_j} = \{ \sup_{n \ge n_j} |X_n| > \varepsilon_j \} \quad \text{for } j \ge 1.$$

Let $p(n) = \max\{n_j: n_j \leq n\}$. From (2) it follows that

$$\int_{P(n)} X_n^+ \to \infty \quad \text{and} \quad \int_{A_{P(n)}} X_n^- \to \infty \quad \text{as } n \to \infty.$$

We put

$$M_n = \frac{1}{4} \int_{A_p(n)} X_n^+ \quad \text{and} \quad N_n = \frac{1}{4} \int_{A_p(n)} X_n^- \quad \text{for } n \ge 1$$

(we can assume that $M_n \ge 1$ and $N_n \ge 1$ for $n \ge 1$).

Let (Y_n) be a sequence of simple random variables of the form

$$Y_n = \sum_{i=1}^{k(n)} \alpha_i(n) \, \mathbf{1}_{G_i(n)} + \sum_{i=1}^{l(n)} \beta_i(n) \, \mathbf{1}_{H_i(n)}$$

and such that

$$\lim_{n \to \infty} Y_n = Y \text{ a.s.},$$

$$\alpha_i(n) > 0 \quad \text{for } n \ge 1, \ i = 1, 2, \dots, k(n),$$

$$\beta_i(n) < 0 \quad \text{for } n \ge 1, \ i = 1, 2, \dots, l(n),$$

(5)

We can also assume that for $n_j \leq n < n_{j+1}$ we have

(6)
$$-N_n \leqslant Y_n(\omega) \leqslant -\varepsilon_j$$
 or $\varepsilon_j \leqslant Y_n(\omega) \leqslant M_n$ for $\omega \in \Omega$.

Since $(\Omega, \mathfrak{F}, P)$ is a non-atomic one, we can consider a random variable Z uniformly distributed on [0, 1].

Before we construct the σ -fields \mathfrak{A}_n we shall prove that for every $n \ge 1$ there exist real numbers $t_1, \ldots, t_{k(n)}$ and $s_1, \ldots, s_{l(n)}$ satisfying

(7)
$$0 \leqslant t_1 \leqslant \ldots \leqslant t_{k(n)} \leqslant 1,$$

$$(8) 0 \leqslant s_1 \leqslant \ldots \leqslant s_{l(n)} \leqslant 1,$$

(9)
$$\alpha_i(n) P(C_i(n)) = \int_{C_i(n)} X_n \text{ for } i = 1, ..., k(n),$$

(10)
$$\beta_i(n) P(D_i(n)) = \int_{D_i(n)} X_n \text{ for } i = 1, ..., l(n),$$

where

(11)
$$C_i(n) = (G_i(n) \setminus A_{p(n)}) \cup (Z^{-1}([t_{i-1}, t_i]) \cap A_{p(n)} \cap \{X_n^+ > 0\}),$$

(12)
$$D_i(n) = (H_i(n) \setminus A_{p(n)}) \cup (Z^{-1}([s_{i-1}, s_i)) \cap A_{p(n)} \cap \{X_n^- > 0\}).$$

From (4) and (6) we deduce that

(13)
$$\varepsilon_j \leq \alpha_i(n) \leq M_n$$
 for $n_j \leq n < n_{j+1}$ and $i = 1, ..., k(n)$

and

(14)
$$|X_n| \leq \varepsilon_j$$
 for $n_j \leq n < n_{j+1}$ and $\omega \notin A_{n_j}$.

For $t_1 \in [0, 1]$ we put

$$T_1(t_1) = \alpha_1(n) P((G_1(n) \setminus A_{p(n)}) \cup (Z^{-1}([0, t_1)) \cap A_{p(n)} \cap \{X_n^+ > 0\}))$$

$$-\int_{(G_1(n)\setminus A_{p(n)})\cup (Z^{-1}([0,t_1))\cap A_{p(n)}\cap \{X_n^+>0\})} X_n$$

It can easily be seen that T_1 is continuous.

From (13) and (14) we get

$$T_1(0) = \alpha_1(n) P(G_1(n) \setminus A_{p(n)}) - \int_{G_1(n) \setminus A_{p(n)}} X_n \ge 0.$$

On the other hand, from (13), (14) and the definition of M_n we have

$$T_1(1) = \alpha_1(n) P((G_1(n) \setminus A_{p(n)}) \cup (A_{p(n)} \cap \{X_n^+ > 0\}))$$

$$-\int_{G_1(n)\setminus A_{P(n)}} X_n - \int_{A_{P(n)}\cap\{X_n^+>0\}} X_n$$

$$\leqslant M_n + 1 - 4M_n \leqslant 0.$$

Thus there exists $t_1 \in [0, 1]$ satisfying (9) for i = 1.

Let us assume that we have found t_1, \ldots, t_{r-1} satisfying (7) and (9). Then from (9) we get

(15)
$$M_{n} \Big[P \left((G_{i}(n) \setminus A_{p(n)}) \cup \left(Z^{-1} \left([t_{i}, t_{i-1}) \right) \cap A_{p(n)} \cap \{ X_{n}^{+} > 0 \} \right) \right) \Big]$$
$$\geqslant \int_{Z^{-1} ([t_{i-1}, t_{i})) \cap A_{p(n)} \cap \{ X_{n}^{+} > 0 \}} X_{n} - \varepsilon_{1} P \left(G_{i}(n) \setminus A_{p(n)} \right), \quad i = 1, ..., r-1.$$

By summing (15) over *i* we obtain

$$M_{n} \geq \int_{Z^{-1}([0,t_{r-1})) \cap A_{p(n)} \cap \{X_{n}^{+} > 0\}} X_{n} - \varepsilon_{1}$$

= $\int_{A_{p(n)} \cap \{X_{n}^{+} > 0\}} X_{n} - \int_{Z^{-1}([t_{r-1},1]) \cap A_{p(n)} \cap \{X_{n}^{+} > 0\}} X_{n} - \varepsilon_{1}$
 $\geq 3M_{n} - \int_{Z^{-1}([t_{r-1},1]) \cap A_{p(n)} \cap \{X_{n}^{+} > 0\}} X_{n}.$

Thus

(16)

6)
$$\int_{Z^{-1}([t_{r-1},1])\cap A_{p(n)}\cap\{X_{n}^{+}>0\}} X_{n} \ge 2M_{n}.$$

We put for $t_r \in [t_{r-1}, 1]$

$$T_r(t_r) = \alpha_r(n) P(C_r(n)) - \int_{G_r(n) \setminus A_{p(n)}} X_n - \int_{Z^{-1}([t_{r-1}, t_r)) \cap A_{p(n)} \cap \{X_n^+ > 0\}} X_n.$$

We have

 $T_r(t_{r-1}) \ge 0$

and from (16) we obtain

$$T_r(1) \leqslant M_n + 1 - 2M_n \leqslant 0.$$

In consequence, there exists $t_r \in [t_{r-1}, 1]$ satisfying (9) for i = r.

By the above arguments we conclude that there exist $t_1, \ldots, t_{k(n)}$ such that (7) and (9) hold. In the same manner we can find $s_1, \ldots, s_{l(n)}$ satisfying (8) and (10).

Let us define now the sequence (\mathfrak{A}_n) . For $n \ge 1$ we put

$$\mathfrak{A}_{n} = \sigma(C_{1}(n), \ldots, C_{k(n)}(n), D_{1}(n), \ldots, D_{l(n)}(n)).$$

From (11) and (12) it can easily be seen that the sets $C_1(n), \ldots, C_{k(n)}(n)$ and $D_1(n), \ldots, D_{l(n)}(n)$ are mutually disjoint. From (9) and (10) it follows immediately that

$$E(X_n | \mathfrak{A}_n)(\omega) = Y_n(\omega) \quad \text{for } \omega \notin A_{p(n)}.$$

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As $\sum_{j=1}^{\infty} P(A_{nj}) < \infty$, we conclude by the Borel-Cantelli lemma that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = \lim_{n\to\infty} Y_n = Y \text{ a.s.}$$

This completes the proof.

One can easily see that if Y is nonnegative, then the assumption that $EX_n^- \to \infty$ as $n \to \infty$ is not used in the proof. Thus the following version of the previous theorem holds:

THEOREM 2. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space. If (X_n) is a sequence of random variables satisfying the conditions

$$\lim_{n\to\infty} X_n = 0 \ a.s. \quad and \quad \lim_{n\to\infty} EX_n^+ = \infty,$$

then for any nonnegative random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = Y \ a.s.$$

The next example shows that we cannot omit the assumption that the probability space is non-atomic.

COUNTEREXAMPLE 1. Let $\Omega = [0, 1]$, $\mathfrak{F} = \sigma((2^{-n}, 2^{-n+1}], n \ge 1)$ and P be the Lebesgue measure on [0, 1]. Let us consider random variables X_n given by

$$X_n = 4^n \mathbb{1}_{(2^{-n}, 2^{-n+1})}$$
 for $n \ge 1$.

Then

$$\lim_{n\to\infty} X_n = 0 \text{ a.s.} \quad \text{and} \quad \lim_{n\to\infty} E X_n = \infty.$$

It can be also easily checked that for any σ -field $\mathfrak{A} \subset \mathfrak{F}$ we have

 $E(X_n \mid \mathfrak{A}) = 0$ or $2^n \leq E(X_n \mid \mathfrak{A}) \leq 4^n$ a.s.

Thus there exists no sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = 1 \text{ a.s.}$$

One can ask whether the assumption that the sequence (X_n) converges to zero almost surely can be replaced by the stochastic convergence. The answer is negative.

COUNTEREXAMPLE 2. Let $\Omega = [0, 1]$, $\mathfrak{F} = \text{Borel}([0, 1])$, and P be the Lebesgue measure on [0, 1]. Let (X_n) be a sequence of random variables given by

$$X_{2^{n+k}} = 4^n \mathbb{1}_{[k2^{-n},(k+1)2^{-n}]}$$
 for $n \ge 0, k = 0, 1, ..., 2^n - 1$.

It can easily be seen that the sequence (X_n) converges to zero in probability but does not converge with probability one. Let us assume now that there exists a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} \mathbb{E}(X_n | \mathfrak{A}_n) = 0 \text{ a.s.}$$

Then we can find $m_0 \ge 1$ such that

$$P\left(\sup_{m\geq m_0} E\left(X_m \mid \mathfrak{A}_m\right) > \frac{1}{2}\right) \leq \frac{1}{2}.$$

Let us put

$$A = \{ \omega \colon \sup_{m \ge m_0} E(X_m | \mathfrak{A}_m)(\omega) > \frac{1}{2} \}.$$

Then for $\omega \in A^c$ and $m \ge m_0$ we have

$$E(X_m \mid \mathfrak{A}_m)(\omega) \leq \frac{1}{2}.$$

Since $P(A^c) \ge \frac{1}{2}$, we can choose $n \ge 1$ and $k = 0, 1, ..., 2^n - 1$ such that $2^n + k \ge m_0$ and

(17)
$$P(A^{c} \cap [k2^{-n}, (k+1)2^{-n}]) \ge 2^{-n-1}.$$

From (17) we get

$$\frac{1}{2} \ge \int_{\{E(X_{2^n+k}|\mathfrak{Y}_{2^n+k}) \le 1/2\}} E(X_{2^n+k}|\mathfrak{Y}_{2^n+k}) = \int_{\{E(X_{2^n+k}|\mathfrak{Y}_{2^n+k}) \le 1/2\}} X_{2^n+k}$$
$$\ge \int_{A^c \cap [k2^{-n},(k+1)2^{-n}]} X_{2^n+k} \ge 4^n 2^{-n-1} \ge 1,$$

which gives a contradiction and proves that there exists no sequence (\mathfrak{A}_n) such that

$$\lim_{n\to\infty} E(X_n \,|\, \mathfrak{U}_n) = 0 \text{ a.s.}$$

We have shown that under the assumptions of Theorem 1 the almost sure convergence cannot be replaced by the stochastic one. Nevertheless, weaker assumptions give a weaker conclusion. We shall prove the following

THEOREM 3. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space. If (X_n) is a sequence of random variables satisfying the conditions

(18)
$$\lim_{n \to \infty} X_n = 0 \text{ in probability,}$$

(19)
$$\lim_{n\to\infty} EX_n^+ = \lim_{n\to\infty} EX_n^- = \infty,$$

then for any random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = Y \text{ in probability.}$$

Proof. From (18) we get immediately

(20)
$$\lim_{n \to \infty} P(|X_n| > \varepsilon) = 0 \quad \text{for every } \varepsilon > 0.$$

Let (ε_j) be a sequence of real numbers such that $0 < \varepsilon_j < 1$ and $\varepsilon_j > 0$ as $j \to \infty$. The equality (20) implies now the existence of an increasing sequence (n_j) of integers such that

$$P(|X_n| > \varepsilon_j) \leq 2^{-j} \quad \text{for } n \geq n_j.$$

We put

$$A_n = \{ \omega \colon |X_n(\omega)| > \varepsilon_j \} \quad \text{for } n_j \leq n < n_{j+1}.$$

It is obvious that $P(A_n) \to 0$ as $n \to \infty$. From (19) we deduce that

$$\int_{A_n} X_n^+ \to \infty \quad \text{and} \quad \int_{A_n} X_n^- \to \infty \quad \text{as } n \to \infty.$$

Now in the same manner as in the proof of Theorem 1 we can find a sequence (Y_n) of simple random variables and a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n \to \infty} Y_n = Y \text{ a.s.} \quad \text{and} \quad E(X_n | \mathfrak{A}_n)(\omega) = Y_n(\omega) \text{ for } \omega \notin A_n.$$

Fix $\varepsilon > 0$. For any $\eta > 0$ we can find $m \ge 1$ such that

 $P(|Y_n - Y| > \varepsilon) < \eta/2$ for $n \ge m$

and

$$P(E(X_n | \mathfrak{A}_n) \neq Y_n) \leq P(A_n) < \eta/2 \quad \text{for } n \geq m.$$

Therefore, we obtain

$$P(|E(X_n|\mathfrak{A}_n) - Y| > \varepsilon) \leq \eta/2 + \eta/2 = \eta \quad \text{for } n \geq m,$$

which means that $E(X_n | \mathfrak{A}_n) \to Y$ in probability as $n \to \infty$. This completes the proof of the theorem m

In a similar way we obtain

THEOREM 4. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space. If (X_n) is a sequence of random variables satisfying the conditions

$$\lim_{n\to\infty} X_n = 0 \text{ in probability} \quad and \quad \lim_{n\to\infty} EX_n^+ = \infty,$$

then for any nonnegative random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that

$$\lim_{n\to\infty} E(X_n | \mathfrak{A}_n) = Y \text{ in probability.}$$

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