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**On the stochastic convergence
of conditional expectations of some
random sequences**

ABSTRACT. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space and (X_n) be a sequence of integrable random variables. We shall indicate several conditions under which the following conclusion holds: for any random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields such that $\mathbb{E}(X_n|\mathfrak{A}_n) \rightarrow Y$ in probability, for $n \rightarrow \infty$.

1. Introduction. Let $(\Omega, \mathfrak{F}, P)$ be a non-atomic probability space and (X_n) be a sequence of integrable random variables. The aim of this paper is to find possibly weakest assumptions on (X_n) under which the following conclusion holds:

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

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

It is easily seen that condition (α) forces:

$$(0) \quad \lim_{n \rightarrow \infty} \mathbb{E}X_n^+ = \lim_{n \rightarrow \infty} \mathbb{E}X_n^- = \infty.$$

However, as shown by Ex. 4.1 in [3], (0) is not sufficient for (α). Results presented in this paper generalize the following ones obtained in [2]:

2000 *Mathematics Subject Classification.* 60A10.

Key words and phrases. Conditional expectation, stochastic convergence.

Theorem 1. Let (X_n) be a sequence of random variables satisfying the following conditions

$$\lim_{n \rightarrow \infty} X_n = 0 \text{ in probability}$$

and

$$\lim_{n \rightarrow \infty} \mathbb{E}X_n^+ = \lim_{n \rightarrow \infty} \mathbb{E}X_n^- = \infty.$$

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

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

Theorem 2. Let (X_n) be a sequence of random variables satisfying the following conditions

$$\lim_{n \rightarrow \infty} X_n = 0 \text{ in probability}$$

and

$$\lim_{n \rightarrow \infty} \mathbb{E}X_n^+ = \infty.$$

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

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

Similar theorems for almost sure convergence can be found in [3].

2. Main results. The following lemma has been proved in [3]:

Lemma 3. Let X be an integrable random variable. For any simple random variable Y of the form

$$Y = \sum_{i=1}^k \alpha_i \mathbf{1}_{A_i} + \beta \mathbf{1}_B, \quad \emptyset \subsetneq B \subsetneq \Omega$$

satisfying

$$\begin{aligned} & \sum_{i=1}^k |\alpha_i| P(A_i) + \max_{i=1, \dots, k} |\alpha_i| P(B) \\ & \leq \min \{ \mathbb{E}X^+ \mathbf{1}_B - \mathbb{E}X^- \mathbf{1}_{B^c}, \mathbb{E}X^- \mathbf{1}_B - \mathbb{E}X^+ \mathbf{1}_{B^c} \} \end{aligned}$$

there exists a σ -field \mathfrak{A} such that

$$\mathbb{E}(X | \mathfrak{A})(\omega) = Y(\omega) \quad \text{a.s. for } \omega \in B^c.$$

Now let us prove the following:

Theorem 4. Let (X_n) be a sequence of integrable random variables such that for some sequence of events (B_n) we have

$$(1) \quad \lim_{n \rightarrow \infty} P(B_n^c) = 1$$

and

$$(2) \quad \mathbb{E}X_n^+ \mathbf{1}_{B_n} - \mathbb{E}X_n^- \mathbf{1}_{B_n^c} \rightarrow \infty \text{ for } n \rightarrow \infty,$$

$$(3) \quad \mathbb{E}X_n^- \mathbf{1}_{B_n} - \mathbb{E}X_n^+ \mathbf{1}_{B_n^c} \rightarrow \infty \text{ for } n \rightarrow \infty.$$

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

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

Proof. For sequences (X_n) and (B_n) satisfying (1), (2) and (3) we have

$$\min \{ \mathbb{E}X_n^+ \mathbf{1}_{B_n} - \mathbb{E}X_n^- \mathbf{1}_{B_n^c}, \mathbb{E}X_n^- \mathbf{1}_{B_n} - \mathbb{E}X_n^+ \mathbf{1}_{B_n^c} \} \rightarrow \infty \text{ for } n \rightarrow \infty.$$

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

$$Y_n = \sum_{i=1}^{k(n)} \alpha_i(n) \mathbf{1}_{A_i(n)} + \beta_n \mathbf{1}_{B_n}$$

such that

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

and

$$\max_{i=1, \dots, k(n)} |\alpha_i(n)| \leq \min \{ \mathbb{E}X_n^+ \mathbf{1}_{B_n} - \mathbb{E}X_n^- \mathbf{1}_{B_n^c}, \mathbb{E}X_n^- \mathbf{1}_{B_n} - \mathbb{E}X_n^+ \mathbf{1}_{B_n^c} \}.$$

Lemma 3 implies now the existence of a sequence (\mathfrak{A}_n) of σ -fields such that

$$\mathbb{E}(X_n | \mathfrak{A}_n)(\omega) = Y_n(\omega) \text{ a.s. for } \omega \in B_n^c.$$

Since $\lim_{n \rightarrow \infty} P(B_n^c) = 1$, we finally get

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

which ends the proof of the theorem. \square

Theorem 5. Let (p_n) be a sequence of probability distributions for which there exist sequences (a_n) and (b_n) of nonnegative real numbers satisfying

$$\lim_{n \rightarrow \infty} p_n((-\infty, -b_n) \cup (a_n, \infty)) = 0$$

and

$$\int_{(a_n, \infty)} x dp_n(x) + \int_{[-b_n, 0]} x dp_n(x) \rightarrow \infty \text{ for } n \rightarrow \infty,$$

$$\int_{(-\infty, -b_n)} x dp_n(x) + \int_{[0, a_n]} x dp_n(x) \rightarrow -\infty \text{ for } n \rightarrow \infty.$$

Then for any sequence (X_n) of integrable random variables such that $p_{X_n} = p_n$ and any random variable Y there exists a sequence (\mathfrak{A}_n) of σ -fields satisfying

$$\mathbb{E}(X_n | \mathfrak{A}_n) \rightarrow Y \text{ in probability.}$$

Proof. Under the assumptions of the theorem we put

$$B_n = X_n^{-1} [(-\infty, -b_n) \cup (a_n, \infty)].$$

Now the conclusion follows from Theorem 4. \square

Theorem 6. Let (X_n) be a sequence of integrable random variables such that the sequence of distributions (p_{X_n}) is tight and

$$(4) \quad \lim_{n \rightarrow \infty} \mathbb{E}X_n^+ = \lim_{n \rightarrow \infty} \mathbb{E}X_n^- = \infty.$$

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

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

Proof. The fact that the sequence (p_{X_n}) is tight means that for any $\varepsilon > 0$ there exists $a > 0$ such that

$$P(|X_n| < a) > 1 - \varepsilon \text{ for } n \geq 1.$$

Let (a_j) be a sequence of real numbers such that

$$P(|X_n| < a_k) > 1 - 2^{-k} \text{ for } n \geq 1.$$

We put

$$B_{n,k} = \{|X_n| \geq a_k\}.$$

From (4) we easily get

$$\mathbb{E}X_n^+ \mathbf{1}_{B_{n,k}} - \mathbb{E}X_n^- \mathbf{1}_{B_{n,k}^c} \rightarrow \infty \text{ for } n \rightarrow \infty$$

and

$$\mathbb{E}X_n^- \mathbf{1}_{B_{n,k}} - \mathbb{E}X_n^+ \mathbf{1}_{B_{n,k}^c} \rightarrow \infty \text{ for } n \rightarrow \infty.$$

Now let (n_k) be an increasing sequence of integers such that

$$(5) \quad \mathbb{E}X_n^+ \mathbf{1}_{B_{n,k}} - \mathbb{E}X_n^- \mathbf{1}_{B_{n,k}^c} \geq k \text{ for } n \geq n_k$$

and

$$(6) \quad \mathbb{E}X_n^- \mathbf{1}_{B_{n,k}} - \mathbb{E}X_n^+ \mathbf{1}_{B_{n,k}^c} \geq k \text{ for } n \geq n_k.$$

Let us put

$$B_n = B_{n,k} \text{ for } n_k \leq n < n_{k+1}.$$

From (5) and (6) it follows that

$$\mathbb{E}X_n^+ \mathbf{1}_{B_n} - \mathbb{E}X_n^- \mathbf{1}_{B_n^c} \rightarrow \infty \text{ for } n \rightarrow \infty$$

and

$$\mathbb{E}X_n^- \mathbf{1}_{B_n} - \mathbb{E}X_n^+ \mathbf{1}_{B_n^c} \rightarrow \infty \text{ for } n \rightarrow \infty.$$

We also easily observe that

$$\lim_{n \rightarrow \infty} P(B_n^c) = 1.$$

The conclusion of the theorem is a direct consequence of Theorem 2.7. \square

The following lemma has been proved in [3]:

Lemma 7. *Let (p_n) be a sequence of probability distributions on the real line weakly convergent to a probability distribution p satisfying*

$$\int_0^\infty tp(dt) = - \int_{-\infty}^0 tp(dt) = \infty.$$

Then

$$\lim_{n \rightarrow \infty} \int_0^\infty tp_n(dt) = - \lim_{n \rightarrow \infty} \int_{-\infty}^0 tp_n(dt) = \infty.$$

The next proposition provides quite a large class of sequences for which condition (α) holds.

Proposition 8. *Let (X_n) be a sequence of integrable random variables weakly convergent to a random variable X such that*

$$\mathbb{E}X^+ = \mathbb{E}X^- = \infty.$$

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

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

Proof. It is well known (see for instance [1]) that if (X_n) is a weakly convergent sequence of random variables then the sequence of probability distributions (p_{X_n}) is tight. The above lemma implies that

$$\lim_{n \rightarrow \infty} \mathbb{E}X_n^+ = \lim_{n \rightarrow \infty} \mathbb{E}X_n^- = \infty.$$

The conclusion follows now from Theorem 6. \square

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received November 21, 2001