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CONVERGENCE OF CONDITIONAL EXPECTATIONS

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Abstract: A simple lemma in which uniform integrability together with convergence in distribution implies convergence in probability is presented. The result provides a generalization to that of D. Gilat (1971) and Štěpán (1971).

Key words and phrases: Bayes estimator, uniform integrability, convergence in distribution, convergence in probability.

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The purpose of this note is to present a result in which uniform integrability together with convergence in distribution implies convergence in probability. The result, which provides a generalization to that of D. Gilat (1971), is designed to show that the sequence of Bayes estimators of a real valued function is consistent with respect to L_r -convergence ($r \geq 1$) if and only if it is consistent with respect to convergence in distribution. Our main result is

Lemma. Let $\{X_n\}$, $\{Y_n\}$ be sequences of integrable random variables such that X_n, Y_n are defined on a probability

1) Part of this work was performed while the author was visiting the Mathematical Institute of the University of Aarhus, Denmark.

space $(\Omega_n, \mathcal{A}_n, P_n)$. Suppose that $E[X_n | \mathcal{E}_n] \leq Y_n$ ¹⁾ where $\mathcal{E}_n \subset \mathcal{A}_n$, $n \geq 1$, are σ -algebras and assume the sequences $\{X_n^-\}$, $\{Y_n^-\}$ to be uniformly integrable. If X_n and Y_n have the same limiting distribution then $X_n - Y_n \xrightarrow{P} 0$ ²⁾.

Moreover, if

(1) $E[X_n | \mathcal{E}_n] = Y_n$, $n \geq 1$ and $|X_n|^r$ is uniformly integrable for some $r \geq 1$,

so is $|Y_n|^r$; hence this lemma implies $E|X_n - Y_n|^r \rightarrow 0$ as $n \rightarrow \infty$.

Proof of Lemma. First ³⁾ consider the stronger set of assumptions (1) putting there $r = 1$. Fix a positive integer k and define Φ by

$$\begin{aligned} \Phi(t) &= t^2 & 0 \leq t \leq k \\ &= 2kt - k^2 & t > k \\ &= \Phi(-t) & t < 0. \end{aligned}$$

Φ is continuous, linear for $|t| \geq k$. Hence the uniform integrability argument (Loeve (1963), page 183) applies to conclude from our assumptions that $E\Phi(X_n) - E\Phi(Y_n) \rightarrow 0$ as $n \rightarrow \infty$.

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- 1) Equalities and inequalities between random variables are meant in the almost sure sense.
 - 2) We write $X_n - Y_n \xrightarrow{P} 0$ and mean that $X_n - Y_n \rightarrow 0$ in probability as $n \rightarrow \infty$, i.e. $P_n[|X_n - Y_n| \geq \varepsilon] \rightarrow 0$ as $n \rightarrow \infty$ for all $\varepsilon > 0$.
 - 3) The method employed in the first part of this proof is due to the referee of the present note. The author's original proof was much more complicated.

Further define Ψ by

$$\begin{aligned} \Psi(x, t) &= 2xt - x^2 & |x| \leq k, t \in \mathbb{R}^1 \\ &= 2kt - k^2 & x > k, t \in \mathbb{R}^1 \\ &= -2kt - k^2 & x < -k, t \in \mathbb{R}^1; \end{aligned}$$

i.e. $t \rightarrow \Psi(x, t)$ is the unique linear function which is $\leq \Phi$ and equal Φ at the point x . Moreover, for any given $\epsilon > 0$ there is some $\sigma > 0$ such that

$$\Phi(t) - \Psi(x, t) \geq \sigma \quad \text{if } |x - t| \geq \epsilon \quad \text{and } |x| \leq k - 1.$$

Since

$$E[\Psi(Y_n, X_n) | \epsilon_n] = \Phi(Y_n), \quad n \geq 1$$

we arrive at

$$[E\Phi(X_n) - E\Phi(Y_n)] \geq \sigma P_n[|X_n - Y_n| \geq \epsilon, |Y_n| \leq k - 1] \rightarrow 0$$

as $n \rightarrow \infty$. Letting $k \rightarrow \infty$ it is easy to argue from the tightness of the sequence $\{Y_n\}$ that $X_n - Y_n \xrightarrow{p} 0$.

Finally, consider $\{X_n\}$, $\{Y_n\}$ satisfying the hypotheses of Lemma. Take $c > 0$ and put

$$\begin{aligned} \Delta(t) &= t & t \leq c \\ &= c & t > c. \end{aligned}$$

The conditional form of Jensen's inequality (Loeve (1963), page 348) provides the argument for the inequality

$$Z_n = E[\Delta(X_n) | \epsilon_n] \leq \Delta(Y_n) \quad n \geq 1$$

since Δ is continuous concave and nondecreasing. From the uniform integrability of $\{X_n^-\}$, $\{Y_n^-\}$ it follows that $E\Delta(X_n) - E\Delta(Y_n) \rightarrow 0$ as $n \rightarrow \infty$. Consequently

$$(2) \quad Z_n - \Delta(Y_n) \xrightarrow{p} 0.$$

To prove that $\Delta(X_n) - \Delta(Y_n) \xrightarrow{p} 0$, which is obviously sufficient for our purposes, we simply apply the proven part

of this lemma to the sequences $\{Z_n\}$, $\{\Delta(X_n)\}$ ($\Delta(X_n)$ is uniformly integrable) and combine the result with (2).

The following example shows that our lemma is not necessarily true if its uniform integrability assumptions are not satisfied. Let the (Ω, \mathcal{A}, P) be the closed unit interval with Lebesgue measure. Denote by I_A the indicator of a set A and put for $n \geq 1$

$$A_n = [0, \frac{1}{2n}), \quad B_n = [\frac{1}{2n}, \frac{1}{2}), \quad C_n = [\frac{1}{2}, 1 - \frac{1}{2n}),$$

$$D_n = [1 - \frac{1}{2n}, 1],$$

$$X_n = -n \cdot I_{A_n} + I_{C_n} + n \cdot I_{D_n}, \quad \varepsilon_n = \sigma(A_n \cup C_n, B_n \cup D_n),$$

$$Y_n = E[X_n | \varepsilon_n].$$

Simple computations show that the sequences X_n, Y_n have the same limiting distribution but the sequence $X_n - Y_n$ fails to converge in probability to zero.

A pair of random variables is said to be fair (subfair) if $E[X|Y] = Y$ ($E[X|Y] \leq Y$). D. Gilat (1971) introduced this concept and proved that if (Y, X) is a subfair pair of integrable random variables then Y and X have the same distribution if and only if $X = Y$. Obviously, our Lemma provides a generalization to this result.

As a corollary we obtain the following comparison of L_r -convergence and convergence in distribution:

Corollary 1 (J. Štěpán (1971)). Consider random variables X, X_1, X_2, \dots whose r -th ($r \geq 1$) absolute moments are finite such that $X_n \rightarrow X$ in distribution as $n \rightarrow \infty$. Then $E|X_n - X|^r \rightarrow 0$ if and only if $E|E[X|X_n] - X_n|^r \rightarrow 0$ as $n \rightarrow \infty$.

Finally, consider a parameter-space Θ which is endowed with a priori probability distribution μ defined on a σ -algebra \mathcal{B} of its subsets and have a sequence of statistical problems where the n -th term of the sequence consists of a measurable sample space (Z_n, ϵ_n) and a family of probability measures $\{P_{n\theta}, \theta \in \Theta$ which are defined on ϵ_n . Moreover, suppose that the mapping $P_{n\theta}(E): \Theta \rightarrow R^1$ is measurable for $E \in \epsilon_n$.

The objects under consideration determine a sequence of probability spaces $(\Omega_n, \mathcal{A}_n, P_n)$, $n \geq 1$ where

$$\Omega_n = Z_n \times \Theta, \quad \mathcal{A}_n = \epsilon_n \times \mathcal{B} \quad \text{and}$$

$$P_n(E \times B) = \int_B P_{n\theta}(E) \mu(d\theta) \quad E \in \epsilon_n, B \in \mathcal{B}.$$

Considering $f: \Theta \rightarrow R^1$, a measurable and integrable function, the sequence of conditional expectations

$$b_n(f) = E_{P_n} [f | \epsilon_n] \quad n \geq 1$$

is called the Bayes estimator of f . (By ϵ_n we mean the natural extension of the original σ -algebra such that $\epsilon_n \subset \mathcal{A}_n$.)

Thus, we may apply the assertion of Lemma to get

Corollary 2. Consider $r \geq 1$ and a function $f: \Theta \rightarrow R^1$ such that $|f|^r$ is integrable. Then the Bayes estimator converges to f in distribution if and only if

$$\lim_{n \rightarrow \infty} E_{P_n} |b_n(f) - f|^r = 0.$$

R e f e r e n c e s

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