

## LETTERS TO THE EDITOR

### A NOTE ON TWO MEASURES OF DEPENDENCE AND MIXING SEQUENCES

MAGDA PELIGRAD,\* *University of Rome*

#### Abstract

In this note we establish an inequality between the maximal coefficient of correlation and the  $\varphi$ -mixing coefficient which is symmetric in its arguments. Motivated by this inequality, we introduce a mixing coefficient which is the product of two  $\varphi$ -mixing coefficients.

We also study an invariance principle under conditions imposed on this new mixing coefficient. As a consequence of this result it follows that the invariance principle holds when either the direct-time process or its time-reversed process is  $\varphi$ -mixing; when both processes are  $\varphi$ -mixing the invariance principle holds for sequences of  $L_2$ -integrable random variables under a mixing rate weaker than that used by Ibragimov.

#### MAXIMAL COEFFICIENT OF CORRELATION

Let  $(\Omega, K, P)$  be a probability space and  $K_1$  and  $K_2$  two  $\sigma$ -algebras contained in the  $\sigma$ -algebra  $K$ . Define the measures of dependence between  $K_1$  and  $K_2$  as follows:

$$\varphi(K_1, K_2) = \sup_{\{A \in K_1, P(A) \neq 0, B \in K_2\}} |P(B | A) - P(B)|$$

and

$$\rho(K_1, K_2) = \sup_{\substack{\{X \in L_2(K_1)\} \\ \{Y \in L_2(K_2)\}}} \frac{|E(X - EX)(Y - EY)|}{E^{\frac{1}{2}}(X - EX)^2 E^{\frac{1}{2}}(Y - EY)^2}.$$

The following well-known inequality ([5], Theorem 17.2.3, p. 309) relates the two measures of dependence.

Suppose  $X$  is a random variable  $K_1$ -measurable and  $Y$  a random variable  $K_2$ -measurable and  $E^{1/p} |X|^p < \infty$ ,  $E^{1/q} |Y|^q < \infty$ , where  $1/p + 1/q = 1$ . Then

$$(1) \quad |EXY - EX \cdot EY| \leq 2(\varphi(K_1, K_2) E |X|^p)^{1/p} (E |Y|^q)^{1/q}$$

whence

$$(2) \quad \rho(K_1, K_2) \leq 2\varphi^{\frac{1}{2}}(K_1, K_2).$$

We notice that in (2)  $\varphi$  is not symmetric in its arguments whereas  $\rho$  is. We shall

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\* Postal address: Istituto Matematico 'Guido Castelnuovo', Università di Roma, Piazzale Aldo Moro, Città Universitaria, 00100 Roma, Italy.

establish the following symmetric inequality which improves (1):

$$(3) \quad |EXY - EX \cdot EY| \leq 2(\varphi(K_1, K_2)E|X|^p)^{1/p}(\varphi(K_2, K_1)E|Y|^q)^{1/q},$$

whence

$$(4) \quad \rho(K_1, K_2) \leq 2\varphi^{\frac{1}{2}}(K_1, K_2)\varphi^{\frac{1}{2}}(K_2, K_1).$$

*Proof of (3).* The proof of (3) follows in the same way as the proof of (1). We approximate  $X$  and  $Y$  by  $X = \sum_i a_i I(A_i)$ ,  $Y = \sum_j b_j I(B_j)$ , where  $(A_i)_i$  and  $(B_j)_j$  are respectively, finite decompositions of  $\Omega$  into disjoint elements of  $K_1$  and  $K_2$  and  $I(A)$  denotes the indicator function of  $A$ . Using Hölder's inequality we obtain

$$\begin{aligned} |EXY - EX \cdot EY| &\leq \left( \sum_i |a_i|^p P(A_i) \right)^{1/p} \\ &\quad \times \left[ \sum_i P(A_i) \left( \sum_j |b_j| |P(B_j | A_i) - P(B_j)| \right)^q \right]^{1/q} \\ &\leq (E|X|^p)^{1/p} \left[ \sum_i P(A_i) \times \left( \sum_j |b_j|^q |P(B_j | A_i) - P(B_j)| \right) \right] \\ &\quad \times \left( \sum_j |P(B_j | A_i) - P(B_j)| \right)^{\frac{q}{p}} \leq (E|X|^p)^{1/p} (E|Y|^q)^{1/q} \\ &\quad \times \max_i \left( \sum_j |P(B_j | A_i) - P(B_j)| \right)^{1/p} \max_j \left( \sum_i |P(A_i | B_j) - P(A_i)| \right)^{\frac{1}{2}}. \end{aligned}$$

If  $C_i^+$  (or  $C_i^-$ ) is the union of those  $B_j$  for which  $P(B_j | A_i) - P(B_j)$  is positive, (or non-positive) then

$$\sum_j |P(B_j | A_i) - P(B_j)| \leq |P(C_i^+ | A_i) - P(C_i^+)| + |P(C_i^- | A_i) - P(C_i^-)| \leq 2\varphi(K_1, K_2).$$

Similarly

$$\sum_i |P(A_i | B_j) - P(A_i)| \leq 2\varphi(K_2, K_1)$$

so (3) holds for simple random variables, and by passing to the limit the inequality remains valid for every  $X \in L_p(K_1)$  and  $Y \in L_q(K_2)$ .

Suppose now  $(X_n, n = 0, \pm 1, \pm 2, \dots)$  is a stationary sequence of random variables and denote by  $F_n^m = \sigma(X_k, n \leq k < m)$ . For each  $n \in N$  define

$$\varphi(n) = \varphi(F_{-\infty}^0, F_n^\infty)$$

$$\rho(n) = \rho(F_{-\infty}^0, F_n^\infty).$$

The sequence  $(X_n)_{n \in \mathbb{Z}}$  is said to be  $\varphi$ -mixing, or  $\rho$ -mixing, respectively, as  $\varphi(n) \rightarrow 0$  or  $\rho(n) \rightarrow 0$ . It is known that there are sequences of random variables that are not  $\varphi$ -mixing, while their reverses are, (see [6], p. 414). For instance let  $(X_n, n = 0, \pm 1, \pm 2, \dots)$  be a stationary Markov chain with transition matrix  $A_{i,j} = 2^{-j}$  and  $A_{i,i-1} = 1$  for  $j, i \geq 1$ . This sequence is not  $\varphi$ -mixing, but its reversed-time sequence, with transition matrix  $B_{i,1} = B_{i,i+1} = \frac{1}{2}$  for all  $i$ , is  $\varphi$ -mixing. Therefore it seems natural to ask if the properties valid for  $\varphi$ -mixing sequences are valid for sequences of random variables with the time-reversed sequence  $\varphi$ -mixing, and the fact that both the direct and the reversed sequence are  $\varphi$ -mixing can improve on the  $\varphi$ -mixing rate in certain limit theorems.

The new relation between  $\rho$  and  $\varphi$  suggests that instead of the mixing coefficient  $\varphi(n)$  we can consider another one, namely the product

$$\varphi(n)\varphi^r(n) = \varphi(F_{-\infty}^0, F_n^\infty)\varphi(F_n^\infty, F_{-\infty}^0).$$

The following theorem gives an invariance principle for stationary sequences of  $L_2$ -integrable random variables under conditions imposed on this new mixing coefficient. From this result we deduce that the invariance principle obtained by Ibragimov [4], Theorem (3.2), also holds for stationary sequences of  $L_2$ -integrable random variables whose time-reversed sequences satisfy a  $\varphi$ -mixing condition. When both the direct-time sequence and its reverse are  $\varphi$ -mixing the  $\varphi$ -mixing rate used in [4], Theorem (3.2), is improved (for instance for reversible  $\varphi$ -mixing sequences). This theorem also yields a functional form for Corollary 5.3. (i) of [3], which is a central limit theorem for sequences of random variables whose reversed-time sequences are  $\varphi$ -mixing. At the same time the mixing rate used there (polynomial) is improved (logarithmic).

Let  $S_n = \sum_{i=1}^n X_i$ , and let  $[t]$  denote the greatest integer  $\leq t$ .

*Theorem.* Let  $(X_n, n = 0, \pm 1, \pm 2, \dots)$  be a stationary sequence of centered random variables which have  $L_2$ -moments and  $ES_n^2 \rightarrow \infty$ . Suppose also that

$$(5) \quad \sum_i [\varphi(2^i)\varphi'(2^i)]^{\frac{1}{2}} < \infty.$$

Then there exists  $\sigma^2, 0 < \sigma^2 < \infty$  such that  $\lim_n ES_n^2/n = \sigma^2$ , and the normalised sample paths  $W_n(t) = S_{[nt]}/n^{\frac{1}{2}}\sigma, (0 \leq t \leq L)$  converge in distribution to the standard Brownian motion process  $W(t), (0 \leq t \leq 1)$ .

*Proof.* By (4) and (5) we have  $\sum_i \rho(2^i) < \infty$ , and, using Theorem 1 in [2], or Theorem (4.1) in [7], we obtain that  $ES_n^2/n$  converges to a positive constant  $\sigma^2 > 0$ . The theorem follows by applying Theorem 19.2 of [1]. First  $W_n(t)$  has asymptotically independent increments (see the proof of Theorem 20.1 of [1]). Then, by Lemma (3.5) of [7] it follows that  $(S_n^2/n, n \geq 1)$  is uniformly integrable, so  $W_n^2(t)$  is uniformly integrable for each  $t$  and obviously  $EW_n(t) = 0$  and  $EW_n^2(t) \xrightarrow{n \rightarrow \infty} t$ . It remains only to verify the tightness condition, namely that for each  $\epsilon > 0$ , there exists  $\lambda > 1$  and an integer  $n_0$  such that  $n \geq n_0$  implies  $P(\max_{1 \leq i \leq n} |S_i| > \lambda \sigma n^{\frac{1}{2}}) \leq \epsilon/\lambda^2$ . Without loss of generality we assume  $\sigma^2 = 1$ . If  $\varphi_n \rightarrow 0$  this condition was verified in [1], pp. 175–176. If  $\varphi'_n \rightarrow 0$ , the proof follows the same lines with the difference that we now denote

$$E_i^n = \left\{ \max_{0 \leq j < i} |S_n - S_j| < 3\lambda n^{\frac{1}{2}} \leq |S_n - S_i| \right\} \in F_i^n.$$

So, we have successively:

$$\begin{aligned} P\left(\max_{i \leq n} |S_i| > 4\lambda n^{\frac{1}{2}}\right) &\leq P(|S_n| > \lambda n^{\frac{1}{2}}) + P\left(\max_{i \leq n-1} |S_n - S_i| > 3\lambda n^{\frac{1}{2}}\right) \\ &\leq 2P(|S_n| > \lambda n^{\frac{1}{2}}) + \sum_{i=1}^{n-1} P(E_i^n \cap \{|S_i| > 2\lambda n^{\frac{1}{2}}\}) \leq 2P(|S_n| > \lambda n^{\frac{1}{2}}) \\ &\quad + \sum_{i=1}^p P(|S_i| > 2\lambda n^{\frac{1}{2}}) + \sum_{i=p+1}^{n-1} P(|S_i - S_{i-p}| > \lambda n^{\frac{1}{2}}) + \sum_{i=p+1}^{n-1} P(E_i^n \cap \{|S_{i-p}| > \lambda n^{\frac{1}{2}}\}) \\ &\leq 2P(|S_n| > \lambda n^{\frac{1}{2}}) + nP(S_p^* > \lambda n^{\frac{1}{2}}) + \sum_{i=p+1}^{n-1} P(E_i^n)(P(|S_{i-p}| > \lambda n^{\frac{1}{2}}) + \varphi'(p)) \end{aligned}$$

where  $p$  and  $S_p^*$  were defined in [1], p. 175. This gives the desired result.

With a similar proof it is easy to see the following.

*Remark.* This theorem can be obtained for some non-stationary sequences of random variables  $(X_n, n \geq 1)$ , namely, we can assume instead of stationarity that  $(X_n^2, n \geq 1)$  is uniformly integrable and  $E\left(\sum_{i=kn}^{(k+1)n} X_i\right)^2 / ES_n^2 \rightarrow 1$  as  $n \rightarrow \infty$  uniformly in  $k$ , the mixing coefficients  $\varphi(n)$  and  $\varphi^r(n)$  being defined by

$$\varphi(n) = \sup_m \varphi(F_0^m, F_{m+n}^\infty) \quad \text{and} \quad \varphi^r(n) = \sup_m \varphi(F_{m+n}^\infty, F_0^m).$$

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