

Let  $K_w = R_w I_w - Q_{\text{even}}$  for  $w \in I$ . Then

$$K_w(f) = \sum_{k=0}^{\infty} \frac{1}{(2k)!} f^{(2k)}(0) \cdot (R_w I_w(\chi_{2k}) - \chi_{2k}) \quad \text{for } f \in A(D).$$

Thus  $K_w$  is compact (even nuclear). Hence the family  $(R_w, I_w)_{w \in I}$  satisfies condition (2.8) of Proposition 2(a). This is an easy consequence of the following facts.

(i) For every  $w \in I$ , if  $z \in B_2 \setminus (F_w \cup F_w^*)$ , then there exists a  $q(z, w)$  such that  $0 \leq q(z, w) < 1$  and  $|R_w I_w(\chi_{2k})(z)| \leq 2q(z, w)^{2k}$  for  $k = 1, 2, \dots$ ;

(ii) If  $w_1 \neq w_2$ , then  $(F_{w_1} \cup F_{w_1}^*) \cap (F_{w_2} \cup F_{w_2}^*) \cap \partial B_2 = \emptyset$  ( $w_1 \in I, w_2 \in I$ ). The desired conclusion follows from Proposition 2a.

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#### The invariance principle for group-valued random variables

by

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**Abstract.** Let  $G$  be a complete, separable metric group. We extend the invariance principle to certain triangular arrays of  $G$ -valued random variables. As an application we examine the invariance principle for triangular arrays of random variables with values in a Fréchet space or in a locally compact abelian group.

Let  $X_1, X_2, \dots$  be a sequence of independent identically distributed random variables defined on a probability space  $(\Omega, \sigma, P)$ . For each integer  $n \geq 1$ , let  $\xi_n(t)$ ,  $0 \leq t \leq 1$ , denote the stochastic process obtained by the linear interpolation of the normalized sums  $S_k/n^{1/2} = (X_1 + X_2 + \dots + X_k)/n^{1/2}$ ,  $k = 1, \dots, n$ ;  $S_0 = 0$ . The invariance principle asserts that if  $X_n$  have mean 0 and variance 1, then the sequence of probability measures on  $C[0, 1]$ , induced by the stochastic processes  $\xi_n(t)$ , converges weakly to the Wiener measure (see e.g. [1]). This result has been generalized by J. Kuelbs to triangular arrays of Banach space valued random variables [8]. In this note we prove that the invariance principle has a more general scope. Let  $G$  be a Polish (complete separable metric) group. We discuss the invariance principle for certain triangular arrays of  $G$ -valued random variables. Instead of  $C[0, 1]$  we use the space  $D_G \subset C^{[0, 1]}$  of all left-continuous functions having the right-hand limits (at every point of  $[0, 1]$ ). As an application of our result we examine the invariance principle for Fréchet space valued random variables and for random variables taking values in any LCA group.

**Preliminaries.** All statements which we quote here are proved in [1] under the assumption that  $G$  is the real line. However, the arguments used there either apply to our general situation without any change or need a slight modification. We shall not repeat those proofs.

Throughout this paper we shall assume that  $G$  is a complete separable metric group having a left-invariant and complete metric  $\varrho$ . It is well known that every two-sided invariant metric is complete whenever  $G$  is topologically complete.

Let  $\|\cdot\|$  denote the distance from the identity  $e$  of  $G$ :  $\|x\| = \varrho(x, e)$ .

Let  $D_G = D_G[0, 1]$  be the space of functions  $f$  on  $[0, 1]$  into  $G$  that are right-continuous and have left-hand limits:

(i) for  $0 \leq t < 1$ ,  $f(t+) = \lim_{s \downarrow t} f(s)$  exists and  $f(t+) = f(t)$ ,

(ii) for  $0 < t \leq 1$ ,  $f(t-) = \lim_{s \uparrow t} f(s)$  exists.

Let  $\mathcal{A}$  denote the class of strictly increasing, continuous mappings of  $[0, 1]$  onto itself taking 0 onto 0. For  $f$  and  $g$  in  $D_G$ , define  $d(f, g)$  to be the infimum of those positive  $\varepsilon$  for which there exists in  $\mathcal{A}$  a  $\lambda$  such that

$$\sup_t |\lambda t - t| \leq \varepsilon$$

and

$$\sup_t \|(f(t))^{-1} g(\lambda t)\| \leq \varepsilon.$$

The space  $D_G$  is the separable, topologically complete metric space in the topology generated by the metric  $d$  (the so-called *Skorohod topology*).

We say that a mapping  $\xi$  from a probability space  $(\Omega, \sigma, P)$  into  $D_G$  is a *random element* iff it is measurable with respect to the Borel  $\sigma$ -field of  $D_G$ . Let us define, for  $t_1, t_2, \dots, t_k$ ;  $t_i \in [0, 1]$ , the natural projection  $\pi_{t_1, \dots, t_k}$  from  $D_G$  to  $G^k$ :

$$\pi_{t_1, \dots, t_k}(f) = (f(t_1), f(t_2), \dots, f(t_k)).$$

By *finite-dimensional sets* we shall mean sets of the form  $\pi_{t_1, \dots, t_k}^{-1}(H)$ , where  $H \in \mathcal{B}_{G^k}$  (the Borel  $\sigma$ -field in  $G^k$ ).

For the random element  $\xi$ , by  $\xi(t)$  we shall denote the composition of  $\xi$  and  $\pi_t$ :

$$\xi(t) = \pi_t(\xi).$$

From the usual arguments regarding this subject, it follows that  $\xi$  is a random element of  $D_G$  iff  $\xi(t)$  is a random variable of  $G$  for each  $t \in [0, 1]$  (i.e.,  $\xi(t)$  is the Borel measurable mapping from the probability space into  $G$ ).

Finally, we say that a random element  $\xi$  has *independent increments* iff, for each  $t_1, t_2, \dots, t_k$ ;  $t_i \in [0, 1]$  such that  $t_1 < t_2 < \dots < t_k$ , the random variables

$$\xi(t_1)^{-1} \xi(t_2), \xi(t_2)^{-1} \xi(t_3), \dots, \xi(t_{k-1})^{-1} \xi(t_k)$$

are independent.

**The invariance principle.** First of all we state some lemmas needed in the sequel. A proof of the first one may be found e.g. in [5].

**LEMMA 1.** Let  $X_1, X_2, \dots, X_n$  be  $n$  independent,  $G$ -valued random variables and  $S_k = X_1 X_2 \dots X_k$ ,  $k = 1, 2, \dots, n$ ,  $S_0 = e$ . Then

$$P(\{\max_{1 \leq k \leq n} \|S_k\| > 2\varepsilon\}) \leq \alpha^{-1} P(\{\|S_n\| > \varepsilon\})$$

whenever

$$P(\{\|S_k^{-1} S_n\| \leq \varepsilon\}) \geq \alpha > 0 \quad \text{for } k = 0, 1, \dots, n-1.$$

The next lemma, which may be found in [4], gives some property of random elements with independent increments and continuous paths.

**LEMMA 2.** Let  $\xi$  be a random element of  $D_G$  with independent increments. If  $\xi$  has continuous paths with probability one, then

$$\sum_{k=1}^{m_n} P(\{\|\xi(t_{k-1}^{(n)})^{-1} \xi(t_k^{(n)})\| \geq \varepsilon\}) \rightarrow 0$$

whenever  $n \rightarrow \infty$ , for every  $\varepsilon > 0$  and every sequence  $\{t_k^{(n)}\}_{k=1}^{m_n}$  of partitions of  $[0, 1]$  such that  $0 = t_0^{(n)} < t_1^{(n)} < \dots < t_{m_n}^{(n)} = 1$  and  $\lambda_n = \max_{1 \leq k \leq m_n} (t_k^{(n)} - t_{k-1}^{(n)}) \rightarrow 0$  if  $n \rightarrow \infty$ .

The arguments needed to prove the following lemma do not differ essentially from those used in the proofs of Theorems 14.3, 15.2 and 15.5, Chapt. 3 of [1], so we shall not repeat them.

**LEMMA 3.** Let  $\{X_j^{(n)}: j = 1, \dots, n; n = 1, 2, \dots\}$  be a triangular array of  $G$ -valued, independent and identically distributed random variables. Let  $S_k^{(n)} = X_1^{(n)} X_2^{(n)} \dots X_k^{(n)}$ . Suppose that for each  $\varepsilon > 0$  and every  $\eta > 0$  there exist:  $\delta, 0 < \delta < 1$ , a positive integer  $n_0$ , and a family  $\{K_t\}_{t \in S}$  of compact subsets of  $G$ , where  $S$  is a dense subset of  $[0, 1]$  containing 1, such that

$$(a) \frac{1}{\delta} P(\{\max_{1 \leq k \leq n_0} \|S_k^{(n)}\| > \varepsilon\}) \leq \eta \text{ for } n \geq n_0,$$

$$(b) P(\{S_{[nt]}^{(n)} \notin K_t \text{ for } t \in S\}) \leq \eta.$$

$$\text{Let } \xi_n(t) = S_{[nt]}^{(n)}, \xi_n(0) = e.$$

Then the sequence of random elements  $\xi_n$  is tight, and if  $\mu$  is the weak limit of a certain subsequence  $\xi_{n'}$ , then  $\mu(C_G) = 1$ , where  $C_G$  denotes the space of all continuous functions  $\subseteq G^{[0,1]}$ .

The following lemma will also be useful in the sequel.

**LEMMA 4.** Let  $\{X_j^{(n)}: j = 1, \dots, n; n = 1, 2, \dots\}$  be a triangular array of  $G$ -valued, independent and identically distributed random variables. Let  $S_k^{(n)} = X_1^{(n)} X_2^{(n)} \dots X_k^{(n)}$ . Suppose that

(1)  $S_{[nt]}^{(n)}$  converges weakly for  $t \in S$ , where  $S$  is a dense subset of  $[0, 1]$  containing 1,

(2) for each  $\varepsilon > 0$  and a certain sequence  $\delta_n$ ,  $0 < \delta_n < 1$ ,  $\delta_n \rightarrow 0$

$$\frac{1}{\delta_k} \limsup_n P(\{\|S_{[n\delta_k]}^{(n)}\| > \varepsilon\}) \rightarrow 0 \quad \text{if} \quad k \rightarrow \infty,$$

(3) for each  $\varepsilon > 0$  there exists  $h > 0$  such that

$$\limsup_n \max_{1 \leq r \leq nh} P(\{\|S_r^{(n)}\| > \varepsilon\}) < 1.$$

Let  $\xi_n(t) = S_{[nt]}^{(n)}$ . Then the sequence of random elements  $\xi_n$  converges weakly to a certain element of  $D_G$  with continuous paths.

**Proof.** First we shall show that conditions (a) and (b) of Lemma 3 are satisfied.

Condition (b) follows directly from assumption (1) and from Prohorov Theorem (see [1]).

Now, let  $\varepsilon, \eta$  be given real, positive numbers. By (3), there exist real positive numbers  $h_1, \alpha$  such that  $\alpha < 1$  and a positive integer  $n_1$  such that

$$\max_{1 \leq r \leq nh_1} P(\{\|S_r^{(n)}\| > \varepsilon\}) \leq \alpha$$

for every  $n \geq n_1$ . Lemma 1 implies that

$$(*) \quad P(\{\max_{1 \leq i \leq n\delta_k} \|S_i^{(n)}\| > \varepsilon\}) \leq \frac{1}{1-\alpha} P\left(\left\{\|S_{[n\delta_k]}^{(n)}\| > \frac{\varepsilon}{2}\right\}\right)$$

for  $n \geq n_1$  and  $\delta_k \leq h_1$ . By (2), there exists a positive integer  $N$  such that if  $k \geq N$  then

$$\limsup_n P\left(\left\{\|S_{[n\delta_k]}^{(n)}\| > \frac{\varepsilon}{2}\right\}\right) < \delta_k \eta (1-\alpha) \quad \text{and} \quad \delta_k \leq h_1.$$

Let  $n_2 = \max(n_1, N)$  and  $\delta = \delta_{n_2}$ . There exists an  $n_0 \geq n_2$  such that

$$(**) \quad P\left(\left\{\|S_{[n\delta]}^{(n)}\| > \frac{\varepsilon}{2}\right\}\right) < \delta \eta (1-\alpha)$$

for  $n \geq n_0$ . From (\*) and (\*\*), we have

$$\frac{1}{\delta} P(\{\max_{1 \leq i \leq n\delta} \|S_i^{(n)}\| > \varepsilon\}) \leq \eta$$

for  $n \geq n_0$  and thus condition (a) of Lemma 3 is satisfied.

Now, let a certain subsequence  $\xi_{n'}$  converge weakly to  $\mu$ . Let  $\xi_{n''}$  be another subsequence converging to a distribution  $\nu$ . Since  $\mu(G_G) = \nu(G_G) = 1$ , it follows (see [1], Chapt. 3, § 15) that, for every  $t \in [0, 1]$ ,  $\xi_{n'}(t)$  and  $\xi_{n''}(t)$  converge weakly to one-dimensional distributions of  $\mu$  and  $\nu$ , respectively. Since, for  $t \in S$ ,  $\xi_n(t)$  also converges weakly, one-dimensional distributions of  $\mu$  and  $\nu$  based on points in  $S$  are equal. Since the limits of  $\xi_{n'}$ ,  $\xi_{n''}$  are random elements with independent increments, the corresponding finite-dimensional distributions of  $\mu$  and  $\nu$  based on points in  $S$  are also equal. Since  $S$  is dense in  $[0, 1]$  and contains 1, we have  $\mu = \nu$  (see Th. 14.5, [1]). This completes the proof.

**PROPOSITION.** Suppose that there exists a homogeneous random element  $W$  on  $D_G$  with independent increments and with continuous paths (with probability one) such that  $W(0) = e$ . Let  $\{X_k^{(n)}; k = 1, 2, \dots, n; n = 1, 2, \dots\}$  be a triangular array of independent, identically distributed random variables with values in  $G$ . Let  $S_k^{(n)} = X_1^{(n)} X_2^{(n)} \dots X_k^{(n)}$ ,  $S_0^{(n)} = e$ . Suppose that

- (i)  $S_{[nt]}^{(n)}$  converges weakly to  $W(t)$  for every  $t \in [0, 1]$ ,
- (ii) for each  $\varepsilon > 0$  there exists  $h > 0$  such that

$$\limsup_n \max_{1 \leq r \leq nh} P(\{\|S_r^{(n)}\| > \varepsilon\}) < 1.$$

Let  $\xi_n(t) = S_{[nt]}^{(n)}$ . Then  $\xi_n$  converges weakly to  $W$ .

**Proof.** From (i) it follows that

$$\limsup_n P(\{\|S_{[nt]}^{(n)}\| \geq \varepsilon\}) \leq P(\{\|W(t)\| \geq \varepsilon\}),$$

which, together with Lemma 2, establishes condition (2) of Lemma 4 and ends the proof.

We now present some applications of these statements.

Let  $G$  be a separable Fréchet space, i.e., a real vector space which is a complete, separable metric group, and is such that the mappings  $\alpha \rightarrow \alpha x$  of  $R$  onto  $G$  and  $x \rightarrow \alpha x$  of  $G$  onto  $G$  are continuous.

We say that the random variable  $X$  with values in  $G$  has a *symmetric Gaussian distribution* iff for any pair  $Y_1, Y_2$  of independent random variables having the distributions of  $X$ , and for every pair of real numbers  $(s, t)$  such that  $s^2 + t^2 = 1$ , the random variables  $sY_1 + tY_2$  and  $tY_1 - sY_2$  are independent and have the distribution of  $X$ . This definition was introduced by Fernique in [4]. If the continuous linear functionals of  $G$  generate the Borel  $\sigma$ -field of  $G$ , then this definition is equivalent to the following one:  $X$  has a *symmetric Gaussian distribution* iff  $f(X)$  is a symmetric real Gaussian random variable, for every continuous linear functional  $f$ .

**THEOREM 1.** Let  $X$  be a symmetric Gaussian random variable on  $G$ . The following conditions are equivalent:

(i) for each  $s > 0$ ,  $nP(\{\|X/n^{1/2}\| \geq s\}) \rightarrow 0$  if  $n \rightarrow \infty$ ,

(ii) there exists a homogeneous Gaussian random element  $W$  on  $D_G$ , with independent increments, having continuous paths with probability one and such that  $W(0) = 0$  and  $W(1)$  has the distribution of  $X$ ,

(iii) if  $\{X_n\}$  is a sequence of independent, identically distributed random variables such that  $(X_1 + X_2 + \dots + X_n)/n^{1/2}$  converges weakly to the distribution of  $X$ , then the sequence  $\xi_n$ ,  $\xi_n(t) = (X_1 + X_2 + \dots + X_{[nt]})/n^{1/2}$  converges weakly to a certain random element on  $D_G$  with continuous paths with probability one.

**Proof.** First we shall prove that (iii) implies (ii). Let  $X_1, X_2, \dots, X_n, \dots$  be a sequence of independent random variables such that  $X_i$  has the distribution of  $X$ . Then  $(X_1 + X_2 + \dots + X_n)/n^{1/2}$  also has the distribution of  $X$ , and if (iii) holds, then  $\xi_n$  converges weakly to a certain random element with the properties as in (ii). Thus (ii) holds.

That (ii) implies (i) follows directly from Lemma 2 and the definition of Gaussian distribution.

Now, suppose that condition (i) holds. Let  $\{X_n\}$  be a sequence of independent, identically distributed random variables such that  $(X_1 + X_2 + \dots + X_n)/n^{1/2}$  converges weakly to the distribution of  $X$ . In order to conclude that  $\xi_n$  converges weakly to a random element with properties as described in (ii), it suffices to show that the assumptions of Lemma 4 are satisfied.

For this purpose let us observe that if

$$\xi_n(t) = \left(\frac{[nt]}{n}\right)^{1/2} \frac{X_1 + \dots + X_{[nt]}}{[nt]^{1/2}}$$

and

$$\xi'_n(t) = t^{1/2} \frac{X_1 + \dots + X_{[nt]}}{[nt]^{1/2}},$$

then  $\xi_n(t) - \xi'_n(t)$  converges to 0 with probability 1, and  $\xi'_n(t)$  converges weakly to the distribution of  $t^{1/2}X$ . Therefore, by [1], Theorem 4.1,  $\xi_n(t)$  also converges weakly to the distribution of  $t^{1/2}X$ , and condition (1) of Lemma 4 holds.

Next, we shall show that for each  $\varepsilon, \eta > 0$  there exists an  $h > 0$  such that

$$(*) \quad \limsup_n \max_{1 \leq r \leq nh} P\left(\left\|\frac{1}{n^{1/2}} \sum_{j=1}^r X_j\right\| \geq \varepsilon\right) \leq \eta.$$

Let

$$Y_n = \frac{X_1 + \dots + X_n}{n^{1/2}}, \quad U = \{x; \|x\| < \varepsilon\}.$$

Since  $Y_n$  converges weakly, there exists a compact subset  $K$  of  $G$  (by the Prohorov Theorem) such that  $P(\{Y_n \notin K\}) < \eta$ . Since  $U$  is

an open neighbourhood of 0, there exists an  $h > 0$  such that for  $t \leq h^{1/2}$  we have  $tK \subset U$  (see [3], Chapt. II). Therefore, for  $r \leq nh$  we have

$$\begin{aligned} \left\{\left\|\frac{1}{n^{1/2}} \sum_{j=1}^r X_j\right\| \geq \varepsilon\right\} &= \left\{\left(\frac{r}{n}\right)^{1/2} \frac{X_1 + \dots + X_r}{r^{1/2}}\right\| \geq \varepsilon\right\} \\ &= \left\{\left(\frac{r}{n}\right)^{1/2} \frac{X_1 + \dots + X_r}{r^{1/2}} \notin U\right\} = \left\{\frac{X_1 + \dots + X_r}{r^{1/2}} \notin \left(\frac{n}{r}\right)^{1/2} U\right\} \subset \{Y_r \notin K\}, \end{aligned}$$

which proves (\*). Thus, condition (3) of Lemma 4 holds.

Finally, since  $(X_1 + X_2 + \dots + X_{[k/n]})/k^{1/2}$  converges to the distribution of  $X/n^{1/2}$  as  $k \rightarrow \infty$ , by condition (i) of this theorem it follows that condition (2) of Lemma 4 is also satisfied. This ends the proof of the theorem.

**Remark.** It is interesting to know which separable Fréchet spaces  $G$  have the property that every symmetric Gaussian random element taking values in  $G$  satisfies condition (i) of Theorem 1. By Theorem 2 in [7] it follows that every separable locally pseudoconvex Fréchet space has this property. The corresponding result for Banach spaces can be derived from an earlier paper of Fernique [4] and also from Landau and Shepp [10]; for locally convex spaces — from Kuelbs [9]. The forthcoming example shows that every symmetric Gaussian random element with values in the space  $S(T, \mathcal{A}, m)$  (all  $\mathcal{A}$ -measurable real functions on  $T$  with the convergence in measure  $m$ ) also satisfies condition (i). The idea of this example is due to Professor C. Ryll-Nardzewski.

However, the author does not know either any characterization of Fréchet spaces having the above property or any example of a symmetric Gaussian random element which does not satisfy (i).

**EXAMPLE.** Let  $(T, \mathcal{A}, m)$  be a finite measure space. Let  $S$  denote the space of all real-valued  $\mathcal{A}$ -measurable functions defined on  $T$ . It is well known that  $S$  with the norm

$$\|f\| = \int_T \frac{|f(t)|}{1 + |f(t)|} dm(t),$$

which induces the topology of convergence in measure  $m$ , is a real Fréchet space. If  $m$  is nonatomic, then  $S$  admits no nonzero continuous linear functionals. Suppose that  $S$  is separable.

Let  $\{\xi(t); t \in T\}$  be a stochastic process defined on a probability space  $(\Omega, \sigma, P)$ ; it is said to be measurable if the map  $\xi, \xi: T \times \Omega \rightarrow \mathbb{R}$  defined by  $(t, \omega) \rightarrow \xi(t, \omega)$  is measurable relative to the  $\sigma$ -algebras  $\mathcal{B}_R$  and  $\mathcal{A} \times \sigma$ . From the measurability of the process and the separability of  $S$  it follows that the mapping  $\omega \rightarrow \xi(\cdot, \omega)$  is measurable relative to  $\mathcal{B}_S$  and  $\sigma$ . The probability measure on  $(S, \mathcal{B}_S)$  induced by this mapping will be denoted by  $\mu_\xi$ .



A stochastic process  $\{\xi(t); t \in T\}$  is said to be *Gaussian* if, for every  $t_1, t_2, \dots, t_k \in T$ ,  $\langle \xi(t_1), \xi(t_2), \dots, \xi(t_k) \rangle$  is a Gaussian random vector with values in  $R^k$ .

The proof that every symmetric Gaussian random element with values in  $(S, \mathcal{B}_S)$  satisfies condition (i) of Theorem 1 is divided into two parts; each of these parts may be of independent interest.

In the first part we shall prove that every symmetric Gaussian measure on  $(S, \mathcal{B}_S)$  is induced by a Gaussian measurable stochastic process. In the second part we shall show that every measurable symmetric Gaussian process is induced by a continuous linear mapping  $\Phi: L_2 \rightarrow S$ , from a measurable Gaussian process with samples in  $L_2$ ; so the induced random element with values in  $S$  is symmetric Gaussian and satisfies condition (i).

I. Let  $\mu$  be a probability measure on  $(S, \mathcal{B}_S)$ . We construct a measurable stochastic process  $\{\xi(t); t \in T\}$  such that the induced measure  $\mu_\xi$  equals  $\mu$ .

For any set  $A \subseteq S$  we denote its diameter by  $\delta(A)$ . Since  $S$  is separable, we can write, for every positive integer  $n$ ,  $S = \bigcup_{k=1}^{\infty} S_k^{(n)}$ , where  $S_k^{(n)}$  are non-empty,  $\delta(S_k^{(n)}) < 1/n$ ,  $S_k^{(n)} \in \mathcal{B}_S$  ( $k = 1, 2, \dots$ ) and  $S_k^{(n)} \cap S_m^{(n)} = \emptyset$  if  $k \neq m$ . Without loss of generality we may assume that  $\{S_k^{(n+1)}\}$  is a refinement of  $\{S_k^{(n)}\}$ . Let us choose, for each  $m$ , an element of  $S_m^{(n)}$ , namely  $h_m^{(n)}$ , and let  $\tilde{h}_m^{(n)}$  be a representative of the equivalence class  $h_m^{(n)}$ . Now, let us define, for  $n = 1, 2, \dots$ ,

$$\xi_n(f, t) = \tilde{h}_k^{(n)}(t) \quad \text{if} \quad f \in S_k^{(n)}.$$

It is easy to check that  $\xi_n: S \times T \rightarrow R$  is  $\mathcal{B}_S \times \mathcal{A}$ -measurable. Next, by the construction of  $\xi_n$  it follows immediately that for  $\varepsilon > 0$

$$m\{t; |\xi_n(f, t) - \xi_m(f, t)| > \varepsilon\} \rightarrow 0$$

if  $n, m \rightarrow \infty$  uniformly with respect to  $f$ . Fubini's Theorem implies that  $\xi_n$  is fundamental in the  $\mu \times m$  measure. So, there is a function  $\xi: S \times T \rightarrow R$ ,  $\mathcal{B}_S \times \mathcal{A}$ -measurable and such that  $\xi_n$  converges to  $\xi$  in the  $\mu \times m$ -measure. Let  $\tilde{\xi}$  denote the mapping of  $S$  into  $S$  induced by  $\xi$  as follows:

$$\tilde{\xi}(f)(\cdot) = \xi(f, \cdot).$$

Let us observe that  $\tilde{\xi} = I$   $\mu$ -a.e., where  $I$  denotes the identity map. For, let  $\xi_{n_k}$  be a subsequence of  $\xi_n$  converging to  $\xi$   $\mu \times m$ -a.e. Then, by Fubini's Theorem,  $\xi_{n_k}(f, \cdot)$  converges  $m$ -a.e. to  $\xi(f, \cdot)$ , for  $\mu$ -almost all  $f$ , and hence in the  $m$ -measure. So,  $\xi_{n_k} \rightarrow \tilde{\xi}$   $\mu$ -a.e. On the other hand,  $\xi_{n_k}(f) \rightarrow I(f)$ , for every  $f \in S$ , which gives the desired conclusion. Next, for every pair  $(s, u)$  of real numbers we have

$$(1) \quad \xi(sf + ug, t) = s\xi(f, t) + u\xi(g, t)$$

for  $\mu \times \mu \times m$ -almost all  $(f, g, t)$ . This follows from the construction of  $\xi_n$  and from the fact that  $S$  is topologized by the norm of convergence in measure  $m$ .

Now, let  $\mu$  be a symmetric Gaussian distribution. Let  $(s, u)$  be a fixed pair of real numbers such that  $0 < s, u < 1$ ,  $s^2 + u^2 = 1$ . Let us define the mapping  $F: S \times S \rightarrow S \times S$ :

$$F(f, g) = (sf + ug, uf - sg).$$

Since  $\mu$  is symmetric Gaussian, we have

$$(2) \quad (\mu \times \mu)(A) = (\mu \times \mu)(F^{-1}(A))$$

for every  $A \in \mathcal{B}_S$ . Let  $N$  be a subset of  $T$  such that  $N \in \mathcal{A}$ ,  $m(N) = 0$  and

$$(3) \quad (\mu \times \mu)\{((f, g); \xi(sf + ug, t_0) \neq s\xi(f, t_0) + u\xi(g, t_0), \\ \xi(uf - sg, t_0) \neq u\xi(f, t_0) - s\xi(g, t_0))\} = 0 \quad \text{for} \quad t_0 \in T \setminus N.$$

Let  $t_1, t_2, \dots, t_k \in T \setminus N$  and let

$$X = (\xi(f, t_1), \xi(f, t_2), \dots, \xi(f, t_k)), \\ Y = (\xi(g, t_1), \xi(g, t_2), \dots, \xi(g, t_k)).$$

Let  $A, B \in \mathcal{B}_{R^k}$  (the Borel  $\sigma$ -field in  $R^k$ ). Let us define

$$A' = \{f; X \in A\}, \quad B' = \{g; Y \in B\}.$$

Clearly, by the measurability of  $\xi$  it follows that  $A', B' \in \mathcal{B}_S$ . By (2), we have

$$(\mu \times \mu)(A' \times B') = (\mu \times \mu)(F^{-1}(A' \times B')).$$

In view of (3) we have

$$\begin{aligned} & (\mu \times \mu)(F^{-1}(A' \times B')) \\ &= (\mu \times \mu)\{((f, g); F(f, g) \in A' \times B')\} \\ &= (\mu \times \mu)\{((f, g); (\xi(sf + ug, t_1), \dots, \xi(sf + ug, t_k)) \in A; (\xi(uf - sg, t_1), \dots, \xi(uf - sg, t_k)) \in B)\} \\ &= (\mu \times \mu)\{((f, g); (sX + uY, uX - sY) \in A \times B)\} \\ &= (\mu \times \mu)\{((f, g); (X, Y) \in F_k^{-1}(A \times B))\} = (\mu_X \times \mu_Y)(F_k^{-1}(A \times B)), \end{aligned}$$

where  $\mu_X, \mu_Y$  denote the distributions of  $X, Y$ , respectively, and  $F_k(w_1, w_2) = (sw_1 + uw_2, uw_1 - sw_2)$  for each  $w_1, w_2 \in R^k$ . Thus we have

$$(\mu_X \times \mu_Y)(A \times B) = (\mu_X \times \mu_Y)(F_k^{-1}(A \times B)).$$

By Theorem 2, § 8. XV, in [5], we infer that  $X$  is a symmetric Gaussian random variable with values in  $R^k$ . Thus, if we define

$$\bar{\xi}(f, t) = \begin{cases} \xi(f, t) & \text{if } t \in T \setminus N, \\ 0 & \text{if } t \in N, \end{cases}$$

then  $\mu_{\bar{\xi}} = \mu_{\xi} = \mu$  and  $\bar{\xi}$  is a symmetric measurable Gaussian process.

II. Next, let  $\{\xi(t); t \in T\}$  be a symmetric measurable Gaussian process. Let us define  $K(s, t) = E(\xi(t)\xi(s))$ . Now, let

$$\eta(\omega, t) = \xi(\omega, t) (1 + K(t, t))^{-1/2}.$$

Then  $E(\eta(t)^2) = K(t, t) (1 + K(t, t))^{-1}$  and so  $\eta(\omega, \cdot) \in L_2(T, \mathcal{A}, m)$  with probability one (see Proposition 3.4 in [12]). Thus, the probability measure  $\mu_{\eta}$  induced on  $L_2$  by  $\eta$  is symmetric Gaussian (one uses the fact that  $L_2$  has sufficiently many continuous functionals and Theorem 3.2 in [12]). Now, let us define a linear mapping from  $L_2$  into  $S$  as follows:

$$\Phi f = f(1 + K)^{1/2}.$$

It is easy to see that  $\Phi$  is continuous,  $\xi(\omega, \cdot) = \Phi\eta(\omega, \cdot)$ , and that  $\mu_{\xi}$  is symmetric Gaussian. Let us denote by  $X, Y$  the random elements induced by the stochastic processes  $\eta, \xi$  on  $L_2, S$ , respectively. Let  $U$  be any open neighbourhood of 0 in  $S$ . Then

$$nP\{Y/n^{1/2} \notin U\} = nP\{\Phi X/n^{1/2} \notin U\} = nP\{X/n^{1/2} \notin \Phi^{-1}U\} \rightarrow 0$$

if  $n \rightarrow \infty$ , since by the continuity of  $\Phi$ ,  $\Phi^{-1}U$  is an open neighbourhood of 0 in  $L_2$  and  $X$  satisfies condition (i) of Theorem 1 (as a symmetric Gaussian random element with values in a Banach space). So,  $Y$  satisfies condition (i) of Theorem 1.

Now, let  $G$  be a locally compact, second countable abelian group and  $\Gamma$  its character group. The random variable  $X$  with values in  $G$  has a symmetric Gaussian distribution if its characteristic function has the form

$$\hat{\mu}(\gamma) = \exp(-\varphi(\gamma)),$$

where  $\gamma \in \Gamma$  and  $\varphi$  is a continuous, nonnegative function on  $\Gamma$  satisfying the equality

$$\varphi(\gamma_1 + \gamma_2) + \varphi(\gamma_1 - \gamma_2) = 2[\varphi(\gamma_1) + \varphi(\gamma_2)]$$

for all  $\gamma_1, \gamma_2$  in  $\Gamma$  (see § 6, IV in [11]).

Let  $X$  be a  $G$ -valued symmetric Gaussian random variable having the characteristic function  $\hat{\mu}(\gamma) = \exp(-\varphi(\gamma))$ , where  $\varphi$  is a certain fixed function on  $\Gamma$  with the properties as described above.

**THEOREM 2.** *The following statements are valid:*

(i)  $nP(\{\|X_n\| \geq \varepsilon\}) \rightarrow 0$  if  $n \rightarrow \infty$ , where  $X_n$  is a random variable with the characteristic function  $\hat{\nu}_n(\gamma) = \exp(-\varphi(\gamma)/n)$ .

(ii) *There exists a homogeneous Gaussian random element  $W$  with independent increments, having, with probability one, continuous paths and such that  $W(1)$  has the distribution of  $X$  and  $W(0) = 0$ .*

(iii) *Let  $\{X_j^{(n)}; j = 1, 2, \dots, n; n = 1, 2, \dots\}$  be an infinitesimal triangular array of symmetric, independent and identically distributed random variables such that  $X_1^{(n)} + \dots + X_n^{(n)}$  converges weakly to the distribution of  $X$ . Let  $\xi_n(t) = X_1^{(n)} + \dots + X_{[nt]}^{(n)}$ . Then  $\xi_n$  converges weakly to  $W$ .*

**Proof.** It follows immediately from [2] that condition (ii) of the theorem is satisfied.

In order to prove that conditions (i), (ii) and (iii) are equivalent we need only show that condition (3) of Lemma 4 holds for the array as is described in (iii). The remaining arguments are almost the same as in the proof of Theorem 1.

Let  $S_k^{(n)} = X_1^{(n)} + \dots + X_k^{(n)}$  and let  $\hat{\mu}_n$  be the characteristic function of  $X_j^{(n)}$ . We shall show that for each compact subset  $K \subseteq \Gamma$

$$\lim_{h \rightarrow 0} \lim_n \max_{h \leq n/h} \sup_{\gamma \in K} |\hat{\mu}_n(\gamma)^k - 1| = 0.$$

Let us observe that  $(\hat{\mu}_n)^{[nh]}$  converges to  $\exp(-h\varphi(\gamma))$  uniformly on each compact set  $K \subseteq \Gamma$  for a fixed  $h$ ,  $0 < h \leq 1$ . Let  $\nu_n$  be the distribution defined by the characteristic function  $\exp(-h\varphi)$ . Notice that if  $h \rightarrow 0$ , then  $\nu_n$  converges to the measure concentrated at the identity of  $G$ . So, given  $\varepsilon > 0$  and a compact subset  $K$  of  $\Gamma$ , there exists an  $h$ ,  $0 < h < 1$ , such that if  $0 < t \leq h$  then

$$\sup_{\gamma \in K} |\exp(-t\varphi(\gamma)) - 1| < \varepsilon/2.$$

Now, by the infinitesimality of array, it follows that there exists a positive integer  $n_0$  such that if  $n \geq n_0$  then  $\hat{\mu}_n(\gamma) > 0$  for  $\gamma \in K$  and

$$\sup_{\gamma \in K} |\hat{\mu}_n(\gamma)^{nh} - \exp(-h\varphi(\gamma))| < \varepsilon/2.$$

Hence

$$\sup_{\gamma \in K} |\hat{\mu}_n(\gamma)^{nh} - 1| < \varepsilon \quad \text{for } n \geq n_0$$

and, since  $\hat{\mu}_n(\gamma)$  is positive if  $\gamma \in K$  and  $n \geq n_0$ , we have

$$\sup_{\gamma \in K} |\hat{\mu}_n(\gamma)^{nt} - 1| = \sup_{\gamma \in K} |(\hat{\mu}_n(\gamma)^{nh})^{t/h} - 1| < \varepsilon$$

for  $t \leq h$  and  $n \geq n_0$ .

We now complete the proof by citing the following lemma, resulting from the remark after Definition 5.1, IV in [11].

**LEMMA 5.** *Let  $\{S_k^{(n)}; k = 1, 2, \dots, n; n = 1, 2, \dots\}$  be a triangular array of  $G$ -valued random variables. The following conditions are equivalent;*

- (i)  $\lim_{h \rightarrow 0} \overline{\lim}_{n \quad k \leq nh} P(\{\|S_k^{(n)}\| \geq \varepsilon\}) = 0$  for each  $\varepsilon > 0$ .  
 (ii) For each compact subset  $K$  of  $\Gamma$

$$\lim_{h \rightarrow 0} \overline{\lim}_{n \quad k \leq nh} \sup_{\gamma \in K} |\psi_k^{(n)}(\gamma) - 1| = 0,$$

where  $\psi_k^{(n)}$  is the characteristic function of  $S_k^{(n)}$ .

Remark. It is easy to see that we can formulate and prove, in a similar way, an analogue of Theorem 1 for  $G_G$ -valued random elements. Obviously, instead of random elements used in the formulation of Theorem 1, we have to deal with the stochastic process obtained by linear interpolation of the sums  $(X_1 + \dots + X_k)/n^{1/2}$ ,  $k = 0, 1, \dots, n$ .

**Added in proof.** I have noticed that the fact that the condition (i) of Theorem 2 holds in every LCA group follows also from the paper of V. V. Sazonov and V. N. Tutubalin *Probability distributions on topological groups*, Theor. Prob. Appl. (1966) (see Theorem 4.11).

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