

## LARGE DEVIATIONS FOR GAUSSIAN STOCHASTIC PROCESSES WITH SAMPLE PATHS IN ORLICZ SPACES

ANNA T. LAWNICZAK

**1. Introduction.** Let  $X$  be a complete, separable metric space, and  $\{\mu_\epsilon: \epsilon \searrow 0\}$  a family of probability measures on the Borel subsets of  $X$ . We say that  $\{\mu_\epsilon: \epsilon \searrow 0\}$  obeys the large deviation principle (LDP) with a rate function  $I(\cdot)$  if there exists a function  $I(\cdot)$  from  $X$  into  $[0, \infty]$  satisfying:

- (i)  $0 \leq I(x) \leq \infty$  for all  $x \in X$ ,
- (ii)  $I(\cdot)$  is lower semicontinuous,
- (iii) for each  $I < \infty$  the set  $\{x: I(x) \leq I\}$  is compact set in  $X$ ,
- (iv) for each closed set  $C \subset X$

$$\limsup_{\epsilon \rightarrow 0} \epsilon \log \mu_\epsilon(C) \leq -\inf_{x \in C} I(x),$$

- (v) for each open set  $U \subset X$

$$\liminf_{\epsilon \rightarrow 0} \epsilon \log \mu_\epsilon(U) \geq -\inf_{x \in U} I(x).$$

It is easy to see that if  $A$  is a Borel set such that

$$\inf_{x \in A^0} I(x) = \inf_{x \in A} I(x) = \inf_{x \in \bar{A}} I(x)$$

then

$$\lim_{\epsilon \rightarrow 0} \epsilon \log \mu_\epsilon(A) = -\inf_{x \in A} I(x)$$

where  $A^0$  and  $\bar{A}$  are respectively the interior and the closure of the Borel set  $A$ .

Let  $\mathbf{R}$  denote the real line and for  $N < \infty$  let  $(\mathbf{R}^N, \mathbf{B}(\mathbf{R}^N))$  be  $N$ -dimensional vector space with Borel  $\sigma$ -algebra  $\mathbf{B}(\mathbf{R}^N)$ . Let  $\mu$  be a mean-zero, Gaussian measure on  $(\mathbf{R}^N, \mathbf{B}(\mathbf{R}^N))$  such that the covariance  $S$  is a positive definite matrix. Let  $\mu_\epsilon = \mu \circ \epsilon^{-1/2}$  (i.e.,  $\mu_\epsilon(A) = \mu(\epsilon^{-1/2}A)$  for any measurable set  $A$ ) then by Cramer's Theorem [5]  $\{\mu_\epsilon: \epsilon \searrow 0\}$  satisfies the large deviation principle with the rate function

$$I(x) = 2^{-1} \langle x, S^{-1}x \rangle, \quad x \in \mathbf{R}^N.$$

---

Received April 30, 1987.

In the case that  $\mu$  is a mean-zero Gaussian measure defined on a separable, real Banach space  $\{\mu_\epsilon: \epsilon \searrow 0\}$  satisfies the LDP. This follows from a result of Donsker, Varadhan [6] in 1976. The above result for locally convex, Hausdorff real vector spaces was discussed by Bahadur, Zabell in 1979 [1].

In this paper we are going to prove the large deviation principle for

$$\{\mu_\epsilon: \mu_\epsilon = \mu \circ \epsilon^{-1/2}, \epsilon \searrow 0\}$$

where  $\mu$  is a mean-zero, non-degenerate Gaussian measure defined on a separable, real Orlicz space  $L_\phi$  which is not necessarily locally-convex vector space. The proof of the main result utilizes Cramer's Theorem for finite dimensional vector spaces and it bases on the series representation of Gaussian random elements with values in  $L_\phi$  spaces as well as on its exponential integrability.

The importance of studying the LDP for the sequence  $\{\mu_\epsilon: \epsilon \searrow 0\}$  defined on some Orlicz space  $L_\phi$  lies in the fact that the Orlicz space  $L_\phi$  can be considered as a space of sample paths of measurable stochastic processes. Necessary and sufficient conditions for a measurable Gaussian stochastic process to have almost all its sample paths in some Orlicz space  $L_\phi$  were established in [2], in terms of the covariance function of the process.

As an application of the LDP for Orlicz spaces we get an extension of Kallianpur's and Oodaira's (1978), Marlow's (1973) results concerning some asymptotic estimates of the probabilities of high level occupation times for Gaussian stochastic processes with sample paths in Orlicz spaces.

**2. Preliminaries.** Let  $(T, \mathbf{F}, m)$  be an arbitrary  $\sigma$ -finite measure space with  $\sigma$ -algebra  $\mathbf{F}$  and a separable measure  $m$ . Let  $S$  be the space of equivalence classes in measure  $m$  of all real valued  $\mathbf{F}$  measurable functions. By  $\phi$  let us denote a continuous, non-negative, non-decreasing function defined for  $u \geq 0$  such that:

- (i)  $\phi(u) = 0$  if and only if  $u = 0$ ,
- (ii) satisfies the so-called  $\Delta_2$  condition, i.e., there is a positive constant  $k$  such that for any  $u$ ,  $\phi(2u) \leq k\phi(u)$ .

For  $x \in S$  let us define

$$R_\phi(x) = \int_T \phi(|x(t)|) m(dt)$$

and let  $L_\phi$  be the set of all  $x \in S$  such that  $R_\phi(x) < \infty$ . The set  $L_\phi$  is a linear space under the usual addition and scalar multiplication. Moreover it becomes a complete, separable metric space under the (usually non-homogeneous) seminorm  $\|\cdot\|_\phi$ .

$$\|x\|_\phi = \inf\{c: c > 0, R_\phi(c^{-1}x) \leq c\}.$$

The space  $(L_\phi, \|\cdot\|_\phi)$  is called an Orlicz space. In the case that  $\phi$  is a convex function  $L_\phi$  is a Banach space [11]. The best known examples of the Orlicz spaces are  $L_p(T, \mathbf{F}, m)$  spaces for  $0 \leq p < \infty$  [11].

For convenience let us recall some necessary facts concerning probability measures on Orlicz spaces  $(L_\phi, \mathbf{B}(L_\phi))$ , where  $\mathbf{B}(L_\phi)$  denotes Borel  $\sigma$ -algebra of subsets of  $L_\phi$ .

A. For each probability measure  $\mu$  on  $(L_\phi, \mathbf{B}(L_\phi))$  one can construct a measurable stochastic process

$$\xi = \{\xi(t):t \in T\} \text{ on } (\Omega, \Sigma, P) = (L_\phi, \mathbf{B}(L_\phi), \mu),$$

with sample paths in  $L_\phi$  such that  $\xi(x) = x \mu$  a.e.; induced measure  $\mu_\xi$  is equal to  $\mu$ , and for every pair  $(s, u)$  or real numbers

$$\xi(t, sx \pm uy) = s\xi(t, x) \pm u\xi(t, y) \quad m \times \mu \times \mu \text{ a.e.}$$

Conversely, each jointly measurable stochastic process  $\xi(t, \omega)$  defined on  $T \times \Omega$  with almost all its sample paths in  $L_\phi$  induces an  $L_\phi(T, \mathbf{F}, m)$  valued random element [2].

B. An  $L_\phi$ -valued random element  $\xi$  (or p.m.  $\mu$  on  $(L_\phi, \mathbf{B}(L_\phi))$ ) is Gaussian if for any pair of independent copies of  $\xi$ ,  $X_1$  and  $X_2$  the random elements  $X_1 + X_2$  and  $X_1 - X_2$  are independent; this is equivalent to: the process  $\xi$  with sample paths in  $L_\phi$  is Gaussian if and only if there exists a measurable subset  $T_0, m(T_0) = 0$  such that for all finite sets  $\{t_1, \dots, t_k\} \subset T \setminus T_0$  the random vectors  $\langle \xi(t_1), \dots, \xi(t_k) \rangle$  are Gaussian [2].

C. Let  $\xi = \{\xi(t):t \in T\}$  be a measurable Gaussian stochastic process and let

$$\theta(t) = E\xi(t), K(s, t) = E(\xi(s) - \theta(s))(\xi(t) - \theta(t)).$$

Then for almost every  $\omega, \xi(\cdot, \omega) \in L_\phi$  if and only if  $\theta(t) \in L_\phi$  and  $K^{1/2}(t, t) \in L_\phi$ . If almost all sample paths of the process  $\xi$  belong to the space  $L_\phi$  then the measure  $\mu_\xi$  induced by  $\xi$  on  $(L_\phi, \mathbf{B}(L_\phi))$  is Gaussian [2].

D. Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure on  $(L_\phi, \mathbf{B}(L_\phi))$  and let  $\xi = \{\xi(t):t \in T\}$  be a measurable stochastic process, such as in A, inducing the measure  $\mu$ . By A there exists a measurable subset  $T_0, m(T_0) = 0$  such that for any  $t \in T \setminus T_0$

$$\xi(t, x \pm y) = \xi(t, x) \pm \xi(t, y) \quad \mu \times \mu \text{ a.e.}$$

Let

$$H_\mu = \overline{\text{lin}\{\xi(t):t \in T \setminus T_0\}}^{L_2(\mu)}.$$

From [8] it follows that the space  $H_\mu$  does not depend on the version of the stochastic process inducing the measure  $\mu$  and consists of all quasi-additive measurable functionals (q.m.f.)  $F$  defined on  $(L_\phi, \mathbf{B}(L_\phi), \mu)$  [8], i.e.,

$$H_\mu = \{F: F: L_\phi \rightarrow \mathbf{R}, \text{ measurable, } F(x \pm y) = F(x) \pm F(y) \ \mu \times \mu \text{ a.e.}\}.$$

For each  $F \in H_\mu$  let

$$\Lambda F(\cdot) = \left[ \int \xi(\cdot, x)F(x)\mu(dx) \right] = [\Lambda_\xi F(\cdot)]$$

where  $[\cdot]$  denotes the class of functions equivalent in  $m$  a.e. In [8] it was shown that  $\Lambda$  is a one-to-one map which embeds continuously the space  $H_\mu$  into  $L_\phi$ , and this embedding does not depend on the version of the stochastic process inducing the measure  $\mu$ .

Let  $\{E_j\}_{j=1}^\infty$  be a C.O.N.S. in  $H_\mu$  and  $\psi_j(t) = \langle \xi(t), E_j \rangle$ , where  $\langle \cdot, \cdot \rangle$  denotes the inner product in  $H_\mu$ , then by [3, 8]

$$\xi(t, x) = \sum_{j=1}^\infty \psi_j(t)E_j(x)$$

$\mu$  a.e. in the  $L_\phi$ -seminorm, implying that  $\text{supp } \mu \subseteq \Lambda \bar{H}_\mu^{L_\phi}$ .

For the remainder of this paper we assume that there exists for certain  $p, 0 < p \leq 1$ , a  $p$ -homogeneous  $F$ -norm  $\|\cdot\|$  equivalent to the original one  $\|\cdot\|_\phi$ . This class of spaces contains for example all  $L_p, 0 < p < \infty$ , spaces or by [11] spaces  $L_\phi$  for which  $\phi$  satisfies additionally the condition

$$\inf_{0 < t < \infty} \inf\{c > 0: 2\phi(ct) \geq \phi(t)\} > 0.$$

PROPOSITION 2.1. Let  $\{X_n\}_{n=1}^\infty$  be a sequence of mean-zero,  $L_\phi$ -valued Gaussian random elements such that  $X_n \rightarrow 0$  in  $P$  (probability) as  $n \rightarrow \infty$ , then

$$\forall n \exists \alpha_n \forall \alpha < \alpha_n \ E \exp(\alpha^2 \|X_n\|^{2/p}) < \infty,$$

and  $\alpha_n \rightarrow \infty$  as  $n \rightarrow \infty$ .

Proof. Let  $\{\eta_j\}_{j=1}^\infty$  be a sequence of i.i.d.  $N(0, 1)$  random variables and

$$G = \overline{\text{lin}\{x\eta_j: x \in L_\phi, j = 1, 2, \dots\}}^{inP}$$

then  $G$  is a linear, separable, complete metric space which consists of mean-zero,  $L_\phi$ -valued Gaussian random elements. From [3] it follows that for each  $X_n$  there exists an  $L_\phi$  valued random element  $Y_n$  such that  $Y_n \in G$  and  $\mathbf{L}(X_n) = \mathbf{L}(Y_n)$  (where  $\mathbf{L}(\cdot)$  denotes the distribution of a random element). By [4] for each  $Y \in G$  there exists  $\alpha > 0$  such that

$$E \exp(\alpha^2 \|Y\|^{2/p}) < \infty.$$

Let  $g: G \rightarrow \mathbf{R}$  be defined as follows

$$g(Y) = \exp\|Y\|^{2/p} - 1,$$

and for  $Y \in G$

$$\|Y\|_g = \inf\{c^{-1} > 0: E(\exp c^2\|Y\|^{2/p} - 1) \leq c^{-1}\},$$

then  $(G, \|\cdot\|_g)$  becomes an Orlicz space in which convergence in  $\|\cdot\|_g$  seminorm is equivalent to the convergence in probability.

Let  $\{Y_n\} \subset G$  and  $Y_n \rightarrow 0$  in  $\|\cdot\|_g$  as  $n \rightarrow \infty$  then  $Y_n \rightarrow 0$  in  $P$  as  $n \rightarrow \infty$ , which follows from the fact that  $\|Y_n\|_g^{-1} \rightarrow \infty$  and the inequality

$$\begin{aligned} (1) \quad P(\|Y_n\| > \alpha) &\leq \exp\{-\alpha^{2/p}(\|Y_n\|_g^{-1} - \epsilon)^2\} \cdot E \exp\{(\|Y_n\|_g^{-1} - \epsilon)^2\|Y_n\|^{2/p}\} \\ &\leq \exp\{-\alpha^{2/p}(\|Y_n\|_g^{-1} - \epsilon)^2\} \cdot \{1 + (\|Y_n\|_g^{-1} - \epsilon)^{-1}\}, \end{aligned}$$

where  $\alpha$  and  $\epsilon$  are arbitrary small positive constants.

The inverse implication follows from the closed graph theorem. Since the convergence in probability is equivalent to the convergence in  $\|\cdot\|_g$  seminorm this implies that  $\|X_n\|_g^{-1} \rightarrow \infty$  as  $n \rightarrow \infty$  and for any  $\alpha < \|X_n\|_g^{-1}$ ,

$$E \exp \alpha^2\|X_n\|^{2/p} < \infty$$

which finishes the proof.

For future purposes let us recall the following fact:

**PROPOSITION 2.2 [13].** *Let  $X$  be a complete, separable metric space and  $\{\mu_\epsilon\}$  a family of probability measures defined on  $X$  which satisfies the large deviation principle with a rate function  $I(\cdot)$ . Let  $u$  be a continuous map from  $X \rightarrow Y$ , where  $Y$  is another complete, separable metric space. If we define  $Q_\epsilon$  on  $Y$  by  $Q_\epsilon = \mu_\epsilon \circ u^{-1}$ , then  $\{Q_\epsilon\}$  satisfies the large deviation principle with the rate function  $J(\cdot)$  defined by*

$$J(y) = \begin{cases} \inf_{x:u(x)=y} I(x) & \text{if } y \in u(X) \\ \infty & \text{if } y \notin u(X). \end{cases}$$

**3. Main theorem.** The proof of the main theorem will be preceded by several lemmas.

**LEMMA 3.1.** *Let  $\{E_k\}_{k=1}^\infty$  be a C.O.N.S. in  $H_\mu$  then  $\{E_k(x)\}_{k=1}^\infty$  is a sequence of i.i.d.  $N(0, 1)$  random variables. For each  $N < \infty$ , let*

$$X_N = \sum_{k=1}^N E_k(x)\Lambda E_k$$

and

$$\mu_{N,\epsilon} = L(\epsilon^{1/2}X_N),$$

then  $\{\mu_{N,\epsilon} : \epsilon \searrow 0\}$  satisfies the large deviation principle with the rate function

$$I_N(x) = \begin{cases} 2^{-1} \|\Lambda^{-1}x\|_{H_\mu}^2 & \text{if } x \in \Lambda(P_N H_\mu) \\ \infty & \text{if } x \notin \Lambda(P_N H_\mu) \end{cases}$$

where  $P_N$  denotes the orthogonal projection onto  $\text{lin}\{E_k : k = 1, \dots, N\}$ .

*Proof.* Let  $N < \infty$  be an arbitrary but fixed and let  $\{e_k\}_{k=1}^N$  be a C.O.N.S. in  $\mathbf{R}_N$ , then

$$Y_N = \sum_{k=1}^N E_k(x)e_k$$

is  $N$ -dimensional, mean-zero Gaussian random vector. For each  $\epsilon > 0$ , let

$$v_{N,\epsilon} = \mathbf{L}(\epsilon^{1/2} Y_N)$$

then by Cramer's Theorem  $\{v_{N,\epsilon} : \epsilon \searrow 0\}$  satisfies the large deviation principle with the rate function

$$\mathbf{J}_N(x) = 2^{-1} \langle x, x \rangle \quad \text{for } x \in \mathbf{R}^N.$$

By  $u$  let us denote a map from  $\mathbf{R}^N$  into  $L_\phi$  defined as follows

$$u\left(\sum_{k=1}^N x_k e_k\right) = \sum_{k=1}^N x_k \Lambda E_k,$$

then  $u$  is a continuous, one-to-one map from  $\mathbf{R}^N$  onto  $\Lambda(P_N H_\mu)$ , such that  $u(Y_N) = X_N$ . Let

$$\mu_{N,\epsilon} = \mathbf{L}(\epsilon^{1/2} X_N),$$

since  $\mu_{N,\epsilon} = v_{N,\epsilon} \circ u^{-1}$  then by Proposition 2.2  $\{\mu_{N,\epsilon} : \epsilon \searrow 0\}$  satisfies the large deviation principle with the rate function

$$I_N(x) = \inf_{y: u(y)=x} \mathbf{J}_N(y).$$

It is easy to see that

$$I_N(x) = \begin{cases} 2^{-1} \|\Lambda^{-1}x\|_{H_\mu}^2 & \text{if } x \in \Lambda(P_N H_\mu) \\ \infty & \text{if } x \notin \Lambda(P_N H_\mu). \end{cases}$$

*Remark.* For the rest of this paper we will use the following notation

$$I_\mu(x) = \begin{cases} 2^{-1} \|\Lambda^{-1}x\|_{H_\mu}^2 & \text{if } x \in \Lambda H_\mu \\ \infty & \text{if } x \notin \Lambda H_\mu. \end{cases}$$

For  $A$  an arbitrary set by  $A^c$  we will denote a complementary set and

$$I_{A,N} = \inf_{x \in A} I_N(x), \quad I_A = \inf_{x \in A} I_\mu(x),$$

$$A_\delta = \{x \in L_\phi : \exists y \in A, \|x - y\| \leq \delta\}.$$

LEMMA 3.2. *Let  $C$  be a closed subset of  $L_\phi$  such that  $C \cap \overline{\Lambda H_\mu}^{L_\phi} \neq \emptyset$  then for  $\delta > 0$  there exists  $N_\delta$  such that for any  $N \geq M \geq N_\delta$*

$$I_{C_\delta} \leq I_{C_{\delta,N}} \leq I_{C_{\delta,M}} < \infty.$$

*Proof.* Since  $C \cap \overline{\Lambda H_\mu}^{L_\phi} \neq \emptyset$  then for any  $\delta > 0$

$$C_\delta \cap \Lambda H_\mu \neq \emptyset.$$

Let

$$\Lambda F \in C_{\delta/2} \cap \Lambda H_\mu,$$

then there exists  $N$  such that

$$\Lambda P_N F \in C_\delta \cap \Lambda H_\mu.$$

Since for any  $M < N$

$$\Lambda P_M H_\mu \subseteq \Lambda P_N H_\mu \subseteq \Lambda H_\mu,$$

then for any  $\delta > 0$  there exists  $N_\delta$  such that for  $N > M \geq N_\delta$

$$\emptyset \neq C_\delta \cap \Lambda P_M H_\mu \subseteq C_\delta \cap \Lambda P_N H_\mu \subseteq C_\delta \cap \Lambda H_\mu$$

implying that

$$I_{C_\delta} \leq I_{C_{\delta,N}} \leq I_{C_{\delta,M}} < \infty.$$

LEMMA 3.3. (i) *The set  $K_r = \{\Lambda F : I_\mu(\Lambda F) \leq r^2\}$ ,  $0 < r < \infty$  is compact in  $L_\phi$ .*

(ii)  *$I_\mu(x)$  is a lower-semicontinuous function on  $\Lambda H_\mu$  with respect to  $\|\cdot\|$ -norm convergence, i.e., if  $\|\Lambda F_n - \Lambda F\| \rightarrow 0$  as  $n \rightarrow \infty$ ,  $F_n, F \in H_\mu$  then*

$$I_\mu(\Lambda F) \leq \liminf_{n \rightarrow \infty} I_\mu(\Lambda F_n).$$

*Proof.* First we show that  $K_r$  is a compact subset of  $L_\phi$  for any  $0 < r < \infty$ . Let  $\{\Lambda F_n\} \subset K_r$  be an arbitrary sequence. By the Banach-Alaoglu Theorem  $\{F_n\}$  contains a subsequence  $\{F_n\}$  which is weakly convergent to  $F$  from  $\Lambda^{-1}K_r$ . Remark D implies that there exists a measurable subset  $T_0$ ,  $m(T_0) = 0$  such that for any  $t \in T \setminus T_0$ ,  $\xi(t) \in H_\mu$  and

$$\Lambda_\xi F_n(t) = \int \xi(t) F_n d\mu \rightarrow \int \xi(t) F d\mu = \Lambda_\xi F(t).$$

Since

$$|\Lambda_{\xi} F_{n'}(t)| \leq K^{1/2}(t, t) \|F_{n'}\|_{H_{\mu}} \leq \sqrt{2r} K^{1/2}(t, t)$$

for  $m$  a.e.  $t$ , and  $K^{1/2}(t, t) \in L_{\phi}$  then by the Lebesgue Dominated Convergence Theorem,  $\Lambda F_{n'} \rightarrow \Lambda F$  in  $L_{\phi}$ , proving that  $K_r$  is a compact subset of  $L_{\phi}$ .

Proof of part (ii). Let us denote by  $\{F_{n'}\}$  a subsequence such that

$$\liminf_n I_{\mu}(\Lambda F_n) = \lim_{n'} I_{\mu}(\Lambda F_{n'}).$$

Since  $\|\Lambda F_{n'} - \Lambda F\| \rightarrow 0$  as  $n' \rightarrow \infty$ , then there exists a subsequence  $\{n''\} \subset \{n'\}$  and a measurable subset  $T_0$ ,  $m(T_0) = 0$  such that for any  $t \in T \setminus T_0$ ,  $\xi(t)$  is a q.m.f. and

$$\langle \xi(t), F_{n''} \rangle = \Lambda_{\xi} F_{n''}(t) \rightarrow \Lambda_{\xi} F(t) = \langle \xi(t), F \rangle$$

where  $\langle \cdot, \cdot \rangle$  denotes the inner product in  $H_{\mu}$ . Let

$$G = \text{lin}\{\xi(t): t \in T \setminus T_0\}$$

then  $G$  is a dense subset of  $H_{\mu}$  [8] and for any  $g \in G$

$$\langle g, F_{n''} \rangle \rightarrow \langle g, F \rangle \quad \text{as } n'' \rightarrow \infty.$$

Since

$$\|F_{n''}\|_{H_{\mu}} = \sup\{\langle g, F_{n''} \rangle: g \in G, \|g\|_{H_{\mu}} = 1\},$$

then for any  $g \in G$ ,  $\|g\|_{H_{\mu}} = 1$

$$\lim_{n''} \|F_{n''}\|_{H_{\mu}} \geq \lim_{n''} \langle g, F_{n''} \rangle = \langle g, F \rangle.$$

This implies that

$$\lim_{n''} \|F_{n''}\|_{H_{\mu}} \geq \sup\{\langle g, F \rangle: g \in G, \|g\|_{H_{\mu}} = 1\} = \|F\|_{H_{\mu}}$$

proving part (ii) because

$$\liminf_n \|F\|_{H_{\mu}} = \lim_{n''} \|F_{n''}\|_{H_{\mu}} \geq \|F\|_{H_{\mu}}.$$

LEMMA 3.4. Let  $C$  be a closed subset of  $L_{\phi}$  then  $\mathbf{I}_{C_{\delta}} \nearrow \mathbf{I}_C$  as  $\delta \searrow 0$ .

*Proof.* Clearly  $\mathbf{I}_{C_{\delta}} \leq \mathbf{I}_C$  and  $\mathbf{I}_{C_{\delta}} \nearrow$  as  $\delta \searrow 0$ . Suppose that  $\mathbf{I}_{C_{\delta}} \leq \mathbf{I}' < \mathbf{I}_C$  for all  $\delta > 0$ . Let  $n_0$  be such that for all  $n > n_0$   $\mathbf{I}' + n^{-1} < \mathbf{I}$ . Let  $\Lambda F_n \in C_{n^{-1}}$  be so that

$$I_{\mu}(\Lambda F_n) \leq \mathbf{I}' + n^{-1} < \mathbf{I}.$$

Since  $\{\Lambda F: I_{\mu}(\Lambda F) \leq \mathbf{I}\}$  is a compact set in  $L_{\phi}$  (Lemma 3.3) then  $\{\Lambda F_n\}$  contains a convergent subsequence  $\{\Lambda F_{n'}\}$  such that  $\Lambda F_{n'} \rightarrow \Lambda F$  in  $L_{\phi}$  where  $\Lambda F \in C$ . By the lower semicontinuity of the function  $I_{\mu}(\cdot)$  (Lemma 3.3)



$$I_\mu(\Lambda F) \cong \liminf_{n'} I_\mu(\Lambda F_{n'}) \cong I'$$

and  $I_C \cong I'$  which is contradictory unless  $I_{C_\delta} \nearrow I_C$  as  $\delta \searrow 0$ .

LEMMA 3.5. *Let  $U$  be an open set such that  $U \cap \overline{\Lambda H_\mu}^{L_\phi} \neq \emptyset$  then there exists  $N_U$  such that for any  $N \cong N_U$*

$$I_{U,N} < \infty \quad \text{and} \quad I_{U,N} \searrow I_U \quad \text{as} \quad N \rightarrow \infty.$$

*Proof.* Since  $U$  is an open set,  $U \cap \overline{\Lambda H_\mu}^{L_\phi} \neq \emptyset$  then

$$U \cap \Lambda H_\mu \neq \emptyset.$$

Let  $\Lambda F \in U \cap \Lambda H_\mu$  where  $F$  is a q.m.f. then there exists  $N_F$  such that for any  $N > N_F$

$$\Lambda P_N F \in U \cap \Lambda H_\mu$$

implying that there exists  $N_U$  such that for any  $N > N_U$   $I_{U,N} < \infty$ . Since for any  $M < N$

$$\Lambda P_M H_\mu \subseteq \Lambda P_N H_\mu \subseteq \Lambda H_\mu$$

then

$$I_{U,M} \cong I_{U,N} \cong I_U.$$

Let  $\epsilon > 0$  be an arbitrary small but fixed and let  $F$  be a q.m.f. such that

$$\Lambda F \in U \cap \Lambda H_\mu, \quad I_U + \epsilon > I_\mu(\Lambda F).$$

There exists  $N_F$  such that for  $N > N_F$

$$\Lambda P_N F \in U \cap \Lambda P_N H_\mu \quad \text{and} \quad I_U + \epsilon > I_\mu(\Lambda P_N F) \cong I_{U,N}$$

implying that  $I_{U,N} \searrow I_U$  as  $N \rightarrow \infty$ .

LEMMA 3.6. *Let  $U$  be an open set such that  $U \cap \Lambda H_\mu \neq \emptyset$  then there exists  $\delta_0 > 0$  such that for any  $\delta < \delta_0$*

$$((U^c)_\delta)^c \cap \Lambda H_\mu \neq \emptyset \quad \text{and} \quad I_{((U^c)_\delta)^c} \searrow I_U \quad \text{as} \quad \delta \searrow 0.$$

*Proof.* Let us observe that  $((U^c)_\delta)^c$  is an increasing sequence of open sets as  $\delta \searrow 0$ . If there is no  $\delta > 0$  such that

$$((U^c)_\delta)^c \cap \Lambda H_\mu \neq \emptyset$$

then

$$\bigcup_\delta ((U^c)_\delta)^c \cap \Lambda H_\mu = \emptyset$$

implying that  $\Lambda H_\mu \subseteq U^c$  which is contradictory to the assumption. Since  $((U^c)_\delta)^c$  is an increasing sequence of sets then

$$I_{((U^c)_\delta)^c} \searrow I \cong I_U \text{ as } \delta \searrow 0.$$

Let  $\epsilon > 0$  be an arbitrary then there exists

$$\Lambda F \in U = \bigcup_{\delta} ((U^c)_\delta)^c$$

and  $\delta_0$  such that for each  $\delta < \delta_0$

$$I_U + \epsilon > I_\mu(\Lambda F) > I_{((U^c)_\delta)^c} \cong I$$

proving that  $I_U = I$ .

**THEOREM 3.7.** *Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_\phi, \mathbf{B}(L_\phi))$  such that there exists a  $p$ -homogeneous  $F$ -norm  $\|\cdot\|$ ,  $0 < p \leq 1$ , equivalent to the original one  $\|\cdot\|_\phi$ . Let  $\mu_\epsilon = \mu \circ \epsilon^{-1/2}$ , then the family  $\{\mu_\epsilon; \epsilon \searrow 0\}$  satisfies the large deviation principle with the rate function  $I_\mu, I_\mu: L_\phi \rightarrow [0, \infty]$ , defined as follows*

$$I_\mu(x) = \begin{cases} 2^{-1} \|\Lambda^{-1}x\|_{H_\mu}^2 & \text{if } x \in \Lambda H_\mu \\ \infty & \text{if } x \notin \Lambda H_\mu \end{cases}$$

*Proof.* Lemma 3.3 implies lower-semicontinuity of the function  $I_\mu$  and the compactness in  $L_\phi$  of the set  $\{x: I_\mu(x) \leq I\}$  for any  $I > 0$ .

Let  $X$  be an  $L_\phi$ -valued random element, generated by  $\xi = \{\xi(t): t \in T\}$  a measurable stochastic process such as in  $D$ , inducing the measure  $\mu$ , then  $L(X) = \mu$  and by  $D$

$$X_N = \sum_{j=1}^N \psi_j E_j \rightarrow \sum_{j=1}^\infty \psi_j E_j = X$$

$\mu$  a.e. in  $L_\phi$ . Let  $A \in \mathbf{B}(L_\phi)$  then for  $\epsilon > 0$

$$\mu_\epsilon(A) = \mu(\epsilon^{-1/2}A) = P(\epsilon^{1/2}X \in A).$$

Upper bound. We want to prove that for an arbitrary closed set  $C$  in  $L_\phi$

$$(2) \limsup_{\epsilon \searrow 0} \epsilon \log \mu_\epsilon(C) \leq -\inf_{x \in C} I_\mu(x).$$

If  $C \cap \overline{\Lambda H_\mu}^{L_\phi} = \emptyset$  then  $I_C = \infty$  and for any  $\epsilon > 0$ ,

$$\epsilon^{-1/2}C \cap \overline{\Lambda H_\mu}^{L_\phi} = \emptyset.$$

Since

$$\text{supp } \mu \subseteq \overline{\Lambda H_\mu}^{L_\phi}$$

then for every  $\epsilon > 0$ ,  $\mu_\epsilon(C) = 0$  proving that (2) is true.

Let us assume that

$$C \cap \overline{\Lambda H}_\mu^{L_\phi} = \emptyset$$

and let  $\delta > 0$  be an arbitrary small but fixed, then

$$P(\epsilon^{1/2}X \in C) \leq P(\epsilon^{1/2}X_N \in C_\delta) + P(\|\epsilon^{1/2}(X - X_N)\| \geq \delta).$$

Let  $\eta > 0$  be an arbitrary small but fixed. By Proposition 2.1 and Lemma 3.2 there exists  $N_0$  such that for any  $N > N_0$

$$(\alpha_N - \eta)^2 \delta^{2/p} > \mathbf{1}_{C_{\delta, N_\delta}} \geq \mathbf{1}_{C_{\delta, N}}$$

and

$$\begin{aligned} P(\|X - X_N\| \geq \delta \epsilon^{-p/2}) &\leq \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} E \exp\{(\alpha_N - \eta)^2 \|X - X_N\|^{2/p}\} \\ &= \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} M_N. \end{aligned}$$

Let  $N > N_0$  be an arbitrary but fixed and  $\mathbf{L}(X_N) = \mu_N$  then by Lemma 3.1  $\{\mu_{N, \epsilon}; \epsilon \searrow 0\}$  satisfies the large deviation principle and for sufficiently small  $\epsilon > 0$

$$P(\epsilon^{1/2}X_N \in C_\delta) \leq \exp(-\epsilon^{-1} \mathbf{1}_{C_{\delta, N}} + \epsilon^{-1} \eta).$$

Therefore for sufficiently small  $\epsilon > 0$

$$\begin{aligned} P(\epsilon^{1/2}X \in C) &\leq \exp(-\epsilon^{-1} \mathbf{1}_{C_{\delta, N}} + \epsilon^{-1} \eta) \\ &\quad + \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} M_N \\ &= \{1 + M_N \exp \epsilon^{-1} [-(\alpha_N - \eta)^2 \delta^{2/p} + \mathbf{1}_{C_{\delta, N}} - \eta]\} \\ &\quad \times \exp(-\epsilon^{-1} \mathbf{1}_{C_{\delta, N}} + \epsilon^{-1} \eta). \end{aligned}$$

Then

$$\limsup_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2}X \in C) \leq -\mathbf{1}_{C_{\delta, N}} + \eta$$

and by Lemma 3.2

$$\limsup_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2}X \in C) \leq -\mathbf{1}_{C_\delta} + \eta.$$

Since  $\delta > 0$  and  $\eta > 0$  were arbitrarily small then by Lemma 3.4

$$\limsup_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2}X \in C) \leq -\mathbf{1}_C$$

equivalently

$$\limsup_{\epsilon \searrow 0} \epsilon \log \mu_\epsilon(C) \leq -\inf_{x \in C} I_\mu(x).$$

Lower bound. We want to prove that for an arbitrary open set  $U$  in  $L_\phi$

$$(3) \liminf_{\epsilon \searrow 0} \epsilon \log \mu_\epsilon(U) \geq -\inf_{x \in U} I_\mu(x).$$

If  $U \cap \overline{\Lambda H_\mu}^{L_\phi} = \emptyset$  then  $\mathbf{1}_U = \infty$  and for any  $\epsilon > 0$ ,

$$\epsilon^{-1/2} U \cap \overline{\Lambda H_\mu}^{L_\phi} = \emptyset.$$

Since  $\text{supp } \mu \subseteq \overline{\Lambda H_\mu}^{L_\phi}$ , then for each  $\epsilon$ ,  $\mu_\epsilon(U) = 0$  proving that (3) is true.

Let us assume that

$$U \cap \overline{\Lambda H_\mu}^{L_\phi} \neq \emptyset$$

then  $U \cap \Lambda H_\mu \neq \emptyset$  and by Lemma 3.6 there exists  $\delta_0$  such that for any  $\delta < \delta_0$

$$((U^c)_\delta)^c \cap \Lambda H_\mu \neq \emptyset.$$

Let  $\delta < \delta_0$  be arbitrary but fixed then

$$\begin{aligned} P(\epsilon^{1/2} X \in U) &\geq P(\epsilon^{1/2} X_N \in ((U^c)_\delta)^c, \|\epsilon^{1/2}(X - X_N)\| < \delta) \\ &\geq P(\epsilon^{1/2} X_N \in ((U^c)_\delta)^c) \\ &\quad + P(\|\epsilon^{1/2}(X - X_N)\| < \delta) - 1 \\ &= P(\epsilon^{1/2} X_N \in ((U^c)_\delta)^c) \\ &\quad - P(\|\epsilon^{1/2}(X - X_N)\| \geq \delta). \end{aligned}$$

Let  $\eta > 0$  be an arbitrary small but fixed. By Proposition 2.1 and Lemma 3.5 there exists  $N_0$  such that for any  $N \geq N_0$

$$(\alpha_N - \eta)^2 \delta^{2/p} - \eta > \mathbf{1}_{((U^c)_\delta)^c, N},$$

$$\mathbf{1}_{((U^c)_\delta)^c} + \eta \geq \mathbf{1}_{((U^c)_\delta)^c, N},$$

and

$$\begin{aligned} P(\|X - X_N\| \geq \delta \epsilon^{-p/2}) &\leq \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} \\ &\quad \times E \exp\{(\alpha_N - \eta)^2 \|X - X_N\|^{2/p}\} \\ &= \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} M_N. \end{aligned}$$

Let  $N > N_0$  be arbitrary but fixed and  $\mathbf{L}(X_N) = \mu_N$  then by Lemma 3.1  $\{\mu_{N,\epsilon} : \epsilon \searrow 0\}$  satisfies the large deviation principle and for sufficiently small  $\epsilon$

$$P(\epsilon^{1/2} X_N \in ((U^c)_\delta)^c) \geq \exp\{-\epsilon^{-1} \mathbf{1}_{((U^c)_\delta)^c, N} - \epsilon^{-1} \eta\}.$$

Then for sufficiently small  $\epsilon$

$$\begin{aligned} P(\epsilon^{1/2} X \in U) &\geq \exp\{-\epsilon^{-1} \mathbf{1}_{((U^c)_\delta)^c, N} - \epsilon^{-1} \eta\} - \exp\{-(\alpha_N - \eta)^2 \delta^{2/p} \epsilon^{-1}\} M_N \end{aligned}$$

$$= \{1 - M_N \exp \epsilon^{-1} [ -(\alpha_N - \eta)^2 \delta^{2/p} + \mathbf{1}_{((U^\epsilon)_\delta)^c, N} + \eta ] \} \\ \times \exp \{ -\epsilon^{-1} \mathbf{1}_{((U^\epsilon)_\delta)^c, N} - \epsilon^{-1} \eta \}.$$

Hence

$$\liminf_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2} X \in U) \geq -\mathbf{1}_{((U^\epsilon)_\delta)^c, N} - \eta \geq -\mathbf{1}_{(U^\epsilon)_\delta^c} - 2\eta.$$

Since  $\delta > 0$  was an arbitrary small then by Lemma 3.6

$$\liminf_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2} X \in U) \geq -\mathbf{1}_U - 3\eta$$

and by going with  $\eta \searrow 0$  we get

$$\liminf_{\epsilon \searrow 0} \epsilon \log P(\epsilon^{1/2} X \in U) \geq -\mathbf{1}_U$$

equivalently

$$\liminf_{\epsilon \searrow 0} \epsilon \log \mu_\epsilon(U) \geq -\inf_{x \in U} I_\mu(x).$$

**COROLLARY 3.8** (Extension of Cramer’s Theorem to Orlicz spaces  $L_\phi$ ). *Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_\phi, \mathbf{B}(L_\phi))$  such that there exists a  $p$ -homogeneous  $F$ -norm  $\|\cdot\|$ ,  $0 < p \leq 1$ , equivalent to the original one  $\|\cdot\|_\phi$ . Let  $\{X_i; i \geq 1\}$  be a sequence of independent,  $L_\phi$ -valued random elements, each with distribution  $\mu$ . Set*

$$S_n = \sum_{i=1}^n X_i$$

and let  $\mu_{n^{-1}}$  be the distribution of  $S_n/n$ , then  $\{\mu_{n^{-1}}; n \nearrow \infty\}$  satisfies the large deviation principle with the rate function  $I_\mu$ .

*Proof.* Since  $\mathbf{L}(S_n/n) = \mathbf{L}(X_1/\sqrt{n})$  then the proof is an immediate consequence of Theorem 3.7 if we take  $\epsilon = n^{-1}$ .

**COROLLARY 3.9.** *Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_\phi, \mathbf{B}(L_\phi))$  such that there exists a  $p$ -homogeneous  $F$ -norm  $\|\cdot\|$ ,  $0 < p \leq 1$ , equivalent to the original one  $\|\cdot\|_\phi$ . Let*

$$a = \inf \{ I_\mu(x) : \|x\| \geq 1 \}$$

then  $0 < a < \infty$  and

$$\lim_{R \rightarrow \infty} R^{-2} \log \mu(\{x : \|R^{-1}x\| > 1\}) = -a.$$

*Proof.* The proof follows from Theorem 3.7 and Proposition 8 in [9].

As a next application of Theorem 3.7 we get an extension of Kallianpur’s and Oodaira’s (1978), Marlow’s (1973) results concerning

some asymptotic estimates of the probabilities of high level occupation times for Gaussian stochastic processes with sample paths in Orlicz spaces. Let

$$D_\beta = \{f(t): f(t) \in L_\phi, m(\{t: f(t) > 1\}) > \beta\}.$$

Then for any  $\beta > 0$ ,  $D_\beta$  is an open set in  $L_\phi$  [9].

**COROLLARY 3.10.** *Let  $\xi = \{\xi(t): t \in T\}$  be a mean-zero, Gaussian stochastic process with almost all sample paths in an Orlicz space  $L_\phi$  such that there exists a  $p$ -homogeneous  $F$ -norm  $\|\cdot\|$ ,  $0 < p \leq 1$ , equivalent to the original one  $\|\cdot\|_\phi$ . Let for any  $\beta > 0$*

$$a_\beta = \inf\{I_\mu(x): x \in D_\beta\}, \quad \bar{a}_\beta = \inf\{I_\mu(x): x \in \bar{D}_\beta\},$$

then

$$\begin{aligned} -a_\beta &\leq \liminf_{\alpha \rightarrow \infty} \alpha^{-2} \log P(\{\omega: m(\{t: \xi(t, \omega) > \alpha\}) > \beta\}) \\ &\leq \limsup_{\alpha \rightarrow \infty} \alpha^{-2} \log P(\{\omega: m(\{t: \xi(t, \omega) > \alpha\}) > \beta\}) \leq -\bar{a}_\beta. \end{aligned}$$

If  $T$  is a metric space with the measure  $m$  such that for any open set  $U$ ,  $m(U) > 0$ , the covariance function  $K(s, t)$  of the process  $\xi = \{\xi(t): t \in T\}$  is continuous and for each  $\beta > 0$

$$m(\{s: m(\{t: K(s, t) > 0\}) > \beta\}) > 0$$

then  $0 < a_\beta < \infty$  and

$$\lim_{\alpha \rightarrow \infty} \alpha^{-2} \log P(\{\omega: m(\{t: \xi(t, \omega) > \alpha\}) > \beta\}) = -a_\beta.$$

*Proof.* This follows from Theorem 3.7 and Theorem 9 in [9].

*Acknowledgement.* The author is thankful to Professor J. Rosinski for pointing out to her Proposition 2.1.

#### REFERENCES

1. R. R. Bahadur and S. L. Zabell, *Large deviations of the sample mean in general vector spaces*, Ann. Probab. 7 (1979), 587-621.
2. T. Byczkowski, *Gaussian measures on  $L_p$  spaces*,  $0 < p < \infty$ , Studia Math. 59 (1977), 249-261.
3. ———, *Norm convergent expansion for  $L_\phi$ -valued Gaussian random elements*, Studia Math. 64 (1979), 87-95.
4. T. Byczkowski and T. Zak, *On the integrability of Gaussian random vectors*, Lecture Notes in Math. 828 (Springer, New York, 1979), 21-30.
5. H. Cramér, *Sur un nouveau théorème-limite de la théorie des probabilités*, Actualités Sci. Indust. 736 (1938), 5-23.
6. M. D. Donsker and S. R. S. Varadhan, *Asymptotic evaluation for certain Markov processes expectations for large-time III*, Comm. Pure Appl. Math. 29 (1976), 389-461.
7. G. Kallianapur and H. Oodaira, *Freidlin-Wentzell type estimates for abstract Wiener spaces*, Sankhya 40, Series A (1978), 116-137.

8. A. T. Lawniczak, *Gaussian measures on Orlicz spaces and abstract Wiener spaces*, Lecture Notes in Math. 939 (Springer, New York, 1982), 81-97.
9. ——— *High level occupation times for Gaussian stochastic processes with sample paths in Orlicz spaces*, Can. J. Math. 39 (1987), 239-256.
10. N. Marlow, *High level occupation times for continuous Gaussian processes*, Ann. Probab. 3 (1973), 388-397.
11. S. Rolewicz, *Metric linear spaces* (Polish Scientific Publishers, PWN, 1972).
12. D. W. Stroock, *An introduction to the theory of large deviations* (Springer-Verlag, New York, 1984).
13. S. R. S. Varadhan, *Large deviations and applications*, SIAM (1984).

*University of Toronto,  
Toronto, Ontario*