## HIGH LEVEL OCCUPATION TIMES FOR GAUSSIAN STOCHASTIC PROCESSES WITH SAMPLE PATHS IN ORLICZ SPACES

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Let X be a complete separable metric space, and  $\{P_{\epsilon}\}$  a family of probability measures on the Borel subsets of X. We say that  $\{P_{\epsilon}\}$  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  $l < \infty$  the set  $\{x: I(x) \leq l\}$  is a compact set in X.
- (iv) For each closed set  $C \subset X$

 $\limsup_{\epsilon \to 0} \operatorname{sup} P_{\epsilon}(C) \leq -\inf_{x \in C} I(x).$ 

(v) For each open set  $G \subset X$ 

 $\liminf_{\epsilon \to 0} \operatorname{elog} P_{\epsilon}(G) \geq -\inf_{x \in G} I(x).$ 

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

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

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

1. PROPOSITION [12]. Let  $P_{\epsilon}$  satisfy the large deviation principle with a rate function  $I(\cdot)$ . Let F be a continuous map from  $X \to L$  where L is another complete separable metric space. Then if we define  $Q_{\epsilon}$  on L by  $Q_{\epsilon} = P_{\epsilon} \circ F^{-1}$ , then  $Q_{\epsilon}$  satisfies the large deviation principle with a rate function  $\mathcal{J}(\cdot)$  defined by

$$\mathcal{J}(y) = \inf_{x:F(x)=y} I(x).$$

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2. PROPOSITION [3, 11]. Let  $(B, \mathcal{B}(B), \mu)$  be a real separable Banach space with a mean-zero Gaussian measure  $\mu$  defined on the Borel  $\sigma$ -algebra  $\mathcal{B}(B)$ . Let  $H_{\mu}$  be the closure in  $L^{2}(\mu)$  of the set  $\{x^{*}(\cdot):x^{*} \in B^{*}\}$ ; and, for  $h \in H_{\mu}$  let us define

$$S_{\mu}(h) = \int xh(x)\mu(dx).$$

Let  $\{X_i\}_{i=1}^{\infty}$  be a sequence of independent B-valued random elements, each with distribution  $\mu$ . Set

$$S_n = \sum_{1}^{n} X_m$$

and let  $\mu_n$  be the distribution of  $S_n/n$ , then  $\{\mu_n: n \ge 1\}$  satisfies the large deviation principle with the rate function  $I_{\mu}(\cdot)$  defined as follows

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

and for any closed set F:

$$\limsup_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-\frac{1}{2}}F) \leq -\inf_{x \in F} I_{\mu}(x),$$

for any open set G:

$$\liminf_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-\frac{1}{2}}G) \geq -\inf_{x \in G} I_{\mu}(x).$$

In this paper we are going to show that Proposition 2 is true for Orlicz spaces  $L_{\phi}$  such that  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave. It is easy to see that this class of Orlicz spaces includes some non-locally convex vector spaces. By applying the L.D.P. for Orlicz spaces we extend Kallianpur's and Oodaira's (1978), Marlow's (1973) results concerning some asymptotic estimates for the probabilities of high level occupation times for continuous Gaussian stochastic processes to the class of Gaussian stochastic processes with sample paths in Orlicz spaces.

Let  $(T, \mathcal{F}, m)$  be an arbitrary  $\sigma$ -finite measure space with  $\sigma$ -algebra  $\mathcal{F}$ and a separable measure m. Let S be the space of equivalence classes in measure m of all real valued  $\mathcal{F}$  measurable functions. By  $\phi$  let us denote a continuous, non-negative, non-decreasing function defined for  $u \ge 0$  such that  $\phi(u) = 0$  if and only if u = 0. We assume additionally that the function  $\phi(u)$  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  $R_{\phi}(ax) < \infty$  for some positive scalar *a*. 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  $\|\cdot\|_{\phi}$ :

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

The space  $(L_{\phi}, \|\cdot\|_{\phi})$  is called an Orlicz space. It is easy to see that convergence in the  $L_{\phi}$  seminorm implies convergence in measure. In the case that  $\phi$  is a convex function  $L_{\phi}$  is a Banach space [10]. We say that  $\phi(\sqrt{u})$  is equivalent to a concave function  $\tilde{\phi}(u)$  if for all  $u \ge 0$ 

$$A\phi(c_1\sqrt{u}) \leq \widetilde{\phi}(u) \leq B\phi(c_2\sqrt{u})$$

for some  $c_1$ ,  $c_2$ , A, B positive constants. In this case Theorem 7.2.5 [5] implies that  $\phi(u)$  satisfies  $\Delta_2$ -condition. The best known examples of the Orlicz spaces are  $L_p(T, \mathcal{F}, m)$  spaces for  $0 \leq p < \infty$  [10].

For convenience let us recall some necessary facts concerning probability measures on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  spaces.

A. For each probability measure  $\mu$  on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  can be constructed a measurable stochastic process  $\xi = \{\xi(t): t \in T\}$  on

$$(\Omega, \Sigma, P) = L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$$

with sample paths in  $L_{\phi}$  such that  $\tilde{\xi}(x) = x \mu$  a.e.; induced measure  $\mu_{\xi}$  is equal to  $\mu$ , and for every pair (s, u) of 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, with almost all its sample paths in  $L_{\phi}$  induces an  $L_{\phi}(T, \mathcal{F}, m)$  valued random element [1].

B. An  $L_{\phi}$ -valued r.e.  $\xi$  (or p.m.  $\mu$  on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  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  $\{t_1, \ldots, t_k\} \subset T \setminus T_0$  the random vector  $\langle \xi(t_1), \ldots, \xi(t_k) \rangle$  is Gaussian [1].

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

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

Then for almost every  $\omega$ ,  $\xi(\cdot, \omega) \in L_{\phi}$  if and only if  $\theta(t) \in L_{\phi}$  and

 $K^{\frac{1}{2}}(t, t) \in L_{\phi}$ . If almost all sample paths of the process  $\xi$  belong to the space  $L_{\phi}$  then the measure  $\mu_{\xi}$  induced by  $\xi$  on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  is Gaussian [1].

D. Let  $\mu$  be a mean-zero non-degenerate Gaussian measure on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  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

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

From [7] it follows that the space  $H_{\mu}$  does not depend on the version of the stochastic process inducing the measure  $\mu$  and consists of quasi-additive measurable functionals (q.m.f.) F [7], i.e.,

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

For each  $F \in H_{\mu}$  let

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

where  $[\cdot]$  denotes the class of functions equivalent *m* a.e. In [7] 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  $\mu$ . When  $L_{\phi}$  is a Banach space from [7] it follows that  $S_{\mu}$  defined in Proposition 2 equals  $\Lambda$  and the rate function  $I_{\mu}(\cdot)$  is expressed as follows:

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

Let  $\{E_i\}$  be a C.O.N.S. in  $H_{\mu}$  and  $\psi_i(t) = \langle \xi(t), E_i \rangle$ , then by [1, 2]

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

 $\mu$  a.e. in the seminorm of  $L_{\phi}$ .

3. PROPOSITION [8]. Any mean-zero, non-degenerate Gaussian measure  $\mu$  defined on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  such that  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave is the image under a continuous linear map of a centered Gaussian measure on a separable real Hilbert space.

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$$R_{\phi}(bK^{\gamma_2}(t,\,t)\,) < \infty$$

then

$$\nu_{\phi}(A) = \int_{A} \widetilde{\phi}(b^{2}c_{2}^{-2}K(t, t))m(dt), \quad A \in \mathscr{P}$$

defines a non-negative, finite measure on F. Let

$$L_{2,\phi} = L_2(T, \mathscr{F}, v_{\phi})$$

be a real, separable Hilbert space and u be a map defined on  $L_{2,\phi}$  as follows:

$$L_{2,\phi} \ni f(t) \mapsto (uf)(t) = f(t)K^{\frac{1}{2}}(t, t)$$

then u is a linear, continuous map with values in  $L_{\phi}$ . Let

$$f_j(t) = \begin{cases} \psi_j(t) K^{-\frac{1}{2}}(t, t) & \text{if } K(t, t) \neq 0\\ 0 & \text{otherwise} \end{cases}$$

then

$$S = \sum_{j=1}^{\infty} f_j(t) E_j$$

is a mean-zero Gaussian random element with values in  $L_{2,\phi}$  such that  $uS = \tilde{\xi}$  a.e.

4. THEOREM. Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$  such that:

(i)  $\phi(t)$  is a convex function,

or

(ii)  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave function. Let  $\{X_i\}_{i=1}^{\infty}$  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  $\mu_n$  be the distribution of  $S_n/n$ , then  $\{\mu_n:n \ge 1\}$  satisfies the large deviation principle with the rate function  $I_{\mu}(x)$ , defined as follows:

$$I_{\mu}(x) = \begin{cases} \frac{1}{2} \int F^{2}(x)\mu(dx) & \text{if } x = \Lambda F \\ \infty & \text{if } x \notin \Lambda H_{\mu}. \end{cases}$$

*Proof.* In the case that  $\phi(t)$  is a convex function  $L_{\phi}$  is a Banach space and the theorem follows from Remark D. We have to prove only the case where  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave function.

In the proof we use the same notation as in the sketch of the proof of Proposition 3. Let

$$L_{\phi,0} = \{ f(t): f(t) \in L_{\phi}, f(t) = 0 \text{ on the set } \{ t: K(t, t) = 0 \} \}.$$

It is easy to see that  $L_{\phi,0}$  is a closed linear subspace of  $L_{\phi}$ , such that  $u(L_{2,\phi}) \subseteq L_{\phi,0}$ . Since the measures  $v_{\phi}$  and *m* are absolutely continuous with respect to each other on the set  $\{t: K(t, t) \neq 0\}$ , then *u* is a one-to-one map with  $u^{-1}$  defined as follows:

$$(u^{-1}f)(t) = \begin{cases} K^{-\frac{1}{2}}(t, t)f(t) & \text{if } K(t, t) \neq 0\\ 0 & \text{if } K(t, t) = 0 \end{cases}$$

for any  $f \in L_{\phi,0}$ .

Proposition  $\frac{1}{3}$  implies that the measure  $\mu$  is concentrated on the subspace  $L_{\phi,0}$  and  $\mu(A) = \mu_S(u^{-1}A)$  for any measurable subset A, where  $\mu_S$  denotes the distribution of the random element S.

Since *u* is a continuous linear map [8],  $\mu = \mu_S \circ u^{-1}$ ,  $\mu_S$  is a mean-zero Gaussian measure defined on the Hilbert space  $L_{2,\phi}$ , then by Propositions 1 and 2 { $\mu_n: n \ge 1$ } satisfies the L.D.P. with the rate function

$$I_{\mu}(x) = \inf_{y:u(y)=x} I_{S}(y)$$

where  $I_S$  is a rate function for the measures  $\mu_{S,n}$ . We will prove that

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

where  $\|\cdot\|$  denotes the norm in the Hilbert space  $H_{\mu}$ .

Let us denote by  $\xi = \{\xi(t): t \in T\}$ ,  $\eta = \{\eta(t): t \in T\}$  a measurable stochastic processes such as in A, inducing the measures  $\mu$  and  $\mu_S$  respectively. Since

$$\eta(t, x) \in x$$
 for  $\mu_S$  a.e. x;

 $\xi(t, x) \in x$  for  $\mu$  a.e. x

then

$$u\eta(\cdot, x) = \xi(\cdot, ux) \qquad m \text{ a.e. for } \mu_S \text{ a.e. } x$$
$$u^{-1}\xi(\cdot, x) = \eta(\cdot, u^{-1}x) \nu_{\phi} \text{ a.e. for } \mu \text{ a.e. } x.$$

Let  $H_{\mu}$  be the space of quasi-additive measurable functionals defined on  $(L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$  and  $H_S$  the space of quasi-additive measurable functionals defined on  $(L_{2,\phi}, \mathscr{B}(L_{2,\phi}), \mu_S)$ . Since

$$|\Lambda_{\xi}F(t)| \leq \left(\int \xi^2(t)d\mu\right)^{\frac{1}{2}} \left(\int F^2d\mu\right)^{\frac{1}{2}} = K^{\frac{1}{2}}(t,t) ||F||$$

then

 $\Lambda H_{\mu} \subseteq L_{\phi,0}.$ 

If F is a q.m.f. on the space  $(L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$  then  $F \circ u$  is a q.m.f. on the space  $(L_{2,\phi}, \mathscr{B}(L_{2,\phi}), \mu_S)$  and if G is a q.m.f. on the space  $(L_{2,\phi}, \mathscr{B}(L_{2,\phi}), \mu_S)$  then  $G \circ u^{-1}$  is a q.m.f. on the space  $(L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$ . These follow from

$$0 = \mu \times \mu(\{(x, y): F(x \pm y) \neq F(x) \pm F(y)\})$$
  
=  $\mu_S \times \mu_S(\{(u^{-1}x, u^{-1}y): F(x \pm y) \neq F(x) \pm F(y)\})$   
=  $\mu_S \times \mu_S(\{(z, s): F(u(z \pm s)) \neq F(uz) \pm F(us)\}).$   
 $\mu \times \mu(\{(x, y): G(u^{-1}(x \pm y)) \neq G(u^{-1}x) \pm G(u^{-1}y)\})$   
=  $\mu_S \times \mu_S(\{(u^{-1}x, u^{-1}y): G(u^{-1}(x \pm y)) \neq G(u^{-1}x) \pm G(u^{-1}x)) \pm G(u^{-1}y)\})$ 

 $= \mu_{S} \times \mu_{S}(\{(z, s): G(z \pm s) \neq G(z) \pm G(s)\}) = 0.$ 

Let G be  $\mu_{S}$  q.m.f., then

$$u(\Lambda_{\eta}G)(t) = u\left(\int \eta(t, x)G(x)\mu_{S}(dx)\right)$$
  
=  $\int K^{\frac{1}{2}}(t, t)\eta(t, x)G(x)\mu_{S}(dx)$   
=  $\int \xi(t, ux)G(x)\mu_{S}(dx) = \int \xi(t, y)G(u^{-1}y)\mu(dy) m \text{ a.e.}$ 

Since  $G \circ u^{-1}$  is  $\mu$  q.m.f., then  $u(\Lambda_{\eta}G) \in \Lambda H_{\mu}$ , which implies  $u(\Lambda_{\eta}H_S) \subseteq \Lambda H_{\mu}$ . We use the same notation for a function and the corresponding equivalence class in measure.

Let F be  $\mu$  q.m.f., then

$$u^{-1}(\Lambda_{\xi}F)(t) = u^{-1}\left(\int \xi(t, x)F(x)\mu(dx)\right)$$
$$= \int K^{-\frac{1}{2}}(t, t)\xi(t, x)F(x)\mu(dx)$$
$$= \int \eta(t, u^{-1}x)F(x)\mu(dx)$$
$$= \int \eta(t, y)F(uy)\mu_{S}(dy) \quad \nu_{\phi} \text{ a.e.}$$

Since  $F \circ u$  is  $\mu_S$  q.m.f., then

$$u^{-1}(\Lambda_{\xi}F) \in \Lambda_{\eta}H_S$$
, and  $u^{-1}(\Lambda_{\xi}H_{\mu}) \subseteq \Lambda_{\eta}H_S$ ,

this implies that

 $u(\Lambda_n H_{\mathcal{S}}) = \Lambda H_{\mu} \subseteq L_{\phi 0}.$ 

From Propositions 1, 2 and Remark D it follows that

$$I_{\mu}(x) = \inf_{y:u(y)=x} I_{S}(y)$$

where  $I_{S}(\cdot)$  is the rate function for the sequence  $\{\mu_{S,n}:n \geq 1\}$ .

If  $x \in \Lambda H_{\mu}$  then there exists a q.m.f. F such that  $x = \Lambda F$ . Since u is a one-to-one map,

$$I_{\mu}(x) = I_{S}(u^{-1}(\Lambda F))$$
  
=  $\frac{1}{2} \int [F(u(z))]^{2} \mu_{S}(dz) = \frac{1}{2} \int F^{2}(y) \mu(dy).$ 

If  $x \notin \Lambda H_{\mu}$  then there is no  $y \in \Lambda_{\eta} H_S$  such that u(y) = x and this implies that  $I_{\mu}(x) = \infty$ . This finishes the proof of the theorem that

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

5. COROLLARY. Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_{\phi}, \mathscr{B}(L_{\phi}))$ , such that  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave, then for any closed subset E

$$\limsup_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-1/2}E) \leq -\inf_{x \in E} I_{\mu}(x)$$

and for any open subset D

$$\liminf_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-\frac{1}{2}}D) \ge -\inf_{x \in D} I_{\mu}(x).$$

*Proof.* The proof of this corollary is an immediate consequence of Theorem 4 and Theorem 3.48 in [11], namely for any closed subset E and an open subset D

$$\limsup_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-\frac{1}{2}}E) = \limsup_{\epsilon \to 0} \epsilon \log \mu_{S}(\epsilon^{-\frac{1}{2}}u^{-1}E)$$
$$\leq -\inf_{x \in u^{-1}E} I_{S}(x) = -\inf_{y \in E} I_{\mu}(y),$$

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 $\liminf_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-\frac{1}{2}}D) = \liminf_{\epsilon \to 0} \epsilon \log \mu_{S}(\epsilon^{-\frac{1}{2}}u^{-1}D)$ 

$$\geq -\inf_{x \in u^{-1}D} I_S(x) = -\inf_{y \in D} I_{\mu}(y),$$

because for any subset B

$$\inf_{x \in u^{-1}B} I_S(x) = \begin{cases} \inf_{x \in u^{-1}(B \cap \Lambda H_{\mu})} I_S(x) & \text{if } B \cap \Lambda H_{\mu} \neq \phi \\ \infty & \text{if } B \cap \Lambda H_{\mu} = \phi \end{cases}$$
$$= \begin{cases} \inf_{y \in B \cap \Lambda H_{\mu}} I_S(u^{-1}y) & \text{if } B \cap \Lambda H_{\mu} \neq \phi \\ \infty & \text{if } B \cap \Lambda H_{\mu} = \phi \end{cases}$$
$$= \begin{cases} \inf_{y \in B \cap \Lambda H_{\mu}} I_{\mu}(y) & \text{if } B \cap \Lambda H_{\mu} \neq \phi \\ \infty & \text{if } B \cap \Lambda H_{\mu} = \phi \end{cases}$$
$$= \inf_{y \in B} I_{\mu}(y).$$

6. PROPOSITION. Let  $(L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$  be an Orlicz space with mean-zero, non-degenerate Gaussian measure  $\mu$  and a rate function  $I_{\mu}(\cdot)$ :

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

then

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

(ii)  $I_{\mu}(y)$  is lower-semicontinuous on  $\Lambda H_{\mu}$  with respect to  $\|\cdot\|_{\phi}$ -norm convergence, i.e., if  $\|\Lambda F_n - \Lambda F\|_{\phi} \to 0$  as  $n \to \infty$ ,  $F_n$ ,  $F \in H_{\mu}$ , then

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

*Proof.* First we show that  $K_r$ , for any  $0 < r < \infty$  is a compact subset of  $L_{\phi}$ . 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

$$\Lambda_{\xi}F_{n'}(t) = \int \xi(t)F_{n'}d\mu \mapsto \int \xi(t)Fd\mu = \Lambda_{\xi}F(t).$$

Since

$$|\Lambda_{\xi}F_{n'}(t)| \leq K^{\frac{1}{2}}(t, t) ||F_{n'}|| \leq \sqrt{2}rK^{\frac{1}{2}}(t, t)$$

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

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

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

Since

 $\|\Lambda F_{n'} - \Lambda F\|_{\phi} \mapsto 0 \text{ as } n' \to \infty,$ 

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

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

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

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

then G is a dense subset of  $H_{\mu}$  [7] and for any  $g \in G$ 

 $\langle g, F_{n''} \rangle \mapsto \langle g, F \rangle$  as  $n'' \to \infty$ .

Since

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

then for any  $g \in G$ , ||g|| = 1

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

This implies that

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

which proves part (ii), because

 $\liminf_{n} ||F_{n}|| = \lim_{n''} ||F_{n''}|| \ge ||F||.$ 

7. Remark. In the case that  $\phi(t)$  satisfies additionally

 $\liminf_{t \to \infty} \inf \{c > 0 : 2\phi(ct) > \phi(t) \} > 0$ 

the space  $L_{\phi}$  is locally bounded (i.e., contains a bounded neighbourhood of zero) and for certain p, 0 , there exists a*p*-homogeneous*F* $-norm <math>\|\cdot\|_1$  equivalent to the original  $\|\cdot\|_{\phi}$  [10].

8. PROPOSITION. Let  $\mu$  be a mean-zero, non-degenerate Gaussian measure defined on  $(L_{\phi}, \mathcal{B}(L_{\phi}))$  such that

(i)  $\phi(t)$  is a convex function, or (ii)  $\phi(\sqrt{t})$  is equivalent to a  $\tilde{\phi}(t)$  concave function and  $\liminf_{t\to\infty} \inf\{c > 0: 2\phi(ct) \ge \phi(t)\} > 0.$  Let  $a = \inf\{I_{\mu}(x): ||x||_1 \ge 1\}$  where  $||\cdot||_1$  is p-homogeneous, 0 , $F-norm equivalent to <math>||\cdot||_{\phi}$ , then  $0 < a < \infty$ , and

$$\lim_{R \to \infty} R^{-2} \log \mu( \{x: ||R^{-1}x||_1 > 1\} ) = -a.$$
  
Proof. Let  $\epsilon = R^{-2}$  and  $B = \{x: ||x||_1 < 1\}$ , then by Corollary 5  

$$\lim_{R \to \infty} R^{-2} \log \mu( \{x: ||R^{-1}x||_1 \ge 1\} )$$

$$= \lim_{\epsilon \to 0} \epsilon \log \mu( \{x: ||\epsilon^{1/2}x||_1 \ge 1\} )$$

$$= \lim_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-1/2}B^c)$$

$$\leq -\inf_{x \in B^c} I_{\mu}(x) = -\inf_{||x||_1 \ge 1} I_{\mu}(x).$$

$$\lim_{R \to \infty} R^{-2} \log \mu( \{x: ||R^{-1}x||_1 > 1\} )$$

$$= \lim_{\epsilon \to 0} \epsilon \log \mu(\epsilon^{-1/2}\overline{B}^c)$$

$$\geq -\inf_{x \in \overline{B}^c} I_{\mu}(x) = -\inf_{||x||_1 \ge 1} I_{\mu}(x).$$

Therefore

$$\inf_{\|x\|_{1} \ge 1} I_{\mu}(x) \leq \lim_{R \to \infty} R^{-2} \log \mu(\{x: \|R^{-1}x\|_{1} > 1\})$$

$$\leq \lim_{R \to \infty} R^{-2} \log \mu(\{x: \|R^{-1}x\|_{1} > 1\})$$

$$\leq \inf_{\|x\|_{1} \ge 1} I_{\mu}(x).$$

Since

$$a = \inf\{I_{\mu}(x):||x||_{1} > 1, x \in \Lambda H_{\mu}\}$$
  
=  $\inf\{c^{2}I_{\mu}(x):||x||_{1} = 1, c > 1, x \in \Lambda H_{\mu}\}$   
=  $\inf\{c^{2}I_{\mu}(x):||x||_{1} = 1, c \ge 1, x \in \Lambda H_{\mu}\}$   
=  $\inf\{I_{\mu}(x):||x||_{1} \ge 1, x \in \Lambda H_{\mu}\}$ 

then

$$\lim_{R\to\infty} R^{-2} \log \mu(\{x: ||R^{-1}x||_1 > 1\}) = -a.$$

If a = 0, then there exists a sequence  $\{F_n\}$  of q.m.f.'s such that  $||\Lambda_{\xi}F_n||_1 > 1$  and  $||F_n|| \le 1/n$ . This implies that

$$|\Lambda_{\xi}F_{n}(t)| \leq K^{\frac{1}{2}}(t, t) ||F_{n}|| \leq \frac{1}{n} K^{\frac{1}{2}}(t, t).$$

Since  $K^{\frac{1}{2}}(t, t) \in L_{\phi}$ , then by the Lebesgue Dominated Convergence Theorem  $\|\Lambda_{\xi}F_n\|_{\phi} \to 0$  as  $n \to \infty$ , which implies that  $\|\Lambda_{\xi}F_n\|_1 \to 0$ as  $n \to \infty$  contradicting the assumption that  $\|\Lambda_{\xi}F_n\|_1 > 1$ , therefore  $0 < a < \infty$ .

LEMMA 1. Let  $(T, \mathcal{F}, m)$  be a measurable space with a  $\sigma$ -finite measure m, and  $L_{\phi}(T, \mathcal{F}, m)$  an Orlicz space, then for any  $\beta > 0$ 

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

is an open set in  $L_{\phi}$ .

*Proof.* It is enough to prove that for some  $\beta > 0$ 

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

is a closed set in  $L_{\phi}$ .

Let  $\{f_n\} \subset D_{\beta}^{e^{\psi}}$  and  $f_n \mapsto f$  in  $L_{\phi}$  as  $n \to \infty$ , then there exists a subsequence  $\{n_k\}$  such that  $f_{n_k}(t) \to f(t)$  m a.e. By Egoroff's theorem [4], there exists an increasing sequence of measurable subsets  $\{E_i\}$  such that the sequence  $\{f_{n_i}\}$  converges uniformly on each  $E_i$  i = 1, 2, ..., and

$$m\left(T\setminus \bigcup_{i=1}^{\infty} E_i\right) = 0.$$

Let

$$T_n = \left\{ t: f(t) > 1 + \frac{1}{n} \right\},$$

then  $T_n$  is an increasing sequence of subsets and

$$S = \{t: f(t) > 1\} = \bigcup_{n=1}^{\infty} T_n.$$

Therefore

$$m(S) = \lim_{n} m(T_n)$$

and we finish the proof by showing that  $m(T_n) \leq \beta$  for each *n*. Let *n* be an arbitrary but fixed, then

$$\forall i \exists n_i \forall n_k > n_i \forall t \in E_i \quad f(t) - \frac{1}{n} < f_{n_k}(t).$$

This implies that

$$\left\{t:f(t) > 1 + \frac{1}{n}\right\} \cap E_i \subset \bigcap_{n_k > n_i} \{t:f_{n_k}(t) > 1\} \cap E_i = A_i.$$

Since  $\{E_i\}$  is an increasing sequence of sets then  $\{n_i\}$  is a non-decreasing sequence implying that  $\{A_i\}$  is an increasing sequence of sets. Since

$$m\left(T\setminus \bigcup_{i=1}^{\infty} E_i\right) = 0,$$

then

$$m\left(\left\{t:f(t)>1+\frac{1}{n}\right\}\right) = m\left(\bigcup_{i=1}^{\infty}\left\{t:f(t)>1+\frac{1}{n}\right\} \cap E_i\right)$$
$$\leq m\left(\bigcup_{i=1}^{\infty}A_i\right) = \lim_{i\to\infty}m(A_i).$$

Since

$$m(A_i) = m(\bigcap_{n_k > n_i} \{t: f_{n_k}(t) > 1\} \cap E_i) \leq \beta \quad \text{for each } i,$$

then

$$m\left(\left\{t:f(t)>1+\frac{1}{n}\right\}\right) \leq \beta \quad \text{for each } n,$$

which proves the lemma.

LEMMA 2. Let  $(T, \mathcal{F}, m)$  be a measurable space with a  $\sigma$ -finite measure m, and  $L_{\phi}(T, \mathcal{F}, m)$  an Orlicz space, then for any  $\beta > 0$  the  $L_{\phi}$ -closure of  $D_{\beta}$ ,  $\overline{D}_{\beta}$  is contained in  $D_{\beta}^{*}$  where

$$D_{\beta}^{*} = \left\{ f(t): f(t) \in L_{\phi}, \\ \forall k = 1, 2, \dots, m\left(\left\{ t: f(t) > 1 - \frac{1}{k} \right\} \right) \ge \beta \right\}.$$

*Proof.* Let  $\{f_n\} \subset D$  and  $f_n \to f$  in  $L_{\phi}$  as  $n \to \infty$ , then  $f_n \to f$  in measure *m* as  $n \to \infty$ , i.e.,

$$\forall k \forall \epsilon \exists n_{k,\epsilon} \forall n > n_{k,\epsilon} m \left( \left\{ t : |f_n(t) - f(t)| \ge \frac{1}{k} \right\} \right) < \epsilon$$

which implies that

$$m\left(\left\{t:f_n(t)>1, |f_n(t)-f(t)|<\frac{1}{k}\right\}\right)>\beta-\epsilon.$$

Therefore

$$m\left(\left\{t:f(t)>1-\frac{1}{k}\right\}\right)>\beta-\epsilon$$
 for each k and  $\epsilon$ 

and

$$m\left(\left\{t:f(t)>1-\frac{1}{k}\right\}\right) \geq \beta$$
 for each  $k$ ,

which proves that  $f \in D^*_{\mathcal{B}}$ .

LEMMA. 3. Let  $(L_{\phi}, \mathscr{B}(L_{\phi}), \mu)$  be an Orlicz space with a mean-zero, non-degenerate Gaussian measure  $\mu$  and a rate function  $I_{\mu}(\cdot)$ :

$$I_{\mu}(x) = \frac{1}{2} ||\Lambda^{-1}x||^2 \quad if \ x \in \Lambda H_{\mu}$$
  
$$\infty \qquad if \ x \notin \Lambda H_{\mu}.$$

Let

$$egin{aligned} a_eta &= \inf\{I_\mu(x): x \in D_eta\},\ \overline{a}_eta &= \inf\{I_\mu(x): x \in \overline{D}_eta\},\ a_eta^* &= \inf\{I_\mu(x): x \in D_eta\}, \end{aligned}$$

then  $0 < a_{\beta}^* \leq \overline{a}_{\beta} \leq a_{\beta}$ . If the covariance function K(s, t) of a measurable stochastic process  $\xi = \{\xi(t): t \in T\}$  inducing the measure  $\mu$  is such that

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

then  $a_{\beta} < \infty$  for every  $\beta > 0$ .

*Proof.* If  $a_{\beta}^* = 0$  then there exists a sequence  $\{\Lambda F_n\} \subset D_{\beta}^*$  such that  $||F_n|| < 1/n$  and for almost every t

$$|\Lambda F_n(t)| \leq K^{\frac{1}{2}}(t, t) ||F_n|| < \frac{1}{n} K^{\frac{1}{2}}(t, t).$$

This implies that for each k

$$\left\{t:\Lambda F_n(t)>1-\frac{1}{k}\right\}\subset \left\{t:\frac{1}{n}K^{\nu_2}(t,t)>1-\frac{1}{k}\right\},\$$

and for each k and n

$$m\left(\left\{t:K^{V_2}(t,\,t)>\left(1-\frac{1}{k}\right)n\right\}\right)\geq\beta.$$

Let k be an arbitrary but fixed and

$$A_n = \left\{ t: K^{\frac{1}{2}}(t, t) > \left(1 - \frac{1}{k}\right)n \right\}$$

for every *n*, since  $K^{\frac{1}{2}}(t, t) \in L_{\phi}$  then there exists a > 0 such that

$$m(A_n)\phi\left(a\left(1-\frac{1}{k}\right)n\right) \leq \int \phi(aK^{\frac{1}{2}}(t,t))m(dt) < \infty$$

which implies that  $m(A_n) < \infty$  for every *n*.

Since  $\{A_n\}$  is a decreasing sequence then

$$\lim_{n} m(A_{n}) = m\left(\bigcap_{n=1}^{\infty} A_{n}\right) \text{ and}$$
$$m\left(\bigcap_{n=1}^{\infty} A_{n}\right) \ge \beta$$

implying that

$$m(\{t:K^{\frac{1}{2}}(t,t)=\infty\}) \geq \beta$$

which is impossible. Therefore  $0 < a^* \leq \overline{a}_{\beta} \leq a_{\beta}$ .

Let  $\xi = \{\xi(t): t \in T\}$  be a measurable stochastic process such as in A, inducing the measure  $\mu$  with the covariance function K(s, t) satisfying (\*). There exists a measurable subset  $T_0$ ,  $m(T_0) = 0$  such that for every  $s \in T \setminus T_0$ ,  $\xi(s)$  is a q.m.f. Let  $\beta > 0$  be an arbitrary but fixed, then there exists a q.m.f.  $\xi(s)$  such that

$$m(\{t:\Lambda\xi(s)(t)>0\})>\beta.$$

Let

or

$$A_n = \left\{ t: \Lambda \xi(s)(t) > \frac{1}{n} \right\},$$
  
$$\left\{ t: \Lambda \xi(s)(t) > 0 \right\} = \bigcup_{n=1}^{\infty} A_n.$$

Since  $\{A_n\}$  is an increasing sequence, then there exists *n* such that  $m(A_n) > \beta$  implying that for a q.m.f.

$$F = n\xi(s), m(\{t: \Lambda F(t) > 1\}) > \beta \text{ and } a_{\beta} \leq \frac{1}{2} ||F||^{2}.$$

9. THEOREM. 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

(i)  $\phi(t)$  is a convex function,

(ii)  $\phi(\sqrt{t})$  is equivalent to  $\tilde{\phi}(t)$  concave function. Let for any  $\beta > 0$ 

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

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

then

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

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

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

*Proof.* Let  $\mu$  denote Gaussian measure generated by the stochastic process  $\xi = \{\xi(t): t \in T\}$ . By Lemma 1  $D_{\beta}$  is an open set. Since

$$\mu(\alpha D) = P(\{\omega: m(\{t:\xi(t, \omega) > \alpha\}) > \beta\})$$

and  $\mu(\alpha \overline{D}) \ge \mu(\alpha D)$  then by Corollary 5

$$-a_{\beta} \leq \lim_{\alpha \to \infty} \alpha^{-2} \log \mu(\alpha D) \leq \lim_{\alpha \to \infty} \alpha^{-2} \log \mu(\alpha D) \leq -\overline{a}_{\beta}$$

Under the additional assumptions by Lemma 3,  $0 < \overline{a}_{\beta} \leq a_{\beta} < \infty$ . Since the covariance function K(s, t) is continuous the space  $\Lambda H_{\mu}$  consists of continuous functions.

To finish the proof of the theorem, by Lemma 3 it is sufficient to show that  $a_{\beta}^* \ge a_{\beta}$ . Let F be an arbitrary q.m.f. such that for each k

$$m\left(\left\{t:\Lambda_{\xi}F(t)>1-\frac{1}{k}\right\}\right) \geq \beta.$$

Let

$$G_k = \left(1 + \frac{1}{k-1}\right)F$$

then for any k

$$m(\{t:\Lambda_{\xi}G_k(t)>1\}) \geq \beta.$$

Since for any open set U, m(U) > 0, and for each  $k \Lambda_{\xi}G_k(t)$  is a continuous function, then for any k and n

$$m\left(\left\{t:1-\frac{1}{n}<\Lambda_{\xi}G_{k}(t)<1\right\}\right)>0.$$

Let

$$H_{n,k} = \left(1 + \frac{1}{n-1}\right)G_k = \left(1 + \frac{1}{n-1}\right)\left(1 + \frac{1}{k-1}\right)F,$$

then

$$m(\{t:\Lambda_{\xi}H_{n,k}(t) > 1\}) > \beta.$$

Since for each  $n, k H_{n,k} \in D_{\beta}, ||H_{n,k}|| \mapsto ||F||$  when  $n \to \infty, k \to \infty$  this implies that

$$\inf\{I_{\mu}(x):x \in D_{\beta}, x \in \Lambda H_{\mu}\} = \inf\{I_{\mu}(x):x \in D_{\beta}^{*}, x \in \Lambda H_{\mu}\}$$

and  $a_{\beta}^* = a_{\beta} = \overline{a}_{\beta}$ .

10. COROLLARY. Let  $(T, \mathcal{F}, m)$  be a real line with Borel  $\sigma$ -algebra  $\mathcal{F}$  and Lebesgue measure m. Let  $\xi = \{\xi(t): t \in T\}$  be a mean-zero Gaussian stochastic process, continuous in probability with almost all its sample paths in  $L_p = L_p(T, \mathcal{F}, m), 0 , such that for any <math>\beta > 0$ 

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

then for any  $\beta > 0$  there exists  $0 < a_{\beta} < \infty$  such that

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

## REFERENCES

- 1. T. Byczkowski, Gaussian measures on  $L_p$  spaces, 0 , Studia Math. 59 (1977), 249-261.
- 2. —— Norm convergent expansion for  $L_{\phi}$ -valued Gaussian random elements, Studia Math. 64 (1979), 87-95.
- 3. M.D. Donsker and S.R.S. Varadhan, Asymptotic evaluation of certain Markov processes expectations for large time III, Comm. Pure Appl. Math. 29 (1976), 389-461.
- 4. P.R. Halmos, Measure theory (New York, Springer-Verlag, 1974).
- 5. E. Hille and R. Phillips, *Functional analysis and semigroups*, AMS Colloquium Publications, 31 (1975).
- G. Kallianpur and H. Oodaira, Freidlin-Wentzell type estimates for abstract Wiener spaces, Sankhyà 40, Series A (1978), 116-137.
- A.T. Lawniczak, Gaussian measures on Orlicz spaces and abstract Wiener spaces, Lecture Notes in Mathematics 939 (Springer-Verlag, 1982), 81-97.
- 8. Gaussian measures on Orlicz spaces, Stoch. Analysis and its Appl. 3 (1985).
- **9.** N. Marlow, *High level occupation times for continuous Gaussian processes*, Ann. of Probab. *3* (1973), 388-397.

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10. S. Rolewicz, Metric linear spaces (Polish Scientific Publishers, PWN, 1972).

- 11. D.W. Strook, An introduction to the theory of large deviations (Springer-Verlag, 1984).
- 12. S.R.S. Varadhan, Large deviations and applications, SIAM (1984).

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