MODERATE DEVIATIONS AND ASSOCIATED LAPLACE APPROXIMATIONS FOR SUMS OF INDEPENDENT RANDOM VECTORS

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ABSTRACT. Let $\{X_j\}$ be an i.i.d. sequence of Banach space valued r.v.'s and let $S_n = \sum_{j=1}^n X_j$. For certain positive sequences $b_n \to \infty$, we determine the exact asymptotic behavior of $E \exp\{(b_n^2/n)\Phi(S_n/b_n)\}$, where Φ is a smooth function. We also prove a large deviation principle for $\{\mathscr{L}(S_n/b_n)\}$.

1. Introduction

Let $\{X_j\}$ be a sequence of independent E-valued random vectors with common distribution μ , where E is a separable Banach space, let $S_n = \sum_{j=1}^n X_j$, and assume that $\{\mathcal{L}(S_n/n^{1/2})\}$ converges weakly. Let $\{b_n\}$ be a positive sequence such that $b_n/n^{1/2} \to \infty$. In this paper we study certain aspects of the asymptotic behavior of $\{\mathcal{L}(S_n/b_n)\}$ (under the further assumption (1.1), $\{P\{S_n/b_n \in A\}\}$ are sometimes referred to in the literature as "probabilities of moderate deviations").

One of our results is a large deviation principle (in the sense of Varadhan [20]) for $\{\mathcal{L}(S_n/b_n)\}$ when $\{b_n\}$ is such that

$$(1.1) b_n/n \to 0.$$

We prove that, under appropriate integrability conditions,

(1.2)
$$\limsup_{n\to\infty} \frac{n}{b_n^2} \log P\{S_n/b_n \in F\} \le -\inf_{x\in F} I(x) \quad \text{for } F \text{ closed},$$

(1.3)
$$\liminf_{n\to\infty} \frac{n}{b_n^2} \log P\{S_n/b_n \in G\} \ge -\inf_{x\in G} I(x) \quad \text{for } G \text{ open.}$$

The rate function I depends on μ only through its covariance structure; this is in contrast to the situation that arises when $b_n = n$ (see [14]; also [8, 7, 2]). The precise statement is given in Theorem 2.3; Theorem 2.2 is a more general result about triangular arrays which we need in §3. Parts of Theorem 2.3 were obtained by Borovkov and Mogulskii [13] (see also de Acosta and

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Kuelbs [5] and Bolthausen [10]). For the case when E is finite-dimensional (and the covariance matrix of X_1 is nonsingular), see Freidlin and Wentzell [17, p. 142]. Of course, in the case $E = \mathbf{R}$ there is an extensive literature on the exact asymptotics of $\{P\{S_n/b_n > x\}\}\$; for classical results in this context see [16, 19].

If $\Phi: E \to \mathbb{R}$ is a bounded continuous function, then by Varadhan's theorem (see [15, p. 51]), (1.2) and (1.3) imply

(1.4)
$$\lim_{n \to \infty} \frac{n}{b_n^2} \log E \exp \left\{ \frac{b_n^2}{n} \Phi \left(\frac{S_n}{b_n} \right) \right\} = \sup_{x \in E} [\Phi(x) - I(x)].$$

Our main result is a refinement of (1.4), giving the exact asymptotic behavior of $E \exp\{(b_n^2/n)\Phi(S_n/b_n)\}$ when Φ is a smooth function subject to a growth condition and (1.1) is replaced by the stronger condition

$$(1.5) b_n/n^{2/3} \to 0.$$

Under appropriate integrability, tightness and nondegeneracy conditions, we prove that

(1.6)
$$E \exp\left\{\frac{b_n^2}{n}\Phi(S_n/b_n)\right\} \sim C \exp\left\{\frac{b_n^2}{n}\sup_{x\in E}[\Phi(x)-I(x)]\right\},$$

where the constant C depends on Φ and the covariance structure of μ . The precise statement is given in Theorem 3.1 and a simple example shows that condition (1.5) cannot in general be relaxed; it should be remarked that (1.5) appears also in classical results on the exact asymptotic behavior of $\{P\{S_n/b_n > x\}\}$ (see e.g. [16]). In the case $b_n = n$, results similar to (1.6) were obtained by Bolthausen [10] for general E and previously by Martin-Löf [18] for $E = \mathbf{R}$; in contrast to (1.6), in their results the rate function I is replaced by the Crámer functional of μ (see [7]). §3 of the present paper is close in spirit to Bolthausen's interesting work (he has also studied in [11] the more complicated situation that occurs in the case $b_n = n$ when the nondegeneracy assumption is dropped). For reference to previous results in the case when μ is Gaussian, see [10].

§2 contains the proof of the moderate deviation results and §3 that of the Laplace approximation (1.6).

Throughout the paper E will denote a separable Banach space and E^* its dual space.

2. Moderate deviations

It will be convenient for the developments in §3 to prove (1.2) and (1.3) in a somewhat more general setting. Let $\{X_{nj}: n \in \mathbb{N}, j=1,\ldots,n\}$ be a triangular array of *E*-valued random vectors, each row of which is independent and identically distributed, and let $S_n = \sum_{j=1}^n X_{nj}$. We will consider the following conditions:

$$(2.1) EX_{n1} = 0 for all n \in \mathbb{N},$$

$$(2.2) {\mathscr{L}(X_{n1})} is tight,$$

(2.3)
$$\sup_{n} E \exp(t||X_{n1}||) < \infty \quad \text{for all } t > 0,$$

(2.4) $\{\mathcal{L}(S_n/n^{1/2})\}$ converges weakly to a centered Gaussian measure γ ,

(2.5) If
$$j_n \in \mathbb{N}$$
, $j_n \to \infty$, $j_n \le n$, then $\left\{ \mathcal{L}\left(\sum_{j=1}^{j_n} X_{nj}/j_n^{1/2}\right) \right\}$ is tight.

If E is a Banach space of type 2, then it is well known that (2.4)–(2.5) follow automatically from (2.1)–(2.3); let us recall that the class of type 2 spaces includes \mathbb{R}^k , Hilbert space and L^p spaces for $p \ge 2$ (see e.g. [6]).

Lemma 2.1. Assume that $\{X_{nj}\}$ satisfies conditions (2.1)–(2.4) and let $\{b_n\}$ be a positive sequence satisfying the conditions $b_n/n^{1/2} \to \infty$ and (1.1). Then

(i) There exist c > 0, $n_0 \in \mathbb{N}$ and a compact convex symmetric set K such that for $n \ge n_0$,

$$E\exp\left\{\frac{b_n}{n}q_K(S_n)\right\}\leq c^{b_n^2/n}\,,$$

where q_K is the Minkowski functional of K.

(ii) For every $\xi \in E^*$,

$$\lim_{n\to\infty}\frac{n}{b_n^2}\log E\exp\left\{\frac{b_n}{n}\langle\xi\,,\,S_n\rangle\right\}=\frac{1}{2}\int\xi^2\,d\gamma\,.$$

Proof. By assumptions (2.3) and (2.4), an easy modification of Theorem 5.1(b) of [4] yields:

(2.6)
$$\sup_{n} E \exp\{t \|S_n/n^{1/2}\|\} < \infty \text{ for all } t > 0.$$

Now using (2.4) and (2.6), and also (2.2) and (2.3), we obtain by Theorem 3.1 of [1]: there exists a compact, convex, symmetric set K such that

(2.7)
$$\sup_{n} E \exp\{q_{K}(S_{n}/n^{1/2})\} < \infty, \qquad \sup_{n} E \exp\{2q_{K}(X_{n1})\} < \infty.$$

We use now a well-known technique of Yurinskii [21]. For fixed n, let $\mathscr{F}_0 = \{\phi, \Omega\}$, $\mathscr{F}_j = \sigma(X_{n1}, \ldots, X_{nj})$ for $1 \le j \le n$. For $j = 1, \ldots, n$ let

$$\eta_j = E[q_K(S_n)|\mathscr{F}_j] - E[q_K(S_n)|\mathscr{F}_{j-1}].$$

Then

$$q_K(S_n) - Eq_K(S_n) = \sum_{j=1}^n \eta_j$$

and

$$(2.8) |\eta_j| \leq q_K(X_{nj}) + Eq_K(X_{nj}).$$

Now for $0 < \lambda \le 1$,

 $E\exp\{\lambda[q_K(S_n) - Eq_K(S_n)]\} = E\exp\left\{\lambda\sum_{i=1}^n \eta_i\right\}$

$$= EE \left[\exp \left\{ \lambda \sum_{j=1}^{n} \eta_{j} \right\} | \mathscr{F}_{n-1} \right] = E \exp \left\{ \lambda \sum_{j=1}^{n-1} \eta_{j} \right\} E[\exp \{\lambda \eta_{n}\} | \mathscr{F}_{n-1}].$$

Next.

(2.10)
$$E[\exp[(\lambda \eta_{n})|\mathscr{F}_{n-1}] \leq E\left[\left(1 + \lambda \eta_{n} + \frac{\lambda^{2} \eta_{n}^{2}}{2} e^{|\eta_{n}|}\right)|\mathscr{F}_{n-1}\right]$$

$$= 1 + \frac{\lambda^{2}}{2} E[\eta_{n}^{2} e^{|\eta_{n}|}|\mathscr{F}_{n-1}]$$

$$\leq 1 + \lambda^{2} E[e^{2|\eta_{n}|}|\mathscr{F}_{n-1}]$$

$$\leq \exp\{\lambda^{2} E e^{2(q_{K}(X_{n1}) + Eq_{K}(X_{n1}))}\},$$

using (2.8). Iterating the same procedure and taking $\lambda = b_n/n$ for sufficiently large n, we have by (2.7), (2.9) and (2.10), for certain constant $\alpha > 0$, $\beta > 0$,

$$E \exp\left\{\frac{b_n}{n}q_K(S_n)\right\} \le \exp\left\{\frac{b_n}{n}Eq_K(S_n)\right\} \exp\left\{n \cdot \frac{b_n^2}{n^2}\alpha\right\}$$

$$= \exp\left\{\frac{b_n}{n^{1/2}}Eq_K(S_n/n^{1/2})\right\} \exp\left\{\alpha\frac{b_n^2}{n}\right\}$$

$$\le \exp\left\{\beta\frac{b_n^2}{n}\right\} \exp\left\{\alpha\frac{b_n^2}{n}\right\}$$

for sufficiently large n. This proves statement (i).

(ii) It follows from (2.1), (2.3) and (2.4) that for all $\xi \in E^*$,

(2.11)
$$\lim_{n\to\infty} E\langle \xi, X_{n1}\rangle^2 = \lim_{n\to\infty} E\langle \xi, S_n/n^{1/2}\rangle^2 = \int \xi^2 d\gamma.$$

For $\lambda \in \mathbf{R}$, we have by (2.1)

$$E \exp\langle \lambda \xi, S_n \rangle = (E \exp\langle \lambda \xi, X_{n1} \rangle)^n$$

= $(1 + \frac{1}{2} \lambda^2 E[\langle \xi, X_{n1} \rangle^2 e^{\theta \lambda \langle \xi, X_{n1} \rangle}])^n$

where $|\theta| \le 1$. Setting $\lambda = b_n/n$, it follows by (2.3), (2.11) and dominated convergence that

$$\lim_{n\to\infty} \left(E \exp\left\langle \frac{b_n}{n} \xi, S_n \right\rangle \right)^{n/b_n^2} = \exp\left(\frac{1}{2} \int \xi^2 \, d\gamma \right). \quad \Box$$

Let us recall that the Crámer functional of γ , defined by

$$I(x) = \sup_{\xi \in E^*} \left[\langle \xi, x \rangle - \frac{1}{2} \int \xi^2 \, d\gamma \right] \,,$$

is given by

(2.12)
$$I(x) = \begin{cases} \frac{1}{2} ||x||_{\gamma}^{2}, & x \in H_{\gamma}, \\ \infty, & x \notin H_{\gamma}, \end{cases}$$

where $(H_{\gamma}, \|\cdot\|_{\gamma})$ is the Hilbert space associated to γ (see [7]; also [12]). We obtain now a large deviation principle for $\{\mathcal{L}(S_n/b_n)\}$.

Theorem 2.2. Let $\{b_n\}$ be a positive sequence satisfying the conditions $b_n/n^{1/2} \to \infty$ and (1.1). Let I be given by (2.12).

(i) Assume that $\{X_{nj}\}$ satisfies (2.1)–(2.4). Then for every closed set F,

$$\limsup_{n\to\infty}\frac{n}{b_n^2}\log P\{S_n/b_n\in F\}\leq -\inf_{x\in F}I(x).$$

(ii) Assume that $\{X_{nj}\}$ satisfies (2.1) and (2.3)–(2.5). Then for every open set G,

$$\liminf_{n\to\infty} \frac{n}{b_n^2} \log P\{S_n/b_n \in G\} \ge -\inf_{x\in G} I(x).$$

Proof. (i) Let $Y_n = b_n S_n/n$, $\phi(\xi) = \frac{1}{2} \int \xi^2 d\gamma$ for $\xi \in E^*$. The result will follow from Theorem 2.1 of [1], taking the normalization in that result to be b_n^2/n instead of n (obviously the result remains valid with this change). By Lemma 2.1(ii), assumption (2.1) of [1] is verified. We show next that assumption (2.3) of [1] is satisfied. Let a > 0, and let K be as in Lemma 2.1(i). By Lemma 2.1(i), for $\alpha > 0$

$$P\left\{\frac{n}{b_n^2}Y_n \notin \alpha K\right\} = P\left\{q_K(Y_n) > \alpha \frac{b_n^2}{n}\right\}$$

$$\leq e^{-\alpha \frac{b_n^2}{n}} E \exp\left\{\frac{b_n}{n} q_K(S_n)\right\}$$

$$\leq \exp\left[-(\alpha - \log c) \frac{b_n^2}{n}\right] \quad \text{for } n \geq n_0$$

$$\leq \exp\left[-a \frac{b_n^2}{n}\right] \quad \text{for } n \geq n_0$$

if $\alpha \ge a + \log c$.

(ii) We first remark that Lemma 3.1 of [5] is valid for triangular arrays under assumptions (2.1) and (2.3)–(2.5) (obviously the integrability assumption may be weakened). The proof is just a reinterpretation of that of Lemma 3.1 of [5]; notice that statements (3.5) and (3.6) in [5] follow from standard facts about triangular arrays; (2.5) is needed to verify (3.6) of [5] (see [3]). Now let G be an open set. We must prove that if $h \in G \cap H_{\gamma}$, then

$$\liminf_{n\to\infty}\frac{n}{b_n^2}\log P\{S_n/b_n\in G\}\geq -\frac{\|h\|_{\gamma}^2}{2}.$$

But this follows from Lemma 3.2 of [5]. \Box

The following result deals with the case of an independent, identically distributed sequence. In order to formulate it, we consider the following conditions on a probability measure μ on E:

(2.14)
$$\int e^{t||x||} \mu(dx) < \infty \quad \text{for all } t > 0,$$

(2.15)
$$\{\mu^{*n}(n^{1/2}(\cdot))\}$$
 is tight.

As is well known (2.15) follows from (2.13) and (2.14) if E is a Banach space space of type 2 (see e.g. [6]). Under condition (2.15), $\{\mu^{*n}(n^{1/2}(\cdot))\}$ converges weakly to a Gaussian measure γ with the same covariance structure as μ . Let $(H_{\mu}, \|\cdot\|_{\mu})$ be the Hilbert space associated to μ ; then $(H_{\mu}, \|\cdot\|_{\mu})$ depends only on the covariance structure of μ and $(H_{\mu}, \|\cdot\|_{\mu}) = (H_{\gamma}, \|\cdot\|_{\gamma})$, and therefore

(2.16)
$$I(x) = \begin{cases} \frac{1}{2} ||x||_{\mu}^{2}, & x \in H_{\mu}, \\ \infty, & x \notin H_{\mu}. \end{cases}$$

Theorem 2.3. Let $\{b_n\}$ be as in Theorem 2.2 and let I be given by (2.16). Let $\{X_j\}$ be an independent sequence with common distribution μ and let $T_n = \sum_{j=1}^n X_j$.

(i) Assume that μ satisfies (2.13)–(2.15). Then for every closed set F,

$$\limsup_{n\to\infty}\frac{n}{b_n^2}\log P\{T_n/b_n\in F\}\leq -\inf_{x\in F}I(x).$$

(ii) Assume that μ satisfies (2.15). Then for every open set G,

$$\liminf_{n\to\infty}\frac{n}{b_n^2}\log P\{T_n/b_n\in G\}\geq -\inf_{x\in G}I(x).$$

Proof. The first statement is a direct corollary of Theorem 2.2(i). The second statement follows from [5], Lemmas 3.1 and 3.2. \Box

3. LAPLACE APPROXIMATIONS

We have adopted in this section the framework of Bolthausen [10] for the case $b_n = n$. The central part of the proof of Theorem 3.1 is, however, different and is based on the decomposition given by Lemma 3.3(i); the use of this decomposition is illustrated in a simple way in the proof of Theorem 3.1(i).

In order to formulate Theorem 3.1 we will consider a probability measure μ on E satisfying the following conditions; let us observe that (3.3) is a strengthening of (2.15) (take $\xi = 0$). Let $\hat{\mu}(\xi) = \int e^{\xi} d\mu$ for $\xi \in E^*$.

$$\int x\mu(dx) = 0,$$

(3.2)
$$\int e^{t||x||} \mu(dx) < \infty \quad \text{for all } t > 0,$$

(3.3) Let
$$\xi \in E^*$$
 and $\alpha_n > 0$, $\alpha_n \to 0$.

Let

$$d\nu_n = \frac{e^{\alpha_n \xi}}{\hat{\mu}(\alpha_n \xi)} d\mu, \qquad v_n = \int x \nu_n(dx).$$

Then

$$\{(\nu_n * \delta_{-\nu_n})^{*n}(n^{1/2}(\cdot))\}$$
 is tight.

Condition (3.3) follows from (3.1)–(3.2) if E is a type 2 space.

We will also consider a function $\Phi: E \to \mathbf{R}$ such that

(3.4)
$$\Phi \in C^2(E)$$
 in the Fréchet sense,

(3.5)
$$\Phi(x) \le a||x|| + b \text{ for certain constants } a > 0, \ b > 0.$$

The following two assumptions should be considered in the light of Lemma 3.2. They also appear in [10].

(3.6) There exists a unique point
$$x^* \in E$$
 such that $\Phi(x^*) - I(x^*)$

$$= \sup_{x \in E} [\Phi(x) - I(x)], \text{ where } I \text{ is given by (2.16)},$$

$$(3.7) \ \ \text{For all} \ \xi \in E^* \ \text{with} \ \int \xi^2 \, d\mu > 0 \, , \ \ \text{setting} \ \Delta(\xi) = \int x \langle \xi \, , \, x \rangle \mu(dx) \, ,$$

$$D^2 \Phi(x^*) (\Delta(\xi) \, , \, \Delta(\xi)) < \int \xi^2 \, d\mu \, .$$

Theorem 3.1. Let μ be a probability measure on E satisfying (3.1)–(3.3), let $\{X_j\}$ be an independent sequence with common distribution μ , and let $S_n = \sum_{j=1}^n X_j$. Let $\Phi: E \to \mathbf{R}$ satisfy (3.4) and (3.5), and assume that (3.6) and (3.7) hold.

(i) Let γ be the Gaussian measure to which $\{\mu^{*n}(n^{1/2}(\cdot))\}$ converges weakly by (3.3) with $\xi = 0$. Then

$$C = \int \exp\left\{\frac{1}{2}D^2\Phi(x^*)(y,y)\right\}\gamma(dy) < \infty.$$

(ii) Let $\{b_n\}$ be a positive sequence satisfying the conditions $b_n/n^{1/2} \to \infty$ and (1.5). Then

(3.8)
$$\lim_{n\to\infty} \exp\left\{-\frac{b_n^2}{n}\sup_{x\in E}[\Phi(x)-I(x)]\right\} E\exp\left\{\frac{b_n^2}{n}\Phi(S_n/b_n)\right\} = C.$$

Remarks. (1) If, furthermore, it is assumed that $\int \xi^3 d\mu = 0$ for all $\xi \in E^*$, then (3.8) is valid for any positive sequence $\{b_n\}$ such that $b_n/n^{1/2} \to \infty$ and $b_n/n^{3/4} \to 0$. This follows from the proof of Theorem 3.1.

- (2) Theorem 2.2(i) is used in the proof of Theorem 3.1(ii), but not Theorem 2.2(ii).
- (3) The following example shows that assumption (1.5) cannot in general be relaxed. Let $\mu = \rho * \delta_{-1}$, where ρ is the standard Poisson distribution with parameter 1; then $\mathcal{L}(S_n/n^{1/2})$ converges weakly to $\gamma = N(0,1)$. Let $\Phi(x) = x$; then $\sup_x [\Phi(x) I(x)] = \sup_x [x \frac{1}{2}x^2] = \frac{1}{2}$. Also $\Phi''(x) = 0$ for all x, so C = 1. Now

$$\exp\left\{-\frac{1}{2}\frac{b_n^2}{n}\right\} E \exp\left\{\frac{b_n^2}{n} \cdot \frac{S_n}{b_n}\right\} = \exp\left\{-\frac{1}{2}\frac{b_n^2}{n} - b_n + n(e^{b_n/n} - 1)\right\},$$

$$= \exp\left\{\frac{1}{6}\left(\frac{b_n}{n^{2/3}}\right)^3 + nO\left(\left(\frac{b_n}{n}\right)^4\right)\right\}.$$

We shall need several lemmas for the proof of Theorem 3.1. For background information on $(H_{\mu}, \|\cdot\|_{\mu})$ we refer again to [7, 12].

Lemma 3.2. (i) $s = \sup_{x \in E} [\Phi(x) - I(x)]$ is attained. Moreover, if $\Phi(x^*) - I(x^*) = s$, then $x^* = \Delta(D\Phi(x^*))$; in particular, $x^* \in \Delta(E^*)$.

(ii) If $\Phi(x^*) - I(x^*) = s$, then

$$D^2\Phi(x^*)(h, h) \leq ||h||_{\mu}^2$$
 for all $h \in H_{\mu}$.

Proof. (i) By (3.5), $\Phi(x) - I(x) \le (a\|x\| + b) - \frac{1}{2}\|x\|_{\mu}^{2}$. Since $\|x\| \le \sigma \|x\|_{\mu}$, where $\sigma = \sup_{\|x\|_{\mu} \le 1} \|x\|$, $\{x \in H_{\mu} \colon \|x\|_{\mu} \le r\}$ is compact for all $r \ge 0$ and (3.4) holds, it follows that s is finite and is attained on H_{μ} .

Let $x^* \in H_{\mu}$ be such that $s = \Phi(x^*) - I(x^*)$. For fixed $h \in H_{\mu}$, $t \in \mathbf{R}$, let

$$\begin{split} f(t) &= \Phi(x^* + th) - I(x^* + th) \\ &= \Phi(x^* + th) - \frac{1}{2} (\|x^*\|_{\mu}^2 + 2t\langle x^*, h \rangle_{\mu} + t^2 \|h\|_{\mu}^2) \,. \end{split}$$

Then $f'(t)=D\Phi(x^*+th)(h)-\langle x\,,\,h\rangle_\mu-t\|h\|_\mu^2$, and $0=f'(0)=D\Phi(x^*)(h)-\langle x^*\,,\,h\rangle_\mu$. It follows that for all $h\in H_\mu$,

$$\langle \Delta(D\Phi(x^*)), h \rangle_{\mu} = D\Phi(x^*)(h) = \langle x^*, h \rangle_{\mu},$$

implying $\Delta(D\Phi(x^*)) = x^*$.

(ii) Similarly,

$$0 \ge f''(0) = D^2 \Phi(x^*)(h, h) - ||h||_{\nu}^2$$
. \square

Conditions (3.6) and (3.7) are a strengthening of the conclusions of Lemma 3.2; the latter one should be viewed as a nondegeneracy condition.

Lemma 3.3. (i) Let $\{\xi_1, \ldots, \xi_k\} \subset E^*$ be such that $\int \xi_i \xi_j d\mu = \delta_{ij}$, and let $e_j = \Delta(\xi_j)$. For $x \in E$, define

$$P_k(x) = \sum_{j=1}^k \langle \xi_j, x \rangle e_j, \qquad Q_k(x) = x - P_k(x).$$

Let A be a continuous symmetric bilinear form on $E \times E$. Then for every $x \in E$, $\beta > 0$,

$$A(x, x) \le d(1 + \beta^2) \|Q_k(x)\|^2 + (\alpha + \beta^{-2} d\sigma^2) \sum_{j=1}^k \langle \xi_j, x \rangle^2,$$

where $d = \sup\{|A(x, y)| \colon \|x\| \le 1, \|y\| \le 1\}$, $\alpha = \sup_{\|x\|_{\mu} \le 1} A(x, x)$, $\sigma = \sup_{\|x\|_{\mu} < 1} \|x\|$.

(ii) Assume that A satisfies: $A(\Delta(\xi), \Delta(\xi)) < \int \xi^2 d\mu$ for all $\xi \in E^*$ with $\int \xi^2 d\mu > 0$. Then $\sup_{\|x\|_{\mu} \le 1} A(x, x) < 1$.

Proof. (i) Since $x = P_k(x) + Q_k(x)$, we have

$$A(x, x) = A(Q_k(x), Q_k(x)) + 2A(Q_k(x), P_k(x)) + A(P_k(x), P_k(x))$$

and therefore

$$(3.9) A(x, x) \le d\|Q_k(x)\|^2 + \alpha \|P_k(x)\|_{\mu}^2 + 2d\|Q_k(x)\| \|P_k(x)\|.$$

Now

$$||P_{k}(x)||_{\mu}^{2} = \sum_{j=1}^{k} \langle \xi_{j}, x \rangle^{2},$$

$$2||Q_{k}(x)|| ||P_{k}(x)|| \le \beta^{2} ||Q_{k}(x)||^{2} + \beta^{-2} ||P_{k}(x)||^{2}$$

$$\le \beta^{2} ||Q_{k}(x)||^{2} + \sigma^{2} \beta^{-2} ||P_{k}(x)||_{\mu}^{2}.$$

The conclusion follows from (3.9)–(3.11).

(ii) By continuity and the $\|\cdot\|_{\mu}$ -density of $\Delta(E^*)$ in H_{μ} , the assumption implies $s=\sup_{\|x\|_{\mu}=1}A(x\,,\,x)\leq 1$. Suppose s=1. By the compactness of $\{x\in H_{\mu}\colon \|x\|_{\mu}\leq 1\}$, there exists $h\in H_{\mu}$ such that $\|h\|_{\mu}=1$ and $A(h\,,\,h)=1$. Now for all $y\in H_{\mu}$, t>0,

$$A(h + ty, h + ty) \le ||h + ty||_{\mu}^{2}$$

and expanding both sides and cancelling we get

$$2tA(h, y) + t^2A(y, y) \le 2t\langle h, y\rangle_{\mu} + t^2||y||_{\mu}^2$$

Dividing by t, letting $t \to 0$ and observing that (-y) may be substituted for y, we have $A(h, y) = \langle h, y \rangle_{\mu}$. But

$$A(h, v) = A(h, \cdot)(v) = \langle \Delta(A(h, \cdot)), v \rangle_{\mu},$$

and therefore for all $y \in H_u$,

$$\langle \Delta(A(h,\cdot)) - h, y \rangle_{\mu} = 0,$$

implying $h = \Delta(A(h, \cdot)) \in \Delta(E^*)$, which contradicts the assumption. \square

Lemma 3.4. Let $\{Y_{nj}: n \in \mathbb{N}, j = 1, ..., n\}$ be a row-wise independent triangular array with

$$d\mathscr{L}(Y_{nj}) = \frac{e^{b_n \varphi/n}}{\hat{\mu}(\frac{b_n}{n}\varphi)} d\mu \quad \text{for } j = 1, \ldots, n,$$

where $\varphi = D\Phi(x^*)$, and let $T_n = \sum_{j=1}^n (Y_{nj} - E(Y_{nj}))$. Let $u_n = (n/b_n)E(Y_{n1}) - x^*$, and define for $v \in E$,

$$\begin{split} f_n(v) &= \exp\left\{\frac{b_n^2}{n}\left[\Phi\left(x^* + u_n + \frac{n^{1/2}}{b_n}v\right) - \Phi(x^*) - \left\langle\varphi\,,\,u_n + \frac{n^{1/2}}{b_n}v\right\rangle\right]\right\}\,,\\ f(v) &= \exp\left\{\frac{1}{2}D^2\Phi(x^*)(v\,,\,v)\right\}\,. \end{split}$$

Then $\{\mathcal{L}(f_n(T_n/n^{1/2}))\}\$ converges weakly to $\gamma \circ f^{-1}$.

Proof. We first show

(i) $\lim_{n\to\infty} (b_n/n^{1/2})u_n = 0$.

By Lemma 3.2(i), $x^* = \int x \langle \varphi, x \rangle \mu(dx)$. Therefore

$$E(Y_{n1}) - \frac{b_n}{n} x^* = \frac{\int x e^{b_n \langle \varphi, x \rangle / n} \mu(dx) - \frac{b_n}{n} \int x \langle \varphi, x \rangle \mu(dx)}{\hat{\mu}(\frac{b_n}{n} \varphi)} + \frac{\frac{b_n}{n} \int x \langle \varphi, x \rangle \mu(dx) (1 - \hat{\mu}(\frac{b_n}{n} \varphi))}{\hat{\mu}(\frac{b_n}{n} \varphi)}.$$

Now by (3.1) and the inequality $|e^y - 1 - y| \le \frac{1}{2}|y|^2e^{|y|}$ $(y \in \mathbb{R})$, we have

$$\begin{split} \left\| \int x e^{b_n \langle \varphi, x \rangle / n} \mu(dx) - \frac{b_n}{n} \int x \langle \varphi, x \rangle \mu(dx) \right\| \\ &= \left\| \int x \left(e^{b_n \langle \varphi, x \rangle / n} - 1 - \frac{b_n}{n} \langle \varphi, x \rangle \right) \mu(dx) \right\| \\ &\leq \frac{1}{2} \frac{b_n^2}{n^2} \int \|x\| \langle \varphi, x \rangle^2 e^{b_n \langle \varphi, x \rangle / n} \mu(dx) \,. \end{split}$$

Similarly,

$$\left|1-\hat{\mu}\left(\frac{b_n}{n}\varphi\right)\right| \leq \frac{1}{2}\frac{b_n^2}{n^2}\int \langle \varphi, x\rangle^2 e^{b_n\langle \varphi, x\rangle/n}\mu(dx).$$

Therefore

$$\left\| \frac{b_n}{n^{1/2}} u_n \right\| = \left\| n^{1/2} \left(E(Y_{n1}) - \frac{b_n}{n} x^* \right) \right\| \le n^{1/2} \frac{b_n^2}{n^2} r_n,$$

with $\{r_n\}$ bounded. Since $n^{1/2}b_n^2/n^2=(b_n/n^{3/4})^2$, claim (i) follows by assumption (1.5).

We prove next

(ii) For every r > 0, $\lim_{n \to \infty} \sup_{\|v\| \le r} |f_n(v) - f(v)| = 0$. In fact, by Taylor's formula,

$$\begin{aligned} |\log f_n(v) - \log f(v)| \\ &= \frac{1}{2} \left| D^2 \Phi \left(x^* + \theta \left(u_n + \frac{n^{1/2}}{b_n} v \right) \right) \left(\frac{b_n}{n^{1/2}} u_n + v , \frac{b_n}{n^{1/2}} u_n + v \right) \\ &- D^2 \Phi(x^*)(v, v) \right| \\ &\leq \frac{1}{2} \left\| D^2 \Phi \left(x^* + \theta \left(u_n + \frac{n^{1/2}}{b_n} v \right) \right) - D^2 \Phi(x^*) \right\| \left\| \frac{b_n}{n^{1/2}} u_n + v \right\|^2 \\ &+ \frac{1}{2} \left| D^2 \Phi(x^*) \left(\frac{b_n}{n^{1/2}} u_n + v , \frac{b_n}{n^{1/2}} u_n + v \right) - D^2 \Phi(x^*)(v, v) \right| , \end{aligned}$$

where $|\theta| \le 1$, and claim (ii) follows by assumption (3.4) and claim (i). By [9, p. 34] and claim (ii), the proof will be completed if we show that

(3.11)
$$\{ \mathcal{L}(T_n/n^{1/2}) \}$$
 converges weakly to γ .

By assumption (3.3), $\{\mathcal{L}(T_n/n^{1/2})\}$ is tight. Therefore it is enough to show that $\{\mathcal{L}(\langle \xi, T_n/n^{1/2} \rangle)\}$ converges weakly to $\gamma \circ \xi^{-1}$ for $\xi \in E^*$. But it is easily shown that

$$E\langle \xi, T_n/n^{1/2} \rangle^2 \to \int \xi^2 \, d\gamma,$$

$$\lim_{n \to \infty} E(\langle \xi, Y_{n1} - E(Y_{n1}) \rangle^2 I_{\{|\langle \xi, Y_{n1} - E(Y_{n1}) \rangle| > n^{1/2} \varepsilon\}}) = 0$$

for every $\varepsilon > 0$, so (3.11) follows by Lindeberg's theorem (for triangular arrays). \Box

Lemma 3.5. Let $\{U_{nj}: j=1,\ldots,n; n\in \mathbb{N}\}$ be a triangular array of E-valued random vectors such that each row is independent and identically distributed. Let $V_n = \sum_{j=1}^n U_{nj}$, and let q be a continuous seminorm on E. Assume

- (1) For all n, $E(U_{n1}) = 0$,
- (2) For all t > 0, $\sup_{n} E \exp\{tq(U_{n1})\} < \infty$,
- (3) $\sup_{n} Eq(V_n/n^{1/2}) < \infty$.

Let $a = \inf_{h \in \mathbb{N}} \limsup_{n \to \infty} (Eq^2(h^{-1/2} \sum_{j=1}^h U_{nj}))^{1/2}$. Then for all $\rho < (8a^2)^{-1}$ and sufficiently small $\delta > 0$,

$$\sup_{n} E(\exp{\{\rho q^{2}(V_{n}/n^{1/2})\}}I_{\{q(V_{n})<\delta n\}}) < \infty.$$

Proof. Let b > a. Choose $h, n_0 \in \mathbb{N}$ such that for $n \ge n_0$,

(3.12)
$$Eq^2 \left(h^{-1/2} \sum_{j=1}^h U_{nj} \right) < b^2.$$

Let $j_n = [\frac{n}{h}]$ and for $n \ge h$, $i = 1, ..., j_n$, define

$$Z_{ni} = h^{-1/2} \sum_{j=(i-1)h+1}^{ih} U_{nj}, \qquad W_{j_n} = \sum_{i=1}^{j_n} Z_{ni}.$$

The triangular array $\{Z_{ni}: n \ge h, i = 1, ..., j_n\}$ has independent, identically distributed rows. Next, for all $n \ge h$, $m \in \mathbb{N}$, m > 2,

(3.13)
$$Eq^{m}(Z_{n1}) = \int_{0}^{\infty} mt^{m-1}P\{q(Z_{n1}) > t\} dt \\ \leq \int_{0}^{\infty} mt^{m-1}e^{-t}E\exp\{q(Z_{n1})\} dt \\ \leq m!L,$$

where $L = \sup_{n \ge h} E \exp\{q(Z_{n1})\} < \infty$ by assumption (2). By (3.12) and (3.13), we have for all $n \ge \max\{h, n_0\}$, $m \ge 2$,

(3.14)
$$Eq^{m}(Z_{n1}) \le \left(\frac{m!}{2}\right) b^{2} H^{m-2}$$

where $H = \max\{2b^{-2}L, 1\}$.

By (3.14), applying to the *n*th row of $\{Z_{ni}\}$ Theorem 2.1 of Yurinskii [21], with $b_j^2 = b^2$ (so $B_n^2 = j_n b^2$), $\beta_n = Eq(W_{j_n})$, we have: for s > 0, setting $c_n = Eq(W_{j_n}/j_n^{1/2})$,

$$P\{q(W_{j_n}) > s j_n^{1/2}\} = P\{q(W_{j_n}) > \left(\frac{s}{b}\right) b j_n^{1/2}\}$$

$$\leq \exp\left\{-\frac{1}{8b^2} (s - c_n)^2 \left(1 + \frac{(s - c_n)H}{2b^2 j_n^{1/2}}\right)^{-1}\right\}.$$

Now for $0 < \lambda < 1$, *n* sufficiently large, t > 0

$$P\left\{q\left(\frac{V_{n}}{n^{1/2}}\right) > t\right\} \leq P\left\{q(h^{1/2}W_{j_{n}}) > \lambda t_{n}^{1/2}\right\}$$

$$+ P\left\{q(V_{n} - h^{1/2}W_{j_{n}}) > (1 - \lambda)tn^{1/2}\right\}$$

$$\leq P\left\{q(W_{j_{n}}) > \lambda t j_{n}^{1/2}\right\}$$

$$+ \sup_{l \leq h} P\left\{q\left(\sum_{j=1}^{l} U_{nj}\right) > (1 - \lambda)tn^{1/2}\right\}.$$

For $\tau > 0$, $l \le h$, n sufficiently large, we have

$$(3.17) P\left\{q\left(\sum_{j=1}^{l} U_{nj}\right) > (1-\lambda)tn^{1/2}\right\} \leq \exp\{-\tau tn^{1/2}\}M,$$

where $M = \sup_n (E \exp\{\tau(1-\lambda)^{-1}q(U_{n1})\})^h < \infty$ by assumption (2). By (3.15)–(3.17) we have for t > 0, $0 < \lambda < 1$ and n sufficiently large,

$$(3.18) P\left\{q\left(\frac{V_n}{n^{1/2}}\right) > t\right\} \le \exp\left\{-\frac{1}{8b^2}(\lambda t - c_n)^2 \left(1 + \frac{(\lambda t - c_n)H}{2b^2 j_n^{1/2}}\right)^{-1}\right\} + M \exp\left\{-\tau t n^{1/2}\right\}.$$

Using assumption (1), it is easily seen that

(3.19)
$$\limsup_{n \to \infty} c_n \le \limsup_{n \to \infty} Eq\left(\frac{V_n}{n^{1/2}}\right) = c, \quad \text{say.}$$

For $\frac{2}{3} < \lambda < 1$ and sufficiently large n, we have by (3.18) and (3.19):

$$E\left(\exp\left\{\rho q^{2}\left(\frac{V_{n}}{n^{1/2}}\right)\right\}I_{\{q(V_{n})<\delta n\}}\right)$$

$$=1+\int_{0}^{\delta n^{1/2}}2\rho t e^{\rho t^{2}}P\left\{q\left(\frac{V_{n}}{n^{1/2}}\right)>t\right\}dt$$

$$\leq 1+\int_{0}^{3c}\cdots+\int_{3c}^{\delta n^{1/2}}2\rho t e^{\rho t^{2}}$$

$$\times\exp\left\{-\frac{1}{8b^{2}}(\lambda t-2c)^{2}\left(1+\frac{\delta \lambda H(2h)^{1/2}}{2b^{2}}\right)^{-1}\right\}dt$$

$$+\int_{3c}^{\delta n^{1/2}}2\rho t e^{\rho \delta n^{1/2}t}M\exp\{-\tau t n^{1/2}\}dt.$$

It is now clear that if $\rho < (8a^2)^{-1}$, by taking b sufficiently close to a, λ sufficiently close to 1, δ sufficiently small and τ sufficiently large, the statement follows. \Box

Lemma 3.6. Let $\{U_{nj}: j=1,\ldots,n\in\mathbb{N}\}$ be a triangular array of E-valued random vectors such that each row is independent and identically distributed, and let $V_n = \sum_{j=1}^n U_{nj}$. Assume

- (1) For all n, $E(U_{n1}) = 0$,
- (2) For every bounded finite dimensional set $A \subset E^*$,

$$\sup_{n} \sup_{\xi \in A} E \exp\{|\langle \xi, U_{n1} \rangle|\} < \infty.$$

(3) $\{\mathcal{L}(U_{n1})\}$ converges weakly to μ . Let $\{\xi_j: j \in \mathbf{N}\}$ be such that $\int \xi_i \xi_j d\mu = \delta_{ij}$. Then for every $\eta < 1$ and every $k \in \mathbf{N}$, there exists $\delta > 0$ such that

$$\sup_n E\left(\exp\left\{\frac{1}{2}\eta^2\sum_{j=1}^k\left\langle\xi_j\,,\,\frac{V_n}{n^{1/2}}\right\rangle^2\right\}I_{\{\|V_n\|<\delta n\}}\right)<\infty\,.$$

Proof. Fix $\eta < 1$ and $k \in \mathbb{N}$. Since $|\langle \xi_j, V_n \rangle| \le \|\xi_j\| \|V_n\|$ for $j = 1, \ldots, k$, it is enough to prove: for sufficiently small $\delta > 0$,

$$(3.20) \qquad \sup_{n} E\left(\exp\left\{\frac{1}{2}\eta^{2}\sum_{j=1}^{k}\left\langle \xi_{j}, \frac{V_{n}}{n^{1/2}}\right\rangle^{2}\right\} I_{\{|\langle \xi_{j}, V_{n}\rangle| < \delta n, j=1, \dots, k\}}\right) < \infty.$$

Let λ be the canonical Gaussian measure in \mathbb{R}^k . Then (3.21)

$$\begin{split} E\left(\exp\left\{\frac{1}{2}\eta^{2}\sum_{j=1}^{k}\left\langle\xi_{j},\frac{V_{n}}{n^{1/2}}\right\rangle^{2}\right\}I_{\{|\langle\xi_{j},V_{n}\rangle|<\delta n,j=1,\ldots,k\}}\right) \\ &=E\left(\left[\int\exp\left\{\eta\sum_{j=1}^{k}z_{j}\left\langle\xi_{j},\frac{V_{n}}{n^{1/2}}\right\rangle\right\}\lambda(dz)\right]I_{\{|\langle\xi_{j},V_{n}\rangle|<\delta n,j=1,\ldots,k\}}\right) \\ &=\int\lambda(dz)E\left(\exp\left\{\eta\sum_{j=1}^{k}z_{j}\left\langle\xi_{j},\frac{V_{n}}{n^{1/2}}\right\rangle\right\}I_{\{|\langle\xi_{j},V_{n}\rangle|<\delta n,j=1,\ldots,k\}}\right), \end{split}$$

where $z = (z_1, \ldots, z_k) \in \mathbf{R}^k$.

For i = 1, ..., n, $z \in \mathbf{R}^k$, let $Y_{ni}^{(z)} = \sum_{j=1}^k z_j \langle \xi_j, U_{ni} \rangle$, and let $W_n^{(z)} = \sum_{i=1}^n Y_{ni}^{(z)}$. Since, setting $|z|^2 = \sum_{j=1}^k z_j^2$, we have

$$\{|\langle \xi_j, V_n \rangle| < \delta n, j = 1, \ldots, k\} \subset \{|W_n^{(z)}| < \delta |z| k^{1/2} n\},$$

by (3.21) in order to prove (3.20) it is enough to show: for sufficiently small $\delta > 0$,

(3.22)
$$\sup_{n} \int \lambda(dz) E\left(\exp\left\{\eta \frac{W_{n}^{(z)}}{n^{1/2}}\right\} I_{\{|W_{n}^{(z)}| < \delta|z|n\}}\right) < \infty.$$

In order to prove (3.22) we estimate $P\{|W_n^{(z)}/n^{1/2}| > t\}$ by means of Bernstein's inequality (see [21, p. 474], or [19, p. 55]). We have $E(Y_{n1}^{(z)}) = 0$ for $n \in \mathbb{N}$ by assumption (1). Also

$$E(Y_{n1}^{(z)})^{2} - \sum_{j=1}^{k} z_{j}^{2} = \sum_{j} \sum_{l} z_{j} z_{l} [E(\langle \xi_{j}, U_{n1} \rangle \langle \xi_{l}, U_{n1} \rangle) - \int \xi_{j} \xi_{l} d\mu]$$

$$\leq \sup_{j,l} \left| E(\langle \xi_{j}, U_{n1} \rangle \langle \xi_{l}, U_{n1} \rangle) - \int \xi_{j} \xi_{l} d\mu \right| \left(\sum_{j} |z_{j}| \right)^{2}$$

$$\leq \sup_{j,l} \left| E(\langle \xi_{j}, U_{n1} \rangle \langle \xi_{l}, U_{n1} \rangle) - \int \xi_{j} \xi_{l} d\mu \right| k \left(\sum_{j} z_{j}^{2} \right)$$

and therefore by assumptions (2) and (3), given $\zeta > 1$ (to be further specified later), there exists n_0 such that for $n \ge n_0$,

$$(3.23) E(Y_{n_1}^{(z)})^2 \le \zeta |z|^2.$$

Next.

$$\begin{split} P\{|Y_{n1}^{(z)}| > t\} &\leq \exp\left\{-\frac{t}{|z|}\right\} E \exp\left\{\frac{1}{|z|}|Y_{n1}^{(z)}|\right\} \\ &\leq a \exp\left\{-\frac{t}{|z|}\right\}, \end{split}$$

where $a = \sup_n \sup_{|w| \le 1} E \exp\{|Y_{n1}^{(w)}|\} < \infty$ by assumption (2). Therefore, for m > 2, $n \in \mathbb{N}$,

(3.24)
$$E|Y_{n1}^{(z)}|^{m} = \int_{0}^{\infty} mt^{m-1}P\{|Y_{n1}^{(z)}| > t\} dt$$

$$\leq \int_{0}^{\infty} mt^{m-1}a \exp\left\{-\frac{t}{|z|}\right\} dt$$

$$= am!|z|^{m} \leq \left(\frac{m!}{2}\right) \zeta|z|^{2} (2a\zeta^{-1}|z|)^{m-2},$$

since we may assume $2a\zeta^{-1} \ge 1$. By (3.23), (3.24) and Bernstein's inequality, for $n \ge n_0$ and t > 0,

$$(3.25) P\left\{\left|\frac{W_n^{(z)}}{n^{1/2}}\right| > t\right\} \le 2 \exp\left\{-\frac{t^2}{2\zeta|z|^2} \left(1 + \frac{2at}{n^{1/2}\zeta^2|z|}\right)^{-1}\right\}.$$

Now by (3.25), for sufficiently large n,

$$\begin{split} E\left(\exp\left\{\eta\left|\frac{W_{n}^{(z)}}{n^{1/2}}\right|\right\} I_{\{|W_{n}^{(z)}/n^{1/2}|<\delta|z|n^{1/2}\}}\right) \\ &= 1 + \int_{0}^{\delta|z|n^{1/2}} \eta \exp(\eta t) P\left\{\left|\frac{W_{n}^{(z)}}{n^{1/2}}\right| > t\right\} dt \\ &\leq 1 + \int_{0}^{\delta|z|n^{1/2}} \eta \exp(\eta t) \cdot 2 \exp\left\{-\frac{t^{2}}{2\zeta|z|^{2}} \left(1 + \frac{2at}{n^{1/2}\zeta^{2}|z|}\right)^{-1}\right\} dt \\ &\leq 1 + 2 \int_{0}^{\infty} \eta \exp(\eta t) \exp\left\{-\frac{t^{2}}{2\zeta|z|^{2}} (1 + 2a\zeta^{-2}\delta)^{-1}\right\} dt \\ &\leq 1 + 2\sqrt{2\pi} \eta \alpha |z| \exp\left\{\frac{1}{2} (\eta \alpha)^{2} |z|^{2}\right\} \end{split}$$

by an elementary computation, where $\alpha = \zeta^{1/2}(1+2a\zeta^{-2}\delta)^{1/2}$. Choosing $\zeta > 1$ and $\delta > 0$ so that $\eta \alpha < 1$, the result follows from (3.22). \square

Proof of Theorem 3.1. (i) The fact that $C < \infty$ follows from the integrability arguments in the proof of (ii). Nevertheless, we will give a simple direct proof because it illustrates the use of the decomposition given by lemma 3.3(i), which is later employed in a more complicated situation in the proof of (ii).

Let $\{\xi_j\colon j\in \mathbf{N}\}\subset E^*$ be such that $\{\Delta(\xi_j)\colon j\in \mathbf{N}\}$ is an orthonormal basis of $H_\gamma=H_\mu$ (see [12]). Let $A=D^2\Phi(x^*)$. By (3.7) and Lemma 3.3(ii), $\alpha=\sup_{\|x\|_\gamma\leq 1}A(x\,,\,x)<1$. Choose $\beta>0$ such that $\rho=\alpha+\beta^{-2}\,d\sigma^2<1$. Since $Q_k(Z)\to 0$ a.s. for a random vector Z with $\mathscr{L}(Z)=\gamma$ (see [12]), one may choose $k\in \mathbf{N}$ such that

$$(3.26) \qquad \int \exp\left(\frac{1}{2}d(1+\beta^2)\|Q_k(x)\|^2\right)\gamma(dx) < \infty.$$

By Lemma 3.3(i)

$$\begin{split} &\int \exp\left\{\frac{1}{2}A(y\,,\,y)\right\}\gamma(dy) \\ &\leq \int \exp\left\{\frac{1}{2}\,d(1+\beta^2)\|Q_k(x)\|^2\right\}\gamma(dx) \\ &\quad \times \int \exp\left\{\frac{1}{2}\rho\sum_{j=1}^k\langle\xi_j\,,\,x\rangle^2\right\}\gamma(dx) < \infty \end{split}$$

since the second integral on the right-hand side is

$$\int_{\mathbf{P}^k} \exp\left\{\frac{1}{2}\rho|z|^2\right\} \lambda(dz) < \infty$$

because $\rho < 1$, where $|\cdot|$ is the usual Euclidean norm in \mathbf{R}^k and λ is the canonical Gaussian measure in \mathbf{R}^k .

(ii) Let x^* be as in (3.6). For the rest of the proof, we set $\varphi = D\Phi(x^*)$. Since for any $\xi \in E^*$, $\sup_{\eta \in E^*} [\langle \eta, \Delta(\xi) \rangle - \frac{1}{2} \int \eta^2 d\mu]$ is attained at $\eta = \xi$, as

is easily proved, we have by Lemma 3.2(i)

$$I(x^*) = \langle \varphi, x^* \rangle - \frac{1}{2} \int \varphi^2 d\mu.$$

Therefore we may write

$$C_{n} = \exp\left\{-\frac{b_{n}^{2}}{n}\left[\Phi(x^{*}) - I(x^{*})\right]\right\} E \exp\left\{\frac{b_{n}^{2}}{n}\Phi\left(\frac{S_{n}}{b_{n}}\right)\right\}$$

$$= \exp\left\{-\frac{b_{n}^{2}}{n}\frac{1}{2}\int\varphi^{2}d\mu\right\} E \left(\exp\left\{\frac{b_{n}^{2}}{n}\left[\Phi\left(x^{*} + \left(\frac{S_{n}}{b_{n}} - x^{*}\right)\right)\right]\right)$$

$$-\Phi(x^{*}) - \left\langle\varphi, \frac{S_{n}}{b_{n}} - x^{*}\right\rangle\right]\right\} \exp\left\{\frac{b_{n}}{n}\langle\varphi, S_{n}\rangle\right\}$$

$$= \left(\hat{\mu}\left(\frac{b_{n}}{n}\varphi\right)\right)^{n} \exp\left\{-\frac{b_{n}^{2}}{n}\frac{1}{2}\int\varphi^{2}d\mu\right\} E \exp\left\{\frac{b_{n}^{2}}{n}\left[\Phi\left(x^{*} + u_{n} + \frac{T_{n}}{b_{n}}\right)\right]\right\}$$

$$-\Phi(x^{*}) - \left\langle\varphi, u_{n} + \frac{T_{n}}{b_{n}}\right\rangle\right]\right\},$$

where $T_n = \sum_{j=1}^n (Y_{nj} - E(Y_{nj}))$, with $\{Y_{nj}: j = 1, ..., n\}$ independent and

$$d\mathscr{L}(Y_{nj}) = \frac{e^{b_n \varphi/n}}{\hat{\mu}(\frac{b_n}{n}\varphi)} d\mu$$
, and $u_n = \frac{n}{b_n} E(Y_{n1}) - x^*$.

Now let $F_n(x) = \Phi(x^* + u_n + x) - \Phi(x^*) - \langle \varphi, u_n + x \rangle$. Since

$$\exp\left\{\frac{b_n^2}{n}F_n\left(\frac{T_n}{b_n}\right)\right\} = f_n(T_n/n^{1/2}),\,$$

where f_n is as in Lemma 3.4, in order to prove $\lim_{n\to\infty} C_n = C$ it is enough by Lemma 3.4 to prove

(a)
$$\lim_{n\to\infty} \left(\hat{\mu}\left(\frac{b_n}{n}\varphi\right)\right)^n \exp\left\{-\frac{b_n^2}{n}\frac{1}{2}\int \varphi^2 d\mu\right\} = 1.$$

(b)
$$\left\{ \exp \left\{ \frac{b_n^2}{n} F_n \left(\frac{T_n}{b_n} \right) \right\} \right\}$$
 is uniformly integrable.

We prove (a) first. A straightforward expansion gives

$$\hat{\mu}\left(\frac{b_n}{n}\varphi\right) = 1 + \frac{1}{2}\left(\frac{b_n}{n}\right)^2 \int \varphi^2 d\mu + \frac{1}{6}\left(\frac{b_n}{n}\right)^3 \int \varphi^3 d\mu + O\left(\left(\frac{b_n}{n}\right)^4\right).$$

On the other hand,

$$\exp\left\{-\frac{1}{2}\frac{b_n^2}{n^2}\int\varphi^2\,d\mu\right\} = 1 - \frac{1}{2}\left(\frac{b_n}{n}\right)^2\int\varphi^2\,d\mu + O\left(\left(\frac{b_n}{n}\right)^4\right).$$

Therefore

$$\hat{\mu}\left(\frac{b_n}{n}\varphi\right)\exp\left\{-\frac{1}{2}\left(\frac{b_n}{n}\right)^2\int\varphi^2\,d\mu\right\}=1-\left(\frac{b_n}{n}\right)^3r_n\,,$$

where $\{r_n\}$ is bounded. From assumption (1.5) it follows that

$$\left(\hat{\mu}\left(\frac{b_n}{n}\varphi\right)\right)^n \exp\left\{-\frac{1}{2}\frac{b_n^2}{n}\int \varphi^2 d\mu\right\} = \left(1 - \frac{(b_n/n^{2/3})^3 r_n}{n}\right)^n \to 1.$$

(It is clear from the argument that if $\int \varphi^3 d\mu = 0$, then the conclusion will hold under the weaker assumption $b_n/n^{3/4} \to 0$.)

We turn now to the proof of (b). In order to prove this assertion, it is clearly enough to prove:

(I) For every $\delta > 0$,

$$\limsup_{n\to\infty} E\left(\exp\left\{\frac{b_n^2}{n}F_n\left(\frac{T_n}{b_n}\right)\right\}I_{\{\|T_n/b_n\|\geq\delta\}}\right)=0.$$

(II) There exist $\delta > 0$, p > 1 such that

$$\sup_{n} E\left(\exp\left\{p\frac{b_n^2}{n}F_n(T_n/b_n)\right\}I_{\{\|T_n/b_n\|<\delta\}}\right)<\infty.$$

In order to prove (I), it suffices to show that for any closed set D,

(3.27)
$$\limsup_{n\to\infty} \frac{n}{b_n^2} \log \int_D \exp\left\{\frac{b_n^2}{n} F_n\right\} d\lambda_n \le \sup_{x\in D} [F(x) - I(x)],$$

where $\lambda_n = \mathcal{L}(T_n/b_n)$ and $F(x) = \Phi(x^* + x) - \Phi(x^*) - \langle \varphi, x \rangle$. Indeed, a simple computation using Lemma 3.2(i) gives

(3.28)
$$F(x) - I(x) = [\Phi(x^* + x) - I(x^* + x)] - [\Phi(x^*) - I(x^*)].$$

Now let $D = \{x \in E : ||x|| \ge \delta\}$. Arguing as in the proof of Lemma 3.2(i) one can see that $s = \sup_{x \in D} [F(x) - I(x)]$ is attained, and assumption (3.6) together with (3.28) imply that s < 0. Thus (3.27) implies (I).

To establish (3.27) we first show that

(3.29)
$$\lim_{t\to\infty} \limsup_{n\to\infty} \frac{n}{b_n^2} \log \int_{\{F_n\geq t\}} \exp\left\{\frac{b_n^2}{n}F_n\right\} d\lambda_n = -\infty.$$

By assumption (3.5) and Lemma 3.4, there exist α , $\beta > 0$ such that $\sup_n F_n(x) \le \alpha ||x|| + \beta$ for all $x \in E$. It follows that

$$\int_{\{F_n \ge t\}} \exp\{b_n^2 F_n\} d\lambda_n \le \exp\left\{-\frac{b_n^2}{n}t\right\} \int \exp\left\{2\frac{b_n^2}{n} F_n\right\} d\lambda_n$$

$$\le \exp\left\{-\frac{b_n^2}{n}t\right\} E \exp\left\{2\frac{b_n^2}{n} \left(\alpha \left\|\frac{T_n}{b_n}\right\| + \beta\right)\right\}.$$

By the argument in Lemma 2.1(i) (with q_K replaced by $\|\cdot\|$), the second factor is bounded by $\exp\{b_n^2d/n\}$ for some constant d>0 and sufficiently large n. Therefore

$$\limsup_{n\to\infty}\frac{n}{b_n^2}\log\int_{\{F_n\geq t\}}\exp\left\{\frac{b_n^2}{n}F_n\right\}\,d\lambda_n\leq -(t-d)\,,$$

which yields (3.29). We proceed now to complete the proof of (3.27). Let $a_n = b_n^2/n$ and fix t > 0. By the argument of the proof of Theorem 2.2(i), given $\tau > 0$ there exist a compact set K_{τ} and $n_0 \in \mathbb{N}$ such that for $n \ge n_0$,

$$P\{T_n/b_n\in K_{\tau}^c\}\leq e^{-a_n\tau}.$$

By Lemma 3.4, $F_n \to F$ uniformly over compact sets. Therefore given $\delta > 0$, we have for sufficiently large n

$$\int_{D} \exp\{a_n \min(F_n, t)\} d\lambda_n$$

$$\leq \int_{D \cap K_{\tau}} \exp\{a_n \min(F_n, t)\} d\lambda_n + \int_{K_{\tau}^c} \exp\{a_n t\} d\lambda_n$$

$$\leq \int_{D \cap K_{\tau}} \exp\{a_n \min(F + \delta, t)\} d\lambda_n + e^{-a_n(\tau - t)}.$$

By [15, p. 321], taking into account Theorem 2.2(i)

$$\limsup_{n \to \infty} a_n^{-1} \log \int_{D \cap K_{\tau}} \exp\{a_n \min(F + \delta, t)\} d\lambda_n$$

$$\leq \sup_{x \in D \cap K_{\tau}} [\min(F(x) + \delta, t) - I(x)]$$

$$\leq \delta + \sup_{x \in D} [F(x) - I(x)]$$

and it follows that

$$\limsup_{n\to\infty} a_n^{-1} \log \int_D \exp\{a_n \min(F_n, t)\} d\lambda_n \le \sup_{x\in D} [F(x) - I(x)].$$

Now a standard argument using (3.29) completes the proof of (3.27), and hence that of (I).

We turn now to the proof of (II). By Taylor's formula,

$$F_n\left(\frac{T_n}{b_n}\right) = \frac{1}{2}D^2\Phi\left(x^* + \theta\left(u_n + \frac{T_n}{b_n}\right)\right)\left(u_n + \frac{T_n}{b_n}, u_n + \frac{T_n}{b_n}\right),$$

where $|\theta| < 1$, and therefore

$$\frac{b_n^2}{n}F_n\left(\frac{T_n}{b_n}\right) = \frac{1}{2}D^2\Phi\left(x^* + \theta\left(u_n + \frac{T_n}{b_n}\right)\right)\left(\frac{b_n}{n^{1/2}}u_n + \frac{T_n}{n^{1/2}}, \frac{b_n}{n^{1/2}}u_n + \frac{T_n}{n^{1/2}}\right).$$

By the assumptions on Φ and Lemma 3.4, simple estimates show that given $\varepsilon > 0$ (to be further specified later), it is possible to choose $\delta_1 > 0$ and $n_1 \in \mathbb{N}$ such that for $n \ge n_1$,

$$(3.30) \quad \frac{b_n^2}{n} F_n\left(\frac{T_n}{b_n}\right) \le \varepsilon + \varepsilon \left\|\frac{T_n}{n^{1/2}}\right\|^2 + \frac{1}{2} A\left(\frac{T_n}{n^{1/2}}, \frac{T_n}{n^{1/2}}\right) \quad \text{on } \left\{\left\|\frac{T_n}{b_n}\right\| < \delta_1\right\},$$

where $A=D^2\Phi(x^*)$. Let p>1, r>1 (to be further specified later), and define s>1 by $r^{-1}+s^{-1}=1$. Let $\{\xi_j\colon j\in \mathbf{N}\}\subset E^*$ be such that $\{\Delta(\xi_j)\colon j\in \mathbf{N}\}$ is an orthonormal basis of H_μ . By Lemma 3.3 and Hölder's inequality, for any $\delta\leq\delta_1$, $n\geq n_1$ and $k\in \mathbf{N}$, we have

$$(3.31) E\left(\exp\left\{p\left(\varepsilon\left\|\frac{T_{n}}{n^{1/2}}\right\|^{2} + \frac{1}{2}A\left(\frac{T_{n}}{n^{1/2}}, \frac{T_{n}}{n^{1/2}}\right)\right)\right\} I_{\{\|T_{n}/b_{n}\|<\delta\}}\right)$$

$$\leq \left(E\left(\exp\left\{sp\left(\varepsilon\left\|\frac{T_{n}}{n^{1/2}}\right\|^{2} + \frac{d}{2}(1+\beta^{2})\left\|Q_{k}\left(\frac{T_{n}}{n^{1/2}}\right)\right\|^{2}\right)\right\} I_{\{\|T_{n}/b_{n}\|<\delta\}}\right)\right)^{1/s}$$

$$\cdot \left(E\left(\exp\left\{r\frac{p}{2}(\alpha+\beta^{-2}d\sigma^{2})\sum_{j=1}^{k}\left\langle\xi_{j}, \frac{T_{n}}{n^{1/2}}\right\rangle^{2}\right\} I_{\{\|T_{n}/b_{n}\|<\delta\}}\right)\right)^{1/r}.$$

Since $\alpha < 1$, we may choose and fix $\beta > 0$, p > 1, r > 1 in such a way that $rp(\alpha + \beta^{-2} d\sigma^2) < 1$. For $x \in E$, let

$$q_k(x) = \varepsilon^{1/2} ||x|| + (\frac{d}{2}(1+\beta^2))^{1/2} ||Q_k(x)||,$$

and let c_k be such that $q_k \le c_k \| \cdot \|$. Then the sth power of the first factor in the right-hand side of (3.31) is bounded by

$$E(\exp\{spq_k^2(T_n/n^{1/2})\}I_{\{q_k(T_n)<\delta c_k n\}})$$

for sufficiently large n, since $b_n/n \to 0$. By [12], one may choose $k \in \mathbb{N}$, $\varepsilon > 0$ such that

 $\int q_k^2 \, d\gamma < (8sp)^{-1} \, .$

Since $\{\mu^{*n}(n^{1/2}(\cdot))\}$ converges weakly to γ , by the results of [4] one may choose $h \in \mathbb{N}$ such that

$$\int q_k^2 d\mu^{*h}(h^{1/2}(\cdot)) < (8sp)^{-1},$$

and since $\{\mathcal{L}(Y_{n1} - EY_{n1})\}\$ converges weakly to μ , we have by [4]

(3.32)
$$\limsup_{n\to\infty} Eq_k^2 \left(h^{-1/2} \sum_{j=1}^h (Y_{nj} - EY_{nj}) \right) < (8sp)^{-1}.$$

The triangular array $\{Y_{nj} - EY_{nj} : n \in \mathbb{N}, j = 1, ..., n\}$ and the seminorm q_k satisfy the assumptions of Lemma 3.5 (assumption (3) follows from Lemma 3.4 and [4]). Therefore for sufficiently small $\delta > 0$,

(3.33)
$$\sup_{n} E(\exp\{spq_{k}^{2}(T_{n}/n^{1/2})\}I_{\{q_{k}(T_{n})<\delta c_{k}n\}}) < \infty.$$

Finally, by Lemma 3.6, it is possible to choose $\delta > 0$ such that the second factor in the right-hand side of (3.31) is bounded uniformly in n. Now assertion (II) follows from (3.30), (3.31) and (3.33). \square

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