

# MODERATE DEVIATIONS FOR MARTINGALES WITH BOUNDED JUMPS <sup>1</sup>

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*Abstract*

*We prove that the Moderate Deviation Principle (MDP) holds for the trajectory of a locally square integrable martingale with bounded jumps as soon as its quadratic covariation, properly scaled, converges in probability at an exponential rate. A consequence of this MDP is the tightness of the method of bounded martingale differences in the regime of moderate deviations.*

## 1 Introduction

Suppose  $\{X_m, \mathcal{F}_m\}_{m=0}^\infty$  is a discrete-parameter real valued martingale with bounded jumps  $|X_m - X_{m-1}| \leq a$ ,  $m \in \mathbb{N}$ , filtration  $\mathcal{F}_m$  and such that  $X_0 = 0$ . The basic inequality for the method of bounded martingale differences is Azuma-Hoeffding inequality (c.f. [1]):

$$\mathbb{P}\{X_k \geq x\} \leq e^{-x^2/2ka^2} \quad \forall x > 0. \quad (1)$$

In the special case of i.i.d. differences  $\mathbb{P}\{X_m - X_{m-1} = a\} = 1 - \mathbb{P}\{X_m - X_{m-1} = -\epsilon a/(1 - \epsilon)\} = \epsilon \in (0, 1)$ , it is easy to see that  $\mathbb{P}\{X_k \geq x\} \leq \exp[-kH(\epsilon + (1 - \epsilon)x/(ak)|\epsilon)]$ , where  $H(q|p) = q \log(q/p) + (1 - q) \log((1 - q)/(1 - p))$ . For  $\epsilon \rightarrow 0$ , the latter upper bound approaches 0, thus demonstrating that (1) may in general be a non-tight upper bound. Let  $B(u) = 2u^{-2}((1 + u) \log(1 + u) - u)$  and

$$\langle X \rangle_m = \sum_{k=1}^m E[(X_k - X_{k-1})^2 | \mathcal{F}_{k-1}]$$

denote the quadratic variation of  $\{X_m, \mathcal{F}_m\}_{m=0}^\infty$ . Then,

$$\mathbb{P}\{X_k \geq x\} \leq \mathbb{P}\{\langle X \rangle_k \geq y\} + e^{-x^2 B(ax/y)/2y} \quad \forall x, y > 0 \quad (2)$$

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(c.f. [4, Theorem (1.6)]). In particular,  $B(0_+) = 1$ , recovering (1) for the choice  $y = ka^2$  and  $x/y \rightarrow 0$ . The inequality (2) holds also for the more general setting of locally square integrable (continuous-parameter) martingales with bounded jumps (c.f. [7, Theorem II.4.5]).

In this note we adopt the latter setting and demonstrate the tightness of (2) in the range of moderate deviations, corresponding to  $x/y \rightarrow 0$  while  $x^2/y \rightarrow \infty$  (c.f. Remark 5 below). We note in passing that for *continuous* martingales [6] studies the tightness of the inequality:

$$\mathbb{P}\{X_k \geq \frac{1}{2} x(1 + \langle X \rangle_k/y)\} \leq e^{-x^2/2y},$$

using Girsanov transformations, whereas we apply large deviation theory and concentrate on martingales with (bounded) jumps, encompassing the case of discrete-parameter martingales. Recall that a family of random variables  $\{Z_k; k > 0\}$  with values in a topological vector space  $\mathcal{X}$  equipped with  $\sigma$ -field  $\mathcal{B}$  satisfies the Large Deviation Principle (LDP) with speed  $a_k \downarrow 0$  and *good rate function*  $I(\cdot)$  if the level sets  $\{x; I(x) \leq \alpha\}$  are compact for all  $\alpha < \infty$  and for all  $\Gamma \in \mathcal{B}$

$$-\inf_{x \in \Gamma^\circ} I(x) \leq \liminf_{k \rightarrow \infty} a_k \log \mathbb{P}\{Z_k \in \Gamma\} \leq \limsup_{k \rightarrow \infty} a_k \log \mathbb{P}\{Z_k \in \Gamma\} \leq -\inf_{x \in \bar{\Gamma}} I(x)$$

(where  $\Gamma^\circ$  and  $\bar{\Gamma}$  denote the interior and closure of  $\Gamma$ , respectively). The family of random variables  $\{Z_k; k > 0\}$  satisfies the *Moderate Deviation Principle* with good rate function  $I(\cdot)$  and critical speed  $1/h_k$  if for every speed  $a_k \downarrow 0$  such that  $h_k a_k \rightarrow \infty$ , the random variables  $\sqrt{a_k} Z_k$  satisfy the LDP with the good rate function  $I(\cdot)$ .

Let  $D(\mathbb{R}^d) (= D(\mathbb{R}_+, \mathbb{R}^d))$  denote the space of all  $\mathbb{R}^d$ -valued *càdlàg* (i.e. right-continuous with left-hand limits) functions on  $\mathbb{R}_+$  equipped with the locally uniform topology. Also,  $C(\mathbb{R}^d)$  is the subspace of  $D(\mathbb{R}^d)$  consisting of continuous functions.

The process  $X \in D(\mathbb{R}^d)$  is defined on a complete stochastic basis  $(\Omega, \mathcal{F}, \mathbf{F} = \mathcal{F}_t, \mathbb{P})$  (c.f. [5, Chapters I and II] or [7, Chapters 1-4] for this and the related definitions that follow). We equip  $D(\mathbb{R}^d)$  hereafter with a  $\sigma$ -field  $\mathcal{B}$  such that  $X : \Omega \rightarrow D(\mathbb{R}^d)$  is measurable ( $\mathcal{B}$  may well be strictly smaller than the Borel  $\sigma$ -field of  $D(\mathbb{R}^d)$ ).

Suppose that  $X \in \mathcal{M}_{\text{loc},0}^2$  is a locally square integrable martingale with bounded jumps  $|\Delta X| \leq a$  (and  $X_0 = 0$ ). We denote by  $(A, C, \nu)$  the triplet predictable characteristics of  $X$ , where here  $A = 0$ ,  $C = (C_t)_{t \geq 0}$  is the  $\mathbf{F}$ -predictable quadratic variation process of the continuous part of  $X$  and  $\nu = \nu(ds, dx)$  is the  $\mathbf{F}$ -compensator of the measure of jumps of  $X$ . Without loss of generality we may assume that

$$\nu(\{t\}, \mathbb{R}^d) = \int_{|x| \leq a} \nu(\{t\}, dx) \leq 1, \quad \int_{|x| \leq a} x \nu(\{t\}, dx) = 0, \quad t > 0 \quad (3)$$

and for all  $s < t$ ,  $(C_t - C_s)$  is a symmetric positive-semi-definite  $d \times d$  matrix. The predictable quadratic characteristic (covariation) of  $X$  is the process

$$\langle X \rangle_t = C_t + \int_0^t \int_{|x| \leq a} x x' d\nu, \quad (4)$$

where  $x'$  denotes the transpose of  $x \in \mathbb{R}^d$ , and  $\|A\| = \sup_{|\lambda|=1} |\lambda' A \lambda|$  for any  $d \times d$  symmetric matrix  $A$ .

Our main result is as follows.

**Proposition 1** *Suppose the symmetric positive-semi-definite  $d \times d$  matrix  $Q$  and the regularly varying function  $h_t$  of index  $\alpha > 0$  are such that for all  $\delta > 0$ :*

$$\limsup_{t \rightarrow \infty} h_t^{-1} \log \mathbb{P}\{\|h_t^{-1}\langle X \rangle_t - Q\| > \delta\} < 0. \quad (5)$$

*Then  $\{h_k^{-1/2} X_k\}$  satisfies the MDP in  $(D[\mathbb{R}^d], \mathcal{B})$  (equipped with the locally uniform topology) with critical speed  $1/h_k$  and the good rate function*

$$I(\phi) = \begin{cases} \int_0^\infty \Lambda^*(\dot{\phi}(t)) \alpha^{-1} t^{(1-\alpha)} dt & \phi \in \mathcal{AC}_0 \\ \infty & \text{otherwise,} \end{cases} \quad (6)$$

*where  $\Lambda^*(v) = \sup_{\lambda \in \mathbb{R}^d} (\lambda'v - \frac{1}{2} \lambda'Q\lambda)$ , and  $\mathcal{AC}_0 = \{\phi : \mathbb{R}_+ \rightarrow \mathbb{R}^d \text{ with } \phi(0) = 0 \text{ and absolutely continuous coordinates}\}$ .*

**Remark 1** Note that both (5) and the MDP are invariant to replacing  $h_t$  by  $g_t$  such that  $h_t/g_t \rightarrow c \in (0, \infty)$  and taking  $cQ$  instead of  $Q$ . Thus, if  $Q \neq 0$  we may take  $h_t = \text{median } \|\langle X \rangle_t\|$ , and in general we may assume with no loss of generality that  $h_t \in D(\mathbb{R}_+)$  is strictly increasing of bounded jumps.

**Remark 2** If  $X$  is a locally square integrable martingale with independent increments, then  $\langle X \rangle$  is a deterministic process, hence suffices that  $h_t^{-1}\langle X \rangle_t \rightarrow Q$  for (5) to hold.

As stated in the next corollary, less is needed if only  $X_k$  (or  $\sup_{s \leq k} X_s$ ) is of interest.

**Corollary 1**

(a) *Suppose that (5) holds for some unbounded  $h_t$  (possibly not regularly varying). Then,  $\{h_k^{-1/2} X_k\}$  satisfies the MDP in  $\mathbb{R}^d$  with critical speed  $1/h_k$  and good rate function  $\Lambda^*(\cdot)$ .*

(b) *If also  $d = 1$ , then  $\{h_k^{-1/2} \sup_{s \leq k} X_s\}$  satisfies the MDP with the good rate function  $I(z) = z^2/(2Q)$  for  $z \geq 0$  and  $I(z) = \infty$  otherwise.*

**Remark 3** For  $d = 1$ , discrete-time martingales, and assuming that  $h_k = \langle X \rangle_k$  is non-random, strong Normal approximation for the law of  $h_k^{-1/2} X_k$  is proved in [9] for the range of values corresponding to  $a_k^3 h_k \rightarrow \infty$ .

**Remark 4** The difference between Proposition 1 and Corollary 1 is best demonstrated when considering  $X_t = B_{h_t}$ , with  $B_s$  the standard Brownian motion. The MDP for  $h_t^{-1/2} B_{h_t}$  in  $\mathbb{R}$  then trivially holds, whereas the MDP for  $h_k^{-1/2} B_{h_{t_k}}$  is equivalent to Schilder's theorem (c.f. [3, Theorem 5.2.3]), and thus holds only when  $h_t$  is regularly varying of index  $\alpha > 0$ .

**Remark 5** When  $d = 1$  and  $Q \neq 0$ , the rate function for the MDP of part (a) of Corollary 1 is  $x^2/(2Q)$ . For  $y = h_k Q(1 + \delta)$ ,  $\delta > 0$  and  $x = x_k = o(y)$  such that  $x^2/y \rightarrow \infty$ , this MDP then implies that  $\mathbb{P}\{X_k \geq x\} = \exp(-(1 + \delta + o(1))x^2/2y)$  while  $P(\langle X \rangle_k \geq y) = o(\exp(-x^2/2y))$  by (5). Consequently, for such values of  $x, y$  the inequality (2) is tight for  $k \rightarrow \infty$  (see also Remark 9 below for non-asymptotic results).

**Remark 6** In contrast with Corollary 1 we note that the LDP with speed  $m^{-1}$  may fail for  $m^{-1} X_m$  even when  $X$  is a real valued discrete-parameter martingale with bounded independent increments such that  $\langle X \rangle_m = m$ . Specifically, let  $b : \mathbb{N} \rightarrow \{1, 2\}$  be a deterministic sequence such that  $p_m = m^{-1} \sum_{k=1}^m 1_{\{b(k)=1\}}$  fails to converge for  $m \rightarrow \infty$  and let  $\mu_i$ ,  $i = 1, 2$  be two probability measures on  $[-a, a]$  such that  $\int x d\mu_i = 0$ ,  $\int x^2 d\mu_i = 1$ ,  $i = 1, 2$  while  $c_1 \neq c_2$  for

$c_i = \log \int e^x d\mu_i$ . Then,  $\Delta X_k$  independent random variables of law  $\mu_{b(k)}$ ,  $k \in \mathbb{N}$ , result with  $X_m$  as above. Indeed,  $m^{-1} \log \mathbb{E}\{\exp(X_m)\} = p_m c_1 + (1 - p_m) c_2$  fails to converge for  $m \rightarrow \infty$ , hence by Varadhan's lemma (c.f. [3, Theorem 4.3.1]), necessarily the LDP with speed  $m^{-1}$  fails for  $m^{-1} X_m$ .

**Remark 7** Corollary 1 may fail when  $X$  is a real valued discrete-parameter martingale with *unbounded* independent increments such that  $\langle X \rangle_m = m$ . Specifically, for  $m_j = 2^{2j^2}$ ,  $j \in \mathbb{N}$  let  $M(m_j) = 2(m_j \log m_j)^{1/2}$  and  $M(k) = 1$  for all other  $k \in \mathbb{N}$ . Let  $Z_k$  be independent Bernoulli( $1/(M(k)^2 + 1)$ ) random variables. Then,  $\Delta X_k = M(k)Z_k - M(k)^{-1}(1 - Z_k)$  result with  $X_m$  as above, with the LDP of speed  $1/\log m$  not holding for  $(m \log m)^{-1/2} X_m$ . Indeed, let  $Y_m$  be the martingale with  $\Delta Y_{m_j}$  i.i.d. and independent of  $X$  such that  $\mathbb{P}\{\Delta Y_{m_j} = 1\} = \mathbb{P}\{\Delta Y_{m_j} = -1\} = 0.5$  and  $\Delta Y_k = \Delta X_k$  for all other  $k \in \mathbb{N}$ . Then,  $(m \log m)^{-1/2} |X_m - Y_m| \rightarrow 0$  for  $m = (m_j - 1)$ ,  $j \rightarrow \infty$ , while  $(m \log m)^{-1/2} (X_m - Y_m) \geq 2Z_m + o(1)$  for  $m = m_j$ ,  $j \rightarrow \infty$ . The LDP with speed  $1/\log m$  and good rate function  $x^2/2$  holds for  $(m \log m)^{-1/2} Y_m$  (c.f. Corollary 1), while  $\log \mathbb{P}\{Z_{m_j} = 1\}/\log m_j \rightarrow -1$  as  $j \rightarrow \infty$ . Consequently, the LDP bounds fail for  $\{(m \log m)^{-1/2} X_m \geq 2\}$ .

Proposition 1 is proved in the next section with the proof of Corollary 1 provided in Section 3. Both results build upon Lemma 1. Indeed, Proposition 1 is a direct consequence of Lemma 1 and [8]. Also, with Lemma 1 holding, it is not hard to prove part (a) of Corollary 1 as a consequence of the Gärtner–Ellis theorem (c.f. [3, Theorem 2.3.6]), without relying on [8].

## 2 Proof of Proposition 1

The cumulant  $G(\lambda) = (G_t(\lambda))_{t \geq 0}$  associated with  $X$  is

$$G_t(\lambda) = \frac{1}{2} \lambda' C_t \lambda + \int_0^t \int_{|x| \leq a} (e^{\lambda' x} - 1 - \lambda' x) \nu(ds, dx), t > 0, \lambda \in \mathbb{R}^d. \quad (7)$$

The stochastic (or the Doléans-Dade) exponential of  $G(\lambda)$ , denoted  $\mathcal{E}(G(\lambda))$  is given by

$$\varphi_t(\lambda) = \log \mathcal{E}(G(\lambda))_t = G_t(\lambda) + \sum_{s \leq t} [\log(1 + \Delta G_s(\lambda)) - \Delta G_s(\lambda)], \quad (8)$$

where

$$\Delta G_s(\lambda) = \int_{|x| \leq a} (e^{\lambda' x} - 1) \nu(\{s\}, dx) = \int_{|x| \leq a} (e^{\lambda' x} - 1 - \lambda' x) \nu(\{s\}, dx). \quad (9)$$

The next lemma which is of independent interest, is key to the proof of Proposition 1.

**Lemma 1** For  $\epsilon > 0$ , let  $v(\epsilon) = 2(e^\epsilon - 1 - \epsilon)/\epsilon^2 \geq 1 \geq v(-\epsilon) - \epsilon^2 v(\epsilon)^2/4 = w(\epsilon)$ . Then, for any  $0 \leq u \leq t < \infty$ ,  $\lambda \in \mathbb{R}^d$

$$\frac{1}{2} w(|\lambda|a) \lambda' (\langle X \rangle_t - \langle X \rangle_u) \lambda \leq \varphi_t(\lambda) - \varphi_u(\lambda) \leq \frac{1}{2} v(|\lambda|a) \lambda' (\langle X \rangle_t - \langle X \rangle_u) \lambda. \quad (10)$$

**Remark 8** Since  $\exp[\lambda' X_t - \varphi_t(\lambda)]$  is a local martingale (c.f. [7, Section 4.13]), Lemma 1 implies that  $\exp[\lambda' X_t - \frac{1}{2} v(|\lambda|a) \lambda' \langle X \rangle_t \lambda]$  is a non-negative super-martingale while  $\exp[\lambda' X_t - \frac{1}{2} w(|\lambda|a) \lambda' \langle X \rangle_t \lambda]$  is a non-negative local sub-martingale. Noting that  $w(|\lambda|a), v(|\lambda|a) \rightarrow 1$  for  $|\lambda| \rightarrow 0$ , these are to be compared with the local martingale property of  $\exp[\lambda' X_t - \frac{1}{2} \lambda' \langle X \rangle_t \lambda]$  when  $X \in \mathcal{M}_{\text{loc},0}^c$  is a *continuous* local martingale (c.f. [7, Section 4.13]).

**Remark 9** For  $d = 1$  it follows that for every  $\lambda \in \mathbb{R}$ ,

$$\mathbb{E}\{\exp[\lambda X_m - \frac{1}{2}v(|\lambda|a)\lambda^2\langle X \rangle_m]\} \leq 1 \quad (11)$$

(c.f. Remark 8). The inequality (2) then follows by Chebycheff's inequality and optimization over  $\lambda \geq 0$ . For the special case of a real-valued discrete-parameter martingale  $X_m$  also

$$\mathbb{E}\{\exp[\lambda X_m - \frac{1}{2}w(|\lambda|a)\lambda^2\langle X \rangle_m]\} \geq 1, \quad (12)$$

and we can even replace  $w(|\lambda|a)$  in (12) by  $v(-|\lambda|a)$  (c.f. [4, (1.4)] where the sub-martingale property of  $\exp(\lambda X_m - \frac{1}{2}v(-|\lambda|a)\lambda^2\langle X \rangle_m)$  is proved).

*Proof:* To prove the upper bound on  $\varphi_t(\lambda) - \varphi_u(\lambda)$  note that  $\log(1+x) - x \leq 0$  implying by (8) that  $\varphi_t(\lambda) - \varphi_u(\lambda) \leq G_t(\lambda) - G_u(\lambda)$ . The required bound then follows from (7) since  $(e^{\lambda'x} - 1 - \lambda'x) \leq \frac{1}{2}v(|\lambda|a)\lambda'(xx')\lambda$  for  $|x| \leq a$ , and  $\lambda'(C_t - C_u)\lambda \geq 0$  for  $u \leq t$ .

To establish the corresponding lower bound, note that since  $\Delta G_s(\lambda) \geq 0$  (see (9)) and  $\log(1+x) - x \geq -x^2/2$  for all  $x \geq 0$ , we have that

$$\varphi_t(\lambda) - \varphi_u(\lambda) \geq G_t(\lambda) - G_u(\lambda) - \frac{1}{2} \sum_{u < s \leq t} \Delta G_s(\lambda)^2.$$

Moreover, again by (9) we see that

$$0 \leq \Delta G_s(\lambda) \leq \frac{1}{2} v(|\lambda|a)\lambda' \left[ \int_{|x| \leq a} xx' \nu(\{s\}, dx) \right] \lambda \leq \frac{1}{2} v(|\lambda|a)^2 (|\lambda|a)^2.$$

Hence,

$$\begin{aligned} \frac{1}{2} \sum_{u < s \leq t} \Delta G_s(\lambda)^2 &\leq \frac{1}{8} v(|\lambda|a)^2 (|\lambda|a)^2 \lambda' \left[ \sum_{u < s \leq t} \int_{|x| \leq a} xx' \nu(\{s\}, dx) \right] \lambda \\ &\leq \frac{1}{8} v(|\lambda|a)^2 (|\lambda|a)^2 \lambda' [\langle X \rangle_t - \langle X \rangle_u] \lambda, \end{aligned}$$

and the required lower bound follows by noting that

$$G_t(\lambda) - G_u(\lambda) \geq \frac{1}{2} v(-|\lambda|a)\lambda' [\langle X \rangle_t - \langle X \rangle_u] \lambda. \quad \blacksquare$$

To prove Proposition 1 we need the following immediate consequence of Lemma 1.

**Lemma 2** *Suppose there exists  $q \in C[0, \infty)$ , a positive-semi-definite matrix  $Q$  and an unbounded function  $h : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  such that for all  $\delta > 0, T < \infty$*

$$\limsup_{k \rightarrow \infty} \frac{1}{h_k} \log \mathbb{P} \left\{ \sup_{u \in [0, T]} \left\| \frac{\langle X \rangle_{uk}}{h_k} - q(u)Q \right\| > \delta \right\} < 0. \quad (13)$$

*Then, for every  $\lambda \in \mathbb{R}^d$  and  $a_k \rightarrow 0$  such that  $h_k a_k \rightarrow \infty$ ,*

$$\limsup_{k \rightarrow \infty} a_k \log \mathbb{P} \left\{ \sup_{u \in [0, T]} \left| a_k \varphi_{uk}(\lambda / \sqrt{h_k a_k}) - \frac{1}{2} q(u) \lambda' Q \lambda \right| > \delta \right\} = -\infty. \quad (14)$$

*Proof:* Use (10), noting that  $a_k = \frac{1}{h_k}(a_k h_k)$  with  $a_k h_k \rightarrow \infty$ , and that  $\lim_{k \rightarrow \infty} v(|\lambda|a/\sqrt{a_k h_k}) = \lim_{k \rightarrow \infty} w(|\lambda|a/\sqrt{a_k h_k}) = 1$ , while  $\sup_{u \in [0, T]} |q(u)| < \infty$ . ■

The next lemma is a simple application of the results of [8], relating (14) with the LDP (with speed  $a_k$ ) of  $\left\{ \sqrt{\frac{a_k}{h_k}} X_{k \cdot} \right\}$ .

**Lemma 3** *When (14) holds, the processes  $\left\{ \sqrt{\frac{a_k}{h_k}} X_{k \cdot}, k > 0 \right\}$  satisfy the LDP in  $(D(\mathbb{R}^d), \mathcal{B})$  with speed  $a_k$  and the good rate function*

$$I(\phi) = \begin{cases} \int_0^\infty \Lambda^* \left( \frac{d\phi}{dq}(t) \right) q(dt) & \phi \ll q, \quad \phi(0) = 0 \\ \infty & \text{otherwise} \end{cases} \quad (15)$$

(where  $q \in M_+(\mathbb{R}_+)$  is the continuous locally finite measure on  $(\mathbb{R}_+, \mathcal{B}_{\mathbb{R}_+})$  such that  $q([0, t]) = q(t)$ ).

*Proof:* For each sequence  $k_n \rightarrow \infty$  we shall apply [8, Theorem 2.2] for the local martingales  $\sqrt{a_{k_n}/h_{k_n}} X_{k_n t}$  replacing  $\frac{1}{n}$  throughout by  $a_{k_n}$ . Cramér's condition [8, (2.6)] is trivially holding in the current setting, while for  $G_t(\lambda) = \frac{1}{2} q(t) \lambda' Q \lambda$  the condition (sup  $\mathcal{E}$ ) of [8, Theorem 2.2] is merely (14). Moreover, for this  $G_t(\lambda)$  the condition [8, (G)] is easily shown to hold (as  $H_{s,t}(\cdot)$  is then a positive-definite quadratic form on the linear subspace  $\text{dom} H_{s,t}$  for all  $s < t$ ). Thus, the LDP in Skorohod topology follows from [8, Theorem 2.2] and the explicit form (15) of the rate function follows from [8, (2.4)] taking there  $g_t(\lambda) = \frac{1}{2} \lambda' Q \lambda$ . Suppose  $I(\phi) < \infty$ . Then,  $\phi \ll q$  and since  $q \in C[0, \infty)$  it follows that  $\phi \in C(\mathbb{R}^d)$ . Hence, by [8, Theorem C] we may replace the Skorohod topology by the stronger locally uniform topology on  $D(\mathbb{R}^d)$ . ■

Proposition 1 follows by combining Lemmas 2 and 3 with the next lemma.

**Lemma 4** *If  $h_t$  is regularly varying of index  $\alpha > 0$  then (5) implies that (13) holds for  $q(u) = u^\alpha$ .*

**Proof:** Fix  $T < \infty$  and  $\delta > 0$ . Since  $h_t$  is regularly varying of index  $\alpha > 0$ , clearly  $h_{uk}/h_k \rightarrow u^\alpha$  for all  $u \in (0, \infty)$  (c.f. [2, page 18]). Take  $\epsilon > 0$  small enough for  $\sup_{0 \leq i \leq \lceil T/\epsilon \rceil} |q(i\epsilon + \epsilon) - q(i\epsilon)| \leq$

$\delta/(3\|Q\|)$ , and  $k_0 < \infty$  such that  $\sup_{0 \leq i \leq \lceil T/\epsilon \rceil} |h_{i\epsilon k}/h_k - q(i\epsilon)| \leq \delta/(3\|Q\|)$  whenever  $k \geq k_0$  (note

that  $q(0) = 0$ ).

The monotonicity of  $\langle X \rangle_{tk}$  in  $t$  (and  $\langle X \rangle_0 = 0$ ) implies that for all  $k \geq k_0$

$$\left\{ \sup_{u \in [0, T]} \left\| \frac{\langle X \rangle_{uk}}{h_k} - q(u)Q \right\| > \delta \right\} \subseteq \left\{ \sup_{1 \leq i \leq \lceil T/\epsilon \rceil} \|\langle X \rangle_{i\epsilon k} - h_{i\epsilon k}Q\| > \frac{1}{3}\delta h_k \right\}.$$

Hence, suffices to show that for every  $i \in \mathbb{N}$ ,  $\epsilon > 0$

$$\limsup_{k \rightarrow \infty} \frac{1}{h_k} \log \mathbb{P} \left\{ \|\langle X \rangle_{i\epsilon k} - h_{i\epsilon k}Q\| > \frac{1}{3}\delta h_k \right\} < 0.$$

Since  $h_{i\epsilon k}/h_k \rightarrow q(i\epsilon) \in (0, \infty)$  this inequality follows from (5). ■

### 3 Proof of Corollary 1

(a) Assume first that  $h_t$  is regularly varying of index 1. Given Proposition 1, this case is easily settled by applying the contraction principle for the continuous mapping  $\phi \mapsto \phi(1) : D[\mathbb{R}^d] \rightarrow \mathbb{R}^d$ . In the general case, we take without loss of generality  $h_t \in D(\mathbb{R}_+)$  strictly increasing of bounded jumps (see Remark 1). Let  $\sigma_s = \inf\{t \geq 0 : h_t \geq s\}$  and  $g_s = h_{\sigma_s}$ . Note that  $g_s - s$  is bounded, while (5) holds for the locally square integrable martingale  $Y_s = X_{\sigma_s}$  of bounded jumps and the regularly varying function  $g_s$  of index 1. Consequently,  $\{g_s^{-1/2}Y_s\}$  satisfies the MDP with the critical speed  $1/g_s$  and the good rate function  $\Lambda^*(\cdot)$ . Since  $h_t$  is strictly increasing and unbounded it follows that  $\sigma(\mathbb{R}_+) = \mathbb{R}_+$ . Hence, this MDP is equivalent to the MDP for  $\{h_k^{-1/2}X_k\}$ .

(b) As in part (a) above suffices to prove the stated MDP for  $h_t$  regularly varying of index 1. Applying the contraction principle for the continuous mapping  $\phi \mapsto \sup_{s \leq 1} \phi(s)$  we deduce the stated MDP from Proposition 1. Since  $\Lambda^*(v) = v^2/(2Q)$ , the good rate function for this MDP is (c.f. (6))

$$I(z) = \frac{1}{2Q} \inf_{\{\phi \in AC_0 : \sup_{s \leq 1} \phi(s) = z\}} \int_0^\infty \dot{\phi}(s)^2 ds \geq \frac{z^2}{2Q}.$$

Clearly,  $\phi(0) = 0$  implies that  $I(z) = \infty$  for  $z < 0$ , while taking  $\phi(s) = (s \wedge 1)z$  we conclude that  $I(z) = z^2/(2Q)$  for  $z \geq 0$ .

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