

Large deviations for two scaled diffusions

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Summary. We formulate large deviations principle (LDP) for diffusion pair $(X^\varepsilon, \zeta^\varepsilon) = (X_t^\varepsilon, \zeta_t^\varepsilon)$, where first component has a small diffusion parameter while the second is ergodic Markovian process with fast time. More exactly, the LDP is established for $(X^\varepsilon, \nu^\varepsilon)$ with $\nu^\varepsilon(dt, dz)$ being an occupation type measure corresponding to ζ_t^ε . In some sense we obtain a combination of Freidlin–Wentzell’s and Donsker–Varadhan’s results. Our approach relies on the concept of the exponential tightness and Puhalskii’s theorem.

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1 Introduction

Let ε be a small positive parameter, $(X^\varepsilon, \zeta^\varepsilon) = (X_t^\varepsilon, \zeta_t^\varepsilon)_{t \geq 0}$ be a diffusion pair defined on some stochastic basis $(\Omega, \mathcal{F}, \mathbf{F} = (\mathcal{F}_t)_{t \geq 0}, \mathbf{P})$ by Itô’s equations w.r.t. independent Wiener processes W_t and V_t :

$$\begin{aligned} dX_t^\varepsilon &= A(X_t^\varepsilon, \zeta_t^\varepsilon) dt + \sqrt{\varepsilon} B(X_t^\varepsilon, \zeta_t^\varepsilon) dW_t, \\ d\zeta_t^\varepsilon &= \frac{1}{\varepsilon} b(\zeta_t^\varepsilon) dt + \frac{1}{\sqrt{\varepsilon}} \sigma(\zeta_t^\varepsilon) dV_t \end{aligned} \quad (1.1)$$

subject to fixed initial point (x_0, z_0) .

Assume $(X^\varepsilon, \zeta^\varepsilon)$ is an ergodic process in the following sense. Let $p(z)$ be the unique invariant density of ζ^ε ,

$$\nu^{(p)}(dt, dz) = p(z) dt dz,$$

and \bar{X}_t is a solution of an ordinary differential equation $\dot{\bar{X}}_t = \bar{A}(\bar{X}_t)$ with $\bar{A}(x) = \int_{\mathcal{R}} A(x, z) p(z) dz$ subject to the same initial point x_0 . Then for any

bounded continuous function $h(t, z)$ and $T > 0$

$$\begin{aligned} \mathbf{P} - \lim_{\varepsilon \rightarrow 0} \int_0^T h(t, \xi_t^\varepsilon) dt &= \int_0^T \int_R h(t, z) v^{(p)}(dt, dz), \\ \mathbf{P} - \lim_{\varepsilon \rightarrow 0} r_T(X^\varepsilon, \bar{X}) &= 0, \end{aligned} \quad (1.2)$$

where r_T is the uniform metric on $[0, T]$. The above-mentioned ergodic property is a motivation to examine LDP for pair $(X^\varepsilon, \xi^\varepsilon)$, or more exactly for pair $(X^\varepsilon, v^\varepsilon)$, where $v^\varepsilon = v^\varepsilon(dt, dz)$ is an occupation measure on $(R_+ \times R, \mathcal{B}(R_+) \otimes \mathcal{B}(R))$ ($\mathcal{B}(R_+)$, $\mathcal{B}(R)$ are the Borel σ -algebras on R_+ and R respectively) corresponding to ξ^ε :

$$v^\varepsilon(\Delta \times \Gamma) = \int_0^\infty I(t \in \Delta, \xi_t^\varepsilon \in \Gamma) dt, \quad \Delta \in \mathcal{B}(R_+), \Gamma \in \mathcal{B}(R). \quad (1.3)$$

A choice of v^ε as the occupation measure is natural since the first ergodic property in (1.2) is nothing but

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \rho_T(v^\varepsilon, v^{(p)}) = 0,$$

where ρ_T is Levy–Prohorov’s distance for restrictions of measures v^ε and $v^{(p)}$ on $[0, T] \times R$. Also the first Itô’s equation in (1.1) and the predictable quadratic variation $\langle M^\varepsilon \rangle_t$ of a martingale $M_t^\varepsilon = \int_0^t B(X_s^\varepsilon, \xi_s^\varepsilon) dW_s$ can be represented in the term of v^ε :

$$\begin{aligned} X_t^\varepsilon &= x_0 + \int_0^t \int_R A(X_s^\varepsilon, z) v^\varepsilon(ds, dz) + \sqrt{\varepsilon} M_t^\varepsilon, \\ \langle M^\varepsilon \rangle_t &= \int_0^t \int_R B^2(X_s^\varepsilon, z) v^\varepsilon(ds, dz). \end{aligned}$$

The random measure v^ε obeys the disintegration $v^\varepsilon(dt, dz) = dt K_{v^\varepsilon}(t, dz)$ with the transition kernel $K_{v^\varepsilon}(t, dz)$ being probabilistic Dirac’s measure that is v^ε values in space $\mathbb{M} = \mathbb{M}_{[0, \infty)}$ of σ -finite (locally in t) measures $v = v(dt, dz)$ on $(R_+ \times R, \mathcal{B}(R_+) \otimes \mathcal{B}(R))$ obeying the disintegration $v(dt, dz) = K_v(t, dz) dt$ with the probabilistic transition kernel $K_v(t, dz)$ ($\int_R K_v(t, dz) \equiv 1$). X^ε values in the space $\mathbb{C} = \mathbb{C}_{[0, \infty)}$ of continuous function. Define metrics r and ρ in \mathbb{C} and \mathbb{M} respectively, letting

$$r(X', X'') = \sum_{k \geq 1} \frac{r_k(X', X'') \wedge 1}{2^k} \quad \text{and} \quad \rho(v', v'') = \sum_{k \geq 1} \frac{\rho_k(v', v'') \wedge 1}{2^k}.$$

Evidently ergodic properties (1.2) are equivalent to

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} [r(X^\varepsilon, \bar{X}) + \rho(v^\varepsilon, v^{(p)})] = 0$$

and so for examination of the LDP for $(X^\varepsilon, v^\varepsilon)$ we choose the metric space $(\mathbb{C} \times \mathbb{M}, r \times \rho)$.

Recall the definition of LDP from Varadhan [1] adapted to our setting. The family $(X^\varepsilon, \nu^\varepsilon)$ obeys the LDP in the metric space $(\mathbf{C} \times \mathbf{M}, r \times \rho)$ if

- (0) there exists a function $L(X, \nu)$, $X \in \mathbf{C}$, $\nu \in \mathbf{M}$, values in $[0, \infty]$, such that its level sets are compacts in $(\mathbf{C} \times \mathbf{M}, r \times \rho)$;
 (1) for any open set G from $(\mathbf{C} \times \mathbf{M}, r \times \rho)$

$$\liminf_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}((X^\varepsilon, \nu^\varepsilon) \in G) \geq - \inf_{(X, \nu) \in G} L(X, \nu);$$

- (2) for any closed set F from $(\mathbf{C} \times \mathbf{M}, r \times \rho)$

$$\limsup_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}((X^\varepsilon, \nu^\varepsilon) \in F) \leq - \inf_{(X, \nu) \in F} L(X, \nu).$$

The function $L(X, \nu)$, meeting in (0), (1), and (2), is named rate function (action functional in the terminology of Freidlin and Wentzell [2] or good rate function in the terminology of Stroock [3]).

Below we recall well known particular results in LDPs related to pair $(X^\varepsilon, \zeta^\varepsilon)$ and give corresponding forms of rate functions which will be inherited by a rate function for our setting. Note at first LDP for family $\mu^\varepsilon(dz) = \nu^\varepsilon([0, 1], dz)$ (on the space of probability measures supplied by Levy–Prohorov’s metric) proved by Donsker and Varadhan [4–7] for a wide class of Markov processes $\zeta_t^\varepsilon = \zeta_{t/\varepsilon}$. Corresponding rate function obeys an invariant form: for any probabilistic measure μ on R

$$I(\mu) = - \inf \int_R \frac{\mathcal{L}u(z)}{u(z)} \mu(dz),$$

where \mathcal{L} is backward Kolmogorov’s operator, respecting to ζ , and where “inf” is taken over all functions $u(z)$ from the domain of definition for the operator \mathcal{L} . For the diffusion case, Gärtner’s type of $I(\mu)$ is well known [8]:

$$I(\mu) = \begin{cases} \frac{1}{8} \int_R \sigma^2(z) \left[\frac{m'(z)}{m(z)} - \frac{p'(z)}{p(z)} \right]^2 m(z) dz, & d\mu = m(z) dz, \quad dm(z) = m'(z) dz, \\ \infty, & \text{otherwise.} \end{cases} \quad (1.4)$$

Freidlin–Wentzell’s result [2] is devoted to LDP for diffusion X^ε with drift $A(x)$ and diffusion $B^2(x)$ (independent of z) in the space of continuous functions on every finite time interval, supplied by the uniform metric. A rate function, say, for $[0, T]$ time interval is given by

$$S(X) = \begin{cases} \frac{1}{2} \int_0^T \frac{[\dot{X}_t - A(X_t)]^2}{B^2(X_t)} dt, & dX_t = \dot{X}_t dt, \quad X_0 = x_0, \\ \infty, & \text{otherwise.} \end{cases} \quad (1.5)$$

Other type of LDP for a degenerate diffusion X^ε defined by the first equation in (1.1) with $B(x, z) \equiv 0$ and $\zeta_t^\varepsilon = \zeta_{t/\varepsilon}$, where ζ_t is Markov process values in a finite state space, also is well known from Freidlin [9]. In this case

rate function has a form similar to (1.5) ($H(y, x)$ is some non negative function):

$$S(X) = \begin{cases} \int_0^T H(\dot{X}_t, X_t) dt, & dX_t = \dot{X}_t dt, X_0 = x_0, \\ \infty, & \text{otherwise.} \end{cases} \quad (1.6)$$

All above-mentioned LDPs are inspired by the examination of the LDP for $(X^\varepsilon, \nu^\varepsilon)$. In some sense, the LDP for $(X^\varepsilon, \nu^\varepsilon)$ is a combination of Donsker–Varadhan’s and Freidlin–Wenzell’s results. Namely LDP for ν^ε is a generalization one for μ^ε while LDP for X^ε is implied by LDP for ν^ε and for a diffusion martingale scaled by $\sqrt{\varepsilon}$. Hence, a rate function for $(X^\varepsilon, \nu^\varepsilon)$, is defined as a sum: $L(X, \nu) = L_1(X, \nu) + L_2(\nu)$, where $L_1(X, \nu)$ and $L_2(\nu)$ respect to X^ε and ν^ε and what is more $L_1(X, \nu)$ has the same form as $S(x)$ in (1.5) with $A(X_t)$ and $B^2(X_t)$ replaced on $A_\nu(t, X_t) = \int_R A(X_t, z) K_\nu(t, dz)$ and $B_\nu^2(t, X_t) = \int_R B^2(X_t, z) K_\nu(t, dz)$, where $K_\nu(t, dz)$ is the transition kernel of measure ν .

Note that $\zeta_t^\varepsilon \in R$ and so the LDP for its occupation measure responds to a non compact diffusion case. Also note that diffusion parameter $B^2(x, z)$ is not assumed to be non singular and consequently $B^2(x, z) \equiv 0$ is admissible. The last allows to derive LDP for a singular diffusion parameter case from LDP for ν^ε using the contraction principle of Varadhan [1] (continuous mapping method of Freidlin [10]). This result extends above-mentioned [9] for non compact case.

In contrast with Freidlin and Wentzell [2], Donsker and Varadhan [4–7], Gärtner [8], and Veretennikov [11,12], and many others (see e.g. Acosta [13], Dupuis and Elis [14]) our method of proof is based on Puhalskii’s theorem [15,16] and relies concepts of exponential tightness and LD relative compactness.

The paper is organized as follows. In Sect. 2, we formulate the general assumptions and the main result. Section 3 contains the method of proving LDP which also has been used in [17]. In Sect. 4, we check the exponential tightness while in Sects. 5 and 6 the upper and lower bounds in local LDP are verified. The main results are proved in Sect. 7. All technical results are gathered in Appendix.

2 Assumptions. Main result

1. We fix the following conditions which are assumed to be fulfilled hereafter.

(A.1) $A(x, z)$ and $B(x, z)$ are continuous in (x, z) , Lipschitz continuous in x uniformly in z , and $\sup_z (|A(0, z)| + |B(0, z)|) < \infty$;

(A.2) $\sigma^2(z)$ is bounded and uniformly positive function; it is continuously differentiable, having bounded and Lipschitz continuous derivative;

(A.3) $b(z)$ is Lipschitz continuous, satisfying

$$\lim_{|z| \rightarrow \infty} b(z) \text{sign } z = -\infty.$$

It would be noted that (A.2) and (A.3) imply, so called, assumption (H*) from [6].

2. It is well known (see [18]) that under (A.2) and (A.3) ξ^ε is ergodic process obeying the unique invariant density

$$p(z) = \text{const.} \frac{\exp\left(2\int_0^z (b(y)/\sigma^2(y)) dy\right)}{\sigma^2(z)}. \quad (2.1)$$

For any ν from \mathbb{M} with the transition kernel $K_\nu(t, dz)$, define $K_\nu(t, dz)$ -averaged drift $A_\nu(t, x) = \int_R A(x, z)K_\nu(t, dz)$ and diffusion parameter $B_\nu^2(t, x) = \int_R B^2(x, z)K_\nu(t, dz)$. If ν is absolutely continuous w.r.t. $\Lambda(dt, dz) = dt dz$, put

$$n(t, z) = \frac{d\nu}{d\Lambda}(t, z). \quad (2.2)$$

If the density $n(t, z)$ is absolutely continuous w.r.t. dz : $d_z n(t, z) = n'_z(t, z) dz$, a function $n'_z(t, z)$ is chosen to be measurable in t, z .

Throughout the paper, we use conventions $0/0 = 0$ and $\min(\inf)(\emptyset) = \infty$.

For every $\nu \in \mathbb{M}$ and $X \in \mathbb{C}$ define two quantities (comp. (1.4) and (1.5)):

$$F(\nu) = \begin{cases} \int_0^\infty \int_R \sigma^2(z) \left[\frac{n'_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]^2 n(t, z) dz dt, & d\nu = n d\Lambda, \quad d_z n = n'_z dz, \\ \infty, & \text{otherwise;} \end{cases} \quad (2.3)$$

$$S(X, \nu) = \begin{cases} \int_0^\infty \frac{[\dot{X}_t - A_\nu(t, X_t)]^2}{B_\nu^2(t, X_t)} dt, & dX = \dot{X} dt, \quad X_0 = x_0, \\ \infty, & \text{otherwise.} \end{cases}$$

3. Now we are in the position to formulate the main result.

Theorem 2.1 Under (A.1), (A.2), and (A.3) the family $(X^\varepsilon, \nu^\varepsilon)$ obeys the LDP in $(\mathbb{C} \times \mathbb{M}, r \times \rho)$ with rate function

$$L(X, \nu) = \frac{1}{2}S(X, \nu) + \frac{1}{8}F(\nu).$$

4. LDPs for families (X^ε) and (ξ^ε) run out from Theorem 2.1.

Corollary 2.1 (ν^ε) obeys the LDP in (\mathbb{M}, ρ) with rate function $\frac{1}{8}F(\nu)$.

Corollary 2.2 (comp. [9]) (X^ε) obeys the LDP in (\mathbb{C}, r) with rate function $S(X) = \inf_{\nu \in \mathbb{M}} L(X, \nu)$. In particular, if $B(x, z) \equiv 0$, it is sufficient to take “inf” over all ν from \mathbb{M} with the transition kernel $K_\nu(t, dz) \equiv \mu(dz)$ with $d\mu = m(z) dz$ such that the density $m(z) = (d\mu/dz)(z)$ is absolutely continuous w.r.t. dz ($m'(z) = dm(z)/dz$). In this case, rate function

$$S(X) = \begin{cases} \frac{1}{8} \int_0^\infty H(\dot{X}_t, X_t) dt, & dX = \dot{X} dt, \quad X_0 = x_0, \\ \infty, & \text{otherwise,} \end{cases} \quad (2.4)$$

where

$$H(y, x) = \inf \int_R \sigma^2(z) \left[\frac{m'(z)}{m(z)} - \frac{p'(z)}{p(z)} \right]^2 m(z) dz, \quad (2.5)$$

and where “inf” is taken over all above-mentioned measures μ such that

$$y = \int_R A(x, z) m(z) dz.$$

As an example, also the LDP for the family of the Donsker and Varadhan occupation measures $\mu^\varepsilon(dz) = v^\varepsilon([0, 1] \times dz)$, corresponding to diffusion case, can be derived from Theorem 2.1. In fact, due to the contraction principle, (μ^ε) obeys the LDP with Gärtner’s type rate function (see (1.4)) $I(\mu) = \inf \frac{1}{8} F(v)$, where “inf” is taken over all $v \in \mathbb{M}$ such that

$$v(dt, dz) = I(1 \geq t) dt \mu(dz) + I(1 < t) v^{(p)}(dt, dz).$$

3 Preliminaries

For proving LDP for the family $(X^\varepsilon, v^\varepsilon)$ in the metric space $(\mathbb{C} \times \mathbb{M}, r \times \rho)$ we apply Dawson–Gärtner’s type theorem (see e.g. [19]). Following it the LDP in $(\mathbb{C} \times \mathbb{M}, r \times \rho)$ is implied by LDPs in the metric spaces $(\mathbb{C}_{[0, n]} \times \mathbb{M}_{[0, n], r_n \times \rho_n})$, $n \geq 1$, where $\mathbb{C}_{[0, n]}$ is the space of continuous functions on the time interval $[0, n]$, $\mathbb{M}_{[0, n]}$ is the space of finite measures on $[0, n] \times R$, having probabilistic transition kernel w.r.t. dt , r_n is the uniform metric, and ρ_n is Levy–Prohorov’s metric. The definition of the LDP in $(\mathbb{C}_{[0, n]} \times \mathbb{M}_{[0, n], r_n \times \rho_n})$ is given in terms of (0), (1), and (2) with obvious modifications. Moreover, if $L_n(X, v)$, $n \geq 1$ are rate functions, corresponding to LDPs in $(\mathbb{C}_{[0, n]} \times \mathbb{M}_{[0, n], r_n \times \rho_n})$, $n \geq 1$, then rate function in $(\mathbb{C} \times \mathbb{M}, r \times \rho)$ is defined as

$$L(X, v) = \sup_n L_n(X, v). \quad (3.1)$$

Hence only the LDP in $(\mathbb{C}_{[0, T]} \times \mathbb{M}_{[0, T], r_T \times \rho_T})$ has to be checked for any $T > 0$. Our approach in proving the LDP in $(\mathbb{C}_{[0, T]} \times \mathbb{M}_{[0, T], r_T \times \rho_T})$, $T > 0$ relies on the concept of the exponential tightness and notions of LD relative compactness and local LDP. Below we give necessary definitions.

Definition 1 *The family $(X^\varepsilon, v^\varepsilon)$ is said to be exponentially tight in the metric space $(\mathbb{C}_{[0, T]} \times \mathbb{M}_{[0, T], r_T \times \rho_T})$, if there exists an increasing sequence of compacts $(K_j)_{j \geq 1}$ such that*

$$\lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}((X^\varepsilon, v^\varepsilon) \in \{\mathbb{C}_{[0, T]} \times \mathbb{M}_{[0, T]}\} \setminus K_j) = -\infty \quad (3.2)$$

(Deuschel and Stroock [20], Lynch and Sethuraman [21]).

Definition 2 *The family $(X^\varepsilon, v^\varepsilon)$ is said to be LD relatively compact in $(\mathbb{C}_{[0, T]} \times \mathbb{M}_{[0, T], r_T \times \rho_T})$, if any decreasing to zero sequence (ε_k) contains*

further subsequence $(\bar{\varepsilon}_k)$ ($(\bar{\varepsilon}_k) \subseteq (\varepsilon_k)$) such that the family $(X^{\bar{\varepsilon}_k}, v^{\bar{\varepsilon}_k})$ obeys the LDP in $(\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T], r_T \times \rho_T})$ (with rate function $L_T(X, v)$). (Puhalskii [15,16]).

Definition 3 The family $(X^\varepsilon, v^\varepsilon)$ is said to obey the local LDP in $(\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T], r_T \times \rho_T})$ with local rate function $\widehat{L}_T(X, v)$, if for any (X, v) from $\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T]}$

$$\begin{aligned} & \lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}((r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ &= \lim_{\delta \rightarrow 0} \underline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}((r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ &= -\widehat{L}_T(X, v) \end{aligned} \quad (3.3)$$

(Freidlin and Wentzell [2]).

The connecting component of these notions used in the proof of the next result is Puhalskii's theorem [15,16]. Below we formulate only the first part of it.

Theorem P If $(X^\varepsilon, v^\varepsilon)$ is exponentially tight family in $(\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T], r_T \times \rho_T})$, then it is LD relatively compact.

The following result is a reformulation of Theorem 1.3 from [17].

Proposition 3.1 The exponential tightness and the local LDP for the family $(X^\varepsilon, v^\varepsilon)$ in $(\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T], r_T \times \rho_T})$ imply the LDP in $(\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T], r_T \times \rho_T})$ for this family with (good) rate function $L_T(X, v) \equiv \widehat{L}_T(X, v)$, where $\widehat{L}_T(X, v)$ is the local rate function.

4 Exponential tightness in $\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T]}$

Theorem 4.1 Under assumptions (A.1), (A.2), and (A.3) the family $(X^\varepsilon, v^\varepsilon)$ is exponentially tight in $\mathbf{C}_{[0,T]} \times \mathbf{M}_{[0,T]}$.

Proof. Following Definition 1, (3.2) has to be checked. It is clear it takes place if

$$\begin{aligned} & \lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(X^\varepsilon \in \mathbf{C}_{[0,T]} \setminus K_j') = -\infty, \\ & \lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(v^\varepsilon \in \mathbf{M}_{[0,T]} \setminus K_j'') = -\infty, \end{aligned} \quad (4.1)$$

where K_j' and K_j'' are appropriate increasing sequences of compacts from $\mathbf{C}_{[0,T]}$ and $\mathbf{M}_{[0,T]}$ respectively. It is natural to use as compacts K_j' increasing sets of uniformly bounded and equicontinuous functions from $\mathbf{C}_{[0,T]}$ parametrized by j . Since the process $(X_t^\varepsilon, \xi_t^\varepsilon)_{t \geq 0}$ is defined on a stochastic basis with the filtration \mathbf{F} one can use Aldous–Puhalskii's type sufficient conditions

(see [15], and also Theorem 3.1 in [17]) for C -exponential tightness:

$$\begin{aligned} \lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P} \left(\sup_{t \leq T} |X_t^\varepsilon| > j \right) &= -\infty, \\ \lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \sup_{\tau \leq T-\delta} \mathbf{P} \left(\sup_{t \leq \delta} |X_{\tau+t}^\varepsilon - X_\tau^\varepsilon| > \eta \right) &= -\infty, \quad \forall \eta > 0, \end{aligned} \quad (4.2)$$

where τ is a stopping time w.r.t. the filtration \mathbf{F} . Following Theorem 3.1 in [17], (4.2) implies the validity of the first part in (4.1) with above-mentioned compacts K'_j of uniformly bounded and equicontinuous functions. Now, choose relevant compacts K''_j , $j \geq 1$:

$$K''_j = \bigcap_{m \geq j} \left\{ v \in \mathbf{M}_{[0, T]} : \int_0^T \int_{|z| > m} v(dt, dz) \leq g(m) \right\}, \quad (4.3)$$

where $g(y)$, $y > 0$ is positive continuous decreasing function with $\lim_{y \rightarrow \infty} g(y) = 0$. In fact, if $v_k \in K''_j$, $k \geq 1$ then we have for any $m \geq j$ $\sup_k \int_0^T \int_{|z| > m} v_k(dt, dz) \leq g(m)$ that is the set K''_j is tight and by Prohorov's theorem (see [22]) is relatively compact. On the other hand, since the set $\{z: |z| > m\}$ is open a limit of any converging sequence from K''_j also belongs to K''_j that is K''_j is compact in $(\mathbf{M}_{[0, T]}, \rho_T)$. Evidently $K''_j \subseteq K''_{j+1}$. Below we choose a special function $g(y)$, suited to assumption (A.3), to satisfy the second part in (4.1).

We check the validity of (4.1) in the next two lemmas.

Lemma 4.1 *Under (A.1) the first relation in (4.1) holds.*

Lemma 4.2 *Under (A.2) and (A.3) the second relation in (4.1) holds.*

Proof of Lemma 4.1. Put

$$Z_t^* = \sup_{t' \leq t} |Z_{t'}|.$$

By virtue of (A.1) we have $|A(x, z)| \leq \ell(1 + |x|)$. Therefore, with $t \leq T$, we derive from (1.1)

$$X_t^{\varepsilon*} \leq |x_0| + \ell \int_0^t (1 + X_s^{\varepsilon*}) ds + \sqrt{\varepsilon} M_T^{\varepsilon*}, \quad (4.4)$$

where $M_t^\varepsilon = \int_0^t B(X_s^\varepsilon, \zeta_s^\varepsilon) dW_s$. Due to Bellman–Gronwall's inequality, (4.4) implies $X_T^{\varepsilon*} \leq \text{const.}(1 + \sqrt{\varepsilon} M_T^{\varepsilon*})$ with const., depending only on $|x_0|, \ell$, and T . Therefore, the first part of (4.2) holds if

$$\lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(M_T^{\varepsilon*} > j) = -\infty. \quad (4.5)$$

On the other hand, by Chebyshev's inequality $\mathbf{P}(M_T^{\varepsilon*} > j) \leq j^{-1/\varepsilon} \mathbf{E}(M_T^{\varepsilon*})^{1/\varepsilon}$ and so, $\varepsilon \log \mathbf{P}(M_T^{\varepsilon*} > j) \leq -\log j + \varepsilon \log \mathbf{E}(M_T^{\varepsilon*})^{1/\varepsilon}$. Thereby (4.5) holds if

$$\overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{E}(M_T^{\varepsilon*})^{1/\varepsilon} < \infty. \quad (4.6)$$

Below we check the validity (4.6). Assuming $1/\varepsilon > 2$ and applying Itô's formula to $|M_t^\varepsilon|^{1/\varepsilon}$, we get

$$\begin{aligned} |M_t^\varepsilon|^{1/\varepsilon} &= \frac{1}{\sqrt{\varepsilon}} \int_0^t |M_s^\varepsilon|^{1/\varepsilon-1} (\text{sign } M_s^\varepsilon) B(X_s^\varepsilon, \zeta_s^\varepsilon) dW_s \\ &\quad + \frac{1-\varepsilon}{2\varepsilon} \int_0^t |M_s^\varepsilon|^{1/\varepsilon-2} B^2(X_s^\varepsilon, \zeta_s^\varepsilon) ds, \end{aligned}$$

that is $|M_t^\varepsilon|^{1/\varepsilon}$ is a submartingale obeying a decomposition: $|M_t^\varepsilon|^{1/\varepsilon} = N_t^\varepsilon + U_t^\varepsilon$ with local martingale N_t^ε and predictable increasing process

$$U_t^\varepsilon = \frac{1-\varepsilon}{2\varepsilon} \int_0^t |M_s^\varepsilon|^{1/\varepsilon-2} B^2(X_s^\varepsilon, \zeta_s^\varepsilon) ds. \quad (4.7)$$

Then, due to a modification of Doob's inequality (see [23], Theorem 1.9.2)

$$\mathbf{E}(M_t^{\varepsilon*})^{1/\varepsilon} \leq \left(\frac{1}{1-\varepsilon} \right)^{1/\varepsilon} \mathbf{E}U_t^\varepsilon. \quad (4.8)$$

Now evaluate from above $|M_s^\varepsilon|^{1/\varepsilon-2} B^2(X_s^\varepsilon, \zeta_s^\varepsilon)$. By virtue of (A.1) $|B(x, z)| \leq \ell(1+|x|)$. Thereby, due to above-mentioned upper bound $X_T^{\varepsilon*} \leq \text{const.}(1+M_T^{\varepsilon*})$ which remains true with replacing T on s for any $s < T$, we arrive at

$$\begin{aligned} |M_s^\varepsilon|^{1/\varepsilon-2} B^2(X_s^\varepsilon, \zeta_s^\varepsilon) &\leq \text{const.}(1 + |M_s^\varepsilon|^{1/\varepsilon-2} + |M_s^\varepsilon|^{1/\varepsilon}) \\ &\leq \text{const.}(1 + (M_s^{\varepsilon*})^{1/\varepsilon}). \end{aligned}$$

Substituting the last upper bound in (4.7) and using (4.8) we obtain ($t \leq T$) $\mathbf{E}(M_t^{\varepsilon*})^{1/\varepsilon} \leq (\text{const.}/\varepsilon) \int_0^t [1 + \mathbf{E}(M_s^{\varepsilon*})^{1/\varepsilon}] ds$. Hence, by Bellman–Gronwall's inequality, an upper bound $\mathbf{E}(M_T^{\varepsilon*})^{1/\varepsilon} \leq (\text{const.}T/\varepsilon) \exp\{\text{const.}T/\varepsilon\}$ holds and implies (4.6). Consequently the first part in (4.2) is valid. To check the second part in (4.2), first use obvious estimates:

$$\begin{aligned} &\mathbf{P} \left(\sup_{t \leq \delta} |X_{\tau+t}^\varepsilon - X_\tau^\varepsilon| > \eta \right) \\ &\leq \mathbf{P} \left(\sup_{t \leq \delta} |X_{\tau+t}^\varepsilon - X_\tau^\varepsilon| > \eta, X_T^{\varepsilon*} \leq j \right) + \mathbf{P}(X_T^{\varepsilon*} > j) \\ &\leq 2 \max \left[\mathbf{P} \left(\sup_{t \leq \delta} |X_{\tau+t}^\varepsilon - X_\tau^\varepsilon| > \eta, X_T^{\varepsilon*} \leq j \right), \mathbf{P}(X_T^{\varepsilon*} > j) \right]. \end{aligned}$$

Thence, due to proved above the first part of (4.2), the validity of the second part follows if

$$\lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \sup_{\tau \leq T-\delta} \mathbf{P} \left(\sup_{t \leq \delta} |X_{\tau+t}^\varepsilon - X_\tau^\varepsilon| > \eta, X_T^{\varepsilon*} \leq j \right) = -\infty, \quad j \geq 1, \eta > 0. \quad (4.9)$$

The simplest way for verifying (4.9) consists in checking the validity of both

$$\begin{aligned} \lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \sup_{\tau \leq T-\delta} \mathbf{P} \left(\sup_{t \leq \delta} \left| \int_{\tau}^{\tau+t} A(X_s^\varepsilon, \zeta_s^\varepsilon) ds \right| > \eta, X_T^{\varepsilon*} \leq j \right) &= -\infty, \\ \lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \sup_{\tau \leq T-\delta} \mathbf{P} \left(\sup_{t \leq \delta} \left| \sqrt{\varepsilon} \int_{\tau}^{\tau+t} B(X_s^\varepsilon, \zeta_s^\varepsilon) dW_s \right| > \eta, X_T^{\varepsilon*} \leq j \right) &= -\infty. \end{aligned} \quad (4.10)$$

Obviously, the first part in (4.10) holds. To verify the second, note that the process $Y_t^\varepsilon = \sqrt{\varepsilon} \int_{\tau}^{\tau+t} B(X_s^\varepsilon, \zeta_s^\varepsilon) dW_s$ is continuous martingale w.r.t. the new filtration $\mathbf{F}^\tau = (\mathcal{F}_{\tau+t})_{t \geq 0}$ (see Chap. 4, Sect. 7 in [23]). It has the predictable quadratic variation $\langle Y^\varepsilon \rangle_t = \varepsilon \int_{\tau}^{\tau+t} B^2(X_s^\varepsilon, \zeta_s^\varepsilon) ds$. Also define a positive continuous local martingale (w.r.t. the same filtration \mathbf{F}^τ)

$$Z_t^\varepsilon = \exp(\lambda Y_t^\varepsilon - \frac{1}{2} \lambda^2 \langle Y^\varepsilon \rangle_t), \quad \lambda \in R, \quad (4.11)$$

which is simultaneously a supermartingale (see [23], Problem 1.4.4) and so for any Markov time σ (w.r.t. \mathbf{F}^τ) $\mathbf{E}Z_\sigma^\varepsilon \leq 1$. Take $\sigma = \inf\{t \leq \delta: |Y_t^\varepsilon| \geq \eta\}$. Evidently the second part of (4.10) holds if

$$\lim_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \sup_{\tau \leq T-\delta} \mathbf{P}(Y_\sigma^\varepsilon \geq \eta \text{ (or } \leq -\eta), \sigma \leq \delta, X_T^{\varepsilon*} \leq j) = -\infty. \quad (4.12)$$

By virtue of an obvious inequality $\mathbf{E}Z_\sigma^\varepsilon I(Y_\sigma^\varepsilon \geq \eta, X_T^{\varepsilon*} \leq j) \leq 1$ we find that

$$\varepsilon \log \mathbf{P}(Y_\sigma^\varepsilon \geq \eta, \sigma \leq \delta, X_T^{\varepsilon*} \leq j) \leq -\sup_{\lambda > 0} \left[\lambda \eta - \text{const.} \frac{\lambda^2}{2} \delta \varepsilon \right] \quad (4.13)$$

and since

$$\sup_{\lambda > 0} \left[\lambda \eta - \text{const.} \frac{\lambda^2}{2} \delta \varepsilon \right] = \frac{\eta^2}{2 \text{const.} \delta \varepsilon}$$

(4.12) with “ $\geq \eta$ ” is implied by (4.13). The validity (4.12) with “ $\leq -\eta$ ” is proved in the same way.

Proof of Lemma 4.2. It is clear that $\{v^\varepsilon \in \mathbb{M}_{[0,T]} \setminus K_j''\} = \{\ell(j, v^\varepsilon) < \infty\}$, where

$$\ell(j, v) = \min \left\{ m \geq j: \int_0^T \int_{|z|>m} v(dt, dz) > g(m) \right\}. \quad (4.14)$$

Therefore, the second part of (4.1) is equivalent to

$$\lim_j \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(\ell(j, v^\varepsilon) < \infty) = -\infty. \quad (4.15)$$

To verify (4.15), choose a special function $g(y)$ satisfying the above-mentioned properties. To this end introduce non linear operator

$$\mathcal{D} = b(z) \frac{\partial}{\partial z} + \frac{\sigma^2(z)}{2} \left[\frac{\partial^2}{\partial z^2} + \left(\frac{\partial}{\partial z} \right)^2 \right] \quad (4.16)$$

and choose a non negative twice continuously differentiable function $u(z)$ such that

$$-\sup_{v \in R} \mathcal{D}u(v) = -d > -\infty, \quad (4.17)$$

$$\lim_{j \rightarrow \infty} \inf_{|z| > j} \left[-\mathcal{D}u(z) + \sup_{v \in R} \mathcal{D}u(v) \right] = \infty.$$

Under assumptions (A.2) and (A.3) one can take any of function $u(z)$ with properties: $u(0) = 0$, $u'(z) = \text{sign } z$, $|z| > 1$, and $0 \leq u''(z) \leq 1$. With chosen $u(z)$ put

$$g(y) = \inf_{|z| > y} \left[-\mathcal{D}u(z) + \sup_v \mathcal{D}u(v) \right]^{-1/2}. \quad (4.18)$$

Introduce a positive continuous local martingale (the martingale property is checked by Itô's formula)

$$Z_t^\varepsilon = \exp \left(u(\xi_t^\varepsilon) - u(\xi_0) - \int_0^t \mathcal{D}u(\xi_s^\varepsilon) ds \right). \quad (4.19)$$

It is simultaneously a supermartingale (see Problem 1.4.4. in [23]) and so $\mathbf{E}Z_T^\varepsilon \leq 1$. The last implies

$$\mathbf{E}I(\ell(j, v^\varepsilon) < \infty) Z_T^\varepsilon \leq 1. \quad (4.20)$$

Inequality (4.20) can be sharpened by changing of Z_T^ε on its lower bound on the set $\{\ell(j, v^\varepsilon) < \infty\}$ which can be chosen non random. Taking into account that $\int_0^T \mathcal{D}u(\xi_s^\varepsilon) ds = \int_0^T \int_R \mathcal{D}u(z) v^\varepsilon(ds, dz)$ and $\ell(j, v^\varepsilon) \geq j$ we arrive at

$$\begin{aligned} \log Z_T^\varepsilon &\geq -u(\xi_0) - \frac{dT}{\varepsilon} + \frac{1}{\varepsilon} \int_0^T \int_{|z| > \ell(j, v^\varepsilon)} [-\mathcal{D}u(z) + d] v^\varepsilon(ds, dz) \\ &\geq -u(\xi_0) - \frac{dT}{\varepsilon} + \frac{1}{\varepsilon} \inf_{|z| > \ell(j, v^\varepsilon)} [-\mathcal{D}u(z) + d] \int_0^T \int_{|z| > \ell(j, v^\varepsilon)} v^\varepsilon(ds, dz) \\ &\geq -u(\xi_0) - \frac{dT}{\varepsilon} + \frac{1}{\varepsilon} \inf_{|z| > \ell(j, v^\varepsilon)} [-\mathcal{D}u(z) + d]^{1/2} \\ &\geq -u(\xi_0) - \frac{dT}{\varepsilon} + \frac{1}{\varepsilon} \inf_{|z| > j} [-\mathcal{D}u(z) + d]^{1/2} (= \log Z_*). \end{aligned}$$

Thereby, from (4.20), with Z_T^ε repalced on Z_* , we derive

$$\varepsilon \log \mathbf{P}(\ell(j, v^\varepsilon) < \infty) \leq \varepsilon u(\xi_0) + dT - \inf_{|z| > j} [-\mathcal{D}u(z) + d]^{1/2},$$

i.e. (4.15) is implied by (4.17).

5 Upper bound for local LDP in $\mathbf{C}_{[0, T]} \times \mathbb{M}_{[0, T]}$

In this section, we consider family $(X^\varepsilon, v^\varepsilon)$ from $\mathbf{C}_{[0, T]} \times \mathbb{M}_{[0, T]}$. Parallel to $F(v)$ and $S(X, v)$, given in (2.3), let us define $F_T(v)$ and $S_T(X, v)$ by changing integrals “ \int_0^∞ ” in (2.3) on “ \int_0^T ”. Put

$$L_T(X, v) = \frac{1}{2}S_T(X, v) + \frac{1}{8}F_T(v). \quad (5.1)$$

Theorem 5.1 *Under (A.1), (A.2), and (A.3) for every (X, v) from $\mathbf{C}_{[0, T]} \times \mathbb{M}_{[0, T]}$*

$$\overline{\lim}_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \leq -L_T(X, v).$$

Proof of this theorem is based on

Lemma 5.1 *Assume (A.1), (A.2), and (A.3). Then for every piece wise constant function $\lambda(t) = \sum_i \lambda(t_i)I(t_i \leq t < t_{i+1})$ (with not overlapping intervals $[t_i, t_{i+1})$), and for every compactly supported in z and continuously differentiable (once in t and twice in z) function $u(t, z)$, and $X \in \mathbf{C}_{[0, T]}$, $v \in \mathbb{M}_{[0, T]}$*

$$\begin{aligned} & \overline{\lim}_{\delta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ & \leq - \left\{ \sum_i \lambda(t_i) [X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] - \int_0^T \int_R \lambda(t) A(X_t, z) v(dt, dz) \right. \\ & \quad \left. - \frac{1}{2} \int_0^T \int_R \lambda^2(t) B^2(X_t, z) v(dt, dz) \right\} + \int_0^T \int_R \mathcal{D}u(t, z) v(dt, dz), \end{aligned}$$

where \mathcal{D} is the non linear operator defined in (4.16).

Proof. The following well known fact will be used hereafter. If N_t ($N_0 = 0$) is continuous local martingale and $\langle N \rangle_t$ is its predictable quadratic variation, then the exponential process $Z_t = \exp(N_t - (1/2)\langle N \rangle_t)$ is a continuous local martingale too, and what is more if N'_t, N''_t are continuous local martingales ($N'_0 = N''_0 = 0$) with the mutual predictable quadratic variation $\langle N', N'' \rangle_t \equiv 0$ and Z'_t, Z''_t are corresponding exponential processes, then the process $Z'_t Z''_t$ is also local martingale which, being positive, is a supermartingale too (Problem 1.4.4 in [23]) and so $\mathbf{E}Z'_t Z''_t \leq 1$, $t \geq 0$.

Let $\lambda(t)$ and $u(t, z)$ be functions involved in Lemma 5.1. Put

$$\begin{aligned} N'_t &= \frac{1}{\sqrt{\varepsilon}} \int_0^t \lambda(s) B(X_s^\varepsilon, \zeta_s^\varepsilon) dW_s, \\ N''_t &= \frac{1}{\sqrt{\varepsilon}} \int_0^t u'_z(s, \zeta_s^\varepsilon) \sigma(\zeta_s^\varepsilon) dV_s. \end{aligned}$$

Evidently

$$\begin{aligned}\langle N' \rangle_t &= \frac{1}{\varepsilon} \int_0^t \lambda^2(s) B^2(X_s^\varepsilon, \xi_s^\varepsilon) ds, \\ \langle N'' \rangle_t &= \frac{1}{\varepsilon} \int_0^t u_z^2(s, \xi_s^\varepsilon) \sigma^2(\xi_s^\varepsilon) ds.\end{aligned}\tag{5.2}$$

Since Wiener processes W_t and V_t are independent and so $\langle N', N'' \rangle_t \equiv 0$ a process

$$Z_t = \exp(N'_t + N''_t - \frac{1}{2}[\langle N' \rangle_t + \langle N'' \rangle_t])\tag{5.3}$$

is local martingale and also a supermartingale with

$$\mathbf{E}Z_t \leq 1, \quad t \geq 0.\tag{5.4}$$

Note that

$$N'_t = \frac{1}{\varepsilon} \int_0^t \lambda(s) [dX_s^\varepsilon - A(X_s^\varepsilon, \xi_s^\varepsilon) ds]\tag{5.5}$$

and also find similar representation for N''_t . Due to Itô's formula we obtain

$$\frac{1}{\sqrt{\varepsilon}} \int_0^t u'_z(s, \xi_s^\varepsilon) \sigma(\xi_s^\varepsilon) dV_s = u(t, \xi_t^\varepsilon) - u(0, \xi_0) - \int_0^t u'_t(s, \xi_s^\varepsilon) ds - \frac{1}{\varepsilon} \int_0^t \mathcal{L}u(s, \xi_s^\varepsilon) ds,$$

where $\mathcal{L} = b(z)(\partial d/\partial z) + (\sigma^2(z)/2)(\partial^2/\partial z^2)$ and consequently

$$N''_t = u(t, \xi_t^\varepsilon) - u(0, \xi_0) + \int_0^t u'_t(s, \xi_s^\varepsilon) ds - \frac{1}{\varepsilon} \int_0^t \mathcal{L}u(s, \xi_s^\varepsilon) ds.\tag{5.6}$$

(5.4) implies an obvious inequality

$$\mathbf{E}I(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) Z_T \leq 1,\tag{5.7}$$

which can be sharpened by changing of Z_T by its lower bound. To this end evaluate from below $\log Z_T$ on the set $\{r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta\}$. For both $N'_T - \frac{1}{2}\langle N' \rangle_T$ and $N''_T - \frac{1}{2}\langle N'' \rangle_T$ we get

$$\begin{aligned}N'_T - \frac{1}{2}\langle N' \rangle_T &\geq \frac{1}{\varepsilon} \sum_i \lambda(t_i) [X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] \\ &\quad - \frac{1}{\varepsilon} \int_0^T \int_R \lambda(t) A(X_t, z) v(dt, dz) - \frac{1}{2} \int_0^T \int_R \lambda^2(t) B^2(X_t, z) v(dt, dz) \\ &\quad - \frac{1}{\varepsilon} \left\{ \sum_i |\lambda(t_i)| [|X_{T \wedge t_{i+1}}^\varepsilon - X_{T \wedge t_{i+1}}| + |X_{T \wedge t_i}^\varepsilon - X_{T \wedge t_i}|] \right. \\ &\quad \left. + \int_0^T |\lambda(t)| |A(X_t^\varepsilon, \xi_t^\varepsilon) - A(X_t, \xi_t^\varepsilon)| ds + \frac{1}{2} \int_0^T \lambda^2(t) |B^2(X_t^\varepsilon, \xi_t^\varepsilon) - B^2(X_t, \xi_t^\varepsilon)| ds \right. \\ &\quad \left. + \left| \int_0^T \int_R [\lambda(t) A(X_t, z) + \lambda^2(t) B^2(X_t, z)] [v^\varepsilon - v](dz, dt) \right| \right\}\end{aligned}\tag{5.8}$$

and

$$\begin{aligned}
& N_T'' - \frac{1}{2} \langle N'' \rangle_T \\
&= -\frac{1}{\varepsilon} \int_0^T [\mathcal{L}u(s, \xi_s^\varepsilon) + \frac{1}{2} u_z^2(s, \xi_s^\varepsilon)] ds + u(T, \xi_T^\varepsilon) - u(0, \xi_0) - \int_0^T u_t(s, \xi_s^\varepsilon) ds \\
&= -\frac{1}{\varepsilon} \int_0^T \int_R \mathcal{D}u(s, z) v^\varepsilon(dz, ds) + u(T, \xi_T^\varepsilon) - u(0, \xi_0) - \int_0^T u_t(s, \xi_s^\varepsilon) ds \\
&\geq -\frac{1}{\varepsilon} \int_0^T \int_R \mathcal{D}u(s, z) v(dz, ds) \\
&\quad - \frac{1}{\varepsilon} \left\{ \left| \int_0^T \int_R \mathcal{D}u(s, z) [v^\varepsilon - v](dz, ds) \right| \right. \\
&\quad \left. + \varepsilon |u(T, \xi_T^\varepsilon)| + \varepsilon |u(0, \xi_0)| + \varepsilon \int_0^T |u_t(s, \xi_s^\varepsilon)| ds \right\}. \tag{5.9}
\end{aligned}$$

The terms in the curly brackets in the right hand sides of (5.8) and (5.9) are random variables. Nevertheless, they can be evaluated from above on the set $\{r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta\}$ by non random quantities. Evidently $\int_0^T \int_R |\lambda(t)| |A(X_t^\varepsilon, \xi_t^\varepsilon) - A(X_t, \xi_t^\varepsilon)| ds \leq \text{const.} T \delta$ and $\frac{1}{2} \int_0^T \lambda^2(t) |B^2(X_t^\varepsilon, \xi_t^\varepsilon) - B^2(X_t, \xi_t^\varepsilon)| ds \leq \text{const.} \delta \int_0^T [1 + |X_t|] ds$. Denote by $H(s, z) = \lambda(s)A(X_s, z) + \lambda^2(s)B^2(X_s, z)/2$. Since $\lambda(s)$ is piece wise constant function without loss of a generality one can assume that it is simply constant. Then function $H(s, z)$ is bounded continuous function and so, by Lemma A.1 (see Appendix) for any $\gamma > 0$ and $k \geq 1$ there exist increasing continuous function $h_k^\gamma(y)$, $y \geq 0$ with $h_k^\gamma(0) = 0$ and decreasing sequence φ_k , $k \geq 1$ with $\lim_k \varphi_k = 0$ both dependent on $H(s, z)$ and v only such that

$$\left| \int_0^T \int_R H(s, z) [v^\varepsilon - v](ds, dz) \right| \leq \gamma + h_k^\gamma(\delta) + \varphi_k.$$

Further, by the remark to Lemma A.1

$$\left| \int_0^T \int_R \mathcal{D}u(t, z) [v^\varepsilon - v](dt, dz) \right| \leq \gamma + h^\gamma(\delta),$$

where $h^\gamma(y)$ is an increasing continuous function with $h^\gamma(0) = 0$ depending on $\mathcal{D}u(s, z)$ and v only.

Hence, we arrive to the lower bounds (with positive const.'s):

$$\begin{aligned}
N_T' - \frac{1}{2} \langle N' \rangle_T &\geq \frac{1}{\varepsilon} \left[\sum_i \lambda(t_i) [X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] - \int_0^T \int_R \lambda(t) A(X_t, z) v(dt, dz) \right. \\
&\quad \left. - \int_0^T \int_R \lambda^2(t) B^2(X_t, z) v(dt, dz) \right] - \frac{\text{const.}}{\varepsilon} (\gamma + h_k^\gamma(\delta) + \varphi_k) \tag{5.10}
\end{aligned}$$

and

$$N_T'' - \frac{1}{2}\langle N'' \rangle_T \geq -\frac{1}{\varepsilon} \int_0^T \int_R \mathcal{D}u(t, z)v(dt, dz) - \frac{\text{const.}}{\varepsilon}(\varepsilon + \gamma + h^\gamma(\delta) + \varphi_k). \quad (5.11)$$

By virtue of (5.10) and (5.11) one can choose a non random lower bound:

$$\begin{aligned} \log Z_T &\geq \frac{1}{\varepsilon} \left[\sum_i \lambda(t_i)[X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] - \int_0^T \int_R \lambda(t)A(X_t, z)v(dt, dz) \right. \\ &\quad \left. - \int_0^T \int_R \lambda^2(t)B^2(X_t, z)v(dt, dz) - \frac{1}{\varepsilon} \int_0^T \int_R \mathcal{D}u(t, z)v(dt, dz) \right] \\ &\quad - \frac{\text{const.}}{\varepsilon}(\varepsilon + \gamma + h^\gamma(\delta) + h_k^\gamma(\delta) + \varphi_k). \\ &= \log Z_*. \end{aligned}$$

Hence and from (5.7), with replacing of Z_T on Z_* , it follows

$$\begin{aligned} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho(v^\varepsilon, v) \leq \delta) \\ &\leq - \left[\sum_i \lambda(t_i)[X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] - \int_0^T \int_R \lambda(t)A(X_t, z)v(dt, dz) \right. \\ &\quad \left. - \frac{1}{2} \int_0^T \int_R \lambda^2(t)B^2(X_t, z)v(dt, dz) \right] + \int_0^T \int_R \mathcal{D}u(t, z)v(dt, dz) \\ &\quad + \text{const.}\{(\varepsilon + \gamma + h^\gamma(\delta) + h_k^\gamma(\delta) + \varphi_k)\}. \quad (5.12) \end{aligned}$$

The desired result holds since the term in the curly brackets of the right hand side in (5.12) goes to zero if limit “ $\lim_{k \rightarrow \infty} \lim_{\gamma \rightarrow 0} \lim_{\delta \rightarrow 0} \lim_{\varepsilon \rightarrow 0}$ ” is taken.

Proof of Theorem 5.1. Follows from Lemmas 5.1, A.2, and A.3 (see Appendix) since

$$\begin{aligned} & - \sup_\lambda \left\{ \sum_i \lambda(t_i)[X_{T \wedge t_{i+1}} - X_{T \wedge t_i}] - \int_0^T \int_R \lambda(t)A(X_t, z)v(dt, dz) \right. \\ & \left. - \frac{1}{2} \int_0^T \int_R \lambda^2(t)B^2(X_t, z)v(dt, dz) \right\} + \inf_u \int_0^T \int_R \mathcal{D}u(t, z)v(dt, dz) \\ & = -[\frac{1}{2}S_T(X, v) + \frac{1}{8}F_T(v)] = -L_T(X, v). \end{aligned}$$

6 Lower bound for local LDP in $\mathbf{C}_{[0, T]} \times \mathbf{M}_{[0, T]}$

Theorem 6.1 *Under (A.1), (A.2), and (A.3), for every (X, v) from $\mathbf{C}_{[0, T]} \times \mathbf{M}_{[0, T]}$*

$$\lim_{\delta \rightarrow 0} \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \geq -L_T(X, v).$$

Evidently for X, v such that $L_T(X, v) = \infty$ it is nothing to prove. Therefore we consider below only the case $L_T(X, v) < \infty$ which distinguishes subsets from $\mathbf{C}_{[0, T]} \times \mathbf{M}_{[0, T]}$:

- (i) $dX_t \ll dt$ and $\frac{1}{2} \mathcal{S}_T(X, v) = \frac{1}{2} \int_0^T ([\dot{X}_t - A_v(t, X_t)]^2 / B_v^2(t, X_t)) dt < \infty$;
(ii) $dv = n d\lambda$, $d_z n = n'_z dz$ and $\frac{1}{8} F_T(v) = \frac{1}{2} \int_0^T \int_R (v_v^2(t, z) / \sigma^2(z)) n(t, z) dz dt < \infty$,
where

$$v_v(t, z) = \frac{\sigma^2(z)}{2} \left[\frac{n'_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]. \quad (6.1)$$

It is convenient to consider further subset (ii') of (ii):

(ii') The function $v_v(t, z)$ is compactly supported in z and continuously differentiable in (t, z) , having bounded partial derivatives.

The central role in proving Theorem 6.1 plays

Lemma 6.1 *Assume (i), (ii'), and $\inf_{x, z} B^2(x, z) > 0$. Then for any $\delta > 0$ and $\gamma > 0$ there exists an increasing continuous function $h_\gamma(y)$ with $h_\gamma(0) = 0$, depending on $v_v^2(s, z) / \sigma^2(z)$ and v only, such that*

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \geq -L_T(X, v) - \gamma - h_\gamma(\delta).$$

Proof. Put

$$\begin{aligned} b_v(t, z) &= b(z) + v_v(t, z), \\ G_v(t, x, z) &= \frac{\dot{X}_t - A_v(t, X_t)}{B_v(t, X_t)} B(x, z) + A(x, z) \end{aligned} \quad (6.2)$$

and parallel to $(X_t^\varepsilon, \xi_t^\varepsilon)$ introduce, on the same stochastic basis, new diffusion pair $(\tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon)$:

$$\begin{aligned} d\tilde{X}_t^\varepsilon &= G_v(t, \tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon) dt + \sqrt{\varepsilon} B(\tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon) dW_t, \\ d\tilde{\xi}_t^\varepsilon &= \frac{1}{\varepsilon} b_v(t, \tilde{\xi}_t^\varepsilon) dt + \frac{1}{\sqrt{\varepsilon}} \sigma(\tilde{\xi}_t^\varepsilon) dV_t \end{aligned} \quad (6.3)$$

subject to the same initial point (x_0, ξ_0) . Also denote by $\tilde{v}^\varepsilon(dt, dz)$ the occupation measure corresponding to $\tilde{\xi}^\varepsilon$: $\tilde{v}^\varepsilon(\Delta \times \Gamma) = \int_0^\infty I(t \in \Delta, \tilde{\xi}_t^\varepsilon \in \Gamma) dt$.

By virtue of the formula $b(z) = \frac{1}{2}(p'(z)/p(z)) + \sigma'(z)\sigma(z)$ (see (2.1)) we get

$$\frac{2b_v(t, z)}{\sigma^2(z)} = \frac{n'_z(t, z)}{n(t, z)} + 2 \frac{\sigma'(z)}{\sigma(z)}$$

and so

$$p_v(t, z) = c(t) \frac{\exp(2 \int_0^z \frac{b_v(t, y)}{\sigma^2(y)} dy)}{\sigma^2(z)},$$

with norming constant $c(t)$ such that $\int_R p_v(t, z) = 1$, coincides with $n(t, z)$. Then by Lemma A.5 (see Appendix)

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \rho_T(\tilde{v}^\varepsilon, v) = 0 \quad \text{and} \quad \mathbf{P} - \lim_{\varepsilon \rightarrow 0} r_T(\tilde{X}^\varepsilon, X) = 0. \quad (6.4)$$

Denote by Q^ε and \tilde{Q}^ε distributions of $(X_t^\varepsilon, \xi_t^\varepsilon)_{t \leq T}$, $(\tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon)_{t \leq T}$ respectively. By Theorem 7.18 (Chap. 7 in [24]) Q^ε is absolutely continuous w.r.t. \tilde{Q}^ε and

$$\frac{dQ^\varepsilon}{d\tilde{Q}^\varepsilon}(\tilde{X}^\varepsilon, \tilde{\xi}^\varepsilon) = \exp\left(\frac{1}{\sqrt{\varepsilon}} M_T^\varepsilon - \frac{1}{2\varepsilon} \langle M^\varepsilon \rangle_T + \frac{1}{\sqrt{\varepsilon}} M_T - \frac{1}{2\varepsilon} \langle M \rangle_T\right), \quad (6.5)$$

where

$$M_t^\varepsilon = -\int_0^t \frac{v_v(s, \tilde{\xi}_s^\varepsilon)}{\sigma(\tilde{\xi}_s^\varepsilon)} dV_s \quad \text{and} \quad M_t = -\int_0^t \frac{\dot{X}_s - A_v(s, X_s)}{B_v(s, X_s)} dW_s,$$

$$\langle M^\varepsilon \rangle_t = \int_0^t \frac{v_v^2(s, \tilde{\xi}_s^\varepsilon)}{\sigma^2(\tilde{\xi}_s^\varepsilon)} ds \quad \text{and} \quad \langle M \rangle_t = \int_0^t \frac{[\dot{X}_s - A_v(s, X_s)]^2}{B_v^2(s, X_s)} ds.$$

By virtue of (6.5) and the rule of changing for probability measure we obtain

$$\mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) = \mathbf{E}\left[\frac{dQ^\varepsilon}{d\tilde{Q}^\varepsilon}(\tilde{X}^\varepsilon, \tilde{\xi}^\varepsilon) I(r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta)\right]. \quad (6.6)$$

The desired lower bound can be derived from (6.6) provided that a relevant lower bound for the right hand side of (6.6) can be found. Use an obvious inequality:

$$I(r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta) \geq I(r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta, |M_T^\varepsilon| \leq k, |M_T| \leq k)$$

and estimate from below $\log(dQ^\varepsilon/d\tilde{Q}^\varepsilon)(\tilde{X}^\varepsilon, \tilde{\xi}^\varepsilon)$ on the set $\{r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta\} \cap \{|M_T^\varepsilon| \leq k\} \cap \{|M_T| \leq k\}$. Noticing that $\frac{1}{2}\langle M \rangle_T = \frac{1}{2}S(X, v)_T$ and $\frac{1}{2}\int_0^T \int_R (v_v^2(s, z)/\sigma^2(z))n(t, z) dz dt = \frac{1}{8}F_T(v)$ we obtain

$$\log \frac{dQ^\varepsilon}{d\tilde{Q}^\varepsilon}(\tilde{X}^\varepsilon, \tilde{\xi}^\varepsilon) \geq -\frac{2k}{\sqrt{\varepsilon}} - \frac{1}{\varepsilon} L_T(X, v) - \frac{1}{2\varepsilon} \left| \int_0^T \int_R \frac{v_v^2(s, z)}{\sigma^2(z)} m[\tilde{v}^\varepsilon(dt, dz) - n(t, z) dz dt] \right|.$$

By the remark to Lemma A.1 (see Appendix), for any $\gamma > 0$ there exists increasing continuous function $h_\gamma(y)$ with $h_\gamma(0) = 0$, depending on $v_v^2(s, z)/\sigma^2(z)$ and v only, such that

$$\frac{1}{2} \left| \int_0^T \int_R \frac{v_v^2(s, z)}{\sigma^2(z)} [\tilde{v}^\varepsilon(dt, dz) - n(t, z) dz dt] \right| \leq \gamma + h_\gamma(\delta).$$

Then the lower bound for the right hand side of (6.6) is the following:
 $-(2k/\sqrt{\varepsilon}) - (1/\varepsilon)[L_T(X, v) + \gamma + h_\gamma(\delta)]$. It implies

$$\begin{aligned} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ \geq -L_T(X, v) - 2k\sqrt{\varepsilon} - \gamma - h_\gamma(\delta) + \varepsilon \log \mathbf{P}(r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta, \\ |M_T^\varepsilon| \leq k, |M_T| \leq k). \end{aligned}$$

Thus the statement of the lemma holds since

$$\lim_k \lim_{\varepsilon \rightarrow 0} \mathbf{P}(r_T(\tilde{X}^\varepsilon, X) + \rho_T(\tilde{v}^\varepsilon, v) \leq \delta, |M_T^\varepsilon| \leq k, |M_T| \leq k) = 1$$

what follows by virtue of (6.4), obvious $\lim_k \mathbf{P}(|M_T| > k) = 0$ and

$$\mathbf{P}(|M_T^\varepsilon| > k) \leq \frac{\mathbf{E}|M_T^\varepsilon|^2}{k^2} = \frac{\mathbf{E}\langle M^\varepsilon \rangle_T}{k^2} \leq \frac{\text{const.}}{k^2} \rightarrow 0, \quad k \rightarrow \infty.$$

Proof of Theorem 6.1. Assume (i), (ii), and $\inf_{x,z} B^2(x, z) > 0$. Due to Lemma A.4 (see Appendix), one can choose a sequence $v^{(k)}$, $k \geq 1$ of measures such that for every k the function $v_{v^{(k)}}(t, z)$ satisfies (ii') and what is more $\rho(v, v^{(k)}) \rightarrow 0$, $L_T(X, v^{(k)}) \rightarrow L_T(X, v)$. On the other hand, by Lemma 6.1 for any $\delta > 0$ and $\gamma > 0$ there exist increasing continuous function $h_{\gamma, k}(y)$ with $h_{\gamma, k}(0) = 0$, depending on $v^{(k)}$, such that

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v^{(k)}) \leq \delta) \geq -L_T(X, v^{(k)}) - \gamma - h_{\gamma, k}(\delta).$$

Choose $k_0(\delta)$ such that for any $k \geq k_0(\delta)$ we have $0 < \delta - \rho_T(v, v^{(k)}) \leq \delta/2$. Then, taking into account the triangular inequality: $\rho_T(v^\varepsilon, v) \leq \rho_T(v^\varepsilon, v^{(k)}) + \rho_T(v, v^{(k)})$, we arrive to a lower bound:

$$\begin{aligned} \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ \geq \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v^{(k)}) \leq \delta/2) \\ \geq -L_T(X, v^{(k)}) - \gamma - h_{\gamma, k}(\delta/2). \end{aligned}$$

The right hand side of the last inequality converges to $-L_T(X, v)$ if limit “ $\lim_k \lim_{\gamma \rightarrow 0} \lim_{\delta \rightarrow 0}$ ” is taken.

Assume only (i) and (ii). Parallel to the process X_t^ε introduce new diffusion $X_t^{\varepsilon, \beta}$, $\beta \neq 0$:

$$dX_t^{\varepsilon, \beta} = A(X_t^{\varepsilon, \beta}, \zeta_t^\varepsilon) dt + \sqrt{\varepsilon}[B(X_t^{\varepsilon, \beta}, \zeta_t^\varepsilon) dW_t + \beta dW_t']$$

subject to the same initial point x_0 , where W_t' is a Wiener process independent of $(W_t, \zeta_t^\varepsilon)$. The diffusion parameter here is $B^2(x, z) + \beta^2$ and so, due to be proved above,

$$\lim_{\delta \rightarrow 0} \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^{\varepsilon, \beta}, X) + \rho_T(v^\varepsilon, v) \leq \delta) \geq -L_T^\beta(X, v),$$

where $L_T^\beta(X, v) = \frac{1}{2}S_T^\beta(X, v) + \frac{1}{8}F_T(v)$, and where

$$S_T^\beta(X, v) = \int_0^\infty \frac{[\dot{X}_t - A_v(v, X_t)]^2}{B_v^2(t, X_t) + \beta^2} dt .$$

Evidently $\lim_{\beta \rightarrow 0} S_T^\beta(X, v) = S_T(X, v)$. On the other hand, by Lemma A.6 (see Appendix)

$$\lim_{\beta \rightarrow 0} \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^{\varepsilon, \beta}, X^\varepsilon) > \eta) = -\infty .$$

To get the desired result, we combine both these facts. Namely, using the triangular inequality: $r_T(X^\varepsilon, X) \leq r_T(X^{\varepsilon, \beta}, X^\varepsilon) + r_T(X^{\varepsilon, \beta}, X)$ and taking $\eta = \delta/2$ we arrive to an upper bound

$$\begin{aligned} & \mathbf{P}(r_T(X^{\varepsilon, \beta}, X) + \rho_T(v^\varepsilon, v) \leq \delta) \\ & \leq \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta/2) + \mathbf{P}(r_T(X^{\varepsilon, \beta}, X^\varepsilon) > \delta/2) \\ & \leq 2 \max[\mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta/2), \mathbf{P}(r_T(X^{\varepsilon, \beta}, X^\varepsilon) > \delta/2)] , \end{aligned}$$

which implies

$$\lim_{\delta \rightarrow 0} \lim_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^\varepsilon, X) + \rho_T(v^\varepsilon, v) \leq \delta) \geq -\lim_{\beta \rightarrow 0} L_T^\beta(X, v) = -L_T(X, v) .$$

Other approach for establishing lower bound with singular diffusion parameter can be found in Puhalskii [25].

7 Proof of main result

Proof of Theorem 3.1. Due to Theorems 4.1, 5.1, and Proposition 3.1 the family $(X^\varepsilon, v^\varepsilon)$ obeys the LDP in $(\mathbb{C}_{[0, k]} \times \mathbb{M}_{[0, k]}, r_k \times \rho_k)$ with rate function $L_k(X, v)$. Then it obeys the LDP in the metric space $(\mathbb{C} \times \mathbb{M}, r \times \rho)$ with rate function $\sup_k L_k(X, v) = L(X, v)$.

Proof of Corollary 2.1. The result holds since $\inf_{X \in \mathbb{C}} S(X, v)$ is attained at X_t^0 , being a solution of a differential equation: $\dot{X}_t = A_v(t, X_t^0)$ subject to $X_0^0 = x_0$, and so $S(X^0, v) = 0$.

Proof of Corollary 2.2. The first statement is obvious.

Assume $B^2(x, z) \equiv 0$. In this case $S(X, v) = 0$ for any X_t being a solution of a differential equation $\dot{X}_t = \int_R A(X_t, z)n(t, z) dz$ subject to $X_0 = x_0$; otherwise $S(X, v) = \infty$. Therefore

$$S(X) = \begin{cases} \frac{1}{8} \inf_{v: \dot{X}_t = \int_R A(X_t, z)n(t, z) dz, X_0 = x_0} F(v) \\ \infty, \end{cases} \quad \text{otherwise .}$$

On the other hand, since $F(v) < \infty$ implies $dv = n d\lambda$, $d_z n = n'_z dz$, assuming measurability in t of function

$$H(t, \dot{X}_t, X_t) = \inf_{\nu: \dot{X}_t = \int_R A(X_t, z) m(t, z) dz, X_0 = x_0} \int_R \sigma^2(z) \left[\frac{n'_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]^2 n(t, z) dz \quad (7.1)$$

we arrive at independent of t function $H(t, y, x) \equiv H(y, x)$, or by other words, “inf” in (7.1) can be taken over all measures ν with densities $n(t, z) \equiv m(z)$. The last means the desired result holds if the function

$$H(y, x) = \inf_{\substack{m: \\ y = \int_R A(x, z) m(z) dz}} \int_R \sigma^2(z) \left[\frac{m'(z)}{m(z)} - \frac{p'(z)}{p(z)} \right]^2 m(z) dz \quad (7.2)$$

is measurable. We check this by showing that level sets of $H(y, x)$ are closed.

Let $c \geq 0$ be fixed and (y_n, x_n) , $n \geq 1$ be a sequence from $\{(y, x) : H(y, x) \leq c\}$ converging to a limit point (y_0, x_0) . Show that $H(y_0, x_0) \leq c$. By virtue of assumption (A.1) the set $\mathcal{A}(y, x) = \{m : y = \int_R A(x, z) m(z) dz\}$ is closed in the Levy–Prohorov metric that is for every fixed (y, x) there exists a density $m^{(y, x)}$ from $\mathcal{A}(y, x)$ such that

$$H(y, x) = \begin{cases} \int_R \sigma^2(z) \left[\frac{(m^{(y, x)}(z))'}{m^{(y, x)}(z)} - \frac{p'(z)}{p(z)} \right]^2 m^{(y, x)}(z) dz, & dm^{(y, x)} = (m^{(y, x)})' dz, \\ \infty, & \text{otherwise.} \end{cases} \quad (7.3)$$

Note that the function $H(y, x)$, defined in (7.3), obeys a following property: there exists a measure $\nu^{(y, x)}$ from $\mathbb{M}_{[0, 1]}$, having density $m^{(y, x)}(z)$ w.r.t. $dt dz$, such that $H(y, x) = F_1(\nu^{(y, x)})$. Since $\frac{1}{8}F_1(\nu)$ is good rate function level sets $\{y, x : H(y, x) \leq c\}$ are compacts. Therefore $H(y_0, x_0) \leq c$.

Appendix

1. Evaluation via Levy–Prohorov’s metric

Lemma A.1 *Let $T > 0$, $\nu', \nu'' \in \mathbb{M}_{[0, T]}$, $\rho_T(\nu', \nu'') = q$, and $f(t, z)$ be bounded continuous function. Then for any $\gamma > 0$ and $k \geq 1$ one can choose increasing continuous function $h_k^\gamma(y)$, $y \geq 0$ with $h_k^\gamma(0) = 0$ and decreasing sequence φ_k , $k \geq 1$ with $\lim_k \varphi_k = 0$ both depending on $f(t, z)$ and only from one of ν' or ν'' such that*

$$\left| \int_0^T \int_R f(t, z) [\nu' - \nu''](dt, dz) \right| \leq \gamma + h_k^\gamma(q) + \varphi_k.$$

Remark. If $f(t, z)$ is bounded compactly supported continuous function, then the statement of the lemma remains true with $h_k^\gamma(y) \equiv h^\gamma(y)$ and $\varphi_k \equiv 0$.

Proof. Assume $f(t, z)$ is continuously differentiable (one in z and twice in (t, z)) and compactly supported in z . Denote by $F'(t, z) = v'([0, t] \times (-\infty, z])$ that is $F'(t, z)$ is the distribution function corresponding to v' . Integrating by parts we get

$$\int_0^T \int_R f(t, z) v'(dt, dz) = - \int_0^T \int_R \left[\frac{\partial f(t, z)}{\partial z} + \frac{\partial^2 f(t, z)}{\partial t \partial z} \right] F'(t, z) dz dt$$

and consequently (F'' is the distribution functions corresponding to v'')

$$\left| \int_0^T \int_R f(t, z) [v' - v''](dt, dz) \right| \leq \int_0^T \int_R |F'(t, z) - F''(t, z)| m(t, z) dz dt,$$

where $m(t, z) = |(\partial/\partial z)f(t, z)| + |(\partial^2/\partial t \partial z)f(t, z)|$.

Assume $f(t, z)$ is compactly supported in z and continuous only. Then, approximating it by compactly supported and continuously differentiable in z function $f^\gamma(t, z)$ in a sense $\sup_{t, z} |f(t, z) - f^\gamma(t, z)| \leq \gamma/2T$, due to the foregoing proof, we get

$$\left| \int_0^T \int_R f(t, z) [v' - v''](dt, dz) \right| \leq \gamma + \int_0^T \int_R |F'(t, z) - F''(t, z)| m^\gamma(t, z) dz dt$$

with $m^\gamma(t, z) = |(\partial/\partial z)f^\gamma(t, z)| + |(\partial^2/\partial t \partial z)f^\gamma(t, z)|$.

In the general case, one can choose a decomposition $f(t, z) = f_k(t, z) + g_k(t, z)$, where $f_k(t, z)$ is continuous compactly supported in z on the interval $[-k, k]$ function while $g_k(t, z) \equiv 0$ on the interval $[-(k - 1/2), (k - 1/2)]$ and is bounded: $|g_k(t, z)| \leq L$. Then by foregoing result we get

$$\begin{aligned} \left| \int_0^T \int_R f(t, z) [v' - v''](dt, dz) \right| &\leq \gamma + \int_0^T \int_R |F'(t, z) - F''(t, z)| m_k^\gamma(t, z) dz dt \\ &\quad + L \int_0^T \int_{|z| > k-1/2} [v' + v''](dt, dz), \end{aligned}$$

where $m_k^\gamma(t, z) = |(\partial/\partial z)f_k^\gamma(t, z)| + |(\partial^2/\partial t \partial z)f_k^\gamma(t, z)|$. Evaluate from above the last integral from the right hand side. To this end, choose an increasing sequences $z_k \nearrow \infty$, $k \rightarrow \infty$ such that $z_k \leq k - 1/2$ and for every k z_k and $-z_k$ are points of continuity for the distribution function $F'(T, z)$. Then

$$\begin{aligned} \int_0^T \int_{|z| > k-1/2} [v' + v''](dt, dz) &\leq 2 \int_0^T \int_{|z| > z_k} v'(dt, dz) + \left| \int_0^T \int_{|z| > z_k} [v' - v''](dt, dz) \right| \\ &\leq 2 \int_0^T \int_{|z| > z_k} v'(dt, dz) + |F'(T, z_k) - F''(T, z_k)| \\ &\quad + |F'(T, -z_k) - F''(T, -z_k)|. \end{aligned}$$

Now evaluate from above $|F'(t, z) - F''(t, z)|$ via q and $F'(t, z)$. From the definition of the Levy–Prohorov metric (see e.g. [22, 26]) it follows: $q + F'(t - q, z - q) - F'(t, z) \leq F'(t, z) - F''(t, z) \leq q + F'(t + q, z + q) - F'(t, z)$ and so

$$|F'(t, z) - F''(t, z)| \leq q + [F'(t + q, z + q) - F'(t - q, z - q)].$$

Hence, combining all obtained upper estimates, we arrive at the desired result with

$$\begin{aligned} h_k^\gamma(y) &= y \left(2L + \int_0^T \int_R m_k^\gamma(t, z) dt dz \right) \\ &+ \int_0^T \int_R [F'(t + y, z + y) - F'(t - y, z - y)] m_k^\gamma(t, z) dz dt \\ &+ L |F'(T + y, z_k + y) - F'(T - y, z_k - y)| \\ &+ L |F'(T + y, -z_k + y) - F'(T - y, -z_k - y)| \end{aligned}$$

and

$$\varphi_k = 2L \int_0^T \int_{|z| > z_k} v'(dt, dz).$$

The same proof takes place with F'' instead of F' .

2. The Fenchel–Legendre transform

Let $\lambda(t) = \sum_i \lambda(t_i) I(t_i \leq t < t_{i+1})$ with non overlapping intervals $[t_i, t_{i+1})$. For any $X \in \mathbf{C}_{[0, T]}$ and $v \in \mathbf{M}_{[0, T]}$ put $\int_0^T \lambda(t) dX_t = \sum_i \lambda(t_i) [X_{T \wedge t_{i+1}} - X_{T \wedge t_i}]$, $A_v(t, X_t) = \int_R A(X_t, z) K_v(t, dz)$, and $B_v^2(t, X_t) = \int_R B^2(X_t, z) K_v(t, dz)$. Let \mathcal{D} be non linear operator defined in (4.16).

Lemma A.2 For any $X \in \mathbf{C}_{[0, T]}$ and $v \in \mathbf{M}_{[0, T]}$

$$\begin{aligned} &\sup \int_0^T [\lambda(t) dX_t - (A_v(t, X_t) - \frac{1}{2} \lambda^2(t) B_v^2(t, X_t))] dt \\ &= \begin{cases} \frac{1}{2} \int_0^T \frac{[\dot{X}_t - A_v(t, X_t)]^2}{B_v^2(t, X_t)} dt, & dX_t = \dot{X}_t dt, \\ \infty, & \text{otherwise,} \end{cases} \end{aligned}$$

where “sup” is taken over all piece wise constant functions $\lambda(t)$.

Lemma A.3 For any $v \in \mathbf{M}_{[0, T]}$

$$\begin{aligned} &\inf \int_0^T \int_R \mathcal{D}u(t, z) v(dt, dz) \\ &= \begin{cases} -\frac{1}{8} \int_0^T \int_R \sigma^2(z) \left[\frac{n'_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]^2 n(t, z) dz dt, & dv = n d\lambda, \quad d_z n = n'_z dz, \\ -\infty, & \text{otherwise,} \end{cases} \end{aligned}$$

where “inf” is taken over all continuously differentiable (once in t and twice in z) compactly supported in z functions $u(t, z)$.

Proof of Lemma A.2. For $dX_t \lll dt$ the result follows from Lemma 6.1 in [17] (see also Lemma 2.1 in [27]). For $dX_t = \dot{X}_t dt$ by Lemma 6.1 [17] “sup \int_0^T ” is equal $\int_0^T \sup_{\lambda \in R} \{ \lambda(\dot{X}_t - A_v(t, X_t)) - \frac{1}{2} \lambda^2 B_v^2(t, X_t) \} dt = \frac{1}{2} \int_0^T ([\dot{X}_t - A_v(t, X_t)] / B_v^2(t, X_t)) dt$.

Proof of Lemma A.3. Assume $dv = n d\lambda$, $d_z n = n'_z dz$. Due to (2.1), $p'(z)/p(z) = (2b(z) - 2\sigma(z)\sigma'(z))/\sigma^2(z)$ and so $b(z) = \frac{1}{2}[\sigma^2(z)(p'(z)/p(z)) + 2\sigma(z)\sigma'(z)]$. Putting $v(t, z) = u'_z(t, z)$ and taking into account the formula for $b(z)$ we arrive at

$$\begin{aligned} \int_0^T \int_R \mathcal{D}u(t, z)n(t, z) dz dt &= \frac{1}{2} \int_0^T \int_R \left\{ \left[\sigma^2(z) \frac{p'(z)}{p(z)} + 2\sigma(z)\sigma'(z) \right] v(t, z) \right. \\ &\quad \left. + \sigma^2(z)(v'_z(t, z) + v^2(t, z)) \right\} n(t, z) dz dt . \end{aligned} \quad (8.1)$$

Then, integrating by parts,

$$\int_R \sigma^2(z)(v'_z(t, z)n(t, z) dz = - \int_R v(t, z)[2\sigma(z)\sigma'(z)n(t, z) + \sigma^2(z)n'_z(t, z)] dz ,$$

we obtain

$$\begin{aligned} &\int_0^T \int_R \mathcal{D}u(t, z)n(t, z) dz dt \\ &= \frac{1}{2} \int_0^T \int_R \sigma^2(z) \left(v^2(t, z)n(t, z) + v(t, z) \left[\frac{p'(z)}{p(z)} n(t, z) - n_z(t, z) \right] \right) dz dt . \end{aligned} \quad (8.2)$$

(8.2) and the method of proving for lemma 6.1 in [17] imply

$$\begin{aligned} &\inf \int_0^T \int_R \mathcal{D}u(t, z)n(t, z) dz dt \\ &= \frac{1}{2} \int_0^T \int_R \sigma^2(z) \inf_{v \in R} \left(v^2 n(t, z) + v \left[\frac{p'(z)}{p(z)} n(t, z) - n_z(t, z) \right] \right) dz dt \\ &= -\frac{1}{8} \int_0^T \int_R \sigma^2(z) \left[\frac{n_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]^2 n(t, z) dz dt . \end{aligned}$$

Thus for “ $dv = n d\lambda, d_z n = n'_z dz$ ”, the result holds.

Assume $dv = n\lambda, d_z n \lll dz$. Show that $\inf \int_0^T \int_R \mathcal{D}u(t, z)n(t, z) dz dt = -\infty$. To this end, take $u(t, z) \equiv u(z)$ and put $v(z) = u'(z)$. The function $v(z)$ is compactly supported and continuously differentiable and, in particular, has the finite total variation. Put $n(z) = \int_0^T n(t, z) dt$ and $w(z) = \frac{1}{2} \sigma^2(z)n(z)$. It is

clear that there exists a positive constant, say, ℓ such that $I(v) = \ell \int_R [v^2(z) + |v(z)|]n(z) dz + \int_R w(z) dv(z)$ is an upper bound for the right hand side of (8.1). Show that $I(v)$ can be chosen less than any negative quantity. Use the fact that $I(v)$ is well defined not only for compactly supported and continuously differentiable function $v(z)$ but also for any compactly supported function $v^\alpha(z)$ obeying finite total variation. Assume that there exists a family of $v^\alpha(z)$, $\alpha \in (0, 1]$ such that

$$\lim_{\alpha \rightarrow 0} I(v^\alpha) = -\infty \quad (8.3)$$

and every function $v^\alpha(z)$ obeys an approximation by $v_m^\alpha(z)$, $m \geq 1$ of continuously differentiable compactly supported functions in a sense

$$\lim_m I(v_m^\alpha) = I(v^\alpha). \quad (8.4)$$

We show that under (8.3) and (8.4) the desired result holds. In fact, for fixed α one can choose a number m_α such that $|I(v^\alpha) - I(v_{m_\alpha}^\alpha)| \leq 1$. Hence we obtain

$$\inf_{0 \leq t \leq T} \int_R \mathcal{D}u(t, z) n(t, z) dz dt \leq I(v_{m_\alpha}^\alpha) \leq 1 + I(v^\alpha) \rightarrow -\infty, \quad \alpha \rightarrow 0.$$

Therefore, only (8.3) and (8.4) have to be checked. Since $d_z n \ll dz$ the function $n(z)$ is not absolutely continuous and $w(z)$ is inherited the same property. Therefore by the definition of the negation for absolute continuity [28] a constant k can be chosen such that for any $\alpha > 0$ there exists a positive constant c and non overlapping intervals $(z'_i, z''_i) \in [-c, c]$, such that $\sum_i |w(z''_i) - w(z'_i)| \geq k$ and $\sum_i \int_{z'_i}^{z''_i} n(z) dz \leq \alpha$. Put

$$v^\alpha(z) = \begin{cases} -\frac{1}{\sqrt{\alpha}} \text{sign}[w(z''_i) - w(z'_i)], & z'_i < z \leq z''_i, \\ 0, & \text{otherwise.} \end{cases}$$

Show that (8.3) holds. Evaluate from above $I(v^\alpha)$:

$$\begin{aligned} I(v^\alpha) &= \ell \int_R [(v^\alpha(z))^2 + |v^\alpha(z)|]n(z) dz + \int_R w(z) dv^\alpha(z) \\ &\leq \ell \left(\frac{1}{\alpha} + \frac{1}{\sqrt{\alpha}} \right) \sum_i \int_{z'_i}^{z''_i} n(z) dz + \sum_i w(z'_i)[v^\alpha(z''_i) - v^\alpha(z'_i)] \\ &\leq \ell(1 + \sqrt{\alpha}) + \sum_i w(z'_i)[v^\alpha(z''_i) - v^\alpha(z'_i)]. \end{aligned}$$

Now, summing by parts, we find $\sum_i w(z'_i)[v^\alpha(z''_i) - v^\alpha(z'_i)] = -\sum_i v^\alpha(z''_i)[w(z''_i) - w(z'_i)]$. On the other hand, from the definition of $v^\alpha(z)$ it follows $\sum_i v^\alpha(z''_i)[w(z''_i) - w(z'_i)] = (1/\sqrt{\alpha})\sum_i |w(z''_i) - w(z'_i)| \geq k/\sqrt{\alpha}$. Thereby

$$I(v^\alpha) \leq \ell(1 + \sqrt{\alpha}) - \frac{k}{\sqrt{\alpha}} \rightarrow -\infty, \quad \alpha \rightarrow 0.$$

Evidently to satisfy (8.4), it is sufficient to choose approximating functions $v_m^\alpha(z)$, $m \geq 1$ which are compactly supported and continuously differentiable and such that $\lim_m v_m^\alpha(z) = v^\alpha(z)$ in every point of continuity of $v^\alpha(z)$.

Assume $v \ll \lambda$. Put $K^v(dz) = \int_0^T K_v(t, dz) dt$ and note that $K^v(dz) \ll dz$. Use Lebesgue's decomposition: $K^v(dz) = q(z) dz + K^\perp(dz)$, where $q(z)$ is a density of absolutely continuous part of $K^v(dz)$ and $K^\perp(dz)$ is its singular part. Taking $u(t, z) \equiv u(z)$ which is compactly supported, say, on $[-c, c]$ we find

$$\int_0^T \mathcal{D}u(z) v(dt, dz) = \int_{-c}^c \mathcal{D}u(z) q(z) dz + \int_{-c}^c \mathcal{D}u(z) K^\perp(dz).$$

Since $|u'(z)| \leq |u'(0)| + \int_{-c}^c |u''(y)| dy$ there exists constant, say, ℓ , such that $\int_{-c}^c \mathcal{D}u(z) q(z) dz \leq \ell(1 + \int_{-c}^c |u''(y)| dy)$ and so we arrive to an upper estimate

$$\int_0^T \mathcal{D}u(z) v(dt, dz) \leq \ell \left(1 + \int_{-c}^c |u''(z)| dz \right) + \frac{1}{2} \int_{-c}^c \sigma^2(z) u''(z) K^\perp(dz).$$

Then, using the singularity of $K^\perp(dz)$ and dz , one can choose $u''(z)$ such that the second integral is less than any negative quantity while the first remains bounded.

3. Approximation of rate function

For “ $dX = \dot{X}_t dt$, $dv = n d\lambda$, $d_z n = n'_z dz$ ” denote by

$$S_T(X, v) = \int_0^T \frac{[\dot{X}_t - A_v(t, X_t)]^2}{B_v^2(t, X_t)} dt,$$

$$F_T(v) = \int_0^T \int_R \sigma^2(z) \left[\frac{n'_z(t, z)}{n(t, z)} - \frac{p'(z)}{p(z)} \right]^2 n(t, z) dz dt.$$

Also note one to one correspondence between density $n(t, z)$ and function $v_v(t, z)$ defined in (6.1):

$$n(t, z) = n(t, 0) \frac{p(z)}{p(0)} \exp \left(2 \int_0^z \frac{v_v(t, y)}{\sigma^2(y)} dy \right). \quad (8.5)$$

Put

$$\phi(t) = \int_R |n'_z(t, y)| dy. \quad (8.6)$$

Lemma A.4 *Let $B^2(x, z) \geq \beta^2 > 0$. If $S_T(X, v) < \infty$, $F_T(v) < \infty$, then v can be approximated by a sequence of measures $v^{(k)}$, $k \geq 1$, satisfying the property: $d v^{(k)} = n^{(k)} d\lambda$, $d_z n^{(k)} = n_z^{(k)} dz$, such that the function $v_{v^{(k)}}(t, z)$, corresponding to $n^{(k)}(t, z)$, is compactly supported in z and continuously differentiable in (t, z)*

and what is more

$$\begin{aligned}\lim_k \rho_T(v, v^{(k)}) &= 0, \\ \lim_k S_T(X, v^{(k)}) &= S_T(X, v), \\ \lim_k F_T(v^{(k)}) &= F_T(v).\end{aligned}\tag{8.7}$$

Proof. Introduce a chain of expanding subclasses of measures v characterized in terms of $n(t, z)$ and $v_v(t, z)$:

0) $v_v(t, z)$ is compactly supported in z and continuously differentiable in (t, z) ;

1) $v_v(t, z)$ is compactly supported in z and bounded;

2) $v_v(t, z)$ is compactly supported, $\inf_{t \leq T, z \in R} (n(t, z)/p(z)) > 0$ and $\sup_{t \leq T} [n(t, 0) + \phi(t)] < \infty$;

3) $v_v(t, z)$ is compactly supported, $\inf_{t \leq T, z \in R} (n(t, z)/p(z)) > 0$;

4) $v_v(t, z)$ is compactly supported;

5) $v_v(t, z)$ satisfies the assumptions of the lemma.

The proof is based on the following fact. If measure v from class “ i ” ($i = 1, \dots, 5$) can be approximated by $v^{(k)}$, $k \geq 1$ from class “ $i - 1$ ” in a sense (8.7), then the statement of the lemma holds.

Assume $v^{(k)}$, $k \geq 1$ is such that

$$\begin{aligned}\Lambda_T - \lim_k n^{(k)}(t, z) &= n(t, z) \quad (\Lambda_T(dt, dz) = I_{[0, T]} dt dz), \\ \lim_k F_T(v^{(k)}) &= F_T(v).\end{aligned}\tag{8.8}$$

Then by Scheffe’s theorem [29, 22] we have $\lim_k \int_0^T \int_R |n(t, z) - n^{(k)}(t, z)| dt dz = 0$ that is $v^{(k)}$ converges to v in the total variation norm which implies convergence in Levy–Prohorov’s metric too: $\rho_T(v, v^{(k)}) \rightarrow 0$. Since $A_{v^{(k)}}(t, X_t) = \int_R A(X_t, z) n^{(k)}(t, z) dz$ and $B_{v^{(k)}}^2(t, X_t) = \int_R B^2(X_t, z) n^{(k)}(t, z) dz$ by Lebesgue dominated theorem $S_T(X, v^{(k)}) \rightarrow S_T(X, v)$. Therefore, for all steps of approximations only (8.8) has to be checked.

Assume v is from class “1”. Approximate $v_v(t, z)$ by $v_v^{(k)}(t, z)$:

$$\lim_k \int_0^T \int_R [v_v(t, z) - v_v^{(k)}(t, z)]^2 (1 + n(t, z)) dt dz = 0,$$

where for all k the functions $v_v^{(k)}(t, z)$, $k \geq 1$ are compactly supported continuously differentiable in (t, z) . Without loss of a generality one can assume that all function are bounded by the same constant. Similarly to (8.5) define a density of $v^{(k)}$:

$$n^{(k)}(t, z) = n^{(k)}(t, 0) \frac{p(z)}{p(0)} \exp\left(\int_0^z \frac{v_v^{(k)}(t, y)}{\sigma^2(z)} dy\right),\tag{8.9}$$

with

$$n^{(k)}(t, 0) = \left(\int_R \frac{p(z)}{p(0)} \exp \left(\int_0^z \frac{v_v^{(k)}(t, y)}{\sigma^2(z)} dy \right) dz \right)^{-1}.$$

Put

$$v^{(k)}(dt, dz) = n^{(k)}(t, z) dt dz.$$

Evidently $v^{(k)}$ belongs to class “0”. It is easy to check that

$$v_v^{(k)}(t, z) \equiv v_{v^{(k)}}(t, z) \quad (8.10)$$

and the validity of the first part in (8.8). To verify the second part in (8.8), note that

$$F_T(v^{(k)}) = 4 \int_0^T \int_R \frac{(v_{v^{(k)}}(t, z))^2}{\sigma^2(z)} n^{(k)}(t, z) dz dt$$

and consequently

$$\begin{aligned} |F_T(v) - F_T(v^{(k)})| &\leq \text{const.} \int_0^T \int_R |n(t, z) - n^{(k)}(t, z)| dt dz \\ &\quad + \text{const.} \int_0^T \int_R |v_v^2(t, z) - (v_{v^{(k)}}(t, z))^2| dt dz \\ &\rightarrow 0, \quad k \rightarrow \infty. \end{aligned}$$

Assume v is from class “2”. For the definiteness assume that there exists positive constant z_0 such that $v_v(t, z) \equiv 0$ out of $[-z_0, z_0]$. Put $v_v^{(k)}(t, z) = v_v(t, z)I(|n'_z(t, z)| \leq k)$, define $n^{(k)}(t, z)$ by (8.9) and take $v^{(k)}$ with this density. It belongs to class “1”. Herewith, $v_v^{(k)}(t, z)$ is defined by (8.10). It is clear that the first part in (8.8) holds and below we check the validity of the second part. We have

$$F_T(v^{(k)}) = 4 \int_0^T \int_{|z| \leq z_0} \frac{v_v^2(t, z)}{\sigma^2(z)} I(|n_z(t, z)| \leq k) n^{(k)}(t, z) dz dt,$$

$$F_T(v) = 4 \int_0^T \int_{|z| \leq z_0} \frac{v_v^2(t, z)}{\sigma^2(z)} n(t, z) dz dt.$$

The required convergence $F_T(v^{(k)}) \rightarrow F_T(v)$ holds by Lebesgue dominated theorem since $n^{(k)}(t, z) \leq p(z) \exp(2\phi(t)) \leq \text{const.} n(t, z)$.

Assume v is from class “3”. Putting $v_v^{(k)}(t, z) = v_v(t, z)I(n(t, 0) + \phi(t) \leq k)$ we arrive at

$$n^{(k)}(t, z) = \begin{cases} n(t, z), & n(t, 0) + \phi(t) \leq k, \\ p(z), & n(t, 0) + \phi(t) > k, \end{cases}$$

and since $n^{(k)}(t, 0) \leq k + p(0)$ and $\phi^{(k)}(t) \leq k + \int_R |p'(z)| dz$ measure $v^{(k)}$ with density $n^{(k)}(t, z)$ belongs to class “2”. It is clear that the first part in (8.8) holds

and

$$\begin{aligned}
|F_T(v) - F_T(v^{(k)})| &= 4 \int_0^T \int_{|z| \leq z_0} \frac{v_v^2(t, z)}{\sigma^2(z)} I(n(t, 0) + \phi(t) > k) p(z) dz dt \\
&\leq \text{const.} \int_0^T \int_{|z| \leq z_0} \frac{v_v^2(t, z)}{\sigma^2(z)} I(n(t, 0) + \phi(t) > k) n(t, z) dz dt \\
&\rightarrow 0, \quad k \rightarrow \infty.
\end{aligned}$$

Assume v is from class “4”. Put $n^{(k)}(t, z) = c^{(k)}(t)(n(t, z) \vee p(z))$, where $c^{(k)}(t) = (\int_R (n(t, z) \vee p(z)) dz)^{-1}$ is norming constant. $v^{(k)}$ with this density belongs to class “3”. The first part in (8.8) holds and what is more $\lim_k c^{(k)}(t) = 1$. On the other hand, since $v_{v^{(k)}}(t, z) = v_v(t, z) I(n(t, z) \geq p(z)/k)$ we obtain

$$\begin{aligned}
F_T(v^{(k)}) &= 4 \int_0^T \int_R \frac{v_v^2(t, z)}{\sigma^2(z)} I(n(t, z) \geq p(z)/k) c^{(k)}(t) n(t, z) dz dt \\
&\rightarrow 4 \int_0^T \int_R \frac{v_v^2(t, z)}{\sigma^2(z)} n(t, z) dz dt = F_T(v).
\end{aligned}$$

Assume v is from class “5”. Put $v_v^{(k)}(t, z) = v_v(t, z) T(|z| \leq k)$ and define $n^{(k)}(t, z)$ by (8.9). Then

$$n^{(k)}(t, z) = n^{(k)}(t, 0) \begin{cases} p(z) \frac{n(t, k)}{p(k)}, & z > k, \\ n(t, z), & |z| \leq k, \\ p(z) \frac{n(t, -k)}{p(-k)}, & z < -k. \end{cases}$$

Taking $v^{(k)}$ with this density and noticing that $\lim_k n^{(k)}(t, 0) = 1$ we find

$$F_T(v^{(k)}) = 4 \int_0^T \int_{|z| \leq k} \frac{(v_v(t, z))^2}{\sigma^2(z)} c^{(k)}(t) n(t, z) dz dt \rightarrow F_T(v),$$

i.e. both parts in (8.8) hold.

4. Ergodic property

Consider diffusion pair $(\tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon)$ defined by Itô's differential equations w.r.t. independent Wiener processes W_t and V_t :

$$\begin{aligned}
d\tilde{X}_t^\varepsilon &= G(t, \tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon) dt + \sqrt{\varepsilon} B(\tilde{X}_t^\varepsilon, \tilde{\xi}_t^\varepsilon) dW_t, \\
d\tilde{\xi}_t^\varepsilon &= \frac{1}{\varepsilon} b(t, \tilde{\xi}_t^\varepsilon) dt + \frac{1}{\sqrt{\varepsilon}} \sigma(\tilde{\xi}_t^\varepsilon) dV_t
\end{aligned} \tag{8.11}$$

subject to (x_0, z_0) , where $B(x, z)$ and $\sigma(z)$ are functions involving in (1.1). Assume $b(t, z)$ is continuous it (t, z) , continuously differentiable in t , Lipschitz

continuous in z uniformly in t , and $zb(t, z)$ is negative for large $|z|$ uniformly in t . Also assume that

$$G(t, x, z) = \frac{\dot{X}_t - A_p(t, x)}{B_p(t, x)} B(x, z) + A(x, z), \quad (8.12)$$

where $A(x, z)$ involves in (1.1), \dot{X}_t is the Radon–Nykodim derivative of absolute continuous function X_t from \mathbf{C} with $X_0 = x_0$, $A_p(t, x) = \int_{\mathbf{R}} A(x, z) p(t, z) dz$, $B_p(t, x) = \sqrt{\int_{\mathbf{R}} B^2(x, z) p(t, z) dz}$, and where (comp. (2.1))

$$p(t, z) = c(t) \frac{\exp(2 \int_0^z \frac{b(t, y)}{\sigma^2(y)} dy)}{\sigma^2(z)}$$

with norming function $c(t)$ such that $\int_{\mathbf{R}} p(t, z) dz = 1$. Introduce an occupation measure $\tilde{v}^\varepsilon(dt, dz)$: $\tilde{v}^\varepsilon(\Delta \times \Gamma) = \int_0^\infty I(t \in \Delta, \tilde{\xi}_t^\varepsilon \in \Gamma) dt$ and put $v(dt, dz) = p(t, z) dz dt$.

Lemma A.5

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \rho_T(\tilde{v}^\varepsilon, v) = 0 \quad \text{and} \quad \mathbf{P} - \lim_{\varepsilon \rightarrow 0} r_T(\tilde{X}^\varepsilon, X) = 0.$$

Proof. It is clear, the first statement of the lemma is equivalent to: for any bounded and continuous function $h(t, z)$ $\int_0^T \int_{\mathbf{R}} h(t, \tilde{\xi}_t^\varepsilon) dt \rightarrow \int_0^T \int_{\mathbf{R}} h(t, z) p(t, z) dz dt$ in probability or, for $h^0(t, z) = h(t, z) - \int_0^T \int_{\mathbf{R}} h(t, y) p(t, y) dy dt$,

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \int_0^T \int_{\mathbf{R}} h^0(t, \tilde{\xi}_t^\varepsilon) dt = 0.$$

First we check it for continuously differentiable in t, z function $h(t, z)$, having bounded partial derivatives. Straightforward calculation brings Kolmogorov's type differential equation (t is fixed):

$$\frac{1}{2} \frac{\partial}{\partial z} (\sigma^2(x) p(t, x)) = b(t, z) p(t, z).$$

A conjugate differential equation

$$\frac{1}{2} \sigma^2(z) \frac{\partial w(t, z)}{\partial z} + b(t, z) = h^0(t, z) \quad (8.13)$$

obeys a solution

$$w(t, z) = \frac{2}{\sigma^2(z) p(t, z)} \int_{-\infty}^z h^0(t, y) p(t, y) dy.$$

It is clear that properties of $h(t, z)$ are inherited by $w(t, z)$ and so function $u(t, z) = \int_0^z w(t, y) dy$ is continuously differentiable once in t and twice

in z and what is more, due to the boundness of $w(t, z)$, there exists a positive constant, say ℓ , such that $|u(t, z)| \leq \ell|z|$ and $|u_t(t, z)| \leq \ell|z|$. Applying Itô's formula to $u(t, \tilde{\xi}_t^\varepsilon)$ and taking into account that $w(t, z)$ is solution of differential equation (8.13) we find $u(T, \tilde{\xi}_T^\varepsilon) = u(0, \xi_0) + \int_0^T u'_t(t, \tilde{\xi}_t^\varepsilon) dt + (1/\sqrt{\varepsilon}) \int_0^T w(t, \tilde{\xi}_t^\varepsilon) \sigma(\tilde{\xi}_t^\varepsilon) dV_t + (1/\varepsilon) \int_0^T h^0(t, \tilde{\xi}_t^\varepsilon) dt$ that is

$$\begin{aligned} \int_0^T h^0(t, \tilde{\xi}_t^\varepsilon) dt &= \varepsilon u(T, \tilde{\xi}_T^\varepsilon) - \varepsilon u(0, \xi_0) \\ &\quad - \varepsilon \int_0^T u'_t(t, \tilde{\xi}_t^\varepsilon) dt - \sqrt{\varepsilon} \int_0^T w(t, \tilde{\xi}_t^\varepsilon) \sigma(\tilde{\xi}_t^\varepsilon) dV_t. \end{aligned} \quad (8.14)$$

The second term in the right hand of (8.14) converges to zero; the last term converges to zero in probability since by Problem 1.9.2 in [23] the mentioned convergence follows from $\varepsilon \int_0^T w^2(t, \tilde{\xi}_t^\varepsilon) \sigma^2(\tilde{\xi}_t^\varepsilon) dt \rightarrow 0$; other two terms converge to zero in probability if $\lim_{\varepsilon \rightarrow 0} \varepsilon^2 \mathbf{E} \sup_{t \leq T} (\tilde{\xi}_t^\varepsilon)^2 = 0$. To check the last, apply Itô's formula to $(\varepsilon \tilde{\xi}_t^\varepsilon)^2$:

$$\begin{aligned} (\varepsilon \tilde{\xi}_t^\varepsilon)^2 &= (\varepsilon \xi_0)^2 + 2\varepsilon \int_0^t b(s, \tilde{\xi}_s^\varepsilon) \tilde{\xi}_s^\varepsilon ds + \varepsilon \int_0^t \sigma^2(\tilde{\xi}_s^\varepsilon) ds \\ &\quad + 2\varepsilon^{3/2} \int_0^t \tilde{\xi}_s^\varepsilon \sigma(\tilde{\xi}_s^\varepsilon) dV_s. \end{aligned}$$

The function $b(s, z)$ is such that $zb(t, z)$ is negative for large $|z|$ what implies $(\varepsilon \tilde{\xi}_t^\varepsilon)^2 \leq (\varepsilon \xi_0)^2 + T\varepsilon \text{const.} + 2\varepsilon^{3/2} \int_0^t \tilde{\xi}_s^\varepsilon \sigma(\tilde{\xi}_s^\varepsilon) dV_s$. Thereby $\mathbf{E}(\varepsilon \tilde{\xi}_t^\varepsilon)^2 \leq (\varepsilon \xi_0)^2 + T\varepsilon \text{const.}$ In turn, using Doob's inequality (see e.g. Theorem 1.9.1 in [23]), we arrive to $\mathbf{E} \sup_{t \leq T} (\varepsilon \tilde{\xi}_t^\varepsilon)^2 \leq (\varepsilon \xi_0)^2 + T\varepsilon \text{const.} + \text{const.} \varepsilon^3 \int_0^T \mathbf{E}(\varepsilon \tilde{\xi}_t^\varepsilon)^2 dt$ and, due to the above obtained upper bound for $\mathbf{E}(\varepsilon \tilde{\xi}_t^\varepsilon)^2$, the result holds.

If $h(t, z)$ is bounded and continuous only, it can be approximated by smooth functions $h_m(t, z)$, $m \geq 1$ in the following sense: for any $k \geq 1$ $\lim_m \sup_{t \leq T, |z| \leq k} |h(t, z) - h_m(t, z)| = 0$. Since for every $h_m(t, z)$ the statement of the lemma is proved, it holds for $h(t, z)$ if

$$\begin{aligned} \lim_m \int_0^T \int_R |h(t, z) - h_m(t, z)| p(t, z) dz dt &= 0, \\ \mathbf{P} - \lim_m \overline{\lim}_{\varepsilon \rightarrow 0} \int_0^T |h(t, \tilde{\xi}_t^\varepsilon) - h_m(t, \tilde{\xi}_t^\varepsilon)| n(t, z) dt &= 0. \end{aligned}$$

The first takes place since $\lim_k \int_0^T \int_{|z| > k} n(t, z) dz dt = 0$ while the second from $\mathbf{P} - \lim_m \overline{\lim}_{\varepsilon \rightarrow 0} \int_0^T I(|\tilde{\xi}_t^\varepsilon| > k) dt = 0$ and the fact that one can choose smooth bounded functions $g_k(z)$, $k \geq 1$ such that $I(|z| > k) \leq g_k(z)$, $\lim_k g_k(z) = 0$, $z \in R$ and by proved above $\int_0^T g_k(\tilde{\xi}_t^\varepsilon) dt \rightarrow \int_0^T \int_R g_k(z) p(t, z) dz dt \rightarrow 0$, $k \rightarrow \infty$.

To check the second statement, put $\Delta_t = \tilde{X}_t^\varepsilon - X_t$. From the first equation in (8.11) we find

$$\begin{aligned} \Delta_t &= \int_0^t \left[\frac{\dot{X}_t - A_p(s, X_s)}{B_p(s, X_s)} B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) + A(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) - \dot{X}_s \right] ds + \sqrt{\varepsilon} \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) dW_s \\ &= \int_0^t \left[\frac{\dot{X}_t - A_p(s, X_s)}{B_p(s, X_s)} (B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) - B(X_s, \tilde{\xi}_s^\varepsilon)) \right] ds \\ &\quad + \int_0^t \left[\frac{\dot{X}_s - A_p(s, X_s)}{B_p(s, X_s)} (B(X_s, \tilde{\xi}_s^\varepsilon) - B_p(s, X_s)) \right] ds \\ &\quad + \int_0^t (A(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) - A(X_s, \tilde{\xi}_s^\varepsilon)) ds + \int_0^t (A(X_s, \tilde{\xi}_s^\varepsilon) - A_p(s, X_s)) ds \\ &\quad + \sqrt{\varepsilon} \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) dW_s . \end{aligned}$$

For brevity put $\varphi_s = (\dot{X}_s - A_p(s, X_s))/B_p(s, X_s)$. Then by the Lipschitz continuity of $A(x, z), B(x, z)$ in x uniformly in z , say, with constant ℓ , we obtain

$$\begin{aligned} |\Delta_t| &\leq \ell \int_0^t (1 + |\varphi_s|) \Delta_s ds + \sup_{t \leq T} \left| \int_0^t \varphi_s (B(X_s, \tilde{\xi}_s^\varepsilon) - B_p(s, X_s)) ds \right| \\ &\quad + \sup_{t \leq T} \left| \int_0^t (A(X_s, \tilde{\xi}_s^\varepsilon) - A_p(s, X_s)) ds \right| + \sqrt{\varepsilon} \sup_{t \leq T} \left| \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) dW_s \right| . \end{aligned}$$

Therefore, by Gronwall–Bellman’s inequality

$$\begin{aligned} \sup_{t \leq T} |\Delta_t| &\leq \exp \left(\ell \int_0^T (1 + |\varphi_s|) ds \right) \\ &\quad \times \left[\sup_{t \leq T} \left| \int_0^t \varphi_s (B(X_s, \tilde{\xi}_s^\varepsilon) - B_p(s, X_s)) ds \right| \right. \\ &\quad \left. + \sup_{t \leq T} \left| \int_0^t (A(X_s, \tilde{\xi}_s^\varepsilon) - A_p(s, X_s)) ds \right| + \sqrt{\varepsilon} \sup_{t \leq T} \left| \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) dW_s \right| \right] . \end{aligned}$$

Hence, the second statement holds if

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \sup_{t \leq T} \sqrt{\varepsilon} \sup_{t \leq T} \left| \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\xi}_s^\varepsilon) dW_s \right| = 0 \quad (8.15)$$

and for any measurable function ψ_s such that $\int_0^T \psi_s^2 ds < \infty$ and any continuous function $C(x, z)$, being Lipschitz’s continuous in x uniformly in z ,

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \sup_{t \leq T} \left| \int_0^t \psi_s (C(X_s, \tilde{\xi}_s^\varepsilon) - C_p(s, X_s)) ds \right| = 0 , \quad (8.16)$$

where $C_p(s, X_s) = \int_R C(X_s, z) p(s, z) dz$. It can be shown (see e.g. the method of proving the statement (2) of Theorem 4.6, Chap. 4 in [24]) that $\sup_{t \leq T} \mathbf{E}(\tilde{X}_t^\varepsilon)^2 \leq \text{const.}$ and so $\mathbf{E} \int_0^T B^2(\tilde{X}_s^\varepsilon, \tilde{\zeta}_s^\varepsilon) ds \leq \text{const.}$ Consequently, by Doob's inequality (see e.g. Theorem 1.9.1 [23]) we get $\mathbf{E} \sup_{t \leq T} |\sqrt{\varepsilon} \int_0^t B(\tilde{X}_s^\varepsilon, \tilde{\zeta}_s^\varepsilon) dW_s|^2 \leq \varepsilon \text{const.}$ that is (8.15) holds. To check the validity of (8.16) with $\psi_s C(x, z) \geq 0$, note that, due to Problem 5.3.2 in [23], it is sufficient to prove

$$\mathbf{P} - \lim_{\varepsilon \rightarrow 0} \int_0^t \psi_s (C(X_s, \tilde{\zeta}_s^\varepsilon) - C_p(s, X_s)) ds = 0, \quad \forall t \leq T, \quad (8.17)$$

and what is more, due to an arbitrariness of ψ_s and $C(x, z)$, (8.17) implies (8.16) in the general case since one can use separately (8.17) for positive $(\psi_s C(x, z))^+$ and negative $(\psi_s C(x, z))^-$ parts. Therefore, (8.17) remains to be verified. If ψ_s is bounded and continuous, (8.17) takes place by virtue of the first statement of the lemma. If only $\int_0^T \psi_s^2 ds < \infty$, approximate ψ_s by bounded and continuous functions $\psi_s^{(k)}$, $k \geq 1$ such that $\lim_k \int_0^T (\psi_s - \psi_s^{(k)})^2 ds = 0$ and, due to the boundness in z of $C(x, z)$ and Cauchy–Schwartz's inequality, we find that

$$\left| \int_0^t (\psi_s - \psi_s^{(k)}) (C(X_s, \tilde{\zeta}_s^\varepsilon) - C_p(s, X_s)) ds \right| \leq \text{const.} \sqrt{\int_0^T (\psi_s - \psi_s^{(k)})^2 ds} \rightarrow 0, \quad k \rightarrow \infty$$

that is (8.17) takes place since it holds for every $\psi_s^{(k)}$.

5. LD-regularization

Parallel to X_t^ε , defined in (1.1), determine new diffusion $X_t^{\varepsilon, \beta}$ with uniformly non singular diffusion parameter $B^2(x, z) + \beta^2$, $\beta^2 > 0$, letting

$$dX_t^{\varepsilon, \beta} = A(X_t^{\varepsilon, \beta}, \zeta_t^\varepsilon) dt + \sqrt{\varepsilon} [B(X_t^{\varepsilon, \beta}, \zeta_t^\varepsilon) dW_t + \beta dW_t'] \quad (8.18)$$

subject to the same initial point x_0 , where W_t' is a Wiener process independent of $(W_t, \zeta_t^\varepsilon)$.

Lemma A.6 *Under assumption (A.1) for every $T > 0$ and $\eta > 0$*

$$\lim_{\beta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P}(r_T(X^{\varepsilon, \beta}, X^\varepsilon) > \eta) = -\infty.$$

Proof. Put $\Delta_t = X_t^{\varepsilon, \beta} - X_t^\varepsilon$, and

$$a_1(x', x'', z) = \frac{A(x'', z) - A(x', z)}{x'' - x'}, \quad a_2(x', x'', z) = \frac{B(x'', z) - B(x', z)}{x'' - x'},$$

where for $x' = x''$ $a_i(x', x', z)$, $i = 1, 2$ are Radon–Nikodym's derivatives. Note that for $x' \neq x''$ $a_i(x', x'', z)$, are bounded, say, by constant ℓ , and so $a_i(x', x', z)$ can be taken bounded by the same constant. For brevity, denote by $\alpha_i(t) = a_i(X_t^{\varepsilon, \beta}, X_t^\varepsilon, \zeta_t^\varepsilon)$, $i = 1, 2$. (8.18) and (1.1) imply: $\Delta_t = \int_0^t \alpha_1(s) \Delta_s ds + \sqrt{\varepsilon} \int_0^t$

$\alpha_2(s)\Delta_s dW_s + \sqrt{\varepsilon}\beta W'_t$. Letting $\mathcal{E}_t = \exp(\int_0^t [\alpha_1(s) - (\varepsilon/2)\alpha_2^2(s)] ds + \sqrt{\varepsilon} \int_0^t \alpha_2(s) dW_s)$ and using Itô's formula, we find that $\Delta_t = \sqrt{\varepsilon}\beta \mathcal{E}_t \int_0^t \mathcal{E}_s^{-1} dW'_s$ and thereby

$$\sup_{t \leq T} |\Delta_t| \leq \sqrt{\varepsilon}\beta \sup_{t \leq T} \mathcal{E}_t \sup_{t \leq T} \left| \int_0^t \mathcal{E}_s^{-1} dW'_s \right|.$$

Put $\Gamma_N = \{1/N \leq \inf_{t \leq T} \mathcal{E}_t \leq \sup_{t \leq T} \mathcal{E}_t \leq N\}$ and use an upper estimate

$$\begin{aligned} \mathbf{P} \left(\sup_{t \leq T} |\Delta_t| > \eta \right) &\leq \mathbf{P} \left(\sup_{t \leq T} |\Delta_t| > \eta, \Gamma_N \right) + \mathbf{P}(\Omega \setminus \Gamma_N) \\ &\leq 2 \max \left[\mathbf{P} \left(\sup_{t \leq T} |\Delta_t| > \eta, \Gamma_N \right), \mathbf{P}(\Omega \setminus \Gamma_N) \right] \end{aligned}$$

which implies, due to the boundness of $\alpha_i(s)$, $i = 1, 2$, the desired statement if

$$\begin{aligned} \lim_{N \rightarrow \infty} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P} \left(\sqrt{\varepsilon} \sup_{t \leq T} \left| \int_0^t \alpha_2(s) dW_s \right| > N \right) &= -\infty, \\ \lim_{\beta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P} \left(\sqrt{\varepsilon}\beta \sup_{t \leq T} \left| \int_0^t \mathcal{E}_s^{-1} dW'_s \right| > \eta, \Gamma_N \right) &= -\infty, \quad \forall N \geq 1. \end{aligned} \quad (8.19)$$

Let $\tau = \{t \leq T : |\int_0^t \alpha_2(s) dW_s| > (N/\sqrt{\varepsilon})\}$ and $\sigma = \{t \leq T : |\int_0^t \mathcal{E}_s^{-1} dW'_s| > (\eta/\sqrt{\varepsilon}\beta)\}$. It is clear that (8.19) is equivalent to:

$$\begin{aligned} \lim_{N \rightarrow \infty} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P} \left(\sqrt{\varepsilon} \int_0^\tau \alpha_2(s) dW_s \geq N \text{ (or } \leq -N) \right) &= -\infty, \\ \lim_{\beta \rightarrow 0} \overline{\lim}_{\varepsilon \rightarrow 0} \varepsilon \log \mathbf{P} \left(\sqrt{\varepsilon}\beta \int_0^\sigma \mathcal{E}_s^{-1} dW'_s \geq \eta \text{ (or } \leq -\eta), \Gamma_N \right) &= -\infty, \quad \forall N \geq 1. \end{aligned} \quad (8.20)$$

Below we check (8.20). To this end with $\lambda \in R$, introduce continuous local martingales: $Z_t^1 = \exp(\lambda \int_0^t \alpha_2(s) dW_s - (\lambda^2/2) \int_0^t \alpha_2^2(s) ds)$ and $Z_t^2 = \exp(\lambda \int_0^t \mathcal{E}_s^{-1} dW'_s - (\lambda^2/2) \int_0^t \mathcal{E}_s^{-2} ds)$, where each of them is a supermartingale too (see Problem 1.4.4. in [23]) that is $\mathbf{E}Z_\tau^1 \leq 1$ and $\mathbf{E}Z_\sigma^2 \leq 1$. Then we use obvious inequalities: $\mathbf{E}I(\sqrt{\varepsilon} \int_0^\tau \alpha_2(s) dW_s \geq N)Z_\tau^1 \leq 1$ and $\mathbf{E}I(\sqrt{\varepsilon}\beta \int_0^\sigma \mathcal{E}_s^{-1} dW'_s \geq \eta, \Gamma_N)Z_\sigma^2 \leq 1$. Since for $\lambda > 0$, $\log Z_\tau^1 \geq (\lambda N/\sqrt{\varepsilon}) - (\lambda^2 \ell^2 T/2)$ and $\log Z_\sigma^2 \geq (\lambda \eta/\sqrt{\varepsilon}\beta) - (\lambda^2 N^2 T/2)$ on sets $\{\sqrt{\varepsilon} \int_0^\tau \alpha_2(s) dW_s \geq N\}$ and $\{\sqrt{\varepsilon}\beta \int_0^\sigma \mathcal{E}_s^{-1} dW'_s \geq \eta, \Gamma_N\}$ respectively, we arrive at (8.20) for “ $\geq N$ ” and “ $\geq \eta$ ”, taking $\lambda^1 = (N/\sqrt{\varepsilon}\ell^2 T)$ and $\lambda^2 = (\eta/\sqrt{\varepsilon}\beta N^2 T)$. For “ $\leq -N$ ” and “ $\leq -\eta$ ” the validity of (8.20) is proved in the same way.

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