

Large deviations for perturbed reflected diffusion processes

Lijun Bo^{1,2} and Tusheng Zhang^{3,*}

1. Department of Mathematics, Xidian University, Xi'an 710071, China
2. School of Mathematical Sciences, Nankai University, Tianjin 300071, China
3. Department of Mathematics, University of Manchester
Oxford Road, Manchester M13 9PL, England, U.K.

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Abstract

In this article, we establish a large deviation principle for the solutions of perturbed reflected diffusion processes. The key is to prove a uniform Freidlin-Ventzell estimates of perturbed diffusion processes.

Keywords: Large deviations; perturbed diffusion processes; uniform Freidlin-Ventzell estimates.

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1 Introduction

In [6], Doney and Zhang obtained the existence and uniqueness of the solutions for the following perturbed diffusion and perturbed reflected diffusion equations:

$$X_t = x + \int_0^t \sigma(X_s) dB_s + \int_0^t b(X_s) ds + \alpha \sup_{0 \leq s \leq t} X_s, \quad t \in [0, 1], \quad (1.1)$$

and

$$Y_t = y + \int_0^t a(Y_s) dB_s + \alpha \sup_{0 \leq s \leq t} Y_s + L_t, \quad t \in [0, 1], \quad (1.2)$$

*Corresponding author. E-mail address: tzhang@maths.man.ac.uk

where $\alpha \in (0, 1)$, $x \in \mathbf{R}$, $y \in \mathbf{R}_+$ are deterministic, $b, \sigma : \mathbf{R} \rightarrow \mathbf{R}$ and $a : \mathbf{R}_+ \rightarrow \mathbf{R}$ are bounded Lipschitz continuous function, $\{L_t, t \in [0, 1]\}$ is non-decreasing with $L_0 = 0$ and

$$\int_0^t \chi_{\{Y_s=0\}} dL_s = L_t, \quad t \in [0, 1],$$

$\{B_t, t \in [0, 1]\}$ is a 1-dimensional standard Brownian motion on a completed probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbf{P})$. The perturbed reflected Brownian motion was first introduced by Le Gall and M. Yor, in [8, 9], and subsequently studied by Carmona, Petit and Yor in [2, 3], Perman and Werner in [12] and Chaumont and Doney in [4]. Consider the small noise perturbations

$$X_t^\varepsilon = x + \sqrt{\varepsilon} \int_0^t \sigma(X_s^\varepsilon) dB_s + \int_0^t b(X_s^\varepsilon) ds + \alpha \sup_{0 \leq s \leq t} X_s^\varepsilon, \quad t \in [0, 1], \quad (1.3)$$

and

$$Y_t^\varepsilon = y + \sqrt{\varepsilon} \int_0^t a(Y_s^\varepsilon) dB_s + \alpha \sup_{0 \leq s \leq t} Y_s^\varepsilon + L_t^\varepsilon, \quad t \in [0, 1]. \quad (1.4)$$

The aim of this paper is to establish a large deviation principle for the laws of X^ε and Y^ε on the space of continuous functions equipped with uniform topology. Our approach will be based on a classical result of Azencott [1], which can be stated as the follows:

Proposition 1.1 *Let (E_i, d_i) ($i = 1, 2$) be two Polish spaces and $\Phi_i^\varepsilon : \Omega \rightarrow E_i$, $\varepsilon > 0$ ($i = 1, 2$) be two families of random variables. Assume that*

(1) $\{\Phi_1^\varepsilon, \varepsilon > 0\}$ *satisfies a large deviation principle with the rate function $I_1 : E_1 \rightarrow [0, \infty]$.*

(2) *There exists a map $K : \{I_1 < \infty\} \rightarrow E_2$ such that for every $a < \infty$, $K : \{I_1 \leq a\} \rightarrow E_2$ is continuous.*

(3) *For any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in E_1$ satisfying $I_1(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,*

$$\mathbf{P}(d_2(\Phi_2^\varepsilon, K(h)) \geq \delta, d_1(\Phi_1^\varepsilon, h) \leq \rho) \leq \exp\left(-\frac{R}{\varepsilon}\right). \quad (1.5)$$

Then $\{\Phi_2^\varepsilon, \varepsilon > 0\}$ satisfies a large deviation principle with the rate function

$$I(g) = \inf\{I_1(h); K(h) = g\}.$$

The estimates (1.5) is also known as the uniform Freidlin-Ventzell estimates.

The rest of the paper is organized as follows. Section 2 is for the large deviation principle of perturbed diffusion processes. The large deviation estimates for the perturbed reflected diffusion processes is in Section 3.

2 LDP for perturbed diffusion processes

In this section, we will give a large deviation principle of the perturbed diffusion processes (1.3). Let H denote the Cameron-Martin space, i.e.

$$H = \left\{ h(t) = \int_0^t \dot{h}(s) ds : [0, 1] \rightarrow \mathbf{R}; \int_0^1 |\dot{h}(s)|^2 ds < +\infty \right\},$$

and $\{\nu_\varepsilon, \varepsilon > 0\}$ be the probability measure induced by X^ε on $C_x([0, 1], \mathbf{R})$, the space of all continuous functions $f : [0, 1] \rightarrow \mathbf{R}$ such that $f(0) = x$, equipped with the supremum norm topology. For $h \in C_0([0, 1], \mathbf{R})$, define $\tilde{I} : C_0([0, 1], \mathbf{R}) \rightarrow [0, \infty]$ by

$$\tilde{I}(h) = \begin{cases} \frac{1}{2} \int_0^1 |\dot{h}(s)|^2 ds, & \text{if } h \in H, \\ +\infty, & \text{otherwise.} \end{cases}$$

The well known Schilder theorem states that the laws μ_ε of $\{\sqrt{\varepsilon}B_t, t \in [0, 1]\}$ satisfies a large deviation principle on $C_0([0, 1], \mathbf{R})$ with the rate function $\tilde{I}(\cdot)$, that is,

(a) for any closed subset $C \subset C_0([0, 1], \mathbf{R})$,

$$\limsup_{\varepsilon \rightarrow 0} \varepsilon \log \mu_\varepsilon(C) \leq - \inf_{g \in C} \tilde{I}(g),$$

(b) for any open subset $G \subset C_0([0, 1], \mathbf{R})$,

$$\liminf_{\varepsilon \rightarrow 0} \varepsilon \log \mu_\varepsilon(G) \geq - \inf_{g \in G} \tilde{I}(g).$$

Let $F(h)$ be the solution of the perturbed equation

$$\begin{aligned} F(h)(t) &= x + \int_0^t \sigma(F(h)(s)) \dot{h}(s) ds + \int_0^t b(F(h)(s)) ds \\ &\quad + \alpha \sup_{0 \leq s \leq t} F(h)(s), \quad h \in H, \quad t \in [0, 1]. \end{aligned} \quad (2.1)$$

Then we have the following main result.

Theorem 2.1 *If $\alpha \in (0, 1)$, then the family $\{\nu_\varepsilon, \varepsilon > 0\}$ obeys a large deviation principle with the rate function*

$$I(g) = \inf \left\{ \tilde{I}(h); F(h) = g \right\}, \quad g \in C_x([0, 1], \mathbf{R}).$$

Before giving the proof of Theorem 2.1, recall the following lemma (see e.g. [13]).

Lemma 2.1 Let $Z_t := \int_s^t C(u)dB_u + \int_s^t D(u)du$ be an Itô process, where $0 \leq s < t < \infty$ and $C, D : [0, \infty) \times \Omega \rightarrow \mathbf{R}$ are $(\mathcal{F}_t)_{t \geq 0}$ progressively measurable random processes. If $|C(\cdot)| \leq M_1$ and $|D(\cdot)| \leq M_2$, then for $T > s$ and $R > 0$ satisfying $R > M_2(T - s)$, we have

$$\mathbf{P} \left(\sup_{s \leq t \leq T} |Z_t| \geq R \right) \leq \exp \left(-\frac{(R - M_2(T - s))^2}{2M_1^2(T - s)} \right). \quad (2.2)$$

Proof of Theorem 2.1. For $h \in H$, let $F(h)$ be defined as in (2.1). First we prove that for any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in C_0([0, 1], \mathbf{R})$ satisfying $\tilde{I}(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,

$$\begin{aligned} & \mathbf{P} \left(\sup_{0 \leq t \leq 1} |X_t^\varepsilon - F(h)(t)| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon}B_t - h(t)| \leq \rho \right) \\ & \leq \exp \left(-\frac{R}{\varepsilon} \right). \end{aligned} \quad (2.3)$$

By (1.3) and (2.1),

$$\begin{aligned} X_t^\varepsilon - F(h)(t) &= \int_0^t \sigma(X_s^\varepsilon) \left(\sqrt{\varepsilon}dB_s - \dot{h}(s)ds \right) \\ &+ \int_0^t (\sigma(X_s^\varepsilon) - \sigma(F(h)(s))) \dot{h}(s)ds \\ &+ \int_0^t (b(X_s^\varepsilon) - b(F(h)(s))) ds \\ &+ \alpha \left(\sup_{0 \leq s \leq t} X_s^\varepsilon - \sup_{0 \leq s \leq t} F(h)(s) \right). \end{aligned} \quad (2.4)$$

Consequently,

$$\begin{aligned} |X_t^\varepsilon - F(h)(t)| &\leq \left| \int_0^t \sigma(X_s^\varepsilon) \left(\sqrt{\varepsilon}dB_s - \dot{h}(s)ds \right) \right| \\ &+ L \int_0^t |X_s^\varepsilon - F(h)(s)| \left(1 + |\dot{h}(s)| \right) ds \\ &+ \alpha \sup_{0 \leq s \leq t} |X_s^\varepsilon - F(h)(s)|, \end{aligned} \quad (2.5)$$

where $L > 0$ is the Lipschitz coefficient, and we also used the fact that

$$\left| \sup_{0 \leq s \leq t} u(s) - \sup_{0 \leq s \leq t} v(s) \right| \leq \sup_{0 \leq s \leq t} |u(s) - v(s)|,$$

for any two continuous functions u and v on \mathbf{R}_+ . Thus it follows from (2.5) that, for $t \in [0, 1]$,

$$\begin{aligned} \sup_{0 \leq u \leq t} |X_u^\varepsilon - F(h)(u)| &\leq \frac{1}{(1-\alpha)} \sup_{0 \leq u \leq t} \left| \int_0^u \sigma(X_s^\varepsilon) (\sqrt{\varepsilon} dB_s - \dot{h}(s) ds) \right| \\ &\quad + \frac{L}{(1-\alpha)} \int_0^t \sup_{0 \leq u \leq s} |X_u^\varepsilon - F(h)(u)| (1 + |\dot{h}(s)|) ds. \end{aligned} \quad (2.6)$$

By the Gronwall lemma this yields that

$$\begin{aligned} \sup_{0 \leq t \leq 1} |X_t^\varepsilon - F(h)(t)| &\leq \frac{1}{(1-\alpha)} \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) (\sqrt{\varepsilon} dB_s - \dot{h}(s) ds) \right| \\ &\quad \times \exp \left(\int_0^1 \frac{L}{(1-\alpha)} (1 + |\dot{h}(s)|) ds \right) \\ &\leq C_1 \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) (\sqrt{\varepsilon} dB_s - \dot{h}(s) ds) \right|, \end{aligned} \quad (2.7)$$

where $C_1 := \frac{1}{1-\alpha} \exp \left(\frac{L(1+\|h\|_H)}{1-\alpha} \right) < \infty$ with $\|h\|_H := \left(\int_0^1 |\dot{h}(s)|^2 ds \right)^{\frac{1}{2}}$ for $h \in H$. Thus to prove (2.3), it suffices to prove that, for any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in C_0([0, 1], \mathbf{R})$ satisfying $\tilde{I}(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,

$$\begin{aligned} &\mathbf{P} \left(\sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) (\sqrt{\varepsilon} dB_s - \dot{h}(s) ds) \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t - h(t)| \leq \rho \right) \\ &\leq \exp \left(-\frac{R}{\varepsilon} \right). \end{aligned} \quad (2.8)$$

For $\varepsilon > 0$, define a probability measure \mathbf{P}^ε on Ω by

$$d\mathbf{P}^\varepsilon = Z_\varepsilon d\mathbf{P} := \exp \left(\frac{1}{\sqrt{\varepsilon}} \int_0^1 \dot{h}(s) dB_s - \frac{1}{2\varepsilon} \int_0^1 |\dot{h}(s)|^2 ds \right) d\mathbf{P}.$$

Then by Girsanov theorem, $\{B_t^\varepsilon := B_t - \frac{1}{\sqrt{\varepsilon}} h(t), t \in [0, 1]\}$ is a standard Brownian motion on $(\Omega, \mathcal{F}, \mathbf{P}^\varepsilon)$. If we let

$$\begin{aligned} A^\varepsilon &= \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) (\sqrt{\varepsilon} dB_s - \dot{h}(s) ds) \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t - h(t)| \leq \rho \right\} \\ &= \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) \sqrt{\varepsilon} dB_s^\varepsilon \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t^\varepsilon| \leq \rho \right\}, \end{aligned}$$

then by the Hölder inequality

$$\begin{aligned}\mathbf{P}(A^\varepsilon) &= \int_{\Omega} Z_\varepsilon^{-1} \chi_{A^\varepsilon}(\omega) \mathbf{P}^\varepsilon(d\omega) \\ &\leq \left(\int_{\Omega} Z_\varepsilon^{-2}(\omega) \mathbf{P}^\varepsilon(d\omega) \right)^{\frac{1}{2}} (\mathbf{P}^\varepsilon(A^\varepsilon))^{\frac{1}{2}}.\end{aligned}$$

Note that

$$\begin{aligned}\int_{\Omega} Z_\varepsilon^{-2}(\omega) \mathbf{P}^\varepsilon(d\omega) &= \mathbf{E}_{\mathbf{P}^\varepsilon} \left[\exp \left(-\frac{2}{\sqrt{\varepsilon}} \int_0^1 \dot{h}(s) dB_s + \frac{1}{\varepsilon} \int_0^1 |\dot{h}(s)|^2 ds \right) \right] \\ &= \mathbf{E}_{\mathbf{P}^\varepsilon} \left[\exp \left(-\frac{2}{\sqrt{\varepsilon}} \int_0^1 \dot{h}(s) dB_s^\varepsilon - \frac{1}{\varepsilon} \int_0^1 |\dot{h}(s)|^2 ds \right) \right] \\ &= \mathbf{E}_{\mathbf{P}^\varepsilon} \left[\exp \left(-\frac{2}{\sqrt{\varepsilon}} \int_0^1 \dot{h}(s) dB_s^\varepsilon - \frac{2}{\varepsilon} \int_0^1 |\dot{h}(s)|^2 ds \right) \right] \\ &\quad \times \exp \left(\frac{1}{\varepsilon} \int_0^1 |\dot{h}(s)|^2 ds \right) \\ &= \exp \left(\frac{1}{\varepsilon} \|h\|_H^2 \right).\end{aligned}$$

Therefore if $\tilde{I}(h) \leq a$, then

$$\mathbf{P}(A^\varepsilon) \leq \exp \left(\frac{a}{\varepsilon} \right) (\mathbf{P}^\varepsilon(A^\varepsilon))^{\frac{1}{2}}. \quad (2.9)$$

Note that under the probability \mathbf{P}^ε , X^ε satisfies the following stochastic differential equation.

$$U_t^\varepsilon = x + \int_0^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_0^t (b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s)) ds + \alpha \sup_{0 \leq s \leq t} U_s^\varepsilon. \quad (2.10)$$

Therefore,

$$\begin{aligned}\mathbf{P}^\varepsilon(A^\varepsilon) &= \mathbf{P}^\varepsilon \left(\sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(X_s^\varepsilon) \sqrt{\varepsilon} dB_s^\varepsilon \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t^\varepsilon| \leq \rho \right) \\ &= \mathbf{P} \left(\sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right),\end{aligned}$$

Thus, in view of (2.9), to prove (2.8), the proof of (2.8) is reduced to show that, for any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in C_0([0, 1], \mathbf{R})$ satisfying $\tilde{I}(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,

$$\mathbf{P} \left(\sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right) \leq \exp \left(-\frac{R}{\varepsilon} \right). \quad (2.11)$$

For $n \in \mathbf{N}$ fixed, set $t_k = \frac{k}{n}$ for $k \in \{0, 1, 2, \dots, n\}$. Define

$$U_t^{\varepsilon, n} := U_{t_k}^\varepsilon, \quad \text{if } t_k \leq t < t_{k+1}, \quad k \in \{0, 1, \dots, n-1\}.$$

Then for $\delta_1 > 0$

$$\begin{aligned} A^\varepsilon &\subset \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t (\sigma(U_s^\varepsilon) - \sigma(U_s^{\varepsilon, n})) \sqrt{\varepsilon} dB_s \right| \geq \frac{\delta}{2}, \sup_{0 \leq t \leq 1} |U_t^\varepsilon - U_t^{\varepsilon, n}| \leq \delta_1 \right\} \\ &\cup \left\{ \sup_{0 \leq t \leq 1} |U_t^\varepsilon - U_t^{\varepsilon, n}| > \delta_1 \right\} \\ &\cup \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma(U_s^{\varepsilon, n}) \sqrt{\varepsilon} dB_s \right| \geq \frac{\delta}{2}, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right\} \\ &:= B^\varepsilon \cup C^\varepsilon \cup D^\varepsilon. \end{aligned}$$

On the set $\left\{ \sup_{0 \leq t \leq 1} |U_t^\varepsilon - U_t^{\varepsilon, n}| \leq \delta_1 \right\}$,

$$\sup_{0 \leq s \leq 1} \varepsilon |\sigma(U_s^\varepsilon) - \sigma(U_s^{\varepsilon, n})|^2 \leq L^2 \varepsilon \delta_1^2.$$

By Lemma 2.1, it follows that

$$\mathbf{P}(B^\varepsilon) \leq \exp\left(-\frac{\delta^2}{8L^2 \varepsilon \delta_1^2}\right) \leq \exp\left(-\frac{R}{\varepsilon}\right), \quad (2.12)$$

if $\delta_1 \leq \frac{\delta}{2L\sqrt{2R}}$.

On the other hand, on the set $\left\{ \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right\}$, for any $t \in [0, 1]$

$$\begin{aligned} \left| \int_0^t \sigma(U_s^{\varepsilon, n}) \sqrt{\varepsilon} dB_s \right| &= \sqrt{\varepsilon} \left| \sum_{j=0}^{n-1} \sigma(U_{t_j}^\varepsilon) (B_{t_{j+1} \wedge t} - B_{t_j \wedge t}) \right| \\ &\leq 2 \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \sum_{j \in \{0, \dots, n-1\}: t_j \leq t} |\sigma(U_{t_j}^\varepsilon)| \\ &\leq 2nM\rho, \end{aligned}$$

where $M > 0$ is a common bound of b and σ . Therefore if $\rho < \frac{\delta}{4nM}$, then

$$D^\varepsilon = \emptyset.$$

To treat C^ε , we note that for $t \in [t_k, t_{k+1})$, $k \in \{0, 1, \dots, n-1\}$,

$$\begin{aligned} |U_t^\varepsilon - U_t^{\varepsilon, n}| &\leq \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t (b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s)) ds \right| \\ &\quad + \alpha \left(\sup_{0 \leq s \leq t} U_s^\varepsilon - \sup_{0 \leq s \leq t_k} U_s^\varepsilon \right). \end{aligned} \quad (2.13)$$

Consider two possible cases:

- The case I: $\sup_{0 \leq s \leq t} U_s^\varepsilon = \sup_{0 \leq s \leq t_k} U_s^\varepsilon$.

We have

$$|U_t^\varepsilon - U_t^{\varepsilon,n}| \leq \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t \left(b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s) \right) ds \right|. \quad (2.14)$$

- The case II: $\sup_{0 \leq s \leq t} U_s^\varepsilon > \sup_{0 \leq s \leq t_k} U_s^\varepsilon$.

Then there exists $u \in (t_k, t]$ such that $U_u^\varepsilon = \sup_{0 \leq s \leq t} U_s^\varepsilon$, which yields that

$$\begin{aligned} |U_t^\varepsilon - U_{t_k}^\varepsilon| &\leq \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t \left(b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s) \right) ds \right| \\ &\quad + \alpha \left(U_u^\varepsilon - \sup_{0 \leq s \leq t_k} U_s^\varepsilon \right) \\ &\leq \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t \left(b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s) \right) ds \right| \\ &\quad + \alpha \left(U_u^\varepsilon - U_{t_k}^\varepsilon \right) \\ &\leq \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t \left(b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s) \right) ds \right| \\ &\quad + \alpha \sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_{t_k}^\varepsilon|. \end{aligned}$$

This implies

$$\begin{aligned} &\sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_t^{\varepsilon,n}| \\ &= \sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_{t_k}^\varepsilon| \\ &\leq C_2 \sup_{t_k \leq t < t_{k+1}} \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t \left(b(U_s^\varepsilon) + \sigma(U_s^\varepsilon) \dot{h}(s) \right) ds \right|, \end{aligned} \quad (2.15)$$

where $C_2 := \frac{1}{1-\alpha}$.

Since $\sup_{t_k \leq s < t_{k+1}} \varepsilon |\sigma(U_s^\varepsilon)|^2 \leq \varepsilon M^2$, $\sup_{t_k \leq s < t_{k+1}} |b(U_s^\varepsilon)| \leq M$ and

$$\sup_{t_k \leq t < t_{k+1}} \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \dot{h}(s) ds \right| \leq M \sqrt{\frac{2a}{n}},$$

it follows from (2.15) and Lemma 2.1 that for $n \geq n_1 := \left\lceil \frac{2MC_2}{\delta_1} \right\rceil + 1 \vee n_2 :=$

$$\begin{aligned}
& \left\lceil \frac{8C_2^2 M^2 a}{\delta_1^2} \right\rceil + 1, \\
& \mathbf{P} \left(\sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_t^{\varepsilon, n}| > \delta_1 \right) \\
& \leq \mathbf{P} \left(\sup_{t_k \leq t < t_{k+1}} \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \sqrt{\varepsilon} dB_s + \int_{t_k}^t b(U_s^\varepsilon) ds \right| > \frac{\delta_1}{2C_2} \right) \\
& \quad + \mathbf{P} \left(\sup_{t_k \leq t < t_{k+1}} \left| \int_{t_k}^t \sigma(U_s^\varepsilon) \dot{h}(s) ds \right| > \frac{\delta_1}{2C_2} \right) \\
& \leq \exp \left(-n \frac{\delta_1^2}{8C_2^2 M^2 \varepsilon} \right). \tag{2.16}
\end{aligned}$$

Therefore

$$\begin{aligned}
\mathbf{P}(C^\varepsilon) &= \mathbf{P} \left(\max_{k \in \{0, 1, \dots, n-1\}} \sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_t^{\varepsilon, n}| > \delta_1 \right) \\
&\leq \sum_{k=0}^{n-1} \mathbf{P} \left(\sup_{t_k \leq t < t_{k+1}} |U_t^\varepsilon - U_t^{\varepsilon, n}| > \delta_1 \right) \\
&\leq n \exp \left(-n \frac{\delta_1^2}{8C_2^2 M^2 \varepsilon} \right). \tag{2.17}
\end{aligned}$$

Now given $R > 0$, choose first $\delta_1 > 0$ such that (2.12) holds. Then choose n large enough so that

$$n \exp \left(-n \frac{\delta_1^2}{8C_2^2 M^2 \varepsilon} \right) \leq \exp \left(-\frac{R}{\varepsilon} \right).$$

Finally for the fixed n , there exists $\rho > 0$ such that $D^\varepsilon = \emptyset$. Combining above inequalities proves (2.11).

Next, we prove the map $F : H \rightarrow C_x([0, 1], \mathbf{R})$ is continuous on $\{h : \tilde{I}(h) \leq a\}$ for $a \in [0, \infty)$. Let $h_n, h \in \{h \in C_0([0, 1], \mathbf{R}) : \tilde{I}(h) \leq a\}$ with $h_n \rightarrow h$ in $C_0([0, 1], \mathbf{R})$. Since $\{h : \tilde{I}(h) \leq a\}$ is weakly compact in H , we conclude that h_n also weakly converges to h in H . By the Lipschitz continuity,

$$\begin{aligned}
& \sup_{0 \leq s \leq t} |F(h_n)(s) - F(h)(s)| \\
& \leq \xi_n(t) + \alpha \sup_{0 \leq s \leq t} |F(h_n)(s) - F(h)(s)| \\
& \quad + L \int_0^t \sup_{0 \leq u \leq s} |F(h_n)(u) - F(h)(u)| (1 + |\dot{h}_n(s)|) ds,
\end{aligned}$$

where

$$\xi_n(t) = \sup_{0 \leq s \leq t} \left| \int_0^s \sigma(F(h)(u))(\dot{h}_n(u) - \dot{h}(u)) du \right| \quad (2.18)$$

Applying the Gronwall's lemma,

$$\begin{aligned} & \sup_{0 \leq t \leq 1} |F(h_n)(t) - F(h)(t)| \\ & \leq \frac{2M}{1-\alpha} \xi_n(1) \times \exp \left(\frac{L \left(1 + \sqrt{2\tilde{I}(h_n)} \right)}{1-\alpha} \right) \\ & \leq \frac{2M}{1-\alpha} \exp \left(\frac{L \left(1 + \sqrt{2a} \right)}{1-\alpha} \right) \xi_n(1). \end{aligned}$$

Since $h_n \rightarrow h$ weakly in H , for every $s \in [0, 1]$,

$$\int_0^s \sigma(F(h)(u))(\dot{h}_n(u) - \dot{h}(u)) du \rightarrow 0, \quad \text{as } n \rightarrow \infty.$$

On the other hand, it is easy to see that there exists a constant C_a such that

$$\left| \int_s^t \sigma(F(h)(u))(\dot{h}_n(u) - \dot{h}(u)) du \right| \leq C_a |t - s|^{\frac{1}{2}}.$$

Therefore, we conclude that $\xi_n(1) \rightarrow 0$, as $n \rightarrow \infty$. Consequently,

$$\sup_{0 \leq t \leq 1} |F(h_n)(t) - F(h)(t)| \rightarrow 0, \quad \text{as } n \rightarrow \infty.$$

which proves the continuity of the mapping F . Now, letting $\Phi_1^\varepsilon = \sqrt{\varepsilon}B$ and $K(\cdot) = F(\cdot)$, Theorem 2.1 follows from Proposition 1.1. \square

3 LDP for perturbed reflected diffusion processes

In this section, we will prove the large deviation principle for the solution of the perturbed reflected diffusion equation (1.4).

For $y \geq 0$ and $f \in C_y([0, 1], \mathbf{R})$, define two operators $\Gamma : C_y([0, 1], \mathbf{R}) \rightarrow C_y([0, 1], \mathbf{R}_+)$ and $K : C_y([0, 1], \mathbf{R}) \rightarrow C_0([0, 1], \mathbf{R}_+)$ by

$$\Gamma f = f + \tilde{f} \quad \text{and} \quad Kf = \tilde{f}, \quad \text{where} \quad \tilde{f}(t) := - \inf_{s \leq t} (f(s) \wedge 0), \quad t \in [0, 1].$$

By the reflection principle, the solution Y^ε of (1.4) is given by

$$Y_t^\varepsilon = (\Gamma Z^\varepsilon)(t) \quad \text{and} \quad L_t^\varepsilon = (KZ^\varepsilon)(t), \quad t \in [0, 1], \quad (3.1)$$

where Z^ε is a solution of the following stochastic equation

$$Z_t^\varepsilon = y + \sqrt{\varepsilon} \int_0^t \sigma((\Gamma Z^\varepsilon)(s)) dB_s + \alpha \sup_{0 \leq s \leq t} (\Gamma Z^\varepsilon)(s), \quad t \in [0, 1].$$

For $h \in H$, let $G(h)$ be the solution of the following equation

$$G(h)(t) = y + \int_0^t \sigma(G(h)(s)) \dot{h}_s ds + \alpha \sup_{0 \leq s \leq t} G(h)(s) + \eta_t, \quad t \in [0, 1], \quad (3.2)$$

where $G(h)$ is continuous, nonnegative, and η is an increasing continuous function satisfying $\eta_t = \int_0^t \chi_{\{G(h)(s)=0\}} d\eta_s$.

Similar as (3.1), $G(h)$ can also be written as

$$G(h)(t) = (\Gamma V(h))(t) \quad \text{and} \quad \eta_t = (KV(h))(t), \quad t \in [0, 1], \quad (3.3)$$

where $V(h)$ satisfies

$$V(h)(t) = y + \int_0^t \sigma((\Gamma V(h))(s)) \dot{h}_s ds + \alpha \sup_{0 \leq s \leq t} (\Gamma V(h))(s), \quad t \in [0, 1].$$

Let ν_ε^1 be the law of Y^ε on $C_y([0, 1], \mathbf{R}_+)$. Our main result in this section is

Theorem 3.1 *If $\alpha \in (0, \frac{1}{2})$, then the family $\{\nu_\varepsilon^1, \varepsilon > 0\}$ obeys a large deviation principle with the rate function*

$$I(g) = \inf \left\{ \tilde{I}(h); G(h) = g \right\}, \quad g \in C_y([0, 1], \mathbf{R}_+).$$

By Proposition 1.1, Theorem 3.1 is the consequence of the following two propositions.

Proposition 3.1 *If $\alpha \in (0, \frac{1}{2})$, then for any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in C_0([0, 1], \mathbf{R})$ satisfying $\tilde{I}(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,*

$$\begin{aligned} & \mathbf{P} \left(\sup_{0 \leq t \leq 1} |Y_t^\varepsilon - G(h)(t)| + \sup_{0 \leq t \leq 1} |L_t^\varepsilon - \eta_t| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t - h(t)| \leq \rho \right) \\ & \leq \exp \left(-\frac{R}{\varepsilon} \right). \end{aligned}$$

Proof. The proof is similar to that of Theorem 2.1. We only highlight the differences. Note that for $f_1, f_2 \in C_y([0, 1], \mathbf{R})$ and $t \in [0, 1]$,

$$\sup_{0 \leq s \leq t} |(\Gamma f_1)(s) - (\Gamma f_2)(s)| \leq 2 \sup_{0 \leq s \leq t} |f_1(s) - f_2(s)|. \quad (3.4)$$

Then by (3.1) and (3.3),

$$\sup_{0 \leq t \leq 1} |Y_t^\varepsilon - G(h)(t)| + \sup_{0 \leq t \leq 1} |L_t^\varepsilon - \eta_t| \leq 3 \sup_{0 \leq t \leq 1} |Z_t^\varepsilon - V(h)(t)|.$$

Therefore, the proof of Proposition 3.1 is reduced to show that for any $R, \delta, a > 0$, there exist $\rho > 0$ and $\varepsilon_0 > 0$ such that, for any $h \in C_0([0, 1], \mathbf{R})$ satisfying $\tilde{I}(h) \leq a$ and $\varepsilon \leq \varepsilon_0$,

$$\mathbf{P} \left(\sup_{0 \leq t \leq 1} |Z_t^\varepsilon - V(h)(t)| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t - h(t)| \leq \rho \right) \leq \exp \left(-\frac{R}{\varepsilon} \right). \quad (3.5)$$

Using (3.4) and the Gronwall's lemma,

$$\sup_{0 \leq t \leq 1} |Z_t^\varepsilon - V(h)(t)| \leq C_3 \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma((\Gamma Z^\varepsilon)(s)) \sqrt{\varepsilon} dB_s^\varepsilon \right|,$$

where $C_3 := \frac{\exp(\frac{2L}{1-2\alpha} \|h\|_H)}{1-2\alpha}$. Thus as in section 2, by the Girsanov's theorem, to prove (3.5), it suffices to prove

$$\mathbf{P} \left(\sup_{0 \leq t \leq 1} \left| \int_0^t \sigma((\Gamma \bar{Z}^\varepsilon)(s)) \sqrt{\varepsilon} dB_s \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right) \leq \exp \left(-\frac{R}{\varepsilon} \right),$$

where \bar{Z}^ε satisfies

$$\begin{aligned} \bar{Z}_t^\varepsilon &= y + \sqrt{\varepsilon} \int_0^t \sigma((\Gamma \bar{Z}^\varepsilon)(s)) dB_s + \int_0^t \sigma((\Gamma \bar{Z}^\varepsilon)(s)) \dot{h}(s) ds \\ &\quad + \alpha \sup_{0 \leq s \leq t} (\Gamma \bar{Z}^\varepsilon)(s), \quad t \in [0, 1]. \end{aligned}$$

For a function f , put $\Gamma(f)^{(n)}(t) = \Gamma(f)(t_k)$, if $t_k \leq t < t_{k+1}$, $k \in \{0, 1, \dots, n-1\}$. Then

$$\begin{aligned} \tilde{A}^\varepsilon &:= \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma((\Gamma \bar{Z}^\varepsilon)(s)) \sqrt{\varepsilon} dB_s \right| \geq \delta, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon} B_t| \leq \rho \right\} \\ &\subset \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t (\sigma((\Gamma \bar{Z}^\varepsilon)(s)) - \sigma((\Gamma \bar{Z}^\varepsilon)^{(n)}(s))) \sqrt{\varepsilon} dB_s \right| \geq \frac{\delta}{2}, \right. \end{aligned}$$

$$\begin{aligned}
& \left. \sup_{0 \leq t \leq 1} |\Gamma(\bar{Z}^\varepsilon)(t) - \Gamma(\bar{Z}^\varepsilon)^{(n)}(t)| \leq \delta_1 \right\} \\
& \cup \left\{ \sup_{0 \leq t \leq 1} |\Gamma(\bar{Z}^\varepsilon)(t) - \Gamma(\bar{Z}^\varepsilon)^{(n)}(t)| > \delta_1 \right\} \\
& \cup \left\{ \sup_{0 \leq t \leq 1} \left| \int_0^t \sigma((\Gamma(\bar{Z}^\varepsilon)^{(n)}(s))\sqrt{\varepsilon}dB_s \right| \geq \frac{\delta}{2}, \sup_{0 \leq t \leq 1} |\sqrt{\varepsilon}B_t| \leq \rho \right\} \\
& := \tilde{B}^\varepsilon \cup \tilde{C}^\varepsilon \cup \tilde{D}^\varepsilon,
\end{aligned}$$

Let us just sketch the proof of

$$\mathbf{P}(\tilde{C}^\varepsilon) \leq \exp\left(-\frac{R}{\varepsilon}\right). \quad (3.6)$$

For $t \in [t_k, t_{k+1})$, it is easy to see that

$$\begin{aligned}
& \sup_{t_k \leq s \leq t} |\Gamma(\bar{Z}^\varepsilon)(s) - \Gamma(\bar{Z}^\varepsilon)^{(n)}(s)| \\
& = \sup_{t_k \leq s \leq t} |\Gamma(\bar{Z}^\varepsilon)(s) - \Gamma(\bar{Z}^\varepsilon)(t_k)| \\
& \leq 2 \sup_{t_k \leq s \leq t} |\bar{Z}^\varepsilon(s) - \bar{Z}^\varepsilon(t_k)|.
\end{aligned} \quad (3.7)$$

But,

$$\begin{aligned}
\left| \bar{Z}_t^\varepsilon - \bar{Z}_{t_k}^\varepsilon \right| & \leq \left| \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\sqrt{\varepsilon}dB_s + \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\dot{h}(s)ds \right| \\
& \quad + \alpha \left| \sup_{0 \leq s \leq t} (\Gamma\bar{Z}^\varepsilon)(s) - \sup_{0 \leq s \leq t_k} (\Gamma\bar{Z}^\varepsilon)(s) \right| \\
& \leq \left| \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\sqrt{\varepsilon}dB_s + \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\dot{h}(s)ds \right| \\
& \quad + \alpha \sup_{t_k \leq s \leq t} |\Gamma(\bar{Z}^\varepsilon)(s) - \Gamma(\bar{Z}^\varepsilon)(t_k)| \\
& \leq \left| \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\sqrt{\varepsilon}dB_s + \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\dot{h}(s)ds \right| \\
& \quad + 2\alpha \sup_{t_k \leq s \leq t} |\bar{Z}^\varepsilon(s) - \bar{Z}^\varepsilon(t_k)|.
\end{aligned} \quad (3.8)$$

Therefore

$$\begin{aligned}
& \sup_{t_k \leq t < t_{k+1}} \left| \bar{Z}_t^\varepsilon - \bar{Z}_{t_k}^\varepsilon \right| \\
& \leq \frac{1}{1-2\alpha} \sup_{t_k \leq t < t_{k+1}} \left| \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\sqrt{\varepsilon}dB_s + \int_{t_k}^t \sigma((\Gamma\bar{Z}^\varepsilon)(s))\dot{h}(s)ds \right|.
\end{aligned}$$

The rest of the proof is the same as the proof of (2.17) of Theorem 2.1. We omit it. \square

Proposition 3.2 *If $\alpha \in (0, \frac{1}{2})$, then $G : H \rightarrow C_y([0, 1], \mathbf{R}_+)$ and $\eta : H \rightarrow C_0([0, 1], \mathbf{R}_+)$ defined by (3.3) are continuous on the compact set $\{h \in C_0([0, 1], \mathbf{R}) : \tilde{I}(h) \leq a\}$ for $a \in [0, \infty)$.*

Proof. By (3.4), it suffices to prove that the conclusion holds for $V : H \rightarrow C_y([0, 1], \mathbf{R})$. Let $h_n, h \in \{h \in C_0([0, 1], \mathbf{R}) : \tilde{I}(h) \leq a\}$ with $h_n \rightarrow h$ in $C_0([0, 1], \mathbf{R})$. Then we have

$$\begin{aligned} & \sup_{0 \leq s \leq t} |V(h_n)(t) - V(h)(t)| \\ & \leq \frac{2L}{1 - 2\alpha} \int_0^t \sup_{0 \leq u \leq s} |V(h_n)(s) - V(h)(s)| |\dot{h}_n(s)| ds + \frac{1}{1 - 2\alpha} \xi_n(t), \end{aligned}$$

where

$$\xi_n(t) := \sup_{0 \leq s \leq t} \left| \int_0^s \sigma((\Gamma V(h))(s)) (\dot{h}_n(s) - \dot{h}(s)) ds \right|.$$

The rest of the proof is similar to the last part of the proof of Theorem 2.1. \square

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References

- [1] R. Azencott (1980): Grandes déviations et applications. *Ecole d'Eté de Probabilités de Saint-Flour VIII, 1978. Lecture Notes in Math.* **779** 1-176. Springer, New York.
- [2] P. Carmona, F. Petit and M. Yor (1994): Some extentions of the arc-sine law as (partial) consequences of the scaling property of Brownian motion. *Probab. Theory Relat. Fields* **100** 1-29.
- [3] P. Carmona, F. Petit and M. Yor (1998): Beta variables as times spent in $[0, \infty)$ by certain perturbed Brownian motions. *J. London Math. Soc.* **58** 239-256.
- [4] L. Chaumont and R. Doney (1999) Pathwise uniqueness for perturbed versions of Brownian motion and reflected Brownian motion. *Probab. Theory Relat. Fields* **113** 519-534.

- [5] R. Doney (1998): Some calculations for perturbed Brownian motion. *Sém. Probab. XXXII, Lecture Notes Math.* **1686** 231-236. Springer, New York.
- [6] R. Doney and T. Zhang (2005): Perturbed Skorohod equations and perturbed reflected diffusion processes. *Ann. I. H. Poincaré-PR* **41** 107-121.
- [7] H. Doss and P. Priouret (1983): Petites perturbations de systems dynamiques avec reflexion. *Lecture Notes in Math.* No. 986, Springer, New York, 1983.
- [8] J.F. Le Gall and M. Yor (1986): Excursions browniennes et carrés de processus de Bessel. *C. R. Acad. Sci. Paris Sér. I* **303** 73-76.
- [9] J.F. Le Gall and M. Yor (1990): Enlacements du mouvement brownien autour des courbes de l'espace. *Trans. Amer. Math. Soc.* **317** 687-722.
- [10] A. Millet, D. Nualart and M. Sanz (1992): Large deviations for a class of anticipating stochastic differential equations. *Ann. Probab.* **20**(4) 1902-1931.
- [11] A. Mohammed and T. Zhang (2006): Large deviations for stochastic systems with memory. *Disc. and Contin. Dyn. Sys.-Series B* **6**(4) 881-893.
- [12] M. Perman and W. Werner (1997): Perturbed Brownian motions. *Probability Theory and Related Fields* **108** 357-383.
- [13] W. Stroock (1983): Some applications of stochastic calculus to partial differential equations. *Ecole d'Eté de Probabilités de Saint-Flour XI, 1981. Lecture Notes in Math.* **976** Springer, New York.