

## DEVIATION INEQUALITIES AND MODERATE DEVIATIONS FOR ESTIMATORS OF PARAMETERS IN AN ORNSTEIN-UHLENBECK PROCESS WITH LINEAR DRIFT

FUQING GAO<sup>1</sup>

*School of Mathematics and Statistics, Wuhan University, Wuhan 430072, P.R.China*

email: fqgao@whu.edu.cn

HUI JIANG

*School of Science, Nanjing University of Aeronautics and Astronautics, Nanjing 210016, P.R.China*

email: huijiang@nuaa.edu.cn

*Submitted December 29, 2008, accepted in final form April 21, 2009*

AMS 2000 Subject classification: 60F12, 62F12, 62N02

Keywords: Deviation inequality, logarithmic Sobolev inequality, moderate deviations, Ornstein-Uhlenbeck process

### *Abstract*

Some deviation inequalities and moderate deviation principles for the maximum likelihood estimators of parameters in an Ornstein-Uhlenbeck process with linear drift are established by the logarithmic Sobolev inequality and the exponential martingale method.

## 1 Introduction and main results

### 1.1 Introduction

We consider the following Ornstein-Uhlenbeck process

$$dX_t = (-\theta X_t + \gamma)dt + dW_t, \quad X_0 = x \tag{1.1}$$

where  $W$  is a standard Brownian motion and  $\theta, \gamma$  are unknown parameters with  $\theta \in (0, +\infty)$ . We denote by  $P_{\theta, \gamma, x}$  the distribution of the solution of (1.1).

It is known that the maximum likelihood estimators (MLE) of the parameters  $\theta$  and  $\gamma$  are (cf.

---

<sup>1</sup>RESEARCH SUPPORTED BY THE NATIONAL NATURAL SCIENCE FOUNDATION OF CHINA (10871153)

[15])

$$\begin{aligned}\hat{\theta}_T &= \frac{-T \int_0^T X_t dX_t + (X_T - x) \int_0^T X_t dt}{T \int_0^T X_t^2 dt - \left( \int_0^T X_t dt \right)^2} \\ &= \theta + \frac{W_T \hat{\mu}_T - \int_0^T X_t dW_t}{T \hat{\sigma}_T^2},\end{aligned}\quad (1.2)$$

$$\begin{aligned}\hat{\gamma}_T &= \frac{-\int_0^T X_t dt \int_0^T X_t dX_t + (X_T - x) \int_0^T X_t^2 dt}{T \int_0^T X_t^2 dt - \left( \int_0^T X_t dt \right)^2} \\ &= \gamma + \frac{W_T}{T} + \frac{\hat{\mu}_T (W_T \hat{\mu}_T - \int_0^T X_t dW_t)}{T \hat{\sigma}_T^2},\end{aligned}\quad (1.3)$$

where

$$\hat{\mu}_T = \frac{1}{T} \int_0^T X_t dt, \quad \hat{\sigma}_T^2 = \frac{1}{T} \int_0^T X_t^2 dt - \hat{\mu}_T^2. \quad (1.4)$$

It is known that  $\hat{\theta}_T$  and  $\hat{\gamma}_T$  are consistent estimators of  $\theta$  and  $\gamma$  and have asymptotic normality (cf. [15]).

For  $\gamma \equiv 0$  case, Florens-Landais and Pham ([9]) calculated the Laplace functional of  $(\int_0^T X_t dX_t, \int_0^T X_t^2 dt)$  by Girsanov's formula and obtained large deviations for  $\hat{\theta}_T$  by Gärtner-Ellis theorem. Bercu and Rouault ([1]) presented a sharp large deviation for  $\hat{\theta}_T$ . Lezaud ([14]) obtained the deviation inequality of quadratic functional of the classical OU processes. We refer to [8] and [11] for the moderate deviations of some non-linear functionals of moving average processes and diffusion processes. In this paper we use the logarithmic Sobolev inequality (LSI) to study the deviation inequalities and the moderate deviations of  $\hat{\theta}_T$  and  $\hat{\gamma}_T$  for  $\gamma \neq 0$  case.

## 1.2 Main results

Throughout this paper, let  $\lambda_T, T \geq 1$  be a positive sequence satisfying

$$\lambda_T \rightarrow \infty, \quad \frac{\lambda_T}{\sqrt{T}} \rightarrow 0. \quad (1.5)$$

**Theorem 1.1.** *There exist finite positive constants  $C_0, C_1, C_2$  and  $C_3$  such that for all  $r > 0$  and all  $T \geq 1$ ,*

$$\begin{aligned}P_{\theta, \gamma, x} \left( |\hat{\theta}_T - \theta| \geq r \right) &\leq C_0 \exp \left\{ -C_1 r T E_{\theta, \gamma, x} (\hat{\sigma}_T^2) \min \{1, C_2 r\} \right\} \\ &\quad + C_0 \exp \left\{ -C_3 T E_{\theta, \gamma, x} (\hat{\sigma}_T^2) \right\}\end{aligned}$$

and

$$\begin{aligned}P_{\theta, \gamma, x} \left( |\hat{\gamma}_T - \gamma| \geq r \right) &\leq C_0 \exp \left\{ -C_1 r T E_{\theta, \gamma, x} (\hat{\sigma}_T^2) \min \{1, C_2 r\} \right\} \\ &\quad + C_0 \exp \left\{ -C_3 T E_{\theta, \gamma, x} (\hat{\sigma}_T^2) \right\}.\end{aligned}$$

**Remark 1.1.** *In this theorem and the remainder of the paper, all the constants involved depend on  $\theta, \gamma$  and the initial point  $x$ .*

**Theorem 1.2.** (1).  $\left\{P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\theta}_T - \theta) \in \cdot\right), T \geq 1\right\}$  satisfies the large deviation principle with speed  $\lambda_T$  and rate function  $I_1(u) = \frac{u^2}{4\theta}$ , that is, for any closed set  $F$  in  $\mathbb{R}$ ,

$$\limsup_{n \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\theta}_T - \theta) \in F\right) \leq -\inf_{u \in F} \frac{u^2}{4\theta}$$

and open set  $G$  in  $\mathbb{R}$ ,

$$\liminf_{n \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\theta}_T - \theta) \in G\right) \geq -\inf_{u \in G} \frac{u^2}{4\theta}.$$

(2).  $\left\{P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\gamma}_T - \gamma) \in \cdot\right), T \geq 1\right\}$  satisfies the large deviation principle with speed  $\lambda_T$  and rate function  $I_2(u) = \frac{\theta u^2}{2(\theta + 2\gamma^2)}$ , that is, for any closed set  $F$  in  $\mathbb{R}$ ,

$$\limsup_{n \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\gamma}_T - \gamma) \in F\right) \leq -\inf_{u \in F} \frac{\theta u^2}{2(\theta + 2\gamma^2)}$$

and open set  $G$  in  $\mathbb{R}$ ,

$$\liminf_{n \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta,\gamma,x}\left(\sqrt{\frac{T}{\lambda_T}}(\hat{\gamma}_T - \gamma) \in G\right) \geq -\inf_{u \in G} \frac{\theta u^2}{2(\theta + 2\gamma^2)}.$$

In  $\gamma = 0$  case, the deviation inequalities of quadratic functionals of the classical OU process are obtained in [14]. For the large deviations and the moderate deviations of  $\hat{\theta}_T$ , we refer to [1], [9] and [11]. The proofs of Theorem 1.1 and Theorem 1.2 are based on the LSI with respect to  $L^2$ -norm in the Wiener space and Herbst's argument (cf. [10], [12]).

## 2 Deviation inequalities

In this section, we give some deviation inequalities for the estimators  $\hat{\theta}_T$  and  $\hat{\gamma}_T$  by the logarithmic Sobolev inequality and the exponential martingale method. For deviation bounds for additive functionals of Markov processes, we refer to [3] and [18].

### 2.1 Moments

It is known that the solution of equation (1.1) has the following expression:

$$X_t = \left(x - \frac{\gamma}{\theta}\right) e^{-\theta t} + \frac{\gamma}{\theta} + e^{-\theta t} \int_0^t e^{\theta s} dW_s. \quad (2.1)$$

From this expression, it is easily seen that for any  $t \geq 0$ ,

$$\mu_t := E_{\theta,\gamma,x}(X_t) = \left(x - \frac{\gamma}{\theta}\right) e^{-\theta t} + \frac{\gamma}{\theta}, \quad (2.2)$$

$$\sigma_t^2 := \text{Var}_{\theta,\gamma,x}(X_t) = \frac{1}{2\theta}(1 - e^{-2\theta t}) \quad (2.3)$$

and for any  $0 \leq s \leq t$ ,

$$\text{Cov}_{\theta,\gamma,x}(X_s, X_t) = \frac{1}{2\theta}(1 - e^{-2\theta s})e^{-\theta(t-s)}. \quad (2.4)$$

Therefore

$$E_{\theta,\gamma,x}(\hat{\mu}_T) = \frac{1}{T}E_{\theta,\gamma,x}\left(\int_0^T X_t dt\right) = \frac{1}{\theta T}\left(x - \frac{\gamma}{\theta}\right)(1 - e^{-\theta T}) + \frac{\gamma}{\theta}, \quad (2.5)$$

$$\begin{aligned} \text{Var}_{\theta,\gamma,x}(\hat{\mu}_T) &= \frac{1}{T^2}E_{\theta,\gamma,x}\left(\left(\int_0^T e^{-\theta t} \int_0^t e^{\theta s} dW_s dt\right)^2\right) \\ &= \frac{1}{\theta^2 T^2}\left(T - \frac{1}{2\theta}(e^{-2\theta T} - 1) + \frac{2}{\theta}(e^{-\theta T} - 1)\right) \end{aligned} \quad (2.6)$$

and so for all  $T \geq 1$ ,

$$\text{Var}_{\theta,\gamma,x}(\hat{\mu}_T) \leq \frac{1}{2\theta^3 T}(2\theta + 1) \quad (2.7)$$

and

$$\begin{aligned} E_{\theta,\gamma,x}(\hat{\sigma}_T^2) &= \frac{1}{2\theta} + \frac{1}{4\theta^2 T}(1 - e^{-2\theta T})\left(-1 + 2\theta\left(x - \frac{\gamma}{\theta}\right)^2\right) \\ &\quad - \frac{1}{\theta^2 T^2}(1 - e^{-\theta T})^2\left(x - \frac{\gamma}{\theta}\right)^2(1 - e^{-\theta T}) \\ &\quad - \frac{1}{\theta^2 T^2}\left(T - \frac{1}{2\theta}(e^{-2\theta T} - 1) + \frac{2}{\theta}(e^{-\theta T} - 1)\right) \end{aligned}$$

which implies

$$\left|E_{\theta,\gamma,x}(\hat{\sigma}_T^2) - \frac{1}{2\theta}\right| \leq \frac{1}{\theta^2 T}\left(\theta\left(x - \frac{\gamma}{\theta}\right)^2 + \frac{2}{\theta}\right). \quad (2.8)$$

**Lemma 2.1.** For any  $0 \leq \alpha \leq \theta^2/4$ , for all  $T \geq 1$ ,

$$E_{\theta,\gamma,x}\left(\exp\left(\alpha \int_0^T X_t^2 dt\right)\right) < \infty,$$

and there exist finite positive constants  $L_1$  and  $L_2$  such that for all  $0 \leq \alpha \leq \theta^2/4$  and  $T \geq 1$ ,

$$E_{\theta,\gamma,x}\left(\exp\left(\alpha \int_0^T X_t^2 dt\right)\right) \leq L_1 e^{L_2 \alpha T}.$$

*Proof.* For any  $0 \leq \alpha \leq \theta^2/4$ , set  $\kappa = \sqrt{\theta^2 - 2\alpha}$ . Then by Girsanov theorem, we have

$$\frac{dP_{\theta,\gamma,x}}{dP_{\kappa,\gamma,x}} = \exp\left\{-\int_0^T (\theta - \kappa)X_t dX_t - \int_0^T (\alpha X_t^2 - \gamma(\theta - \kappa)X_t)dt\right\}$$

and so

$$\begin{aligned}
& E_{\theta, \gamma, x} \left( \exp \left( \alpha \int_0^T X_t^2 dt \right) \right) \\
&= E_{\kappa, \gamma, x} \left( \frac{dP_{\theta, \gamma, x}}{dP_{\kappa, \gamma, x}} \exp \left\{ \alpha \int_0^T X_t^2 dt \right\} \right) \\
&= E_{\kappa, \gamma, x} \left( \exp \left\{ (-\theta + \kappa) \int_0^T X_t dX_t + \gamma \int_0^T (\theta - \kappa) X_t dt \right\} \right) \\
&= E_{\kappa, \gamma, x} \left( \exp \left\{ \frac{-(\theta - \kappa)}{2} (X_T^2 - T) + \gamma \int_0^T (\theta - \kappa) X_t dt \right\} \right) \\
&\leq \exp \left\{ \frac{(\theta - \kappa)T}{2} \right\} E_{\kappa, \gamma, x} \left( \exp \left\{ \gamma \int_0^T (\theta - \kappa) X_t dt \right\} \right)
\end{aligned}$$

where the last inequality is due to  $\theta \geq \kappa$ . Now we have to estimate  $E_{\kappa, \gamma, x}(\exp\{\gamma \int_0^T (\theta - \kappa) X_t dt\})$ . Since under  $P_{\kappa, \gamma, x}$ ,

$$\hat{\mu}_T \sim N \left( \frac{1}{\kappa T} (x - \frac{\gamma}{\kappa})(1 - e^{-\kappa T}) + \frac{\gamma}{\kappa}, \frac{1}{\kappa^2 T^2} \left( T - \frac{1}{2\kappa}(e^{-2\kappa T} - 1) + \frac{2}{\kappa}(e^{-\kappa T} - 1) \right) \right),$$

we have

$$\begin{aligned}
& E_{\kappa, \gamma, x} \left( \exp \left\{ \gamma \int_0^T (\theta - \kappa) X_t dt \right\} \right) \\
&= \exp \left\{ \frac{\gamma(\theta - \kappa)}{\kappa} \left( \left( x - \frac{\gamma}{\kappa} \right) (1 - e^{-\kappa T}) + \gamma T \right) \right\} \\
&\quad \cdot \exp \left\{ \frac{\gamma^2(\theta - \kappa)^2}{2\kappa^2} \left( T - \frac{1}{2\kappa}(e^{-2\kappa T} - 1) + \frac{2}{\kappa}(e^{-\kappa T} - 1) \right) \right\}.
\end{aligned}$$

Noting  $\theta/\sqrt{2} \leq \kappa \leq \theta$ ,  $0 \leq \theta - \kappa = 2\alpha/(\theta + \kappa) \leq 2\alpha/\theta$  and  $(\theta - \kappa)^2 \leq \alpha\theta$  for all  $0 \leq \alpha \leq \theta^2/4$ , we complete the proof of the lemma.  $\square$

## 2.2 Logarithmic Sobolev inequality

Since the LSI with respect to the Cameron-Martin metric does not produce the concentration inequality of correct order in large time  $T$  for the functionals

$$F(X) := \frac{1}{\sqrt{T}} \left( \int_0^T g(X_s) ds - \mathbb{E} \left( \int_0^T g(X_s) ds \right) \right),$$

in order to get the concentration inequality of correct order for the functionals  $F(X)$ , as pointed out by Djellout, Guillin and Wu ([7]) we should establish the LSI with respect to the  $L^2$ -metric.

Let us introduce the logarithmic Sobolev inequality on  $W$  with respect to the gradient in  $L^2([0, T], \mathbb{R})$  ([10]). Let  $\mu$  be the Wiener measure on  $W = C([0, T], \mathbb{R})$ . A function  $f : W \rightarrow \mathbb{R}$  is said to be

differentiable with respect to the  $L^2$ -norm, if it can be extend to  $L^2([0, T], \mathbb{R})$  and for any  $w \in W$ , there exists a bounded linear operator  $Df(w) : g \rightarrow D_g f(w)$  on  $L^2([0, T], \mathbb{R})$  such that

$$\lim_{\|g\|_{L^2} \rightarrow 0} \frac{|f(w + g) - f(w) - D_g f(w)|}{\|g\|_{L^2}} = 0.$$

If  $f : W \rightarrow \mathbb{R}$  is differentiable with respect to the  $L^2$ -norm, then there exists a unique element  $\nabla f(w) = (\nabla_t f(w), t \in [0, T])$  in  $L^2([0, T], \mathbb{R})$  such that

$$D_g f(w) = \langle \nabla f(w), g \rangle_{L^2}, \text{ for all } g \in L^2([0, T], \mathbb{R}).$$

Denote by  $C_b^1(W/L^2)$  the space of all bounded function  $f$  on  $W$ , differentiable with respect to the  $L^2$ -norm, such that  $\nabla f$  is also continuous and bounded from  $W$  equipped with  $L^2$ -norm to  $L^2([0, T], \mathbb{R})$ . Applying Theorem 2.3 in [10] to the Ornstein-Uhlenbeck process with linear drift, we have

$$Ent_{P_{\theta, \gamma, x}}(f^2) \leq \frac{2}{\theta^2} E_{\theta, \gamma, x} \left( \int_0^T |\nabla_t f|^2 dt \right), \quad f \in C_b^1(W/L^2) \tag{2.9}$$

where the entropy of  $f^2$  is given by

$$Ent_{P_{\theta, \gamma, x}}(f^2) = E_{\theta, \gamma, x}(f^2 \log f^2) - E_{\theta, \gamma, x}(f^2) \log E_{\theta, \gamma, x}(f^2).$$

**Lemma 2.2.** For any  $|\alpha| \leq \theta^2/4$ ,

$$E_{\theta, \gamma, x} \left( \exp \left\{ \alpha \left( \int_0^T X_t^2 dt - E_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt \right) \right) \right\} \right) \leq E_{\theta, \gamma, x} \left( \exp \left\{ \frac{4\alpha^2}{\theta^2} \int_0^T X_t^2 dt \right\} \right)$$

and

$$E_{\theta, \gamma, x} \left( \exp \left\{ \alpha T (\hat{\mu}_T^2 - E_{\theta, \gamma, x}(\hat{\mu}_T^2)) \right\} \right) \leq E_{\theta, \gamma, x} \left( \exp \left\{ \frac{4\alpha^2}{\theta^2} \int_0^T X_t^2 dt \right\} \right).$$

*Proof.* We apply Theorem 2.7 in [12] to prove the conclusions of the lemma. Take  $\mathcal{A}_1 = \{\alpha f; |\alpha| \leq \theta^2/4\}$  and  $\mathcal{A}_2 = \{ah; |\alpha| \leq \theta^2/4\}$ , where

$$f(w) = \int_0^T w_t^2 dt, \quad h(w) = \frac{1}{T} \left( \int_0^T w_t dt \right)^2.$$

Define

$$\Gamma_1(g_1) = \frac{4}{\theta^2} \frac{g_1^2}{f}, \quad g_1 \in \mathcal{A}_1; \quad \Gamma_2(g_2) = \frac{4}{\theta^2} \frac{g_2^2}{h}, \quad g_2 \in \mathcal{A}_2.$$

Then for any  $\lambda \in [-1, 1]$ ,  $g_1 \in \mathcal{A}_1$  and  $g_2 \in \mathcal{A}_2$ ,  $\lambda g_1 \in \mathcal{A}_1$ ,  $\lambda g_2 \in \mathcal{A}_2$ ,  $\Gamma_1(\lambda g_1) = \lambda^2 \Gamma_1(g_1)$ ,  $\Gamma_2(\lambda g_2) = \lambda^2 \Gamma_2(g_2)$  and by Lemma 2.1

$$E_{\theta, \gamma, x}(\exp\{\lambda \Gamma_1(g_1)\}) < \infty, \quad E_{\theta, \gamma, x}(\exp\{\lambda \Gamma_2(g_2)\}) < \infty.$$

Choose a sequence of real  $C^\infty$ -functions  $\Phi_n, n \geq 1$  with compact support such that  $\lim_{n \rightarrow \infty} \sup_{|x| \leq M} |\Phi_n(x) - e^x| = 0$  for all  $M \in (0, \infty)$ . For any  $g_1 = \alpha f \in \mathcal{A}_1$  and  $g_2 = ah \in \mathcal{A}_2$ , set

$$F_n(w) = \Phi_n(g_1(w)/2), \quad H_n(w) = \Phi_n(g_2(w)/2).$$

Then for any  $g \in L^2([0, T], \mathbb{R})$ ,

$$\lim_{\|g\|_{L^2} \rightarrow 0} \frac{|F_n(w+g) - F_n(w) - \alpha \Phi'_n(g_1(w)/2) \langle w, g \rangle_{L^2}|}{\|g\|_{L^2}} = 0$$

and

$$\lim_{\|g\|_{L^2} \rightarrow 0} \frac{|H_n(w+g) - H_n(w) - \alpha \Phi'_n(g_2(w)/2) \frac{1}{T} \int_0^T w_t dt \int_0^T g_t dt|}{\|g\|_{L^2}} = 0.$$

Therefore,  $F_n, H_n \in C_b^1(W/L^2)$ ,  $\nabla F_n = \alpha \Phi'_n(g_1(w)/2) w$ , and

$$\nabla H_n = \frac{\alpha}{T} \int_0^T w_t dt \Phi'_n(g_2(w)/2)$$

and so by (2.9), we have

$$Ent_{P_{\theta, \gamma, x}}(F_n^2) \leq \frac{2}{\theta^2} E_{\theta, \gamma, x} \left( \int_0^T |\alpha w_t|^2 dt (\Phi'_n(g_1(w)/2))^2 \right)$$

and

$$Ent_{P_{\theta, \gamma, x}}(H_n^2) \leq \frac{2}{\theta^2} E_{\theta, \gamma, x} \left( \frac{1}{T} \left( \alpha \int_0^T w_t dt \right)^2 (\Phi'_n(g_2(w)/2))^2 \right).$$

Letting  $n \rightarrow \infty$  and by Lemma 2.1, we get

$$Ent_{P_{\theta, \gamma, x}}(e^{g_1}) \leq \frac{1}{2} E_{\theta, \gamma, x}(\Gamma_1(g_1)e^{g_1}), \quad Ent_{P_{\theta, \gamma, x}}(e^{g_2}) \leq \frac{1}{2} E_{\theta, \gamma, x}(\Gamma_2(g_2)e^{g_2}), \quad (2.10)$$

and so the conclusions of the lemma hold by Theorem 2.7 in [12] and  $T\hat{\mu}_T^2 \leq \int_0^T X_t^2 dt$ .  $\square$

### 2.3 Deviation inequalities

Since  $X_T \sim N(\mu_T, \sigma_T^2)$ , and under  $P_{\theta, \gamma, x}$

$$\hat{\mu}_T \sim N\left(\frac{1}{\theta T}(x - \frac{\gamma}{\theta})(1 - e^{-\theta T}) + \frac{\gamma}{\theta}, \frac{1}{\theta^2 T^2} \left(T - \frac{1}{2\theta}(e^{-2\theta T} - 1) + \frac{2}{\theta}(e^{-\theta T} - 1)\right)\right),$$

it is easily to get from Chebyshev inequality, for any  $r > 0$ ,

$$P_{\theta, \gamma, x}(|X_T - E_{\theta, \gamma, x}(X_T)| \geq r) \leq 2 \exp\{-\theta r^2\}, \quad (2.11)$$

$$P_{\theta, \gamma, x}(|\hat{\mu}_T - E_{\theta, \gamma, x}(\hat{\mu}_T)| \geq r) \leq 2 \exp\left\{-\frac{\theta^3 T r^2}{2\theta + 1}\right\} \quad (2.12)$$

where we used (2.7).

**Lemma 2.3.** *There exist finite positive constants  $C_0, C_1, C_2$  such that for all  $r > 0$  and all  $T \geq 1$ ,*

$$P_{\theta, \gamma, x} \left( \left| \int_0^T X_t^2 dt - E_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt \right) \right| \geq rT \right) \leq C_0 \exp \{-C_1 rT \min \{1, C_2 r\}\}$$

and

$$P_{\theta, \gamma, x} \left( \left| \hat{\mu}_T^2 - E_{\theta, \gamma, x}(\hat{\mu}_T^2) \right| \geq r \right) \leq C_0 \exp \{-C_1 rT \min \{1, C_2 r\}\}.$$

In particular, there exist finite positive constants  $C_0, C_1, C_2$  such that for all  $r > 0$  and all  $T \geq 1$ ,

$$P_{\theta, \gamma, x} \left( \left| \hat{\sigma}_T^2 - E_{\theta, \gamma, x}(\hat{\sigma}_T^2) \right| \geq r \right) \leq C_0 \exp \{-C_1 rT \min \{1, C_2 r\}\}.$$

*Proof.* We only prove the first inequality. By Lemma 2.2 and Lemma 2.1, there exist finite positive constants  $L_1$  and  $L_2$  such that for all  $T \geq 1$ , for any  $|\alpha| \leq \theta^2/4$ ,

$$E_{\theta, \gamma, x} \left( \exp \left\{ \alpha \left( \int_0^T X_t^2 dt - E_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt \right) \right) \right\} \right) \leq L_1 e^{L_2 \alpha^2 T}.$$

Therefore, by Chebyshev inequality, for any  $r > 0$ ,  $T \geq 1$  and  $|\alpha| \leq \theta^2/4$ ,

$$P_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt - E_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt \right) \geq rT \right) \leq L_1 e^{-(\alpha r - L_2 \alpha^2)T}$$

and

$$P_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt - E_{\theta, \gamma, x} \left( \int_0^T X_t^2 dt \right) \leq -rT \right) \leq L_1 e^{-(\alpha r - L_2 \alpha^2)T}.$$

Now, by

$$\sup_{|\alpha| \leq \theta^2/4} \{\alpha r - L_2 \alpha^2\} \geq \frac{\theta^2 r}{8} \min \left\{ 1, \frac{2r}{L_2 \theta^2} \right\},$$

we obtain the first inequality of the lemma from the above estimates.  $\square$

**Lemma 2.4.** *There exist finite positive constants  $C_0, C_1$  and  $C_2$  such that for all  $r > 0$  and all  $T \geq 1$ ,*

$$P_{\theta, \gamma, x} \left( \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \right| \geq rT \right) \leq C_0 \exp \{-C_1 rT \min \{1, C_2 r\}\}.$$

*Proof.* Since for any  $r > 0$  and  $T \geq 1$ ,

$$\begin{aligned} & \left\{ \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \right| \geq rT \right\} \\ & \subset \left\{ \left| W_T (\hat{\mu}_T - E_{\theta, \gamma, x}(\hat{\mu}_T)) \right| \geq rT/2 \right\} \cup \left\{ \left| W_T \left( E_{\theta, \gamma, x}(\hat{\mu}_T) - \frac{\gamma}{\theta} \right) \right| \geq rT/2 \right\} \\ & \subset \left\{ |W_T| \geq \sqrt{r}T/2 \right\} \cup \left\{ \left| (\hat{\mu}_T - E_{\theta, \gamma, x}(\hat{\mu}_T)) \right| \geq \sqrt{r} \right\} \cup \left\{ \left| W_T \right| \geq \frac{\theta r T}{2 \left| x - \frac{\gamma}{\theta} \right|} \right\}, \end{aligned}$$



by (2.12) and  $W_T \sim N(0, T)$ , we get

$$\begin{aligned} & P_{\theta, \gamma, x} \left( \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \right| \geq rT \right) \\ & \leq 2 \exp \left\{ -\frac{Tr}{8} \right\} + 2 \exp \left\{ -\frac{\theta^3 Tr}{2\theta + 1} \right\} + 2 \exp \left\{ -\frac{\theta^2 r^2 T}{8 \left( x - \frac{\gamma}{\theta} \right)^2} \right\}. \end{aligned}$$

□

**Lemma 2.5.** For each  $\beta \in \mathbb{R}$  fixed, there exist finite positive constants  $C_0, C_1, C_2$  such that for all  $r > 0$  and all  $T \geq 1$ ,

$$P_{\theta, \gamma, x} \left( \left| \int_0^T (X_t - \beta) dW_t \right| \geq rT \right) \leq C_0 \exp \{ -C_1 rT \min \{ 1, C_2 r \} \}.$$

*Proof.* It is known that for  $\alpha \in \mathbb{R}$ ,

$$M_T^{(\alpha)} = \exp \left\{ \alpha \int_0^T (X_t - \beta) dW_t - \frac{\alpha^2}{2} \int_0^T (X_t - \beta)^2 dt \right\}, \quad T \geq 0$$

is  $\mathcal{F}_T$ -martingale, where  $\mathcal{F}_T := \sigma(W_t, t \leq T)$ . Therefore, by Hölder inequality, we can get that for any  $\epsilon \in (0, 1]$ ,

$$\begin{aligned} & E_{\theta, \gamma, x} \left( \exp \left\{ \alpha \int_0^T (X_t - \beta) dW_t \right\} \right) \\ & \leq \left( E_{\theta, \gamma, x} \left( \exp \left\{ \frac{(1+\epsilon)^2 \alpha^2}{2\epsilon} \int_0^T (X_t - \beta)^2 dt \right\} \right) \right)^{\frac{\epsilon}{1+\epsilon}} \left( E_{\theta, \gamma, x} \left( M_T^{((1+\epsilon)\alpha)} \right) \right)^{\frac{1}{1+\epsilon}} \\ & = \left( E_{\theta, \gamma, x} \left( \exp \left\{ \frac{(1+\epsilon)^2 \alpha^2}{2\epsilon} \int_0^T (X_t - \beta)^2 dt \right\} \right) \right)^{\frac{\epsilon}{1+\epsilon}}. \end{aligned}$$

In particular, take  $\epsilon = 1$ , then by Lemma 2.1, there exists finite positive constants  $L_1 = L_1(\theta, \beta, \gamma, x)$  and  $L_2 = L_2(\theta, \beta, \gamma, x)$  such that for all  $T \geq 1$ , for any  $\alpha^2 \leq \theta^2/16$ , by Cauchy-Schwartz inequality,

$$\begin{aligned} & E_{\theta, \gamma, x} \left( \exp \left\{ \alpha \int_0^T (X_t - \beta) dW_t \right\} \right) \\ & \leq \left( E_{\theta, \gamma, x} \left( \exp \left\{ 2\alpha^2 \int_0^T (X_t - \beta)^2 dt \right\} \right) \right)^{\frac{1}{2}} \\ & \leq \left( E_{\theta, \gamma, x} \left( \exp \left\{ 4\alpha^2 \int_0^T X_t^2 dt \right\} \right) \right)^{\frac{1}{4}} \left( E_{\theta, \gamma, x} \left( \exp \left\{ 4\alpha^2 \int_0^T (-2\beta X_t + \beta^2) dt \right\} \right) \right)^{\frac{1}{4}} \\ & \leq L_1 e^{L_2 \alpha^2 T}. \end{aligned}$$

Therefore, by Chebyshev inequality, the conclusion of the lemma holds.

□

### Proof of Theorem 1.1

We only show the first inequality. The second one is similar. By

$$\hat{\theta}_T - \theta = \frac{W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) - \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t}{T \hat{\sigma}_T^2}$$

for any  $r > 0$  and  $T \geq 1$ ,

$$\begin{aligned} & P_{\theta, \gamma, x} \left( |\hat{\theta}_T - \theta| \geq r \right) \\ & \leq P_{\theta, \gamma, x} \left( \left| \hat{\sigma}_T^2 - E_{\theta, \gamma, x}(\hat{\sigma}_T^2) \right| \geq E_{\theta, \gamma, x}(\hat{\sigma}_T^2)/2 \right) \\ & \quad + P_{\theta, \gamma, x} \left( \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) - \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq E_{\theta, \gamma, x}(\hat{\sigma}_T^2) r T / 2 \right) \end{aligned}$$

Therefore, by Lemmas 2.3, 2.4 and 2.5, we obtain the first inequality of the theorem.  $\square$

### 3 Moderate deviations

In this section, we show Theorem 1.2. By (1.2) and (1.3), we have the following estimates

$$\begin{aligned} & \left| (\hat{\theta}_T - \theta) + \frac{2\theta}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \\ & \leq \frac{|W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right)|}{T \hat{\sigma}_T^2} + \frac{|2\theta \hat{\sigma}_T^2 - 1| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right|}{T \hat{\sigma}_T^2} \end{aligned} \quad (3.1)$$

and for

$$\begin{aligned} & \left| (\hat{\gamma}_T - \gamma) - \frac{W_T}{T} + \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \\ & \leq \frac{|\hat{\mu}_T| |W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right)|}{T \hat{\sigma}_T^2} + \frac{|2\gamma \hat{\sigma}_T^2 - \hat{\mu}_T| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right|}{T \hat{\sigma}_T^2}. \end{aligned} \quad (3.2)$$

**Lemma 3.1.** (1). For any  $r > 0$ ,

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| \hat{\mu}_T - \frac{\gamma}{\theta} \right| |W_T| \geq \sqrt{T \lambda_T} r \right) = -\infty,$$

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| \hat{\mu}_T - \frac{\gamma}{\theta} \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \sqrt{T \lambda_T} r \right) = -\infty$$

and

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| \hat{\sigma}_T^2 - \frac{1}{2\theta} \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \sqrt{T \lambda_T} r \right) = -\infty.$$

(2). For any  $\delta > 0$ ,

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| (\hat{\theta}_T - \theta) - \frac{2\theta}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \sqrt{\frac{\lambda_T}{T}} \right) = -\infty$$

and

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| (\hat{\gamma}_T - \gamma) - \frac{W_T}{T} - \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \sqrt{\frac{\lambda_T}{T}} \right) = -\infty.$$

*Proof.* (1). We only give the proof of the third assertion in (1). The rest is similar. For any  $L > 0$ ,

$$\begin{aligned} & \left\{ \left| \hat{\sigma}_T^2 - \frac{1}{2\theta} \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \sqrt{T\lambda_T r} \right\} \\ & \subset \left\{ \left| \hat{\sigma}_T^2 - \frac{1}{2\theta} \right| \geq \frac{r}{L} \right\} \cup \left\{ \frac{1}{\sqrt{T\lambda_T}} \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq L \right\}. \end{aligned}$$

By Lemma 2.3, and Lemma 2.5, we have

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| \hat{\sigma}_T^2 - \frac{1}{2\theta} \right| \geq \frac{r}{L} \right) = -\infty$$

and

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \frac{1}{\sqrt{T\lambda_T}} \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq L \right) \leq -L^2 C_1 C_2.$$

Hence,

$$\limsup_{T \rightarrow \infty} \frac{1}{\lambda_T} \log P_{\theta, \gamma, x} \left( \left| \hat{\sigma}_T^2 - \frac{1}{2\theta} \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \sqrt{T\lambda_T r} \right) \leq -L^2 C_1 C_2.$$

Letting  $L \rightarrow \infty$ , we obtain the third conclusion.

(2). It follows from (3.1) and (3.2) that

$$\begin{aligned} & \left( \left| (\hat{\theta}_T - \theta) - \frac{2\theta}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \sqrt{\frac{\lambda_T}{T}} \right) \\ & \subset \left\{ \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \right| \geq \delta \hat{\sigma}_T^2 \frac{\sqrt{T\lambda_T}}{2} \right\} \cup \left\{ \left| 2\theta \hat{\sigma}_T^2 - 1 \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \hat{\sigma}_T^2 \frac{\sqrt{T\lambda_T}}{2} \right\} \\ & \subset \left\{ \left| W_T \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \right| \geq \delta E_{\theta, \gamma, x}(\hat{\sigma}_T^2) \frac{\sqrt{T\lambda_T}}{4} \right\} \cup \left\{ \left| \hat{\sigma}_T^2 - E_{\theta, \gamma, x}(\hat{\sigma}_T^2) \right| \geq E_{\theta, \gamma, x}(\hat{\sigma}_T^2)/2 \right\} \\ & \cup \left\{ \left| 2\theta \hat{\sigma}_T^2 - 1 \right| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta E_{\theta, \gamma, x}(\hat{\sigma}_T^2) \frac{\sqrt{T\lambda_T}}{4} \right\} \end{aligned}$$

and

$$\begin{aligned}
& \left( \left| (\hat{\gamma}_T - \gamma) - \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \sqrt{\frac{\lambda_T}{T}} \right) \\
& \subset \left\{ |\hat{\mu}_T| |W_T| \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \geq \delta \hat{\sigma}_T^2 \frac{\sqrt{T\lambda_T}}{2} \right\} \cup \left\{ |2\gamma \hat{\sigma}_T^2 - \hat{\mu}_T| \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta \hat{\sigma}_T^2 \frac{\sqrt{T\lambda_T}}{2} \right\} \\
& \subset \left\{ \left| \hat{\mu}_T - \frac{\gamma}{\theta} \right| \geq \frac{\gamma}{2\theta} \right\} \cup \left\{ |\hat{\sigma}_T^2 - E_{\theta,\gamma,x}(\hat{\sigma}_T^2)| \geq E_{\theta,\gamma,x}(\hat{\sigma}_T^2)/2 \right\} \\
& \cup \left\{ \frac{3\gamma}{2\theta} |W_T| \left( \hat{\mu}_T - \frac{\gamma}{\theta} \right) \geq \delta E_{\theta,\gamma,x}(\hat{\sigma}_T^2) \frac{\sqrt{T\lambda_T}}{4} \right\} \\
& \cup \left\{ \left( \left| 2\gamma \hat{\sigma}_T^2 - \frac{\gamma}{\theta} \right| + \left| \hat{\mu}_T - \frac{\gamma}{\theta} \right| \right) \left| \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right| \geq \delta E_{\theta,\gamma,x}(\hat{\sigma}_T^2) \frac{\sqrt{T\lambda_T}}{4} \right\}.
\end{aligned}$$

Therefore, by Lemmas 2.3 and (1), we get the conclusions.  $\square$

**Lemma 3.2.** For each  $\beta, \kappa \in \mathbb{R}$  fixed,  $\left\{ P_{\theta,\gamma,x} \left( \frac{\kappa}{\sqrt{T\lambda_T}} \int_0^T (X_t - \beta) dW_t \in \cdot \right), T \geq 1 \right\}$  satisfies the LDP with speed  $\lambda_T$  and rate function  $J(u) = \frac{\theta^2 u^2}{\kappa^2(\theta + 2(\gamma - \theta\beta)^2)}$ .

*Proof.* By (2.12) and Lemma 2.3, we can get for any  $\delta > 0$ ,

$$\lim_{T \rightarrow \infty} \frac{1}{T} \log P_{\theta,\gamma,x} \left( \left| \frac{1}{T} \int_0^T (X_t - \beta)^2 dt - \left( \frac{1}{2\theta} + \frac{1}{\theta^2} (\gamma - \theta\beta)^2 \right) \right| \geq \delta \right) < 0. \quad (3.3)$$

Therefore, Proposition 1 in [4] yields the conclusion of the lemma.  $\square$

### Proof of Theorem 1.2

By Lemma 3.1,  $\{P_{\theta,\gamma,x}(\sqrt{\frac{T}{\lambda_T}}(\hat{\theta}_T - \theta) \in \cdot), T \geq 1\}$  and  $\{P_{\theta,\gamma,x}(\sqrt{\frac{T}{\lambda_T}}(\hat{\gamma}_T - \gamma) \in \cdot), T \geq 1\}$  are exponential equivalent to

$$\left\{ P_{\theta,\gamma,x} \left( \sqrt{\frac{T}{\lambda_T}} \frac{2\theta}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \in \cdot \right), T \geq 1 \right\}$$

and

$$\left\{ P_{\theta,\gamma,x} \left( \sqrt{\frac{T}{\lambda_T}} \left( \frac{W_T}{T} + \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t \right) \in \cdot \right), T \geq 1 \right\},$$

respectively. Noting for  $\gamma \neq 0$ ,  $\frac{W_T}{T} + \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} \right) dW_t = \frac{2\gamma}{T} \int_0^T \left( X_t - \frac{\gamma}{\theta} + \frac{1}{2\gamma} \right) dW_t$ , Theorem 1.2 follows from Lemma 3.2.

□

**Acknowledgments** The authors are grateful to referees for their comments and suggestions.

## References

- [1] B. Bercu, A. Rouault. Sharp large deviations for the Ornstein-Uhlenbeck process. *Theory of Prob. and its Appl.*, 46(2002), 1-19. MR1968706
- [2] S. G. Bobkov, F. Götze. Exponential Integrability and Transportation Cost Related to Logarithmic Sobolev Inequalities. *Journal of Functional Analysis*, 163(1999), 1-28. MR1682772
- [3] P. Cattiaux, A. Guillin. Deviation bounds for additive functionals of Markov process. *ESAIM: Probability and Statistics*. 12(2008), 12-29. MR2367991
- [4] A. Dembo. Moderate Deviations for Martingales with Bounded Jumps. *Electronic Communications in Probability*, 1 (1996),11-17. MR1386290
- [5] A. Dembo, D. Zeitouni. *Large Deviations Techniques and Applications*, Springer-Verlag, 1998. MR1619036
- [6] J. D. Deuschel, D. W. Stroock. *Large Deviations*, New York, 1989. MR0997938
- [7] H. Djellout, A. Guillin, L. M. Wu. Transportation cost-information inequalities and applications to random dynamical system and diffusions. *Ann. Probab.*, 32(2004), 2702-2732. MR2078555
- [8] H. Djellout, A. Guillin, L. M. Wu. Moderate deviations for non-linear functionals and empirical spectral density of moving average processes. *Ann. Inst. H. Poincaré Probab. Statist.*, 42(2006), 393-416. MR2242954
- [9] D. Florens-Landais, H. Pham, Large deviations in estimate of an Ornstein-Uhlenbeck model. *Journal of Applied Probability*, 36(1999), 60-77. MR1699608
- [10] M. Gourcy, L. M. Wu, Logarithmic Sobolev inequalities of diffusions for the  $L^2$  metric. *Potential Analysis*, 25(2006), 77-102. MR2238937
- [11] A. Guillin, R. Liptser, Examples of moderate deviation principles for diffusion processes. *Discrete and continuous dynamical systems-series B*, 6(2006), 803-828. MR2223909
- [12] M. Ledoux, Concentration of Measure and Logarithmic Sobolev Inequalities. Séminaire de probabilités XXXIII, Lecture Notes in Mathematics, 1709(1999), 120-216. MR1767995
- [13] M. Ledoux, *The Concentration of Measure Phenomenon*. Mathematical Surveys and Monographs 89, American Mathematical Society, 2001. MR1849347
- [14] P. Lezaud, Chernoff and Berry-Essen's inequalities for Markov processes. *ESAIMS: Probability and Statistics*, 5(2001), 183-201. MR1875670
- [15] Yury A. Kutoyants, *Statistical Inference for Ergodic Diffusion Processes*. Springer Series in Statistics, London, 2004, MR2144185

- 
- [16] B. L. S. Prakasa Rao, *Statistical Inference for Diffusion Type Processes*. Oxford University Press, New York, 1999.
- [17] D. Revuz, M. Yor, *Continuous Martingales and Brownian Motion*. Springer-Verlag, 1991. MR1083357
- [18] L. M. Wu, A deviation inequality for non-reversible Markov processes. *Ann. Inst. Henri Poincaré, Probabilités et Statistiques*. 36(2000), 435-445. MR1785390