

# Parameter Estimation for Controlled Semilinear Stochastic Systems: Identifiability and Consistency<sup>1</sup>

B. Goldys

*The University of New South Wales, Sydney, Australia*

E-mail: B.Goldys@unsw.edu.au

and

B. Maslowski

*Czech Academy of Sciences, Praha, Czech Republic*

E-mail: maslow@math.cas.cz

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We consider a controlled stochastic semilinear evolution equation with the drift depending on the unknown parameter. We show that the maximum likelihood estimator is strongly consistent for a class of bounded predictable controls. © 2001

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## 1. INTRODUCTION

The aim of this paper is to study the maximum likelihood estimator of the parameter  $\alpha_0 \in \mathbb{R}^d$  in a controlled stochastic semilinear equation

$$\begin{cases} dX_t = (AX_t + f(\alpha_0, X_t) - u_t) dt + \sqrt{Q} dW_t, & t \geq 0, \\ X_0 = x, \end{cases} \quad (1.1)$$

considered in a separable Hilbert space  $H$ . The process  $W$  is a cylindrical Wiener process on  $H$  and  $(u_t)$  is a uniformly bounded progressively measurable process. We assume that  $f: H \rightarrow H$  is Lipschitz and Gateaux

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differentiable and  $A$  generates a strongly continuous semigroup on  $H$ . It is well known (see monograph [14] for a thorough discussion of this subject) that a large variety of stochastic partial differential equations may be studied as special cases of (1.1), see also an example at the end of this paper.

Estimation of the drift parameter in a stochastic differential equation of type (1.1) with  $u=0$  has been studied for a long time in the case when  $H = \mathbb{R}^d$ ; see [28] and references therein for an extensive study of this problem. The first proofs of consistency of the Maximum Likelihood Estimators of  $\alpha_0$  in finite dimensional case have been obtained in the seminal papers [29, 30]. The estimation problem was also studied in the context of adaptive control of diffusion processes in [7–9].

Most of the results on parameter estimation for stochastic evolution equations in infinite dimensional spaces we are aware of concern the linear systems. For example, the least squares method has been used in [15, 16] to define a strongly consistent family of estimators for controlled linear systems of stochastic PDE's with distributed and boundary controls, respectively. The maximum likelihood method has been employed in [2–4] to estimate unknown parameters (which can be vectors in  $\mathbb{R}^d$  or functions) in stochastic parabolic PDEs; see also the recent paper [6], where the unknown parameter is a function (in general discontinuous) involved in the second order term of the parabolic operator. The hyperbolic case was treated in [5]. Consistency of the maximum likelihood Estimators (MLE) in a different sense from ours was recently studied in interesting papers [22, 23] for certain linear stochastic parabolic PDEs. A study of the Bayes estimators for the parameters of linear stochastic partial differential equations has been recently initiated in [31].

It seems that few results are available in the case of nonlinear stochastic infinite-dimensional systems. For an uncontrolled potential type parameter-dependent system of stochastic reaction-diffusion equations a minimum contrast estimator is proposed and its strong consistency is proved in [26]; the method is based on an earlier result for linear systems obtained in [24].

The aim of this paper is to show that if  $\alpha_0$  is identifiable and additionally  $f$  is Hölder continuous in  $\alpha$  then the maximum likelihood estimator  $\alpha_t$  of  $\alpha_0$  is strongly consistent, that is,

$$\mathbb{P}(\alpha_t \rightarrow \alpha_0) = 1,$$

under some reasonable assumptions on the control  $(u_t)$ . Our work is motivated by the recent works on the ergodic adaptive control of stochastic partial differential equations, see [17]; see Remark 4.4 for more details.

## 2. SETTING OF THE PROBLEM AND SOME AUXILIARY FACTS

We consider an equation

$$\begin{cases} dX_t = (AX_t + f(\alpha_0, X_t) - u_t) dt + \sqrt{Q} dW_t, & t \geq 0 \\ X_0 = x, \end{cases} \quad (2.1)$$

on a separable Hilbert space  $H$  with the norm  $|\cdot|$  and the inner product  $\langle \cdot, \cdot \rangle$ . The following is a standing hypothesis for the rest of the paper.

**HYPOTHESIS 2.1.** (i)  $A$  generates a  $C_0$ -semigroup  $(S_t)$  on  $H$ . Moreover, there exists  $\omega \in \mathbb{R}$  such that

$$\|S_t\| \leq e^{\omega t}, \quad t \geq 0. \quad (2.2)$$

(ii)  $W$  is a standard cylindrical Wiener process on  $H$  defined on a stochastic basis  $(\Omega, \mathcal{F}, (\mathcal{F}_t), \mathbb{P})$  and  $Q = Q^*$  is a bounded operator on  $H$  with bounded inverse.

(iii)  $\mathcal{A}$  is a compact subset of  $\mathbb{R}^d$  and  $f: \mathcal{A} \times H \rightarrow H$  is such that  $f(\alpha, \cdot)$  is Gateaux differentiable for each  $\alpha \in \mathcal{A}$ . Moreover, there exists  $K > 0$  such that

$$|f(\alpha, x)| \leq K(1 + |x|), \quad x \in H, \quad (2.3)$$

and

$$\langle Ax + f(\alpha, x + y), x \rangle \leq -K|x|^2 + a(|y|)|x|, \quad x \in \mathcal{D}(A), y \in H, \quad (2.4)$$

where  $a: \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is a continuous increasing function.

(iv) There exists  $T > 0$  such that

$$\int_0^T \|S_t\|_{HS}^2 dt < \infty,$$

where  $\|\cdot\|_{HS}$  denotes the Hilbert–Schmidt norm of an operator.

(v) Controls  $u$  are taken from the set of admissible controls  $\mathcal{U}$  which consists of progressively measurable processes  $u: \mathbb{R}_+ \times \Omega \rightarrow H$  such that  $\mathbb{P}(u_t \in B(r_0)) = 1$  for all  $t \geq 0$ , where  $B(r_0) \subset H$  stands for a centered ball with the radius  $r_0$ .

*Remark 2.2.* (a) Condition (iv) which may seem unnecessarily restrictive is a direct consequence of Condition (ii). If the noise is cylindrical then (iv) is necessary for the existence of an  $H$ -valued solution to (1.1).

(b) Assumption (v) which allows bounded controls only, excludes unstable systems stabilised by the adaptive control. This condition is technical and it would be interesting to remove it. This will be a subject of a future paper.

Let  $(Z_t)$  be a solution to the linear equation

$$\begin{cases} dZ_t = AZ_t dt + \sqrt{Q} dW_t, \\ Z_0 = x, t \geq 0. \end{cases} \tag{2.5}$$

In view of condition (iii) we can assume in (2.2) without loss of generality that  $\omega < 0$ . Taking (iv) into account we find in this case that the process  $(Z_t)$  has a unique invariant measure  $\mu = N(0, Q_\infty)$ , where the operator

$$Q_\infty = \int_0^\infty S_t Q S_t^* dt$$

is of trace class. Therefore we shall always assume that the following hypothesis holds.

**HYPOTHESIS 2.3.** (i) *There exists  $\omega > 0$  such that*

$$\langle Ax, x \rangle \leq -\omega |x|^2, \quad x \in \mathcal{D}(A). \tag{2.6}$$

(iii) *The function  $a$  introduced in Hypothesis 2.1 enjoys the property*

$$\int_H a \left( |y| + \frac{r_0}{\omega} \right) \mu(dy) < \infty. \tag{2.7}$$

The parameter  $\alpha \in \mathcal{A}$  remains fixed throughout the rest of this section and for simplicity of notation it is omitted in all formulae.

If Hypothesis 2.1 holds then the equation

$$\begin{cases} dY_t^x = (AY_t^x + f(Y_t^x)) dt + \sqrt{Q} dW_t, & t \geq 0, \\ Y_0^x = x, \end{cases} \tag{2.8}$$

has a unique mild solution, see Theorem 7.4 in [14]. Let  $P_t^f \phi(x) = \mathbb{E} \phi(Y_t^x)$  be a transition semigroup of the Markov process  $(Y_t^x)$ . By Hypothesis 2.1 the semigroup  $(P_t^f)$  has a unique invariant measure  $\mu^f$  [19] and therefore, the semigroup  $(P_t^f)$  extends to a  $C_0$ -semigroup of contractions on  $L^2(H, \mu^f)$  with the generator  $L^f$ . The domain of  $L^f$  in  $L^2(H, \mu^f)$  will be denoted by  $\mathcal{D}(L^f)$ .

Let  $\mathcal{K} = \{k_n : n \geq 1\} \subset \mathcal{D}(A^*)$  be a countable set such that  $\mathcal{K}_l = \text{lin}(\mathcal{K})$  is dense in  $\text{dom}(A^*)$  when endowed with the graph norm. Let  $\mathcal{G} = \{g_n : n \geq 1\} \subset \bigcup_{m=1}^\infty C_0^2(\mathbb{R}^m)$  has the property, that  $\mathcal{G}_l = \text{lin}\{g_n : n \geq 1\}$  is

dense in each  $C_0^2(\mathbb{R}^m)$ ,  $m \geq 1$ , in the topology of uniform convergence on compacts,

$$\mathcal{D}_0 = \{g \circ P : g \in \mathcal{G}_1, P(H) \subset \mathcal{K}_1, \dim P(H) < \infty\},$$

where  $P$  stands for an orthogonal projection on  $H$  and  $C_0^2(\mathbb{R}^n)$  is the space of twice differentiable functions on  $\mathbb{R}^m$  with compact supports. Let  $v: H \rightarrow \mathbb{R}$  be a bounded Borel function. If Hypothesis 2.1 holds then by the Girsanov Theorem the equation

$$\begin{cases} dY_t^{v,x} = (AY_t^{v,x} + f(Y_t^{v,x}) + v(Y_t^{v,x})) dt + \sqrt{Q} dW_t, \\ Y_0^{v,x} = x \in H, \end{cases}$$

has a unique weak solution. For every  $\phi \in \mathcal{D}_0$  a differential operator

$$L^{f+v}\phi(x) = \frac{1}{2} \text{tr}(QD^2\phi(x)) + \langle x, A^*D\phi(x) \rangle + \langle f(x) + v(x), D\phi(x) \rangle$$

is well defined and  $L^{f+v}\phi \in L^2(H, \mu^f)$ . The next proposition is a version of the results proved in [12, 13]. For the reader's convenience we sketch its proof.

**PROPOSITION 2.4.** *Assume Hypotheses 2.1 and 2.3 and let  $v: H \rightarrow \mathbb{R}$  be a bounded Borel function. Then the operator  $(L^{f+v}, \mathcal{D}_0)$  in  $L^2(H, \mu^f)$  has a unique extension to a generator of a  $C_0$ -semigroup  $(P_t^{f+v})$ , still denoted by  $L^{f+v}$  and  $\mathcal{D}(L^{f+v}) = \mathcal{D}(L^f)$ . Moreover, the semigroup  $(P_t^{f+v})$  may be identified with the transition semigroup of the process  $Y^v$ .*

*Proof.* *Step 1.* In this part of the proof we construct a core for the generator  $L^f$  in  $L^2(H, \mu^f)$ . Let us recall first some properties of the semigroup  $(P_t^f)$  acting the spaces of continuous functions. It is well known, see [14], that the semigroup  $(P_t^f)$  is Feller, that is,  $P_t^f(C_b(H)) \subset C_b(H)$ , where  $C_b(H)$  stands for the Banach space of bounded continuous functions on  $H$  with the norm

$$\|\phi\|_\infty = \sup_{x \in H} |\phi(x)|.$$

Following [10] we say that a sequence  $(\phi_n) \subset C_b(H)$  is  $\mathcal{K}$ -convergent to  $\phi \in C_b(H)$ ,  $\phi_n \xrightarrow{\mathcal{K}} \phi$ , if

$$\sup_{n \geq 1} \|\phi_n\|_\infty < \infty, \tag{2.9}$$

and

$$\phi_n \rightarrow \phi, \quad \text{uniformly on compacts.} \tag{2.10}$$

The domain  $\mathcal{D}_0(L^f)$  of the operator  $L^f$  in  $C_b(H)$  is defined as the set of all functions  $\phi \in C_b(H)$  such that

$$\frac{P_t^f \phi - \phi}{t} \xrightarrow[t \rightarrow 0]{\mathcal{X}} \psi \quad (\text{say}). \tag{2.11}$$

It is easy to see that  $\mathcal{D}_0(L^f) \subset \mathcal{D}(L^f)$ . Let

$$\mathcal{D}_1 = \left\{ g \circ P : g \in \bigcup_{m=1}^{\infty} C_b^2(\mathbb{R}^m), P(H) \subset \mathcal{D}(A^*), \dim P(H) < \infty \right\},$$

where  $C_b^2(\mathbb{R}^m)$  is a space of twice differentiable bounded functions on  $\mathbb{R}^m$ . Let  $C_1(H)$  denote the space of continuous functions  $\phi: H \rightarrow \mathbb{R}$  such that

$$\|\phi\|_1 = \sup_{x \in H} \frac{|\phi(x)|}{1 + |x|} < \infty.$$

The space  $C_1(H)$  with the norm  $\|\cdot\|_1$  is a Banach space and  $(P_t^f)$  is a semi-group of bounded operators on  $C_1(H)$ . If for a certain  $\phi \in C_1(H)$

$$\frac{P_t^f \phi - \phi}{t} \xrightarrow[t \rightarrow 0]{\mathcal{X}} \psi \in C_1(H)$$

then we say that  $\phi \in \mathcal{D}_1(L^f)$ , the domain of the generator  $L^f$  of the semi-group  $(P_t^f)$  in  $C_1(H)$ , and we put  $\psi = L^f \phi$ . It has been proved in [20] that for every  $\phi \in \mathcal{D}_1(L^f)$  there exists a sequence  $(\phi_n) \subset \mathcal{D}_1$ , such that

$$\phi_n \xrightarrow{\mathcal{X}} \phi \quad \text{and} \quad L^f \phi_n \xrightarrow{\mathcal{X}} L^f \phi. \tag{2.12}$$

It is easy to see that for every  $\phi \in \mathcal{D}_1$  we can find  $(\phi_n) \subset \mathcal{D}_0$  such that the above convergence holds. Therefore, for every  $\phi \in \mathcal{D}_1(L^f)$  there exists a sequence  $(\phi_n) \subset \mathcal{D}_0$  such that (2.12) holds. Finally, it follows that for  $\phi \in \mathcal{D}_0(L^f)$ , the domain of  $L^f$  in  $C_b(H)$  (2.12) holds as well.

*Step 2.* In this part of the proof we use the same notation  $\|\cdot\|$  for the norm in the space  $L^2(H, \mu^f)$  and in the space  $L^2(H, \mu^f; H)$  of  $H$ -valued functions. It follows from Step 1 that  $\mathcal{D}_0$  is a core for the generator  $L^f$  of the semigroup  $(P_t^f)$  considered in  $L^2(H, \mu^f)$ . Then proceeding in the same way as in [13] one can show that for every  $T > 0$  and  $\phi \in \mathcal{D}_0$

$$\int_0^T \|Q^{1/2} DP_t^f \phi\|^2 dt = \|\phi\|^2 - \|P_T^f \phi\|^2. \tag{2.13}$$

Let

$$B\phi(x) = \langle v(x), D\phi(x) \rangle, \quad \phi \in \mathcal{D}_0.$$

Since  $Q$  is boundedly invertible, we have

$$B\phi(x) = \langle Q^{-1/2}v(x), Q^{1/2}D\phi(x) \rangle, \quad \phi \in \mathcal{D}_0.$$

It follows from (2.13) and the Hölder inequality that for  $\phi \in \mathcal{D}_0$

$$\int_0^T \|BP_t^f \phi\| dt \leq c \sqrt{T} \left( \int_0^T \|Q^{1/2} DP_t^f \phi\|^2 dt \right)^{1/2} \leq c \sqrt{T} \|\phi\|,$$

where  $c = \|Q\|^{-1/2} \|v\|_\infty$ . For  $T$  sufficiently small we have  $c\sqrt{T} < 1$ , and therefore by the result in [32] the operator  $(L^f + B, \mathcal{D}_0)$  extends to a generator  $L^{f+v}$  of a  $C_0$ -semigroup  $(P_t^{f+v})$  on  $L^2(H, \mu^f)$  and

$$\mathcal{D}(L^f) = \mathcal{D}(L^{f+v}).$$

It follows that the graph norms of  $L^f$  and  $L^{f+v}$  are equivalent, hence  $\mathcal{D}_0$  is a core for  $L^{f+v}$  which implies that the extension of  $(L^f + B, \mathcal{D}_0)$  to a generator of  $C_0$  semigroup is unique. Finally, we note (but omit the details of the proof) that taking into account the Girsanov Theorem one can show that  $P_t^{f+v} \phi(x) = \mathbb{E}\phi(Y_t^{v,x})$  for every  $\phi \in L^2(H, \mu^f)$ . ■

**LEMMA 2.5.** *Assume Hypotheses 2.1 and 2.3. Then the semigroup  $(P_t^{f+v})$  has a unique invariant measure  $\mu^{f+v}$ . Moreover, if  $\nu$  is a probability measure on  $H$ , such that*

$$\int_H L^{f+v} \psi(x) \nu(dx) = 0, \quad \psi \in \mathcal{D}_0, \quad (2.14)$$

then  $\nu = \mu^{f+v}$ .

*Proof.* We need to show that  $\nu$  is an invariant measure for  $(P_t^{f+v})$ , that is,

$$\langle \nu, P_t^{f+v} \phi - \phi \rangle = 0 \quad (2.15)$$

for a measure determining set of bounded functions  $\phi$ . To this end it is enough to show that for every  $\beta > 0$

$$\int_0^\infty e^{-\beta t} \langle \nu, P_t^{f+v} \phi - \phi \rangle dt = 0. \quad (2.16)$$

Since for  $\phi \in \mathcal{D}(L^{f+v})$

$$P_t^{f+v}\phi - \phi = \int_0^t L^{f+v} P_s^{f+v}\phi \, ds,$$

we find that (2.16) holds for a certain  $\phi \in \mathcal{D}(L^{f+v})$  if and only if

$$\langle v, L^{f+v} J_\beta \phi \rangle = 0, \tag{2.17}$$

where  $J_\beta = (\beta - L^{f+v})^{-1}$ . Note that for every  $\psi \in \mathcal{D}_0$  the function

$$(\beta - L^{f+v}) \psi(x) = (\beta - L^f) \psi(x) - \langle v(x), D\psi(x) \rangle$$

is bounded and  $(\beta - L^{f+v}) \mathcal{D}_0$  is dense in  $L^2(H, \mu^f)$  by Proposition 2.4. Hence (2.17) follows which proves (2.15) for every  $\phi \in (\beta - L^{f+v}) \mathcal{D}_0$ . ■

### 3. IDENTIFIABILITY OF $\alpha$

**DEFINITION 3.1.** The parameter  $\alpha_0$  is said to be identifiable if for every  $\alpha \in \mathcal{A}$ ,  $\alpha \neq \alpha_0$  there exists  $x \in H$  such that

$$f(\alpha, x) \neq f(\alpha_0, x). \tag{3.1}$$

Note that (3.1) is, in some sense, a necessary and sufficient condition for the identifiability. Indeed, assume that  $f(\alpha_1, \cdot) = f(\alpha_0, \cdot)$  for some  $\alpha_1 \neq \alpha_0$ . Then  $\alpha_1$  and  $\alpha_0$  cannot be distinguished using the observations  $\{X_t : t > 0\}$ .

For a given control  $u \in \mathcal{U}$  we define a measure

$$\mu_t(\omega)(\Gamma) = \frac{1}{t} \int_0^t I_\Gamma(u_s(\omega)) \, ds, \quad \Gamma \in \mathcal{B}(B(r_0)),$$

where  $\mathcal{B}(B(r_0))$  is the  $\sigma$ -algebra of Borel subsets of  $B(r_0)$ . In the sequel we omit  $\omega$  and write simply  $\mu_t(\Gamma)$ .

For any  $t \geq 1$  and  $\zeta = (\zeta_1, \zeta_2) \in L^2_{loc}(\mathbb{R}_+, B(r_0)) \times L^2_{loc}(\mathbb{R}_+, H)$  we define a measure  $\nu_t^\zeta$  by the equation

$$\int_{B(r_0) \times H} \phi(y, x) \nu_t^\zeta(dy, dx) = \frac{1}{t} \int_0^t \phi(\zeta_1(s), \zeta_2(s)) \, ds, \quad \phi \in C_b(B(r_0) \times H). \tag{3.2}$$

The existence of  $\nu_t^\zeta$  follows from the fact that the functional defined by the right hand side of (3.2) is bounded and nonnegative on  $C_b(B(r_0) \times H)$  and tends to zero if  $\phi_n \downarrow 0$ . Therefore, the representing measure  $\nu_t^\zeta$  is non-negative,  $\sigma$ -additive and clearly  $\nu_t^\zeta(B(r_0) \times H) = 1$ .

LEMMA 3.2. *Assume Hypotheses 2.1 and 2.3. Moreover, assume that the set of measures  $\{\mu_t : t \geq 1\}$  is tight  $\mathbb{P}$ -as. Then the family of measures  $\{v_t^{(u, X)} : t \geq 1\}$  is  $\mathbb{P}$ -a.s. tight. In particular the lemma holds if  $u_s = h(\alpha_s, X_s)$ , where  $(\alpha_s)$  is an  $\mathcal{A}$ -valued predictable process and  $h: \mathcal{A} \times H \rightarrow B(r_0)$  is continuous.*

*Proof.* For every  $\zeta = (\zeta_1, \zeta_2) \in L_{loc}^2(\mathbb{R}_+, B(r_0)) \times L_{loc}^2(\mathbb{R}_+, H)$  and compact sets  $K_1 \subset B(r_0)$  and  $K_2 \subset H$  we have

$$\begin{aligned} & v_t^\zeta((B(r_0) \times H) \setminus (K_1 \times K_2)) \\ &= \frac{1}{t} \int_0^t I_{(B(r_0) \times H) \setminus (K_1 \times K_2)}(\zeta_1(s), \zeta_2(s)) ds \\ &\leq \frac{1}{t} \int_0^t I_{B(r_0) \setminus K_1}(\zeta_1(s)) ds + \frac{1}{t} \int_0^t I_{H \setminus K_2}(\zeta_2(s)) ds. \end{aligned} \quad (3.3)$$

Thus by the Prokhorov Theorem it suffices to show that there exists a sequence  $(K_n)$  of compact sets in  $H$  such that

$$\lim_{n \rightarrow \infty} \sup_{t \geq 1} \left( \frac{1}{t} \int_0^t I_{H \setminus K_n}(X_s) ds \right) = 0, \quad \mathbb{P}\text{-a.s.} \quad (3.4)$$

*Step 1.* We show first that

$$\limsup_{t \rightarrow \infty} \left( \frac{1}{t} \int_0^t |X_s|^2 ds \right) < \infty, \quad \mathbb{P}\text{-a.s.} \quad (3.5)$$

Setting

$$Z_t = \int_0^t S_{t-s} \sqrt{Q} dW_s$$

and

$$\tilde{Z}_t = - \int_0^t S_{t-s} u_s ds + Z_t, \quad Y_t = X_t - \tilde{Z}_t,$$

we obtain

$$Y_t = S_t x + \int_0^t S_{t-s} f(\alpha_0, Y_s + \tilde{Z}_s) ds, \quad t \geq 0. \quad (3.6)$$

Approximating  $Y$  by strong solutions to the equation

$$\frac{d}{dt} Y_t^\lambda = AY_t^\lambda + f_\lambda(Y_t^\lambda + \tilde{Z}_t),$$

where  $f_\lambda = \lambda(\lambda I - A)^{-1} f$  and using (2.4) we obtain

$$|Y_t| \leq e^{-\omega t} |x| + \int_0^t e^{-k(t-s)} a(|\tilde{Z}_s|) ds; \quad (3.7)$$

see [19] for a similar proof. Since the Ornstein–Uhlenbeck process  $Z^x$  is strongly Feller, it satisfies the Ergodic Theorem; see [26]. Furthermore, we have

$$\left| \int_0^t S_{t-s} u_s ds \right| \leq \int_0^\infty e^{-\omega(t-s)} r_0 ds = \frac{r_0}{\omega}, \quad (3.8)$$

hence by the Young Inequality we find that

$$\begin{aligned} \frac{1}{T} \int_0^T |X_t|^2 dt &\leq \frac{2}{T} \int_0^T |Y_t|^2 dt + \frac{2}{T} \int_0^T |\tilde{Z}_t|^2 dt \\ &\leq \frac{4}{T} |x|^2 \int_0^\infty e^{-2\omega t} dt + \frac{4}{T} \int_0^\infty e^{-2kt} dt \int_0^T a(|\tilde{Z}_t|) dt \\ &\quad + \frac{4}{T} \int_0^T |Z_t|^2 dt + \frac{4r_0^2}{\omega^2 T} \\ &\leq C_1 + \frac{C_2}{T} \int_0^T a\left(|Z_t| + \frac{r_0}{\omega}\right) dt + \frac{4}{T} \int_0^T |Z_t|^2 dt \end{aligned} \quad (3.9)$$

for some constants  $C_1, C_2$  independent of  $T$ . Thereby by (2.7)

$$\begin{aligned} \limsup_{T \rightarrow \infty} \left( \frac{1}{T} \int_0^T |X_t|^2 dt \right) \\ \leq C_1 + C_2 \int_H a\left(|x| + \frac{r_0}{\omega}\right) \mu(dx) + 4 \int_H |x|^2 \mu(dx) < \infty \end{aligned} \quad (3.10)$$

$\mathbb{P}$ -a.s. Note that from (3.10) it follows that

$$\begin{aligned} \limsup_{T \rightarrow \infty} \left( \frac{1}{T} \int_0^T \int_t^{t+1} |X_s|^2 ds dt \right) &= \limsup_{T \rightarrow \infty} \left( \frac{1}{T} \int_0^T \int_0^T I_{[t, t+1]}(s) |X_s|^2 dt ds \right) \\ &\leq \limsup_{T \rightarrow \infty} \left( \frac{1}{T} \int_0^T |X_s|^2 ds \right) < \infty, \quad \mathbb{P}\text{-a.s.} \end{aligned} \quad (3.11)$$

*Step 2.* We show that there exists a measurable set  $\hat{\Omega} \subset \Omega$  such that  $\mathbb{P}(\hat{\Omega}) = 1$  and for every  $\omega \in \hat{\Omega}$  there exists a sequence  $(C_n(\omega))$  of compact sets in  $H$  with the property

$$\lim_{n \rightarrow \infty} \sup_{t \geq 1} \left( \frac{1}{t} \int_0^t I_{H \setminus C_n(\omega)}(Z_s(\omega)) ds \right) = 0. \quad (3.12)$$

Let

$$\kappa_t(\Gamma) = \frac{1}{t} \int_0^t I_\Gamma(Z_s) ds, \quad \Gamma \in \mathcal{B}(H), t > 0. \quad (3.13)$$

The Ergodic Theorem for the Ornstein–Uhlenbeck process  $Z$  yields for  $t \rightarrow \infty$

$$\frac{1}{t} \int_0^t \phi(Z_s) ds = \int_H \phi(y) \kappa_t(dy) \rightarrow \int_H \phi(y) \mu(dy), \quad \mathbb{P}\text{-a.s.} \quad (3.14)$$

for each  $\phi \in BUC(H)$ , where  $BUC(H)$  stands for a subspace of  $C_b(H)$  consisting of functions which are uniformly continuous on  $H$  and endowed with the sup-norm. It is easy to see that there exists a countable subset  $\mathcal{M}$  of the unit ball in  $BUC(H)$  with the following property: for every  $\phi$  in the unit ball there exists a sequence  $(\phi_n) \subset \text{lin}(\mathcal{M})$  such that  $\phi_n \xrightarrow{\mathcal{X}} \phi$ . Therefore, we can find a measurable set  $\hat{\Omega} \subset \Omega$  such that  $\mathbb{P}(\hat{\Omega}) = 1$  and for the sample points from  $\hat{\Omega}$  and each  $\phi \in \text{lin}(\mathcal{M})$  (3.14) holds. It follows that  $\kappa_t \rightarrow \mu$  in the narrow topology. Therefore, by the Prokhorov Theorem for each sample point in  $\hat{\Omega}$  there exists a sequence  $(C_n)$  of compact sets in  $H$  satisfying (3.12).

*Step 3.* Fix a sample point from  $\hat{\Omega}$ . Then

$$X_{t+1} = S_1 x + J(f(\alpha_0, X_{\cdot+t}) + u_{\cdot+t}) + Z_{t+1} - S_1 Z_t, \quad t \geq 0, \quad (3.15)$$

where  $J: L^1((0, 1), H) \rightarrow H$  is given by the formula

$$Jh = \int_0^1 S_{1-s} Q^{1/2} h(s) ds.$$

Note that  $J$  is a compact operator and for  $s, t \geq 0$

$$|Q^{-1/2}(f(\alpha_0, X_{s+t}) + u_{s+t})| \leq K(1 + |X_{s+t}|) + r_0. \quad (3.16)$$

Set

$$A_n^t = \{s \in [0, t] : |X_s| > n\},$$

$$B_n^t = \{s \in [0, t] : Z_s \notin C_n\},$$

$$C_n^t = \{s \in [0, t] : \|Q^{-1/2}(f(\alpha_0, X_{\cdot+s}) + u_{\cdot+s})\|_{L^1((0,1); H)} > n\}.$$

By (3.12) we have for  $n \rightarrow \infty$

$$\begin{aligned} \sup_{t \geq 0} \left( \frac{1}{t} |B_n^t| \right) &= \sup_{t \geq 0} \left( \frac{1}{t} \int_{B_n^t} I_{H \setminus C_n}(Z_s) ds \right) \\ &= \sup_{t \geq 0} \left( \frac{1}{t} \int_0^t I_{H \setminus C_n}(Z_s) ds \right) \rightarrow 0, \end{aligned} \tag{3.17}$$

where  $|\cdot|$  above stands for the Lebesgue measure. It is also easy to see that (3.5) implies

$$\lim_{n \rightarrow \infty} \sup_{t \geq 0} \left( \frac{1}{t} |A_n^t| \right) = 0. \tag{3.18}$$

Indeed, otherwise there would exist  $\varepsilon > 0$ ,  $t_n \rightarrow \infty$  and  $k_n \rightarrow \infty$  such that

$$\frac{1}{t_n} |A_{k_n}^{t_n}| > \varepsilon \tag{3.19}$$

and this inequality would imply for  $n \rightarrow \infty$

$$\frac{1}{t_n} \int_0^{t_n} |X_s| ds \geq \frac{1}{t_n} \int_{A_{k_n}^{t_n}} |X_s| ds \geq \frac{k_n}{t_n} |A_{k_n}^{t_n}| \geq \varepsilon k_n \rightarrow \infty. \tag{3.20}$$

It follows from the proof of (3.5) that (3.5) and (3.11) hold for all sample points from  $\hat{\Omega}$ . Hence, (3.20) contradicts (3.5). Similarly, it can be shown that (3.11) and (3.16) imply

$$\lim_{n \rightarrow \infty} \sup_{t \geq 1} \left( \frac{1}{t} |C_n^t| \right) = 0. \tag{3.21}$$

Now, we define sets

$$K_n = S_1 \overline{B(n)} + \overline{JB_1(n)} + C_n + S_1 C_n, \quad n \geq 1,$$

where

$$B_1(n) = \{ \psi \in L^1((0, 1), H) : \|\psi\|_{L^1((0, 1), H)} < n \}.$$

Clearly, each  $K_n$  is compact in  $H$  and

$$\begin{aligned} \frac{1}{t} \int_0^t I_{H \setminus K_n}(X_{s+1}) ds &\leq \frac{1}{t} \int_0^t (I_{H \setminus \overline{B(n)}}(X_s) \\ &\quad + I_{L^1((0,1), H \setminus \overline{B_1(n)})}(Q^{-1/2}(f(\alpha_0, X_{\cdot+s}) + u_{\cdot+s})) \\ &\quad + I_{H \setminus C_n}(Z_{s+1}) + I_{H \setminus C_n}(Z_s)) ds \\ &\leq \frac{1}{t} (|A_n^t| + |C_n^t| + |B_n^{t+1}| + |B_n^t|) \rightarrow 0, \quad n \rightarrow \infty, \end{aligned} \quad (3.22)$$

uniformly in  $t \geq 1$ , by (3.17), (3.18), and (3.21). Since the sample point in  $\hat{\mathcal{Q}}$  was arbitrary, we obtain (3.4). If  $u_s = h(\alpha_s, X_s)$  then it remains to show that the family of measures  $\{\mu_t: t \geq 1\}$  is  $\mathbb{P}$ -a.s. tight. Take a sample point from  $\hat{\mathcal{Q}}$  as above and set  $\hat{K} = h(\mathcal{A}, K_n)$ . Then  $\hat{K}_n \subset H$  is compact for each  $n$  and

$$\mu_t(B(r_0) \setminus \hat{K}_n) = \frac{1}{t} \int_0^t I_{B(r_0) \setminus \hat{K}_n}(h(\alpha_s, X_s)) ds = \frac{1}{t} \int_0^t I_{H \setminus K_n}(X_s) ds,$$

hence  $\{\mu_t: t \geq 1\}$  is relatively compact by (3.22). ■

**PROPOSITION 3.3.** *Assume Hypotheses 2.1 and 2.3. Moreover, assume that  $\alpha_0$  is identifiable and the family of measures  $\{\mu_t: t \geq 1\}$  is  $\mathbb{P}$ -a.s. tight. Then*

$$\liminf_{t \rightarrow \infty} \frac{1}{t} \int_0^t |Q^{-1/2}(f(\alpha, X_s) - f(\alpha_0, X_s))|^2 ds > 0 \quad \mathbb{P}\text{-a.s.} \quad (3.23)$$

for each  $\alpha \neq \alpha_0$ .

*Proof.* Let

$$\phi(x) = g_m(\langle x, k_1 \rangle, \dots, \langle x, k_n \rangle) \quad (3.24)$$

for certain  $g_m \in \mathcal{G}$  and  $k_1, \dots, k_n \in \mathcal{X}$ . By the Ito formula we obtain for  $t \geq 0$

$$\begin{aligned} \frac{1}{t} \left( \phi(X_t) - \phi(x) - \int_0^t \langle Q^{1/2} D\phi(X_s), dW_s \rangle \right) \\ = \frac{1}{t} \int_0^t (L^{f_0} \phi(X_s) - \langle u_s, D\phi(X_s) \rangle) ds \quad \mathbb{P}\text{-a.s.}, \end{aligned} \quad (3.25)$$

where  $f_0$  above stands for  $f(\alpha_0, \cdot)$ . Therefore,

$$\begin{aligned} & \frac{1}{t} \left( \phi(X_t) - \phi(x) - \int_0^t \langle Q^{1/2} D\phi(X_s), dW_s \rangle \right) \\ &= \int_{B(r_0) \times H} \psi(y, x) \nu_t^{(u, X)}(dy, dx), \quad \mathbb{P}\text{-a.s.}, \end{aligned} \quad (3.26)$$

where

$$\psi(y, x) = L^{f_0} \phi(x) - \langle y, D\phi(x) \rangle. \quad (3.27)$$

The function  $\psi$  is continuous on  $B(r_0) \times H$  and

$$|\psi(y, x)| \leq k(1 + |x|), \quad x \in H, \quad (3.28)$$

for a certain  $k > 0$ . Using the Strong Law of Large Numbers for Martingales we find that the left hand side of (3.26) tends to zero  $\mathbb{P}$ -a.s. as  $t \rightarrow \infty$  and thereby

$$\lim_{t \rightarrow \infty} \int_{B(r_0) \times H} \psi(y, x) \nu_t^{(u, X)}(dy, dx) = 0, \quad \mathbb{P}\text{-a.s.} \quad (3.29)$$

We can find a measurable  $\Omega_1 \subset \Omega$  such that  $\mathbb{P}(\Omega_1) = 1$  and (3.29) holds for every  $\omega \in \Omega_1$  and any  $\phi$  defined by (3.24). Take a sample path  $(u(\omega), X(\omega)) = (u(\cdot)(\omega), X(\cdot)(\omega))$  for a fixed  $\omega \in \Omega_1$ . By Lemma 3.2 the family of measures  $\{\nu_t^{(u(\omega), X(\omega))} : t \geq 1\}$  is tight. Let  $\nu^{(u(\omega), X(\omega))}$  denote any limit point of this set. The measure  $\nu^{(u(\omega), X(\omega))}$  may be disintegrated to the form

$$\nu^{(u(\omega), X(\omega))} = \nu_1(dy | x) \nu_2(dx),$$

where the dependence on  $\omega$  on the right hand side is suppressed.

Our next aim is to show that

$$\overline{\text{supp}(\nu_2)} = H. \quad (3.30)$$

Since  $H$  is a Polish space the function  $x \rightarrow \nu_1(U|x)$  is Borel-measurable for every Borel set  $U \subset H$  and we can define a bounded Borel function

$$v(x) = \int_{B(r_0)} y \nu_1(dy | x).$$

By (3.29)

$$\begin{aligned} & \int_{B(r_0) \times H} (L^{f_0} \phi(x) - \langle y, D\phi(x) \rangle) v^{(u(\omega), X(\omega))}(dy, dx) \\ &= \int_H (L^{f_0} \phi(x) - \langle v(x), D\phi(x) \rangle) v_2(dx) = 0, \end{aligned} \quad (3.31)$$

for every  $\phi$  given by (3.24). Therefore, by Proposition 2.5 (3.31) holds for every  $\phi \in \mathcal{D}_0$ , hence by Lemma 3.1  $v_2$  is an invariant measure for the equation

$$dX_t = (AX_t + f(\alpha_0, X_t) + v(X_t)) dt + \sqrt{Q} dW_t. \quad (3.32)$$

By the Girsanov Theorem the measure  $v_2$  is equivalent to  $\mu$  and  $\overline{\text{supp}(\mu)} = H$ , hence (3.30) follows. Note that the function

$$x \rightarrow |Q^{-1/2}(f(\alpha, x) - f(\alpha_0, x))|^2$$

is continuous on  $H$ , has at most quadratic growth and for  $\alpha \neq \alpha_0$  is strictly positive on a nonempty open set. For any  $\omega \in \Omega_1$  and any sequence  $t_n \rightarrow \infty$  we can find a subsequence (denoted again by  $(t_n)$ ) and a limit point  $v^{(u, X)}$  of the set  $\{v_{t_n} : n \geq 1\}$  such that

$$\begin{aligned} & \lim_{n \rightarrow \infty} \frac{1}{t_n} \int_0^{t_n} |Q^{-1/2}(f(\alpha, X_s) - f(\alpha_0, X_s))|^2 ds \\ &= \int_{B(r_0) \times H} |Q^{-1/2}(f(\alpha, x) - f(\alpha_0, x))|^2 v^{(u, X)}(dy, dx), \end{aligned}$$

where  $v^{(u, X)}(dy, dx) = v_1^x(dy) v_2(dx)$ . Hence by (3.30)

$$\begin{aligned} & \lim_{n \rightarrow \infty} \frac{1}{t_n} \int_0^{t_n} |Q^{-1/2}(f(\alpha, X_s) - f(\alpha_0, X_s))|^2 ds \\ &= \int_H |Q^{-1/2}(f(\alpha, x) - f(\alpha_0, x))|^2 v_2(dx) > 0. \end{aligned}$$

Since  $\omega \in \Omega_1$  was arbitrary and  $\mathbb{P}(\Omega_1) = 1$ , we obtain (3.23).  $\blacksquare$

## 4. CONSISTENCY OF MLE ESTIMATOR

Following the finite dimensional case (cf. [7, 8]) we can use Proposition 1 to establish strong consistency of the family of maximum likelihood estimators of  $\alpha$ . Let

$$l_t^\alpha = e^{m_t^\alpha - 1/2 a_t^\alpha}, \quad (4.1)$$

where

$$m_t^\alpha = \int_0^t \langle Q^{-1/2}(f(\alpha, X(t)) - f(\alpha_0, X(s))), dW(s) \rangle, \quad (4.2)$$

and

$$a_t^\alpha = \int_0^t |Q^{-1/2}(f(\alpha, X(t)) - f(\alpha_0, X(s)))|^2 ds. \quad (4.3)$$

We define the estimator  $\alpha_t$  as a maximiser of the function  $\alpha \rightarrow l_t^\alpha$ :

$$\alpha_t = \operatorname{argmax}_{\alpha \in \mathcal{A}} l_t^\alpha. \quad (4.4)$$

The fact that  $\alpha_t$  can be realized as an adapted, measurable process follows from Lemma 4.2 below. We will need in this section an additional

**HYPOTHESIS 4.1.** (a)  $\mathcal{A} \subset \mathbb{R}^d$  is a compact set.

(b) There exist  $p > 0$ ,  $k < \infty$  and  $\theta \in (0, 1]$  such that

$$|f(\alpha_1, x) - f(\alpha_2, x)| \leq k |\alpha_1 - \alpha_2|^\theta (1 + |x|^p), \quad \alpha_1, \alpha_2 \in \mathcal{A}, \quad x \in H.$$

(c) For every  $r > 0$

$$\int_H a^r \left( \frac{r_0}{\omega} + |y| \right) \mu(dy) < \infty.$$

**LEMMA 4.2.** Assume Hypotheses 2.1, 4.1. Then the map  $(t, \alpha) \rightarrow m_t^\alpha$  has a continuous modification on  $(0, \infty) \times \mathcal{A}$ , the continuity in  $\alpha$  being uniform with respect to  $t \in [\delta, \infty)$  for each  $\delta > 0$ .

*Proof.* Obviously,  $t \rightarrow m_t^\alpha$  is a.s. continuous for every  $\alpha \in \mathcal{A}$ . thus it suffices to prove a.s. continuity of the function  $\alpha \rightarrow \frac{1}{t} m_t^\alpha$ , uniformly in  $t \in [\delta, \infty)$ . We will prove first that for every  $q \geq 0$

$$\sup_{t \geq 0} \mathbb{E} |X_t|^{2q} < \infty. \quad (4.5)$$

This clearly true if  $f = 0$ . For  $f \neq 0$  we will use the notation  $X_t = Y_t + \tilde{Z}_t$  as in the previous section. Then, similarly as in Lemma 3.2 assumption (iii) of Hypothesis 2.1 yields for  $t \geq 0$

$$\mathbb{E} |Y_t|^{2q} \leq c_1 |x|^{2q} e^{-2kqt} + c_2 \int_0^t e^{-2kq(t-s)} \mathbb{E} |Y_s|^{2q-1} a \left( \frac{r_0}{\omega} + |Z_s| \right) ds \quad (4.6)$$

for any  $q \geq \frac{1}{2}$  and suitable constants  $c_1, c_2$ . for  $q = \frac{1}{2}$  we get immediately

$$\sup_{t \geq 0} \mathbb{E} |Y_t| < \infty, \quad (4.7)$$

since

$$\lim_{s \rightarrow \infty} \mathbb{E} a \left( \frac{r_0}{\omega} + |Z_s| \right) = \int_H a \left( \frac{r_0}{\omega} + |y| \right) \mu(dy) < \infty.$$

For  $q > \frac{1}{2}$  we can use induction with the step  $\frac{1}{2}$ . By the Hölder inequality applied to (4.6) we obtain

$$\begin{aligned} \mathbb{E} |Y_t|^{2q} &\leq c_1 |x|^{2q} e^{-2kqt} \\ &+ c_2 \int_0^t e^{-2kq(t-s)} (\mathbb{E} |Y_s|^{2q-1/2})^{1/b} \left( \mathbb{E} a^{b'} \left( \frac{r_0}{\omega} + |Z_s| \right) \right)^{1/b'} ds, \end{aligned} \quad (4.8)$$

where  $b = \frac{4q-1}{2(2q-1)}$  and  $b' = \frac{b}{b-1}$ . Since  $\mathbb{E} a^b(|Z_s|)$  is bounded by Hypothesis 4.1 we obtain

$$\sup_{t \geq 0} \mathbb{E} |X_t|^{2q} \leq c_3 (\sup_{t \geq 0} \mathbb{E} |Y_t|^{2q} + \sup_{t \geq 0} \mathbb{E} |\tilde{Z}_t|^{2q}) < \infty, \quad (4.9)$$

for a certain  $c_3 > 0$ . Now we have for  $r \geq 1$  by Lemma 4.12 of [25]

$$\begin{aligned} \mathbb{E} \left| \frac{1}{t} m_t^{\alpha_1} - \frac{1}{t} m_t^{\alpha_2} \right|^{2r} &\leq \frac{1}{t^{2r}} \mathbb{E} \left| \int_0^t \langle Q^{-1/2} (f(\alpha_1, X_s) - f(\alpha_2, X_s)), dW_s \rangle \right|^{2r} \\ &\leq \frac{(2r-1)^r r^r t^{r-1}}{t^{2r}} \int_0^t \mathbb{E} |Q^{-1/2} (f(\alpha_1, X_s) - f(\alpha_2, X_s))|^{2r} ds \\ &\leq C_1 t^{-r-1} \int_0^t |\alpha_1 - \alpha_2|^{2\theta r} \mathbb{E} (1 + |X_s|^p)^{2r} ds \\ &\leq C_2 t^{-r} |\alpha_1 - \alpha_2|^{2\theta r}, \end{aligned} \quad (4.10)$$

for some  $C_1, C_2 > 0$ . Taking  $r$  large enough so that  $2\theta r > d$ , and using the Kolmogorov–Totoki criterion for the existence of versions with continuous sample paths (see, e.g., [18]) we conclude the proof. ■

**THEOREM 4.3.** *Let Hypotheses 2.1 and 4.1 hold. Assume that (3.1) is satisfied and the control  $(u_t)$  satisfies the conditions from Proposition 3.3. Then the family of estimators  $(\alpha_t)$  defined by (4.4) is strongly consistent, that is*

$$\mathbb{P}(\lim_{t \rightarrow \infty} \alpha_t = \alpha_0) = 1.$$

*Proof.* Since the process  $t \rightarrow m_t^\alpha$  is a martingale we can find a Wiener process  $\beta^\alpha$  such that  $m_t^\alpha = \beta^\alpha(a_t^\alpha)$  for  $t \geq 0$ . By the Strong Law Large Numbers  $\frac{1}{t}\beta^\alpha(t) \rightarrow 0$   $\mathbb{P}$ -a.s. for  $t \rightarrow \infty$ . Furthermore, by (2.3) we have

$$\mathbb{P}\left(\limsup_{t \rightarrow \infty} \frac{1}{t} a_t^\alpha < \infty\right) = 1, \quad \alpha \in \mathcal{A}.$$

It follows that for every  $\alpha \in \mathcal{A}$

$$\lim_{t \rightarrow \infty} \frac{1}{t} m_t^\alpha = \lim_{t \rightarrow \infty} \frac{\beta^\alpha(a_t^\alpha)}{a_t^\alpha} \frac{a_t^\alpha}{t} = 0, \quad P\text{-a.s.} \tag{4.11}$$

By Lemma 4.2 convergence in (4.11) is uniform with respect to  $\alpha \in \mathcal{A}$ , hence

$$\frac{1}{t} m_t^\alpha \rightarrow 0, \quad \mathbb{P}\text{-a.s.} \tag{4.12}$$

Then the definition of  $l_t^\alpha$  and  $\alpha_t$  yields

$$m_t^{\alpha_t} - \frac{1}{2} a_t^{\alpha_t} \geq m_t^{\alpha_0} - \frac{1}{2} a_t^{\alpha_0}$$

and therefore -a.s.

$$\lim_{t \rightarrow \infty} \frac{1}{t} a_t^{\alpha_t} = \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t |Q^{-1/2}(f(\alpha_s, X_s) - f(\alpha_0, X_s))|^2 ds = 0,$$

which together with Proposition 3.3 completes the proof of the theorem. ■

*Remark 4.4.* Note that our result together with the result from [17] proves the self-optimizing property of the adaptive control defined in [17]

for the case considered there. Indeed, assume that the conditions from [17] (that is, (A1)–(A5) and (3.9)) and, in addition, our Hypothesis 4.1 are satisfied. Moreover, assume (3.1). The adaptive control can be defined in the feedback form

$$u_t = DH(Dv_{\alpha_t}(X_t)), \quad (4.13)$$

where  $H$  is the Hamiltonian of the problem,  $(v_\alpha, \lambda_\alpha)$  is the solution of the ergodic Hamilton–Jacobi–Bellman equation corresponding to the parameter  $\alpha$  (cf. [17, 21] for details) and  $\alpha_t$  is the MLE defined above. Our Hypotheses 2.1, 2.3, and 4.1 are satisfied in that case and from Proposition 3.4 and Corollary 3.1 of [17] it follows that the function  $(\alpha, x) \rightarrow DH(Dv_\alpha(x))$  is continuous on  $\mathcal{A} \times H$ . Therefore, the control  $u_t$  (4.13) satisfies conditions of Proposition (3.3) and we obtain  $\alpha_t \rightarrow \alpha_0$   $\mathbb{P}$ -a.s. by Theorem 4.3. By Proposition 3.5 of [17] it follows that the adaptive control  $(u_t)$  is self-optimizing.

4.1. EXAMPLE. We provide below a simple example of a nonlinear stochastic partial differential equations for which the theory of maximum likelihood estimators developed above is valid.

Let  $A$  be realization in  $H = L^2(0, 1)$  of a differential operator

$$A_0 x(\zeta) = \frac{\partial}{\partial \zeta} \left( a \frac{\partial x}{\partial \zeta} \right) (\zeta) + b(\zeta) \frac{\partial x}{\partial \zeta} (\zeta) + c_0(\zeta) x(\zeta), \quad \zeta \in [0, 1],$$

endowed with the Dirichlet boundary conditions. We assume that  $a$  is Lipschitz on  $[0, 1]$  and  $b, c_0 \in L^\infty(0, 1)$ . Moreover, we assume that

$$a^2(\zeta) \geq m > 0, \quad \zeta \in [0, 1].$$

Under these assumptions  $A$  generates an analytic  $C_0$ -semigroup of contractions on  $L^2(0, 1)$ , see for example [27] for the results on generation of semigroups by differential operators. Under the present assumptions it also known, see [1], that  $A$  generates a semigroup  $S(t)$  of Hilbert–Schmidt operators on  $L^2(0, 1)$  and for all  $T > 0$

$$\int_0^T \|S(t)\|_{HS}^2 ds < \infty.$$

Let  $\mathcal{A} \subset \mathbb{R}^d$  be a compact set and let

$$f_0: \mathcal{A} \times \mathbb{R} \rightarrow \mathbb{R}$$

be such a mapping that  $f_0$  is continuously differentiable in the second argument and

$$\sup_{\zeta, \alpha} \left| \frac{\partial f_0}{\partial \zeta}(\alpha, \zeta) \right| < \infty.$$

We will assume also that

$$|f_0(\alpha_1, \zeta) - f_0(\alpha_2, \zeta)| \leq k |\alpha_1 - \alpha_2|^\theta (1 + |\zeta|), \quad \alpha_1, \alpha_2 \in \mathcal{A}, \zeta \in [0, 1],$$

and

$$\zeta f_0(\alpha, \zeta) \leq -k_1 \zeta^2 + k_2, \quad \alpha \in \mathcal{A}, \zeta_1, \zeta_2 \in [0, 1],$$

where  $k_1 + m\pi^2 > \|c_0\|_\infty$ . Then we define the Nemytskii operator  $f: L^2(0, 1) \rightarrow L^2(0, 1)$ :

$$f(x)(\zeta) = f_0(x(\zeta)), \quad \zeta \in [0, 1].$$

We assume also that the random field  $\{u_t(\zeta) : t \geq 0, \zeta \in [0, 1]\}$  satisfies the following conditions: the process

$$[0, \infty) \ni t \rightarrow u_t(\cdot) \in L^2(0, 1) \text{ is progressively measurable}$$

and

$$\sup_{t \geq 0} \int_0^1 |u_t(\zeta)|^2 d\zeta \leq r_0^2.$$

It is easy to check that Hypotheses 2.1, 2.3 and 4.1 are satisfied in the present case and therefore the stochastic partial differential equation

$$\left\{ \begin{array}{l} \frac{\partial X}{\partial t}(t, \zeta) = \frac{\partial}{\partial \zeta} \left( a \frac{\partial X}{\partial \zeta} \right)(t, \zeta) + b(\zeta) \frac{\partial X}{\partial \zeta}(t, \zeta) \\ \quad + c_0(\zeta) X(t, \zeta) + f_0(\alpha_0, X(t, \zeta)) - u_t(\zeta) + \frac{\partial^2 W}{\partial t \partial \zeta}, \\ X(0, \zeta) = x(\zeta), \zeta \in [0, 1], X(t, 0) = X(t, 1) = 0, \end{array} \right.$$

has a unique solution in the sense defined in Section 2. In the above equation  $W$  stands for the cylindrical Wiener process on  $L^2(0, 1)$ , that is,

$$\mathbb{E} W(t_1, \zeta_1) W(t_2, \zeta_2) = \min(t_1, t_2) \min(\zeta_1, \zeta_2), \quad t_i \geq 0, \zeta_i \in [0, 1].$$

As a consequence we find the strong consistency of the maximum likelihood estimator  $\alpha_t$  of the parameter  $\alpha_0$  provided the family of measures  $\{\mu_t; t \geq 1\}$  is relatively compact on  $L^2(0, 1)$ . In particular, this condition is satisfied if

$$u_t(\zeta) = K(\alpha_t, X(t, \zeta)),$$

where the function  $K: \mathcal{A} \times \mathbb{R} \rightarrow \mathbb{R}$  is uniformly bounded and continuous.

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