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# Convergence rate of maximum likelihood estimator of parameter in stochastic partial differential equation\*



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#### ABSTRACT

Using the recent results obtained by combining Malliavin calculus and Stein's method, we study the rate of convergence of the distribution of the maximum likelihood estimator of a parameter appearing in a stochastic partial differential equation. The aim of this paper is to develop the new techniques, allowing us to improve the rate, given by Mishra and Prakasa Rao (2004), to O(1/N).

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#### 1. Introduction

We investigate asymptotic properties of the maximum likelihood estimators for parameter  $\theta > 0$  occurring in parabolic SPDEs of the following form

$$du(t,x) = \theta \Delta u(t,x) + dW_0(t,x) \tag{1}$$

with the initial and boundary conditions given by

$$\begin{cases} u(0,x) = f(x), & 0 < x < 1, f \in L^2((0,1)) \\ u(t,0) = u(t,1) = 0, & 0 \le t \le T. \end{cases}$$
 (2)

Here  $\Delta = \frac{\partial^2}{\partial x^2}$ , and Q is a nuclear covariance operator for the Wiener process  $W_Q(t,x)$  taking values in  $L^2((0,1))$  so that  $W_Q(t,x) = Q^{1/2}W(t,x)$ , where W is a cylindrical Brownian motion in  $L^2((0,1))$ . It is a standard fact (see Rozovskii, 1990) that  $W_Q(t,x)$  can be represented as

$$dW_{\mathbb{Q}}(t,x) = \sum_{i=1}^{\infty} \sqrt{q_i} e_i(x) W_i(t) \quad \text{a.s.,}$$

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where  $W_i(t)$ ,  $i=1,2,\ldots$  are independent one-dimensional Wiener processes and  $\{e_i\}$ ,  $i=1,2,\ldots$ , is a complete orthonormal system (CONS) in  $L^2((0,1))$  consisting of the eigenvectors of Q, and  $q_i$  are the corresponding eigenvalues of Q. In this paper, we consider the special covariance operator Q with CONS  $e_i = \sin(i\pi x)$ . Then  $\{e_i\}$  is a CONS with the eigenvalues  $q_i = (1-\lambda_i)^{-1}$ ,  $i=1,2,\ldots$ , for the operator Q, where  $\lambda_i = (\pi i)^2$  and  $Q = (I-\Delta)^{-1}$  (see Hübner, Khasminskii, & Rozovskii, 1993).

We define a solution u(t, x) of Eq. (1) with initial and boundary conditions (2) as a formal sum

$$u(t,x) = \sum_{i=1}^{\infty} u_i(t)e_i(x),$$

where the Fourier coefficients  $u_i(t)$  satisfy the following stochastic differential equation:

$$\begin{cases} du_i(t) = -\theta \lambda_i u_i(t) dt + \frac{1}{\sqrt{1+\lambda_i}} dW_i(t), & t > 0, \\ u_i(0) = \nu_i, & \nu_i = \int_0^1 f(x) e_i(x) dx. \end{cases}$$
(3)

Let  $\mathbb{P}_{\theta}$  be the probability measure generated by the process u on C([0,T]). It can be shown that  $\mathbb{P}_{\theta}$  is singular with respect to  $\mathbb{P}_{\theta_0}$  when  $\theta \neq \theta_0$ . Let  $u^N$  be the projection of u on C([0,T]) onto the subspace spanned by  $\{e_1,\ldots,e_N\}$  in  $L^2((0,1))$  and  $\mathbb{P}^N_{\theta}$  be the probability measure corresponding to the process  $u^N$ . Then likelihood ratio for the projection  $u^N$  can be expressed as

$$\log \frac{d\mathbb{P}_{\theta}}{d\mathbb{P}_{\theta_0}}(u^N) = -\sum_{i=1}^N \lambda_i(\lambda_i + 1) \left\{ (\theta - \theta_0) \int_0^T u_i(t) (du_i(t) + \theta_0 \lambda_i u_i(t) dt) + \frac{1}{2} (\theta - \theta_0)^2 \lambda_i \int_0^T u_i^2(t) dt \right\}. \tag{4}$$

It is clear from (4) that the MLE  $\hat{\theta}_N$  of  $\theta$  based on  $u^N$  is given by

$$\hat{\theta}_{N} = -\frac{\sum_{i=1}^{N} \lambda_{i} \sqrt{\lambda_{i} + 1} \int_{0}^{T} u_{i}(t) du_{i}(t)}{\sum_{i=1}^{N} \lambda_{i}^{2} (\lambda_{i} + 1) \int_{0}^{T} u_{i}^{2}(t) dt}.$$
(5)

In the paper Mishra and Prakasa Rao (2004), the authors obtain a Berry–Esseen type bound for the MLE  $\hat{\theta}_N$ :

**Theorem 1** (Mishra and Prakasa Rao). There exists a constant C depending on  $\theta_0$ ,  $||f||_{L^2((0,1))}^2$  and T such that for any  $\gamma > 0$  and  $N \ge N_0$ , depending on  $\theta_0$  and T,

$$\sup_{z \in \mathbb{R}} \left| \mathbb{P}_{\theta_0} \left( \sqrt{\varphi_N(\theta_0)} (\hat{\theta}_N - \theta_0) \le z \right) - \mathbb{P}(Z \le z) \right| \le C N^{3+\gamma} \left( \frac{1 + \sqrt{T}}{T N^3 + \sum_{k=1}^N k^4 \nu_k^2} \right) + 3\sqrt{N^{-\gamma}}, \tag{6}$$

where the random variable Z has the normal distribution with the zero mean and unit variance, and the normalizing factor  $\varphi_N(\theta)$  is

$$\varphi_N(\theta) = \frac{1}{2\theta} \sum_{i=1}^N \lambda_i (\lambda_i + 1) \left\{ \nu_i^2 (1 - e^{-2\theta \lambda_i T}) + \frac{T}{\lambda_i + 1} \right\}.$$

In Remark 4.4 in Mishra and Prakasa Rao (2004), the authors argue that the bound in Theorem 1 is of order  $O(N^{\gamma-2}) + O(N^{-\gamma/2})$  provided  $\sum_{k=1}^{N} k^4 v_k^2 \ge g(N) = O(N^5)$ , and in such case, the upper bound can be obtained to be of order  $O(N^{-2/3})$  by choosing  $\gamma = 4/3$ . However, if  $\sum_{k=1}^{N} k^4 v_k^2 \le g(N) = O(N^3)$  (for example, f = 0 i.e.  $v_k = 0$  for all k = 1, 2, ...), then the upper bound in (6) is given by

$$\sup_{z \in \mathbb{R}} \left| \mathbb{P}_{\theta_0} \left( \sqrt{\varphi_N(\theta_0)} (\hat{\theta}_N - \theta_0) \le z \right) - \mathbb{P}(Z \le z) \right| \le C N^{\gamma}. \tag{7}$$

In such case, from the upper bound (7) of the *Kolmogorov distance*, we cannot show that the normal approximation of the MLE  $\hat{\theta}_N$  holds. Hence the sharp upper bound is needed to prove the normal approximation through the *Kolmogorov distance*. Let  $\{F_N\}$  be a sequence of zero-mean real-valued random variables with the form of a functional of an infinite dimensional Gaussian fields. In the paper Nourdin and Peccati (2009b), the authors, by combining Malliavin calculus with Stein's method, obtain explicit bounds of the type

$$\sup_{z \in \mathbb{R}} \left| \mathbb{P}(F_N \le z) - \mathbb{P}(Z \le z) \right| \le \phi(N), \tag{8}$$

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