

ASYMPTOTIC PROPERTIES OF AN ESTIMATOR  
FOR THE DRIFT COEFFICIENT  
OF A STOCHASTIC DIFFERENTIAL EQUATION  
WITH FRACTIONAL BROWNIAN MOTION

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ABSTRACT. A stochastic differential equation with respect to fractional Brownian motion is considered. We study the maximum likelihood estimator for the drift coefficient. We assume that the coefficient belongs to a given compact set of functions and prove the strong consistency of the estimator and its asymptotic normality.

Let  $(\Omega, F, P)$  be a probability space and let a real stochastic process

$$\{x(t), t \geq 0\}$$

and a fractional Wiener process (fractional Brownian motion)  $\{Z(t), t \geq 0\}$  be defined on  $(\Omega, F, P)$ , where  $E Z(t) = 0$ ,  $Z(0) = 0$ , and

$$E\{Z(t)Z(\tau)\} = \frac{1}{2}(t^{2H} + \tau^{2H} - |t - \tau|^{2H}), \quad \frac{1}{2} < H < 1.$$

Assume that a stochastic process  $\{y(t), t \geq 0\}$  possesses the stochastic differential

$$(1) \quad dy(t) = s_0(t)x(t) dt + dZ(t), \quad t \geq 0,$$

where  $s_0$  is a certain unknown function.

The problem considered in this paper is to estimate the function  $s_0$  from the observations  $\{(x(t), y(t)), 0 \leq t \leq T\}$ .

Note that an analogous problem is considered in [1], where a standard Wiener process is substituted for  $Z$  in equation (1).

1. CONSISTENCY AND THE ASYMPTOTIC DISTRIBUTION OF ESTIMATORS

Below we list the main conditions to be imposed on the function  $s_0$  and the stochastic processes  $\{x(t), t \geq 0\}$  and  $\{Z(t), t \geq 0\}$ .

1. The function  $s_0$  belongs to the family  $K$  of all  $2\pi$ -periodic functions  $s: \mathbf{R} \rightarrow \mathbf{R}$  whose Fourier coefficients

$$c_k(s) = \frac{1}{2\pi} \int_0^{2\pi} s(t)e^{ikt} dt, \quad k \in \mathbf{Z},$$

are such that  $|c_0(s)| \leq L$  and  $|c_k(s)| \leq L|k|^{-a}$ ,  $k \neq 0$ , for some constants  $L > 0$  and  $a > 3$ .

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It is obvious that the functions of the family  $K$  are continuously differentiable and that  $K$  is a compact set with respect to the uniform convergence of functions.

For functions  $s \in K$ , we introduce the norm

$$\|s\| = \left( \frac{1}{2\pi} \int_0^{2\pi} s^2(t) dt \right)^{1/2}.$$

We say that  $s_0 \in K$  is an interior point of  $K$  if

$$|c_0(s_0)| \leq \tilde{L}, \quad |c_k(s_0)| \leq \frac{\tilde{L}}{|k|^a}, \quad k \neq 0,$$

for some constant  $\tilde{L} < L$ .

2. The processes  $\{x(t), t \geq 0\}$  and  $\{Z(t), t \geq 0\}$  are independent.
3. There exists a constant  $c > 0$  such that

$$\mathbf{E} \{(x(t))^2\} \leq c$$

for all  $t \geq 0$ .

4. The trajectories of the process  $\{x(t), t \geq 0\}$  are continuously differentiable with probability 1.
5. The process  $\{(x(t))^2, t \geq 0\}$  is stationary in the wide sense.

Denote by  $r(t)$  the covariance function of the process  $\{(x(t))^2, t \geq 0\}$ :

$$r(t) = \mathbf{E} \{((x(t))^2 - \mathbf{E}(x(0))^2) ((x(0))^2 - \mathbf{E}(x(0))^2)\}.$$

6. For some  $L_1 > 0$  and  $\gamma > 0$ ,

$$\int_0^T |r(t)| dt \leq L_1 T^{1-\gamma}, \quad T > 0.$$

Using conditions 2–4 and integration by parts [2], we define the following stochastic integral:

$$\int_a^b s(\tau)x(\tau) dZ(\tau), \quad 0 \leq a \leq b,$$

for an arbitrary continuously differentiable function  $s: \mathbf{R} \rightarrow \mathbf{C}$  (where  $\mathbf{C}$  is the set of complex numbers).

In what follows we need the following properties of the latter integral:

$$(2) \quad \mathbf{E} \left\{ \left[ \int_0^t s(\tau)x(\tau) dZ(\tau) \right]^2 \right\} \leq c_1 \left( \int_0^t |s(\tau)|^{1/H} d\tau \right)^{2H}, \quad t > 0,$$

where  $c_1$  is a constant, and

$$(3) \quad \mathbf{E} \left\{ \sup_{a \leq t \leq b} \left[ \int_a^t s(\tau)x(\tau) dZ(\tau) \right]^2 \right\} \leq c_2 (b-a) \left( \int_a^b |s(\tau)|^{2/(2H-1)} d\tau \right)^{2H-1},$$

$$0 \leq a < b,$$

where  $c_2$  is another constant.

Note that the constants  $c_1$  and  $c_2$  depend on  $H$ .

To prove (2) and (3) one uses well-known properties of the stochastic integral with respect to a fractional Wiener process ([3, 4]) and the mutual independence of the processes  $\{x(t), t \geq 0\}$  and  $\{Z(t), t \geq 0\}$ .

With probability one, we have

$$\begin{aligned} & \mathbb{E} \left\{ \left[ \int_0^t s(\tau)x(\tau) dZ(\tau) \right]^2 / \sigma \{x(\tau), \tau \geq 0\} \right\} \\ &= H(2H-1) \int_0^t \int_0^t s(\tau_1)x(\tau_1)s(\tau_2)x(\tau_2)|\tau_1-\tau_2|^{2H-2} d\tau_1 d\tau_2. \end{aligned}$$

Thus

$$\begin{aligned} & \mathbb{E} \left\{ \left[ \int_0^t s(\tau)x(\tau) dZ(\tau) \right]^2 \right\} \\ &= H(2H-1) \int_0^t \int_0^t s(\tau_1)s(\tau_2)\mathbb{E}\{x(\tau_1)x(\tau_2)\}|\tau_1-\tau_2|^{2H-2} d\tau_1 d\tau_2 \\ &\leq c \int_0^t \int_0^t |s(\tau_1)||s(\tau_2)||\tau_1-\tau_2|^{2H-2} d\tau_1 d\tau_2 \\ &\leq c_1 \left( \int_0^t |s(\tau)|^{1/H} d\tau \right)^{2H}. \end{aligned}$$

Note that the latter inequality is proved in [3].

Therefore property (2) is proved.

Further, it is shown in [4] that, with probability 1,

$$\begin{aligned} & \mathbb{E} \left\{ \sup_{a \leq t \leq b} \left[ \int_a^t s(\tau)x(\tau) dZ(\tau) \right]^2 / \sigma \{x(\tau), \tau \geq 0\} \right\} \\ &\leq c_3 \left( \int_a^b |s(\tau)x(\tau)|^{1/H} d\tau \right)^{2H} \\ &\leq c_3 \left( \left[ \int_a^b (|s(\tau)|^{1/H})^{\alpha/(\alpha-1)} d\tau \right]^{(\alpha-1)/\alpha} \left[ \int_a^b (|x(\tau)|^{1/H})^\alpha d\tau \right]^{1/\alpha} \right)^{2H} \\ &= c_3 \left( \int_a^b |s(\tau)|^{2/(2H-1)} d\tau \right)^{2H-1} \int_a^b |x(\tau)|^2 d\tau \end{aligned}$$

via the Hölder inequality with  $\alpha = 2H$  and  $\beta = 2H/(2H-1)$ .

Then

$$\begin{aligned} \mathbb{E} \left\{ \sup_{a \leq t \leq b} \left[ \int_a^t s(\tau)x(\tau) dZ(\tau) \right]^2 \right\} &\leq c_3 \left( \int_a^b |s(\tau)|^{2/(2H-1)} d\tau \right)^{2H-1} \int_a^b \mathbb{E} \{ |x(\tau)|^2 \} d\tau \\ &\leq c_2 (b-a) \left( \int_a^b |s(\tau)|^{\frac{2}{2H-1}} d\tau \right)^{2H-1}. \end{aligned}$$

Hence property (3) is proved too.

We turn to the estimation of the function  $s_0$ . Consider the estimator defined as the point of maximum of the functional

$$(4) \quad Q_T(s) = \frac{1}{T} \int_0^T s(t)x(t) dy(t) - \frac{1}{2T} \int_0^T s^2(t)x^2(t) dt, \quad s \in K.$$

This estimator exists with probability one. Denote by  $s_T$  an arbitrary point of maximum of (4). As in the paper [5], the properties of the family  $K$  imply that the function  $s_T(t, \omega)$  can be chosen to be a separable measurable process.

**Lemma 1.1.** *Let conditions 1–4 hold. Then*

$$\mathbb{P} \left\{ \lim_{T \rightarrow \infty} \max_{s \in K} \left| \frac{1}{T} \int_0^T s(t)x(t) dZ(t) \right| = 0 \right\} = 1.$$

*Proof.* Put

$$\eta_T = \max_{s \in K} \left| \frac{1}{T} \int_0^T s(t)x(t) dZ(t) \right|.$$

Expanding  $s$  in the Fourier series we obtain

$$\begin{aligned} \mathbb{E} \left\{ (\eta_T)^2 \right\} &= \mathbb{E} \left\{ \max_{s \in K} \left( \frac{1}{T} \sum_{k=-\infty}^{\infty} \left[ c_k(s) \int_0^T e^{ikt} x(t) dZ(t) \right] \right)^2 \right\} \\ (5) \quad &\leq \mathbb{E} \left\{ \left( \frac{1}{T} \sum_{k=-\infty}^{\infty} \left[ \frac{L}{|k|^a} \left| \int_0^T e^{ikt} x(t) dZ(t) \right| \right] \right)^2 \right\}. \end{aligned}$$

By definition, the denominator of the term corresponding to  $k = 0$  in the latter sum (and in similar sums throughout below) is equal to 1 but not  $|k|^a$ .

Applying the Cauchy–Bunyakovskiĭ inequality and the first property of the stochastic integral, we derive from relation (5) that

$$(6) \quad \mathbb{E} \left\{ (\eta_T)^2 \right\} \leq \left( \frac{1}{T} \sum_{k=-\infty}^{\infty} \frac{L}{|k|^a} \left[ \mathbb{E} \left| \int_0^T e^{ikt} x(t) dZ(t) \right|^2 \right]^{1/2} \right)^2 \leq \frac{c_4}{T^{2(1-H)}},$$

where  $c_4 = c_1 L^2 \left( \sum_{k=-\infty}^{\infty} |k|^{-a} \right)^2$ .

It is clear that there exists a positive integer number  $p$  such that  $2p(1-H) > 1$ . Consider the sequence  $T(n) = n^p$ ,  $n \in \mathbf{N}$ . According to bound (6) and the Borel–Cantelli lemma,

$$(7) \quad \mathbb{P} \left( \lim_{n \rightarrow \infty} \eta_{T(n)} = 0 \right) = 1.$$

For  $T_0 \geq 1$ ,

$$(8) \quad \sup_{T > T_0} \eta_T \leq \sup_{n: T(n+1) > T_0} \sup_{T(n) \leq T \leq T(n+1)} \eta_T.$$

For all  $n$ ,

$$\begin{aligned} \sup_{T(n) \leq T \leq T(n+1)} \eta_T &= \eta_{T(n)} + \sup_{T(n) \leq T \leq T(n+1)} (\eta_T - \eta_{T(n)}) \\ &\leq \eta_{T(n)} \\ &\quad + \sup_{T(n) \leq T \leq T(n+1)} \max_{s \in K} \left| \frac{1}{T} \int_0^T s(t)x(t) dZ(t) - \frac{1}{T(n)} \int_0^{T(n)} s(t)x(t) dZ(t) \right| \\ (9) \quad &\leq \eta_{T(n)} + \frac{T(n+1) - T(n)}{(T(n))^2} \max_{s \in K} \left| \int_0^{T(n)} s(t)x(t) dZ(t) \right| \\ &\quad + \frac{1}{T(n)} \sup_{T(n) \leq T \leq T(n+1)} \max_{s \in K} \left| \int_{T(n)}^T s(t)x(t) dZ(t) \right| \\ &= \frac{T(n+1)}{T(n)} \eta_{T(n)} + \zeta_n, \end{aligned}$$

where

$$\zeta_n = \frac{1}{T(n)} \sup_{T(n) \leq T \leq T(n+1)} \max_{s \in K} \left| \int_{T(n)}^T s(t)x(t) dZ(t) \right|.$$

We have

$$\frac{T(n+1)}{T(n)} = \left(1 + \frac{1}{n}\right)^p \rightarrow 1, \quad n \rightarrow \infty.$$

Taking into account equality (7) we get

$$(10) \quad \mathbb{P} \left\{ \lim_{n \rightarrow \infty} \frac{T(n+1)}{T(n)} \eta_{T(n)} = 0 \right\} = 1.$$

Performing elementary transformations and using properties of the stochastic integral, we obtain

$$\begin{aligned} \mathbb{E} \left\{ (\zeta_n)^2 \right\} &\leq \frac{1}{(T(n))^2} \mathbb{E} \left( \sum_{k=-\infty}^{\infty} \left[ \frac{L}{|k|^a} \sup_{T(n) \leq T \leq T(n+1)} \left| \int_{T(n)}^T e^{ikt} x(t) dZ(t) \right| \right] \right)^2 \\ &\leq \frac{1}{(T(n))^2} \left( \sum_{k=-\infty}^{\infty} \left[ \frac{L}{|k|^a} \left( \mathbb{E} \left\{ \sup_{T(n) \leq T \leq T(n+1)} \left| \int_{T(n)}^T e^{ikt} x(t) dZ(t) \right|^2 \right\} \right)^{1/2} \right] \right)^2 \\ &\leq \frac{(T(n+1) - T(n))^{2H}}{(T(n))^2} c_2 L^2 \left( \sum_{k=-\infty}^{\infty} \frac{1}{|k|^a} \right)^2 \\ &= \frac{c_5}{n^{2p(1-H)}} \left( \left(1 + \frac{1}{n}\right)^p - 1 \right)^{2H} \\ &\leq \frac{c_6}{n^{2p(1-H)}}, \end{aligned}$$

where  $c_5$  and  $c_6$  are some constants. This implies that

$$(11) \quad \mathbb{P} \left( \lim_{n \rightarrow \infty} \zeta_n = 0 \right) = 1.$$

Now the lemma follows from relations (8)–(11).  $\square$

*Remark 1.1.* If the assumptions of Lemma 1.1 hold, then

$$\mathbb{P} \left\{ \lim_{T \rightarrow \infty} \max_{s \in K} \left| \frac{1}{T} \int_0^T (s(t) - s_0(t)) x(t) dZ(t) \right| = 0 \right\} = 1.$$

**Lemma 1.2** ([1]). *Let  $\{\xi(t), t \in \mathbf{R}\}$  be a real wide sense stationary stochastic process with mean  $\mathbb{E} \xi(t) = 0$  and whose covariance function  $r(t) = \mathbb{E}\{\xi(t)\xi(0)\}$ ,  $t \in \mathbf{R}$ , is such that*

$$\int_0^T |r(t)| dt \leq LT^{1-\gamma}$$

for all  $T > 0$  and some positive numbers  $L$  and  $\gamma$ . Then

$$\mathbb{P} \left\{ \lim_{T \rightarrow \infty} \sup_{s \in K} \left| \frac{1}{T} \int_0^T s(t)\xi(t) dt \right| = 0 \right\} = 1.$$

*Remark 1.2* ([1]). Lemma 1.2 remains true if the difference of two functions of the family  $K$  is substituted for  $s \in K$  in the above integral.

*Remark 1.3.* Lemma 1.2 holds also for the square of the difference of two functions of the family  $K$ .

Remark 1.3 can be proved by an observation that the square of the difference of two functions belonging to  $K$  can be used in the proof of Lemma 2.2 in [1].

**Theorem 1.1.** *Let the assumptions of Lemma 1.1 as well as conditions 5 and 6 hold. Then*

$$\mathbb{P} \left\{ \lim_{T \rightarrow \infty} \sup_{t \in \mathbf{R}} |s_T(t) - s_0(t)| = 0 \right\} = 1.$$

*Proof.* Note that

$$(12) \quad \begin{aligned} Q_T(s_T) - Q_T(s_0) &= \frac{1}{T} \int_0^T (s_T(t) - s_0(t)) x(t) dZ(t) \\ &\quad - \frac{1}{2T} \int_0^T (s_T(t) - s_0(t))^2 (x(t))^2 dt. \end{aligned}$$

By the definition of the estimator  $s_T$ ,

$$Q_T(s_T) \geq Q_T(s_0).$$

Then

$$(13) \quad \begin{aligned} &\max_{s \in K} \left| \frac{1}{T} \int_0^T (s(t) - s_0(t)) x(t) dZ(t) \right| \\ &\quad + \max_{s \in K} \left| \frac{1}{2T} \int_0^T (s_T(t) - s_0(t))^2 [(x(t))^2 - \mathbb{E} \{(x(0))^2\}] dt \right| \\ &\geq \frac{1}{2T} \mathbb{E} \{(x(0))^2\} \int_0^T (s_T(t) - s_0(t))^2 dt \end{aligned}$$

by equality (12).

Lemmas 1.1 and 1.2 together with Remarks 1.1 and 1.3 and relation (13) imply that

$$(14) \quad \mathbb{P} \left\{ \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (s_T(t) - s_0(t))^2 dt = 0 \right\} = 1,$$

whence

$$(15) \quad \mathbb{P} \left( \lim_{T \rightarrow \infty} \|s_T - s_0\| = 0 \right) = 1.$$

Now relation (15) yields Theorem 1.1. □

In what follows we will make use of the following conditions.

7. The process  $\{x(t), t \geq 0\}$  is equal to 1 for all  $t$ .
8. Let a function  $\varphi \in K$  be such that
  - 1)  $\frac{1}{2\pi} \int_0^{2\pi} \varphi^2(t) dt = 1$ ,
  - 2)  $\lim_{T \rightarrow \infty} H(2H-1)T^{-2H} \int_0^T \int_0^T \varphi(t_1) \varphi(t_2) |t_1 - t_2|^{2H-2} dt_1 dt_2 = \Delta$ .
9. The function  $s_0$  is an interior point of the set  $K$ .

**Theorem 1.2.** *Let conditions 1 and 7–9 hold. Then the random variable*

$$\frac{T^{1-H}}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt$$

*converges in distribution as  $T \rightarrow \infty$  to the Gaussian law  $N(0, \Delta)$  with mean 0 and variance  $\Delta$ .*

*Proof.* By Theorem 1.1, the function  $s_T$  is an interior point of the family  $K$  with probability converging to 1 as  $T \rightarrow \infty$ . It is easy to show that the functional  $Q_T$  is differentiable at the point  $s_T$  with the same probability. Indeed, we evaluate the weak differential of  $Q_T$  in some neighborhood of  $s_T$  as follows:

$$\begin{aligned} DQ_T(s, h) &= \left. \frac{d}{d\varepsilon} Q_T(s + \varepsilon h) \right|_{\varepsilon=0} \\ &= \frac{1}{T} \int_0^T h(t) dZ(t) + \frac{1}{T} \int_0^T h(t) s_0(t) dt - \frac{1}{T} \int_0^T h(t) s(t) dt. \end{aligned}$$

Properties of the differential  $DQ_T(s, h)$  imply that the strong differential of  $Q_T$  at the point  $s_T$  exists and it coincides with the weak differential [6].

By the necessary condition for the existence of an extremum,

$$DQ_T(s_T, \varphi) = 0$$

with probability converging to 1 as  $T \rightarrow \infty$ .

Thus, with the same probability,

$$(16) \quad \frac{1}{T} \int_0^T \varphi(t) dZ(t) + \frac{1}{T} \int_0^T \varphi(t) (s_0(t) - s_T(t)) dt = 0$$

as  $T \rightarrow \infty$ .

We add to and subtract from the left hand side of (16) the following expression:

$$\frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt - \frac{1}{T} \int_0^T \varphi^2(t) dt.$$

Note that

$$\begin{aligned} & - \frac{1}{T} \int_0^{2\pi \lfloor \frac{T}{2\pi} \rfloor} \varphi(t) (s_T(t) - s_0(t)) dt + \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt - \frac{1}{T} \int_0^{2\pi \lfloor \frac{T}{2\pi} \rfloor} \varphi^2(t) dt \\ &= - \frac{1}{T} \left[ \frac{T}{2\pi} \right] \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt \\ & \quad + \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt - \frac{1}{T} \left[ \frac{T}{2\pi} \right] \int_0^{2\pi} \varphi^2(t) dt \\ &= 0. \end{aligned}$$

Thus

$$\begin{aligned} & \frac{1}{T} \int_0^T \varphi(t) dZ(t) + \frac{1}{T} \int_{2\pi \lfloor \frac{T}{2\pi} \rfloor}^T \varphi(t) (s_0(t) - s_T(t)) dt \\ & \quad + \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt - \frac{1}{T} \int_{2\pi \lfloor \frac{T}{2\pi} \rfloor}^T \varphi^2(t) dt \\ & \quad - \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt - \frac{1}{T} \int_0^T \varphi^2(t) dt \\ &= 0 \end{aligned}$$

with probability converging to 1 as  $T \rightarrow \infty$ . Hence, with the same probability,

$$\begin{aligned}
& T^{1-H} \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt \\
&= \left( \frac{1}{T} \int_0^T \varphi^2(t) dt \right)^{-1} \\
(17) \quad & \times \left( \frac{1}{T^H} \int_0^T \varphi(t) dZ(t) + \frac{1}{T^H} \int_{2\pi[\frac{T}{2\pi}]}^T \varphi(t) (s_0(t) - s_T(t)) dt \right. \\
& \quad \left. + \frac{1}{2\pi} \int_0^{2\pi} \varphi(t) (s_T(t) - s_0(t)) dt \frac{1}{T^H} \int_{2\pi[\frac{T}{2\pi}]}^T \varphi^2(t) dt \right).
\end{aligned}$$

Consider

$$\frac{1}{T} \int_0^T \varphi^2(t) dt = \frac{1}{T} \left( \int_0^{2\pi[\frac{T}{2\pi}]} \varphi^2(t) dt + \int_{2\pi[\frac{T}{2\pi}]}^T \varphi^2(t) dt \right).$$

Then

$$\begin{aligned}
\frac{1}{T} \int_0^{2\pi[\frac{T}{2\pi}]} \varphi^2(t) dt &= \frac{1}{T} \left[ \frac{T}{2\pi} \right] \int_0^{2\pi} \varphi^2(t) dt = \frac{1}{T} \left[ \frac{T}{2\pi} \right] 2\pi \rightarrow 1, \quad T \rightarrow \infty, \\
\frac{1}{T} \int_{2\pi[\frac{T}{2\pi}]}^T \varphi^2(t) dt &\rightarrow 0, \quad T \rightarrow \infty.
\end{aligned}$$

Hence

$$(18) \quad \frac{1}{T} \int_0^T \varphi^2(t) dt \rightarrow 1, \quad T \rightarrow \infty.$$

Since the functions of the family  $K$  are bounded, the second and third terms in the expression on the right hand side of (17) converge to 0 with probability 1 as  $T \rightarrow \infty$ .

Now we study the random variable

$$\xi_T = \frac{1}{T^H} \int_0^T \varphi(t) dZ(t).$$

Its distribution is normal [3] with mean 0 and variance

$$\frac{H(2H-1)}{T^{2H}} \int_0^T \int_0^T \varphi(t_1) \varphi(t_2) |t_1 - t_2|^{2H-2} dt_1 dt_2.$$

The assumptions of the theorem and properties of Gaussian random variables imply that  $\xi_T$  weakly converges to  $N(0, \Delta)$  as  $T \rightarrow \infty$ .

Taking into account (18) and the preceding reasoning, we prove that the right hand side of equality (17) converges in distribution to  $N(0, \Delta)$  as  $T \rightarrow \infty$ . This completes the proof of Theorem 1.2.  $\square$

## 2. CONCLUDING REMARKS

The results concerning the asymptotic behavior of the estimator of the drift diffusion of a stochastic differential equation with respect to fractional Brownian motion obtained above imply that the proposed estimators are optimal, and this allows one to use them for solving various applied problems.



## BIBLIOGRAPHY

1. A. Ya. Dorogovtsev, *The Theory of Estimates of the Parameters of Random Processes*, Vyscha Shkola, Kiev, 1982. (Russian) MR668517 (84h:62122)
2. M. Zähle, *Integration with respect to fractal functions and stochastic calculus*, Probab. Theory Relat. Fields **III** (1998), 333–374. MR1640795 (99j:60073)
3. J. Mémin, Yu. Mishura, and E. Valkeila, *Inequalities for the moments of Wiener integrals with respect to a fractional Brownian motion*, Statist. Probab. Lett. **51** (2001), 197–206. MR1822771 (2002b:60096)
4. Yu. V. Kravych and Yu. S. Mishura, *Some maximal inequalities for moments of Wiener integrals with respect to fractional Brownian motion*, Teor. Imovir. Mat. Stat. **61** (1999), 72–83; English transl. in Theory Probab. Math. Statist. **61** (2000), 75–86. MR1866968 (2002h:60072)
5. J. Pfanzagl, *On the measurability and consistency of minimum contrast estimates*, Metrika **14** (1969), 249–272.
6. A. N. Kolmogorov and S. V. Fomin, *Elements of the Theory of Functions and Functional Analysis*, Nauka, Moscow, 1968; English transl. from the first (1960) Russian ed., Graylock Press, Albany, N.Y., 1961. MR0234241 (38:2559)

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