

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 (Ω, 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 (Ω, 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 2π -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

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

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). □

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 γ . 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 Δ .

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 ξ_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.

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