

LAN property for fractional SDEs

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A joint work with Eulalia Nualart and Samy Tindel.

Outline

- 1 Introduction
- 2 Local Asymptotic Normality property (LAN)
- 3 The Ornstein-Uhlenbeck case - Efficiency of MLE

Presentation Outline

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Introduction

Consider \mathbb{R}^d -valued equation:

$$Y_t = y_0 + \int_0^t b(Y_s; \theta) ds + \sigma B_t, \quad t \in [0, \tau].$$

- $y_0 \in \mathbb{R}^d$: given initial.
- $B = (B^1, \dots, B^d)$: driving process.
- $\theta \in \mathbb{R}^q$: *unknown* parameter.
- $\{b(\cdot; \theta), \theta \in \Theta\}$: known family of drift coefficients. $\Theta \subset \subset \mathbb{R}^q$.
- $\sigma \in \mathbb{R}^{d \times d}$: known diffusion coefficient.

\hookrightarrow We are concerned with the estimation problem of θ while Y is observed.

Fractional Brownian motion

Definition

A fractional Brownian motion $B = (B^1, \dots, B^d)$ with Hurst parameter $H \in (0, 1)$:

- centered Gaussian process
- $\mathbb{E}(B_s^i B_t^i) = \frac{1}{2}(|t|^{2H} + |s|^{2H} - |t - s|^{2H})$
- $\mathbb{E}(B_s^i B_t^j) = 0$ for $i \neq j$

Motivations: finance, geophysics (water cycle, atmosphere, and so on), ethernet traffic, some inverse problems in mathematical biology, and so on.

Parameter estimation results

Available parameter estimation results:

- $b(x; \theta)$ linear in both x and θ : [Belfadli, Es-Sebaiy, Ouknine \(2011\)](#); [Hu, Nualart \(2010\)](#); [Kleptsyna, Le Breton \(2002\)](#); [Le Breton \(1998\)](#), etc.
- $b(x; \theta)$ linear and discrete observation: [Brouste, Lacus \(2013\)](#); [Xiao, Zhang, Xu \(2011\)](#); etc.
- $b(x; \theta)$ linear in θ : [Kozachenko, Melnikov, Mishura \(2004\)](#); [Tudor, Viens \(2007\)](#); [Hu, Nualart, Zhou \(2018\)](#); etc.

Question: parameter estimation for a general drift $b(x; \theta)$ and for multidimensional case?

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LAN property

Denote P_θ (P_θ^τ): probability law of the solution y in the space $C^\lambda(\mathbb{R}_+; \mathbb{R}^d)$ ($C^\lambda([0, \tau]; \mathbb{R}^d)$).

Definition

The parametric statistical model $\{P_\theta, \theta \in \Theta\}$ satisfies LAN property at $\theta \in \Theta$ if there exist:

- A $q \times q$ invertible matrix $\varphi_\tau(\theta)$,
- A $q \times q$ positive definite matrix $\Sigma(\theta)$,

such that for any $u \in \mathbb{R}^q$,

$$\log \left(\frac{dP_{\theta + \varphi_\tau(\theta)u}^\tau}{dP_\theta^\tau} \right) \xrightarrow{\mathcal{L}} u^T \mathcal{N}(0, \Sigma(\theta)) - \frac{1}{2} u^T \Sigma(\theta) u \quad \text{as } \tau \rightarrow \infty.$$

Minimax theorem

Theorem

Suppose

- $(P_\theta)_{\theta \in \Theta}$ satisfies the LAN property at some θ .
- $(\hat{\theta}_\tau)_{\tau \geq 0}$: a family of estimators of the parameter θ .

Then for any loss function ℓ we have:

$$\lim_{\delta \rightarrow 0} \liminf_{\tau \rightarrow \infty} \sup_{\theta' \in \Theta: |\theta' - \theta| < \delta} \mathbb{E}_{\theta'} \left[\ell \left(\varphi_\tau^{-1}(\theta)(\hat{\theta}_\tau - \theta') \right) \right] \geq \mathbb{E}[\ell(Z)],$$

where $Z \sim \mathcal{N}(0, \Sigma(\theta)^{-1})$.

Example of a loss function: $\ell(u) = |u|^p, p > 0$.

Stability assumption

Hypothesis

- regularity assumptions on the drift coefficient b .
- linear growth and inward assumptions on b .
- diffusion coefficient σ invertible.

Ergodic properties

Theorem

Suppose the hypothesis holds.

Then

(i) *There exists an ergodic process \bar{Y} such that*

$$\lim_{t \rightarrow \infty} |Y_t - \bar{Y}_t| = 0, \quad \text{a.s.}$$

(ii) *For a Frechet differentiable functional F on $C^\beta([-\rho, 0])$, $\beta < H$, we have:*

$$\lim_{\tau \rightarrow \infty} \frac{1}{\tau} \int_0^\tau F(Y|_{[t-\rho, t]}) dt = \mathbb{E}[F(\bar{Y}|_{[-\rho, 0]})], \quad \text{a.s.}$$

A Girsanov-type lemma

Lemma

Suppose

- B : a fractional Brownian motion with $H > 1/2$
- g : a process in $I_+^{H-1/2}(L^2(\mathbb{R}))$
- $Q_t = B_t + \int_0^t g_s ds$
- $D_+^{H-1/2}g$ satisfies a Novikov-type condition
- W : the Wiener process such that $B_t = \int_{\mathbb{R}_-} (-r)^{H-1/2} [dW_{t+r} - dW_r]$.
- \tilde{P} : the probability defined by $\left. \frac{d\tilde{P}}{dP} \right|_{[0, \tau]} = e^{-L}$, where

$$L = \frac{1}{c(H)} \int_0^\tau [D_+^{H-1/2}g]_u dW_u + \frac{1}{2c(H)^2} \int_0^\tau [D_+^{H-1/2}g]_u^2 du$$

- $c(H)$: a constant dependent of H .

Then Q is a fractional Brownian motion on $[0, \tau]$ under \tilde{P} .

A first step

Take $\theta_\tau = \theta + \tau^{-1/2}u$. Then by our Girsanov-type lemma:

$$\begin{aligned} & \log \left(\frac{dP_{\theta_\tau}^\tau}{dP_\theta^\tau} \right) \\ &= -\frac{1}{c(H)} \int_0^\tau \left\langle \sigma^{-1} \left([D_+^{H-1/2} b(Y; \theta)]_t - [D_+^{H-1/2} b(Y; \theta_\tau)]_t \right), dW_t \right\rangle \\ & \quad - \frac{1}{2c(H)^2} \int_0^\tau \left| \sigma^{-1} \left([D_+^{H-1/2} b(Y; \theta)]_t - [D_+^{H-1/2} b(Y; \theta_\tau)]_t \right) \right|^2 dt \\ & \approx I(\tau) + J(\tau), \end{aligned}$$

where

$$\begin{aligned} I(\tau) &= -\frac{1}{2c(H)^2\tau} \int_0^\tau \left| \sigma^{-1} [D_+^{H-1/2} \hat{b}(Y; \theta)]_t u \right|^2 dt, \\ J(\tau) &= \frac{1}{c(H)\tau^{1/2}} \int_0^\tau \left\langle \sigma^{-1} [D_+^{H-1/2} \hat{b}(Y; \theta)]_t u, dW_t \right\rangle \end{aligned}$$

and $\hat{b} = \partial_\theta b$.

Convergence of $I(\tau)$

Lemma

Denote

$$\Sigma(\theta) = \tilde{c}_H \int_{\mathbb{R}_+^2} \frac{\mathbb{E}_\theta \left[\left(\hat{b}(\bar{Y}_0; \theta) - \hat{b}(\bar{Y}_{r_1}; \theta) \right)^T (\sigma^{-1})^T \sigma^{-1} \left(\hat{b}(\bar{Y}_0; \theta) - \hat{b}(\bar{Y}_{r_1}; \theta) \right) \right]}{r_1^{1/2+H} r_2^{1/2+H}} dr_1 dr_2$$

Then $\Sigma(\theta)$ is integrable, and we have

$$I \rightarrow -\frac{1}{2} u^T \Sigma(\theta) u, \quad \tau \rightarrow \infty.$$

ingredients of proof: fractional calculus, ergodic theorem.

Convergence of $J(\tau)$

By (1) the CLT for Brownian martingales; and (2) the convergence of $I(\tau)$, we obtain:

Lemma

$$J(\tau) \xrightarrow{\mathcal{L}} u^T \mathcal{N}(0, \Sigma(\theta)), \quad \tau \rightarrow \infty.$$

Main result

Putting together the convergences of $I(\tau)$ and $J(\tau)$:

Theorem

Assume the ergodicity property hypothesis holds. Then for any $\theta \in \Theta$ and $u \in \mathbb{R}^q$ fixed, we have

$$\log \frac{dP_{\theta + \frac{u}{\sqrt{\tau}}}^\tau}{dP_\theta^\tau} \xrightarrow{\mathcal{L}} u^T \mathcal{N}(0, \Sigma(\theta)) - \frac{1}{2} u^T \Sigma(\theta) u, \quad \tau \rightarrow \infty.$$

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Ornstein-Uhlenbeck case

Consider

$$Y_t = -\theta \int_0^t Y_s ds + B_t, \quad t \in [0, \tau], \quad \theta > 0.$$

The stationary solution is given by

$$\bar{Y}_t = \int_{-\infty}^t e^{-\theta(t-s)} dB_s.$$

LAN property - Ornstein-Uhlenbeck process

Proposition

Let Y be the fractional Ornstein-Uhlenbeck process solution. Then the LAN property is satisfied with

$$\Sigma(\theta) = \frac{1}{2\theta}.$$

Remark: in this case $\Sigma(\theta)$ does not depend on the Hurst parameter H .

Efficiency of MLE

Facts: The MLE of a fractional Ornstein-Uhlenbeck process is:
(1) uniformly consistent; (2) uniformly asymptotically normal.

Proposition

The MLE is asymptotically minimax efficient for any loss function $\ell(u) = |u|^p$, $p > 0$.

Thank you very much for your attention!