

LAN property for sde's with additive fractional noise and continuous time observation

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The Ornstein-Uhlenbeck process

- $dX_t = -\theta X_t dt + dB_t$, $t \in [0, \tau]$, $\theta > 0$.
- B_t is a standard Brownian motion.
- Let $\hat{\theta}_\tau$ be the MLE of θ from the continuous observation of X in $[0, \tau]$.
- Then, it is well-known

$$\lim_{\tau \rightarrow \infty} \hat{\theta}_\tau = \theta \quad \text{a.s.}$$

and that

$$\mathcal{L}(\mathbf{P}_\theta) - \lim_{\tau \rightarrow \infty} \sqrt{\tau}(\hat{\theta}_\tau - \theta) = \mathcal{N}(0, 2\theta).$$

- where \mathbf{P}_θ is the probability law of the solution in the space $\mathcal{C}(\mathbf{R}_+; \mathbf{R})$.

The LAN property for the Ornstein-Uhlenbeck process

- The parametric statistical model $\{\mathbf{P}_\theta, \theta \in \Theta\}$ satisfies the LAN property at $\theta \in \Theta$ with rate $\sqrt{\tau}$ since for any $u \in \mathbf{R}$, as $\tau \rightarrow \infty$:

$$\log \left(\frac{d\mathbf{P}_{\theta + \frac{u}{\sqrt{\tau}}}}{d\mathbf{P}_\theta} \right) \xrightarrow{\mathcal{L}(\mathbf{P}_\theta)} u\mathcal{N} \left(0, \frac{1}{2\theta} \right) - \frac{u^2}{4\theta},$$

where \mathbf{P}_θ^τ is probability law of the solution in the space $\mathcal{C}([0, \tau]; \mathbf{R})$.

- The local log likelihood ratio is asymptotically normal, with a locally constant covariance matrix and a mean equal to minus one half the variance.

Consequence of the LAN property

- **Minimax Theorem** : Let $(\hat{\theta}_\tau)_{\tau \geq 0}$ be a family of estimators of the parameter θ . Then

$$\lim_{\delta \rightarrow 0} \liminf_{\tau \rightarrow \infty} \sup_{|\theta' - \theta| < \delta} \mathbf{E}_{\theta'} \left[\tau (\hat{\theta}_\tau - \theta')^2 \right] \geq 2\theta.$$

- In particular, the MLE is asymptotic minimax efficient.
- The LAN property is an important tool in order to quantify the identifiability of a system. Started by Le Cam'60. Parallel theory to Cramér-Rao bound.

LAN property for ergodic diffusions

- Consider a non-linear d -dimensional ergodic diffusion

$$dX_t = b(X_t; \theta)dt + \sigma(X_t)dB_t, \quad t \in [0, \tau], \quad \theta \in \Theta \subset \mathbf{R}^q.$$

- Under regularity, ellipticity, and ergodic assumptions, for any $\theta \in \Theta$ and $u \in \mathbf{R}^q$, as $\tau \rightarrow \infty$:

$$\log \left(\frac{d\mathbf{P}_{\theta + \frac{u}{\sqrt{\tau}}}}{d\mathbf{P}_{\theta}} \right) \xrightarrow{\mathcal{L}(\mathbf{P}_{\theta})} u^T \mathcal{N}(0, \Gamma(\theta)) - \frac{1}{2} u^T \Gamma(\theta) u,$$

where \bar{X} is the ergodic limit of X , and

$$\Gamma(\theta) = \mathbf{E}_{\theta}[\partial_{\theta} b(\bar{X}; \theta)^T \sigma^{-1}(\bar{X})^T \sigma^{-1}(\bar{X}) \partial_{\theta} b(\bar{X}; \theta)].$$

- Proof** : Girsanov's theorem, CLT for martingales and ergodicity.
- Consequence : Minimax theorem** :

$$\lim_{\delta \rightarrow 0} \liminf_{\tau \rightarrow \infty} \sup_{|\theta' - \theta| < \delta} \mathbf{E}_{\theta'} \left[\tau (\hat{\theta}_{\tau} - \theta')^2 \right] \geq \Gamma(\theta)^{-1}.$$

The fractional Ornstein-Uhlenbeck process

- $X_t = -\theta \int_0^t X_s ds + B_t$, $t \in [0, \tau]$, $\theta > 0$.
- B_t fractional Brownian motion with Hurst parameter $H > 1/2$.
- \mathbf{P}_θ is the probability law of the solution in the space $\mathcal{C}^\lambda(\mathbf{R}_+; \mathbf{R})$, for any $\lambda < H$.
- Let $\hat{\theta}_\tau$ be the MLE of θ from the continuous observation of X in $[0, \tau]$.
- Then, it is well-known

$$\lim_{\tau \rightarrow \infty} \hat{\theta}_\tau = \theta \quad \text{a.s.}$$

and that

$$\mathcal{L}(\mathbf{P}_\theta) - \lim_{\tau \rightarrow \infty} \sqrt{\tau}(\hat{\theta}_\tau - \theta) = \mathcal{N}(0, 2\theta).$$

- This suggests that the LAN property holds with the same rate $\sqrt{\tau}$.

Ergodic sde's with additive fractional noise

$$X_t = x_0 + \int_0^t b(X_s; \theta) ds + \sum_{j=1}^d \sigma_j B_t^j, \quad t \in [0, \tau].$$

- $\theta \in \Theta$, where Θ is compactly embedded in \mathbf{R}^q .
- ergodicity condition : $\langle b(x; \theta) - b(y; \theta), x - y \rangle \leq -\alpha |x - y|^2$.
- \hat{b} is the Jacobian matrix $\partial_\theta b$.
- assumptions : $\partial_x b, \partial_{xx} b, \partial_x \hat{b}, \partial_{xx} \hat{b}$ bounded, b, \hat{b} linear growth, \hat{b} Lipschitz in θ and x , and σ invertible.
- The solution converges for $t \rightarrow \infty$ a.s. to a unique stationary process $(\bar{X}_t, t \geq 0)$.
- \mathbf{P}_θ^τ is the probability laws of the solution in the spaces $\mathcal{C}^\lambda([0, \tau]; \mathbf{R}^d)$, for any $\lambda < H$.

The LAN property

Theorem : For any $\theta \in \Theta$ and $u \in \mathbf{R}^q$, as $\tau \rightarrow \infty$,

$$\log \left(\frac{d\mathbf{P}_{\theta + \frac{u}{\sqrt{\tau}}}}{d\mathbf{P}_{\theta}} \right) \xrightarrow{\mathcal{L}(\mathbf{P}_{\theta})} u^T \mathcal{N}(0, \Gamma(\theta)) - \frac{1}{2} u^T \Gamma(\theta) u,$$

where the matrix $\Gamma(\theta)$ is defined by

$$\Gamma(\theta) = \int_{\mathbf{R}_+^2} \frac{\mathbf{E}_{\theta}[(\hat{b}(\bar{X}_0; \theta) - \hat{b}(\bar{X}_{r_1}; \theta))^T (\sigma^{-1})^T \sigma^{-1} (\hat{b}(\bar{X}_0; \theta) - \hat{b}(\bar{X}_{r_2}; \theta))]}{r_1^{1/2+H} r_2^{1/2+H}} dr_1 dr_2.$$

Remark : The efficiency of the MLE in the fractional Ornstein-Uhlenbeck case remains open....

Steps of the proof of the LAN property

- Use the representation of the fBm given in Hairer'05 (introduced by Mandelbrot and Van Ness'68) which is suitable to get the desired ergodic properties.
- Apply Girsanov's theorem for the fBm following Moret and Nualart'02.
- Handle the singularities popping out the fractional derivatives in the Girsanov exponent.
- Get ergodic results in Hölder type norms for our process X .
- In order to apply a CLT for Brownian martingales, we use Malliavin calculus techniques : derive concentration properties for the Girsanov exponents by means of a Poincaré type inequality (Üstunel'95), which needs to conveniently upper bound some Malliavin derivatives.

Mandelbrot and Van Ness representation of fBm

Let W be a two sided Wiener process, then the following defines a two-sided fBm : for any $t \in \mathbf{R}$:

$$\begin{aligned} B_t &= c_H \int_{\mathbf{R}_-} (-r)^{H-1/2} [dW_{t+r} - dW_r] \\ &= c_H \left\{ \int_{-\infty}^0 [(-(r-t))^{H-1/2} - (-r)^{H-1/2}] dW_r - \int_0^t (-(r-t))^{H-1/2} dW_r \right\}. \end{aligned}$$

Abstract Wiener space : $(\mathcal{B}, \bar{\mathcal{H}}, \mathbf{P})$, where

$$\mathcal{B} = \left\{ f \in \mathcal{C}(\mathbf{R}; \mathbf{R}^d); \frac{|f_t|}{1+|t|} < \infty \right\},$$

\mathbf{P} is the law of our fBm, and h is an element of the Cameron-Martin space $\bar{\mathcal{H}}$ iff there exists an element X_h in the first chaos such that

$$h_t = \mathbf{E}[B_t X_h], \quad \text{and} \quad \|h\|_{\bar{\mathcal{H}}} = \|X_h\|_{L^2(\Omega)}.$$

Properties of the SDE

- **Proposition** : There exists a unique continuous pathwise solution on any arbitrary interval $[0, \tau]$ such that :
- The map $X : (x_0, B) \in \mathbf{R}^d \times \mathcal{C}([0, \tau]; \mathbf{R}^d) \rightarrow \mathcal{C}([0, \tau]; \mathbf{R}^d)$ is locally Lipschitz continuous.
- For any $\theta \in \Theta$, $p \geq 1$, and $s, t \geq 0$,

$$\mathbf{E} [|X_t|^p] \leq c_p, \quad \text{and} \quad \mathbf{E} [|\delta X_{st}|^p] \leq k_p |t - s|^{pH},$$

where δ denotes the increment.

- For all $\varepsilon \in (0, H)$ there exists a random variable $Z_\varepsilon \in \cap_{p \geq 1} L^p(\Omega)$ such that a.s.

$$|X_t| \leq Z_\varepsilon (1 + t)^{2\varepsilon}, \quad \text{and} \quad |\delta X_{st}| \leq Z_\varepsilon (1 + t)^{2\varepsilon} |t - s|^{H-\varepsilon},$$

uniformly for $0 \leq s \leq t$.

Ergodic properties of the SDE

- Garrido-Atienza, Kloeden and Neuenkirch'09 :
- Shift operators $\theta_t : \Omega \rightarrow \Omega : \theta_t \omega(\cdot) = \omega(\cdot + t) - \omega(t)$, $t \in \mathbb{R}$, $\omega \in \Omega$.
- The shifted process $(B_s(\theta_t \cdot))_{s \in \mathbb{R}}$ is still a d -dimensional fractional Brownian motion and for any integrable random variable $F : \Omega \rightarrow \mathbb{R}$

$$\lim_{\tau \rightarrow \infty} \frac{1}{\tau} \int_0^\tau F(\theta_t(\omega)) dt = \mathbf{E}[F],$$

for \mathbf{P} -almost all $\omega \in \Omega$.

- **Theorem** : There exists a random variable $\bar{X} : \Omega \rightarrow \mathbb{R}^d$ such that

$$\lim_{t \rightarrow \infty} |X_t(\omega) - \bar{X}(\theta_t \omega)| = 0$$

for \mathbf{P} -almost all $\omega \in \Omega$. Moreover, we have $\mathbf{E}[|\bar{X}|^p] < \infty$ for all $p \geq 1$.

Ergodic properties of the SDE

- **Theorem :** For any $\theta \in \Theta$ and any $f \in C^1(\mathbb{R}^d; \mathbb{R})$ such that

$$|f(x)| + |\partial_x f(x)| \leq c \left(1 + |x|^N\right), \quad x \in \mathbb{R}^d,$$

we have

$$\lim_{\tau \rightarrow \infty} \frac{1}{\tau} \int_0^\tau f(X_t) dt = \mathbf{E}[f(\bar{X})], \quad \mathbf{P}\text{-a.s.}$$

- **Proposition :** Let $\alpha \in (0, H)$. There exists a random variable Z admitting moments of any order such that for all $0 \leq s \leq t$

$$\left|X_t - \bar{X}_t\right| \leq Z e^{-cs} \quad \text{and} \quad \left|\delta \left[X - \bar{X}\right]_{st}\right| \leq Z e^{-cs} (t-s)^\alpha.$$

Operator that transforms W into B

Proposition : For $w \in C_c^\infty(\mathbf{R})$ and $H \in (0, 1)$, set

$$[K_H w]_t = c_H \int_{\mathbf{R}_-} (-r)^{H-1/2} [\dot{w}_{t+r} - \dot{w}_r] dr.$$

Then : (i) There exists a constant c_H such that

$$[K_H w]_t = \begin{cases} -c_H \left([I_+^{H-1/2} w]_t - [I_+^{H-1/2} w]_0 \right), & \text{for } H > \frac{1}{2} \\ -c_H \left([D_+^{1/2-H} w]_t - [D_+^{1/2-H} w]_0 \right), & \text{for } H < \frac{1}{2}, \end{cases}$$

where

$$[D_+^\alpha \varphi]_t = c_\alpha \int_{\mathbf{R}_+} \frac{\varphi_t - \varphi_{t-r}}{r^{1+\alpha}} dr, \quad \text{and} \quad [I_+^\alpha \varphi]_t = \tilde{c}_\alpha \int_{\mathbf{R}_+} \varphi_{t-r} r^{\alpha-1} dr.$$

(ii) For $H > 1/2$, K_H can be extended as an isometry from $L^2(\mathbf{R})$ to $I_+^{H-1/2}(L^2(\mathbf{R}))$.

(iii) There exists a constant c_H such that $K_H^{-1} = c_H K_{1-H}$.

Girsanov's transformation

Proposition : For a given $\theta \in \Theta$, consider the d -dimensional process

$$Q_t = \int_0^t \sigma^{-1} b(X_s; \theta) ds + B_t.$$

Then Q is a d -dimensional fractional Brownian motion under the probability $\tilde{\mathbf{P}}_\theta$ defined by $\frac{d\tilde{\mathbf{P}}_\theta}{d\mathbf{P}_\theta} \Big|_{[0, \tau]} = e^{-L}$, with

$$L = \int_0^\tau \langle \sigma^{-1} [D_+^{H-1/2} b(X; \theta)]_u, dW_u \rangle + \frac{1}{2} \int_0^\tau |\sigma^{-1} [D_+^{H-1/2} b(X; \theta)]_u|^2 du.$$

Proof : show that $D_+^{H-1/2} b(X; \theta)$ is well defined on $[0, \tau]$, and Novikov's condition : there exists $\lambda > 0$ such that

$$\sup_{t \in [0, \tau]} \mathbf{E}_\theta \left[\exp \left(\lambda \int_0^t |\sigma^{-1} [D_+^{H-1/2} b(X; \theta)]_s|^2 ds \right) \right] < \infty.$$

Proof of the LAN property

- **Step 1** : Apply Girsanov's theorem. Fix $\theta \in \Theta$, and set $\theta_\tau = \theta + \tau^{-1/2}u$. Then

$$\log \left(\frac{d\mathbf{P}_{\theta_\tau}^\tau}{d\mathbf{P}_\theta^\tau} \right) = - \int_0^\tau \langle \sigma^{-1}([D_+^{H-1/2}b(X; \theta_\tau)]_t - [D_+^{H-1/2}b(X; \theta)]_t), dW_t \rangle \\ - \frac{1}{2} \int_0^\tau |\sigma^{-1}([D_+^{H-1/2}b(X; \theta_\tau)]_t - [D_+^{H-1/2}b(X; \theta)]_t)|^2 dt.$$

- **Step 2** : Linearize this relation : add and subtract the d -dimensional vector

$$[D_+^{H-1/2}\hat{b}(X; \theta)]_t(\theta_\tau - \theta) = \frac{1}{\sqrt{\tau}} [D_+^{H-1/2}\hat{b}(X; \theta)]_t,$$

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

Step 2 : linearization

$$\log \left(\frac{d\mathbf{P}_{\theta_\tau}^\tau}{d\mathbf{P}_\theta^\tau} \right) = l_1 - l_2 - \frac{1}{2}l_3 - l_4,$$

where

$$l_1 = \frac{1}{\sqrt{\tau}} \int_0^\tau \langle \sigma^{-1} [D_+^{H-1/2} \hat{b}(X; \theta)]_t u, dW_t \rangle - \frac{1}{2\tau} \int_0^\tau |\sigma^{-1} [D_+^{H-1/2} \hat{b}(X; \theta)]_t u|^2 dt$$

$$l_2 = \int_0^\tau \langle \sigma^{-1} ([D_+^{H-1/2} b(X; \theta_\tau)]_t - [D_+^{H-1/2} b(X; \theta)]_t - [D_+^{H-1/2} \hat{b}(X; \theta)]_t (\theta_\tau - \theta)), dW_t \rangle$$

$$l_3 = \int_0^\tau |\sigma^{-1} ([D_+^{H-1/2} b(X; \theta_\tau)]_t - [D_+^{H-1/2} b(X; \theta)]_t - [D_+^{H-1/2} \hat{b}(X; \theta)]_t (\theta_\tau - \theta))|^2 dt$$

$$l_4 = \int_0^\tau \langle \sigma^{-1} ([D_+^{H-1/2} b(X; \theta_\tau)]_t - [D_+^{H-1/2} b(X; \theta)]_t - [D_+^{H-1/2} \hat{b}(X; \theta)]_t (\theta_\tau - \theta)), \sigma^{-1} [D_+^{H-1/2} \hat{b}(X; \theta)]_t (\theta_\tau - \theta) \rangle dt.$$

Remaining steps of the proof

- **Step 3** : Main contribution to our log-likelihood : we show that as $\tau \rightarrow \infty$

$$\frac{1}{\tau} \int_0^\tau |\sigma^{-1} [D_+^{H-1/2} \hat{b}(X; \theta)]_t u|^2 dt \xrightarrow{\mathbf{P}_\theta} u^T \Gamma(\theta) u.$$

- Together with multivariate central limit theorem for Brownian martingales implies that as $\tau \rightarrow \infty$

$$l_1 \xrightarrow{\mathcal{L}(\mathbf{P}_\theta)} u^T \mathcal{N}(0, \Gamma(\theta)) - \frac{1}{2} u^T \Gamma(\theta) u.$$

- **Step 4** : Negligible contributions : We show that the terms l_2 , l_3 and l_4 converge to zero in \mathbf{P}_θ -probability as $\tau \rightarrow \infty$.
- For l_3 apply Taylor's expansion and some computations, l_3 is the quadratic variation of the martingale l_2 , and by Cauchy-Schwarz inequality, l_4 is bounded by l_3 .

Proof of Step 3

Step 3a : We have that

$$\frac{1}{\tau} \int_0^\tau |\sigma^{-1} [D_+^{H-1/2} \hat{b}(X; \theta)]_t u|^2 dt \equiv \frac{1}{\tau} J_\tau(X) = \frac{1}{\tau} \int_0^\tau |\sigma^{-1} N_t(X)|^2 dt,$$

where

$$\begin{aligned} N_t(X) &= \int_{\mathbf{R}_+} \frac{(\hat{b}(X_t; \theta) - \hat{b}(X_{t-r}; \theta))u}{r^{H+1/2}} dr = N_{1,t}(X) + N_{2,t}(X) \\ &= \int_0^t \frac{(\hat{b}(X_t; \theta) - \hat{b}(X_{t-r}; \theta))u}{r^{H+1/2}} dr + \int_t^\infty \frac{(\hat{b}(X_t; \theta) - \hat{b}(x_0; \theta))u}{r^{H+1/2}} dr \end{aligned}$$

where we have set $X_t = x_0$ for all $t \leq 0$.

We denote by $J_\tau(\bar{X})$, $N_t(\bar{X})$, $N_{1,t}(\bar{X})$, $N_{2,t}(\bar{X})$ the same quantities with X replaced by \bar{X} .

Proof of Step 3

Step 3b : We show that

- $N_t(X) - N_t(\bar{X})$ is of order $t^{-\eta}Z$ with $\eta > 0$ and $Z \in \cap_{p \geq 1} L^p(\Omega)$.
- Hence $J_\tau(X) - J_\tau(\bar{X})$ is of order $\tau^{1-2\eta}$, which is a negligible term on the scale τ .
- This allows us to consider the limiting behavior of $J_\tau(\bar{X})$ instead of $J_\tau(X)$.

Proof of Step 3

Step 3c : This step is devoted to reduce our computations to an evaluation for the expected value.

- We show that

$$\lim_{\tau \rightarrow \infty} \frac{1}{\tau} \mathbf{E}_{\theta} [\|J_{\tau}(X) - \mathbf{E}_{\theta}[J_{\tau}(X)] \|] = 0,$$

- **Poincaré type inequality :** Let $F : \mathcal{B} \rightarrow \mathbf{R}$ be a functional in $\mathbb{D}^{1,2}$. Then,

$$\mathbf{E} [\|F - \mathbf{E}[F]\|] \leq \frac{\pi}{2} \mathbf{E} [\|DF\|_{\tilde{\mathcal{H}}}]$$

- This reduces to show that

$$\lim_{\tau \rightarrow \infty} \frac{\mathbf{E}_{\theta} [\|DJ_{\tau}(X)\|_{\tilde{\mathcal{H}}}]}{\tau} = 0.$$

Proof of Step 3

Proposition : For all $t > 0$, X_t belongs to $\mathbb{D}^{1,2}$, and the Malliavin derivative satisfies that

$$\|DX_t\|_{\tilde{\mathcal{H}}} \leq c \exp\left(-\frac{\alpha t}{2}\right),$$

uniformly in $t \in \mathbf{R}_+$. Moreover, for $0 \leq u \leq v$,

$$\|D(\delta X_{uv})\|_{\tilde{\mathcal{H}}} \leq c_1 \exp\left(-\frac{\alpha u}{2}\right) (v - u)^{H/2},$$

uniformly in u and v .

Idea of proof : Derive contraction properties of the map $h \rightarrow X^h$, $h \in \tilde{\mathcal{H}}$, where X^h is the solution to our SDE driven by $B + h$:

$$|X_t^h - X_t| \leq c \exp\left(-\frac{\alpha t}{2}\right) \|h\|_{\tilde{\mathcal{H}}},$$

uniformly in $t \in \mathbf{R}_+$.

Proof of Step 3

Step 3d : We are now reduced to the analysis of the quantity $\mathbf{E}_\theta[J_\tau(\bar{X})]$.

- This is equal to $u^T \Psi u$ where Ψ equals the matrix

$$\int_0^\tau dt \int_{\mathbf{R}_+^2} \frac{\mathbf{E}_\theta[(\hat{b}(\bar{Y}_t; \theta) - \hat{b}(\bar{Y}_{t-r_1}; \theta))^T (\sigma^{-1})^T \sigma^{-1} (\hat{b}(\bar{Y}_t; \theta) - \hat{b}(\bar{Y}_{t-r_2}; \theta))]}{r_1^{1/2+H} r_2^{1/2+H}} dr_1 dr_2$$

- By stationarity of \bar{Y} , the expected value does not depend on t and

$$|\mathbf{E}_\theta[(\hat{b}(\bar{Y}_0; \theta) - \hat{b}(\bar{Y}_{r_1}; \theta))^T (\sigma^{-1})^T \sigma^{-1} (\hat{b}(\bar{Y}_0; \theta) - \hat{b}(\bar{Y}_{r_2}; \theta))]| \lesssim (r_1^H \wedge 1) (r_2^H \wedge 1)$$

- We obtain that Ψ is a convergent integral and

$$\mathbf{E}_\theta[J_\tau(\bar{Y})] = \tau u^T \Gamma(\theta) u.$$

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