

ASYMPTOTIC PROPERTIES OF MAXIMUM LIKELIHOOD ESTIMATOR FOR PARTIALLY OBSERVED FRACTIONAL DIFFUSION SYSTEM

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Statistique Asymptotique des Processus Stochastiques VII

OUTLINE

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COMPLETE OBSERVATION PROBLEM

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System (1) has a uniquely defined solution process X which is Gaussian but neither Markovian nor a semimartingale for $H \neq \frac{1}{2}$.

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- ▶ $Y^T = X^T = (X_t, 0 \leq t \leq T)$ (complete observation problem) ;
- ▶ Y^T defined by $dY_t = \mu X_t dt + dV_t^H$, $Y_0 = 0$, $0 \leq t \leq T$.
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For a fixed value of the parameter ϑ , let \mathbf{P}_ϑ^T denote the probability measure, induced by (X^T, Y^T) on the function space $\mathcal{C}_{[0, T]} \times \mathcal{C}_{[0, T]}$ and let \mathcal{F}_t^Y be the natural filtration of Y , $\mathcal{F}_t^Y = \sigma(Y_s, 0 \leq s \leq t)$.

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Let $\mathcal{L}(\vartheta, Y^T)$ be the likelihood, *i.e.* the Radon-Nikodym derivative of \mathbf{P}_ϑ^T , restricted to \mathcal{F}_T^Y with respect to some reference measure on $\mathcal{C}_{[0, T]}$.

THEOREM (PARTIALLY OBSERVED PROBLEM)

The MLE $\hat{\vartheta}_T$ is uniformly on compacts $\mathbb{K} \subset \mathbb{R}_*^+$ consistent, uniformly asymptotically normal

$$\sqrt{T} \left(\hat{\vartheta}_T - \vartheta \right) \xrightarrow{\text{law}} \mathcal{N} \left(\mathbf{0}, \mathcal{I}(\vartheta)^{-1} \right)$$

where $\mathcal{I}(\vartheta)$ **does not depend on H** :

$$\mathcal{I}(\vartheta) = \frac{1}{2\vartheta} - \frac{2\vartheta}{\alpha(\alpha + \vartheta)} + \frac{\vartheta^2}{2\alpha^3}$$

and $\alpha = \sqrt{\mu^2 + \vartheta^2}$. We have the uniform on $\vartheta \in \mathbb{K}$ convergence of the moments: for any $p > 0$,

$$\lim_{T \rightarrow \infty} \mathbf{E}_{\vartheta} \left| \sqrt{T} \left(\hat{\vartheta}_T - \vartheta \right) \right|^p = \mathbf{E} \left| \mathcal{I}(\vartheta)^{-\frac{1}{2}} \zeta \right|^p \quad \zeta \sim \mathcal{N}(0, 1).$$

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In particular, defining for $0 < s < t$, $H > \frac{1}{2}$,

$$k_H(t, s) = \kappa_H^{-1} s^{\frac{1}{2}-H} (t-s)^{\frac{1}{2}-H}, \quad \kappa_H = 2H\Gamma\left(\frac{3}{2}-H\right)\Gamma\left(\frac{1}{2}+H\right),$$

$$M_t = \int_0^t k_H(t, s) dW_s^H,$$

then the process $M = (M_t, t \geq 0)$ is a Gaussian martingale, the *fundamental martingale*.

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then the process $M = (M_t, t \geq 0)$ is a Gaussian martingale, the *fundamental martingale*.

Moreover, the natural filtration of the martingale M coincides with the natural filtration of the fBm W^H .

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Let us introduce $Z = (Z_t, t \geq 0)$ the *fundamental semimartingale* associated to X , namely

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Then $\zeta = (\zeta_t, t \geq 0) = \left(\int_0^t s^{2H-1} dZ_s \right)$ is the solution of

$$d\zeta_t = -\vartheta \lambda \mathbf{A}(t) \zeta_t d\langle M \rangle_t + b(t) dM_t, \quad \zeta_0 = 0,$$

with

$$\mathbf{A}(t) = \begin{pmatrix} t^{2H-1} & 1 \\ t^{4H-2} & t^{2H-1} \end{pmatrix} \quad \text{and} \quad b(t) = \begin{pmatrix} 1 \\ t^{2H-1} \end{pmatrix}.$$

SKETCH OF THE PROOF

We have

$$\begin{aligned} \mathcal{L}(\vartheta, X^T) &= \frac{d\mathbf{P}_\vartheta}{d\mathbf{P}_0}(\zeta^T) \\ &= \exp\left(-\vartheta\lambda \int_0^T (\mathbf{A}\zeta_s)^* \mathbf{B}^+ d\zeta_s - \frac{\vartheta^2\lambda^2}{2} \int_0^T (\mathbf{A}\zeta_s)^* \mathbf{B}^+ \mathbf{A}\zeta_s d\langle M \rangle_s\right) \end{aligned}$$

where $\mathbf{B}^+ = b(b^*b)^{-2}b^*$ and, by derivating w.r.t. ϑ

$$\sqrt{T}(\hat{\vartheta}_T - \vartheta) = -\frac{\frac{1}{\sqrt{T}} \int_0^T \lambda l(s)^* \zeta_s dM_s}{\frac{1}{T} \int_0^T \lambda^2 \zeta_s^* l(s) l(s)^* \zeta_s d\langle M \rangle_s} \quad l(t) = \begin{pmatrix} t^{2H-1} \\ 1 \end{pmatrix}.$$

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and the explicit expression of the Laplace transform

$$L_T(a, \vartheta) = \mathbf{E}_\vartheta \exp \left\{ -a \lambda^2 \int_0^T \zeta_s^* I(s) I(s)^* \zeta_s d\langle M \rangle_s \right\}$$

that implies that

$$\lim_{T \rightarrow \infty} L_T \left(\frac{a}{T}, \vartheta \right) = \exp \left\{ -a \frac{1}{2\vartheta} \right\},$$

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we have the convergence of the following integrals:

$$\frac{1}{T} \int_0^T \lambda^2 \zeta_s^* I(s) I(s)^* \zeta_s d\langle M \rangle_s \longrightarrow \frac{1}{2\vartheta} \quad \text{a.s.},$$

$$\frac{1}{\sqrt{T}} \int_0^T \lambda I(s)^* \zeta_s dM_s \xrightarrow{\text{law}} \mathcal{N} \left(0, \frac{1}{2\vartheta} \right).$$

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It is governed by the dynamic

$$dZ_t^O = \mu \lambda l(t)^* \zeta_t d\langle N \rangle_t + dN_t, \quad Z_0^O = 0,$$

where, for recall, $\zeta = (\zeta_t, t \geq 0)$ is the solution of:

$$d\zeta_t = -\vartheta \lambda \mathbf{A}(t) \zeta_t d\langle M \rangle_t + b(t) dM_t, \quad \zeta_0 = 0,$$

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Classical Girsanov theorem and the general filtering theorem give the following likelihood **explicit** function

$$\mathcal{L}_T(\vartheta, Z^{O,T}) = \exp \left\{ \mu \lambda \int_0^T I^* \pi_t(\zeta) dZ_t^0 - \frac{\mu^2 \lambda^2}{2} \int_0^T \pi_t(\zeta) I I^* \pi_t(\zeta)^* d\langle N \rangle_t \right\}$$

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where the conditional expectation $\pi_t(\zeta) = \mathbf{E}_\vartheta(\zeta_t | \mathcal{F}_t^Y)$ satisfies:

$$d\pi_t(\zeta) = (-\vartheta\lambda \mathbf{A} - \mu^2 \lambda^2 \gamma_{\zeta, \zeta} l l^*) \pi_t(\zeta) d\langle N \rangle_t + \mu\lambda \gamma_{\zeta \zeta} l dZ_t^O, \quad \pi_0(\zeta) = 0.$$

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and the filtering error $\gamma_{\zeta,\zeta}(t) = \mathbf{E}_\vartheta(\zeta_t - \pi_t(\zeta))^* (\zeta_t - \pi_t(\zeta))$ is the solution of the Riccati equation: $\gamma_{\zeta,\zeta}(0) = 0$ and

$$d\gamma_{\zeta,\zeta}(t) = (-\vartheta\lambda (\mathbf{A}\gamma_{\zeta,\zeta} + \gamma_{\zeta,\zeta} \mathbf{A}^*) + b b^* - \mu^2 \lambda^2 \gamma_{\zeta,\zeta} l l^* \gamma_{\zeta,\zeta}) d\langle N \rangle_t.$$

TRANSFORMATION OF THE OBSERVATION MODEL

Conditional expectation dynamic can be rewritten in the equivalent form

$$d\pi_t(\zeta) = -\vartheta\lambda\mathbf{A}\pi_t(\zeta)d\langle N \rangle_t + \mu\lambda\gamma_{\zeta\zeta}ld\nu_t$$

where the innovation process $(\nu_t, t \geq 0)$ is defined by:

$$d\nu_t = dZ_t^O - \mu\lambda l(t)^* \pi_t(\zeta) d\langle N \rangle_t, \quad \nu_0 = 0.$$

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Then, the likelihood ratio $\mathcal{Z}_T(\vartheta_1, \vartheta_2, \mathbf{Z}^{O,T}) =$

$$\begin{aligned} &= \frac{\mathcal{L}_T(\vartheta_2, \zeta^{O,T})}{\mathcal{L}_T(\vartheta_1, \zeta^{O,T})} = \frac{d\mathbf{P}_{\vartheta_2}^T}{d\mathbf{P}_{\vartheta_1}^T} / \mathcal{F}_T^Y, \\ &= \exp \left\{ \mu\lambda \int_0^T l^* \delta_{\vartheta_1, \vartheta_2} d\nu_t^{\vartheta_1} - \frac{\mu^2\lambda^2}{2} \int_0^T \delta_{\vartheta_1, \vartheta_2}^* \Pi^* \delta_{\vartheta_1, \vartheta_2} d\langle N \rangle_t \right\} \end{aligned}$$

where $\delta_{\vartheta_1, \vartheta_2}(t)$ is the difference $\pi_t^{\vartheta_2}(\zeta) - \pi_t^{\vartheta_1}(\zeta)$.

IBRAGIMOV-KHASMINSKII PROGRAM

From Ibragimov-Khasminskii, it is sufficient to check the three following conditions: (here $Z_T(u, Z^{O,T}) = Z_T(\vartheta, \vartheta + \frac{u}{\sqrt{T}}, Z^{O,T})$)

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$$Z_T(u, Z^{O,T}) \xrightarrow{\text{law}} \underbrace{\exp \left\{ u \cdot \eta - \frac{u^2}{2} \mathcal{I}(\vartheta) \right\}}_{Z(u)} \quad \text{with} \quad \eta \sim \mathcal{N}(0, \mathcal{I}(\vartheta)) ,$$

(A.2) for some $C, \chi > 0$: for all u such that $\vartheta + \frac{u}{\sqrt{T}} \in \mathbf{K} \subset \mathbb{R}_*^+$,

$$\mathbf{E}_\vartheta \sqrt{Z_T(u, Z^{O,T})} \leq C \exp(-\chi u^2) ,$$

(A.3) there exists $C > 0$ such that, for all $|u_1| < R$ and $|u_2| < R$,

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Let $L_T(a, \vartheta_1, \vartheta_2)$ be the Laplace transform of the integral of the quadratic form of the difference $\delta_{\vartheta_1, \vartheta_2}(t) = \pi_t^{\vartheta_2}(\zeta) - \pi_t^{\vartheta_1}(\zeta)$:

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Let us introduce the following condition (L): There exists $a_0 < 0$ such that for all $a > a_0$, $\forall u_1, u_2 \in \mathbb{R}$,

$$\lim_{T \rightarrow \infty} L_T\left(a, \vartheta + \frac{u_1}{\sqrt{T}}, \vartheta + \frac{u_2}{\sqrt{T}}\right) = \exp \left(-a \frac{(u_2 - u_1)^2}{2} \mathcal{I}(\vartheta) \right),$$

and for all T , $L_T\left(a, \vartheta + \frac{u_1}{\sqrt{T}}, \vartheta + \frac{u_2}{\sqrt{T}}\right) \leq C \exp \left(-a \chi (u_1 - u_2)^2 \right)$.

Let $L_T(a, \vartheta_1, \vartheta_2)$ be the Laplace transform of the integral of the quadratic form of the difference $\delta_{\vartheta_1, \vartheta_2}(t) = \pi_t^{\vartheta_2}(\zeta) - \pi_t^{\vartheta_1}(\zeta)$:

$$L_T(a, \vartheta_1, \vartheta_2) = \mathbf{E}_{\vartheta_1} \exp \left\{ -a \frac{\mu^2 \lambda^2}{2} \int_0^T \delta_{\vartheta_1, \vartheta_2}^* \mathbb{I}^* \delta_{\vartheta_1, \vartheta_2} d\langle N \rangle_t \right\}.$$

Let us introduce the following condition (L): There exists $a_0 < 0$ such that for all $a > a_0$, $\forall u_1, u_2 \in \mathbb{R}$,

$$\lim_{T \rightarrow \infty} L_T(a, \vartheta + \frac{u_1}{\sqrt{T}}, \vartheta + \frac{u_2}{\sqrt{T}}) = \exp \left(-a \frac{(u_2 - u_1)^2}{2} \mathcal{I}(\vartheta) \right),$$

and for all T , $L_T(a, \vartheta + \frac{u_1}{\sqrt{T}}, \vartheta + \frac{u_2}{\sqrt{T}}) \leq C \exp \left(-a \chi (u_1 - u_2)^2 \right)$.

PROPOSITION

Suppose condition (L) is satisfied. Then properties (A.1.–A.3) hold.

PROOF OF PROPOSITION

Actually, (A.1) is a direct consequence of (L). Indeed, for $u_1 = 0$ and $u_2 = u$, we have:

$$\lim_{T \rightarrow \infty} L_T(a, \vartheta, \vartheta + \frac{u}{\sqrt{T}}) = \exp\left(-a \frac{u^2}{2} \mathcal{I}(\vartheta)\right).$$

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It gives the convergence of the following integrals:

$$\frac{\mu^2 \lambda^2}{2} \int_0^T \delta_{\vartheta, u, T}^* I^* \delta_{\vartheta, u, T} d\langle N \rangle_t \longrightarrow \frac{u^2}{2} \mathcal{I}(\vartheta) \quad \text{a.s.}$$

and

$$\mu \lambda \int_0^T I^* \delta_{\vartheta, u, T} d\nu_t \xrightarrow{\text{law}} \mathcal{N}(0, u^2 \mathcal{I}(\vartheta)),$$

which achieves the proof of (A.1).

PROOF OF PROPOSITION

The condition (A.2) holds thanks to: $\mathbf{E}_\vartheta \sqrt{\mathcal{Z}_T(u)} =$

$$\begin{aligned}
&= \mathbf{E}_\vartheta \exp \left(\frac{\mu\lambda}{2} \int_0^T I^* \delta_{\vartheta,u,T} d\nu_t^\vartheta - \frac{\mu^2\lambda^2}{4} \int_0^T \delta_{\vartheta,u,T}^* I I^* \delta_{\vartheta,u,T} d\langle N \rangle_t \right) \\
&= \mathbf{E}_\vartheta \exp \left(\frac{\mu\lambda}{2} \int_0^T I^* \delta_{\vartheta,u,T} d\nu_t^\vartheta - \frac{\mu^2\lambda^2 p}{8} \int_0^T \delta_{\vartheta,u,T}^* I I^* \delta_{\vartheta,u,T} d\langle N \rangle_t \right) \times \\
&\quad \times \exp \left(\frac{\mu^2\lambda^2}{4} \left(\frac{p}{2} - 1 \right) \int_0^T \delta_{\vartheta,u,T}^* I I^* \delta_{\vartheta,u,T} d\langle N \rangle_t \right) \\
&\stackrel{(a)}{\leq} \left(\mathbf{E}_\vartheta \exp \left(\frac{\mu\lambda p}{2} \int_0^T I^* \delta_{\vartheta,u,T} d\nu_t^\vartheta - \frac{\mu^2\lambda^2 p^2}{8} \int_0^T \delta_{\vartheta,u,T}^* I I^* \delta_{\vartheta,u,T} d\langle N \rangle_t \right) \right)^{\frac{1}{p}} \\
&\quad \times \left(\mathbf{E}_\vartheta \exp \left(\frac{\mu^2\lambda^2 q}{4} \left(\frac{p}{2} - 1 \right) \int_0^T \delta_{\vartheta,u,T}^* I I^* \delta_{\vartheta,u,T} d\langle N \rangle_t \right) \right)^{\frac{1}{q}}
\end{aligned}$$

PROOF OF PROPOSITION

To prove (A.3), let us note that

$$\begin{aligned} \mathbf{E}_{\vartheta} \left(\sqrt{\mathcal{Z}_T(u_1)} - \sqrt{\mathcal{Z}_T(u_2)} \right)^2 &= 2 \left(1 - \mathbf{E}_{\vartheta} \mathcal{Z}_T(u_1) \sqrt{\frac{\mathcal{Z}_T(u_2)}{\mathcal{Z}_T(u_1)}} \right) \\ &= 2 \left(1 - \mathbf{E}_{\vartheta_1} \sqrt{\mathcal{Z}_T(\vartheta_1, \vartheta_2)} \right) \\ &\leq 2 \left(1 - \exp \left(-\chi (u_2 - u_1)^2 \right) \right) \\ &\leq C |u_1 - u_2|^2. \end{aligned}$$

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SATISFYING CONDITION (L)

The computation of the Laplace transform is based on the Cameron-Martin formula.

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Let us recall that, for $\vartheta = \vartheta_1$, the optimal filter $\pi_t^{\vartheta_1}(\zeta)$ and the difference $\delta_{\vartheta_1, \vartheta_2}(t) = \pi_t^{\vartheta_2}(\zeta) - \pi_t^{\vartheta_1}(\zeta)$ are governed by:

$$d\tilde{\pi}_t = \mathcal{A}(t)\tilde{\pi}_t d\langle N \rangle_t + \mathcal{B}(t)d\nu_t^{\vartheta_1},$$

where $\tilde{\pi}_t = \begin{pmatrix} \pi_t^{\vartheta_1}(\zeta) \\ \delta_{\vartheta_1, \vartheta_2} \end{pmatrix}$, $\underline{\mathbf{A}}^{\vartheta_2}(t) = -\vartheta_2 \lambda \mathbf{A}(t) - \mu^2 \lambda^2 \gamma_{\zeta, \zeta}^{\vartheta_2} \mathbf{I}^*$,

$$\mathbf{D}_{\gamma}^{\vartheta_1, \vartheta_2} = \gamma_{\zeta \zeta}^{\vartheta_2}(t) - \gamma_{\zeta \zeta}^{\vartheta_1}(t),$$

$$\mathcal{A}(t) = \begin{pmatrix} -\vartheta_1 \lambda \mathbf{A} & \mathbf{0} \\ -(\vartheta_2 - \vartheta_1) \lambda \mathbf{A} & \underline{\mathbf{A}}^{\vartheta_2} \end{pmatrix} \quad \text{and} \quad \mathcal{B}(t) = \mu \lambda \begin{pmatrix} \gamma_{\zeta \zeta}^{\vartheta_1}(t) \\ \mathbf{D}_{\gamma}^{\vartheta_1, \vartheta_2} \end{pmatrix} l(t).$$

SATISFYING CONDITION (L)

Then, $L_T(a, \vartheta_1, \vartheta_2) =$

$$\begin{aligned}
 &= \mathbf{E}_{\vartheta_1} \exp \left\{ -a \frac{\mu^2 \lambda^2}{2} \int_0^T \delta_{\vartheta_1, \vartheta_2}^* \Pi^* \delta_{\vartheta_1, \vartheta_2} d\langle N \rangle_t \right\} \\
 &= \mathbf{E}_{\vartheta_1} \exp \left\{ -a \frac{\mu^2 \lambda^2}{2} \int_0^T \begin{pmatrix} \pi_t^{\vartheta_1}(\zeta) \\ \delta_{\vartheta_1, \vartheta_2} \end{pmatrix}^* \mathcal{M}(t) \begin{pmatrix} \pi_t^{\vartheta_1}(\zeta) \\ \delta_{\vartheta_1, \vartheta_2} \end{pmatrix} d\langle N \rangle_t \right\} \\
 &= \exp \left\{ -a \frac{\mu^2 \lambda^2}{2} \int_0^T \text{trace}(\mathcal{H}(t) \mathcal{M}(t)) d\langle N \rangle_t \right\},
 \end{aligned}$$

where $\mathcal{M}(t) = \begin{pmatrix} 0 & 0 \\ 0 & \Pi^* \end{pmatrix}$ and $\mathcal{H}(t)$ is the solution of Riccati differential equation: $\mathcal{H}(0) = 0$,

$$\frac{d\mathcal{H}(t)}{d\langle N \rangle_t} = \mathcal{A}(t)\mathcal{H}(t) + \mathcal{H}(t)\mathcal{A}(t)^* + \mathcal{B}(t)\mathcal{B}(t)^* - a\lambda^2\mu^2\mathcal{H}(t)\mathcal{M}(t)\mathcal{H}(t).$$

SATISFYING CONDITION (L)

It is known that solution $\mathcal{H}(t)$ can be written as $\mathcal{H}(t) = \Psi_1^{-1}(t)\Psi_2(t)$, where the pair of 4×4 matrices (Ψ_1, Ψ_2) satisfies the system of linear differential equations:

$$\begin{aligned}\frac{d\Psi_1(t)}{d\langle N \rangle_t} &= -\Psi_1(t)\mathcal{A}(t) + a\lambda^2\mu^2\Psi_2(t)\mathcal{M}(t) & \Psi_1(0) &= \mathcal{I}d \\ \frac{d\Psi_2(t)}{d\langle N \rangle_t} &= \Psi_1(t)\mathcal{B}(t)\mathcal{B}(t)^* + \Psi_2(t)\mathcal{A}^*(t) & \Psi_2(t) &= 0,\end{aligned}$$

and $\mathcal{I}d$ is the 4×4 identity matrix.

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and Id is the 4×4 identity matrix.

Now,

$$L_T(a, \vartheta_1, \vartheta_2) = \exp \left\{ -\frac{1}{2} \int_0^T \text{trace } \mathcal{A}(t) d\langle N \rangle_t \right\} (\det \Psi_1(T))^{-\frac{1}{2}}.$$

The explicit representation of $\gamma_{\zeta, \zeta}^{\vartheta}$ imply that we can replace $\gamma_{\zeta, \zeta}^{\vartheta}$ by $\Delta^{-1} \gamma_{\infty} \Delta^{-1}$ in the coefficients of Equation and therefore

$$\lim_{T \rightarrow \infty} L_T(\mathbf{a}, \vartheta_1, \vartheta_2) = \lim_{T \rightarrow \infty} \exp \left\{ -\frac{1}{2} \int_0^T \text{trace } \mathcal{A}_{\infty}(t) d\langle N \rangle_t \right\} (\det \Psi_{1, \infty}(T))^{-\frac{1}{2}}$$

where

$$\frac{d\Psi_{1, \infty}(t)}{d\langle N \rangle_t} = -\Psi_{1, \infty}(t) \mathcal{A}_{\infty}(t) + a \lambda^2 \mu^2 \Psi_{2, \infty}(t) \mathcal{M}(t) \quad \Psi_{1, \infty}(0) = \mathcal{I}d$$

$$\frac{d\Psi_{2, \infty}(t)}{d\langle N \rangle_t} = \Psi_{1, \infty}(t) \mathcal{B}_{\infty}(t) \mathcal{B}_{\infty}(t)^* + \Psi_{2, \infty}(t) \mathcal{A}_{\infty}^*(t) \quad \Psi_{2, \infty}(t) = 0,$$

with

$$\mathcal{A}_{\infty}(t) = \begin{pmatrix} -\vartheta_1 & 0 \\ -(\vartheta_2 - \vartheta_1) & -\alpha_2 \end{pmatrix} \otimes \lambda \mathbf{A}, \quad \mathcal{B}_{\infty} \mathcal{B}_{\infty}^* = \begin{pmatrix} g_1^2 & g_1 g_2 \\ g_1 g_2 & g_2^2 \end{pmatrix} \otimes \lambda \mathbf{A} \mathbf{J},$$

$\alpha_2 = \sqrt{\vartheta_2^2 + \mu^2}$, $g_1 = \frac{\mu}{\sqrt{\lambda(\alpha_1 + \vartheta_1)}}$, $g_2 = \frac{\mu}{\sqrt{\lambda(\alpha_2 + \vartheta_2)}} - \frac{\mu}{\sqrt{\lambda(\alpha_1 + \vartheta_1)}}$ and \otimes is the Kronecker product.

Linear system (1) can be rewritten as

$$\frac{d(\Psi_1(t), \Psi_2(t) \otimes \mathbf{J})}{d\langle N \rangle_t} = (\Psi_1(t), \Psi_2(t) \otimes \mathbf{J}) \cdot (\mathfrak{J} \otimes \lambda \mathbf{A}(t))$$

where

$$\mathfrak{J} = \begin{pmatrix} \vartheta_1 & 0 & g_1^2 & g_1 g_2 \\ (\vartheta_2 - \vartheta_1) & \alpha_2 & g_1 g_2 & g_2^2 \\ 0 & 0 & -\vartheta_1 & -(\vartheta_2 - \vartheta_1) \\ 0 & a\lambda\mu^2 & 0 & -\alpha_2 \end{pmatrix}.$$

Clearly, system (5) has an explicit solution:

$$(\Psi_1(t), \Psi_2(t) \otimes \mathbf{J}) = (\mathbf{Id}, 0) \cdot (\mathcal{P} \otimes \mathbf{Id}) \mathcal{G} (\mathcal{P}^{-1} \otimes \mathbf{Id})$$

where $\mathcal{G} = \text{diag}(\mathbf{G}_1, \mathbf{G}_2, \mathbf{G}_3, \mathbf{G}_4)$ and

$$\frac{d\mathbf{G}_i(t)}{d\langle N \rangle_t} = \lambda x_i \mathbf{G}_i \mathbf{A} \quad \mathbf{G}_i(0) = \mathbf{Id}, \quad i = 1 \dots 4,$$

with $(x_i)_{i=1\dots 4}$ the eigenvalues of matrix \mathbf{A} and \mathcal{P} the matrix of its eigenvectors.

For $\vartheta_1 = \vartheta + \frac{u_1}{\sqrt{T}}$ and $\vartheta_2 = \vartheta + \frac{u_2}{\sqrt{T}}$, we can compute the eigenvalues and we obtain:

$$x_1 = \vartheta_1 + C_1 \frac{(u_2 - u_1)^2}{T} + o\left(\frac{1}{T}\right),$$

$$x_2 = -\vartheta_1 + C_2 \frac{(u_2 - u_1)^2}{T} + o\left(\frac{1}{T}\right),$$

$$x_3 = \alpha_2 + C_3 \frac{(u_2 - u_1)^2}{T} + o\left(\frac{1}{T}\right),$$

$$x_4 = -\alpha_2 + C_4 \frac{(u_2 - u_1)^2}{T} + o\left(\frac{1}{T}\right),$$

where $C_1 = \frac{a\mu^2}{2\vartheta\alpha^2(\alpha+\vartheta)^2} (\alpha^2 + 2\alpha\vartheta)$ and $C_3 = \frac{a\mu^2}{2\alpha^3(\alpha+\vartheta)^2} (-\vartheta^2 - 2\alpha\vartheta)$.

It can be easily checked that

$$\begin{aligned} \det \Psi_{1,\infty}(T) &= \det(G_1 \cdot G_3) \left(1 + o\left(\frac{1}{T}\right)\right) \\ &= \exp((x_1 + x_3)T) \left(1 + o\left(\frac{1}{T}\right)\right). \end{aligned}$$

Finally,

$$\begin{aligned} \lim_{T \rightarrow \infty} L_T\left(\mathbf{a}, \vartheta + \frac{u_1}{\sqrt{T}}, \vartheta + \frac{u_2}{\sqrt{T}}\right) &= \lim_{T \rightarrow \infty} \exp\left\{-\frac{1}{2}(x_1 + x_3 - \alpha_2 - \vartheta_1)T\right\} \\ &= \lim_{T \rightarrow \infty} \exp\left\{-\frac{1}{2}(C_1 + C_3)(u_2 - u_1)^2\right\} \\ &= \exp\left(-a \frac{(u_2 - u_1)^2}{2} \mathcal{I}(\vartheta)\right). \end{aligned}$$

CASE $H < \frac{1}{2}$

Thanks to Jost, for $H < 1/2$, we have the link between fBm processes of indexes H and $1 - H$:

$$W_t^H = \left(\frac{2H}{\Gamma(2H)} \Gamma(3 - 2H) \right)^{\frac{1}{2}} \int_0^t (t - s)^{2H-1} dW_s^{1-H}.$$

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The observation model becomes:

$$\begin{cases} d\tilde{X}_t = -\vartheta \tilde{X}_t dt + dw_t^{1-H}, & \tilde{X}_0 = 0, \\ d\tilde{Y}_t = \mu \tilde{X}_t dt + dv_t^{1-H}, & \tilde{Y}_0 = 0, \end{cases}$$

with, for instance,

$$\tilde{X}_t = \left(\frac{2H}{\Gamma(2H)} \Gamma(3 - 2H) \right)^{\frac{1}{2}} \int_0^t (t-s)^{1-2H} dX_s.$$

DEPENDENT NOISES CASE

$$\begin{cases} dX_t &= -\vartheta X_t dt + dW_t, & X_0 = 0, \\ Y_t &= \mu X_t + V_t, \end{cases}$$

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In this case, we can show that,

$$(1 + \mu^2) \frac{d\gamma_{\zeta, \zeta}(t)}{d\langle N \rangle_t} = -\vartheta\lambda (\mathbf{A}\gamma_{\zeta, \zeta} + \gamma_{\zeta, \zeta} \mathbf{A}^*) + bb^* - (\mu\vartheta)^2 \lambda^2 \gamma_{\zeta, \zeta} \mathbf{I}^* \gamma_{\zeta, \zeta}.$$

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Thank you for your attention