

# On a Stationarity Theory for Stochastic Volterra Integral Equation with affine drift: Finite-Time, Functional Weak Asymptotics and Applications.

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Contributed Talk at the "5th Barcelona Summer School of Stochastic Analysis and Quantitative Finance 2025", Barcelona, Spain.

Tuesday 22<sup>nd</sup> July, 2025



The *short/rough* and *long-term memory* of Volterra Equations  
Centre de Recerca Matemàtica - Universitat de Barcelona.

# Stochastic Volterra Integral Equation (Convolutional kernel)

We are interested in the stochastic Volterra integral equation:

$$X_t = X_0\phi(t) + \int_0^t K(t-s)(\mu(s) - \lambda X_s)ds + \int_0^t K(t-s)\sigma(s, X_s)dW_s, \quad X_0 \perp\!\!\!\perp W. \quad (1)$$

where  $\mu : [0, T] \rightarrow \mathbb{R}$  is a bounded Borel function (hence having a well-defined finite Laplace transform on  $\mathbb{R}_+$ ).

- $\lambda > 0$ ,  $\sigma : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$  Lipschitz continuous function,  $\phi$  a **deterministic continuous function**.
- $(W_t)_{t \geq 0}$  is a standard Brownian motion, independent of  $X_0$ , both defined on a probability space  $(\Omega, \mathcal{A}, P)$ .
- Let  $\mathcal{F}_t \supset \mathcal{F}_{t, X_0, W}$  be a filtration satisfying the usual conditions.
- $K$  is a convolutional kernel, i.e. a kernel  $K : \{(s, t) \in \mathbb{R}_+^2 : 0 \leq s < t\} \rightarrow \mathbb{R}_+$  satisfying

$$\forall s, t \geq 0, s < t, \quad K(s, t) = K(0, t - s) \quad (2)$$

④ **Resolvent of a Convolution Kernel:** The  $\lambda$ -resolvent  $R_\lambda$  is defined as the unique solution – if it exists – to

$$\forall t \geq 0, \quad R_\lambda(t) + \lambda \int_0^t K(t-s)R_\lambda(s) ds = 1 \quad \implies \quad R_\lambda = \sum_{k \geq 0} (-1)^k \lambda^k (1 * K^{*k}) \quad (3)$$

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**Example (Focus on fractional kernels  $K(t) = K_\alpha(t) = \frac{t^{\alpha-1}}{\Gamma(\alpha)} \mathbf{1}_{\mathbb{R}}(t)$ )**

- $K_\alpha * K_{\alpha'} = K_{\alpha+\alpha'}$  implies  $t \geq 0$ ,

$$R_{\alpha,\lambda}(t) = \sum_{k \geq 0} (-1)^k \frac{\lambda^k t^{\alpha k}}{\Gamma(\alpha k + 1)} = E_\alpha(-\lambda t^\alpha), \quad \text{So that} \quad f_{\alpha,\lambda}(t) = -R'_{\alpha,\lambda}(t) = \lambda t^{\alpha-1} \sum_{k \geq 0} (-1)^k \lambda^k \frac{t^{\alpha k}}{\Gamma(\alpha(k+1))}.$$

- where  $E_\alpha$  denotes the standard **Mittag-Leffler function**

# Summary of Main Tools

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② **Wiener-Hopf Equation:** Let  $g : \mathbb{R}_+ \rightarrow \mathbb{R}$  be a Borel locally bounded function. Then, the **Wiener-Hopf equation**

$$\forall t \geq 0, \quad x(t) = g(t) - \lambda \int_0^t K(t-s)x(s) ds \quad (4)$$

has a **unique solution** given by

$$\forall t \geq 0, \quad x(t) = g(t) + \int_0^t R'_\lambda(t-s)g(s) ds = g(t) - \int_0^t f_\lambda(t-s)g(s) ds \quad \text{where} \quad f_\lambda = -R'_\lambda \quad (5)$$

- **Standing assumption on the kernel  $K$ :**

Assumption ( $\lambda$ -resolvent  $R_\lambda$  of the kernel for every  $\lambda > 0$ .)

$$(\mathcal{K}) \quad \left\{ \begin{array}{l} (i) \quad R_\lambda \text{ } \mathbb{R}^+ \text{-differentiable } \lim_{t \rightarrow +\infty} R_\lambda(t) = a \in [0, 1[, \quad R_\lambda(0) = 1, \\ (ii) \quad f_\lambda \in \mathcal{L}_{loc}^2(\mathbb{R}_+, \text{Leb}_1), \text{ Where } f_\lambda = -R'_\lambda \text{ for } t > 0, \\ (iii) \quad \phi \in \mathcal{L}_{\mathbb{R}_+}^1(\text{Leb}_1), \text{ a continuous function satisfying } \lim_{t \rightarrow \infty} (\phi(t) - (f_\lambda * \phi)(t)) = a \\ (iv) \quad \mu \text{ is a } C^1\text{-function such that } \|\mu\|_{sup} < \infty \text{ and } \lim_{t \rightarrow +\infty} \mu(t) = \mu_\infty \in \mathbb{R}, \end{array} \right. \quad (6)$$

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- **As a consequence of Wiener-Hopf and stochastic/ordinary Fubini's theorems, equation (1) reads:**

$$X_t = X_0(\phi(t) - (f_\lambda * \phi)(t)) + \frac{1}{\lambda} \int_0^t f_\lambda(t-s)\mu(s) ds + \frac{1}{\lambda} \int_0^t f_\lambda(t-s)\sigma(s, X_s) dW_s. \quad (7)$$

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**Looking for Stationarity !**

- Either in the **classical sense**, where the distribution of the process is invariant under time shifts?
- or in a **weaker sense** ?

By standard arguments (BDG inequality and  $R_\lambda(t) = 1 - \int_0^t f_\lambda(t-s) ds$ ), the mean  $\mathbb{E}[X_t]$  satisfies:

$$\forall t \geq 0, \quad (\phi(t) - 1)\mathbb{E}[X_0] + \int_0^t f_\lambda(t-s) \left( \frac{\mu(s)}{\lambda} - \phi(t)\mathbb{E}[X_0] \right) ds = 0. \quad (8)$$

**Proposition** ( $\mathbb{E}(X_t)$  is constant if and only if:)

- $\phi(t) = 1$  a.s.,  $\mu(t) = \mu_0 = \mu_\infty$  a.s.,  $\mathbb{E}[X_0] = \frac{\mu_0}{\lambda}$ .

*In this case, equation (7) reads:*

$$\forall t \geq 0, \quad X_t = \frac{\mu_0}{\lambda} + \left( X_0 - \frac{\mu_0}{\lambda} \right) R_\lambda(t) + \frac{1}{\lambda} \int_0^t f_\lambda(t-s) \sigma(s, X_s) dW_s. \quad (9)$$

- $\phi(t) = 1 - \frac{\int_0^t f_\lambda(s) \left( \frac{\mu(t-s)}{\mu_\infty} - 1 \right) ds}{\int_t^\infty f_\lambda(s) ds}$ ,  $\mathbb{E}[X_0] = \frac{\mu_\infty}{\lambda}$ .

*In this case:*

$$\forall t \geq 0, \quad X_t = X_0 - \left( X_0 - \frac{\mu_\infty}{\lambda} \right) \int_0^t f_\lambda(t-s) \frac{\mu(s)}{\mu_\infty} ds + \frac{1}{\lambda} \int_0^t f_\lambda(t-s) \sigma(s, X_s) dW_s. \quad (10)$$

## Stationarity of the Variance (time homogenous case):

- Assume  $\sigma(t, x) := \sigma(x)$
- If  $X_0 \in L^2(\mathbb{P})$ : a **necessary condition for stationarity** is  $(\mathbb{E}[X_t] = \frac{\mu_\infty}{\lambda})$  and

$$\forall t \geq 0, \quad \mathbb{E}[X_t] = \frac{\mu_\infty}{\lambda}, \quad \text{Var}(X_t) = v_0 \quad \text{and} \quad \mathbb{E}[\sigma^2(X_t)] = \bar{\sigma}^2, \quad \text{are constant.} \quad (11)$$

By standard arguments (Itô's isomorphism and Fubini's Theorem) and assuming constant mean, the variance  $\text{Var}[X_t]$  satisfies:

$$\forall t \geq 0, \quad \text{Var}(X_t) = \text{Var}(X_0)\phi^2(t)R_\lambda^2(t) + \frac{1}{\lambda^2}(f_\lambda^2 * \bar{\sigma}^2)(t)$$

**Theorem** (Time homogenous diffusion coefficient  $\sigma$ . Let  $\sigma(t, x) = \sigma(x)$  in equation (7))

Assume that  $X_0 \in L^2(\mathbb{P})$  with  $\mathbb{E}[X_0] = \frac{\mu_\infty}{\lambda}$ . Suppose the conditions (11) hold for all  $t \geq 0$ :

Then, only two situations can occur: either

- $\bar{\sigma}^2 > 0$  and the kernel  $K$  is constant so that  $(X_t)_{t \geq 0}$  is a (Markov) Brownian diffusion process.

or

- $\bar{\sigma}^2 = 0$  and  $\forall t \geq 0, \quad X_t = \frac{\mu_\infty}{\lambda} \quad \mathbb{P} - a.s$

True Volterra equations have no stationary regime !

Then, why not fake stationary regimes? ...

We investigate the case where:  $\sigma(t, x) = \underbrace{\varsigma(t)}_{\text{stabilizer?!}} \times \sigma(x)$ ,  $\varsigma(t), \sigma(x) > 0$ .

**Definition (Fake Stationary Regime of type I and II [2] G.Pagès 2024)**

- 1 The process  $(X_t)_{t \geq 0}$  starting from  $X_0 \in L^2(\mathbb{P})$ , exhibit a fake stationary regime of type I if:

$$\forall t \geq 0, \quad \mathbb{E} X_t = c^{ste}, \quad \text{Var}(X_t) = c^{ste} \geq 0 \quad \text{and} \quad \bar{\sigma}^2(t) := \mathbb{E} \sigma^2(X_t) = c^{ste} := \bar{\sigma}_0^2 \geq 0. \quad (12)$$

- 2 The process  $(X_t)_{t \geq 0}$  starting from  $X_0 \in \bigcap_{p > 0} L^p(\mathbb{P})$ , exhibit a fake stationary regime of type II if  $(X_t)_{t \geq 0}$  has the same marginal distribution, i.e.,  $X_t \stackrel{d}{=} X_0$  for every  $t \geq 0$ .

Then, why not fake stationary regimes? ...

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**Theorem (Time-Dependent Volatility  $\sigma$ . Let  $\sigma(t, x) = \zeta(t)\sigma(x)$  in equation (7))**

Assume that  $X_0 \in L^2(\mathbb{P})$  with  $\mathbb{E}[X_0] = \frac{\mu_\infty}{\lambda}$ . Suppose the following conditions hold for all  $t \geq 0$ :

$$\mathbb{E}[X_t] = \frac{\mu_\infty}{\lambda}, \quad \text{Var}(X_t) = \text{Var}(X_0) = v_0 \geq 0, \quad \text{and} \quad \mathbb{E}[\sigma^2(X_t)] =: \bar{\sigma}^2(t) = \bar{\sigma}_0^2 \geq 0. \quad (13)$$

Then, a necessary condition for (13) to hold is that  $(v_0, \bar{\sigma}_0^2, \zeta(t))$  must satisfy the following **Master equation** :

$$(E_{\lambda, c}): \quad \forall t \geq 0, \quad c\lambda^2(1 - \phi^2(t)R_\lambda^2(t)) = (f_\lambda^2 * \zeta^2)(t) \quad \text{where} \quad c = \frac{v_0}{\bar{\sigma}_0^2} \quad \text{so that} \quad \zeta = \zeta_{\lambda, c}. \quad (14)$$

## Examples of fake stationary regimes I and II:

Let's consider a **squared trinomial diffusion coefficient** given below:

$$\sigma(x) = \sqrt{\kappa_0 + \kappa_1(x-a) + \kappa_2(x-b)^2} \quad \text{with} \quad \kappa_i \geq 0, \quad i = 0, 2, \quad \kappa_1^2 \leq 4\kappa_2(\kappa_0 + \kappa_1(b-a)). \quad (15)$$

### Proposition (Fake stationary regimes (types I and II) and asymptotics: [3])

Consider a stabilized Volterra Equation with the diffusion coefficient  $\sigma$  given by (15) and let  $\lambda > 0$ ,  $\mu_0 \in \mathbb{R}$ , where  $\varsigma = \varsigma_{\lambda, c}$ , assumed to be the unique continuous solution to Equation  $(E_{\lambda, c})$  in (14) for some  $c \in (0, \frac{1}{[\sigma]_{Lip}^2})$ . Then, let  $X_0 \in L^2(P)$  with

$$\mathbb{E}[X_0] = \frac{\mu_\infty}{\lambda}, \quad \text{and} \quad \text{Var}(X_0) = v_0 = \frac{c\sigma^2(\frac{\mu_\infty}{\lambda})}{1 - c\kappa_2}:$$

① **Case  $\sigma(x) = \sigma$  is constant.** Here, the solution  $(X_t)_{t \geq 0}$  has a **fake stationary regime of type I** with mean  $\frac{\mu_\infty}{\lambda}$  and variance  $v_0$ .

• This represents a fake stationary regime of type II if  $X_0 \sim \nu^* := \mathcal{N}(\frac{\mu_\infty}{\lambda}, v_0)$ , so that  $X_t \stackrel{d}{=} X_0 \quad \forall t \geq 0$ .

② **Case where  $\sigma$  is not constant and not degenerated (i.e. with  $\sigma(\frac{\mu_\infty}{\lambda}) \neq 0$ ).** The solution  $(X_t)_{t \geq 0}$  to (7) has a **fake stationary regime of type I**, i.e., for all  $t \geq 0$ ,

$$\forall t \geq 0, \quad \mathbb{E}[X_t] = \frac{\mu_\infty}{\lambda}, \quad \text{Var}(X_t) = v_0 = \frac{c\sigma^2(\frac{\mu_\infty}{\lambda})}{1 - c\kappa_2}, \quad \text{and} \quad \mathbb{E}[\sigma^2(X_t)] = \bar{\sigma}_0^2 = \frac{\sigma^2(\frac{\mu_\infty}{\lambda})}{1 - c\kappa_2}.$$

Moreover, if  $a = 0$ , for any starting value  $X_0 \in L^2(P)$ , the process  $X$  mixes i.e.  $\mathbb{E}[X_t] \rightarrow \frac{\mu_\infty}{\lambda}$  and  $\text{Var}(X_t) \rightarrow v_0$  as  $t \rightarrow +\infty$  ( $L^2$ -confluence).

## Proposition ( $L^2$ -confluence/Contraction Properties)

Assume  $(\mathcal{K})$  is in force, let  $\sigma(t, x) := \varsigma(t)\sigma(x)$  where  $\varsigma = \varsigma_{\lambda, c}$  is a non-negative, continuous and bounded solution to  $E_{\lambda, c}$  for some fixed  $\lambda, c > 0$  and  $\sigma : \mathbb{R} \rightarrow \mathbb{R}$  is a (Locally) Lipschitz continuous function. Let  $X_0, X'_0 \in L^2(\mathbb{P})$ , we consider the solutions to Volterra equation (1) denoted  $(X_t)_{t \geq 0}$  and  $(X'_t)_{t \geq 0}$  starting from  $X_0$  and  $X'_0$  respectively.

Set  $\rho := c[\sigma]_{Lip}^2$ . Then, one has, for  $c \in (0, \frac{1}{[\sigma]_{Lip}^2})$  :

- (a) There exists a continuous non-negative function  $\varphi_{\infty}^{\lambda, c, K, \phi} =: \varphi_{\infty} : \mathbb{R}^+ \rightarrow [0, 1]$ , such that  $\varphi_{\infty}(0) = 1$ ,  $\lim_{t \rightarrow +\infty} \varphi_{\infty}(t) = \frac{a^2}{1 - \rho(1 - a^2)}$ , only depending on  $\lambda, c, \phi$ , and the kernel  $K$ , such that :

$$\forall t \geq 0, \mathbb{E} \left( \left| X_t - X'_t \right| \right)^2 \leq \varphi_{\infty}(t) \mathbb{E} \left( \left| X_0 - X'_0 \right| \right)^2.$$

- (b) This result can be written using the 2-Wasserstein distance between  $X$  and  $X'$  :

$$\forall t \geq 0, W_2([X'_t], [X_t]) \leq \varphi_{\infty}(t)^{1/2} W_2([X'_0], [X_0]),$$

- (c) In particular, whenever  $a = 0$ ,

- if  $X$  has a fake stationary regime of type I,  $\mathbb{E}X'_t \rightarrow \frac{\mu_{\infty}}{\lambda}$ ,  $\text{Var}(X'_t) \rightarrow v_0$  as  $t \rightarrow +\infty$ . (Finite-dimensional  $W_2$ -convergence.)
- In case if  $X$  has a fake stationary regime of type II, its marginal distribution is unique.

## Theorem (Long run theorem-a)

- Let  $\lambda, c > 0$ . Assume  $(\mathcal{K})$  is in force and  $f_\lambda$  satisfy

$$\exists \beta > 1 : \int_0^{+\infty} f_\lambda(t)^{2\beta} dt < +\infty \quad \text{for some } \beta > 1, \quad \exists \vartheta \in (0, 1] : \max_{i=1,2} \left[ \int_0^{+\infty} |f_\lambda(u + \bar{\delta}) - f_\lambda(u)|^i du \right]^{\frac{1}{i}} \leq C \bar{\delta}^\vartheta.$$

- Assume  $\sigma$  Lipschitz,  $(E_{\lambda, c})$  in force with a stabilizer  $\varsigma_{\lambda, c}$  and  $c[\sigma]_{\text{Lip}}^2 < 1$
- Moreover, for some  $\delta > 0$ , for any  $p > 0$  and  $T > 0$ ,  $t \rightarrow x_0(t) = X_0\phi(t)$  is absolutely continuous, and satisfy

$$\mathbb{E} \left( \sup_{t \in [0, T]} |x_0(t)|^p \right) < +\infty, \quad \mathbb{E} [ |x_0(t') - x_0(t)|^p ] \leq C_{T, p} \left( 1 + \mathbb{E} \left[ \sup_{t \in [0, T]} |x_0(t)|^p \right] \right) |t' - t|^{\delta p}.$$

- (a) The family of shifted processes  $(X_{t+u})_{u \geq 0}$  is  $\mathcal{C}$ -tight and uniformly square integrable as  $t \rightarrow +\infty$ . There exists a process  $X^\infty$  with a  $(\delta \wedge \vartheta \wedge \frac{\beta-1}{2\beta} - \frac{1}{p} - \eta)$ -Hölder pathwise continuous sample paths for sufficiently small  $\eta > 0$  such that

$$(X_{t+u})_{t \geq 0} \Rightarrow (X_t^\infty)_{t \geq 0} \quad \text{weakly in } C(\mathbb{R}_+; \mathbb{R}) \text{ as } u \rightarrow \infty.$$

## Theorem (Long run theorem b-c)

If  $(X_t)_{t \geq 0}$  of the SVIE (1) has a fake stationary regime of type I, starting from a random variable  $X_0 \in L^2(P)$  with mean  $\frac{\mu_\infty}{\lambda}$  and variance  $v_0$ ,

(b) **Functional weak long-run behavior.** For  $t_1, t_2 \geq 0$ ,  $t_1 \leq t_2$ ,

$$\text{Cov}(X_{t+t_1}, X_{t+t_2}) \xrightarrow{t \rightarrow +\infty} \bar{C}_{f_\lambda}(t_1, t_2) := a^2 v_0 + \frac{(1-a^2)v_0}{\int_0^{+\infty} f_\lambda^2(s) ds} \int_0^{+\infty} f_\lambda(t_2 - t_1 + u) f_\lambda(u) du. \quad (16)$$

Thus, under any limiting distribution  $P$ , on the canonical space  $\Omega_0 := \mathcal{C}(\mathbb{R}_+, \mathbb{R})$ , the canonical process

$Y_t(\omega) = \omega(t)$ ,  $\omega \in \Omega_0$  is a (weak)  $L^2$ -stationary process with mean  $\frac{\mu_\infty}{\lambda}$  and covariance function  $\bar{C}_{f_\lambda}(s, t)$ ,  $s, t \geq 0$ .

(c) **Stationary Gaussian Case.** If  $\sigma(x) = \sigma > 0$  (constant) and  $X_0$  has a Gaussian distribution, say  $X_0 \sim \mathcal{N}(\frac{\mu_\infty}{\lambda}, v_0)$ , then  $(X_t)_{t \geq 0}$  satisfies

$$X_{t+} \xrightarrow{(C)} \Xi(f_\lambda) \text{ as } t \rightarrow +\infty,$$

where  $\Xi(f_\lambda)$  is the stationary Gaussian process with mean  $\frac{\mu_\infty}{\lambda}$  and covariance function  $\bar{C}_{f_\lambda}(\cdot)$ .

Applications:  $\alpha$ -fractional kernels  $K(t) = K_\alpha(t) = \frac{t^{\alpha-1}}{\Gamma(\alpha)} \mathbf{1}_{\mathbb{R}}(t)$ :  $\alpha \in (\frac{1}{2}, \frac{3}{2}) \subset (0, 2)$ ,  $\lambda > 0$ .

Remark ( **Fractional integration kernel**:  $K(t) = K_\alpha(t) = \frac{t^{\alpha-1}}{\Gamma(\alpha)} \mathbf{1}_{\mathbb{R}}(t)$ ,  $\alpha > 0$ .)

The  $\lambda$ -resolvent  $R_{\alpha,\lambda}$  of the kernel  $K$  and its derivative  $f_{\alpha,\lambda} := -R'_{\alpha,\lambda}$  defined on  $(0, +\infty)$  satisfy the *scaling property*:  $R_{\alpha,1} = e_\alpha$ ,  $R_{\alpha,\lambda} = R_{\alpha,1}(\lambda^{1/\alpha} \cdot)$ ,  $f_{\alpha,\lambda} = -\lambda^{\frac{1}{\alpha}} e'_\alpha(\lambda^{\frac{1}{\alpha}} \cdot)$  with  $e_\alpha(t) := E_\alpha(-t^\alpha)$ ,  $t \geq 0$  (*alternate Mittag-Leffler function*)

The representation of  $e_\alpha(t)$ , is obtained by applying **the Cauchy residue theorem using the Hankel path  $H_\alpha(\varepsilon)$** .

$$e_\alpha(t) = F_\alpha(t) + G_\alpha(t), \quad t \geq 0, \quad (17)$$

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The  $\lambda$ -resolvent  $R_{\alpha,\lambda}$  of the kernel  $K$  and its derivative  $f_{\alpha,\lambda} := -R'_{\alpha,\lambda}$  defined on  $(0, +\infty)$  satisfy the scaling property:  $R_{\alpha,1} = e_\alpha$ ,  $R_{\alpha,\lambda} = R_{\alpha,1}(\lambda^{1/\alpha} \cdot)$ ,  $f_{\alpha,\lambda} = -\lambda^{\frac{1}{\alpha}} e'_\alpha(\lambda^{\frac{1}{\alpha}} \cdot)$  with  $e_\alpha(t) := E_\alpha(-t^\alpha)$ ,  $t \geq 0$  (**alternate Mittag-Leffler function**)

The representation of  $e_\alpha(t)$ , is obtained by applying the Cauchy residue theorem using the Hankel path  $H_\alpha(\varepsilon)$ .

$$e_\alpha(t) = F_\alpha(t) + G_\alpha(t), \quad t \geq 0, \quad (17)$$

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$$F_\alpha(t) := \frac{1}{2\pi i} \int_{H_\alpha(\varepsilon)} e^{zt} \frac{z^{\alpha-1}}{z^\alpha + 1} dz = \int_0^\infty e^{-tu} H_\alpha(u) du, \quad \text{with} \quad H_\alpha(u) = \frac{\sin(\alpha\pi)}{\pi} \frac{u^{\alpha-1}}{u^{2\alpha} + 2u^\alpha \cos(\pi\alpha) + 1}, u \in \mathbb{R}_+$$

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$$G_\alpha(t) := \sum_k e^{z_k t} \operatorname{Res} \left[ \frac{z^{\alpha-1}}{z^\alpha + 1} \right]_{z_k} = \frac{1}{\alpha} \sum_k e^{z_k t} = \frac{2}{\alpha} e^{t \cos(\frac{\pi}{\alpha})} \cos \left( t \sin \left( \frac{\pi}{\alpha} \right) \right), \quad \text{where} \quad z_k = \exp \left( i \frac{(2k+1)\pi}{\alpha} \right)$$

$$e_\alpha(t) = \int_0^{+\infty} e^{-tu} H_\alpha(u) du + \frac{2}{\alpha} e^{t \cos(\frac{\pi}{\alpha})} \cos \left( t \sin \left( \frac{\pi}{\alpha} \right) \right), \quad 0 < \alpha < 2.$$

Applications: Case of  $\alpha$ -fractional kernels  $K(t) = K_\alpha(t) = \frac{t^{\alpha-1}}{\Gamma(\alpha)} \mathbf{1}_{\mathbb{R}}(t)$ :  $\alpha \in (\frac{1}{2}, \frac{3}{2})$ ,  $\lambda > 0$ .

Proposition ( $\lambda > 0$  and  $\alpha \in (1, 2)$ ,  $K_\alpha * K_{\alpha'} = K_{\alpha+\alpha'}$ . [4])

(a) The  $\lambda$ -resolvent  $R_{\alpha,\lambda} = R_{\alpha,1}(\lambda^{1/\alpha} \cdot)$  is  $\mathcal{C}^\infty$  on  $\mathbb{R}_+$ ,  $R_{\alpha,\lambda}(0) = 1$ ,  $a = \lim_{t \rightarrow +\infty} R_{\alpha,\lambda}(t) = 0$ ,  $\lim_{t \rightarrow +\infty} f_{\alpha,\lambda}(t) = 0$

$$f_{\alpha,\lambda}(t) := R'_{\alpha,\lambda}(t) = \lambda^{\frac{1}{\alpha}} \int_0^{+\infty} e^{-\lambda^{\frac{1}{\alpha}} tu} u H_\alpha(u) du - \frac{2}{\alpha} e^{t\lambda^{\frac{1}{\alpha}} \cos(\frac{\pi}{\alpha})} \cos \left[ t\lambda^{\frac{1}{\alpha}} \sin \left( \frac{\pi}{\alpha} \right) - \frac{\pi}{\alpha} \right].$$

(b) For  $i \in \{1, 2\}$  and  $\alpha \in (\frac{1}{2}, \frac{3}{2})$ ,  $f_{\alpha,\lambda} \in \mathcal{L}^{2\beta}(\text{Leb}_1) \forall \beta > 0$  and for every  $\vartheta \in (0, \alpha - \frac{1}{i})$ , there exists a real constant  $C_{\vartheta,\lambda} > 0$  such that:  $\forall \delta > 0$ ,  $\left[ \int_0^{+\infty} (f_{\alpha,\lambda}(t+\delta) - f_{\alpha,\lambda}(t))^i \right]^{1/i} \leq C_{\vartheta,\lambda} \delta^\vartheta$ .

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Corollary (Let  $\sigma(t, x) = \varsigma(t)\sigma(x)$  with  $\varsigma = \varsigma_{\alpha,\lambda,c}$  [2] and [3])

Let  $\sigma$  be given by equation (15),  $X_0 \in L^2(\mathbb{P})$  such that  $\mathbb{E}[X_0] = \frac{\mu_\infty}{\lambda}$  and  $\text{Var}(X_0) = v_0 = \frac{c\sigma^2(\frac{\mu_\infty}{\lambda})}{1-c\kappa_2}$ . Then,

- 1 The solution  $(X_t)_{t \geq 0}$  to the SVIE (7) starting from  $X_0$  has a fake stationary regime of type I:
- 2 If  $c \in \left(0, \frac{1}{[\sigma]_{Lip}^2}\right)$ , for every  $X'_0 \in L^2(P)$ , the solution to (7) starting from  $X'_0$  satisfies  $\|X'_t - X_t\|_2 \rightarrow 0$  as  $t \rightarrow +\infty$ .
- 3 The family of shifted processes  $X_{t+\cdot}$ ,  $t \geq 0$ , is  $C$ -tight as  $t \rightarrow +\infty$ , and its (functional) limiting distributions are all  $L^2$ -stationary processes.

# Applications: Computing the Stabilizer $\varsigma_{\alpha,\lambda,c}$ for the $\alpha$ -fractional kernels $K_\alpha(t) = \frac{t^{\alpha-1}}{\Gamma(\alpha)} \mathbf{1}_{\mathbb{R}}(t)$ :

**Total Variation ( Tauberian theorem)** on Laplace transforms of the **Master Equation** suggests to search  $\varsigma^2(t)$  as an expansion of the form (*Exponential Power Series Ansatz*): Set  $a_k = \frac{1}{\Gamma(\alpha k + 1)}$ ,  $b_k = \frac{1}{\Gamma(\alpha(k+1))}$ ,  $k \geq 0$ .

$$\varsigma_{\alpha,\lambda,c}^2(t) = c\lambda^{2-\frac{1}{\alpha}} \varsigma_\alpha^2(\lambda^{\frac{1}{\alpha}} t) \quad \text{where} \quad \varsigma_\alpha^2(t) := 2 t^{1-\alpha} \sum_{k \geq 0} (-1)^k c_k t^{\alpha k}. \quad (18)$$

with

$$c_0 = \frac{\Gamma(\alpha)^2}{\Gamma(2\alpha - 1)\Gamma(2 - \alpha)} \frac{\mu(0)}{\mu_\infty}, \quad \text{and for every } k \geq 1, c_k \text{ is defined inductively by:}$$

$$c_k = \frac{\Gamma(\alpha)^2 B(\alpha(k+1), 2(1-\alpha))}{\Gamma(2(1-\alpha))\Gamma(2\alpha-1)} \left[ (a * b)_k - \alpha(k+1) \sum_{\ell=1}^k B(\alpha(\ell+2) - 1, \alpha(k-\ell-1) + 2) (b^{*2})_\ell c_{k-\ell} \right].$$

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## Proposition (Existence and Properties of the function $\varsigma_{\alpha,\lambda,c}^2$ for $\alpha \in (\frac{1}{2}, \frac{3}{2})$ )

The **convergence radius** of the fractional power series (18) that defines  $\zeta_{\alpha,\lambda,c}$  is **infinite** and  $\zeta_{\alpha,\lambda,c}$  is **positive** on  $(0, +\infty]$  so that  $\zeta_{\alpha,\lambda,c}$  is **well-defined**: The stabilizer  $\zeta_{\alpha,\lambda,c}^2$  exists as a non-negative function, such that:

$$\textcircled{1} \lim_{t \rightarrow 0} \zeta_{\alpha,\lambda,c} = \begin{cases} 0 & \text{if } \alpha \leq 1, \\ +\infty & \text{if } \alpha > 1 \text{ provided } \frac{\mu(0)}{\mu_\infty} > 0. \end{cases} \quad \text{and} \quad \lim_{t \rightarrow +\infty} \zeta_{\alpha,\lambda,c}(t) = \frac{\sqrt{c(1-a^2)}\lambda}{\|f_{\alpha,\lambda}\|_{L^2(\text{Leb}_1)}}.$$

## Example I: On The Stabilized Long-Memory Quadratic Volterra dynamic, $\alpha \in (1, \frac{3}{2}) \subset (1, 2)$

We consider a squared trinomial diffusion coefficient  $\sigma$  of the form (15) for a certain  $\lambda > 0$ , and let  $c$  such that  $c[\sigma]_{Lip}^2 < 1$ :

$$\sigma(x) = \sqrt{\kappa_0 + \kappa_1 \left(x - \frac{\mu_0}{\lambda}\right) + \kappa_2 \left(x - \frac{\mu_0}{\lambda}\right)^2} \quad \text{with} \quad \kappa_i \geq 0, \quad i = 0, 2, \quad \kappa_1^2 \leq 4\kappa_2\kappa_0, \quad \lambda > 0. \quad (19)$$

We introduce a semi-integrated Euler-Maruyama scheme on the time grid  $t_k = t_k^n = \frac{kT}{n}, k = 0, \dots, n$ :

$$\bar{X}_{t_k} = \frac{\mu_0}{\lambda} + \left(X_0 - \frac{\mu_0}{\lambda}\right) R_\lambda(t_k) + \sum_{\ell=1}^k \frac{1}{\lambda} \int_{t_{\ell-1}}^{t_\ell} f_\lambda(t_k - s) \varsigma_{H+\frac{1}{2}, \lambda, c}(t_\ell) \sigma(\bar{X}_{t_{\ell-1}}) dW_s$$

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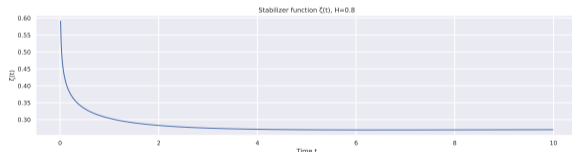


Figure: Graph of the stabilizer  $t \rightarrow \varsigma_{\alpha, \lambda, c}(t)$  over time interval  $[0, T]$ ,  $T = 10$  for a value of the Hurst exponent  $H = 0.8$ ,  $\lambda = 0.2$ ,  $c = 0.3$ .

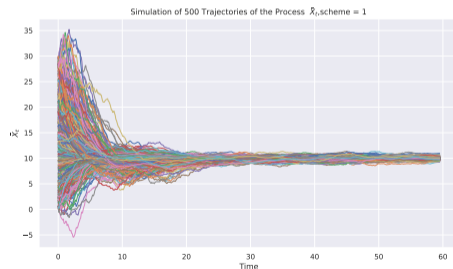


Figure: Confluence from a  $[0,30]$ -Uniform Distribution,  $T=60$ ,  $H = 0.8$ ,  $\lambda = 0.2$ ,  $c = 0.36$ .

## Example II: On The Stabilized Quadratic Rough Heston model

$$\frac{dS_t}{S_t} = \sigma(X_t)dW_t, \quad \sigma(x) = \sqrt{\kappa_0 + \kappa_1 \left(x - \frac{\mu_\infty}{\lambda}\right) + \kappa_2 \left(x - \frac{\mu_\infty}{\lambda}\right)^2} \quad \text{with} \quad \kappa_i \geq 0, \quad i = 1, 2, \quad \kappa_1^2 \leq 4\kappa_0\kappa_2.$$

where  $W$  is a BM, and  $\mu_\infty, \lambda$  are some positive constants, and  $X_t$  follows a quadratic Volterra stochastic process.

$$X_t = X_0 - \left(X_0 - \frac{\mu_\infty}{\lambda}\right) \int_0^t f_{\alpha, \lambda}(t-s) \frac{\mu(s)}{\mu_\infty} ds + \frac{1}{\lambda} \int_0^t f_{\alpha, \lambda}(t-s) \varsigma_{H+\frac{1}{2}, \lambda, c}(s) \sigma(X_s) dW_s, \quad \varsigma_{\alpha, \lambda, c}(t), \sigma(x) \geq 0. \quad (20)$$

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- **The Stabilized Quadratic Rough Heston volatility dynamic** ( $V_t = \sigma^2(X_t)$  and  $\alpha \in (\frac{1}{2}, 1)$ ): joint calibration of the S&P 500 and VIX smile, accounting for the so-called Zumbach effect. [2] and [3]

Curves  $\text{Var}(X_t)$  and  $\mathbb{E}[\sigma^2(X_t)]$  as function of time,  $\sigma(x) = \sqrt{\kappa_0 + \kappa_1(x - \frac{\mu_0}{\lambda}) + \kappa_2(x - \frac{\mu_0}{\lambda})^2}$ ,  $H=0.1$ ,  $c = 0.316$ ,  $1.5e+05$  Samples, scheme = 1

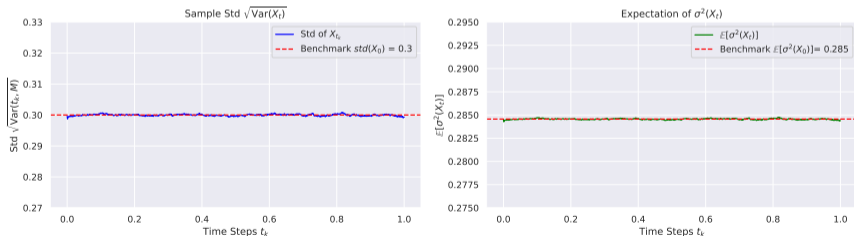


Figure: Graph of  $t_k \mapsto \text{StdDev}(t_k, M)$  and  $t_k \mapsto \mathbb{E}[\sigma^2(X_{t_k}, M)]$  over the time interval  $[0, T]$ ,  $T = 1$ ,  $H = 0.1$ ,  $\mu_0 = 2$ ,  $\lambda = 0.2$ ,  $v_0 = 0.09$  and  $\text{StdDev}(X_0) = 0.3$ . Number of steps:  $n = 800$ , Simulation size:  $M = 150000$ .

- ① J. Gatheral, P. Jusselin and M. Rosenbaum. The quadratic rough Heston model and the joint S&P 500/VIX smile calibration problem
- ② G. Pagès. Volterra equations with affine drift: looking for stationarity. Application to quadratic rough Heston model.
- ③ E. Gnabeyeu and G. Pagès. On a Stationarity Theory for FSVIEs : Finite-Time and Asymptotic Analysis, working paper.
- ④ R. Gorenflo and F. Mainardi. Fractional calculus: Integral and differential equations of fractional order.

Thanks For Your Attention!

Questions ?