

Short dated option pricing under rough volatility

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Based on joint works with
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Implied volatility

- Asset price process: $(S_t = e^{X_t})_{t \geq 0}$, with $X_0 = 0$.
- Black-Scholes-Merton (BSM) framework:

$$C_{\text{BS}}(\tau, k, \sigma) := \mathbb{E}_0 \left(e^{X_\tau} - e^k \right)_+ = \mathcal{N}(d_+) - e^k \mathcal{N}(d_-),$$

$$d_\pm := -\frac{k}{\sigma\sqrt{\tau}} \pm \frac{1}{2}\sigma\sqrt{\tau}.$$

- Spot implied volatility $\sigma_\tau(k)$: the unique (non-negative) solution to

$$C_{\text{observed}}(\tau, k) = C_{\text{BS}}(\tau, k, \sigma_\tau(k)).$$

- Implied volatility: unit-free measure of option prices.

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$$C_{\text{observed}}(\tau, k) = C_{\text{BS}}(\tau, k, \sigma_\tau(k)).$$

- Implied volatility: unit-free measure of option prices.

Implied volatility is not available in closed form generally.
Its asymptotic behaviour is available via (small/large k, τ) approximations.

Literature

Implied volatility ($\sigma_\tau(k)$) asymptotics as $|k| \uparrow \infty$, $\tau \downarrow 0$ or $\tau \uparrow \infty$:

- Hagan-Kumar-Lesniewski-Woodward (2003/2015), Obłój (2008): small-maturity for the SABR model.
- Berestycki-Busca-Florent (2004): small- τ using PDE methods for diffusions.
- Henry-Labordère (2009): small- τ asymptotics using differential geometry.
- Forde et al.(2012), Jacquier et al.(2012): small/ large τ using large deviations.
- Lee (2003), Benaim-Friz (2009), Gulisashvili (2010-2012), Caravenna-Corbetta (2016), De Marco-Jacquier-Hillairet (2013): $|k| \uparrow \infty$.
- Laurence-Gatheral-Hsu-Ouyang-Wang (2012): small- τ in local volatility models.
- Mijatović-Tankov (2012): small- τ for jump models.
- Bompis-Gobet (2015): asymptotic expansions in the presence of both local and stochastic volatility using Malliavin calculus.
- Fouque et al.(2000-2011): perturbation techniques for slow and fast mean-reverting stochastic volatility models.

Related works:

- Deuschel-Friz-Jacquier-Violante (CPAM 2014), De Marco-Friz (2014): small-noise expansions using Laplace method on Wiener space (Ben Arous-Bismut approach).
- Baudoin-Ouyang (2015): small-noise expansions in a (fully) fractional setting
- Gatheral-Jaisson-Rosenbaum (2014), and Bayer-Gatheral-Friz (2015),
- Forde-Zhang (2015): large deviations in a fractional stochastic volatility setting
- Fukasawa (2011,2015), Alós-León-Vives (2007) small-time (fractional) skew
- Guennoun-Jacquier-Roome (2015), El Euch-Rosenbaum (2016) fractional Heston.

"Classical" case: ($H_i = \frac{1}{2}$) case: Deuschel-Friz-Jacquier-Violante (2011)

$$d\mathbf{X}_t^\varepsilon = b(\varepsilon, \mathbf{X}_t^\varepsilon)dt + \varepsilon \sum_{i=1}^m \sigma_i(\mathbf{X}_t^\varepsilon) dW_t^i, \quad \mathbf{X}_0^\varepsilon = \mathbf{x}_0^\varepsilon \in \mathbb{R}^d$$

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Fractional case: ($H_i = H \in (\frac{1}{4}, 1)$) case: Baudoin-Ouyang (2015)

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Our main interest: $H_1 = \frac{1}{2}, H_2 \neq \frac{1}{2}$.

Rough volatility models

- Short-term data suggests a time decay of the ATM skew proportional to $\tau^{-\theta}$, with $\theta \in (0, 1/2)$ while classical stochastic volatility models generate a constant short-maturity skew.
- **Gatheral-Jaisson-Rosenbaum** and **Bayer-Gatheral-Friz** (2014,1015) proposed a fractional volatility model:

$$\begin{aligned}dS_t &= S_t(\sigma_t dZ_t + \mu_t dt), \\ \sigma_t &= \exp(Y_t),\end{aligned}\tag{1}$$

where

$$dY_t = \mu dW_t^H - b(Y_t - m)dt,$$

for $\mu, b > 0$, $m \in \mathbb{R}$ for a Bm Z and a fBm motion W^H with Hurst parameter H .

- Time series of the Oxford-Man SPX realised variance as well as implied volatility smiles of the SPX suggest that $H \in (0, 1/2)$: **short-memory** volatility.
- Main drawback: loss of Markovianity ($H \neq 1/2$) rules out PDE techniques, and Monte Carlo is computationally intensive. One way out is an efficient "Hybrid scheme" of Bennedsen, Lunde and Pakkanen (2015).

Rough volatility models: our setting

$$\begin{aligned}dX_t &= b_1(Y_t)dt + \sigma_1(Y_t)dW_t \\dY_t &= b_2(Y_t)dt + \sigma_2 dW_t^H.\end{aligned}\tag{2}$$

$\sigma_2 > 0$ (extendible to bounded, elliptic) and $H \in (0, 1)$, particular interest in $H < 1/2$.

Forde-Zhang '16: $b_1 \equiv b_2 \equiv 0$, and $\sigma_1 \in C^\alpha$, $\alpha \in (0, 1]$ (second part of the talk);

Fractional Stein-Stein: $b_1(y) \equiv -y^2/2$, $\sigma_1(y) \equiv y$, $b_2(y) \equiv a - by$ (first part).

Gatheral-Jaisson-Rosenbaum '14: $b_1(y) \equiv -e^{2y}/2$, $\sigma_1(y) \equiv e^y$, $b_2(y) \equiv a - by$.

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To introduce correlation we consider \tilde{B} and B independent, and set $W = \tilde{\rho}\tilde{B} + \rho B$ and

$$W_t^H = \int_0^t K(t, s)dB_s$$

where K the Volterra kernel of the (standard) fBm W^H .

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An intuitive remark: Mandelbrot-van Ness representation for fBm:

At time zero, the volatility process in (2) has already accumulated some randomness.

⇒ Hot-start of the process Y (with A. Jacquier and C. Lacombe).

1 Introduction

2 Approach via density asymptotics

Varadhan-type asymptotics for fractional SDEs

Rescalings and density asymptotics for fractional models

Corollaries: Short-time/tail expansion in fractional models

Implied volatility asymptotics

Idea of the proof

3 Bypassing density asymptotics

Refined expansions and moderate regimes

A non-Markovian extension of Osajima's energy expansion

Implied volatility asymptotics

Varadhan-type asymptotics

Recall the Black-Scholes density expansion: heat-kernel asymptotics

$$f_{\text{BS}}(t, x) \sim t^{-1/2} \exp\left(-\frac{1}{2t} \left(\frac{x}{\sigma}\right)^2\right), \quad \text{as } t \rightarrow 0.$$

In the **'homogenous' (fractional) case**: $H_1 = H_2 = \dots = H_d = H$ (Baudoin-Ouyang)

$$d\mathbf{X}_t = b(\mathbf{X}_t)dt + \sum_{i=1}^m \sigma_i(\mathbf{X}_t) d(W^H)^i_t, \quad \mathbf{X}_0 = \mathbf{x}_0 \in \mathbb{R}^d \quad (3)$$

(Extended) Varadhan formula

$$f_{\mathbf{X}}(t, x) \sim \text{cst } t^{-H} \exp\left(-\frac{d^2(x_0, x)}{2t^{2H}}\right), \quad \text{as } t \rightarrow 0.$$

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What if Hurst parameters are different $H_1 \neq H_2$?

Rescalings

Key to asymptotic expansions for $H_1 \neq H_2$: Rescalings

Recall the considered processes

$$dX_t = b_1(Y_t)dt + \sigma_1(Y_t)dW_t$$

$$dY_t = b_2(Y_t)dt + \sigma_2 dW_t^H.$$

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$$dY_t = b_2(Y_t)dt + \sigma_2 dW_t^H.$$

Define appropriate rescalings

$$dX_t^\varepsilon = b_1(\varepsilon^{\kappa_1}, Y_t^\varepsilon)dt + \varepsilon^\beta \sigma_1(Y_t^\varepsilon)dW_t$$

$$dY_t^\varepsilon = b_2(\varepsilon^{\kappa_2}, Y_t^\varepsilon)dt + \varepsilon^\beta \sigma_2 dW_t^H.$$

Rescalings

$$dX_t^\varepsilon = b_1(\varepsilon^{\kappa_1}, Y_t^\varepsilon)dt + \varepsilon^\beta \sigma_1(Y_t^\varepsilon)dW_t \quad dY_t^\varepsilon = b_2(\varepsilon^{\kappa_2}, Y_t^\varepsilon)dt + \varepsilon^\beta \sigma_2 dW_t^H. \quad (4)$$

Fractional Stein-Stein: Consider $(X_0, Y_0) = (0, y_0)$ and

$$dX_t = -\frac{Y_t^2}{2}dt + Y_t dW_t, \quad dY_t = (a - bY_t)dt + cdW_t^H. \quad (5)$$

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- **Rescaling 1 (short-time):** $(X_t^\varepsilon, Y_t^\varepsilon) := (\varepsilon^{2H-1}X_{\varepsilon^{2t}}, Y_{\varepsilon^{2t}}) \Rightarrow (4)$ with $\kappa_1 = 2H + 1$, $\kappa_2 = 2$ and $\beta = 2H$, $(x_0^\varepsilon, y_0^\varepsilon) = (0, y_0)$:

$$dX_t^\varepsilon = -\varepsilon^{2H+1} \frac{(Y_t^\varepsilon)^2}{2} dt + \varepsilon^{2H} Y_t^\varepsilon dW_t, \quad dY_t^\varepsilon = \varepsilon^2 (a - bY_t^\varepsilon) dt + \varepsilon^{2H} cdW_t^H. \quad (6)$$

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- **Rescaling 2 (tails):** $(X_t^\varepsilon, Y_t^\varepsilon) := (\varepsilon^{2H}X_t, \varepsilon^H Y_t) \Rightarrow (4)$ with $\kappa_1 = 0$, $\kappa_2 = \beta = H$, $(x_0^\varepsilon, y_0^\varepsilon) = (0, \varepsilon^H y_0)$:

$$dX_t^\varepsilon = -\frac{(Y_t^\varepsilon)^2}{2} dt + \varepsilon^H Y_t^\varepsilon dW_t, \quad dY_t^\varepsilon = (a\varepsilon^H + bY_t^\varepsilon)dt + \varepsilon^H cdW_t^H. \quad (7)$$

Theorem (Harms-H-Jacquier)

Consider an SDE of the form (4). Then the density of X_T^ε admits an expansion

$$f_\varepsilon(T, x) = \exp\left(-\frac{\Lambda(x)}{\varepsilon^{2\beta}} + \frac{\Lambda'(x)\widehat{X}_T}{\varepsilon^\beta}\right) \varepsilon^{-\min(\kappa_1, \beta)} \left(c_0 + \mathcal{O}(\varepsilon^{\delta(\kappa_1, \beta)})\right), \quad \text{as } \varepsilon \rightarrow 0,$$

where

$$\Lambda(x) = \inf\left\{\frac{1}{2}\|k\|_{\mathcal{H}_H}^2, k \in \mathcal{K}_{x_0^0, y_0^0}^x\right\} = \frac{1}{2}\|k_0\|_{\mathcal{H}_H}^2,$$

and

$$d\widehat{X}_t = \left[\partial_x b_1(0, \phi_t^{k_0}) + \partial_x \sigma_1(\phi_t^{h_0}) \cdot \dot{k}_0(t)\right] \widehat{X}_t dt + \partial_{\varepsilon\beta} b_1(0, \phi_t^{k_0}) dt, \quad \widehat{X}_0 = \partial_{\varepsilon\beta} x_0^\varepsilon|_{\varepsilon=0},$$

where ϕ^{k_0} denotes the ODE solution of the same SDE (4) replacing $\varepsilon^\beta dW$ by \dot{k}_0 and x_0^ε by x_0^0 .

Notations

- \mathcal{H} : absolutely continuous paths $[0, T] \rightarrow \mathbb{R}^2$ starting at 0 such that $\|\dot{h}\|_{\mathcal{H}}^2 < \infty$.
- $\mathcal{H}_H := K_H \mathcal{H}$ and $k := K_H h$, where K_H denotes the Volterra kernel.
- For fixed $(x_0, y_0) \in \mathbb{R}^2$, ϕ^k is the (unique) ODE solution to

$$\dot{\phi}_t^k = b_1(0, \phi_t^k) dt + \sigma_1(\phi_t^h) dk_t^1 + \sigma_2(\phi_t^h) dk_t^2, \quad \phi_0^k = (x_0^0, y_0^0).$$

- Denote $\psi^k := \Pi_1 \phi^k$ its projection on to the first coordinate X .
- $\mathcal{K}_a := \{k \in \mathcal{H}_H : \psi_T^k = a \in \mathbb{R}\} \neq \emptyset$ ("by Hörmander condition").
- $\Lambda(a) := \inf \left\{ \frac{1}{2} \|k\|_H^2 : k \in \mathcal{K}^a \right\}$.

Corollary: Varadhan-type asymptotics

Corollary (short-time asymptotics in Stein-Stein) $dY_t = (a - bY_t)dt + cdW_t^H$

In the fractional Stein-Stein model (X_t, Y_t) with $X_0 = 0$, $Y_0 = y_0 > 0$ the density of X_t satisfies in a neighbourhood of (x_0, y_0) the following asymptotic expansion as $t \rightarrow 0$

$$f_X(t, x) = \exp\left(-\frac{\Lambda(x)}{t^{2H}}\right) t^{-H} \left(\frac{1}{2\pi} + \mathcal{O}(t^{\delta(H, H+1/2)})\right)$$

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Proof: Take $T = 1$, $\varepsilon^2 = t$ and consider $(X_t^\varepsilon, Y_t^\varepsilon) := (\varepsilon^{2H-1}X_{\varepsilon^2 t}, Y_{\varepsilon^2 t})$ with $X_0^\varepsilon = 0$, $Y_0^\varepsilon = y_0 > 0$. \Rightarrow Short-time scaling:

$$dX_t^\varepsilon = -\varepsilon^{2H+1} \frac{(Y_t^\varepsilon)^2}{2} dt + \varepsilon^{2H} Y_t^\varepsilon dW_t, \quad dY_t^\varepsilon = \varepsilon^2 (a + bY_t^\varepsilon) dt + \varepsilon^{2H} cdW_t^H, \quad (6)$$

Note that the drift vanishes in the limit $\varepsilon \rightarrow 0$ and $x_0^\varepsilon = x_0 = 0$.

$\Rightarrow (\widehat{X}_t, \widehat{Y}_t) \equiv 0$, so that there is no $1/\varepsilon^\beta = 1/t^{\beta/2}$ term in the exponential.

Corollary: tail asymptotics

Corollary (tail expansion in Stein-Stein) $dY_t = (a + bY_t)dt + cdW_t^H$

Consider the fractional Stein-Stein model with $X_0 = 0$, $Y_0 = y_0 > 0$. Then as $x \rightarrow \infty$,

$$f_X(T, x) = \exp\left(-c_1 x + c_2 x^{1/2}\right) \frac{1}{x^{1/2}} \left(c_0 + \mathcal{O}\left(x^{1/2}\right)\right)$$

where $c_1 := \Lambda(1)$, $c_2 := \widehat{X}_T \Lambda'(1)$.

Note that the expression on the RHS is independent of the Hurst-parameter!

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Proof: Consider $(X_T^\varepsilon, Y_T^\varepsilon) := (\varepsilon^{2H} X_T, \varepsilon^H Y_T)$ with $X_0^\varepsilon = \varepsilon^{2H} X_0$ and $Y_0^\varepsilon = \varepsilon^H Y_0$.

$$dX_t^\varepsilon = -\frac{(Y_t^\varepsilon)^2}{2} dt + \varepsilon^H Y_t^\varepsilon dW_t, \quad dY_t^\varepsilon = (a\varepsilon^H + bY_t^\varepsilon)dt + \varepsilon^H cdW_t^H, \quad (7)$$

Note that $X_T^\varepsilon \stackrel{\Delta}{=} \varepsilon^{2H} X_T$. $\Rightarrow \mathbb{P}(X_T^\varepsilon \geq y) = \mathbb{P}(X_T \geq y/\varepsilon^{2H})$, \Rightarrow
 $f_X(T, y/\varepsilon^{2H}) = \varepsilon^{2H} f_\varepsilon(T, y)$. Take $y = 1$, that is $x := \varepsilon^{-2H}$. By the theorem,
 $f_\varepsilon(T, 1) \approx \exp\left(-\frac{\Lambda(1)}{\varepsilon^{2H}} + \dots\right) \frac{1}{\varepsilon^H}$, hence
 $f_X(T, x) \approx \exp\left(-\frac{\Lambda(1)}{\varepsilon^{2H}} + \dots\right) \varepsilon^H = \exp(-\Lambda(1)x + \dots) \frac{1}{x^{1/2}}$.

From density to implied volatility: small-time

Recall the Black-Scholes density expansion:

$$f_{\text{BS}}(t, x) \sim t^{-1/2} \exp\left(-\frac{1}{2t} \left(\frac{x}{\sigma}\right)^2\right), \quad \text{as } t \rightarrow 0, \text{ for any } x \in \mathbb{R}.$$

The corollary (Varadhan-type asymptotics) implies that in the fractional Stein-Stein model

$$f_X(t, x) \sim \text{cst } t^{-H} \exp\left(-\frac{d^2(x_0, y_0; x)}{2t^{2H}}\right), \quad \text{as } t \rightarrow 0.$$

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Matching the leading-orders gives

$$\sigma_{\text{BS}}(t, x) \sim \frac{|x|}{d(x_0, y_0; x)} t^{H-1/2} \quad \text{as } t \rightarrow 0.$$

Skew explodes with rate $H - 1/2$ in the short end whenever $H < 1/2$.

From density to implied volatility: tails

Recall the Black-Scholes density expansion:

$$f_{\text{BS}}(t, x) \sim \exp\left(-\frac{x^2}{2\sigma^2 t} - \frac{x}{4}\right) \quad \text{as } x \rightarrow \infty, \text{ for any } t > 0.$$

Our theorem (corollary) says that in the fractional Stein-Stein model (5), we have

$$f_X(t, x) \sim \frac{cst}{x^{1/2}} \exp(-c_1 x + c_2 \sqrt{x}), \quad \text{as } x \rightarrow \infty.$$

From density to implied volatility: tails

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$$f_{\text{BS}}(t, x) \sim \exp\left(-\frac{x^2}{2\sigma^2 t} - \frac{x}{4}\right) \quad \text{as } x \rightarrow \infty, \text{ for any } t > 0.$$

Our theorem (corollary) says that in the fractional Stein-Stein model (5), we have

$$f_X(t, x) \sim \frac{\text{cst}}{x^{1/2}} \exp(-c_1 x + c_2 \sqrt{x}), \quad \text{as } x \rightarrow \infty.$$

Matching the leading-orders gives

$$-c_1 x + c_2 \sqrt{x} \sim -\frac{x^2}{2\sigma^2 t} - \frac{x}{4},$$

and we recover Roger Lee's formula independently of the Hurst exponent in (5).

Proof of Theorem 1

$$dX_t = -\varepsilon^{2H+1} \frac{1}{2} Y_t^2 dt + \varepsilon^{2H} Y_t dW_t, \quad dY_t = \varepsilon^{2H} dW_t^H,$$

with the same initial condition $X_0 = Y_0 = 0$.

$$\text{Density: } f_\varepsilon(T, x) = \exp \left[-\frac{\Lambda(x)}{\varepsilon^{4H}} + \frac{\Lambda'(x) \widehat{X}_T}{\varepsilon^{2H}} \right] \varepsilon^{-2H} \left(c_0 + \mathcal{O}(\varepsilon^{2H}) \right).$$

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Proof: Take $x \in \mathbb{R}$ and a C^∞ -bounded function F such that $F(x) = 0$.

$$f_\varepsilon(T, x) e^{-F(x)/\varepsilon^{4H}} = \frac{1}{2\pi\varepsilon^{2H}} \int_{\mathbb{R}} \mathbb{E} \left\{ \exp \left[i(\zeta, 0) \cdot \left(\frac{X_T^\varepsilon - (x, 0)}{\varepsilon^{2H}} \right) - \frac{F(X_T^\varepsilon)}{\varepsilon^{4H}} \right] \right\} d\zeta.$$

Choose F such that $F(\cdot) + \Lambda_{x_0}(\cdot)$ has a non-degenerate minimum at z . This implies that $k \mapsto F(\phi_T^k(x_0, y_0)) + \frac{1}{2} \|k\|_{\mathcal{H}_H}^2$ has a non-degenerate minimum at $k_0 \in \mathcal{H}_H$.

(For instance $F(z) = \lambda|z - x|^2 - [\Lambda_{x_0, y_0}(z) - \Lambda_{x_0, y_0}(x)]$ with $\lambda > 0$).

Proof of Theorem 1

Replace $\varepsilon^{2H} dB$ ($B := (W, W^H)$) in the SDE by $\varepsilon^{2H} dW + \dot{k}_0$.

Call the corresponding Girsanov-transformed process $\tilde{Z}_t^\varepsilon = (\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon)$:

$$d\tilde{X}_t^\varepsilon = -\varepsilon^{2H+1} \frac{1}{2} \tilde{Y}_t^{\varepsilon 2} dt + \tilde{Y}_t^\varepsilon (\varepsilon^{2H} dW_t + (\dot{k}_0)_1), \quad d\tilde{Y}_t^\varepsilon = \varepsilon^{2H} dW_t^H + (\dot{k}_0)_2.$$

Girsanov factor

$$\mathcal{G} = \exp \left(-\frac{1}{\varepsilon^{2H}} \int_0^T \psi(k_0)_t dB_t - \frac{1}{2\varepsilon^{4H}} \|k_0\|_{\mathcal{H}_H}^2 \right).$$

Therefore

$$\begin{aligned} f(x, T) e^{-F(x)/4\varepsilon^{4H}} &= \frac{1}{2\pi\varepsilon^{2H}} \int_{\mathbb{R}} \mathbb{E} \left[e^{\varepsilon^{2H} i\zeta(\tilde{X}_T - x) - \varepsilon^{-4H} F(\tilde{X}_T)} \mathcal{G} \right] d\zeta \\ &= \frac{1}{2\pi\varepsilon^{2H}} \int_{\mathbb{R}} \mathbb{E} \left[e^{(*)} \right] d\zeta \end{aligned}$$

where

$$(*) = \varepsilon^{2H} i\zeta(\tilde{X}_T - x) - \varepsilon^{-4H} F(\tilde{X}_T) - \varepsilon^{-2H} \int_0^T \psi(\gamma)_t dB_t - \varepsilon^{-4H} \frac{1}{2} \|\gamma\|_{\mathcal{H}_H}^2.$$

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By a stochastic Taylor expansion of $\tilde{Z}_t^\varepsilon = (\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon)$ for $\varepsilon^{2H} \rightarrow 0$,

$$\exp \left(\frac{-F(\tilde{X}_t^\varepsilon)}{\varepsilon^{4H}} \right) = \exp \left[\frac{-1}{\varepsilon^{4H}} \left(F(x) - \varepsilon^{2H} \int_0^T \psi(k_0)_t dB_t - \varepsilon^{2H} \hat{X}_T \cdot \Lambda'_{x_0}(x) + \mathcal{O}(\varepsilon^{4H}) \right) \right]$$

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The rest of the proof follows Ben Arous' proof for X_T^ε .

1 Introduction

2 Approach via density asymptotics

Varadhan-type asymptotics for fractional SDEs
Rescalings and density asymptotics for fractional models
Corollaries: Short-time/tail expansion in fractional models
Implied volatility asymptotics
Idea of the proof

3 Bypassing density asymptotics

Refined expansions and moderate regimes
A non-Markovian extension of Osajima's energy expansion
Implied volatility asymptotics

Bypassing density asymptotics

Direct call price expansion: For $x \geq 0$ the option price satisfies

$$\begin{aligned} c(\varepsilon^{1-2H}x, t) &:= E \left[(\exp(X_t) - \exp(\varepsilon^{1-2H}x))^+ \right] \\ &= E \left[(\exp(Z_t) - \exp(\varepsilon^{1-2H}x))^+ G|_* \right] \end{aligned}$$

where Z is the controlled process around the optimal path k and $G|_* = e^{\frac{-I(x)}{\varepsilon^{4H}}} e^{\frac{-I'(x)g_1}{\varepsilon^{2H}}}$ is the Girsanov factor for the optimal path, and g_1 a Gaussian random variable. Then

$$\begin{aligned} c(\varepsilon^{1-2H}x, t) &= \exp\left(-\frac{I(x)}{\varepsilon^{4H}}\right) \exp\left(x\varepsilon^{1-2H}\right) J(\varepsilon, x), \quad \text{where } \hat{U} := \hat{Z}_1^\varepsilon - x \text{ and} \\ J(\varepsilon, x) &:= E \left[\exp\left(\frac{-I'(x)}{\varepsilon^{2H}} \hat{U}\right) \left(\exp(\varepsilon^{1-2H}\hat{U}) - 1\right) \exp(I'(x)R_2) \mathbf{1}_{\hat{U} \geq 0} \right]. \end{aligned}$$

⇒ Expansion (uniformly in x) via expansion of the “energy” I directly (Osajima).

Moderate regimes

- Moderate Regimes (in the sense of Friz-Gerhold-Pinter '16) **interpolate** between **out-of-the-money** calls with fixed strike $\left(\log \frac{K}{S_0}\right) = k > 0$ and **at-the-money** $k = 0$ calls: Now $k_t = ct^\theta \Rightarrow$ MOTM (for $0 < \theta < \frac{1}{2}$) and AATM (for larger θ)
- **Reflects market reality**: options closer expiry \Rightarrow strikes closer to the money first observed by Mijatović-Tankov on FX markets
- The moderate regime (MOTM) **permits explicit computations** for the rate function $\Lambda(k)$ in terms of the model parameters
Moderate deviations \Rightarrow **Advantage over large deviations** (OTM) case where the $\Lambda(k)$ often related to geodesic distance problems and not explicitly available.
- MOTM expansions naturally involve **quantities very familiar to practitioners**, notably spot (implied) volatility, implied volatility skew ...
- In some cases (fractional volatility models) the scaling θ permits a **fine-tuning** to understand the behavior and derivatives of the energy function.

Moderate regimes for rough volatility

Rescalings \implies We tacitly agreed to consider $\mathbb{P}(X_t \approx t^{1/2-H_X})$. Now it is only a small step to consider instead (for some suitable small $\theta > 0$)

$$\mathbb{P}(X_t \approx t^{1/2-H+\theta_X}).$$

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Theorem (Bayer-Friz-Gulisashvili-H-Stemper)

Consider a moderately out-of-the-money call $k_t = xt^{1/2-H+\theta}$; $\theta \in (0, H)$ resp. $\theta \in (0, \frac{2H}{3})$. Then as $t \rightarrow 0$, the following holds

$$\log c(k_t, t) \approx \frac{1}{2} \Lambda''(0) \frac{x^2}{t^{2H-2\theta}} + \frac{1}{6} \Lambda'''(0) \frac{x^3}{t^{2H-3\theta}},$$

where we have explicit expressions: $\Lambda''(0) = \frac{1}{\sigma_0^2}$ and $\Lambda'''(0) = -\rho \frac{6\sigma_0'}{\sigma_0^4} \langle K, 1 \rangle$.

Here K denotes the Volterra kernel and $\langle K, 1 \rangle := \int_0^1 \int_0^t K(t, s) ds dt$.

Moderate regimes for rough volatility

Corollary (MOTM Implied volatility skew)

In the moderately out-of-the-money case ($k_t = xt^{1/2-H+\theta}$; $\theta \in (0, H)$; $\theta \in (0, \frac{2H}{3})$) the implied volatility satisfies the expansion

$$\sigma_{impl}(k_t, t) = \sigma_0 - \rho \frac{\sigma'_0}{\sigma_0} \int_0^1 \int_0^t K(t, s) ds dt k_t t^{H-1/2} (1 + o(1)).$$

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Proof: The statement follows from the Theorem via matching components in the asymptotic expansions and by (see Gao-Lee (2014)) using

$$t\sigma_{impl}^2(k_t, t) \approx \frac{-k_t^2}{2 \log c(k_t, t)}.$$

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This formula for the skew is in accord with ones previously derived by Alòs-León-Vives (2007) and Fukasawa (2011, 2016) in different settings.

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Thank you!