

Existence for a calibrated Regime Switching Local Volatility model

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Outline

- 1 Processes matching given marginals
 - Motivation
 - Simulation of calibrated LSV models and theoretical results
- 2 A new *fake* Brownian motion
 - The studied problem
 - Main Result
 - Ideas of proof
- 3 Existence of Calibrated RSLV models
 - The calibrated RSLV model
 - Main results

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Trying to match marginals

- The market gives the prices of European Calls $C(T, K)$ for $T, K > 0$ (idealized situation; in practice only $(C(T_i, K_i))_{1 \leq i \leq I}$).
- A model $(S_t)_{t \geq 0}$ is **calibrated** to European options if

$$\forall T, K \geq 0, C(T, K) = \mathbb{E} \left[e^{-rT} (S_T - K)^+ \right].$$

- By Breeden and Litzenberger (1978), {prices of European Call options for all $T, K > 0$ } \iff {marginal distributions of $(S_t)_{t \geq 0}$ }.
- Dupire Local Volatility model (1992), matching market marginals:

$$dS_t = rS_t dt + \sigma_{Dup}(t, S_t) S_t dW_t$$

$$\sigma_{Dup}(T, K) = \sqrt{2 \frac{\partial_T C(T, K) + rK \partial_K C(T, K)}{K^2 \partial_{KK}^2 C(T, K)}}$$

LSV models

- **Motivation:** get processes with richer dynamics (e.g. take into account volatility risk) and satisfying marginal constraints.
- Lipton (2002) and Piterbarg (2007),...: Local and Stochastic Volatility (**LSV**) model

$$dS_t = rS_t + f(Y_t)\sigma(t, S_t)S_t dW_t$$

- “Adding uncertainty” to LV models by a random multiplicative factor $f(Y_t)$ where $(Y_t)_{t \geq 0}$ is a stochastic process and $f > 0$.

Calibration of LSV Models

- By Gyongy's theorem (1988), the LSV model is calibrated to $C(T, K), \forall T, K > 0$ if

$$\mathbb{E} [(f(Y_t)\sigma(t, S_t)S_t)^2 | S_t] = (\sigma_{Dup}(t, S_t)S_t)^2$$

$$\sigma(t, x) = \frac{\sigma_{Dup}(t, x)}{\sqrt{\mathbb{E}[f^2(Y_t) | S_t = x]}}$$

- The obtained SDE is **nonlinear** in the sense of McKean:

$$dS_t = rS_t dt + \frac{f(Y_t)}{\sqrt{\mathbb{E}[f^2(Y_t) | S_t]}} \sigma_{Dup}(t, S_t) S_t dW_t.$$

Simulation results

- Ren, Madan and Qian (2007): solve numerically the associated Fokker-Planck PDE, and get the joint-law of (S_t, Y_t) .
- Guyon and Henry-Labordère (2011): efficient calibration procedure based on kernel approximation of the conditional expectation.
Subsequent extension to stochastic interest rates, stochastic dividends, multidimensional local correlation models,...
- However, calibration errors seem to appear when the range of $f(Y)$ is **too large**.

Theoretical results

- Abergel and Tachet (2010): perturbation of the constant f case (Dupire) \longrightarrow existence for the restriction to a compact spatial domain of the associated Fokker-Planck equation when $\sup f - \inf f$ small.
- Global existence and uniqueness to LSV models remain an open problem.

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A simpler SDE

- The stochastic factor Y is a r.v. (does not evolve with time) with values in $\mathcal{Y} := \{y_1, \dots, y_d\}$ such that

$$\forall i \in \{1, \dots, d\}, \alpha_i = \mathbb{P}(Y = y_i) > 0.$$

It is independent from S_0 and $(W_t)_{t \geq 0}$.

- We suppose $\sigma_{Dup} \equiv 1$, $r = \frac{1}{2}$ and consider $X_t = \ln(S_t)$.
- Since $\mathbb{E}[f^2(Y)|\mathcal{S}_t] = \mathbb{E}[f^2(Y)|X_t]$ and neglecting the drift term $\frac{1}{2} - \frac{1}{2} \frac{f^2(Y)}{\mathbb{E}[f^2(Y)|X_t]}$ with vanishing conditional expectation given X_t , we get

$$(FBM) \begin{cases} dX_t = \frac{f(Y)}{\sqrt{\mathbb{E}[f^2(Y)|X_t]}} dW_t \\ X_0 \sim \mu_0. \end{cases}$$

with X_0 , Y and $(W_t)_{t \geq 0}$ independent.

- If $X_0 = 0$, we expect that $X_t \sim \mathcal{N}_1(0, t)$.

Fake Brownian motion

Definition

A fake Brownian motion is a martingale $(M_t)_{t \geq 0}$ such that for all $t \geq 0$, $M_t \sim \mathcal{N}_1(0, t)$ and $(M_t)_{t \geq 0}$ is not a Brownian motion.

Discontinuous fake Brownian motions : Madan and Yor (2002), Hamza and Klebaner (2007), Henry-Labordère, Tan and Touzi (2016),...

Continuous fake Brownian motions : Albin (2010), Oleszkiewicz (2010), Baker, Donati-Martin and Yor (2011),...

Fake exponential Brownian motion : Hobson (2013)

A new continuous fake Brownian motion

Lemma

If the positive function f is not constant on \mathcal{Y} , then any solution to the SDE

$$dX_t = \frac{f(Y)}{\sqrt{\mathbb{E}[f^2(Y)|X_t]}} dW_t, \quad X_0 = 0$$

with Y and $(W_t)_{t \geq 0}$ indep. is a continuous fake Brownian motion.

- By Gyongy's theorem, $\forall t \geq 0, X_t \sim \mathcal{N}_1(0, t)$.
- If $(X_t)_{t \geq 0}$ is a Brownian motion then a.s. $\forall t \geq 0$, $\langle X \rangle_t = t$ i.e. ds a.e. $\frac{f^2(Y)}{\mathbb{E}[f^2(Y)|X_s]} = 1 = \frac{f(Y)}{\sqrt{\mathbb{E}[f^2(Y)|X_s]}}$ so that a.s. $\forall t \geq 0, X_t = W_t$. Therefore $X_t \perp Y$, $\mathbb{E}[f^2(Y)|X_t] = \mathbb{E}[f^2(Y)]$ and $f^2(Y) = \mathbb{E}[f^2(Y)]$ is constant.

Existence to SDE (FBM) and fake Brownian motion

We define for $i \in \{1, \dots, d\}$, $\lambda_i := f^2(y_i)$.

Theorem

Condition (C) : there exists a symmetric positive definite $\Gamma \in \mathbb{R}^{d \times d}$ such that for all $k \in \{1, \dots, d\}$, the $d \times d$ matrix

$$\Gamma_{ij}^{(k)} = \frac{\lambda_i + \lambda_j}{2} (\Gamma_{ij} + \Gamma_{kk} - \Gamma_{ik} - \Gamma_{jk}) \text{ is positive definite on } e_k^\perp.$$

Under (C), there exists a weak solution to the SDE (FBM).

if $d = 2$, (C) is satisfied : choice $\Gamma = I_2$,

if $d = 3$, (C) $\Leftrightarrow \frac{1}{\beta_1\beta_2} + \frac{1}{\beta_2\beta_3} + \frac{1}{\beta_3\beta_1} > \frac{1}{4}$ with

$$\beta_1 = \left| \sqrt{\frac{\lambda_2}{\lambda_3}} - \sqrt{\frac{\lambda_3}{\lambda_2}} \right|, \beta_2 = \left| \sqrt{\frac{\lambda_3}{\lambda_1}} - \sqrt{\frac{\lambda_1}{\lambda_3}} \right|, \beta_3 = \left| \sqrt{\frac{\lambda_1}{\lambda_2}} - \sqrt{\frac{\lambda_2}{\lambda_1}} \right|$$

if $d \geq 4$, $\max_{1 \leq k \leq d} \sum_{i \neq k} \lambda_i \sum_{i \neq k} \frac{1}{\lambda_i} \leq (d+1)^2 \Rightarrow$ (C):
 $\Gamma = I_d$.

The associated Fokker Planck system

- For $i \in \{1, \dots, d\}$, define p_i s.t., for $\phi \geq 0$ and measurable, $\mathbb{E} [\phi(X_t) 1_{\{Y=y_i\}}] = \int_{\mathbb{R}} \phi(x) p_i(t, x) dx$.
- The associated Fokker-Planck system is:

$$\forall i \in \{1, \dots, d\}, \partial_t p_i = \frac{1}{2} \partial_{xx}^2 \left(\frac{\sum_j p_j}{\sum_j \lambda_j p_j} \lambda_i p_i \right)$$

$$p_i(0) = \alpha_i \mu_0$$

- $\sum_j p_j$ is solution to the **heat equation**.

Rewriting into divergence form

The system satisfied by $p = (p_1, \dots, p_d)$ can be rewritten in divergence form:

$$\partial_t p = \frac{1}{2} \partial_x \left(\sum_{k=1}^d w_k(p) M^{(k)}(p) \partial_x p \right)$$

where $w_k(p) := \frac{\lambda_k p_k}{\sum_j \lambda_j p_j}$, $\sum_{k=1}^d w_k(p) = 1$ and for $\bar{\lambda}(p) := \frac{\sum_j \lambda_j p_j}{\sum_j p_j}$,

$$M^k(p) := \begin{pmatrix} \frac{\lambda_1}{\bar{\lambda}} & & & & & & \\ & \cdot & & & & & \\ & & \frac{\lambda_{k-1}}{\bar{\lambda}} & & & & \\ \left(1 - \frac{\lambda_1}{\bar{\lambda}}\right) & \cdot & \left(1 - \frac{\lambda_{k-1}}{\bar{\lambda}}\right) & 1 & \left(1 - \frac{\lambda_{k+1}}{\bar{\lambda}}\right) & \cdot & \left(1 - \frac{\lambda_d}{\bar{\lambda}}\right) \\ & & & \frac{\lambda_{k+1}}{\bar{\lambda}} & & & \\ & & & & \cdot & & \\ & & & & & & \frac{\lambda_d}{\bar{\lambda}} \end{pmatrix} \leftarrow \text{row } k$$

Computing Energy Estimates

- Multiply the system by $p^*(J_d + \epsilon\Gamma)$ where $(J_d)_{ij} = 1$ and $\epsilon > 0$, and integrate in $x \in \mathbb{R}$ by parts :

$$\frac{1}{2}\partial_t \left(\int_{\mathbb{R}} p^*(J_d + \epsilon\Gamma) p dx \right) = -\frac{1}{2} \int_{\mathbb{R}} \sum_{k=1}^d w_k(p) \partial_x p^*(J_d + \epsilon\Gamma) M^{(k)}(p) \partial_x p dx.$$

- By (C), $\exists \eta > 0$, $\forall y = (y_1, \dots, y_d) \in \mathbb{R}^d$ with $\sum_{j=1}^d y_j = 0$,

$$\bar{\lambda}(p) y^* \Gamma M^{(k)}(p) y = (y - y_k e_k)^* \Gamma^{(k)} (y - y_k e_k) \geq \eta |y|^2.$$
- Since $J_d M^{(k)}(p) = J_d$ so that $y^* J_d M^{(k)}(p) y = (\sum_{i=1}^d y_i)^2$, we deduce that for ϵ small enough,

$$\inf_{p, y \in \mathbb{R}^d} \frac{1}{|y|^2} \sum_{k=1}^d w_k(p) y^* (J_d + \epsilon\Gamma) M^{(k)}(p) y > 0.$$

- \Rightarrow **EE** in $L^2([0, T], (H^1(\mathbb{R}))^d) \cap L^\infty([0, T], (L^2(\mathbb{R}))^d)$.

Step 1/3: Existence to an approximate PDS when $\mu_0(dx) = p_0(x)dx$, $p_0 \in L^2(\mathbb{R})$

For $\epsilon > 0$, use **Galerkin's** method to solve an approximate PDE:

$$\partial_t p^\epsilon = \frac{1}{2} \partial_x (M^\epsilon(p^\epsilon) \partial_x p^\epsilon) \quad p^\epsilon(0) = (\alpha_1, \dots, \alpha_d) p_0.$$

$$\text{where } \sum_{k=1}^d w_k(\rho) M_{ii}^{(k)}(\rho) = 1 + \frac{\sum_{l \neq i} \lambda_l \rho_l \sum_l (\lambda_i - \lambda_l) \rho_l}{(\sum_l \lambda_l \rho_l)^2}$$

$$\longrightarrow M_{ii}^\epsilon(\rho) = 1 + \frac{\sum_{l \neq i} \lambda_l \rho_l^+ \sum_l (\lambda_i - \lambda_l) \rho_l^+}{(\epsilon \vee \sum_l \lambda_l \rho_l^+)^2},$$

$$\text{and for } j \neq i, \sum_{k=1}^d w_k(\rho) M^{(k)}(\rho)_{ij} = \frac{\lambda_i \rho_i \sum_l (\lambda_l - \lambda_j) \rho_l}{(\sum_l \lambda_l \rho_l)^2}$$

$$\longrightarrow M_{ij}^\epsilon(\rho) = \frac{\lambda_i \rho_i^+ \sum_l (\lambda_l - \lambda_j) \rho_l^+}{(\epsilon \vee \sum_l \lambda_l \rho_l^+)^2}.$$

Step 1/3: Existence to an approximate PDS when $\mu_0 \in L^2(\mathbb{R})$

- $\rho \mapsto M^\epsilon(\rho)$ locally Lipschitz and bounded $\rightarrow \exists!$ solution p_m^ϵ to a projection of the equation in dimension m .
- coercivity uniform in ϵ under (C) : \exists solution p^ϵ satisfying uniform in ϵ EE by taking the limit $m \rightarrow \infty$.
- Taking p_ϵ^- as test function, we show that $p_\epsilon \geq 0$.
- $\forall \epsilon, \forall i, \sum_j M_{ji}^\epsilon = 1 \implies \sum_j p_j^\epsilon$ solves the heat equation \rightarrow lower bound uniform in ϵ (but not t, x) for $\sum_j \lambda_j p_j^\epsilon$.
- $\epsilon \rightarrow 0$, existence of a solution to the original PDS.

Step 2/3: Existence to the PDS when $\mu_0 \in \mathcal{P}(\mathbb{R})$

- By **mollification** of μ_0 , we use the results of Step 1 to extract a solution to the PDS when $\mu_0 \in \mathcal{P}(\mathbb{R})$.
- We use the fact that $\sum_j p_j$ is solution to the heat equation to control the rate of explosion of $t \mapsto \int_{\mathbb{R}} \sum_{i=1}^d p_i^2(t, x) dx$ as $t \rightarrow 0$ uniformly in the mollification parameter.

Step 3/3: Existence of a weak the SDE (FBM)

Theorem (Figalli (2008))


For $a : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}_+$ and $b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ meas. and bounded let $L_t \varphi(x) = \frac{1}{2} a(t, x) \varphi''(x) + b(t, x) \varphi'(x)$.

If $[0, T] \ni t \mapsto \mu_t \in \mathcal{M}_+(\mathbb{R})$ is weakly continuous and solves the Fokker-Planck equation $\partial_t \mu_t = L_t^* \mu_t$ in the sense of distributions then there exists a probability measure P on $C([0, T], \mathbb{R})$ with marginals $(P_t = \mu_t)_{t \in [0, T]}$ such that

$\forall \varphi \in C_b^2(\mathbb{R})$, $\varphi(X_t) - \int_0^t L_s \varphi(X_s) ds$ is a P -martingale.

\Rightarrow for $i \in \{1, \dots, d\}$, there exists a probab. P^i on $C([0, T], \mathbb{R})$ with $P_0^i = \mu_0$ and $P_t^i = \frac{p_i(t, x) dx}{\alpha_i}$ for $t \in (0, T]$ and $\forall \varphi \in C_b^2(\mathbb{R})$,

$$\varphi(X_t) - \int_0^t \frac{f^2(y_i) \sum_{j=1}^d p_j}{\sum_{j=1}^d f^2(y_j) p_j} (s, X_s) \varphi''(X_s) ds \text{ is a } P^i\text{-martingale.}$$

$P(dX, dY) = \sum_{i=1}^d \alpha_i P^i(dX) \otimes \delta_{y_i}(dY)$ weak solution. 

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Presentation

- We consider the following dynamics (RSLV):

$$dS_t = rS_t dt + \frac{f(Y_t)}{\sqrt{\mathbb{E}[f^2(Y_t)|S_t]}} \sigma_{Dup}(t, S_t) S_t dW_t,$$

where $(Y_t)_{t \geq 0}$ takes values in \mathcal{Y} , and

$$\mathbb{P}(Y_{t+dt} = y_j | Y_t = y_i, S_t = x) = q_{ij}(x) dt.$$

- **Switching** diffusion, special case of **LSV** model.
- Jump distributions and intensities are functions of the asset level.

Assumptions

- (C), (**Coerc. 1**): f satisfies condition (C).
- (HQ), (**Bounded I**) $\exists \bar{q} > 0$, s.t. $\forall x \in \mathbb{R}, |q_{ij}(x)| \leq \bar{q}$.

We define $\tilde{\sigma}_{Dup}(t, x) := \sigma_{Dup}(t, e^x)$.

- (H1), (**Bounded vol.**) $\tilde{\sigma}_{Dup} \in L^\infty([0, T], W^{1,\infty}(\mathbb{R}))$.
- (H2), (**Coerc. 2**) $\exists \underline{\sigma} > 0$ s.t. $\underline{\sigma} \leq \tilde{\sigma}_{Dup}$ a.e. on $[0, T] \times \mathbb{R}$.
- (H3), (**Regul. 1**) $\exists \eta \in (0, 1], \exists H_0 > 0$, s.t.
 $\forall s, t \in [0, T], \forall x, y \in \mathbb{R}$,

$$|\tilde{\sigma}_{Dup}(s, x) - \tilde{\sigma}_{Dup}(t, y)| \leq H_0 (|x - y|^\eta + |t - s|^\eta).$$

(HQ), (H1) and (H2) permit to generalize the energy estimations to the Fokker-Planck system associated with $((\ln(S_t), Y_t))_{t \in [0, T]}$

With (H3), uniqueness and Aronson estimates for the Fokker-Planck equation associated with $(\ln(S_t^{Dup}))_{t \in [0, T]}$ where

$$dS_t^{Dup} = \sigma_{Dup}(t, S_t^{Dup}) S_t^{Dup} dW_t + r S_t^{Dup} dt, \quad S_0^{Dup} = S_0.$$

→ replaces the heat equation

Main results

Theorem

Under Conditions (H1)-(H3), (HQ) and (C) there exists a weak solution to the SDE (RSLV). Moreover, it has the same marginals as the solution to the local volatility SDE

$$dS_t^{Dup} = \sigma_{Dup}(t, S_t^{Dup}) S_t^{Dup} dW_t + r S_t^{Dup} dt, \quad S_0^{Dup} = S_0.$$

- We generalize the results of Figalli to the regime switching case.
- Uniqueness?

Main obstacle to deal with usual LSV models

Corresponding generalized (FBM) equation

$$dX_t = \frac{f(Y_t)}{\sqrt{\mathbb{E}[f^2(Y_t)|X_t]}} dW_t$$

$$dY_t = \eta(Y_t)(\rho dW_t + \sqrt{1 - \rho^2} dB_t) + b(Y_t) dt$$

with $(B_t)_{t \geq 0}$ Brownian motion indep. of $(W_t)_{t \geq 0}$.

Typically $|\rho| \neq 1$ and (X_t, Y_t) should admit a density $p(t, x, y)$.
Even when $0 < \inf f \leq \sup f < \infty$, in the divergence form of the term

$$\begin{aligned} \partial_{xx} \left(\frac{f^2(y) \int p(t, x, z) dz}{\int f^2(z) p(t, x, z) dz} p(t, x, y) \right) &= \partial_x \left(\frac{f^2(y) \int p(t, x, z) dz}{\int f^2(z) p(t, x, z) dz} \partial_x p(t, x, y) \right) \\ &+ \partial_x \left(\frac{f^2(y) p(t, x, y)}{\int f^2(z) p(t, x, z) dz} \int \partial_x p(t, x, z) dz - \frac{f^2(y) p(t, x, y) \int p(t, x, z) dz}{(\int f^2(z) p(t, x, z) dz)^2} \int f^2(z) \partial_x p(t, x, z) dz \right) \end{aligned}$$

the **red factors** replacing $\frac{f^2(y_i) p_i(t, x)}{\sum_{j=1}^d f^2(y_j) p_j(t, x)}$ and $\frac{f^2(y_i) p_i(t, x) \sum_{j=1}^d p_j(t, x)}{(\sum_{j=1}^d f^2(y_j) p_j(t, x))^2}$
are no longer bounded (same problem for kernel approximations of $\mathbb{E}[f^2(Y_t)|X_t]$).

Thank you for your attention!