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SAM

Wavelet FEM for option pricing in stochastic volatility models

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Outline

SV models

Variational formulation

Discretization

Numerical results

Conclusion

References

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- Bates, D. S.: *Jumps and stochastic volatility: the exchange rate process implicit in Deutsche Mark options* Review of Financial Studies, **9**, 69–107 (1996).
- von Petersdorff, T. and Schwab, C.: *Numerical solution of parabolic equations in high dimensions*. ESAIM: Mathematical Modelling and Numerical Analysis **38** N°1 (2004) 93–127.

Goal: Numerical pricing of options under a general class of SV models by the FE method

$(Z_t)_{t \geq 0}$ a \mathbb{R}^d -valued strong Markov process.

Find arbitrage-free price $u(z, t)$ of contract on Z ,

$$u(z, t) = \mathbb{E}^{\mathbb{Q}}[e^{-r(T-t)}g(Z_T) \mid Z_t = z].$$

$u(z, t)$ is (viscosity) solution of

$$-\partial_t u + \mathcal{A}u + ru = 0, \quad u(z, T) = g,$$

where \mathcal{A} is the infinitesimal generator of Z .

Dynamics of $Z = (X, Y)$

X \mathbb{R}^n -valued, log-price dynamics of underlyings.

Y \mathbb{R}^p -valued, dynamics of volatility, $p \geq n$.

$$dZ_t = \gamma(Z_{t-}, t)dt + Q(Z_{t-}, t)dW_t + \int_E \sigma(Z_{t-}, t, \xi)\tilde{N}(dt, d\xi),$$
$$Z_0 = z.$$

- $\gamma : \mathbb{R}^{n+p} \times [0, T] \rightarrow \mathbb{R}^{n+p}$
- $Q : \mathbb{R}^{n+p} \times [0, T] \rightarrow \mathbb{R}^{(n+p) \times m}$
- $\sigma : \mathbb{R}^{n+p} \times [0, T] \times \mathbb{R}^k \rightarrow \mathbb{R}^{(n+p) \times \ell}$
- W \mathbb{R}^m -valued Brownian motion, \tilde{N} compensated martingale measure of ℓ -dim. Poisson random measure

What is \mathcal{A} ?

$$\mathcal{A} = \mathcal{A}_Q + \mathcal{A}_\gamma + \mathcal{A}_\nu$$

$$\mathcal{A}_Q(u)(z) := -\frac{1}{2} \operatorname{tr} [(QQ^\top)(z) D^2 u(z)]$$

$$\mathcal{A}_\gamma(u)(z) := -\langle \gamma(z), Du(z) \rangle$$

$$\mathcal{A}_\nu(u)(z) := -\sum_{j=1}^{\ell} \mathcal{A}_{\nu,j}(u)(z)$$

$$\mathcal{A}_{\nu,j}(u)(z) := \int_E [u(z + \sigma_j(z, \xi)) - u(z) - \langle \sigma_j(z, \xi), Du(z) \rangle] \nu_j(d\xi)$$

Two examples ($n = p = 1$)

■ Heston ('93)

$$\gamma(\mathbf{z}) = \begin{pmatrix} -y/2 \\ \alpha(m - y) + \lambda(x, y) \end{pmatrix}, \quad \alpha, m > 0,$$

$$Q(\mathbf{z}) = \begin{pmatrix} \sqrt{y} & 0 \\ \beta\rho\sqrt{y} & \beta\sqrt{1-\rho^2}\sqrt{y} \end{pmatrix}, \quad \beta > 0, \rho \in [-1, 1].$$

■ Barndorff-Nielsen and Shepard ('01)

$$\gamma(\mathbf{z}) = \begin{pmatrix} -y/2 + \alpha \\ -\lambda y + \beta \end{pmatrix}, \quad \alpha, \beta \in \mathbb{R}, \lambda > 0,$$

$$Q(\mathbf{z}) = \begin{pmatrix} \sqrt{y} \\ 0 \end{pmatrix}, \quad \sigma(\mathbf{z}, \xi) = \begin{pmatrix} \rho\xi \\ \xi \end{pmatrix}, \quad \rho \leq 0.$$

Analysis and numerics depend on \mathcal{Q}

Three classes: $\mathcal{Q}\mathcal{Q}^\top$ is ¹

- (A) positive definite: $\exists c > 0 : \mathbf{z}^\top (\mathcal{Q}\mathcal{Q}^\top)(\mathbf{z}) \geq c \mathbf{z}^\top \mathbf{z},$
 $\forall \mathbf{z} \in \mathbb{R}^d$
 - (B) positive semidefinite: $\exists \mathbf{z} \in \mathbb{R}^d : \text{rank}((\mathcal{Q}\mathcal{Q}^\top)(\mathbf{z})) < d$
 - (C) positive semidefinite: $\forall \mathbf{z} \in \mathbb{R}^d : \text{rank}((\mathcal{Q}\mathcal{Q}^\top)(\mathbf{z})) < d$
- (B) & (C) \rightarrow : P(l)DE with nonnegative characteristic form
 - typically, (B) holds exactly for one \mathbf{z} , usually $\mathbf{z} = 0$
 - (C) means that there is at least one coordinate direction without diffusion component

¹ $d := n + p$

Analytical framework for models of class (B)

Given $f \in L^2(0, T; V^*)$, find $u \in L^2(0, T; V) \cap H^1(0, T; V^*)$ such that

$$\langle \partial_t u, v \rangle + \underbrace{\langle \mathcal{A}u, v \rangle}_{a(u,v)} = \langle f, v \rangle, \quad \forall v \in V$$

Typically, V is a **weighted** H^1 -Sobolev space.

Theorem (Heston, change of var. $\tilde{u}(x, \tilde{y}) := u(x, 1/4\tilde{y}^2)$)

Assume $1 - 4\sqrt{2}|2\alpha\beta^{-2} - 1/2| > \rho^2$. Then $a(\cdot, \cdot)$ satisfies:

$\exists C, c, c_0 > 0$ such that for all $u, v \in V$

$$|a(u, v)| \leq C \|u\|_V \|v\|_V, \quad a(u, u) \geq c \|v\|_V^2 - c_0 \|v\|^2,$$

where $V = \overline{C_0^\infty(\mathbb{R}^2)}^{\|\cdot\|_V}$, $\|v\|_V^2 := \|y\partial_x v\|^2 + \|\partial_y v\|^2 + \|v\|^2$.

Analytical framework for models of class (C)

Treat time t as **additional space variable**. Consider

$$\begin{aligned} -\operatorname{div}(\mathcal{Q}(\mathbf{z})\nabla u) + \langle \gamma(\mathbf{z}), \nabla u \rangle + cu + \mathcal{A}_\nu(u) &= f \quad \text{in } \mathcal{D} \subset \mathbb{R}^d \\ u &= 0 \quad \text{in } \Gamma_0 \cup \Gamma_- \end{aligned}$$

$\mathcal{V} := \{u \in H^1(\mathcal{D}) : \gamma_0 u|_{\Gamma_0} = 0\}$. The weak form reads: Find $u \in \mathcal{H}$ such that

$$B(u, v) = (f, v), \quad \forall v \in \mathcal{V}$$

Bilinear form $B : \mathcal{H} \times \mathcal{V} \rightarrow \mathbb{R}$

$$B(u, v) = (\mathcal{Q}\nabla u, \nabla v) - (u, \operatorname{div}(\gamma v)) + c(u, v) + \langle u, v \rangle_+ + a_\nu(u, v).$$

Example: BNS model

Theorem

Assume the Lévy measure has the form $\nu(d\xi) = C e^{-\kappa\xi} \xi^{-1-\alpha}$, $C, \kappa > 0$, $\alpha \in (0, 1)$, $\xi \in \mathbb{R}_+$. Then the bilinear form associated to \mathcal{A}_ν satisfies: $\exists C_1, C_2, C_3 > 0$ such that for all $u, v \in V = \tilde{H}^{\alpha/2}(\mathcal{D})$

$$|a_\nu(u, v)| \leq C_1 \|u\|_V \|v\|_V, \quad a_\nu(u, u) \geq C_2 \|v\|_V^2 - C_3 \|v\|^2,$$

Theorem

Let $\omega > 0$ satisfy $c > \lambda/2 + C_3 + \omega$. Then \exists Hilbert space $\hat{\mathcal{H}} \subset \mathcal{H}$ and unique $u \in \hat{\mathcal{H}}$ solving $B(u, v) = (f, v)$, $\forall v \in \mathcal{V}$. Here

$$\mathcal{H} := \overline{\mathcal{V}}^{\|\cdot\|_H}, \quad \|w\|_{\tilde{\mathcal{H}}}^2 := (x_2 \partial_{x_1} w, \partial_{x_1} w) + (w, w) + \langle w, w \rangle_{\Gamma_+} + (w, w)_{\alpha/2}.$$

FEM for models of class (B)

$V^L \subset V$, $\dim V^L < \infty$. The Galerkin approximation: Find $u^L \in L^2(0, T; V^L) \cap H^1(0, T, (V^L)^*)$ such that

$$\langle \partial_t u^L, v \rangle + \langle \mathcal{A}u^L, v \rangle = \langle f, v \rangle, \quad \forall v \in V^L$$

is equivalent to the ODE: Find $\underline{U}_L(t) \in \mathbb{R}^{\dim V^L}$ such that

$$\mathbf{M} \dot{\underline{U}}_L(t) + \mathbf{A} \underline{U}_L(t) = \underline{F}(t)$$

where, with $V^L = \text{span}\{\Phi_j\}_{j=1}^{\dim V^L}$,

$$\mathbf{M} = ((\Phi_i, \Phi_j))_{1 \leq i, j \leq \dim V^L}, \quad \mathbf{A} = (a(\Phi_i, \Phi_j))_{1 \leq i, j \leq \dim V^L}.$$

Note: V^L is a sparse tensor product wavelet space, with $\dim V^L = O(2^L L^{d-1})$ in contrast to classical FE spaces $\dim V_c^L = O(2^{dL})$.

Convergence rates

hp-discontinuous Galerkin discretization in time +
preconditioned GMRES to approximatively solve linear systems

⇒

Theorem

The fully discrete Galerkin scheme with polynomial degree r of wavelets yields approximate option prices at maturity $U_L(T)$ with accuracy ($N_L = \dim V^L$, $d = n + p$)

$$\|U_L(T) - u(T)\| \leq CN_L^{-s} (\log N_L)^{(d-1)s}, \quad s = r + \frac{r}{d(r+1) - 1}$$

and can be computed with at most $O(N_L(\log N_L)^7)$ operations (if $\nu \equiv 0$).

Note: for classical FE spaces we have $s = \frac{r+1}{d}$

SDFEM for models of class (C)

Basic idea of streamline diffusion FEM: add **extra diffusion** in the direction of streamline: take test functions $v + \delta_L \langle \gamma, \nabla v \rangle$ rather than just v . δ_L : SD parameter to be chosen later.

Find $u_L \in V^L$ such that

$$B_\delta(u_L, v) = \ell_\delta(v), \quad \forall v \in V^L,$$

where

$$B_\delta(\varphi, \phi) = B(\varphi, \phi) + \delta_L \sum_{\kappa \in \mathcal{T}^L} (-\operatorname{div}(\mathcal{Q}\nabla\varphi) + \mathbf{c}\varphi, \langle \gamma, \nabla\phi \rangle)_\kappa \\ + \delta_L (\langle \gamma, \nabla\varphi \rangle, \langle \gamma, \nabla\phi \rangle)_S + \delta_L \mathbf{a}_\nu(\varphi, \langle \gamma, \nabla\phi \rangle),$$

$$\ell_\delta(\phi) = (f, \phi) + \delta_L \sum_{\kappa \in \mathcal{T}^L} (f, \langle \gamma, \nabla\phi \rangle)_\kappa$$

Choice of δ_L

Assume $\exists c_0 : c - 1/2 \operatorname{div} \gamma - C_3 \geq c_0 > 0$.

Streamline diffusion norm $\| \cdot \|_{SD}$

$$\|v\|_{SD}^2 := \|\sqrt{Q} \nabla v\|^2 + c_0 \|v\|^2 + \|v\|_{\Gamma_- \cup \Gamma_+}^2 + C_2 \|v\|_S^2 + \delta_L \|\langle \gamma, \nabla v \rangle\|_S^2$$

Theorem

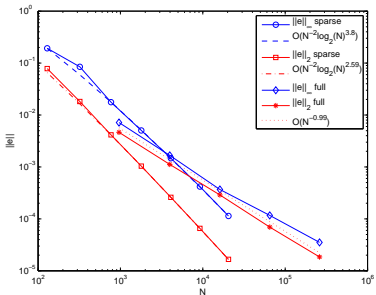
Let δ_L satisfy

$$0 \leq \delta \leq 1/8 \min \left\{ \frac{h_L^2}{(n+p)r^4 |\sqrt{Q}|^2}, 3 \frac{C_2}{C_1^2}, 3 \frac{c_0}{c^2} \right\}.$$

Then $B_\delta(\cdot, \cdot)$ is coercive on $V^L \times V^L$, i.e. $B_\delta(v, v) \geq \frac{1}{4} \|v\|_{SD}^2$.

Convergence rates: European call in Heston model

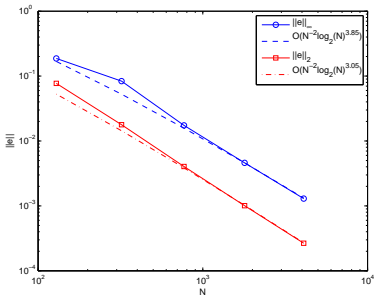
Set $(\alpha, \beta, \rho, m) = (2.5, 0.5, -0.5, 0.025)$, $T = 0.5$, $K = 1$,
 $e_L := U_L(T) - u(T)$. For $q \in \{2, \infty\}$ we find



$$\|e_L\|_q = O(N_L^{-2} (\log N_L)^{c_q}), \quad c_2 \approx 2.59, \quad c_\infty \approx 3.8.$$

Convergence rates: European call in Bates model

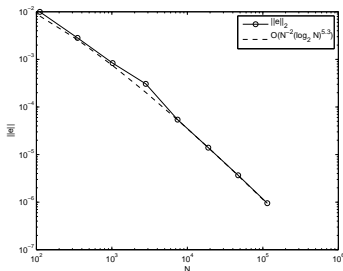
$\mathcal{A}_\nu(u)(x) := -\lambda \int_{\mathbb{R}} [u(x_1 + \xi, x_2) - u(x)] k(\xi) d\xi$. Set
 $(\alpha, \beta, \rho, m, \lambda) = (2.5, 0.5, -0.5, 0.025, 0.5)$, $T = 0.5$, $K = 1$



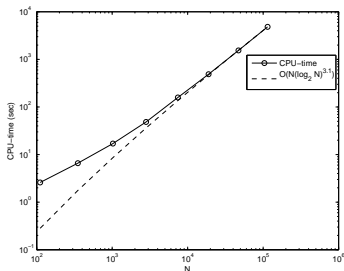
$$\|e_L\|_q = O(N_L^{-2} (\log N_L)^{c_q}), \quad c_2 \approx 3.05, \quad c_\infty \approx 3.85.$$

Linear complexity of pricing methodology

Example: twoscale SV (pure diffusion) model, i.e. $n = 1$, $p = 2$



$$\|e_L\|_2 = O(N_L^{-2}(\log N_L)^{5.3}), \quad t_{\text{CPU}} = O(N_L(\log N_L)^{3.1})^2$$

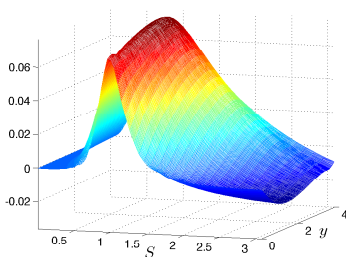
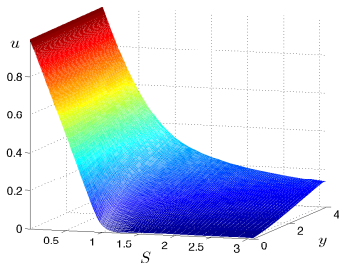


²Matlab, 2GHz CPU, 2G RAM

European put in BNS model

- $T = 0.5, K = 1, r = 0, \lambda = 2.5$, IG(γ, δ)-Lévy kernel

$$k(\xi) = \frac{\delta}{2\sqrt{2\pi}} \xi^{-\frac{3}{2}} (1 + \gamma\xi) e^{-\frac{1}{2}\gamma\xi}, \quad \gamma = 2, \delta = 0.0872.$$



Left: Price for $\rho = -4$.

Right: Difference of prices for $\rho = -0.01$ and $\rho = -4$.

Conclusion

Wavelet FEM for stochastic volatility models

- Derivative pricing
- PIDEs of mixed parabolic/hyperbolic type
- Sparse tensor product spaces
- Stabilization of method due absence of diffusion
- Optimal convergence rates (up to log-terms)
- Linear complexity of numerical scheme (up to log-terms)