

Bergische Universität Wuppertal

Fachbereich Mathematik und Naturwissenschaften

Institute of Mathematical Modelling, Analysis and Computational Mathematics  
(IMACM)

Preprint BUW-IMACM 15/28

Bertram Düring, Christof Heuer

**Essentially high-order compact schemes with application to  
stochastic volatility models on non-uniform grids**

June 24, 2015

<http://www.math.uni-wuppertal.de>

# Essentially high-order compact schemes with application to stochastic volatility models on non-uniform grids

Bertram Düring\*      Christof Heuer†

June 24, 2015

## Abstract

We present high-order compact schemes for a linear second-order parabolic partial differential equation (PDE) with mixed second-order derivative terms in two spatial dimensions. The schemes are applied to option pricing PDE for a family of stochastic volatility models. We use a non-uniform grid with more grid-points around the strike price. The schemes are fourth-order accurate in space and second-order accurate in time for vanishing correlation. In our numerical convergence study we achieve fourth-order accuracy also for non-zero correlation. A combination of Crank-Nicolson and BDF-4 discretisation is applied in time. Numerical examples confirm that a standard, second-order finite difference scheme is significantly outperformed.

## 1 Introduction

We consider the following parabolic partial differential equation for  $u = u(x_1, x_2, t)$  in two spatial dimensions and time,

$$(1) \quad du_\tau + a_1 u_{x_1 x_1} + a_2 u_{x_2 x_2} + b_{12} u_{x_1 x_2} + c_1 u_{x_1} + c_2 u_{x_2} = 0 \quad \text{in } \Omega \times ]0, T] =: Q_T,$$

subject to suitable boundary conditions and initial condition  $u(x_1, x_2, 0) = u_0(x_1, x_2)$  with  $T > 0$  and  $\Omega = [x_{\min}^{(1)}, x_{\max}^{(1)}] \times [x_{\min}^{(2)}, x_{\max}^{(2)}] \subset \mathbb{R}^2$  with  $x_{\min}^{(i)} < x_{\max}^{(i)}$  for  $i = 1, 2$ . The functions  $a_i = a_i(x_1, x_2, \tau) < 0$ ,  $b_{12} = b_{12}(x_1, x_2, \tau)$ ,  $c_i = c_i(x_1, x_2, \tau)$ ,  $d = d(x_1, x_2, \tau)$  map  $Q_T$  to  $\mathbb{R}$ , and  $a_i(\cdot, \tau)$ ,  $b(\cdot, \tau)$ ,  $c_i(\cdot, \tau)$ , and  $d(\cdot, \tau)$  are assumed to be in  $C^2(\Omega)$  and  $u(\cdot, t) \in C^6(\Omega)$  for all  $\tau \in ]0, T]$ . We define a uniform spatial grid  $G$  with step size  $\Delta x_k$  in  $x_k$  direction for  $k = 1, 2$ . Setting  $f = -du_\tau$  and applying a standard, second-order central difference approximation leads to the elliptic problem

$$(2) \quad f = A_0 - \frac{a_1(\Delta x_1)^2}{12} \frac{\partial^4 u}{\partial x_1^4} - \frac{a_2(\Delta x_2)^2}{12} \frac{\partial^4 u}{\partial x_2^4} - \frac{b_{12}(\Delta x_1)^2}{6} \frac{\partial^4 u}{\partial x_1^3 \partial x_2} - \frac{b_{12}(\Delta x_2)^2}{6} \frac{\partial^4 u}{\partial x_1 \partial x_2^3} - \frac{c_1(\Delta x_1)^2}{6} \frac{\partial^3 u}{\partial x_1^3} - \frac{c_2(\Delta x_2)^2}{6} \frac{\partial^3 u}{\partial x_2^3} + \varepsilon,$$

with  $A_0 := a_1 D_1^c D_1^c U_{i_1, i_2} + a_2 D_2^c D_2^c U_{i_1, i_2} + b_{12} D_1^c D_2^c U_{i_1, i_2} + c_1 D_1^c U_{i_1, i_2} + c_2 D_2^c U_{i_1, i_2}$ , where  $D_k^c$  denotes the central difference operator in  $x_k$  direction, and  $\varepsilon \in \mathcal{O}(h^4)$  if  $\Delta x_k \in \mathcal{O}(h)$  for  $h > 0$ . We call a finite difference scheme high-order compact (HOC) if its consistency error is of order  $\mathcal{O}(h^4)$  for  $\Delta x_1, \Delta x_2 \in \mathcal{O}(h)$  for  $h > 0$ , and it uses only points on the compact stencil,  $U_{k,p}$  with  $k \in \{i_1 - 1, i_1, i_1 + 1\}$  and  $p \in \{i_2 - 1, i_2, i_2 + 1\}$ , to approximate the solution at  $(x_{i_1}, x_{i_2}) \in \overset{\circ}{G}$ .

## 2 Auxiliary relations for higher derivatives

Our aim is to replace the third- and fourth-order derivatives in (2) which are multiplied by second-order terms by equivalent expressions which can be approximated with second order on the compact

\*Email: b.during@sussex.ac.uk, Department of Mathematics, University of Sussex, Pevensey II, Brighton, BN1 9QH, United Kingdom

†Email: cheuer@uni-wuppertal.de, Lehrstuhl für Angewandte Mathematik und Numerische Analysis, Fachbereich C, Bergische Universität Wuppertal, Gaußstr. 20, 42119 Wuppertal, Germany

stencil. Indeed, if we differentiate (1) (using  $f = -du_\tau$ ) once with respect to  $x_k$  ( $k = 1, 2$ ), we obtain relations

$$(3) \quad \frac{\partial^3 u}{\partial x_1^3} = A_1, \quad \frac{\partial^3 u}{\partial x_2^3} = A_2,$$

where we can discretise  $A_i$  with second order on the compact stencil using the central difference operator. Analogously, we obtain

$$(4) \quad \begin{aligned} \frac{\partial^4 u}{\partial x_1^4} = B_1 - \frac{b_{12}}{a_1} \frac{\partial^4 u}{\partial x_1^3 \partial x_2} &\iff \frac{\partial^4 u}{\partial x_1^3 \partial x_2} = \frac{a_1}{b_{12}} B_1 - \frac{a_1}{b_{12}} \frac{\partial^4 u}{\partial x_1^4}, \\ \frac{\partial^4 u}{\partial x_2^4} = B_2 - \frac{b_{12}}{a_2} \frac{\partial^4 u}{\partial x_1 \partial x_2^3} &\iff \frac{\partial^4 u}{\partial x_1 \partial x_2^3} = \frac{a_2}{b_{12}} B_2 - \frac{a_2}{b_{12}} \frac{\partial^4 u}{\partial x_2^4}, \\ \frac{\partial^4 u}{\partial x_1^3 \partial x_2} = C_1 - \frac{a_2}{a_1} \frac{\partial^4 u}{\partial x_1 \partial x_2^3} &\iff \frac{\partial^4 u}{\partial x_1 \partial x_2^3} = C_2 - \frac{a_1}{a_2} \frac{\partial^4 u}{\partial x_1^3 \partial x_2}, \end{aligned}$$

where we can approximate  $B_k$  and  $C_k$  with second order on the compact stencil using the central difference operator. A detailed derivation can be found in [4].

### 3 Derivation of high-order compact schemes

In general it is not possible to obtain a HOC scheme for (1), since there are four fourth-order derivatives in (2), but only three auxiliary equations for these in (4). Hence, we propose four different versions of the numerical schemes, where only one of the fourth-order derivatives in (2) is left as a second-order remainder term. Using (3) and (4) in (2) we obtain as *Version 1* scheme

$$(5) \quad \begin{aligned} f = A_0 - \frac{c_1(\Delta x_1)^2}{6} A_1 - \frac{c_2(\Delta x_2)^2}{6} A_2 - \frac{a_2(\Delta x_2)^2}{12} B_2 - \frac{b_{12}(\Delta x_2)^2}{12} C_2 \\ - \frac{a_1(2a_2(\Delta x_1)^2 - a_1(\Delta x_2)^2)}{12a_2} B_1 + \frac{a_1(a_2(\Delta x_1)^2 - a_1(\Delta x_2)^2)}{12a_2} \frac{\partial^4 u}{\partial x_1^4} + \varepsilon, \end{aligned}$$

as *Version 2* scheme

$$(6) \quad \begin{aligned} f = A_0 - \frac{c_1(\Delta x_1)^2}{6} A_1 - \frac{c_2(\Delta x_2)^2}{6} A_2 - \frac{a_1(\Delta x_1)^2}{12} B_1 - \frac{b_{12}(\Delta x_1)^2}{12} C_1 \\ - \frac{a_2(2a_1(\Delta x_2)^2 - a_2(\Delta x_1)^2)}{12a_1} B_2 + \frac{a_2(a_1(\Delta x_2)^2 - a_2(\Delta x_1)^2)}{12a_1} \frac{\partial^4 u}{\partial x_2^4} + \varepsilon, \end{aligned}$$

as *Version 3* scheme

$$(7) \quad \begin{aligned} f = A_0 - \frac{c_1(\Delta x_1)^2}{6} A_1 - \frac{c_2(\Delta x_2)^2}{6} A_2 - \frac{a_1(\Delta x_1)^2}{12} B_1 - \frac{a_2(\Delta x_2)^2}{12} B_2 \\ - \frac{b_{12}(\Delta x_2)^2}{12} C_2 + \frac{b_{12}(a_1(\Delta x_2)^2 - a_2(\Delta x_1)^2)}{12a_2} \frac{\partial^4 u}{\partial x_1^3 \partial x_2} + \varepsilon, \end{aligned}$$

and, finally, as *Version 4* scheme

$$(8) \quad \begin{aligned} f = A_0 - \frac{c_1(\Delta x_1)^2}{6} A_1 - \frac{c_2(\Delta x_2)^2}{6} A_2 - \frac{a_1(\Delta x_1)^2}{12} B_1 - \frac{a_2(\Delta x_2)^2}{12} B_2 \\ - \frac{b_{12}(\Delta x_1)^2}{12} C_1 + \frac{b_{12}(a_2(\Delta x_1)^2 - a_1(\Delta x_2)^2)}{12a_1} \frac{\partial^4 u}{\partial x_1 \partial x_2^3} + \varepsilon. \end{aligned}$$

Employing the central difference operator with  $\Delta x = \Delta y = h$  for  $h > 0$  to discretise  $A_i$ ,  $B_i$ ,  $C_i$ , in (5)–(8) and neglecting the remaining lower-order term leads to four semi-discrete (in space) schemes. A more detailed description of this approach can be found in [4]. When  $a_1 \equiv a_2$  or  $b_{12} \equiv 0$  these schemes are fourth-order consistent in space, otherwise second-order.

In time, we apply the implicit BDF4 method on an equidistant time grid with stepsize  $k \in \mathcal{O}(h)$ . The necessary starting values are obtained using a Crank-Nicolson time discretisation,

where we subdivide the first timesteps with a step size  $k' \in \mathcal{O}(h^2)$  to ensure the fourth-order time discretisation in terms of  $h$ .

With additional information on the solution of (1) even better results are possible. If the specific combination of pre-factors in (1) and the higher derivatives in the second-order terms is sufficiently small, the second-order term dominates the computational error only for very small step-sizes  $h$ . Before this error term becomes dominant one can observe a fourth-order numerical convergence. In this case we call the scheme essentially high-order compact (EHOC).

## 4 Application to option pricing

In this section we apply our numerical schemes to an option pricing PDE in a family of stochastic volatility models, with a generalised square root process for the variance with nonlinear drift term,

$$dS_t = \mu S_t dt + \sqrt{v_t} S_t dW_t^{(1)}, \quad dv_t = \kappa v_t^\alpha (\theta - v_t) dt + \sigma \sqrt{v_t} dW_t^{(2)},$$

with  $\alpha \geq 0$ , a correlated, two-dimensional standard Brownian motion,  $dW_t^{(1)} dW_t^{(2)} = \rho dt$ , as well as drift  $\mu \in \mathbb{R}$  of the stock price  $S$ , long run mean  $\theta > 0$ , mean reversion speed  $\kappa > 0$ , and volatility of volatility  $\sigma > 0$ . For  $\alpha = 0$  one obtains the standard Heston model, for  $\alpha = 1$  the SQRN model, see [1]. Using Itô's lemma and standard arbitrage arguments, the option price  $V = V(S, v, t)$  solves

$$(9) \quad \frac{\partial V}{\partial t} + \frac{v S^2}{2} \frac{\partial^2 V}{\partial S^2} + \rho \sigma v S \frac{\partial^2 V}{\partial S \partial v} + \frac{\sigma^2 v}{2} \frac{\partial^2 V}{\partial v^2} + r S \frac{\partial V}{\partial S} + \kappa v^\alpha (\theta - v) \frac{\partial V}{\partial v} - rV = 0,$$

where  $S, \sigma > 0$  and  $t \in [0, T[$  with  $T > 0$ . For a European Put with exercise price  $K$  we have the final condition  $V(S, T) = \max(K - S, 0)$ . The transformations  $\tau = T - t$ ,  $u = e^{r\tau} V/K$ ,  $\hat{S} = \ln(S/K)$ ,  $y = v/\sigma$  as well as  $\hat{S} = \varphi(x)$  [2], lead to

$$\varphi_x^3 u_\tau + \frac{\sigma y}{2} [\varphi_x u_{xx} + \varphi_x^3 u_{yy}] - \rho \sigma y \varphi_x^2 u_{xy} + \left[ \frac{\sigma y \varphi_{xx}}{2} + \left( \frac{\sigma y}{2} - r \right) \varphi_x^2 \right] u_x - \kappa \sigma^\alpha y^\alpha \frac{\theta - \sigma y}{\sigma} \varphi_x^3 u_y = 0,$$

with initial condition  $u(x, y, 0) = \max(1 - e^{\varphi(x)}, 0)$ . The function  $\varphi$  is considered to be four times differentiable and strictly monotone. It is chosen in such a way that grid points are concentrated around the exercise price  $K$  in the  $S-v$  plane when using a uniform grid in the  $x-y$  plane.

Dirichlet boundary conditions are imposed at  $x = x_{\min}$  and  $x = x_{\max}$  similarly as in [2],

$$u(x_{\min}, y, \tau) = u(x_{\min}, y, 0), \quad u(x_{\max}, y, \tau) = u(x_{\max}, y, 0) \quad \forall \tau \in [0, \tau_{\max}] \quad \forall y \in [y_{\min}, y_{\max}].$$

At the boundaries  $y = y_{\min}$  and  $y = y_{\max}$  we employ the discretisation of the interior spatial domain and extrapolate the resulting ghost-points using

$$U_{i,-1} = 3U_{i,0} - 3U_{i,1} + U_{i,2} + \mathcal{O}(h^3), \quad U_{i,M+1} = 4U_{i,M} - 6U_{i,M-1} + 4U_{i,M-2} - U_{i,M-3} + \mathcal{O}(h^4),$$

for  $i = 0, \dots, N$ . Third-order extrapolation is sufficient here to ensure overall fourth-order convergence [3].

## 5 Numerical experiments

We employ the function  $\varphi(x) = \sinh(c_2 x + c_1(1-x))/\zeta$ , where  $c_1 = \text{asinh}(\zeta \hat{S}_{\min})$ ,  $c_2 = \text{asinh}(\zeta \hat{S}_{\max})$  and  $\zeta > 0$ . We use  $\kappa = 1.1$ ,  $\theta = 0.2$ ,  $v = 0.3$ ,  $r = 0.05$ ,  $K = 100$ ,  $T = 0.25$ ,  $v_{\min} = 0.1$ ,  $v_{\max} = 0.3$ ,  $S_{\min} = 1.5$ ,  $S_{\max} = 250$ ,  $\rho = 0, -0.4$  and  $\zeta = 7.5$ . Hence,  $x_{\max} - x_{\min} = y_{\max} - y_{\min} = 1$ . For the Crank-Nicolson method we use  $k'/h^2 = 0.4$ , for the BDF4 method  $k/h = 0.1$ . We smooth the initial condition according to [5], so that the smoothed initial condition tends towards the original initial condition for  $h \rightarrow 0$ . We neglect the case  $\alpha = 0$  (Heston model), since a numerical study of that case has been performed in [2]. In the numerical convergence plots we use a reference solution  $U_{\text{ref}}$  on a fine grid ( $h = 1/320$ ) and report the absolute  $l^2$ -error compared to  $U_{\text{ref}}$ . The numerical convergence order is computed from the slope of the linear least square fit of the points in the log-log plot.

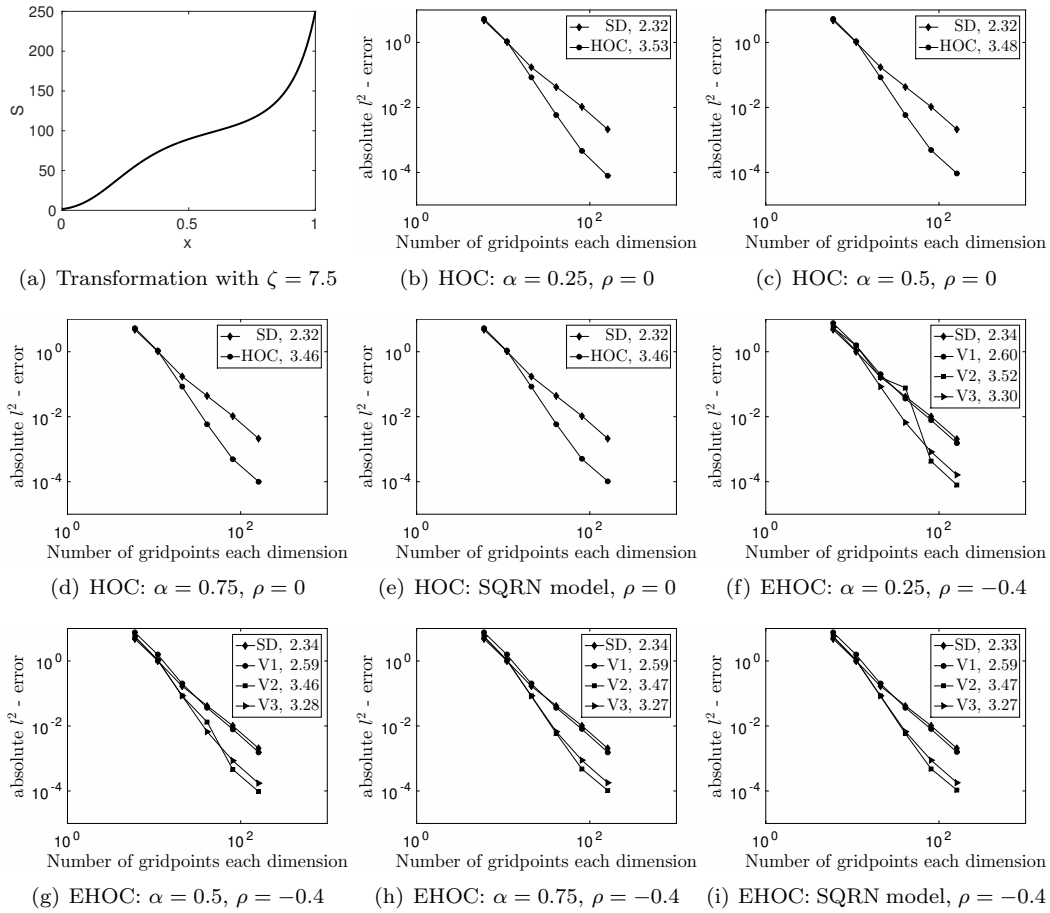


Figure 1: Transformation of the spatial grid and numerical convergence plots.

Figure 1(a) shows the transformation from  $x$  to  $S$ . The transformation focuses on the region around the strike price. Figures 1(b), 1(c), 1(d) and 1(e) show that the HOC schemes lead to a numerical convergence order of about 3.5, whereas the standard, second-order central difference discretisation (SD) leads to convergence orders of about 2.3, in the case of vanishing correlation. In all cases with non-vanishing correlation ( $\rho \neq 0$ ) we observe only slightly improved convergence for Version 1 (V1) when comparing it to the standard discretisation. Version 2 (V2) and Version 3 (V3), however, lead to similar convergence orders as the HOC scheme, even for non-vanishing correlation. Results of Version 4 are not shown as this scheme shows instable behaviour in this example.

In summary, we obtain high-order compact schemes for vanishing correlation and achieve high-order convergence also for non-vanishing correlation for the family (9) of stochastic volatility model. A standard, second-order discretisation is significantly outperformed in all cases.

## Acknowledgment

This research was supported by the European Union in the FP7-PEOPLE-2012-ITN Program under Grant Agreement Number 304617 (FP7 Marie Curie Action, Project Multi-ITN STRIKE – Novel Methods in Computational Finance).

## References

- [1] P. Christoffersen, K. Jacobs, and K. Mimouni. Models for S&P500 dynamics: evidence from realized volatility, daily returns, and option prices. *Rev. Financ. Stud.*, 23(8):3141–3189, 2010.

- [2] B. Düring, M. Fournié, and C. Heuer. High-order compact finite difference schemes for option pricing in stochastic volatility models on non-uniform grids. *J. Comput. Appl. Math.*, 271(18):247–266, 2014.
- [3] B. Gustafsson. The convergence rate for difference approximations to general mixed initial-boundary value problems. *SIAM J. Numer. Anal.*, 18(2):179–190, 1981.
- [4] C. Heuer. *High-order compact finite difference schemes for parabolic partial differential equations with mixed derivative terms and applications in computational finance*. PhD thesis, University of Sussex, August 2014.
- [5] H.O. Kreiss, V. Thomee, and O. Widlund. Smoothing of initial data and rates of convergence for parabolic difference equations. *Commun. Pure Appl. Math.*, 23:241–259, 1970.