

FDM for Cheyette Local Volatility

November 2021

Contents

1 Implementation	1
1.1 Dynamics	1
1.2 Discretization	1
1.3 Time discretization and operator splitting	2
1.4 Boundary conditions in equation (10a)	3

1 Implementation

1.1 Dynamics

We consider the Cheyette model:

$$dx_t = (y_t - \kappa(t)x(t))dt + \sigma(t, x_t, y_t)dW_t, \quad (1a)$$

$$dy_t = (\sigma^2(t, x_t, y_t) - 2\kappa(t)y_t)dt. \quad (1b)$$

Let $V(t, x, y)$ the the time t value of a derivative with terminal payoff $V(x_T, y_T)$:

$$V(t, x, y) = \mathbf{E} \left(e^{-\int_t^T r(s)ds} V(x_T, y_T) \mid x(t) = x, y(t) = y \right). \quad (2)$$

Then $V(t, x, y)$ satisfies the PDE:

$$\partial_t V + (L_x + L_y)V = 0, \quad (3a)$$

$$L_x = (y - \kappa(t)x)\partial_x + \frac{1}{2}\sigma^2(t, x, y)\partial_x^2 - (x + f(0, t)), \quad (3b)$$

$$L_y = (\sigma^2(t, x, y) - 2\kappa(t)y)\partial_y. \quad (3c)$$

1.2 Discretization

We will solve the pde on the rectangular grid $[\underline{M}_1, \overline{M}_1] \times [\underline{M}_2, \overline{M}_2]$. We consider a uniform grid:

$$\begin{aligned} \Gamma_{h_1, h_2} = \{ (x^{j_1}, y^{j_2}) : x^{j_1} = \underline{M}_1 + j_1 h_1, \quad y^{j_2} = \underline{M}_2 + j_2 h_2, \\ j_1 = 0, \dots, m_1 + 1, \quad j_2 = 0, \dots, m_2 + 1 \}. \end{aligned} \quad (4)$$

We put:

$$V_{j_1, j_2}(t) = V(t, x^{j_1}, y^{j_2}), \quad (5a)$$

$$\delta_x V_{j_1, j_2}(t) = \frac{V_{j_1+1, j_2}(t) - V_{j_1-1, j_2}(t)}{2h_1}, \quad (5b)$$

$$\delta_y V_{j_1, j_2}(t) = \frac{V_{j_1, j_2+1}(t) - V_{j_1, j_2-1}(t)}{2h_2}, \quad (5c)$$

$$\delta_{xx} V_{j_1, j_2}(t) = \frac{V_{j_1+1, j_2}(t) - 2V_{j_1, j_2}(t) + V_{j_1-1, j_2}(t)}{h_1^2}. \quad (5d)$$

We pass from the differential operators L_x, L_y to finite difference operators

$$\widehat{L}_x(t) = (y - k(t)x)\delta_x + \frac{1}{2}\sigma^2(t, x, y)\delta_x^2 - (x + f(0, t)), \quad (6a)$$

$$\widehat{L}_y(t) = (\sigma^2(t, x, y) - 2\kappa(t)y)\delta_y, \quad (6b)$$

acting on functions on the grid Γ_{h_1, h_2} . For a moment we omit the definitions on the boundary points. We get a second order accurate approximation of the operator $L_x + L_y$ on the grid:

$$(L_x + L_y)V(t, x, y) = (\widehat{L}_x + \widehat{L}_y)V(t, x, y) + O(h_1^2 + h_2^2). \quad (6c)$$

1.3 Time discretization and operator splitting

The pricing PDE has no mixed derivatives and can be solved using a splitting method. We use a uniform time grid $\{t_i\}$ with step h_t . On the interval $[t_i, t_{i+1}]$ we replace the time derivative with the first difference and we replace the operators L_x, L_y by their discretizations $\widehat{L}_x(t_i), \widehat{L}_y(t_i)$ evaluated at the middle of the interval $\widehat{t}_i = (t_i + t_{i+1})/2$, as in the Crank-Nicolson method. After regrouping the terms, we get:

$$\begin{aligned} \left(1 - h_t \frac{1}{2}(\widehat{L}_x + \widehat{L}_y)\right) V_{j_1, j_2}(t_i) &= \left(1 + h_t \frac{1}{2}(\widehat{L}_x + \widehat{L}_y)\right) V_{j_1, j_2}(t_{i+1}) \\ &+ O(h_t(h_1^2 + h_2^2 + h_t^2)). \end{aligned} \quad (7)$$

Note that up to an error of order $O(h_t(h_1^2 + h_2^2 + h_t^2))$:

$$\left(1 \pm h_t \frac{1}{2}(\widehat{L}_x + \widehat{L}_y)\right) \approx \left(1 \pm h_t \frac{1}{2}\widehat{L}_x\right) \left(1 \pm h_t \frac{1}{2}\widehat{L}_y\right). \quad (8a)$$

This leads to the following equality, up to $O(h_t(h_1^2 + h_2^2 + h_t^2))$ error:

$$\left(1 - h_t \frac{1}{2}\widehat{L}_x\right) \left(1 - h_t \frac{1}{2}\widehat{L}_y\right) V_{j_1, j_2}(t_i) \approx \left(1 + h_t \frac{1}{2}\widehat{L}_x\right) \left(1 + h_t \frac{1}{2}\widehat{L}_y\right) V_{j_1, j_2}(t_{i+1}). \quad (9)$$

Using basic algebra¹, we can split this second order equation into two consecutive first order equations, forgetting about the error terms:

$$\left(1 - h_t \frac{1}{2} \widehat{L}_x\right) U_{j_1, j_2} = \left(1 + h_t \frac{1}{2} \widehat{L}_y\right) V_{j_1, j_2}(t_{i+1}), \quad (10a)$$

$$\left(1 + h_t \frac{1}{2} \widehat{L}_y\right) V_{j_1, j_2}(t_i) = \left(1 - h_t \frac{1}{2} \widehat{L}_x\right) U_{j_1, j_2}, \quad (10b)$$

with intermediate variables U_{j_1, j_2} . We first determine U_{j_1, j_2} from $V_{j_1, j_2}(t_{i+1})$ using the first equation indexed by j_2 , and then determine V_{j_1, j_2} from the second equation indexed by j_1 .

This approach reduces the original second order difference system with a general sparse matrix to a sequence of two first order difference equations with tridiagonal matrices with computational cost $O(m_1 m_2)$ instead of $O((m_1 m_2)^{5/4})$. The resulting method is called after Peaceman and Rachford. Besides, the first equation is perfectly parallelizable in j_2 and the second one if perfectly parallelizable in j_1 reducing the optimal parallel complexity to $O(m_1 + m_2)$.

1.4 Boundary conditions in equation (10a)

To impose boundary conditions in system (10a) we will assume that:

$$U \approx V(\widehat{t}_i), \quad \widehat{t}_i = \frac{t_i + t_{i+1}}{2}. \quad (11)$$

A better approximation is possible.

Dirichlet conditions. Dirichlet conditions in the x variable are:

$$V(t, x_0, y) = \underline{V}_1(t, y), \quad V(t, x_{m_1+1}, y) = \overline{V}_1(t, y), \quad (12)$$

for given functions $\underline{V}_1(t, y)$, $\overline{V}_1(t, y)$. To impose the lower Dirichlet boundary condition we need to do the following modifications to system (10a):

1. Replace the first row in the system matrix by $(1, 0, \dots, 0)$.
2. Replace the first element in the right-hand-side by $\underline{V}(\widehat{t}_i, y)$.

¹Consider a formal equation of the form:

$$(1 - A)(1 - B)x = (1 + A)(1 + B)b$$

for the unknown x where A, B are abstract linear operators. Assuming that $1 - A$ is invertible and multiplying both sides by $(1 - A)^{-1}$, we can write:

$$(1 - B)x = (1 + A)(1 - A)^{-1}(1 + B)b,$$

Now, setting $u = (1 - A)^{-1}(1 + B)b$, we get two equations:

$$\begin{aligned} (1 - B)x &= (1 + A)u, \\ (1 - A)u &= (1 + B)b. \end{aligned}$$

In a similar way, to impose the upper Dirichlet boundary condition, the following modifications are made to system (10a):

1. Replace the last row in the system matrix by $(0, \dots, 0, 1)$.
2. Replace the last element in the right-hand-side by $\bar{V}(\hat{t}_i, y)$.

A possible choice for functions $\underline{V}_1(t, y)$ and $\bar{V}_1(t, y)$ is the inner value of the product, assuming zero volatility.