# Local Volatility Model with Stochastic Rates

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# Kolmogorov Forward & Backward Equation

Multidimensional SDE

$$dX_t = A(t, X_t)dt + B(t, X_t)dW_t, \quad \rho dt = dW_t dW_t'$$

- $p(t, x | s, \alpha)$  is the transition probability density function having  $X_t = x$  at t given  $X_s = \alpha$  at s for s < t
- Forward PDE (a.k.a Fokker-Planck equation)

$$\frac{\partial p}{\partial t} + \sum_{i=1}^{m} \frac{\partial (A_i p)}{\partial x_i} - \frac{1}{2} \sum_{i,j=1}^{m} \frac{\partial^2 (\Sigma_{ij} p)}{\partial x_i \partial x_j} = 0, \quad \Sigma = B \rho B'$$

$$\lim_{t \to s} p(t, x | s, \alpha) = \delta(x - \alpha)$$

Backward PDE

$$\frac{\partial p}{\partial t} + \sum_{i=1}^{m} A_i \frac{\partial p}{\partial x_i} + \frac{1}{2} \sum_{i,j=1}^{m} \sum_{i,j=1}^{m} \sum_{i,j=1}^{m} \frac{\partial^2 p}{\partial x_i \partial x_j} = 0$$

$$\lim_{t \to T} p(T, \beta | t, x) = \delta(x - \beta)$$

#### General Dupire Local Vol – Part 1

• Spot dynamics, volatility  $\eta_t$  takes a general form

$$\frac{dX_t}{X_t} = \mu_t dt + \eta_t d\widetilde{W}_t, \quad \mu_t = r_t - q_t$$

European call price

$$\mathcal{C}_{t,K} = \mathbb{E}_{s} \left[ D_{s,t} (X_{t} - K)^{+} \right] = \mathbb{E}_{s} \left[ D_{s,t} (X_{t} - K) \Theta(X_{t} - K) \right]$$

$$\frac{\partial \mathcal{C}_{t,K}}{\partial K} = -\mathbb{E}_{s} \left[ D_{s,t} \Theta(X_{t} - K) \right]$$

$$\frac{\partial^{2} \mathcal{C}_{t,K}}{\partial K^{2}} = \mathbb{E}_{s} \left[ D_{s,t} \delta(X_{t} - K) \right] = \mathbb{E}_{s} \left[ D_{s,t} | X_{t} = K \right] \mathbb{E}_{s} \left[ \delta(X_{t} - K) \right]$$

The last equality comes from Bayes' rule

Heaviside step function and Dirac delta function

$$\Theta(x) = \begin{cases} 0, & x < 0 \\ 1/2, & x = 0, \\ 1, & x > 0 \end{cases} \quad \Theta(x) = \int_{-\infty}^{x} \delta(u) du, \quad \delta(x) = \frac{d\Theta(x)}{dx}$$
$$\delta(x) = \begin{cases} \infty, & x = 0 \\ 0, & x \neq 0 \end{cases} \quad \text{subject to} \quad \int_{-\infty}^{\infty} \delta(u) du = 1$$

Analogous to the CDF and PDF of a normal distribution with zero mean and infinitesimal variance

# General Dupire Local Vol – Part 2 (Ch 3.2)

Dynamics of payoff function (not a smooth function; applying Ito-Tanaka formula)

$$d(D_{s,t}(X_t - K)^+) = -r_t D_{s,t}(X_t - K)^+ dt + D_{s,t} \Theta(X_t - K) dX_t + \frac{1}{2} D_{s,t} \delta(X_t - K) (dX_t)^2$$

Take expectation of both sides

$$\begin{split} &\frac{\partial \mathcal{C}_{t,K}}{\partial t} = \mathbb{E}_{s} \left[ \frac{d \left( D_{s,t} (X_{t} - K)^{+} \right)}{dt} \right] \\ &= \mathbb{E}_{s} \left[ D_{s,t} \left( -r_{t} (X_{t} - K) \Theta(X_{t} - K) + \mu_{t} X_{t} \Theta(X_{t} - K) + \frac{1}{2} X_{t}^{2} \eta_{t}^{2} \delta(X_{t} - K) \right) \right] \\ &= \mathbb{E}_{s} \left[ D_{s,t} \left( r_{t} K - q_{t} X_{t} \right) \Theta(X_{t} - K) \right] + \frac{1}{2} \mathbb{E}_{s} \left[ D_{s,t} X_{t}^{2} \eta_{t}^{2} \delta(X_{t} - K) \right] \\ &= K \mathbb{E}_{s} \left[ D_{s,t} r_{t} \Theta(X_{t} - K) \right] - \mathbb{E}_{s} \left[ D_{s,t} q_{t} X_{t} \Theta(X_{t} - K) \right] + \frac{1}{2} K^{2} \mathbb{E}_{s} \left[ D_{s,t} \eta_{t}^{2} | X_{t} = K \right] \mathbb{E}_{s} \left[ \delta(X_{t} - K) \right] \end{split}$$

Finally

$$\frac{\mathbb{E}_{S}[D_{S,t}\eta_{t}^{2}|X_{t}=K]}{\mathbb{E}_{S}[D_{S,t}|X_{t}=K]} = \frac{\frac{\partial \mathcal{C}_{t,K}}{\partial t} - K\mathbb{E}_{S}[D_{S,t}r_{t}\Theta(X_{t}-K)] + \mathbb{E}_{S}[D_{S,t}q_{t}X_{t}\Theta(X_{t}-K)]}{\frac{1}{2}K^{2}\frac{\partial^{2}\mathcal{C}_{t,K}}{\partial K^{2}}}$$

• If we write the volatility term  $\eta_t$  as a pure local volatility  $g(t,X_t)$ , which is a deterministic function of  $X_t$ 

$$g_{t,K}^2 = \frac{\mathbb{E}_s[D_{s,t}\eta_t^2|X_t = K]}{\mathbb{E}_s[D_{s,t}|X_t = K]} = \frac{\frac{\partial \mathcal{C}_{t,K}}{\partial t} - K\mathbb{E}_s[D_{s,t}r_t\Theta(X_t - K)] + \mathbb{E}_s[D_{s,t}q_tX_t\Theta(X_t - K)]}{\frac{1}{2}K^2\frac{\partial^2 \mathcal{C}_{t,K}}{\partial K^2}}$$

#### Relationship to Classic Dupire Local Vol (Ch 4.2 & 4.3)

• Assuming deterministic rates  $ar{r}_t$  and  $ar{q}_t$ , we get the Classic Dupire local vol

$$\xi_{t,K}^2 = \mathbb{E}_{s}[\eta_t^2 | X_t = K] = \frac{\frac{\partial \mathcal{C}_{t,K}}{\partial t} + \overline{\mu}_t K \frac{\partial \mathcal{C}_{t,K}}{\partial K} + \overline{q}_t \mathcal{C}_{t,K}}{\frac{1}{2} K^2 \frac{\partial^2 \mathcal{C}_{t,K}}{\partial K^2}}$$

Dupire local variance = Conditional expectation of the (stochastic) variance

• Difference  $\beta_{t,K}$  between the general Dupire local volatility  $g(t,X_t)$  and the classic Dupire local volatility  $\xi(t,K)$ 

$$\beta_{t,K} = g_{t,K}^2 - \xi_{t,K}^2 = \frac{\mathbb{E}_s[D_{s,t}(q_t - \bar{q}_t)X_t\Theta(X_t - K)] - K\mathbb{E}_s[D_{s,t}(r_t - \bar{r}_t)\Theta(X_t - K)]}{\frac{1}{2}K^2\frac{\partial^2 C_{t,K}}{\partial K^2}}$$

the spread  $\beta_{t,K}$  becomes zero if the rates  $r_t$  and  $q_t$  are deterministic

- The general Dupire local vol is constructed on a time and strike 2D grid, which can be calibrated by estimating the two expectations along with the readily available classic Dupire local vols
- Next, we assume Hull-White model for the stochastic rates and general Dupire local vol for the spot process, and attempt to estimate the two expectations numerically

# Local Vol Model with Stochastic Rates (Ch 2)

• 3-Factor model: Hull White model for rates, general local vol for spot

$$\begin{split} r_t &= f_{s,t} + \int_s^t b_{u,t} \beta_{u,t} du + x_t \\ x_t &= \int_s^t \beta_{u,t} dW_{1,u} \,, \qquad dx_t = -\lambda_t x_t dt + \sigma_t dW_{1,t} \\ \hat{r}_t &= \hat{f}_{s,t} + \int_s^t \hat{b}_{u,t} \hat{\beta}_{u,t} du + \hat{x}_t \\ \hat{x}_t &= -\int_s^t \rho_u^{23} \hat{\beta}_{u,t} g_{u,y} du + \int_s^t \hat{\beta}_{u,t} dW_{2,u} \,, \qquad d\hat{x}_t = -\left(\rho_t^{23} \hat{\sigma}_t g_{t,y} + \hat{\lambda}_t \hat{x}_t\right) dt + \hat{\sigma}_t dW_{2,t} \\ dy_t &= \left(r_t - \hat{r}_t - \frac{1}{2} g_{t,y}^2\right) dt + g_{t,y} dW_{3,t}, \qquad dW_{i,t} dW_{j,t} = \rho_t^{ij} dt \end{split}$$

- The  $x_t$ ,  $\hat{x}_t$  and  $y_t$  are the state variables
- Define discounted transition probability density function

$$h(t,x,\hat{x},y|s,x_s,\hat{x}_s,y_s) = D_{s,t}p(t,x,\hat{x},y|s,x_s,\hat{x}_s,y_s), \qquad D_{s,t} = \exp\left(-\int_s^t r_u du\right)$$

 $D_{s,t}$  is not a function of any of the state variables

• Next, we briefly review the Hull-White model and then derive the forward PDE, i.e. *Fokker-Planck* equation, for the model

### 1-Factor Hull-White Model (Ch 2)

1-Factor Hull White model for both domestic and foreign rates

$$r_t = f_{s,t} + \int_s^t b_{u,t} \beta_{u,t} du + x_t, \quad x_t = \int_s^t \beta_{u,t} dW_u, \quad dx_t = -\lambda x_t dt + \sigma_t dW_t$$
$$\beta_{u,t} = e^{-\lambda(t-u)} \sigma_u, \quad b_{u,t} = \int_u^t \beta_{u,v} dv = \frac{1 - e^{-\lambda(t-u)}}{\lambda} \sigma_u$$

 $eta_{u,t}$  is the volatility of the instantaneous forward rate  $f_{u,t}$ 

 $b_{u,t}$  is the volatility of the zero coupon bond  $P_{u,t}$ 

Writing the model in the most familiar form, it will be

$$dr_t = \lambda(\theta_t - r_t)dt + \sigma_t dW_t, \qquad \theta_t = f_{s,t} + \frac{1}{\lambda} \frac{\partial f_{s,t}}{\partial t} + \frac{1}{\lambda} \int_s^t \beta_{u,t}^2 du$$

In practical applications, it usually takes a time-invariant mean reversion rate  $\lambda$  along with a (deterministic) piecewise constant short rate volatility  $\sigma_t$ 

• The model can calibrate to caplets or to co-terminal swaptions (via Jamshidian decomposition) knowing that a forward starting zero coupon bond under T-forward measure is a lognormal martingale (usually fix  $\lambda$  while calibrating  $\sigma_t$ )

$$P_{t,T,V} = \frac{P_{t,V}}{P_{t,T}} = \frac{P_{s,V}}{P_{s,T}} \exp\left(-\int_{s}^{t} \frac{b_{u,V}^{2} - b_{u,T}^{2}}{2} du - \int_{s}^{t} (b_{u,V} - b_{u,T}) dW_{u}\right)$$

$$= \frac{P_{s,V}}{P_{s,T}} \exp\left(-\int_{s}^{t} \frac{(b_{u,V} - b_{u,T})^{2}}{2} du - \int_{s}^{t} (b_{u,V} - b_{u,T}) dW_{u}^{T}\right), \quad dW_{u}^{T} = dW_{u} + b_{u,T} du$$

### Calibration by Forward PDE (Ch 3.1 & Ch 4.1.2)

Forward PDE

$$\begin{split} \frac{\partial h}{\partial t} &= -rh + \frac{\partial (\lambda_t x_t h)}{\partial x} + \frac{\partial \left( (\rho_t^{23} \widehat{\sigma}_t g_{t,y} + \widehat{\lambda}_t \widehat{x}_t) h \right)}{\partial \widehat{x}} - \frac{\partial \left( (r_t - \widehat{r}_t - \frac{1}{2} g_{t,y}^2) h \right)}{\partial y} + \frac{1}{2} \frac{\partial^2 (\sigma_t^2 h)}{\partial x^2} + \frac{1}{2} \frac{\partial^2 (\widehat{\sigma}_t^2 h)}{\partial \widehat{x}^2} \\ &\quad + \frac{1}{2} \frac{\partial^2 (g_{t,y}^2 h)}{\partial y^2} + \frac{\partial^2 (\rho_t^{12} \sigma_t \widehat{\sigma}_t h)}{\partial x \partial \widehat{x}} + \frac{\partial^2 (\rho_t^{13} \sigma_t g_{t,y} h)}{\partial x \partial y} + \frac{\partial^2 (\rho_t^{23} \widehat{\sigma}_t g_{t,y} h)}{\partial \widehat{x} \partial y} \\ &\lim_{t \to s} h(t, x, \widehat{x}, y | s, x_s, \widehat{x}_s, y_s) = \delta(x - x_s, \widehat{x} - \widehat{x}_s, y - y_s) \end{split}$$

- The 3D PDE can be solved by Alternating Direction Implicit (ADI) method
- Steps of calibrating general Dupire local vol g(t, y)
  - Start from  $t_0$ , using  $g(t_0,y)=\xi(t_0,y)$  and 3D Dirac delta function for  $h(t_0)$
  - For each time step from  $t_i$  to  $t_{i+1}$ , use forward PDE to evolve the density from  $h(t_i)$  to  $h(t_{i+1})$ , using the previously calculated  $g(t_i, y)$
  - Use the resulting  $h(t_{i+1})$  at  $t_{i+1}$  to compute the adjustment  $\beta(t_{i+1}, y)$ , basically evaluate the two expectations in the numerator
  - Use the computed adjustment  $\beta(t_{i+1},y)$  along with the classic Dupire local volatility  $\xi(t_{i+1},y)$  to compute the general local volatility  $g(t_{i+1},y)$  for the next time interval from  $t_{i+1}$  to  $t_{i+2}$
  - Repeat steps 2 to 4 until we have evolved the density function all the way to maturity

#### Pricing by Backward PDE

 The pricing is done through solving the backward PDE using the general Dupire local volatility, along with proper terminal conditions and boundary conditions

$$\begin{split} &\frac{\partial V}{\partial t} = rV + \lambda_t x_t \frac{\partial V}{\partial x} + \left(\rho_t^{23} \hat{\sigma}_t g_{t,y} + \hat{\lambda}_t \hat{x}_t\right) \frac{\partial V}{\partial \hat{x}} - \left(r_t - \hat{r}_t - \frac{1}{2} g_{t,y}^2\right) \frac{\partial V}{\partial y} - \frac{\sigma_t^2}{2} \frac{\partial^2 V}{\partial x^2} \\ &- \frac{\hat{\sigma}_t^2}{2} \frac{\partial^2 V}{\partial \hat{x}^2} - \frac{g_{t,y}^2}{2} \frac{\partial^2 V}{\partial y^2} - \rho_t^{12} \sigma_t \hat{\sigma}_t \frac{\partial^2 V}{\partial x \partial \hat{x}} - \rho_t^{13} \sigma_t g_{t,y} \frac{\partial^2 V}{\partial x \partial y} - \rho_t^{23} \hat{\sigma}_t g_{t,y} \frac{\partial^2 V}{\partial \hat{x} \partial y} \end{split}$$

Again, this can be solved by ADI method

# Stochastic Local Vol with Stochastic Rates (Ch 5)

• 4 Factor model: assuming stochastic local vol for spot process. The stochastic volatility component is driven by another stochastic process  $z_t$ 

$$dy_t = \left(r_t - \hat{r}_t - \frac{1}{2}\eta_t^2\right)dt + \eta_t dW_{3,t}, \qquad \eta_t = \gamma_{t,y}\psi_{t,z}$$
$$dz_t = a_{t,z}dt + b_{t,z}dW_{4,t}$$

• Local vol component  $\gamma_{t,y}$  can be derived as

$$\begin{split} &\frac{\mathbb{E}_{s}[D_{s,t}\eta_{t}^{2}|y_{t}=k]}{\mathbb{E}_{s}[D_{s,t}|y_{t}=k]} = \frac{\mathbb{E}_{s}[D_{s,t}\gamma_{t,y}^{2}\psi_{t,z}^{2}|y_{t}=k]}{\mathbb{E}_{s}[D_{s,t}|y_{t}=k]} = \gamma_{t,k}^{2} \frac{\mathbb{E}_{s}[D_{s,t}\psi_{t,z}^{2}|y_{t}=k]}{\mathbb{E}_{s}[D_{s,t}|y_{t}=k]} \\ & \Longrightarrow \gamma_{t,k}^{2} = g_{t,k}^{2} \frac{\mathbb{E}_{s}[D_{s,t}|y_{t}=k]}{\mathbb{E}_{s}[D_{s,t}|y_{t}=k]} = g_{t,k}^{2} \frac{\int_{\Omega} h(t,x,\hat{x},y=k,z|s,x_{s},\hat{x}_{s},y_{s},z_{s})d\Omega}{\int_{\Omega} h(t,x,\hat{x},y=k,z|s,x_{s},\hat{x}_{s},y_{s},z_{s})\psi_{t,z}^{2}d\Omega} \end{split}$$

- Calibration can be challenging: PDE method is basically infeasible for 4-Factor model
- In order to calibrate  $\gamma_{t,y}$  for each time step, we must estimate 4 expectations using Monte Carlo method
  - Two for  $g_{t,k}^2$ :  $\mathbb{E}_s[D_{s,t}r_t\Theta(X_t-K)]$  and  $\mathbb{E}_s[D_{s,t}q_tX_t\Theta(X_t-K)]$
  - Two as above:  $\mathbb{E}_s [D_{s,t} | y_t = k]$  and  $\mathbb{E}_s [D_{s,t} \psi_{t,z}^2 | y_t = k]$

The first two expectation is relatively straight forward to calculate. The second two are conditional expectations, which may need special treatment

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#### Reference:

• Deelstra, G. and Rayée, G., *Local Volatility Pricing Models for Long-dated FX Derivatives*, 2012 https://arxiv.org/pdf/1204.0633