# Stochastic Local Volatility

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March 23, 2014

# 1 The Model

### 1.1 General description

Let T be a finite time horizon (the final maturity of the deal we want to price, for example) Let S be an asset of drift  $\mu_t$  under the domestic risk-neutral probability measure. We assume the following dynamics under  $\mathbb{Q}_d$ 

$$\frac{dS_t}{S_t} = \mu_t dt + \sigma(S_t, t)e^{X_t} dW_t \tag{1}$$

where  $(S,t) \mapsto \sigma(S,t)$  is the local volatility function (deterministic function of spot and time, calibrated to the vanilla option market at time 0. We will develop later a methodology to calibrate that function  $\sigma$ . X is an Ornstein-Ulhenbeck process so that:

$$\forall t \in [0, T], \mathbb{E}\left[e^{X_t}\right] = 1 \tag{2}$$

The goal of the stochastic volatility would be to calibrate to the forward smile (to be consistent with the

# 1.2 Stochastic volatility part

Let us assume that X follows the following dynamics under the domestic risk-neutral probability measure

$$dX_t = (\theta(t) - \lambda X_t)dt + \sigma_X dW_t^X$$

$$X_0 = 1$$
(3)

The coefficients are kept constant for now for simplicity.

We have

$$X_t = e^{-\lambda t} + \int_0^t \theta(s)e^{-\lambda(t-s)}ds + \int_0^t \sigma_X e^{-\lambda(t-s)}dW_s^X$$
(4)

Then,

$$\begin{split} \exp\left(e^{-\lambda t} + \int_0^t \theta(s)e^{-\lambda(t-s)}ds + \frac{1}{2}\int_0^t \sigma_X^2 e^{-2\lambda(t-s)}ds\right) &= 1 \\ e^{-\lambda t} + \int_0^t \theta(s)e^{-\lambda(t-s)}ds + \frac{1}{2}\int_0^t \sigma_X^2 e^{-2\lambda(t-s)}ds &= 0 \\ e^{-\lambda t} + \int_0^t \theta(s)e^{\lambda s}dse^{-\lambda t} + \frac{1}{2}\int_0^t \sigma_X^2 e^{-2\lambda(t-s)}ds &= 0 \\ -\lambda e^{-\lambda t} - \int_0^t \theta(s)e^{\lambda s}ds\lambda e^{-\lambda t} + \sigma_X^2 e^{-2\lambda t} + \theta(t) &= 0 \end{split}$$

Setting  $f(t) = \theta(t)e^{\lambda t}$  it yields

$$\lambda \int_0^t f(s)ds + f(t) = c(t)$$

$$c(t) = \lambda - \frac{\sigma_X^2}{2}e^{-\lambda t}$$
(5)

Then f satisfies the following ODE

$$f'(t) + \lambda f(t) = -\frac{\sigma_X^2 \lambda}{2} e^{-\lambda t}$$

$$f(0) = \lambda - \frac{\sigma_X^2}{2}$$
(6)

Then

$$f(t) = Ce^{-\lambda t} - \frac{\sigma_X^2 \lambda}{2} t e^{-\lambda t}$$

$$= \left(\lambda - \frac{\sigma_X^2}{2}\right) e^{-\lambda t} - \frac{\sigma_X^2 \lambda}{2} t e^{-\lambda t}$$

$$f(t) = \lambda e^{-\lambda t} - \frac{\sigma_X^2}{2} (1 + \lambda t) e^{-\lambda t}$$
(7)

Then

$$\theta(t) = \lambda e^{-2\lambda t} - \frac{\sigma_X^2}{2} (1 + \lambda t) e^{-2\lambda t}$$
(8)

The stochastic volatility X can be computed quite easily and be simulated exactly

$$X_t \sim \mathcal{N}\left(e^{-\lambda t} + \int_0^t \theta(s)e^{-\lambda(t-s)}ds, \sigma^2 \frac{1 - e^{-2\lambda t}}{2\lambda}\right)$$
(9)

#### 1.3 Calibration of the local volatility

**Theorem 1.1** (Markovian Projection). Put here the theorem of the markovian projection of general volatility models into a local volatility

In our case, we have

$$\sigma_D^2(x,t) = \mathbb{E}\left[e^{2X_t}\sigma^2(S_t,t)\middle|S_t = S\right]$$

$$= \mathbb{E}\left[e^{2X_t}\middle|S_t = S\right]\sigma^2(S,t)$$
(10)

Let us introduce the Green function  $G(S_0, X_0, t_0; S, X, t)$ . Let us set  $b(X) = e^X$ . G satisfies

$$\frac{\partial G}{\partial T} - \frac{1}{2} \frac{\partial^2}{\partial S^2} \left( \sigma^2 b^2 S^2 G \right) - \frac{\partial^2}{\partial X^2} \left( \sigma_X^2 G \right) - \frac{\partial^2}{\partial S \partial X} \left( \rho b \sigma S \sigma_X G \right) + \frac{\partial}{\partial X} \left( (\theta - \lambda X) G \right) + \frac{\partial}{\partial S} (\mu S G) = 0 \tag{11}$$

Let us assume (in relation of the previous section), that

- 1.  $\sigma_X$ ,  $\theta$  and  $\lambda$  are not functions of X
- 2.  $\mu$  is not a function of S
- 3.  $\rho$  is neither a function of S nor X

The calibration PDE degenerates into

$$\frac{\partial G}{\partial T} - \frac{1}{2}e^{2X}\frac{\partial^{2}}{\partial S^{2}}\left(\sigma^{2}S^{2}G\right) - \sigma_{X}^{2}\frac{\partial^{2}}{\partial X^{2}}\left(G\right) - \rho\frac{\partial^{2}}{\partial S\partial X}\left(b\sigma S\sigma_{X}G\right) + \left(\theta - \lambda X\right)\frac{\partial}{\partial X}\left(G\right) + \mu\frac{\partial}{\partial S}\left(SG\right) = 0 \tag{12}$$