

# A Multiscale Model with Stochastic Elasticity

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# Motivation

- Although Black and Scholes assumed that the volatility is constant over all strike prices and maturities, in real market, the implied volatility curve (or surface) shows a curvature;
- Volatility level shows a negative correlation with the underlying asset price.



# The CEV(Constant Elasticity of Variance) Model

## CEV

First introduced by Cox(1975), Cox and Ross(1976).

The model assumes that the dynamics of the underlying is given as follows:

$$dX_t = \mu X_t dt + \sigma_0 X_t^{\theta/2} dW_t,$$

where  $\theta$  is responsible for the elasticity of variance which is assumed to be constant.

# Option Price under CEV Model

Under this model, it is known that the option price is given by the following formula(Schroder(1989)):

## CEV Options Price

$$C_{CEV}(t, X_t) = X_t \sum_{n=0}^{\infty} g(n+1, x) G(n+1 + \frac{1}{2-\theta}, kK^{2-\theta}) - Ke^{-r(T-t)} \sum_{n=0}^{\infty} g(n+1 + \frac{1}{2-\theta}, x) G(n+1, kK^{2-\theta}),$$

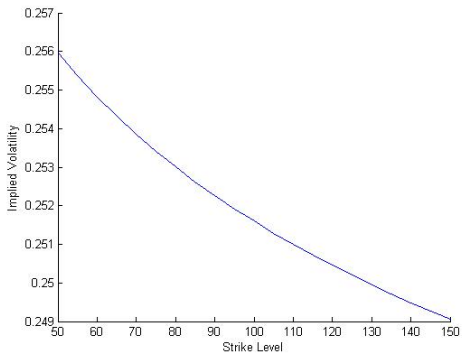
where

$$g(m, v) = \frac{e^{-v} v^{m-1}}{\Gamma(m)}, \quad G(m, v) = \int_v^{\infty} g(m, u) du,$$

$$k = \frac{2r}{\sigma_0^2(2-\theta)(e^{r(2-\theta)(T-t)} - 1)}, \quad x = kX_t^{2-\theta} e^{r(2-\theta)(T-t)}$$

# Benefits from CEV

It captures the volatility skew phenomenon.



# Downsides of CEV

It tells us a false dynamics of implied volatility which may lead unstable hedges.

The correct dynamics of volatility is such that the volatility curve shifts in the **same direction** of the underlying asset movement(Hagan(2002)).

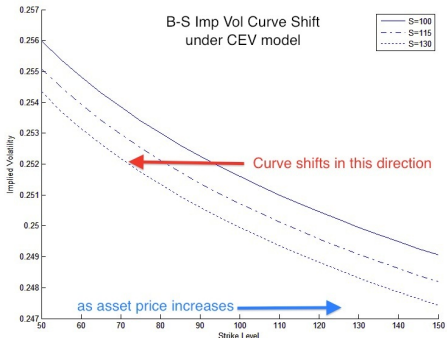


Figure: Wrong dynamics by CEV

# The Stochastic Elasticity of Variance Model

# The SEV(Stochastic Elasticity of Variance) Model

## Introduction of the Stochastic Elasticity

The elasticity of variance parameter in CEV model :  $\theta/2$

Some empirical studies show that  $\theta > 2$  (Bollerslev et al.(1988), Campbell and Hentschel(1992), Ghysels et al.(2005)).

Other studies show that  $\theta < 2$  (Campbell(1987), Breen et al(1989), Glosten et al(1993), Brandt and Kang(2004)).

It also is shown to be time-varying (Harvey(2001), Ghysels et al(1996)).



# The SEV(Stochastic Elasticity of Variance) Model

## CEV to SEV

$$\theta/2 \implies 1 - \gamma f(Y_t)$$

## SEV Model dynamics:

Our model is governed by the SDEs

$$dX_t = \mu X_t dt + \sigma X_t^{1-\gamma f(Y_t)} dB_t^x, \quad (2.1)$$

$$dY_t = \alpha(m - Y_t)dt + \beta dB_t^y, \quad (2.2)$$

with assumption that the Brownian motions  $B_t^x$  and  $B_t^y$  are correlated each other and  $f$  is a bounded function.

Notice that as  $\gamma$  approaches to zero, the model becomes that of Black and Scholes. Hence, we expect the Black-Scholes price to be the leading-order term in our solution.

# Multiple Scales

## Assumption

$Y_t$  has an invariant distribution with variance  $\frac{\beta^2}{2\alpha}$ .

Assume that  $\nu := \frac{\beta}{\sqrt{2\alpha}}$  is an  $O(1)$ -term.

## Assumptions on Multiple Scales

- Fast mean-reverting of the process  $Y_t$ . (*i.e.*  $\alpha$  being large.)
- $1 - \gamma f(Y_t)$  (elasticity) becomes stable to be 1. (*i.e.*  $\gamma$  being small.)
- $\alpha$  being large while  $\gamma$  being small.

# Multiple Scales

Small parameters for the job

$$\begin{aligned}\epsilon &= \frac{1}{\alpha}, \\ \delta &= \gamma^2\end{aligned}$$

Motivation

$\delta$ : We want to have the Black-Scholes price as a leading order term.

$\epsilon$ : Fast mean-reverting OU process is analytically more tractable.

# Change of Measure

## Our SDEs under the Risk-Neutral Measure

Under risk-neutral measure, they become

$$dX_t = rX_t dt + \sigma X_t^{1-\sqrt{\delta}f(Y_t)} dW_t^x, \quad (3.1)$$

$$dY_t = \left[ \frac{1}{\epsilon}(m - Y_t) - \frac{1}{\sqrt{\epsilon}}\nu\sqrt{2}\Lambda^\delta(X_t, Y_t) \right] dt + \frac{1}{\sqrt{\epsilon}}\nu\sqrt{2} dW_t^y \quad (3.2)$$

where the correlation of Brownian motions  $W_t^x$  and  $W_t^y$  are given by

$$d\langle W^x, W^y \rangle_t = \rho dt$$

# The Option Price

## The Option Price

Under the risk-neutral measure, the option price is given by

$$P(t, X_t, Y_t) = E^*[e^{-r(T-t)}h(X_T)|X_t, Y_t], \quad (3.3)$$

where  $h(X_T)$  is the corresponding payoff function for either put or call option.

# Feynman-Kac's Formula

From the Feynman-Kac's formula, one can obtain the pricing PDE

$$\begin{aligned}\mathcal{L}^{\epsilon,\delta} P^{\epsilon,\delta} &= 0 \\ \mathcal{L}^{\epsilon,\delta} &:= \frac{\partial}{\partial t} + \mathcal{L}_{X,Y}^{\epsilon,\delta} - r \cdot, \\ P^{\epsilon,\delta}(T, x, y) &= h(x),\end{aligned}\tag{3.4}$$

where  $\mathcal{L}_{X,Y}^{\epsilon,\delta}$  is the infinitesimal generator of the diffusion process  $(X_t, Y_t)$  given by

$$\begin{aligned}\mathcal{L}_{X,Y}^{\epsilon,\delta} &= \frac{1}{2} \sigma^2 x^{2(1-f(y)\sqrt{\delta})} \frac{\partial^2}{\partial x^2} + rx \frac{\partial}{\partial x} \\ &+ \frac{1}{\sqrt{\epsilon}} \left( \rho \nu \sqrt{2} \sigma x^{1-f(y)\sqrt{\delta}} \frac{\partial^2}{\partial x \partial y} - \nu \sqrt{2} \Lambda^\delta(x, y) \frac{\partial}{\partial y} \right) \\ &+ \frac{1}{\epsilon} \left( \nu^2 \frac{\partial^2}{\partial y^2} + (m - y) \frac{\partial}{\partial y} \right).\end{aligned}\tag{3.5}$$

# Asymptotic Analysis

Singular perturbation for  $\epsilon$  and regular perturbation for  $\delta$ .

Separating the Operator  $\mathcal{L}^{\epsilon,\delta}$

$$\mathcal{L}^{\epsilon,\delta} = \frac{1}{\epsilon} \mathcal{L}_0 + \frac{1}{\sqrt{\epsilon}} \mathcal{L}_1^\delta + \mathcal{L}_2^\delta, \quad (4.1)$$

$$\mathcal{L}_0 := \nu^2 \frac{\partial^2}{\partial y^2} + (m - y) \frac{\partial}{\partial y}, \quad (4.2)$$

$$\mathcal{L}_1^\delta := \rho\nu\sqrt{2}\sigma x^{1-f(y)\sqrt{\delta}} \frac{\partial^2}{\partial x \partial y} - \nu\sqrt{2}\Lambda^\delta(x, y) \frac{\partial}{\partial y}, \quad (4.3)$$

$$\mathcal{L}_2^\delta := \frac{\partial}{\partial t} + \frac{1}{2}\sigma^2 x^{2(1-f(y)\sqrt{\delta})} \frac{\partial^2}{\partial x^2} + r(x \frac{\partial}{\partial x} - \cdot) \quad (4.4)$$

# Regular Perturbation for $\delta$

Handling Dependencies of  $\mathcal{L}_1^\delta$  and  $\mathcal{L}_2^\delta$  on  $\delta$

$$\mathcal{L}_1^\delta := \rho\nu\sqrt{2}\sigma x^{1-f(y)\sqrt{\delta}} \frac{\partial^2}{\partial x \partial y} - \nu\sqrt{2}\Lambda^\delta(x, y) \frac{\partial}{\partial y},$$

$$\mathcal{L}_2^\delta := \frac{\partial}{\partial t} + \frac{1}{2}\sigma^2 x^{2(1-f(y)\sqrt{\delta})} \frac{\partial^2}{\partial x^2} + r(x \frac{\partial}{\partial x} - \cdot)$$

Use the expansion

$$x^{1-f\sqrt{\delta}} = x(1 - \sqrt{\delta}f \log x + \delta \frac{(f \log x)^2}{2!} - \delta\sqrt{\delta} \frac{(f \log x)^3}{3!} + \dots)$$



# Asymptotic Analysis

## Further Expansion in Powers of $\sqrt{\delta}$

$$\mathcal{L}_1^\delta = \mathcal{L}_{10} + \sqrt{\delta}\mathcal{L}_{11} + \delta\mathcal{L}_{12} + \cdots, \quad (4.5)$$

$$\mathcal{L}_{10} := \nu\sqrt{2\rho\sigma}x \frac{\partial^2}{\partial x\partial y}, \quad (4.6)$$

$$\mathcal{L}_{11} := -\nu\sqrt{2\rho\sigma}xf \log x \frac{\partial^2}{\partial x\partial y}, \quad (4.7)$$

$$\mathcal{L}_{12} := \nu\sqrt{2\rho\sigma}x \frac{(f \log x)^2}{2!} \frac{\partial^2}{\partial x\partial y}, \quad (4.8)$$

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# Asymptotic Analysis

Further Expansion in Powers of  $\sqrt{\delta}$  -cont

and

$$\mathcal{L}_2^\delta = \mathcal{L}_{20} + \sqrt{\delta}\mathcal{L}_{21} + \delta\mathcal{L}_{22} + \dots, \quad (4.9)$$

$$\mathcal{L}_{20} := \frac{\partial}{\partial t} + \frac{1}{2}\sigma^2 x^2 \frac{\partial^2}{\partial x^2} + r\left(x\frac{\partial}{\partial x} - \cdot\right) = \mathcal{L}_{BS}, \quad (4.10)$$

$$\mathcal{L}_{21} := -\sigma^2 x^2 f \log x \frac{\partial^2}{\partial x^2}, \quad (4.11)$$

$$\mathcal{L}_{22} := \sigma^2 x^2 (f \log x)^2 \frac{\partial^2}{\partial x^2}, \quad (4.12)$$

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# Asymptotic Analysis

## Asymptotic Analysis

We plug the expansion

$$P^{\epsilon, \delta} = P_0^\epsilon + \sqrt{\delta} P_1^\epsilon + \delta P_2^\epsilon + \dots \quad (4.13)$$

into the PDE (3.4) and obtain the following hierarchy:

$$\delta^0 : \left( \frac{1}{\epsilon} \mathcal{L}_0 + \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{10} + \mathcal{L}_{20} \right) P_0^\epsilon = 0, \quad (4.14)$$

$$\delta^{\frac{1}{2}} : \left( \frac{1}{\epsilon} \mathcal{L}_0 + \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{10} + \mathcal{L}_{20} \right) P_1^\epsilon + \left( \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{11} + \mathcal{L}_{21} \right) P_0^\epsilon = 0, \quad (4.15)$$

$$\begin{aligned} \delta : \left( \frac{1}{\epsilon} \mathcal{L}_0 + \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{10} + \mathcal{L}_{20} \right) P_2^\epsilon + \left( \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{11} + \mathcal{L}_{21} \right) P_1^\epsilon \\ + \left( \frac{1}{\sqrt{\epsilon}} \mathcal{L}_{12} + \mathcal{L}_{22} \right) P_0^\epsilon = 0, \end{aligned} \quad (4.16)$$

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and impose the final conditions  $P_0^\epsilon(T, x, y) = h(x)$  and  $P_1^\epsilon(T, x, y) = 0 \dots$

# Asymptotic Analysis

Analysis of  $\delta^0$  Power w.r.t. the Parameter  $\epsilon$

Expand  $P_0^\epsilon$  as follows

$$P_0^\epsilon = P_{0,0} + \sqrt{\epsilon}P_{0,1} + \epsilon P_{0,2} + \epsilon\sqrt{\epsilon}P_{0,3} + \dots$$

with the final conditions  $P_{0,0}(T, x, y) = h(x)$ ,  $P_{0,1}(T, x, y) = 0$ ,  
 $P_{0,2}(T, x, y) = 0 \dots$

Then we have the following sequence of PDEs:

$$\begin{aligned}\mathcal{L}_0 P_{0,k} + \mathcal{L}_{10} P_{0,k-1} + \mathcal{L}_{20} P_{0,k-2} &= 0, \\ P_{0,-2} &:= 0, \\ P_{0,-1} &:= 0,\end{aligned}\tag{4.17}$$

where  $k = 0, 1, 2, \dots$

# Asymptotic Analysis

## A Few More Assumptions

For  $k = 0$ , the PDE,

$$\mathcal{L}_0 P_{0,0} = 0,$$

is assumed to admit only solutions that do not grow so much as

$$\frac{\partial P_{0,0}}{\partial y} \sim e^{\frac{y^2}{2}}, \quad y \rightarrow \infty,$$

then  $P_{0,0}$  becomes a function of  $t$  and  $x$  only;  $P_{0,0} = P_{0,0}(t, x)$

Similar assumption for  $k = 1$  gives the same result on  $P_{0,1}$ ;  
 $P_{0,1} = P_{0,1}(t, x)$ .

# Asymptotic Analysis

## A Few More Assumptions -cont

We impose the centering condition on  $\mathcal{L}_{2,0}$ , i.e.

$$\langle \mathcal{L}_{20} P_{0,0} \rangle = \mathcal{L}_{20} P_{0,0} = 0,$$

where

$$\langle g \rangle := \frac{1}{\sqrt{2\pi\nu^2}} \int_{-\infty}^{\infty} g(y) e^{-\frac{(y-m)^2}{2\nu^2}} dy$$

in order to admit the solutions only with reasonable growth.

# Asymptotic Analysis

## A Few More Assumptions -cont

We obtain the final value problem:

$$\begin{aligned}\mathcal{L}_{20}P_{0,0} &:= \mathcal{L}_{BS}P_{0,0} = 0, \\ P_{0,0}(T, x, y) &= h(x)\end{aligned}\tag{4.18}$$

thus

$$\mathcal{L}_0P_{2,0} = 0\tag{4.19}$$

As we repeat this methodology, and properly apply the centering condition, we obtain that  $P_{0,0}$  is the Black-Scholes price and  $P_{0,1} = 0$ , and  $P_{0,2} = 0$ .

# Asymptotic Analysis

Analysis of  $\delta^{\frac{1}{2}}$  Power w.r.t. the Parameter  $\epsilon$

Expand  $P_1^\epsilon$  as follows

$$P_1^\epsilon = P_{1,0} + \sqrt{\epsilon}P_{1,1} + \epsilon P_{1,2} + \epsilon\sqrt{\epsilon}P_{1,3} + \dots$$

with the final conditions  $P_{1,0}(T, x, y) = h(x)$ ,  $P_{1,1}(T, x, y) = 0$ ,  
 $P_{2,1}(T, x, y) = 0 \dots$



# Asymptotic Analysis

As we apply the same logic, we obtain the following equation:

$$\begin{aligned}\mathcal{L}_{BS}P_{1,0}^\delta &= V_{1,0}^\delta x^2 \log x \frac{\partial^2 P_{0,0}}{\partial x^2}, \\ P_{1,0}^\delta(T, x, y) &= 0, \\ V_{1,0}^\delta &:= \gamma \sigma^2 \langle f \rangle.\end{aligned}\tag{4.20}$$

# The Group Parameter: $V_{1,0}^\delta$

The leading order correction is entirely captured by the group parameter  $V_{1,0}^\delta := \gamma\sigma^2\langle f \rangle$  that we introduced earlier.

There is no need for the knowledge of the parameters  $\delta, \epsilon, \nu, m, \rho$ , and specific form of the function  $f$ .

$V_{1,0}^\delta$  can be obtained by calibrating the model to the option prices on the market.

# Integral Representation

The price  $P^{\epsilon, \delta}$  as a solution of the final value problem (4.20) with  $h(x) = (x - K)^+$  (i.e. for a call option) is asymptotically given by

$$\begin{aligned} P^{\epsilon, \delta} &\sim x\Phi(d_1) - K\Phi(d_2) \\ &\quad - V_{1,0}^{\delta}(K/\sigma^2) \int_0^{\tau} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi s(\tau-s)}} e^{-\frac{(\tilde{y}-\tilde{x})^2}{4(\tau-s)} + \tilde{x}} \\ &\quad * (\tilde{y} + \log(K)) \phi(d_1) e^*(\tilde{y} - \tilde{x}, \tau - s) d\tilde{y} ds, \end{aligned} \quad (4.21)$$

where

$$\begin{aligned} \tilde{x} &:= \log(x/K), & \tau &:= \frac{1}{2}\sigma^2(T-t), \\ e^*(x, t) &:= e^{\frac{1}{2}(k+1)x - \frac{1}{4}(k+1)^2 t}, & k &:= \frac{2r}{\sigma^2}, \end{aligned}$$

and  $\Phi$  is the cumulative normal distribution and  $\phi$  is the standard normal density function and  $d_1, d_2$  are as appeared in Black-Scholes model.

# Integral Representation

## Proof

One can solve PDE (4.20) for  $P_{1,0}^\delta$  by transforming it into the heat equation with a source term via the transformation

$$\tilde{x} := \log(x/K), \quad \tau := \frac{1}{2}\sigma^2(T-t)$$

for independent variables and the transformation

$$u(\tau, \tilde{x}) = \frac{P_{1,0}^\delta}{K} e^{\frac{1}{2}(k-1)\tilde{x} + \frac{1}{4}(k+1)^2\tau}$$

for dependent variable.

# Integral Representation

## Proof-cont

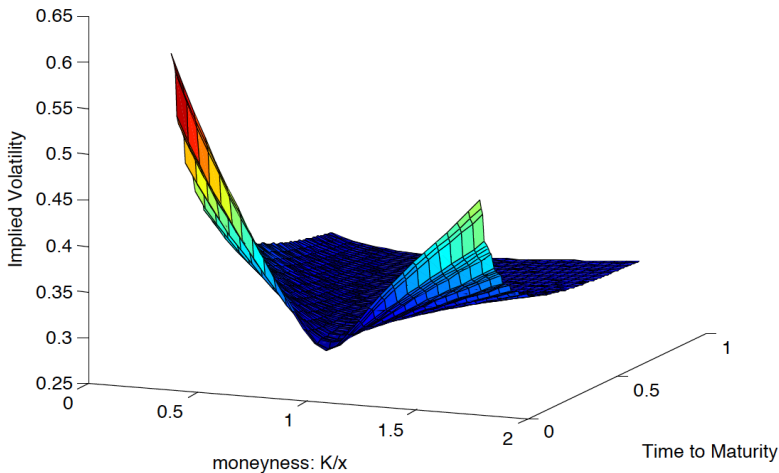
Since  $\frac{\partial^2 P_{0,0}}{\partial x^2}$  is merely the Gamma of an option, the resultant equation for  $u$  as follows:

$$\begin{aligned}\frac{\partial u}{\partial \tau} &= \frac{\partial^2 u}{\partial \tilde{x}^2} - V_{1,0}^\delta \frac{\sqrt{2}}{\sigma^2 \sqrt{\tau}} (\tilde{x} + \log K) \phi(d_1) e^{\frac{1}{2}(k+1)\tilde{x} + \frac{1}{4}(k+1)^2 \tau} \\ u(0, \tilde{x}) &= 0.\end{aligned}$$

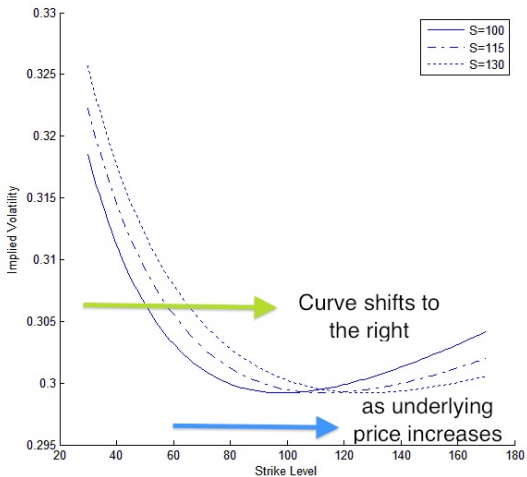
This is the heat equation with zero initial condition and a nonzero source term and its solution  $u$  and subsequently  $P_{1,0}^\delta$  is well-known.

## Some Simulation Results

# The Implied Volatility Surface

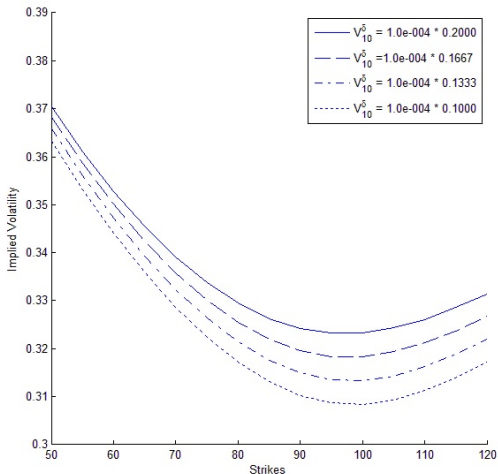


# Dynamics of Implied Volatility





# The Implied Volatility Curves with Various $V_{1,0}^\delta$ 's



# Conclusion

## Conclusion and Works to be done

- SEV model overcomes the major drawback of the Black-Scholes model and gives us a smile curve.
- It fits observed market behavior of volatility curve shift and overcomes problems that CEV model has.
- A model with the CEV price in the leading order term.

**Thank you for your attention!**

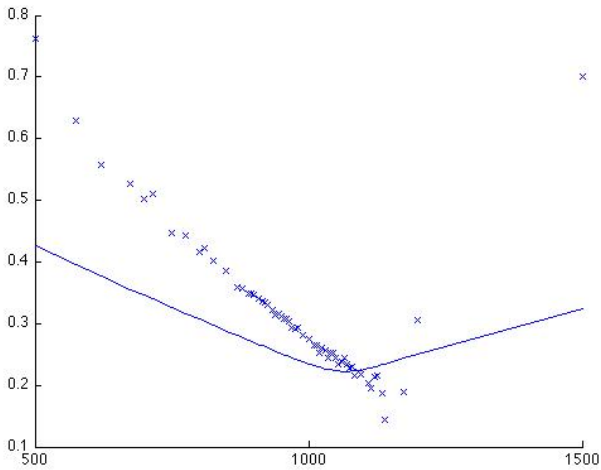


Figure: Fitted implied volatility curve with Spot: S&P500