

Pricing and Hedging with Constant Elasticity and Stochastic Volatility

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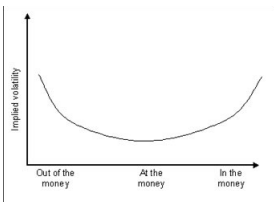
Outline

- 1 Introduction
 - Background
 - Purpose
- 2 Stochastic Volatility CEV
 - Dynamics
 - Characteristics
 - Corrected Price
 - Asymptotic theory
- 3 Numerical Implementation
 - P_0 and P_1
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Drawback of Black Scholes Model

- Smile curve



- Rubinstein (1985)
- Jackwerth and Rubinstein (1996)
- In B.S. Model, Implied Volatility curve is flat.
- We need to use the implied volatility which explicitly depends on the option strike and maturity.



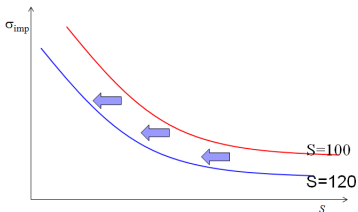
Local Volatility Model

Local Volatility Model

- One needs volatility to depend on underlying
- Local Volatility Models

$$\frac{dS_t}{S_t} = \mu(S_t, t)dt + \sigma(S_t, t)dW_t$$

- The dynamics of Implied Volatility in Local Volatility model.



This is opposite to real market. (Hagan, 2002)



CEV Model

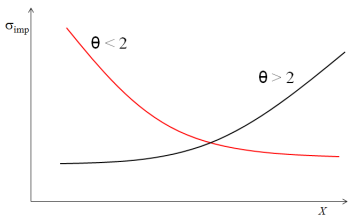
- CEV (constant elasticity of variance) diffusion model
- X_t stock price s.t.

$$dX_t = \mu X_t dt + \sigma X_t^{\frac{\theta}{2}} dW_t$$

- Introduced by Cox and Ross(1976)
- Studied by Beckers (1980), Schroder(1989), Boyle and Tian(1999), Davydov and Linetsky(2001), Delbaen and Schirakawa(2002), Heath and Platen(2002), Carr and Linetsky(2006) and etc.
- Analytic tractability



CEV Model cont.



- When $\theta = 2$ the model is Black-Scholes case.
- When $\theta < 2$ volatility falls as stock price rises.
 \Rightarrow realistic, can generate a fatter left tail.
- When $\theta > 2$ volatility rise as stock price rises.
 \Rightarrow (futures option)



CEV Model cont.

Theorem (Lipton ,2001)

The call option price C_{CEV} for $X_t = x$ is given by

$$C_{CEV}(t, x) = e^{-r(T-t)} x \int_{\check{K}}^{\infty} \left(\frac{\check{X}}{y} \right)^{\frac{1}{2(2-\theta)}} e^{-(\check{x}+y)} I_{\frac{1}{2-\theta}}(2\sqrt{\check{x}y}) dy \\ + e^{-r(T-t)} K \int_{\check{K}}^{\infty} \left(\frac{y}{\check{X}} \right)^{\frac{1}{2(2-\theta)}} e^{-(\check{x}+y)} I_{\frac{1}{2-\theta}}(2\sqrt{\check{x}y}) dy,$$

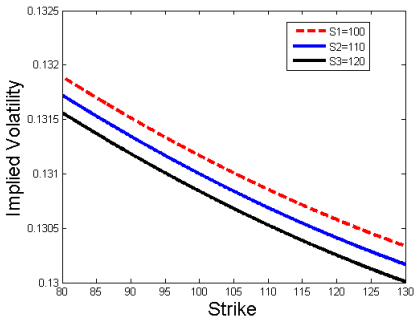
where

$$\check{x} = \frac{2xe^{r(2-\theta)(T-t)}}{(2-\theta)^2\chi}, \quad \chi = \frac{\sigma^2}{(2-\theta)r}(e^{r(2-\theta)T} - e^{r(2-\theta)t}) \\ \check{K} = \frac{2K^{2-\theta}}{(2-\theta)^2\chi}$$



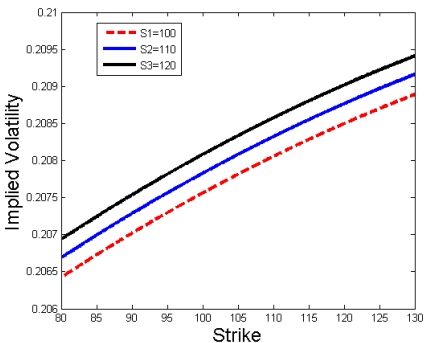
Dynamics of Implied Volatility for CEV model

$$\theta = 1.9$$



Dynamics of Implied Volatility for CEV model cont.

$$\theta = 2.1$$



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Characteristics

- The new volatility is given by the multiplication of a function of a new process
- The new process is taken to be an Ito process (O-U process):

$$dY_t = \alpha(m - Y_t)dt + \beta d\hat{Z}_t.$$

- α = rate of mean reversion.
Assume that mean reversion is fast.
So, α is **large enough**.



Corrected Pricing

Use Risk Neutral Valuation method

Equivalent martingale measure Q , Option price is given by the formula

$$P(t, x, y) = E^Q[e^{-r(T-t)}h(X_T)|X_t = x, Y_t = y] \quad (5)$$



Corrected Pricing Theorem

Theorem 3.1

The option price $P(t, x, y)$ defined by (5) satisfies the Kolmogorov PDE

$$P_t + \frac{1}{2}f^2(y)x^\theta P_{xx} + \rho\beta f(y)x^{\frac{\theta}{2}} P_{xy} + \frac{1}{2}\beta^2 P_{yy} + rxP_x + (\alpha(m - y) - \beta\Lambda(t, x, y))P_y - rP = 0, \quad (6)$$

where

$$\Lambda(t, x, y) = \rho \frac{\mu - r}{f(y)x^{\frac{\theta-2}{2}}} + \sqrt{1 - \rho^2} \gamma(y).$$



Asymptotic theory

- Develop an asymptotic theory on fast mean reversion
- Introduce A **small parameter** ϵ

$$\epsilon = \frac{1}{\alpha}$$

- Assume $\nu = \frac{\beta}{\sqrt{2\alpha}}$ is fixed in scale as ϵ become zero.

$$\alpha \sim \mathcal{O}(\epsilon^{-1}), \quad \beta \sim \mathcal{O}(\epsilon^{-1/2}), \quad \text{and } \nu \sim \mathcal{O}(1).$$



Singular Perturbation

To solve the PDE (6), we use Singular Perturbation method.

Procedure

- Substituting the asymptotic series

$$P(x; \epsilon) \approx \sum_{n=0}^{\infty} \epsilon^n P_n(x)$$

into the differential equation.

- Expanding all quantities in a power series in ϵ .
- Collecting terms with same powers of ϵ and equating them to zero.
- Solving this hierarchy of the problem sequentially.



Asymptotic theory

After rewritten in terms of ϵ , the PDE (6) becomes

$$P_t + \frac{1}{2}f^2(y)x^\theta P_{xx} + \rho \frac{\sqrt{2}\nu}{\sqrt{\epsilon}} f(y)x^{\frac{\theta}{2}} P_{xy} + \frac{1}{2} \frac{2\nu^2}{\epsilon} P_{yy} \\ + rxP_x + \left(\frac{1}{\epsilon}(m-y) - \beta \frac{\sqrt{2}\nu}{\sqrt{\epsilon}} \Lambda(t, x, y)\right) P_y - rP = 0.$$

Collecting by ϵ order,

$$\frac{1}{\epsilon} (\nu^2 P_{yy} + (m-y)P_y) + \frac{1}{\sqrt{\epsilon}} (\sqrt{2}\rho\nu f(y)x^{\frac{\theta}{2}} P_{xy} - \sqrt{2}\nu \Lambda P_y) \\ + (P_t + \frac{1}{2}f^2(y)x^\theta P_{xx} + rxP_x - rP) = 0$$



Asymptotic theory cont.

Define operators \mathcal{L}_0 , \mathcal{L}_1 , \mathcal{L}_2 as

$$\mathcal{L}_0 = \nu^2 \partial_{yy}^2 + (m - y) \partial_y, \quad (7)$$

$$\mathcal{L}_1 = \sqrt{2} \rho \nu f(y) x^{\frac{\theta}{2}} \partial_{xy}^2 - \sqrt{2} \nu \Lambda(t, x, y) \partial_y, \quad (8)$$

$$\mathcal{L}_2 = \partial_t + \frac{1}{2} f(y)^2 x^\theta \partial_{xx}^2 + r(x \partial_x - \cdot). \quad (9)$$

then, the PDE (6) can be written as

$$\left(\frac{1}{\epsilon} \mathcal{L}_0 + \frac{1}{\sqrt{\epsilon}} \mathcal{L}_1 + \mathcal{L}_2 \right) P^\epsilon = 0 \quad (10)$$



Asymptotic Expansion

Expand P^ϵ in powers of $\sqrt{\epsilon}$:

$$P^\epsilon = P_0 + \sqrt{\epsilon}P_1 + \epsilon P_2 + \dots \quad (11)$$

Here, the choice of the power unit $\sqrt{\epsilon}$ in the power series expansion was determined by the method of matching coefficient.



Asymptotic Expansion cont.

Substituting the PDE (10),

$$\begin{aligned} & \frac{1}{\epsilon} \mathcal{L}_0 P_0 + \frac{1}{\sqrt{\epsilon}} (\mathcal{L}_0 P_1 + \mathcal{L}_1 P_0) \\ & + (\mathcal{L}_0 P_2 + \mathcal{L}_1 P_1 + \mathcal{L}_2 P_0) + \\ & \sqrt{\epsilon} (\mathcal{L}_0 P_3 + \mathcal{L}_1 P_2 + \mathcal{L}_2 P_1) + \cdots = 0, \end{aligned} \quad (12)$$

which holds for arbitrary $\epsilon > 0$.



Asymptotic theory cont.

Lemma 3.1

If solution to the Poisson equation

$$\mathcal{L}_0\chi(y) + \psi(y) = 0 \quad (13)$$

exists, then the following centering (solvability) condition must satisfy $\langle \psi \rangle = 0$, where $\langle \cdot \rangle$ is the expectation with respect to the invariant distribution of Y_t . If then, solutions of (13) are given by the form

$$\chi(y) = \int_0^t E^y[\psi(Y_t)] dt + \text{constant}. \quad (14)$$

Note:

$$\langle \psi \rangle = \int_{-\infty}^{\infty} \psi(y) f(y) dy, \quad f(y) = \frac{1}{\sqrt{2\pi\nu^2}} \exp\left(-\frac{(y-m)^2}{2\nu^2}\right)$$



Asymptotic Expansion cont.

From the asymptotic expansion (12) $1/\epsilon$ order, we first have

$$\mathcal{L}_0 P_0 = 0. \quad (15)$$

Solving this equation yields

$$P_0(t, x, y) = c_1(t, x) \int_0^y e^{\frac{(m-z)^2}{2\nu^2}} dz + c_2(t, x)$$

for some functions c_1 and c_2 independent of y .

- $c_1 = 0$ is required.
- $P_0(t, x, y)$ must be a function of only t and x

$$P_0 = P_0(t, x).$$



Asymptotic Expansion cont.

- From the expansion (12) $1/\sqrt{\epsilon}$ order ,

$$\mathcal{L}_0 P_1 + \mathcal{L}_1 P_0 = 0$$

- Known $\mathcal{L}_1 P_0 = 0$
- Get $\mathcal{L}_0 P_1 = 0$

$$P_1 = P_1(t, x) \tag{16}$$



Asymptotic Expansion P_0

Theorem 3.2

The leading term $P_0(t, x)$ is given by the solution of the PDE

$$\frac{\partial P_1}{\partial t} + \frac{1}{2} \langle f^2 \rangle x^\theta \frac{\partial^2 P_1}{\partial x^2} + r(x \frac{\partial P_1}{\partial x} - P_1) = 0 \quad (17)$$

with the terminal condition $P_0(T, x) = h(x)$.



Proof of Theorem 3.2

Proof

From the expansion (12), the PDE

$$\mathcal{L}_0 P_2 + \mathcal{L}_1 P_1 + \mathcal{L}_2 P_0 = 0 \quad (18)$$

Since $\mathcal{L}_1 P_1 = 0$, then

$$\mathcal{L}_0 P_2 + \mathcal{L}_2 P_0 = 0 \quad (19)$$

which is a Poisson equation.



Proof of Theorem 3.2 cont.

From Lemma 3.1 with $\psi = \mathcal{L}_2 P_0$, $P_0(t, x)$ has to satisfy the centering condition

$$\langle \mathcal{L}_2 \rangle P_0 = 0 \quad (20)$$

with the terminal condition $P_0(T, x) = h(x)$, where

$$\langle \mathcal{L}_2 \rangle = \partial_t + \frac{1}{2} \langle f^2 \rangle x^\theta \partial_{xx}^2 + r(x \partial_x - \cdot).$$

Thus P_0 solves the PDE (17). \square



Asymptotic Expansion P_1 cont.

Theorem 3.3

The first correction $P_1(t, x)$ is given by the solution of the PDE

$$\begin{aligned} \frac{\partial P_1}{\partial t} + \frac{1}{2} \langle f^2 \rangle x^\theta \frac{\partial^2 P_1}{\partial x^2} + r(x \frac{\partial P_1}{\partial x} - P_1) = \\ V_3 x \frac{\partial}{\partial x} (x^2 \frac{\partial^2 P_0}{\partial x^2}) + V_2 x^2 \frac{\partial^2 P_0}{\partial x^2} \end{aligned} \quad (21)$$

with the final condition $P_1(T, x) = 0$, where V_3 and V_2 are given by (22) and (23), respectively.



Asymptotic Expansion P_1

For convenience,

$$V_3(x; \theta) = \frac{\nu}{\sqrt{2}} \rho x^{\frac{\theta-2}{2}} \langle f \psi_y \rangle, \quad (22)$$

$$V_2(x; \Lambda; \theta) = \frac{\nu}{\sqrt{2}} \left(\rho x^{\frac{\theta}{2}} \langle f \psi_{xy} \rangle - \langle \Lambda \psi_y \rangle \right), \quad (23)$$

where $\psi(t, x, y)$ is solution of the Poisson equation

$$\mathcal{L}_0 \psi = \nu^2 \psi_{yy} + (m - y) \psi_y = (f^2 - \langle f^2 \rangle) x^{\theta-2}. \quad (24)$$



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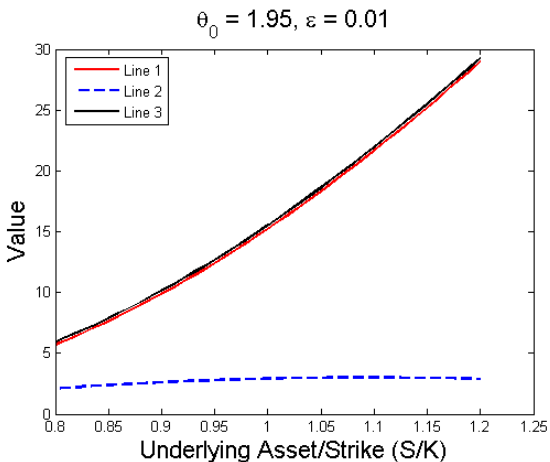
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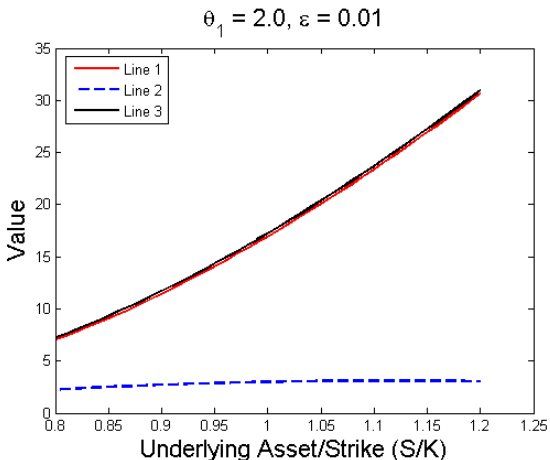
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$\theta = 1.95$ and $\epsilon = 0.01$ 

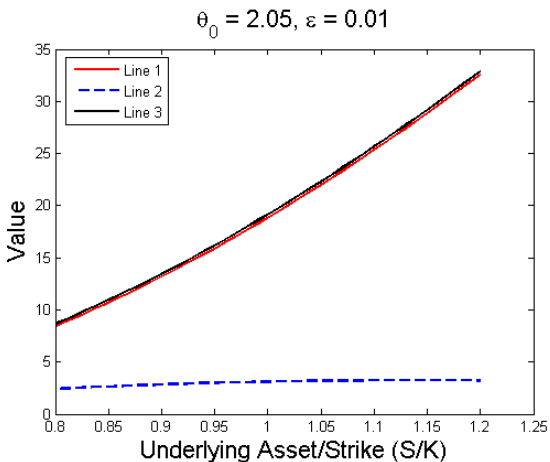
Line 1 = P_0 , Line 2 = P_1 , Line 3 = $P_0 + \sqrt{\epsilon}P_1$

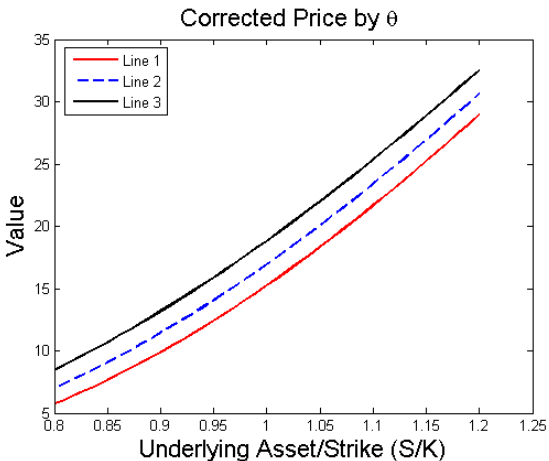


$\theta = 2.00$ and $\epsilon = 0.01$ 

$P_0(t, x)$ with $K = 100$, $T = 1$, $\bar{\sigma} = 0.165$, and $r = 0.02$. It is computed with terminal condition $h(x) = (x - K)^+$.



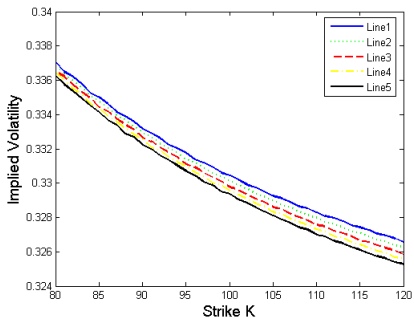
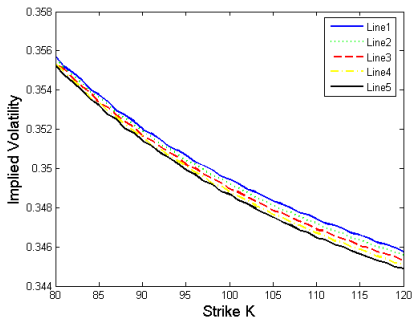
$\theta = 2.05$ and $\epsilon = 0.01$ 



Line 1 : $\theta = 1.95$, Line 2 : $\theta = 1.95$, Line 3 : $\theta = 2.05$



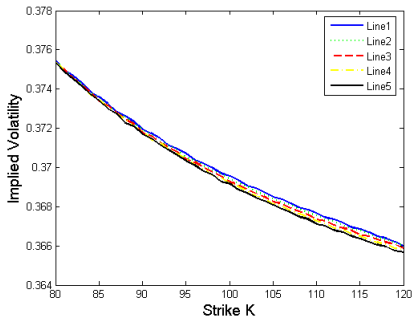
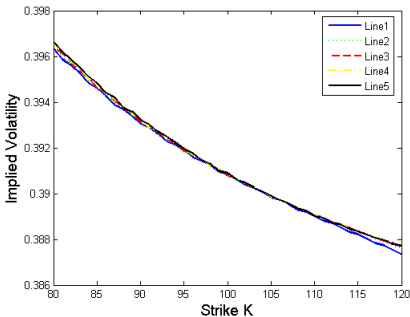
Dynamics of Implied Volatility ($\theta = 1.9$ and $\theta = 1.925$)

(a) $\theta = 1.9$ (b) $\theta = 1.925$

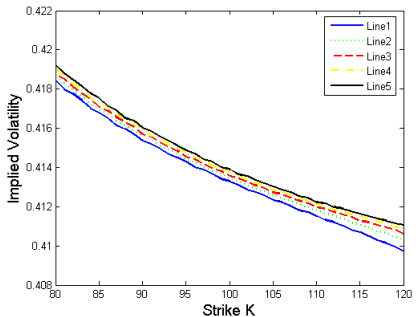
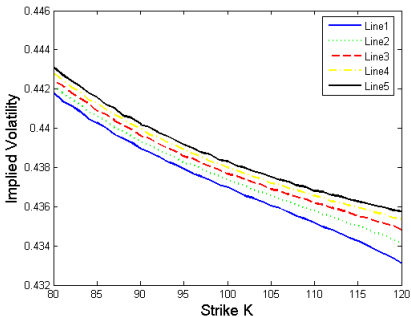
Line 1 : $X_0 = 90$, Line 2 : $X_0 = 95$, Line 3 : $X_0 = 100$,
Line 4 : $X_0 = 105$, Line 5 : $X_0 = 110$



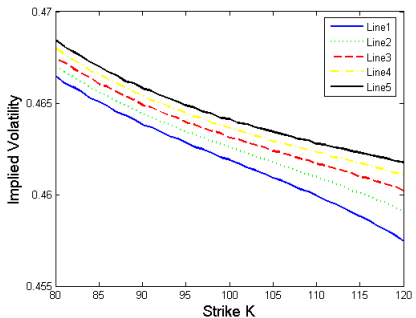
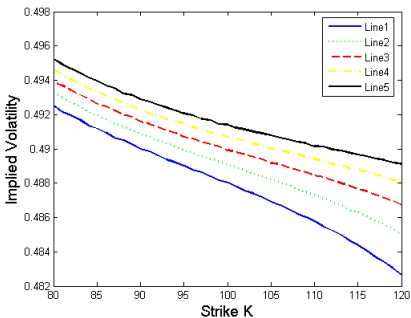
Dynamics of Implied Volatility ($\theta = 1.95$ and $\theta = 1.975$)

(c) $\theta = 1.95$ (d) $\theta = 1.975$ 

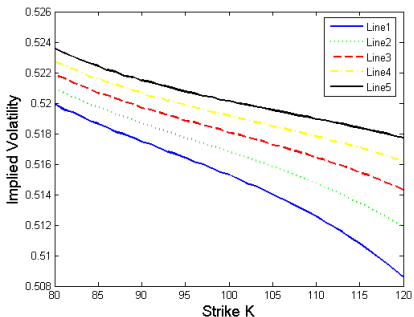
Dynamics of Implied Volatility ($\theta = 2.00$ and $\theta = 2.025$)

(e) $\theta = 2.0$ (f) $\theta = 2.025$ 

Dynamics of Implied Volatility ($\theta = 2.05$ and $\theta = 2.075$)

(g) $\theta = 2.05$ (h) $\theta = 2.075$ 

Dynamics of Implied Volatility ($\theta = 2.1$)



(i) $\theta = 2.1$

Remark

- Implied volatility curve move from left to right for $\theta \geq 1.975$.
- For $\theta \geq 2$, The implied volatility curve seems to be skew, unlike CEV model.



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Conclusion

Conclusion

- Corrected Price (A new hybrid model)
- Right dynamics of Implied Volatility
- Stability of Hedging
- Still ongoing research(Fitting to Market Data)



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Thank you for your attention!

