

Small time asymptotics for implied volatilities

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Outline

- Brief introduction on option pricing theory and implied volatility
- Implied volatility in terms of local volatility
 - The heat kernel approach
 - The BBF approximation
 - BBF to higher orders
- One expansion, two approaches
 - Laplace asymptotic formula
 - Expansion of time value
- Numerical tests
- Summary and conclusions

References



[1] Henri Berestycki, Jérôme Busca, and Igor Florent
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Quantitative Finance Vol 2, pp.61-69, 2002.



[2] Jim Gatheral.
The Volatility Surface: A Practitioner's Guide.
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[3] Jim Gatheral, Elton P Hsu, Peter Laurence, Cheng Ouyang, and Tai-Ho Wang
Asymptotics of implied volatility in local volatility models
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[4] Pierre Henry-Labordère,
Analysis, Geometry, and Modeling in Finance: Advanced Methods in Option Pricing.
CRC Press, 2009.

Option

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- Payoff of a call option (with strike K) = $(S - K)^+$

Black-Scholes model

- Assume the price of the underlying asset follows the SDE

$$\frac{dS_t}{S_t} = \mu dt + \sigma dW_t$$

μ : expected return, σ : volatility (μ and σ are constants)

dW : standard Brownian motion

- For each time t , S_t is lognormally distributed. More precisely,

$$S_t \sim S_0 \exp \left[\left(\mu - \frac{\sigma^2}{2} \right) t + \sigma \sqrt{t} N(0, 1) \right]$$

where $N(0, 1)$ is the standard normal distribution.

What is the fair price?

Assume the price of a call option C is a (smooth enough) function of the calendar time t and the underlying asset S . Consider the portfolio Π consists of selling a call option and holding Δ amounts of S .

- The value of Π at time t is

$$\Pi_t = C(t, S_t) - \Delta S_t$$

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- the infinitesimal change of Π reads

$$d\Pi_t = dC_t - \Delta dS_t$$

Itô's formula

- Itô's formula yields

$$\begin{aligned}dC(t, S_t) &= C_t dt + C_S dS_t + \frac{1}{2} C_{SS} (dS_t)^2 \\ &= C_S \sigma S dW_t + \left(C_t + \frac{1}{2} \sigma^2 S^2 C_{SS} + \mu S C_S \right) dt\end{aligned}$$

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- Hence the infinitesimal change of Π at time t becomes

$$\begin{aligned} d\Pi_t &= dC_t - \Delta dS_t \\ &= \left[C_t + \frac{1}{2} \sigma^2 S^2 C_{SS} + \mu S (C_S - \Delta) \right] dt + \sigma S (C_S - \Delta) dW_t \end{aligned}$$

Delta hedge

Let $\Delta = C_S$, i.e., hold this amount $C_S(t, S_t)$ of underlying assets in the portfolio Π . Then the infinitesimal change of Π becomes

- $d\Pi_t = \left(C_t + \frac{1}{2}\sigma^2 S^2 C_{SS} \right) dt$

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- $d\Pi_t = \left(C_t + \frac{1}{2}\sigma^2 S^2 C_{SS} \right) dt$
- On the other hand, with this choice of Δ , Π is riskless (non-random) hence must be like cash in bank account (Arbitrage Pricing Theory), i.e.,

$$d\Pi_t = r\Pi_t dt = r(C - \Delta S)dt = r(C - C_S S)dt,$$

where r is the interest rate (assumed constant).

Black-Scholes terminal-boundary value problem

Hence we conclude that the price C of a call option satisfies

$$\frac{\partial C}{\partial t} + \frac{\sigma^2}{2} S^2 \frac{\partial^2 C}{\partial S^2} + rS \frac{\partial C}{\partial S} - rC = 0, \text{ for } 0 < S < \infty, \quad 0 < t < T$$

with terminal condition

$$C(T, S) = (S - K)^+$$

and boundary conditions

$$C(t, 0) = 0$$

$$C(t, S) \sim S - Ke^{-r(T-t)} \text{ as } S \rightarrow \infty$$

Black-Scholes equation

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- $\tau = T - t$

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- $\xi = \ln S$

$$\frac{\partial C}{\partial \tau} = \frac{\sigma^2}{2} \frac{\partial^2 C}{\partial \xi^2} + \left(r - \frac{\sigma^2}{2} \right) \frac{\partial C}{\partial \xi} - rC$$

Black-Scholes equation

- $c(\xi, \tau) = e^{r\tau} C(\xi, \tau)$

$$\frac{\partial c}{\partial \tau} = \frac{\sigma^2}{2} \frac{\partial c^2}{\partial \xi^2} + \left(r - \frac{\sigma^2}{2} \right) \frac{\partial c}{\partial \xi}$$

Black-Scholes equation

- $c(\xi, \tau) = e^{r\tau} C(\xi, \tau)$

$$\frac{\partial c}{\partial \tau} = \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial \xi^2} + \left(r - \frac{\sigma^2}{2} \right) \frac{\partial c}{\partial \xi}$$

- $x = \xi + \left(r - \frac{\sigma^2}{2} \right) \tau$

$$\frac{\partial c}{\partial \tau} = \frac{\sigma^2}{2} \frac{\partial^2 c}{\partial x^2}$$

Black-Scholes equation

In total, we have done the transformation

$$\tau = T - t$$

$$x = \ln S + \left(r - \frac{\sigma^2}{2} \right) (T - t)$$

$$c = e^{r(T-t)} C$$

which transforms Black-Scholes equation to heat equation.

Black-Scholes formula

Hence the price $C(t, S)$ of a call option is found to be

$$C(t, S) = SN(d_1) - Ke^{-r(T-t)}N(d_2),$$

where $N(\cdot)$ is the distribution function of the standard normal random variable, i.e.,

$$N(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\xi^2/2} d\xi$$

and

$$d_1 = \frac{\ln \frac{S}{K} + (r + \frac{\sigma^2}{2})(T - t)}{\sigma\sqrt{T - t}}$$

$$d_2 = d_1 - \sigma\sqrt{T - t}$$

Merton

Merton's arguments in his 1973 paper imply more generally that the arbitrage-free value C of many derivatives satisfies

$$\frac{\partial C}{\partial t} + \frac{\sigma^2(S, t)}{2} S^2 \frac{\partial^2 C}{\partial S^2} + \mu(S, t) \frac{\partial C}{\partial S} - r(S, t) C = 0$$

with three variable coefficients $\sigma(S, t)$, $\mu(S, t)$ and $r(S, t)$. However, closed form solutions as explicit as in Black-Scholes model is in general not available.

Implied volatility

- In Black-Scholes' world, to compute the price of a call option, the only unknown parameter is σ , which is termed as *volatility*. Therefore, if we assume the underlying asset follows a geometric Brownian motion and somehow we manage to estimate the volatility σ , then the option price is given by the Black-Scholes' formula.

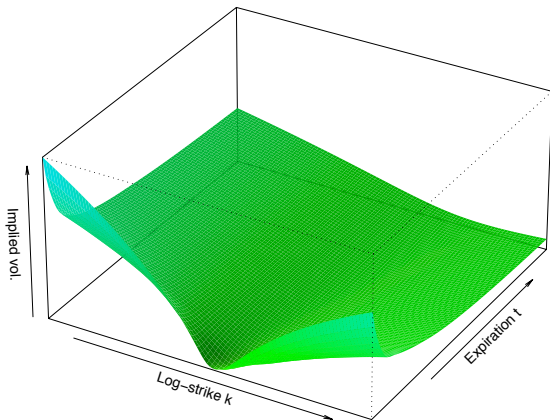
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- On the other hand, if we are given a call price, we can invert the Black-Scholes' formula to fetch out the volatility σ , assuming all the other parameters stays the same. (Exercise: Black-Scholes' formula as a function of σ is strictly increasing.) This is termed as *implied volatility*.

Implied volatility

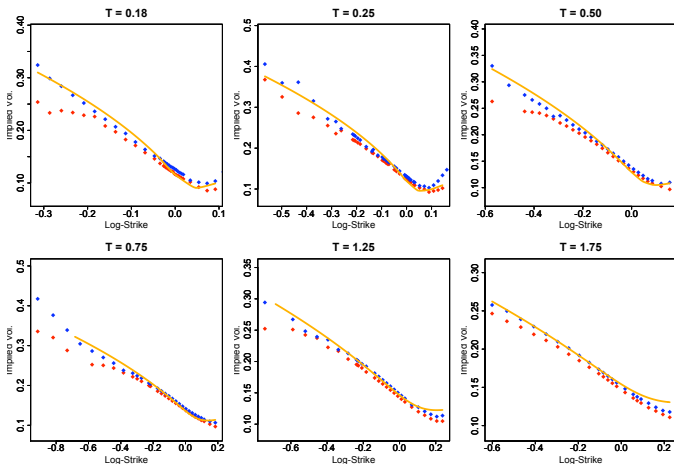
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- On the other hand, if we are given a call price, we can invert the Black-Scholes' formula to fetch out the volatility σ , assuming all the other parameters stays the same. (Exercise: Black-Scholes' formula as a function of σ is strictly increasing.) This is termed as *implied volatility*.
- Therefore, should the market quotes behave as Black-Scholes postulated, the implied volatility would have been flat, i.e., no matter what K and T are, their implied volatilities would be more or less the same. In real world, this is not the case, such nonflat implied volatilities phenomena is dubbed *volatility skew/smile*.

A 3D plot of the SPX volatility surface as of September 15, 2005



$k := \log K/F$ is the log-strike and t is time to expiry.

Slices of the SPX volatility surface



Orange lines are from PDE computations, red and blue points are empirical bid and offered vols respectively.

Objective

Given a local volatility process

$$\frac{dS}{S} = \sigma(S, t) dW_t,$$

with $\sigma(S, t)$ depending only on the underlying level S and the time t , we want to compute implied volatilities $\sigma_{bs}(K, T)$ such that

$$C_{bs}(s, t, K, T, \sigma_{bs}(K, T)) = \mathbb{E} [(S_T - K)^+ | S_t = s]$$

or in words, we want to efficiently compute implied volatility from local volatility.

- Knowing how to get implied volatility from local volatility helps us get accurate approximations to implied volatility in more complex models such as SABR.
 - Efficient calibration of complex models becomes practical.

Call price

Let $p(t, s; t', s')$ be the transition probability density. Then

$$\begin{aligned} C(s, t, K, T) &= \mathbb{E} [(S_T - K)^+ | S_t = s] \\ &= \int (s' - K)^+ p(t, s; T, s') ds' \end{aligned}$$

As a function of t and s , p satisfies the backward Kolmogorov equation:

$$Lp := p_t + \frac{1}{2} s^2 \sigma^2(s, t) p_{ss} = 0,$$

Subindices refer to respective partial derivatives.

Two to approximate

$$C(s, t, K, T) = \int (s' - K)^+ p(t, s; T, s') ds'$$

- Approximate transition density by heat kernel expansion.
- Approximate the integral.
 - Two approaches for approximating the integral lead to one expansion.
- The smaller the time to maturity, the better the approximation, for both approximations.

Heat kernel expansion

Heat kernel expansion for transition density $p(t, s; t', s')$ when $t' - t$ is small:

$$p(t, s; t', s') \sim \frac{e^{-\frac{d^2(s, s', t)}{2(t' - t)}}}{\sqrt{2\pi(t' - t)s'\sigma(s', t')}} \left[\sum_{k=0}^n H_k(t, s, s')(t' - t)^k \right]$$

- $d(s, s', t) = \left| \int_s^{s'} \frac{d\xi}{\xi\sigma(\xi, t)} \right|$: geodesic distance between s to s'
- $H_0(t, s, s') = \sqrt{\frac{s\sigma(s, t)}{s'\sigma(s', t)}} \exp \left[\int_s^{s'} \frac{d_t(\eta, s', t)}{\eta\sigma(\eta, t)} d\eta \right]$
- $H_i(t, s, s') = \frac{H_0(t, s, s')}{d^i(s, s', t)} \int_{s'}^s \frac{d^{i-1}(\eta, s', t) LH_{i-1}}{H_0(\eta, s', t)a(\eta, t)} d\eta$

Heat kernel expansion for Black-Scholes

Heat kernel expansion for Black-Scholes transition density $p_{bs}(t, s; t', s')$ when $t' - t$ is small:

$$p_{bs}(t' - t, s, s') = \frac{e^{-\frac{d_{bs}^2(s, s')}{2(t' - t)}}}{\sqrt{2\pi(t' - t)}\sigma_{bs}s'} \sqrt{\frac{s}{s'}} \sum_{k=0}^{\infty} \frac{(-1)^k}{k!} \left[\frac{\sigma_{bs}^2(t' - t)}{8} \right]^k$$

- $d_{bs}(s, s') = \left| \int_s^{s'} \frac{d\xi}{\sigma_{bs}\xi} \right| = \frac{1}{\sigma_{bs}} \left| \log \frac{s'}{s} \right|$
- $H_0^{bs}(t, s, s') = \sqrt{\frac{s}{s'}}$

Main idea

Implied volatility σ_{bs} is defined as the unique solution to

$$C(s, t, K, T) = C_{bs}(s, t, K, T, \sigma_{bs})$$

- Substitute the transition density by the heat kernel expansion for both the model price C and the Black-Scholes price C_{bs}
- Expand in terms of $T - t$ on both sides of the resulting equation
- Further expand on Black-Scholes side the implied volatility

$$\sigma_{bs}(K, T) \approx \sigma_{bs,0} + \sigma_{bs,1}(T - t) + \sigma_{bs,2}(T - t)^2$$

- Match the corresponding coefficients

Two approaches

- Directly substitute the transition density by heat kernel expansion to call price. Use Laplace asymptotic formula to approximate the resulting integral.
- Rewrite call price as intrinsic value + time value. Further rewrite time value as an integral of transition density over time, i.e., the Carr-Jarrow formula:

$$C(s, t, K, T) = (s - K)^+ + \int_t^T K^2 \sigma^2(K, u) p(s, t; K, u) du$$

Laplace asymptotic formula

Asymptotic expansion of the integral as $\tau \rightarrow 0^+$

$$\int_0^{\infty} e^{-\frac{\phi(x)}{\tau}} f(x) dx \sim \tau^2 e^{-\frac{\phi(x^*)}{\tau}} \left[\frac{f'(x^*)}{[\phi'(x^*)]^2} + \left(\frac{f'(x^*)}{[\phi'(x^*)]^3} \right)' \tau \right]$$

Assumptions:

- f is identically zero when $0 \leq x \leq x^*$.
- ϕ is increasing in $[x^*, \infty)$.

Laplace asymptotic for call price

Let $\tau = T - t$.

$$\begin{aligned}
 C(s, t, K, T) &= \int_0^\infty (s - K)^+ p(t, s; T, s') ds' \\
 &\sim \frac{1}{\sqrt{2\pi\tau}} \int_0^\infty (s' - K)^+ \frac{e^{-\frac{d^2(s, s', t)}{2\tau}}}{s' \sigma(s', T)} \sum_{k=0}^n H_k(t, s, s') \tau^k ds' \\
 &= \frac{1}{\sqrt{2\pi\tau}} \sum_{k=0}^n \int_K^\infty e^{-\frac{d^2(s, s', t)}{2\tau}} G_k(t, s, T, s') ds' \cdot \tau^k
 \end{aligned}$$

- $G_k(t, s, T, s') = (s' - K) \frac{H_k(t, s, s')}{s' \sigma(s', T)}$

Laplace asymptotic for call price

Assume $s < K$.

$$\frac{1}{\sqrt{2\pi\tau}} \int_K^\infty e^{-\frac{d^2(s,s',t)}{2\tau}} G_k(t, s, T, s') ds'$$

$$\sim \frac{\tau^{\frac{3}{2}}}{\sqrt{2\pi}} e^{-\frac{d^2}{2T}} \left[\frac{G'_k}{(dd')^2} + \left(\frac{G'_k}{(dd')^3} \right)' \tau \right],$$

- $d = d(s, K, t)$, $d' = \frac{\partial d}{\partial s'}(s, K, t)$, and $d'' = \frac{\partial^2 d}{\partial (s')^2}(s, K, t)$
- $G'_k = \frac{\partial G_k}{\partial s'}(t, s, T, K) = \frac{H_k(t, s, K)}{K\sigma(K, T)}$

Laplace asymptotic for call price

Laplace asymptotic for model price:

$$C(s, t, K, T) \sim \frac{\tau^{\frac{3}{2}}}{\sqrt{2\pi}} e^{-\frac{d^2}{2\tau}} \left[\frac{G'_0}{(dd')^2} + \left\{ \left(\frac{G'_0(K)}{(dd')^3} \right)' + \frac{G'_1(K)}{(dd')^2} \right\} \tau \right].$$

- $d = d(s, K, t)$, $d' = \frac{\partial d}{\partial s'}(s, K, t)$, and $d'' = \frac{\partial^2 d}{\partial (s')^2}(s, K, t)$
- $G'_k = \frac{\partial G_k}{\partial s'}(t, s, T, K) = \frac{H_k(t, s, K)}{K\sigma(K, T)}$

Laplace asymptotic for Black-Scholes: $k = \log \frac{K}{s}$

$$C_{bs}(s, t, K, T, \sigma_{bs}) \sim \frac{Ke^{-\frac{k}{2}}}{\sqrt{2\pi}} e^{-\frac{k^2}{2\sigma_{bs}^2\tau}} \frac{\sigma_{bs}^3\tau^{\frac{3}{2}}}{k^2} \left[1 - \left(\frac{1}{8} + \frac{3}{k^2} \right) \sigma_{bs}^2\tau \right]$$

Match the coefficients

Let $\sigma_{bs} = \sigma_{bs,0} + \sigma_{bs,1}\tau + \sigma_{bs,2}\tau^2 + \dots$ and set

$$\begin{aligned} & e^{-\frac{d^2}{2\tau}} \left[\frac{G'_0}{(dd')^2} + \left\{ \left(\frac{G'_0(K)}{(dd')^3} \right)' + \frac{G'_1(K)}{(dd')^2} \right\} \tau \right] \\ &= e^{-\frac{k^2}{2\sigma_{bs}^2\tau}} \frac{K\sigma_{bs}^3}{k^2 e^{\frac{k}{2}}} \left[1 - \left(\frac{1}{8} + \frac{3}{k^2} \right) \sigma_{bs}^2 \tau \right] \end{aligned}$$

- Exponential term: $d^2 = \frac{k^2}{\sigma_{bs,0}^2} \implies \sigma_{bs,0} = \frac{k}{d} = \frac{\log K - \log s}{d(s, K, t)}$

- Zeroth order term:

$$\frac{G'_0}{(dd')^2} = e^{\frac{k^2\sigma_{bs,1}}{\sigma_{bs,0}^3}} \frac{K\sigma_{bs,0}^3}{k^2 e^{\frac{k}{2}}} \implies \sigma_{bs,1} = \frac{k}{d^3} \log \left[\frac{dG'_0 e^{-\frac{k}{2}}}{Kk(dd')^2} \right]$$

Time value

Recall

$$C(s, t, K, T) = (s - K)^+ + \int_t^T K^2 \sigma^2(K, u) p(s, t; K, u) du$$

$$\sim (s - K)^+ + \sum_{k=0}^n \int_t^T \frac{e^{-\frac{d^2(s, K, t)}{2(u-t)}}}{\sqrt{2\pi(u-t)}} K \sigma(K, u) (u-t)^k du \cdot H_k(t, s, K)$$

Moreover, denote $d = d(s, K, t)$,

$$\int_t^T e^{-\frac{d^2}{2(u-t)}} \sigma(K, u) (u-t)^{k-\frac{1}{2}} du$$

$$\sim \int_t^T e^{-\frac{d^2}{2(u-t)}} [\sigma(K, t) + \sigma_t(K, t)(u-t)] (u-t)^{k-\frac{1}{2}} du$$

Expansion for call price

$$\text{Let } \Phi_k(d, \tau) = \int_0^t u^{k-\frac{1}{2}} e^{-\frac{d^2}{2u}} du.$$

$$\begin{aligned} & C(s, t, K, T) - (s - K)^+ \\ & \sim \frac{1}{2\sqrt{2\pi}} \{ K\sigma(K, t)\Phi_0(d, \tau)H_0(t, s, K) \\ & \quad + K[\sigma_t(K, t)H_0(t, s, K) + \sigma(K, t)H_1(t, s, K)]\Phi_1(d, \tau) \} \end{aligned}$$

Moreover, on Black-Scholes side,

$$\begin{aligned} & C_{bs}(s, t, K, T) - (s - K)^+ \\ & \sim \frac{\sqrt{sK}}{2\sqrt{2\pi}} \left[\sigma_{bs}\Phi_0(d_{bs}, \tau) - \frac{\sigma_{bs}^3}{8}\Phi_1(d_{bs}, \tau) \right] \end{aligned}$$

Auxiliary expansion and matching

Expanding the Φ_i 's:

- $\Phi_0(d, \tau) \sim 2\tau^{\frac{3}{2}} \left[\frac{1}{d^2} - 3\frac{\tau}{d^4} \right] e^{-\frac{d^2}{2\tau}}$

- $\Phi_1(d, \tau) = \frac{2}{3}\tau^{\frac{3}{2}} e^{-\frac{d^2}{2\tau}} - \frac{d^2}{3}\Phi_0(d, \tau) \sim \frac{2\tau^{\frac{5}{2}}}{d^2} e^{-\frac{d^2}{2\tau}}$

Matching

$$e^{-\frac{d^2(s, K, t)}{2\tau}} \left\{ \frac{K\sigma H_0}{d^2} + \left[\frac{K\sigma_t H_0 + K\sigma H_1}{d^2} - \frac{3K\sigma H_0}{d^4} \right] \tau \right\}$$

$$= e^{-\frac{d_{bs}^2(s, K, t)}{2\tau}} \sqrt{sK} \left[\sigma_{bs} \Phi_0(d_{bs}, \tau) - \frac{\sigma_{bs}^3}{8} \Phi_1(d_{bs}, \tau) \right]$$

Asymptotic expansion once again

$$\sigma_{bs} = \sigma_{bs,0} + \sigma_{bs,1}(T - t) + \sigma_{bs,2}(T - t)^2 + \mathcal{O}(T - t)^3.$$

$$d(s, K, t) = \int_s^K \frac{d\xi}{\xi\sigma(\xi, t)},$$

$$H_0(s, K, t) = \sqrt{\frac{s\sigma(s, t)}{K\sigma(K, t)}} \exp \left[\int_s^K \frac{d_t(\eta, K, t)}{\eta\sigma(\eta, t)} d\eta \right].$$

- $\sigma_{bs,0} = \frac{|\log K - \log s|}{d(s, K, t)}$. (BBF)
- $\sigma_{bs,1} = \frac{k}{d^3} \log \left[\frac{dH_0\sqrt{K}\sigma(K, t)}{k\sqrt{s}} \right]$, where $k = \log K - \log s$.
- $\sigma_{bs,2}$? Too complicated to reproduce here.

Henry-Labordère's approximation

Henry-Labordère also presents a heat kernel expansion based approximation to implied volatility in equation (5.40) on page 140 of his book [4]:

$$\sigma_{BS}(K, T) \approx \sigma_0(K) \left\{ 1 + \frac{T}{3} \left[\frac{1}{8} \sigma_0(K)^2 + Q(f_{av}) + \frac{3}{4} G(f_{av}) \right] \right\} \quad (1)$$

with

$$Q(f) = \frac{C(f)^2}{4} \left[\frac{C''(f)}{C(f)} - \frac{1}{2} \left(\frac{C'(f)}{C(f)} \right)^2 \right]$$

and

$$G(f) = 2 \partial_t \log C(f) = 2 \frac{\partial_t \sigma(f, t)}{\sigma(f, t)}$$

where $C(f) = f \sigma(f, t)$ in our notation, $f_{av} = (S_0 + K)/2$ and the term $\sigma_0(K)$ is the BBF approximation from [1].

How well do these approximations work?

We consider the following explicit local volatility models:

- The square-root CEV model:

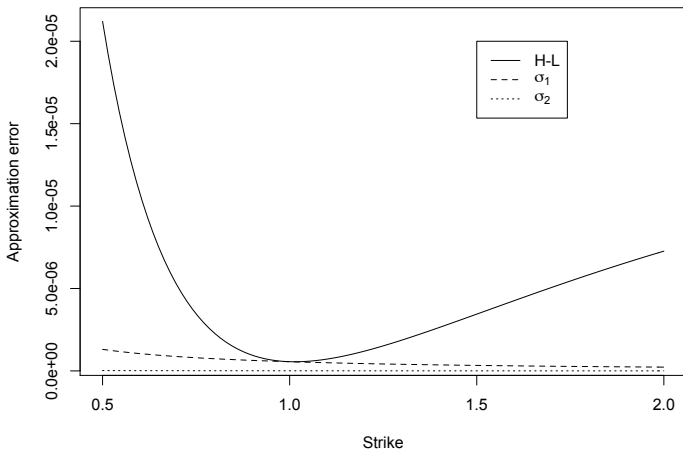
$$dS_t = e^{-\lambda t} \sigma \sqrt{S_t} dW_t$$

- The quadratic model:

$$dS_t = e^{-\lambda t} \sigma \left\{ 1 + \psi (S_t - 1) + \frac{\gamma}{2} (S_t - 1)^2 \right\} dW_t$$

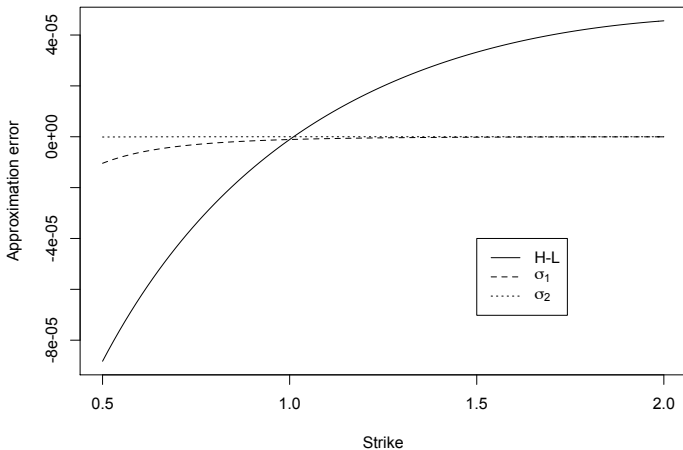
- Parameters are: $\sigma = 0.2$, $\psi = -0.5$ and $\gamma = 0.1$. In each case $S_0 = 1$ and $T = 1$.
- $\lambda = 0$ gives a time-homogeneous local volatility surface and $\lambda = 1$ a time-inhomogeneous one.
- We compare implied volatilities from the approximations and the closed-form solution.

Time-homogeneous Square Root CEV

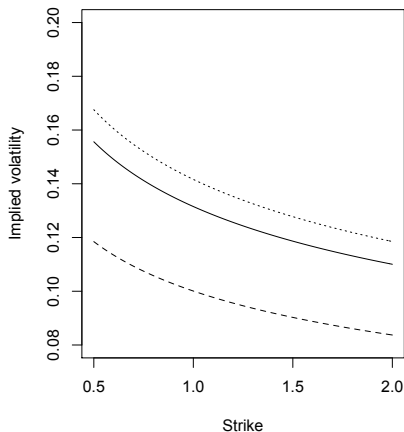
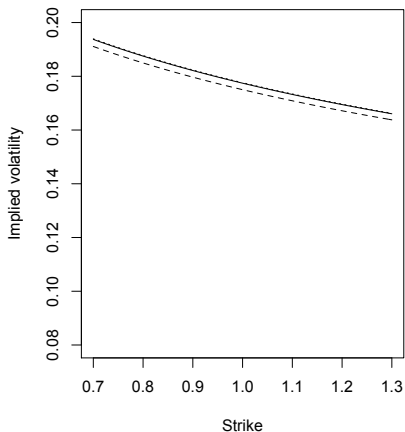


Note that all errors are tiny!

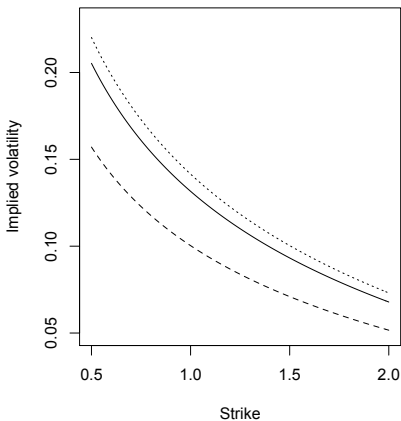
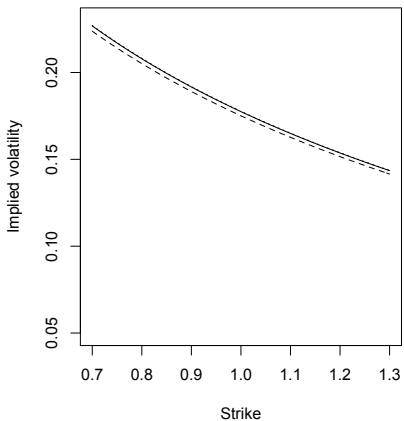
Time-homogeneous Quadratic Model



Time-inhomogeneous Square Root CEV



Time-inhomogeneous Quadratic Model



Summary

- Small-time expansions are useful for generating closed-form expressions for implied volatility from simple models.
- Direct substitute approach is easier for generalization to higher dimensions, e.g., stochastic volatility models.

$$\sigma_{bs} \sim \frac{\log K - \log s}{d_M(s, v)},$$

where $d_M(s, v)$ is the “distance to the money”, i.e., shortest geodesic distance from the spot (s, v) to the line $\{s = K\}$ in the price-volatility plane.

- Application: Short time implied vol in delta is flat! (Joint work with Carr and Lee).

Summary II

- Time value approach is easier for getting higher order terms.
- Refinement of $\sigma_{bs,0}$ (joint work with Gatheral):

$$\sigma_{bs} \sim \left[\frac{\sqrt{T-t}}{|\log K - \log s|} \sqrt{\int_t^T \left| \frac{s'(\tau)}{a(s(\tau), \tau)} \right|^2 d\tau} \right]^{-1}$$

where the integral is along the “most likely path” $s(\tau)$.

- If we take the “likely path” as $s(\tau) = \varphi_t^{-1} \left(\frac{\tau}{T} \varphi_t(K) \right)$, where $\varphi_t(x) = \int_s^x \frac{d\xi}{a(\xi, t)}$, then BBF is recovered.

THANK YOU FOR YOUR ATTENTION.