Smile interpolation and calibration of the local volatility model

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Abstract

The Black-and-Scholes formula provides a correspondence between the price of a plain option and the underlying asset volatility. Volatilities implied from prices of quoted plain options are, in general, not constant and depend on the strike and maturity of the option. We describe a new construction of an implied volatility surface from a discrete set of implied volatilities which is arbitrage-free and satisfies some smoothness conditions. Our algorithm allows the calibration to the smile of the local volatility model, a standard extension of the Black-and-Scholes model known to be hard to calibrate in practice. Plain options prices calculated from our local volatility surface using deterministic schemes or Monte Carlo simulation closely match input prices. This allows the pricing of exotic options in a way consistent with the price of quoted plain options.

1 Introduction

Plain or vanilla European options deliver payoffs dependent on the underlying asset price at the option maturity. The Black-and-Scholes formula determines the price of a vanilla European option by assuming the underlying volatility to be constant. It is well known, however, that implied volatilities of quoted European options are non-constant and depend both on the strike and maturity of the option, a phenomenon often referred to as the "smile". Various models based on jump-diffusion, local or stochastic volatility, have been proposed to explain and calibrate the smile (see (Andersen and Andreasen 2000, Derman and Kani 1994, Dumas, Fleming and Whaley 1998, Dupire 1994, Heston 1993, Lagnado and Osher 1997, Li 2001, Rebonato 1999, Rubinstein 1994) and references therein.) Several problems arise in the presence of smile. First, arbitrage may exist among the quoted options. Another problem is to price European options for strikes and maturities not quoted in the market. Standard interpolation techniques may give rise to arbitrage in the interpolated volatility surface even if there is no arbitrage in the original set. A related problem is to price non-vanilla options by taking the smile into account.

This paper presents a new interpolation method for implied volatilities. If the market volatilities are arbitrage-free we calculate an interpolating surface of the market volatilities for all strikes and maturities up to the last maturity that is arbitrage-free and satisfies some smoothness conditions. The basis block for
our interpolation is a single-maturity interpolation with the following properties (the first holds only if there are no absolute discrete dividends):

- If the input implied volatilities are constant so are the interpolated volatilities.
- The second derivative of the call price with respect to the strike is positive and continuous, as shown by numerical experiments.

The motivation behind the second property is that the second derivative of the call price is proportional to the implied density of the spot.

Our single-maturity interpolation does not depend on the shape of the discrete volatilities and applies to index, equity, forex and interest rate options. It takes seconds to calibrate a $10 \times 10$ volatilities matrix on a 800 Mhz processor and the quality of the fit is excellent. While our interpolation is of independent interest one of its main applications is the calibration of the local volatility model (Derman and Kani 1994, Dupire 1994, Rubinstein 1994), where the volatility of the spot is a deterministic function of the spot and time. The local volatilities can be calculated from the implied volatility surface via Dupire’s formula (Dupire 1994) which is very sensitive to the interpolation used. It is well known (Avellaneda, Friedman, Holmes and Samperi 1997) that, for standard interpolation methods, Dupire’s formula often leads to instabilities in the local volatilities. Our interpolated volatility surface has been designed to calibrate Dupire’s model. Numerical experiments show that prices of plain options calculated via our local volatility surface and deterministic schemes or Monte Carlo simulation are very close to input prices. This allows the pricing of exotic options, including options on several assets, in a way consistent with the smile.

An alternative approach (Achdou and Pironneau 2002, Avellaneda, Friedman, Holmes and Samperi 1997, Coleman, Li and Verma 1999, Crépey 2003a, Crépey 2003b, Lagnado and Osher 1997) to calibrating Dupire’s model is to calculate directly a local volatility surface that satisfies a regularity condition and produces prices close to input prices. This approach generates an arbitrage-free implied volatility surface but is more time-consuming than ours.

The rest of the paper is organised as follows. Section 2 contains preliminary results. Section 3 gives an arbitrage-free one-dimensional interpolation which has a provably continuous first derivative. Section 4 gives an arbitrage-free one-dimensional interpolation which numerical experiments show to have a continuous second derivative. Section 5 describes our two-dimensional interpolation algorithm and its application to the local volatility model. Sections 2 through 5 assume we are in the equity market and there are no interest rates and no dividends. We point out how dividends and interest rates are taken into account in Section 6. We show in Section 7 that our method can be used in the forex market and, in the one-dimensional case, in the interest rate market. An example from (Avellaneda, Friedman, Holmes and Samperi 1997) on the USD/DEM exchange rate is given in Section 8. An example on the S&P 500 has been given in a short version of our paper (Kahale 2004). Section 9 contains concluding remarks.
2 Preliminaries

We use the following definition of arbitrage, which is slightly different from the one usually found in the literature. We define an arbitrage as a self-financing portfolio of securities that has a negative value today and a nonnegative value at a given time in the future independently of the market behavior. Thus one is certain to make profit by buying such a portfolio.

It can be shown that if the input implied volatilities are given for all maturities and strikes there is no arbitrage in the input if and only if the following conditions hold:

1. For a given maturity the call price is non-increasing and convex with respect to the strike.
2. The call price is a non-decreasing function of time.

The convexity condition follows from the convexity of the payoff at maturity. Based upon this result one can find in linear time whether there exists an arbitrage within a discrete set of implied volatilities. In particular the following can be shown in the one-dimensional case.

**Lemma 1.** Consider a sequence \((k_i, c_i)_{0 \leq i \leq n+1}\) such that

\[
0 = c_{n+1} = k_0 < k_1 < \cdots < k_n < k_{n+1} = \infty
\]

and \(c_i, 0 \leq i \leq n,\) is the price of a call with strike \(k_i.\) There is no arbitrage among these prices if and only if \(c_0\) is equal to the current spot, \(c_n \geq 0\) and

\[
-1 \leq \frac{c_i - c_{i-1}}{k_i - k_{i-1}} \leq \frac{c_{i+1} - c_i}{k_{i+1} - k_i} \leq 0 \text{ for } 1 \leq i < n.\]

3 A one-dimensional \(C^1\) interpolation method

We give in this section a \(C^1\) arbitrage-free interpolation method for a given maturity. Like the cubic interpolation, our method is based on the concatenation of several functions. Moreover, these functions are convex. Our construction is inspired from the Black-and-Scholes formula. We start with the following lemma:

**Lemma 2.** Given \(f > 0, \Sigma > 0, a\) and \(b,\) the function

\[
c(k) = cf,\Sigma,a,b(k) = fN(d_1) - kN(d_2) + ak + b,
\]

where

\[
d_1 = \frac{\log(f/k) + \Sigma^2/2}{\Sigma}
\]

and \(d_2 = d_1 - \Sigma,\) is a convex function of \(k\) for \(k > 0.\)

**Proof.** As in the Black-and-Scholes formula, we can check that the second derivative is positive using simple differentiation:

\[
\begin{align*}
c'(k) & = -N(d_2) + a \quad (4) \\
c''(k) & = \frac{N'(d_2)}{k\Sigma} \quad (5)
\end{align*}
\]

\(\square\)
Lemma 3. Let \( g \) be a real function defined for all real numbers such that \( g'(x) \) exists and is positive for any real number \( x \) and \( 1/g' \) is strictly convex. For \( \lambda \in [0,1] \) and any real numbers \( x_0 < x_1 \), the function \( h(a) = g(\lambda g^{-1}(a + x_0) + (1 - \lambda) g^{-1}(a + x_1)) - a \) has a positive derivative with respect to \( a \) on the (possibly empty) interval on which it is defined.

Proof. Let \( y = g^{-1}(a + x_0) \) and \( z = g^{-1}(a + x_1) \). By standard calculus

\[
h'(a) = g'(\lambda y + (1 - \lambda) z) \left( \frac{\lambda}{g'(y)} + \frac{1 - \lambda}{g'(z)} \right) - 1.
\]

The equation \( h'(a) > 0 \) is equivalent to

\[
\frac{\lambda}{g'(y)} + \frac{1 - \lambda}{g'(z)} > \frac{1}{g'(\lambda y + (1 - \lambda) z)}.
\]

which follows from the strict convexity of \( 1/g' \). \(\square\)

Theorem 1. For all real numbers \( k_0, k_1, c_0, c_1, c_0', \) and \( c_1' \) such that \( 0 < k_0 < k_1 \) and

\[
c_0' < \frac{c_1 - c_0}{k_1 - k_0} < c_1' < 1 + c_0'
\]

there exists a unique vector \((f, \Sigma, a, b)\) with \( f > 0, \Sigma > 0 \) such that the function \( c = c_{f, \Sigma, a, b} \) satisfies the following conditions: \( c(k_0) = c_0 \), \( c(k_1) = c_1 \), \( c'(k_0) = c_0' \) and \( c'(k_1) = c_1' \). The vector \((f, \Sigma, a, b)\) is continuous with respect to \((k_0, k_1, c_0, c_1, c_0', c_1')\) and can be calculated numerically.

Proof. Given \( a \in ]c_1', 1 + c_0[ \), let \( d_0^0, d_2^0, \alpha \) and \( \beta \) be the unique real numbers such that

\[
c_0' = - N(d_2^0) + a
\]

and

\[
d_2^0 = \alpha \log(k_1) + \beta,
\]

for \( i \in \{0, 1\} \). The existence of \( d_2^0 \) is a consequence of the inequalities \( c_0' < a < c_1' + 1 \) which follow from Eq. 6. Since \( c_0' < c_1' \) we have \( d_2^0 > d_2^1 \) by Eq. 7, and thus

\[
\alpha = \frac{d_2^0 - d_2^1}{\log(k_0) - \log(k_1)} < 0.
\]

Hence there exist \( f > 0 \) and \( \Sigma > 0 \) such that \( \alpha = - \frac{1}{\Sigma} \) and \( \beta = (\log f)/\Sigma - \Sigma/2 \).

It follows that

\[
d_2^0 = \frac{\log(f/k_0) - \Sigma^2/2}{\Sigma}.
\]

Note that \( d_2^0, \alpha, \beta, f \) and \( \Sigma \) are continuous functions of \( a \) as \( a \) ranges in \( ]c_1', 1 + c_0[ \). Consider the function \( c = c_{f, \Sigma, a, b} \), where \( b \) is chosen so that \( c(k_0) = c_0 \). It follows from Eqs. 9 and 4 that \( c'(k_i) = - N(d_2^0) + a \), and so \( c'(k_i) = c_i' \) by Eq. 7. We now show that, for some \( a \in ]c_1', 1 + c_0[ \), \( c(k_1) = c_1 \). The ratio \( (c(k_1) - c(k_0))/(k_1 - k_0) \) is a continuous function of \( a \). It follows from Eq. 4 that, for \( k_0 < k < k_1 \),

\[
c'(k) = a - N(\lambda N^{-1}(a - c_0) + (1 - \lambda) N^{-1}(a - c_1)),
\]

where \( \lambda = \log(k_1/k_0)/\log(k_1/k_0) \). By applying Lemma 3 to the Normal function, we infer that \( c'(k) \) is a strictly decreasing and has a negative derivative with
Lemma 5. For all real numbers \( k_1, c_0, c_1 \) and \( c'_1 \) such that \( 0 < k_1 \) and \( c_0 > 0 \) there exist three unique parameters \( f > 0 \) and \( b > 0 \) such that the function \( c = c_{f, \Sigma, b} \) satisfies the following conditions: \( c(k) \to c_0 \) as \( k \to 0 \), \( c'(k) = c'_1 \), and \( c(k_1) = c_1 \). Moreover, \( c'(k) \to -1 \) as \( k \to 0 \). The vector \((f, \Sigma, b)\) is continuous with respect to \((k_1, c_0, c_1, c'_1)\) and can be calculated numerically.
Proof. Let \( d_2^1 \) be the unique real number such that \( c_1' = -N(d_2^1) \). For \( \Sigma > 0 \) let
\[
f = k_1 \exp(\Sigma d_1^1 + \Sigma^2/2)
\]  
and
\[
b = c_0 - f.
\]  
The function \( c = c_f, \Sigma, 0, b \) has a derivative equal to \( c_1' \) at \( k_1 \). Moreover, \((c(k), c'(k)) \to (c_0, -1)\) as \( k \to 0 \) by standard calculations. It remains to show that \( c(k_1) = c_1 \) for some \( \Sigma > 0 \). Since \( c(k_1) = fN(d_1) + k_1 c_1' + b \), with \( d_1 = d_2^1 + \Sigma \),
\[
c(k_1) - c_0 = -fN(-d_2^1 - \Sigma) + k_1 c_1'
\]  
by Eq. 12. Eq. 11 determines \( f \) as a function of \( \Sigma \). As \( \Sigma \to 0 \), \( f \to k_1 \) and
\[
c(k_1) - c_0 \to -k_1N(-d_2^1) + k_1 c_1' = -k_1.
\]  
Since \( N(x) \sim N'(x)/|x| \) as \( x \to -\infty \),
\[
c(k_1) - c_0 \to k_1 c_1' \text{ as } \Sigma \to \infty.
\]  
By continuity and Eq. 10 it follows that \( c(k_1) = c_1 \) when \( \Sigma = \Sigma_0 \), for some \( \Sigma_0 > 0 \). As in Lemma 4, it can be shown that the derivative of \( c(k) \) with respect to \( \Sigma \) is positive. This implies the uniqueness and continuity of \( \Sigma_0 \) with respect to \( (k_1, c_0, c_1, c_1') \).

Combining Lemmas 2, 4, 5 and Theorem 1, we obtain the following definition and theorem.

**Definition 1.** For all real numbers \( k_0, k_1, c_0, c_1, c_0', \) and \( c_1' \) such that \( 0 < k_0 < k_1 \) and Eq. 6 holds, denote by \( \phi(k_0, c_0, c_0', k_1, c_1, c_1') \) the function \( c = c_f, \Sigma, 0, b \) that satisfies the following conditions: \( c(k_0) = c_0, \ c(k_1) = c_1, \ c'(k_0) = c_0' \) and \( c'(k_1) = c_1' \). Similarly, for all real numbers \( k_0, c_0, \) and \( c_0' \) such that \( 0 < k_0, -1 < c_0 < 0 \) and \( c_0 > 0 \), denote by \( \phi(k_0, c_0, c_0', \infty, 0, 0) \) the function \( c = c_f, \Sigma, 0, 0 \) that satisfies the following conditions: \( c(k_0) = c_0, \ c'(k_0) = c_0' \). Finally, for all real numbers \( k_1, c_0, c_1 \) and \( c_1' \) such that \( 0 < k_1 \) and Eq. 10 holds, denote by \( \phi(0, c_0, -1, k_1, c_1, c_1') \) the function \( c = c_f, \Sigma, 0, b \) that satisfies the following conditions: \( c(k) \to c_0 \) as \( k \to 0, \ c'(k_1) = c_1', \ c(k_1) = c_1 \).

It follows from Definition 1 that for all real or infinite numbers \( k_0, k_1, c_0, c_1, c_0', \) and \( c_1' \) such that the function \( c = \phi(k_0, c_0, c_0', k_1, c_1, c_1') \) is defined, \( (c(k), c'(k)) \to (c_i, c_i') \) as \( k \to k_i, \ i \in \{0, 1\} \). By combining the interpolated and extrapolated methods for a series of strikes, we obtain the following.

**Theorem 2.** For all sequences \( (k_i)_{0 \leq i \leq n+1}, (c_i)_{0 \leq i \leq n+1} \) and \( (c_i')_{0 \leq i \leq n+1} \) such that Eq. 1 holds together with the limit conditions
\[
c_0' = -1, \ c_n' < c_{n+1}' = 0 < c_n,
\]  
and the convexity conditions
\[
c_i' \leq \frac{c_{i+1}' - c_i'}{k_{i+1} - k_i} < c_{i+1}' \text{ for } 0 \leq i < n,
\]  
6
there exist a $C^1$ convex function $c(k)$, $k > 0$, and a unique sequence $(f_i, \Sigma_i, a_i, b_i)_{0 \leq i \leq n}$ such that $c(k) = c_{f_i, \Sigma_i, a_i, b_i}(k)$ on the interval $[k_i, k_{i+1}] - \{0, \infty\}$, $c(k_i) = c_i$ and $c'(k_i) = c'_i$ for $1 \leq i \leq n$. Moreover,

$$\lim_{k \to 0} (c(k), c'(k)) = (c_0, -1)$$  \hspace{1cm} (16)

and

$$\lim_{k \to \infty} (c(k), c'(k)) = (0, 0).$$  \hspace{1cm} (17)

The sequence $(f_i, \Sigma_i, a_i, b_i)_{0 \leq i \leq n}$ is continuous with respect to $(c_0, (k_i, c_i, c'_i)_{1 \leq i \leq n})$ and can be calculated numerically.

There are $4(n+1)$ unknown parameters $(f_i, \Sigma_i, a_i, b_i)_{0 \leq i \leq n}$ that define the function $c$. Each equation $c(k_i) = c_i$, $1 \leq i \leq n$, accounts for two conditions on these parameters, because it holds both for $c_{f_i, \Sigma_i, a_i, b_i}(k_i)$ and $c_{f_{i-1}, \Sigma_{i-1}, a_{i-1}, b_{i-1}}(k_i)$. The same holds for the equation $c'(k_i) = c'_i$, $1 \leq i \leq n$. Together with the limit conditions as $k \to 0$ and as $k \to \infty$, there are $4n + 4$ conditions, which is equal to the number of parameters.

4 A one-dimensional $C^2$ interpolation method

The algorithm in Theorem 2 gives a $C^1$ convex interpolating curve at a given maturity. But, since the second derivative of the call price with respect to the strike is proportional to the density of the strike in the risk-neutral word, a $C^2$ interpolating curve is desired. We give in this section an interpolating algorithm for the call prices at a given maturity which we conjecture will converge towards a $C^2$ interpolating curve with the properties mentioned in Section 1. We start with a few lemmas to motivate our algorithm.

**Lemma 6.** Let $g$ be a convex $C^1$ function defined on the interval $[x_0, x_1]$, and $x_2 \in [x_0, x_1]$. The following holds:

$$g'(x_1) - g'(x_2) \leq \eta \left(\frac{g'(x_1) - g'(x_2)}{x_1 - x_0}\right),$$  \hspace{1cm} (18)

where $\eta = (x_1 - x_0)/(x_2 - x_0)$.

**Proof.** Since $g(x_2) - g(x_0) \leq g'(x_2)(x_2 - x_0)$ and $g(x_1) - g(x_2) \leq g'(x_1)(x_1 - x_2)$,

$$g(x_1) - g(x_0) \leq g'(x_2)(x_2 - x_0) + g'(x_1)(x_1 - x_2),$$

and so

$$g(x_1) - g(x_0) - g'(x_1)(x_1 - x_0) \leq (g'(x_2) - g'(x_1))(x_2 - x_0)$$

which is equivalent to Eq. 18. \hfill \Box

**Lemma 7.** For any $\gamma > 0$ there exists $\epsilon_0 > 0$ such that, for $\epsilon < \epsilon_0$, $\delta > 0$ and all $u$, if

$$N(u + \delta) - N(u) < \epsilon \quad \text{and} \quad N(u + 2\delta) - N(u) > \gamma$$  \hspace{1cm} (19)

then $\delta N'(u) < \epsilon$.  \hspace{1cm} (20)
Proof. Assume without loss of generality that $\epsilon_0 < \gamma/2 < 1$. By Eq. 20, $N(u + 2\delta) - N(u + \delta) > \gamma/2$ and so
\[ u + \delta < z_0, \tag{21} \]
where $z_0 = N^{-1}(1 - \gamma/2)$. Let $z \in [u, u + \delta]$ be such that $N(u + \delta) - N(u) = \delta N'(z)$. By Eq. 19,
\[ \delta N'(z) < \epsilon. \tag{22} \]
Since the function $N'$ is upper bounded by 1, it follows from Eq. 20 that $\delta > \gamma/2$, and so $N'(z) < 2\epsilon/\gamma$. But $z < z_0$ by Eq. 21. Let $\epsilon_0 = \gamma N'(z_0)/2$. If $\epsilon < \epsilon_0$ then $N'(z) < N'(z_0)$ and so $z < 0$. Hence $N'(u) \leq N'(z)$. By Eqs. 22 it follows that $\delta N'(u) < \epsilon$. \qed

Lemma 8. Given any real numbers $k_0$, $k_1$, $c_0$, $c_1$, $c'_0$ such that $0 < k_0 < k_1$ and
\[ c'_0 = \frac{c_1 - c_0}{k_1 - k_0} < 1 + c'_0, \tag{23} \]
let $c'_1 \in [(c_1 - c_0)/(k_1 - k_0), 1 + c'_0]$ and $c = \phi(k_0, c_0, c'_0, k_1, c_1, c'_1)$. Then $c''(k_1) \to 0$ as $c'_1 \to (c_1 - c_0)/(k_1 - k_0)$.

Proof. We show there exists $\epsilon_0, \theta > 0$ such that, for $0 < \epsilon < \epsilon_0$ if
\[ \frac{c_1 - c_0}{k_1 - k_0} < c'_1 < \frac{c_1 - c_0}{k_1 - k_0} + \epsilon \]
then $c''(k_1) < \theta \epsilon$. By choosing $\epsilon_0 < 1 + c'_0 - (c_1 - c_0)/(k_1 - k_0)$, the existence of $c = c_{f, \Sigma, a, b}$ follows from Theorem 1. Let $\gamma = (c_1 - c_0)/(k_1 - k_0) - c'_0$. By convexity $c''(k_1) - c''(k_0) > \gamma$. Using the same notation as in Theorem 1 and Eq. 4, it follows that
\[ N(d'_1) - N(d'_0) > \gamma. \]
Let $k_2 = \sqrt{k_0 k_1}$ By Lemma 6, $c''(k_1) - c''(k_2) < \eta \epsilon$, where $\eta = (k_1 - k_0)/(k_2 - k_0)$, and so
\[ N\left(\frac{d'_2 + d'_1}{2}\right) - N(d'_2) < \eta \epsilon. \]
By Lemma 7, it follows that $(d'_2 - d'_1)N'(d'_2) < 2\eta \epsilon$ for a suitable $\epsilon_0$. Using Eq. 5, it follows after some calculations that $c''(k_1) < \theta \epsilon$, where $\theta = 2\eta/(k_1 \log(k_1/k_0))$. \qed

Lemma 8 shows that $c''(k_1) \to 0$ as $c'_1(k_1)$ goes to its lower limit. Lemma 9 below shows a similar result holds in the limit case when $k_0 = 0$.

Lemma 9. For all real numbers $k_1$, $c_0$ and $c_1$ such that $0 < k_1$ and
\[ -1 < \frac{c_1 - c_0}{k_1} < 0, \]
let $c'_1$ range in $[c_1 - c_0]/k_1, 0]$ and $c$ the function $\phi(0, c_0, -1, k_1, c_1, c'_1)$. Then $c''(k_1) \to 0$ as $c'_1 \to (c_1 - c_0)/k_1$.

Proof. We use the same notation as in Lemma 5, from which follows the existence of $c$. Eq. 13 implies that, as $c'_0 \to (c_1 - c_0)/k_1$, $f N(-d'_1 - \Sigma) \to 0$. Since, by Eq. 11 $f$ is lower-bounded by $k_1 \exp(-(d'_2)^2/2)$, it follows that $\Sigma \to \infty$. Thus $c''(k_0) \to 0$ by Eq. 5. \qed
Lemma 10. Given any real numbers $k_0$, $k_1$, $c_0$, $c_1$, $c'_1$ such that $0 < k_0 < k_1$ and
\[ c'_1 - 1 < \frac{c_1 - c_0}{k_1 - k_0} < c'_1 \] let $c'_0 \in]c'_1-1, (c_1-c_0)/(k_1-k_0)[\) and $c = \phi(k_0, c_0, c'_0, k_1, c_1, c'_1)$. Then $c''(k_0) \to 0$ as $c'_0 \to (c_1 - c_0)/(k_1 - k_0)$.

Proof. The proof is similar to the proof of Lemma 8 and is omitted. \hfill \Box

Lemma 11. For all real numbers $k_0$, $c_0$ such that $0 < k_0$ and $c_0 > 0$, let $c'_0$ range in $]-1,0[$ and $c = c_{f,\Sigma,0,0}$ be the function $\phi(k_0, c_0, c'_0, \infty, 0, 0)$. Then $c''(k_0) \to 0$ as $c'_0 \to 0$.

Proof. Using the same notation as in Lemma 4, standard algebra shows that, as $c'_0 \to 0$, $a_1^2 \to -\infty$, $fN(d_1) \to c_0$, $\Sigma \to \infty$, and thus $c''(k_0) \to 0$ by Eq. 5. \hfill \Box

Theorem 3 below shows how to construct a concatenated function that will be twice continuous at a given point $k_j$, for some $j$ in the interval $[1, n]$. Step 2 of algorithm A described in Subsection 4.1 guarantees the concatenated function to be twice continuous at all points $k_j$, for any $j$ in the interval $[1, n]$.

**Theorem 3.** Let $j$ be an integer in the interval $[1, n]$. For all sequences $(k_i)_{0 \leq i \leq n+1}$, $(c_i)_{0 \leq i \leq n+1}$ and $(c'_i)_{0 \leq i \leq n+1}$ such that Eqs. 1, 14 and 15 hold, there exist a $C^1$ convex function $c(k)$, $k > 0$, and a sequence $(\Sigma, a_j, b_j)_{0 \leq j \leq n}$ such that $c(k) = c_{f,\Sigma,0,0}(k)$ on the interval $[k_i, k_{i+1}]$ and $c(k_i) = c_i$ for $0 < k_i$ and $c_0 > 0$, let $c'_0$ range in $]-1,0[$ and $c = c_{f,\Sigma,0,0}$ be the function $\phi(k_0, c_0, c'_0, \infty, 0, 0)$. Then $c''(k_0) \to 0$ as $c'_0 \to 0$.

Proof. For $1 \leq i \leq n$, let $l_i = (c_i - c_{i-1})/(k_i - k_{i-1})$. By Theorem 2, for any $\gamma \in [l_i, l_{i+1}]$, there exist a $C^1$ convex function $c(k)$, $k > 0$, and a sequence $(\Sigma, a_j, b_j)_{0 \leq j \leq n}$ such that $c(k) = c_{f,\Sigma,0,0}(k)$ on the interval $[k_i, k_{i+1}]$ and $c(k_i) = c_i$ for $1 \leq i \leq n$, $c'(k_j) = \gamma$ and $c'(k'_j) = c'_i$ for $1 \leq i \leq n$, $i \neq j$, and the limit conditions defined in Eqs. 16 and 17 hold. We show that for some $\gamma \in [l_j, l_{j+1}]$, $c$ has a continuous second derivative at $k_j$.

As $\gamma$ goes to $l_j$, the left second derivative $c''(k_j^-)$ of $c$ at $k_j$ goes to 0 by Lemmas 8 and 9. But the right second derivative $c''(k_j^+)$ of $c$ at $k_j$ goes to $\psi''(k_j)$, where $\psi$ is the function $\phi(k_j, c_j, l_j, k_{j+1}, c_{j+1}, c'_{j+1})$. Hence, $c''(k_j^+)$ has a positive limit as $\gamma$ goes to $l_j$. Similarly, it can be shown that $c''(k_j^-)$ has a negative limit as $\gamma$ goes to $l_{j+1}$. By continuity, $c''(k_j^+)$ is constant.

As in Theorem 2, the number of constraints on the parameters in Theorem 3 is equal to the number of parameters because there are still four constraints at $k_j$: $c(k_j^+) = c(k_j^-) = c_j$, $c''(k_j^+) = c''(k_j^-)$ and $c'^{(2)}(k_j^+) = c'^{(2)}(k_j^-)$. Unlike Theorem 2, though, it is not clear whether the function constructed in Theorem 3 is unique.
4.1 The algorithm description

We are now ready to describe our algorithm. Consider sequences \((k_i)_{0 \leq i \leq n+1}\) and \((c_i)_{0 \leq i \leq n+1}\) such that Eq. 1 holds, \(c_n > 0\) and

\[
-1 < \frac{c_i - c_{i-1}}{k_i - k_{i-1}} < \frac{c_{i+1} - c_i}{k_{i+1} - k_i} < 0 \quad \text{for} \ 1 \leq i < n. \tag{25}
\]

Note that Eq. 25 is the same as Eq. 2 except that inequalities have been replaced by strict inequalities. Let \(\epsilon > 0\) be an error parameter. Algorithm A consists of the following procedures:

1. Initialization Step. Let \(c'_0 = -1, c'_{n+1} = \infty\) and \(c'_i = (l_i + l_{i+1})/2\), for \(1 \leq i \leq n\), where \(l_i = (c_i - c_{i-1})/(k_i - k_{i-1})\).

2. Loop. For \(1 \leq j \leq n\), let \(\gamma_j = c'_j(k_j)\), where \(c_j\) is a function calculated in Theorem 3 for the index \(j\). Replace simultaneously \((c'_j)\) by \(\gamma_j\), \(1 \leq j \leq n\). Repeat this step until \(\max_{1 \leq j \leq n}(|c''(k^+_j) - c''(k^-_j)|) < \epsilon\).

Conjecture 1. For all sequences \((k_i)_{0 \leq i \leq n+1}\), \((c_i)_{0 \leq i \leq n+1}\) such that Eqs. 1 and 25 hold, the sequences calculated by Algorithm A converge in the limit towards a sequence \((f_i, \Sigma, s, b_i)_{0 \leq i \leq n}\). There exists a \(C^2\) convex function \(c(k)\), \(k > 0\), such that \(c(k) = c_{f_i, \Sigma, s, b_i}(k)\) on the interval \([k_i, k_{i+1}] - \{0, \infty\}\) and \(c(k_i) = c_i\) for \(1 \leq i \leq n\). Moreover, the limit properties in Eqs. 16 and 17 hold.

The number of constraints on the parameters in conjecture 1 is equal to the number of parameters. To see this, we note there are still four constraints at \(k_j\), \(1 \leq j \leq n\): \(c(k_j^+) = c(k_j^-) = c_j\), \(c'(k_j^+) = c'(k_j^-)\) and \(c''(k_j^+) = c''(k_j^-)\). Since there are four limit constraints by Eq. 17, the total number of constraints is equal to \(4n + 4\), which is equal to the number of parameters. It is not clear, though, whether there exists a unique function \(c\) satisfying the properties in conjecture 1. Several methods can be used to improve the numerical stability and speed of convergence of Algorithm A. We have tested a variant of Algorithm A where the \(c'_j\) and \(\gamma_j\) are updated in parallel using a Newton-Raphson method. Our experiments support our conjecture.

5 A two-dimensional interpolation method and Dupire’s model

If no arbitrage is found in the input implied volatilities we calculate an arbitrage-free interpolating volatility surface using Algorithm B that consists of the following steps:

1. We generate a one-dimensional arbitrage-free interpolation for each input maturity \(t\) using the above variant of Algorithm A.

2. For each maturity \(t \epsilon [t_i, t_{i+1}]\) and each strike \(K\), we calculate the implied volatility \(\sigma_{imp}(K, T)\) so that \(\sigma^2_{imp}(K, T)\) is a linear interpolation of \(\sigma^2_{imp}(K, t_i)t_i\) and \(\sigma^2_{imp}(K, t_{i+1})t_{i+1}\).

3. We make the necessary adjustments so that the entire volatility surface is arbitrage-free.
Our interpolation in the time domain ensures that if the no-arbitrage condition $C(K, t_i) < C(K, t_{i+1})$ holds, where $C(K, T)$ is today’s price of a European call with strike $K$ and maturity $T$, $C(K, T)$ is an increasing function of $T$ between $t_i$ and $t_{i+1}$. This is because, in the Black-and-Scholes formula, $C(K, T)$ depends on time only through $\sigma_{\text{imp}}^2(K, T)$. Moreover, $C(K, T)$ is an increasing function of $\sigma_{\text{imp}}^2(K, T)T$. Thus the no-arbitrage condition with respect to time holds if and only if $\sigma_{\text{imp}}^2(K, T)T$ is an increasing function of $T$. Step 3 is shown in detail in Subsection 5.1.

Because of the adjustments made in the last step, the smoothness properties of the one-dimensional interpolation method described in Section 1 may not hold for the two-dimensional interpolation method. However, numerical experiments in the two-dimensional case show that, for any maturity up to the last input maturity, the second derivative of the call price with respect to the strike exists and is continuous and positive except for a few points. Moreover, the derivative with respect to time is continuous except at input maturities. These properties help to ensure the stability of pricing exotic options using Dupire’s model. In Dupire’s model (Dupire 1994) the spot follows the following stochastic differential equation:

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

where $W_t$ is a Brownian motion and $\sigma(S, t)$ is a deterministic function. Dupire has shown that if the implied volatilities are known for all strikes and maturities then the local volatility surface is uniquely determined. More precisely,

$$\sigma^2(K, T) = 2 \frac{\partial^2 C(K, T)}{\partial K^2} \frac{\partial C(K, T)}{\partial T}.$$

Several pros and cons of Dupire’s model can be found in the literature. A classical problem in implementing the model is the instability of the local volatilities calculation. According to the practical cases we tested our interpolating algorithm is well suited to calibrating Dupire’s model. This is because the second derivative with respect to the strike of the call prices generated by our algorithm exists in practice and is continuous and positive except for a few points. Moreover, the call prices derivative with respect to the maturity exists and is continuous except at input maturities. The local volatility surface is therefore, in general, continuous except at input maturities, and can be calculated approximately using finite difference approximations of derivatives.

It is known that the price of a contingent claim on $S$ obeys the following PDE:

$$\frac{\partial u(S, t)}{\partial t} + \frac{1}{2} \sigma^2(S, t) S^2 \frac{\partial^2 u(S, t)}{\partial S^2} = 0,$$

where $u(S, t)$ is the price at time $t$ of a contingent claim if the spot price is $S$ at time $t$. The local volatility surface can thus be used to calculate option prices using finite difference schemes such as the Crank-Nicholson’s algorithm.

5.1 Making the entire volatility surface arbitrage-free

Once the implied volatility surface has been calculated in Step 2 for all maturities up to the last input maturity $t_h$, we calculate the associated call prices $C(K, T)$.
for $T \leq t_h$. Then, for each maturity $T \leq t_h$, we replace $C(K,T)$ by $C^*(K,T)$, where $C^*(K,T)$ is the supremum of all convex functions $\tilde{C}(K,T)$ with respect to $K$ such that $\tilde{C}(K,T) \leq C(K,s)$ for $T \leq s \leq t_h$. The function $C^*(K,T)$ is convex with respect to $K$ since the supremum of convex functions is convex. It is not hard to show that $C^*(K,T)$ is increasing with respect to $T$. Hence the call surface $C^*(K,T)$ is arbitrage-free for all $K$ and $T \leq t_h$.

In our implementation we have inverted steps 2 and 3 of Algorithm B and used a variant of step 3 that guarantees the restriction of the volatility surface to the union of input maturities to be arbitrage-free. Experiments we performed on market volatilities show that the entire volatility surface generated by this variant of Algorithm B is, for practical purposes, arbitrage-free.

5.2 Deep in the money and out of the money volatilities

For a given maturity, the extrapolation methods in Sections 3 and 4 yield asymptotically constant implied volatilities as the strike goes to infinity. This follows from standard calculus and the asymptotics

$$
c_f,\Sigma_{0,0}(k) \sim f N'(d_1) \sum_{d_1,d_2},
$$

as $k \to \infty$. Eq. 26 can be shown using the expansion

$$
N(z) \sim \frac{N'(z)}{|z|}(1 - \frac{1}{z^2} + O(\frac{1}{z^4}))
$$

as $z \to -\infty$. A similar result holds when the strike goes to 0. In practice, for a given maturity, we determine a strike which is deep out of the money and a strike deep in the money and keep the implied volatilities constant outside these strikes. The calculated local volatilities are also constant deep in and deep out of the money for a given maturity.

6 Dealing with dividends and interest rates

We assume that interest rates are deterministic and that dividends are of one or a combination of the following types:

- Absolute dividend with a predetermined cash amount at a predetermined date.
- Proportional dividend at a predetermined date with an amount proportional to the spot.
- Continuous dividend with a predetermined dividend rate.

We further assume that absolute dividends exist up to a certain maturity, with no restrictions on proportional or continuous dividends. We reduce the computation of the local volatility surface to the case where interest rates and dividends are null using a transformation similar to the one in (Overhaus, Ferraris, Knudsen, Milward, Nguyen-Ngoc and Schindlmayr 2002, Section 4.6). Let $T^*$ be a maturity larger than all absolute dividends and input options maturities and let $S^*_t$ be the forward of $S_t$ at maturity $T^*$. Thus $S^*_t$ is continuous and driftless.
and can be considered as an underlying in an interest rate and dividend-free world. By a classical calculation (see e.g. (Overhaus, Ferraris, Knudsen, Millward, Nguyen-Ngoc and Schindlmayr 2002, Section 4.6)) \( S_t = a(t)S^*_t + b(t) \), where \( a(t) \) and \( b(t) \) are deterministic functions of time. Let \( C^*(K^*, T) \) be the expected value under the risk neutral distribution of \( \max(0, S^*_t - K^*) \). Thus \( C(K, T) = B(t_0, t)a(t)C^*(K^*, T) \), where \( K = a(t)K^* + b(t) \) and \( B(t_0, t) \) is the price at initial time \( t_0 \) of a zero-coupon maturing at \( t \). A discrete set of option prices on \( S \) determines a discrete set of option prices on \( S^* \). Using results in previous sections, we can calculate a local volatility surface \( \sigma(S^*, t) \) that matches the options prices on \( S^* \). Using Ito’s lemma, the local volatility surface \( \sigma(S, t) = a(t)S^*\sigma(S^*, t)/S \) matches the options prices on \( S \). It can be shown that the local volatility surface \( \sigma(S, t) \) remains unchanged if \( T^* \) is replaced by a larger maturity.

In general, by non-arbitrage, volatility can no longer be a continuous function of time at dividend dates. For the same reason volatility cannot be constant everywhere in the presence of absolute dividends. Our algorithm generates an arbitrage-free implied volatility surface that has the same smoothness properties mentioned in the preceding sections except at discrete dividend dates. In practice the error rate of the calculated prices is generally higher if there are discrete dividends.

7 The foreign exchange and interest rate markets

Our algorithm can readily be used in the foreign exchange derivatives market because the foreign exchange dynamics are similar to the equity dynamics. The continuous dividend rate is simply replaced by the foreign exchange rate.

Our one-dimensional algorithm can be used to interpolate volatilities on options expiring at a given maturity \( T \) on a swap or interest rate spanning a given period \( I \). A slight modification is needed however if we assume interest rates can be negative.

8 Example

We tested our algorithm on the market implied volatilities of the USD/DEM exchange rate on August 23, 1995 given in (Avellaneda, Friedman, Holmes and Samperi 1997) and shown in Fig. 1. The DEM deposit rate is 4.27% while the US deposit rate is 5.91%.

We plotted in Fig. 2 the interpolated volatility surface, in Fig. 3 the call price, in Fig. 4 the risk-neutral density function of the spot, and in Fig. 5 the local volatility surface.

We compared the Black-and-Scholes prices calculated via the input implied volatilities to those calculated via the interpolated implied volatilities. The maximum relative error (price differences divided by the initial spot) obtained is of order \( 10^{-7} \). We also compared the prices calculated via the input implied volatilities to the prices obtained via the Crank-Nicholson method and our local volatility surface with 50 time-steps and 500 space-steps. The maximum relative error obtained is of order \( 10^{-5} \). Finally, we compare in Fig. 6 the prices
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Figure 1: Implied volatilities for the USD/DEM exchange rate on August 23, 1995.
calculated via the input implied volatilities to those obtained via Monte Carlo simulation together with our local volatility surface. We used 50 time-steps and 100000 paths in the Monte Carlo simulation. The maximum relative error obtained is of order $10^{-4}$. Our Monte Carlo simulation algorithm can be generalised to several dimensions, thereby allowing the pricing of options on several assets in a way consistent with the smile.

9 Conclusion

We have designed a one-dimensional interpolation algorithm for implied volatilities that is robust and has good smoothness properties. Our algorithm applies to equity, forex and interest rate options. It can be extended to the two-dimensional case for equity and forex options. In practice the regularity properties of our interpolation scheme ensure it can be used to calibrate Dupire’s model. Vanilla options prices calculated from our local volatility surface using PDE schemes or Monte Carlo simulation closely match input prices. Our method can be used to price options on one or several assets in a way consistent with the smile.

References

Figure 3: Call price on the USD/DEM exchange rate on August 1995 in terms of the maturity and the strike.

Figure 4: Implied risk-neutral density function of the spot of the USD/DEM exchange rate on August 1995 in terms of the maturity and the strike.
Figure 5: Local volatilities on the USD/DEM exchange rate on August 1995 in terms of the time and spot.


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Figure 6: Black-and-Scholes prices and prices obtained via MC simulation on the USD/DEM exchange rate on August 1995.


