



Towards a generalization of Dupire's equation for several assets

P. Amster^{a,b}, P. De Nápoli^{a,b}, J.P. Zubelli^{c,*}

^a Departamento de Matemática, Facultad de Ciencias Exactas y Naturales, Universidad de Buenos Aires, Ciudad Universitaria, Pabellón I, Buenos Aires, Argentina

^b CONICET, Argentina

^c IMPA, Est. D. Castorina 110, Rio de Janeiro, RJ22460-320, Brazil

ARTICLE INFO

Article history:

Received 2 October 2007

Available online 29 January 2009

Submitted by L. Guo

ABSTRACT

We pose the problem of generalizing Dupire's equation for the price of call options on a basket of underlying assets. We present an analogue of Dupire's equation that holds in the case of several underlying assets provided the volatility is time dependent but not asset-price dependent. We deduce it from a relation that seems to be of interest on its own.

© 2009 Elsevier Inc. All rights reserved.

1. Introduction

A fundamental problem in Financial Mathematics is that of calibrating the underlying model from market data. This is crucial, for example, in hedging and portfolio optimization. Such data may consist of underlying asset prices, or, as in many applications, derivative prices on such assets. An example, of central importance herein is an *European call option*. It gives the bearer the right, but not the obligation, of buying an asset B for a given strike price K at a certain maturity date T .

In the present work we are concerned with the problem of determining the model's volatility based on the quoted prices of a basket option for arbitrary values of the strike, the weights, and the maturity. Although, this is a highly idealized situation, it already poses some very interesting mathematical challenges, as we shall see in the sequel. The results presented here should be valuable for the development of effective methods to estimate the local volatility in multi-asset markets where a sufficiently large set of basket options is traded.

In the standard Black and Scholes [2] model for option pricing, the underlying asset is assumed to follow a dynamics described by the stochastic differential equation

$$\frac{dS}{S} = \mu dt + \sigma dW,$$

where W is a Brownian motion, μ is a drift coefficient, and σ is the volatility of the underlying asset. In the classical Black–Scholes theory, σ is assumed to be constant. Despite the enormous success of such model, it is known that in practice it cannot consistently price options with different strike prices and maturity dates, as the volatility empirically appears not to be constant over time. Furthermore, if one computes the *implied volatility* from the quoted price one verifies empirically that different strikes and maturities lead to different implied volatilities for options on a given asset. This is known as the *smile effect* and was discussed in a pioneering paper by B. Dupire [5].

Due to the smile effect, volatility estimates based on historical data are considered not to be reliable. Another approach consists in trying to determine the volatility from the option prices in the market. This leads to a challenging inverse problem. See, for example, [1,3,6,10].

* Corresponding author.

E-mail addresses: pamster@dm.uba.ar (P. Amster), pdenapo@dm.uba.ar (P. De Nápoli), zubelli@impa.br (J.P. Zubelli).

In [5], Dupire considered a model for the dynamics of the underlying asset in which the volatility depends both on the time t and on the stock price S . More precisely,

$$\frac{dS}{S} = \mu dt + \sigma(t, S) dW. \tag{1}$$

This type of model is known as a local volatility model. Other approaches have been proposed in which the volatility follows another stochastic process.

Dupire has shown that in the local volatility model, the volatility can, in principle, be recovered from market data if the price of European options on the underlying asset were known for all the strike prices K and maturity dates T .

The celebrated Dupire equation for the case of a single asset reads as follows

$$\frac{\partial C}{\partial T} = \frac{\sigma^2(K, T)K^2}{2} \frac{\partial^2 C}{\partial K^2} + (r(T) - D(T)) \left(C - K \frac{\partial C}{\partial K} \right),$$

or in other words

$$\sigma = \sqrt{\frac{\frac{\partial C}{\partial T} - (r(t) - D(t))(C - K \frac{\partial C}{\partial K})}{\frac{K^2}{2} \frac{\partial^2 C}{\partial K^2}}}.$$

Here, $C(t, S_t, T, K)$ is the undiscounted European call option price, $r(t)$ is the risk-free interest rate and $D(t)$ is the dividend rate. The price C satisfies, under the usual assumptions of liquidity, absence of arbitrage, and transaction costs (perfect market), the Black–Scholes equation

$$\begin{cases} \frac{\partial C}{\partial t} + \frac{1}{2} \sigma^2(t, S) S^2 \frac{\partial^2 C}{\partial S^2} + r \left(S \frac{\partial C}{\partial S} - C \right) = 0, & S > 0, t < T, \\ C(S, T) = (S - K)^+. \end{cases} \tag{2}$$

In practice, however, the option prices are known only for a few maturity dates and strike prices and some interpolation is needed. The computed volatility depends strongly on the interpolation used. Due to the ill-posed character of this inverse problem, some regularization strategy has to be used to ensure the numerical stability of the reconstruction. See [3,6]. In any case, Dupire’s formula plays a fundamental role in several methods that have been proposed to tackle this problem.

Let us now consider, the multi-asset situation, which is very important in practice. In particular, it could be applied to index options.

Here, the dynamics is given by

$$\frac{dS_i}{S_i} = \mu_i dt + \sum_{j=1}^N \sigma_{ij} dW_j, \tag{3}$$

where W denotes the N -dimensional Brownian motion with respect to the risk-neutral measure. Here $\sigma_{ij} = \sigma_{ij}(t, S)$ is the volatility matrix $\mu_i = \mu_i(t)$ is the risk-neutral drift, with $\mu_i(t) = r(t) - D_i(t)$ where D_i is the dividend rate of the i th asset, and $W = (W_1, \dots, W_N)$ is a standard N -dimensional Brownian motion.

For technical reasons, we shall assume throughout this paper that the volatility matrix $((\sigma_{ij}(t, S)))$ and the drift vector $\mu_j(t)$ are smooth and bounded, i.e.,

$$|\mu_j(t)| \leq C \quad \text{and} \quad |\sigma_{ij}(t, S)| \leq C. \tag{4}$$

Furthermore, we shall assume that the matrix $A = (a_{ij}) = \frac{1}{2} \sigma \sigma^t$ satisfies the uniform ellipticity condition: there exist constants $\lambda, \Lambda > 0$ such that

$$\lambda |y|^2 \leq \sum_{i,j} a_{ij}(t, S) y_i y_j \leq \Lambda |y|^2. \tag{5}$$

Given a vector of weights $w = (w_1, w_2, \dots, w_N)$ with $w_i \geq 0$, we consider an *European basket option*, that is, a contract giving the holder the right to buy a basket composed of w_i units of the i th asset at a maturity date T upon paying a strike price K .

Here, the value

$$B = \sum_{j=1}^N w_j S_j$$

is called the *basket price (or index) composed of the stocks S_i* .

The fair price of such an option is

$$P(S_t, t, K, T) = e^{-\int_t^T \mu_i(\tau) d\tau} E_t^* \left[\left(\sum_{i=1}^N w_i S_{i,T} - K \right)^+ \right]$$

where E_t^* denotes the expected value at time t under the so-called risk-neutral probability. It turns out to be simpler to work with the undiscounted call-price

$$C_w = e^{\int_t^T \mu_i(\tau) d\tau} P = E_t^* \left[\left(\sum_{i=1}^N w_i S_{i,T} - K \right)^+ \right].$$

Our goal is to address the following natural question:

Is there a generalization of Dupire's equation for the multi-asset context?

We have a partial answer to this question, under additional assumptions, the most restrictive of all being that of having an asset-price independent volatility. More precisely, our main result reads as follows:

Theorem 1. *Assume that the volatility matrix σ_{ij} is a deterministic locally integrable function of time. Then the fair price C_w of the European basket call option satisfies*

$$\frac{\partial C_w}{\partial T} = \sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} + \sum_{i,j=1}^N a_{ij} w_i w_j \frac{\partial^2 C_w}{\partial w_i \partial w_j}, \tag{6}$$

where $A = (a_{ij})$ denotes the matrix given by $A = \frac{1}{2} \sigma \sigma^t$.

The proof of this result will be the subject of Section 3 as well as that of Appendix A.

Let p denote the transition probability density corresponding to the stochastic process defined by Eq. (3), and let \mathbf{s} denote the surface measure in the set

$$L_w \stackrel{\text{def}}{=} \left\{ (S_1, \dots, S_N) \mid \sum_{j=1}^N w_j S_j = K, S_j \geq 0 \right\}. \tag{7}$$

Theorem 1 relies on the following remarkable relation, that seems to be of interest in its own:

$$\sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} = \frac{\partial C_w}{\partial T} - \sum_{i,j=1}^N \int_{L_w} a_{ij} S_{i,T} S_{j,T} p(S_t, t, S_T, T) \frac{w_i w_j}{|w|} ds. \tag{8}$$

Remark 2. If no dividends are paid then $\mu_i = r$ for all i , and using the Euler's equation (12) we can re-write (6) as

$$\frac{\partial C_w}{\partial T} = r \left(C_w - K \frac{\partial C_w}{\partial K} \right) + \sum_{i,j=1}^N a_{ij} w_i w_j \frac{\partial^2 C_w}{\partial w_i \partial w_j}.$$

2. Review of Dupire's equation and related facts

A key point in the derivation of the one-dimensional Dupire's equation is that one may express the price of an European call option as

$$C(t, S_t, T, K) = \int_{-\infty}^{\infty} p(S_t, t, S_T, T) (S - K)^+ dS_T$$

where $p(t, S_t, \tilde{t}, S_{\tilde{t}})$ is the transition probability density corresponding to the stochastic process defined by Eq. (1). From the PDE viewpoint, p is fundamental solution associated to the N -dimensional Black–Scholes equation (2). Using the fundamental theorem of calculus we deduce that

$$\frac{\partial C}{\partial K} = - \int_K^{\infty} p(S_t, t, S_T, T) dS_T.$$

Hence, we may recover the transition probability by computing the second derivative of the call price with respect to K

$$\frac{\partial^2 C}{\partial K^2} = p. \tag{9}$$

For comparison with the multi-dimensional case, it is convenient to consider a more general (discounted) call option C_w for buying w units of the stock with strike price K . Then

$$C_w = E_{t_0}^* [(wS_T - K)^+].$$

Thus, C_w is plainly a homogeneous function of degree one, with respect to the variables K and w . Hence, it satisfies Euler's equation, namely

$$K \frac{\partial C_w}{\partial K} + w \frac{\partial C_w}{\partial w} = C_w.$$

Differentiating this equation with respect to K and w we get

$$K \frac{\partial^2 C_w}{\partial K^2} + w \frac{\partial^2 C_w}{\partial w \partial K} = 0$$

and

$$K \frac{\partial^2 C_w}{\partial K \partial w} + w \frac{\partial^2 C_w}{\partial w^2} = 0.$$

Hence,

$$K^2 \frac{\partial C_w}{\partial K^2} = w^2 \frac{\partial^2 C_w}{\partial w^2},$$

and we conclude that Dupire's equation can be written in an equivalent form as

$$\frac{\partial C_w}{\partial T} = \mu w \frac{\partial C_w}{\partial w} + \frac{1}{2} \sigma^2 w^2 \frac{\partial^2 C_w}{\partial w^2}.$$

3. The multi-asset case

We now present a proof of Theorem 1. As before, the price of the basket option can be written as

$$C_w(S_t, t, K, T) = \int_{\mathbb{R}_+^N} p(S_t, t, S_T, T) \left(\sum_{i=1}^N w_i S_{i,T} - K \right)^+ dS_T,$$

where $p(t, S_t, \tilde{t}, S_{\tilde{t}})$ is now the transition probability density associated to the stochastic process defined by (3), or from the PDE's viewpoint the fundamental solutions to the multi-dimensional Black–Scholes equation:

$$\frac{\partial C}{\partial T} + \sum_i \mu_i(t, S) S_i \frac{\partial C}{\partial S_i} + \sum_{i,j=1}^N a_{ij}(t, S) S_i S_j \frac{\partial^2 C}{\partial S_i \partial S_j} = 0. \tag{10}$$

The standard theory of parabolic equations does not apply directly to (10). However, under the usual change of variables $\tau = T - t$ and $X_i = \log S_i$, Eq. (10) transforms into a non-degenerate parabolic equation.

Under the technical conditions (4) and (5), it can be proved that (10) admits a fundamental solution p that is at least of class $C^{1,2}$ and decays exponentially when $\|S\| \rightarrow \infty$, together with its first and second order derivatives. This fact will be crucial in the following computations, since this ensures that all the boundary terms at infinity vanish.

The proof of the existence of the fundamental solutions under these assumptions can be done by using the so-called *parametrix method*, introduced by E. Levi [9] in 1907. We remark that our technical conditions (4) and (5), and the smoothness requirement on the coefficients could be certainly relaxed. See, for example, [4] for a construction of the fundamental solution in the unbounded coefficient case, using Levi's method. However, as our main interest in this paper is the financial significance of our results, we do not intend to state the most general conditions under which our computations are still valid.

We introduce the region

$$H_w \stackrel{\text{def}}{=} \left\{ S \in \mathbb{R}_+^N \mid \sum_{i=1}^N w_i S_i \geq K \right\}.$$

Thus,

$$C_w(S_t, t, K, T) = \int_{H_w} p(S_t, t, S_T, T) \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T. \tag{11}$$

We note that C_w is homogeneous of degree one in the variables $(w_1, w_2, \dots, w_n, K)$. Hence, it satisfies Euler equation

$$\sum_{i=1}^N w_i \frac{\partial C_w}{\partial w_i} + K \frac{\partial C_w}{\partial K} = C_w. \tag{12}$$

In order to be able to compute the derivatives of C_w , it is convenient to re-write Eq. (11) as an integral over a region independent of w . For this purpose, we introduce the change of variables

$$B = \sum_{i=1}^N w_i S_{i,T},$$

and

$$Q_i = \frac{w_i S_{i,T}}{\sum_{i=1}^N w_i S_{i,T}}$$

for $i = 1, \dots, N - 1$. Therefore, $Q \in \Delta_N$ where

$$\Delta_N = \left\{ Q = (Q_1, Q_2, \dots, Q_{N-1}): Q_i \geq 0, \sum_{i=1}^{N-1} Q_i \leq 1 \right\}$$

is the $(N - 1)$ -dimensional simplex. Thus,

$$S_T := S(Q, B) = \left(\frac{Q_1 B}{w_1}, \dots, \frac{Q_{N-1} B}{w_{N-1}}, \frac{(1 - \sum_{i=1}^{N-1} Q_i) B}{w_N} \right).$$

The Jacobian of the change of variables

$$(S_{1,T}, \dots, S_{N,T}) \mapsto (Q_1, \dots, Q_{N-1}, B)$$

is given by

$$J = \frac{\partial(S_{1,T}, \dots, S_{N,T})}{\partial(Q_1, \dots, Q_{N-1}, B)} = \frac{B^{N-1}}{w_1 w_2 \dots w_N}.$$

Thus, we obtain:

$$C_w(S_t, t, K, T) = \int_K^\infty \int_{\Delta_N} p(S_t, t, S(Q, B), T) (B - K) \frac{B^{N-1}}{w_1 w_2 \dots w_N} dQ dB.$$

Hence

$$\begin{aligned} \frac{\partial C_w}{\partial K} &= \int_{\Delta_N} \left[p(S_t, t, S(Q, B), T) (B - K) \frac{B^{N-1}}{w_1 w_2 \dots w_N} \right]_{B=K} dQ - \int_K^\infty \int_{\Delta_N} p(S_t, t, S(Q, B), T) \frac{B^{N-1}}{w_1 w_2 \dots w_N} dQ dB \\ &= - \int_K^\infty \int_{\Delta_N} p(S_t, t, S(Q, B), T) \frac{B^{N-1}}{w_1 w_2 \dots w_N} dQ dB, \end{aligned}$$

and

$$\frac{\partial^2 C_w}{\partial K^2} = \int_{\Delta_N} p(S_t, t, S(Q, K), T) \frac{K^{N-1}}{w_1 w_2 \dots w_N} dQ.$$

Going back to the S_T -coordinates we easily obtain the following identity:

$$\frac{\partial^2 C_w}{\partial K^2} = \frac{1}{|w|} \int_{L_w} p(S_t, t, S_T, T) ds, \tag{13}$$

where L_w is defined as in the introduction. This identity relates the second derivative of the call price C_w with respect to strike price K , to the integral of the probability density p over the set L_w .

Eq. (13) is the multi-dimensional analogue of Eq. (9); in probabilistic terms, the integral term expresses the probability that the basket B has a price K at the maturity date T , given that the price vector has the value S_t at time t , namely

$$\frac{\partial^2 C_w}{\partial K^2} = \frac{1}{|w|} P[B_T = K | S_t].$$

However, this relationship does not seem to yield a suitable multi-dimensional generalization of Dupire's equation. For this reason, we also compute the derivatives $\frac{\partial C_w}{\partial w_i}$ to get

$$\begin{aligned}
 w_i \frac{\partial C_w}{\partial w_i} &= - \int_K \int_{\Delta_N} \frac{\partial p}{\partial S_{i,T}}(S_t, t, S(Q, B), T)(B - K) \frac{Q_i B}{w_i} \frac{B^{N-1}}{w_1 w_2 \dots w_N} dQ dB \\
 &\quad - \int_K \int_{\Delta_N} p(S_t, t, S(Q, B), T)(B - K) \frac{B^{N-1}}{w_1 w_2 \dots w_N} dQ dB
 \end{aligned} \tag{14}$$

for $i = 1, \dots, N - 1$. It is straightforward to notice that upon extending the above notation so that $Q_N = 1 - \sum_{i=1}^{N-1} Q_i$, relation (14) also holds for $i = N$. Then,

$$\begin{aligned}
 \sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} &= - \int_{H_w} \sum_{i=1}^N \mu_i \left[S_{i,T} \frac{\partial p}{\partial S_{i,T}} + p \right] \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T \\
 &= - \int_{H_w} \sum_{i=1}^N \frac{\partial}{\partial S_{i,T}} [\mu_i S_i p] \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T.
 \end{aligned}$$

Now, we use the fact that p satisfies the multi-dimensional Fokker–Planck equation (see e.g. [11]):

$$\frac{\partial p}{\partial T} + \sum_{i=1}^N \frac{\partial}{\partial S_i} [\mu_i S_i p] - \sum_{i,j=1}^N \frac{\partial^2}{\partial S_i \partial S_j} [a_{ij} S_i S_j p] = 0.$$

Thus we obtain

$$\sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} = \int_{H_w} \left\{ \frac{\partial p}{\partial T} + \sum_{i,j=1}^N \frac{\partial^2}{\partial S_{i,T} \partial S_{j,T}} [a_{ij} S_{i,T} S_{j,T} p] \right\} \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T.$$

On the other hand, we compute the derivative of C_w with respect to the maturity date

$$\frac{\partial C_w}{\partial T} = \int_{H_w} \frac{\partial p}{\partial T}(S_t, t, S_T, T) \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T,$$

and then

$$\sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} = \frac{\partial C_w}{\partial T} + \int_{H_w} \sum_{i,j=1}^N \frac{\partial^2}{\partial S_{i,T} \partial S_{j,T}} [a_{ij} S_{i,T} S_{j,T} p] \left(\sum_{i=1}^N w_i S_{i,T} - K \right) dS_T.$$

Upon applying the divergence theorem, and using the fact that the boundary integral over ∂H_w vanishes, we get

$$\sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} = \frac{\partial C_w}{\partial T} - \int_{H_w} \sum_{i,j=1}^N \frac{\partial}{\partial S_{j,T}} [a_{ij} S_{i,T} S_{j,T} p] w_i dS_T.$$

As the exterior normal vector to L_w is given by $\frac{-w}{|w|}$, we obtain:

$$\sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} = \frac{\partial C_w}{\partial T} - \sum_{i,j=1}^N \int_{L_w} a_{ij} S_{i,T} S_{j,T} p(S_t, t, S_T, T) \frac{w_i w_j}{|w|} ds. \tag{15}$$

On the other hand, after changing variables and integrating by parts identity (14) we also deduce that

$$\frac{\partial C_w}{\partial w_i} = \int_{H_w} p(S_t, t, S_T, T) S_{i,T} dS_T.$$

Then

$$\begin{aligned}
 w_i w_j \frac{\partial^2 C_w}{\partial w_j \partial w_i} &= - \int_K \int_{\Delta_N} \frac{\partial p}{\partial S_{j,T}}(S_t, t, S(Q, B), T) Q_j B \frac{Q_i B}{w_j} \frac{B^{N-1}}{w_1 \dots w_N} dQ dB \\
 &\quad - (1 + \delta_{ij}) \int_K \int_{\Delta_N} p(S_t, t, S(Q, B), T) Q_i B \frac{B^{N-1}}{w_1 \dots w_N} dQ dB,
 \end{aligned}$$

where δ_{ij} is the Kronecker's delta. As before, using the fact that $\frac{\partial}{\partial S_{j,T}}(pS_{i,T}S_{j,T}) = \frac{\partial p}{\partial S_{j,T}}S_{i,T}S_{j,T} + p(1 + \delta_{ij})S_{i,T}$, we deduce that

$$\frac{\partial^2 C_w}{\partial w_i \partial w_j} = \frac{1}{|w|} \int_{L_w} p S_{i,T} S_{j,T} ds. \quad (16)$$

Thus, if a_{ij} are time-dependent-only, we obtain:

$$\frac{\partial C_w}{\partial T} = \sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} + \sum_{i,j=1}^N a_{ij} w_i w_j \frac{\partial^2 C_w}{\partial w_i \partial w_j}.$$

This concludes the proof of Theorem 1.

4. Conclusions

The problem of calibrating the model parameters so that pricing of different derivative instruments would be consistent with the market observed prices is a crucial one in financial mathematics. Basket options play an increasingly important role in many markets. One reason being that several indices and bench-marks are defined in terms of baskets. Assuming the well-known multi-asset Black–Scholes model with volatility that varies with time only, the present work provides a theoretical answer to the calibration problem of the volatility matrix by using basket option prices for arbitrary weights and times in the region under consideration. This solution is achieved by means of Eq. (6). The latter is one possible multi-dimensional generalization of Dupire's seminal equation [5]. As is well known, not even in the one-dimensional situation Dupire's equation should be confused with Kolmogorov's backward equation. Yet, many of the one-dimensional derivations pass through such equation by using the fact that the fundamental solution of Black–Scholes equation is given by the second derivative with respect to the strike of the call price. See, for example, Section 2 of [3]. In more than one dimension, the fundamental solution is not characterized by the second derivative with respect to the strike and this is the main difficulty we had to overcome in proving the result.

As in the one-dimensional Dupire approach, the practical implementation of the ideas presented herein would have to face the usual difficulties in inverse problems. Namely, the need for regularization and the scarcity of data. The need for regularization arises because one has to deal with derivatives and has been the subject of a large number of articles. See, for example, [7] and references therein. The scarcity of data, which is already a problem in the one-dimensional case, could not possibly be better here. Although the results are rather restricted in two ways, first, the volatility matrix must be constant with respect to the asset, second, the result (1.6) requires one to be able to observe basket options with several weights (yet a finite number). More precisely, if we want to theoretically reconstruct the volatility matrix we need the prices and its derivatives with respect to the weights up to second order at the time interval under consideration. This coincides with the one-dimensional case that theoretically needs the prices and the derivatives of the prices with respect to the strikes up to second-order. In our multi-asset case, the derivatives with respect to the strikes are replaced by the derivatives with respect to the weights. Despite the difficulty that the number of basket options with different weights is still rather limited, as more financial products become available, such as different options on commodity indices, currency indices, and exchange-traded funds (ETFs), we believe the first step provided by our generalization of Dupire's formula could become of some practical relevance.

Although the assumption of time-dependence-only for the volatility might seem artificial we remark that this is an important step in dealing with such problems. Indeed, in [6] the time dependence of volatility was used as a simplifying assumption. Later on, in [7] such approach was extended to a more realistic context associated with a volatility and term structure. In particular, they assumed a multiplicative structure for the local volatility, which was motivated by the specific data situation, and showed that the inverse problem could be decomposed into two separate sub-problems. In this respect, we feel that the result presented here is relevant to further developments.

One way of geometrically interpreting Dupire's equation is the following: If one looks at the surface generated by the call option prices as the strike and the maturity vary, then such surface is subject to a constraint which determines uniquely the volatility at each point in the (K, T) plane. Furthermore, the volatility depends only on the local behavior of the prices and its derivatives of order up to 2. Our generalization says that if we now consider time-dependent-only volatility for the multi-asset case and look at basket options parameterized by the relative weights and the time to maturity, then once again the (symmetric) volatility tensor is uniquely determined by each point (w, T) . Such determination is unique provided a sufficiently large yet finite number of weights are chosen and the volatility depends only on the derivatives up to order 2 of the prices.

A natural continuation of the present work would be to extend the results presented herein to a situation where σ depends also on the underlying asset prices. Although at this moment we do not have such generalization, we believe that it should somehow rely on Eqs. (15) and (16). One might even speculate that it would involve a non-local operator. Another idea would be to consider contracts whose payoff is $\max(S_1 - K_1, S_2 - K_2, \dots, S_N - K_N)$ and join such ideas with the present results. Yet another natural continuation of the present work would be to use the results obtained herein to develop effective numerical methods to compute the matrix $A = \frac{1}{2} \sigma \sigma^t$. In this respect, the numerical ideas presented in [8] might be very helpful.

Acknowledgments

P. Amster and P. De Nápoli were supported by PIP 5477-CONICET. J.P. Zubelli was supported by CNPq under grant 30.1003 and 200258. The three authors were also supported by the PROSUL program under grant number 490300/2004-9. The authors wish to thank the anonymous referees for a number of helpful comments and questions.

Appendix A. An alternative derivation

In this appendix we present yet another derivation of the main result. We believe that the techniques employed herein provide a complementary view of the problem. For simplicity, throughout this section we shall write S to denote the stock price at time t .

Consider as before a basket option, with a pay-off function given by:

$$f = \left(\sum_{j=1}^N w_j S_j - K \right)^+.$$

Ito–Tanaka formula [12] reads as

$$df = \sum_{i=1}^N \frac{\partial f}{\partial S_i} dS_i + \sum_{i,j=1}^N a_{ij} S_i S_j \frac{\partial^2 f}{\partial S_i \partial S_j} dt$$

with $A = (a_{ij})$ as before. Note that

$$\frac{\partial f}{\partial S_i} = H\left(\sum_{j=1}^N w_j S_j - K\right) w_i,$$

where H denotes the Heaviside function given by $H(s) = 1$ if $s > 0$ and zero otherwise. Furthermore,

$$\frac{\partial^2 f}{\partial S_i \partial S_j} = \delta\left(\sum_{j=1}^N w_j S_j - K\right) w_i w_j.$$

Hence,

$$\begin{aligned} f(T) = f(t_0) &+ \sum_{i=1}^N \int_{t_0}^T H\left(\sum_{j=1}^N w_j S_j - K\right) w_i S_i \mu_i dt + \sum_{i,j=1}^N \int_{t_0}^T H\left(\sum_{j=1}^N w_j S_j - K\right) w_i \sigma_{ij} dW_j \\ &+ \sum_{i,j=1}^N \int_{t_0}^T \delta\left(\sum_{j=1}^N w_j S_j - K\right) a_{ij} S_i S_j w_i w_j dt. \end{aligned}$$

Now we take the expected value $E_{t_0}^*$ at time t_0 to get

$$C_w(t_0) = f(t_0) + \sum_{i=1}^N \int_{t_0}^T E_{t_0}^* \left[H\left(\sum_{j=1}^N w_j S_j - K\right) w_i S_i \mu_i \right] dt + \sum_{i,j=1}^N w_i w_j \int_{t_0}^T E_{t_0}^* \left[\delta\left(\sum_{j=1}^N w_j S_j - K\right) a_{ij} S_i S_j \right] dt. \tag{A.1}$$

In the sequel, we make use of the following

Lemma 3. Let $g : \mathbb{R}_+^N \rightarrow \mathbb{R}$. Then,

$$\int_{\mathbb{R}_+^N} g(S) \delta\left(\sum_{j=1}^N w_j S_j - K\right) p(S_{t_0}, t_0, S, t) dS = \frac{1}{|w|} \int_{L_w} g(S) p(S_{t_0}, t_0, S, t) ds.$$

Proof. Let us define, in a similar way to that of Section 3, B and Q by

$$B = \sum_{i=1}^N w_i S_i,$$

and

$$Q_i = \frac{w_i S_i}{\sum_{i=1}^N w_i S_i}$$

for $i = 1, \dots, N - 1$. Then

$$\begin{aligned} \int_{\mathbb{R}_+^N} g(S) \delta \left(\sum_{j=1}^N w_j S_j - K \right) p(S_{t_0}, t_0, S, t) dS &= \int_K^\infty \int_{\Delta_N} g(S(Q, B)) \delta(B - K) \frac{B^{N-1}}{w_1 w_2 \dots w_N} dB dQ \\ &= \int_{\Delta_N} g(S(Q, K)) p(S_{t_0}, t_0, S, t) \frac{K^{N-1}}{w_1 w_2 \dots w_N} dQ \\ &= \frac{1}{|w|} \int_{L_n} g(S) p(S_{t_0}, t_0, S, t) dS. \quad \square \end{aligned}$$

Back to Eq. (A.1), we get

$$E_{t_0}^* \left[\delta \left(\sum_{j=1}^N w_j S_j - K \right) a_{ij} S_i S_j \right] = \int_{\mathbb{R}_+^N} a_{ij} S_i S_j \delta \left(\sum_{j=1}^N w_j S_j - K \right) p(S_{t_0}, t_0, S, t) dS$$

where, as before, p denotes the transition probability density. From the previous lemma,

$$E_{t_0}^* \left[\delta \left(\sum_{j=1}^N w_j S_j - K \right) a_{ij} S_i S_j \right] = \frac{1}{|w|} \int_{L_w} a_{ij} S_i S_j p(S_{t_0}, t_0, S, t) dS.$$

Furthermore,

$$E_{t_0}^* \left[H \left(\sum_{j=1}^N w_j S_j - K \right) w_i S_i \mu_i \right] = \int_{\mathbb{R}_+^N} \mu_i w S_{i,t} H \left(\sum_{j=1}^N w_j S_j - K \right) p(S_{t_0}, t_0, S, t) dS.$$

On the other hand, upon computing the derivatives

$$\frac{\partial C_w}{\partial w_i} = \int_{\mathbb{R}_+^N} H \left(\sum_{j=1}^N w_j S_j - K \right) S_i p(S_{t_0}, t_0, S, t) dS,$$

we deduce that

$$E_{t_0}^* \left[H \left(\sum_{j=1}^N w_j S_j - K \right) w_i S_i \mu_i \right] = \mu_i w_i \frac{\partial C_w}{\partial w_i}.$$

Finally, from the identity

$$C_w(t_0) = f(t_0) + \sum_{i=1}^N \mu_i w_i \int_{t_0}^T \frac{\partial C_w}{\partial w_i} dt + \sum_{i,j=1}^N \frac{w_i w_j}{|w|} \int_{t_0}^T \int_{L_w} a_{ij} S_i S_j p(S_{t_0}, t_0, S, t) ds dt,$$

we get

$$\frac{\partial C_w}{\partial T} = \sum_{i=1}^N \mu_i w_i \frac{\partial C_w}{\partial w_i} + \sum_{i,j=1}^N \frac{w_i w_j}{|w|} \int_{L_w} a_{ij} S_i S_j p(S_{t_0}, t_0, S, t) ds.$$

Now, since

$$\frac{\partial^2 C_w}{\partial w_i \partial w_j} = \int_{\mathbb{R}_+^N} \delta \left(\sum_{j=1}^N w_j S_j - K \right) S_i S_j p(S_{t_0}, t_0, S, t) dS = \frac{1}{|w|} \int_{L_w} S_i S_j p(S_{t_0}, t_0, S, t) ds,$$

we deduce once again that if the diffusion coefficients a_{ij} are deterministic functions depending only on time, then the generalized Dupire’s equation (6) holds.

References

- [1] M. Avellaneda, D. Boyen-Olson, J. Busca, P. Fritz, Reconstructing volatility, *Risk* (2002) 91–95.
- [2] F. Black, M. Scholes, The pricing of options and corporate liabilities, *J. Polit. Economy* 81 (1973) 637–659.
- [3] I. Bouchoev, V. Isakov, Uniqueness, stability and numerical methods for the inverse problem that arises in financial markets, *Inverse Problems* 15 (1999) R95–R116.
- [4] T. Deck, S. Kruse, Parabolic differential equations with unbounded coefficients: A generalization of the parametrix method, *Acta Appl. Math.* 74 (2002) 71–91.
- [5] B. Dupire, Pricing with a smile, *Risk* 7 (1994) 18–20.
- [6] T. Hein, B. Hofmann, On the nature of ill-posedness of an inverse problem arising in option pricing, *Inverse Problems* 19 (2003) 1319–1338.
- [7] H. Egger, T. Hein, B. Hofmann, On decoupling of volatility smile and term structure in inverse option pricing, *Inverse Problems* 22 (4) (2006) 1247–1259.
- [8] S.D. Howison, N.A. Jackson, Computation of deterministic volatility surfaces, *J. Comp. Finance* 2 (1999) 5–32.
- [9] E. Levi, Sulle equazioni lineari totalmente ellittiche alle derivate parziali, *Rend. Circ. Mat. Palermo* 24 (1907) 275–317.
- [10] J. Lishang, T. Youshan, Identifying the volatility of underlying assets from option prices, *Inverse Problems* 17 (2001) 137–155.
- [11] B. Øksendal, *Stochastic Differential Equations*, fifth ed., Springer-Verlag, 1998.
- [12] L.C.G. Rogers, D. Williams, *Diffusions, Markov Processes, and Martingales*, vol. 2, Cambridge Univ. Press, Cambridge, 2000.