The n-fold compound option

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Abstract

This paper revisits the compound options as introduced by R. Geske [2]. Geske presented a theory for pricing an option on an option which he defined as a compound option. He developed a closed form expression for this kind of options. In this paper we will extend the notion of compound option to the n-fold compound option or compound option of order n.

Moreover an interesting relationship between a k-variate normal distribution function and a (k+1)-variate normal distribution function is proved for this intention.

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1 Introduction

As was mentioned by Geske [2], any opportunity with a choice whose value depends on an underlying asset can be viewed as an option.

The specific opportunity for an option are its boundary conditions. Many opportunities have a sequential nature, where latter opportunities are available only if earlier opportunities are undertaken. Such is the nature of the compound option (by Geske [2]) or option on an option.

We reintroduce the concept of a compound (call) option.

A 2-fold compound call option (or compound option of order 2) is a call option on a call option i.e. a call option with the underlying being a call option itself. So such a contract entitles one to the following payoff at t_1

$$\max \{C(t_1, S(t_1), t_2, K_2), K_1\},\$$

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that is, at t_1 , the investor (holder) receives the maximum of the amount K_1 and the value of a European call on the asset S with exercise date and price given by t_2 and K_2 respectively. In other words the investor holds a call, exercisable at t_1 , on the underlying call which is exercisable at t_2 .

In this paper this idea is generalized to a compound call of order n (with exercise date and price given by t_1 and K_1) with as underlying asset a compound call of order n-1 (with exercise date and price given by t_2 and K_2) which itself is a call on a call of order n-2 ... until the final underlying asset, a European call, to be a call of order 1 (with exercise date and price given by t_n and K_n).

The price at time t_0 of such a call of order n is denoted by

$$C^{(n)}(t_0, S(t_0; (t_j, K_j)_{j=1}^n)),$$

with (t_j, K_j) the exercise date and price of the call of order (n-j+1), underlying the call of order n.

In section 2 the valuation equation for this n-fold compound option (compound of order n) is presented and proved by induction. The proof is based on the PDE representation form and involves a result on a relationship between the k-variate and (k+1)-variate normal distribution function which is treated in appendix A. The reason for the consideration of these n-fold compound options is the possible application of such derived financial products in the field of

- growing business
- insurance business where the relevance of such products was shown by Simon and Van Wouwe [6] to leave the insured the opportunity to get out of a life insurance contract on certain surrender dates and to be in the possibility to put a price on such an opportunity.

Notations:

V: current market value of the firm,

S: current market value of the stock, viewed as a call option on the value of the firm V,

C: current value of the compound call-option,

t: current time,

 t^* : maturity date of investment for the compound call option C,

T: maturity date of investment for the call option S,

r: risk-free rate of interest,

 σ : instantaneous variance of the return on the assets of the firm,

K: exercise price for the compound call option C,

M: exercise price for the call option S,

 $N_1(.)$: univariate cumulative normal distribution function,

 $N_2(h, k : \rho)$: bivariate cumulative normal distribution function

with h and k as upper limits and ρ as the correlation coefficient,

 \overline{V} : solution of the equation $S(V, t^*) = K$.

The economic assumptions, which are to be considered for the valuation equation of a compound option in a continuous time and using a hedging argument, are:

- there is no credit risk, only market risk,
- the market is maximally efficient, i.e. it is infinitely liquid and does not exhibit any friction,
- continuous trading is possible,
- the time evolution of the asset price is stochastic and exhibits geometric Brownian motion,
- the risk-free interest rate r and the volatility σ are constant,
- the underlying pays no dividends,
- the underlying is arbitrarely divisible,
- the market is arbitrage-free.

According to the article of Geske [2] we have the following partial differential equations:

$$\frac{\partial C}{\partial t} = rC - rV \frac{\partial C}{\partial V} - \frac{1}{2}\sigma^2 V^2 \frac{\partial^2 C}{\partial V^2}$$
 (1)

$$\frac{\partial S}{\partial t} = rS - rV \frac{\partial S}{\partial V} - \frac{1}{2}\sigma^2 V^2 \frac{\partial^2 S}{\partial V^2}, \qquad (2)$$

with boundary conditions:

$$C_{t^*} = \max(0, S_{t^*} - K) \tag{3}$$

$$S_T = \max(0, V_T - M). \tag{4}$$

Solving equation (2) leads to the well-known Black-Scholes-Merton equation:

$$S(V,t) = V.N_1(d_1) - M.e^{-r(T-t)}.N_1(d_2),$$

with
$$d_2 = \frac{ln\frac{V}{M} + (r - \frac{\sigma^2}{2}).(T - t)}{\sigma\sqrt{T - t}}$$

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

However, there is a slight difference for the value of the compounded call option, which was showed by Geske [2]:

$$C(V,t) = V.N_2(h_1, d_1; \rho) - M.e^{-r(T-t)}.N_2(h_2, d_2; \rho) - K.e^{-r(t^*-t)}.N_1(h_2),$$

with
$$h_2 = \frac{ln\frac{V}{\overline{V}} + \left(r - \frac{1}{2}\sigma^2\right)(t^* - t)}{\sigma\sqrt{t^* - t}}$$

$$h_1 = h_2 + \sigma\sqrt{t^* - t}.$$

2 Valuation of the n-fold compound option

Definition 2.1 By induction:

A compound call option (of order 2) is a call option on a call option. This can be generalized to a compound call of order k + 1 (with exercise date and price given by t_1 and K_1) with an underlying call of order k.

Now consider an expansion of the symbols from paragraph 1:

 t_i : maturity date of investment for the compound call option C_i ,

 K_i : exercise price for the compound call option C_i ,

 C_i : current value of the compound call option on the option C_{i+1} ,

 $N_k(a_1, a_2, \dots, a_k; A)$: k-variate cumulative normal distribution function with a_i as upper limits and A as the correlation matrix.

Because each C_i is function of the value of the firm V and the time t, these calls all have the same PDE:

$$\frac{\partial C_i}{\partial t} = r.C_i - r.V.\frac{\partial C_i}{\partial V} - \frac{1}{2}\sigma^2.V^2.\frac{\partial^2 C_i}{\partial V^2},$$

but with a different boundary condition:

$$C_i(V, t_i) = \max(0, C_{i+1}(V, t_i) - K_i)$$
.

The most outer call is simply derived according to Black, Scholes and Merton, while the next one can be defined following the method of Geske [2]. Further we repeatedly add a time step and solve the corresponding PDE. This results in the following theorem:

Theorem 2.1 Suppose for s = k + 1, k, ..., 2 the calls C_s are known and given by:

$$C_{s} = V.N_{k+2-s} \left(a_{s}, a_{s+1}, ..., a_{k+1}; A_{s}^{k+2-s} \right)$$

$$- \sum_{m=s}^{k+1} K_{m} e^{-r(t_{m}-t)} .N_{m+1-s} \left(b_{s}, b_{s+1}, ..., b_{m}; A_{s}^{m+1-s} \right),$$

where we use the notations

$$a_{\ell} = b_{\ell} + \sigma \sqrt{t_{\ell} - t} \qquad \ell = 2, 3, ..., k + 1$$

$$b_{\ell} = \frac{\ln \frac{V}{V_{\ell}} + (r - \frac{\sigma^{2}}{2})(t_{\ell} - t)}{\sigma \sqrt{t_{\ell} - t}} \qquad \ell = 2, 3, ..., k + 1$$

 \overline{V}_{ℓ} solution of the equation $C_{\ell+1}(V, t_{\ell}) = K_{\ell}$ $\ell = 2, 3, ..., k$

$$\begin{array}{lll} \overline{V}_{k+1} &=& M \\ \\ \rho_{ij} &=& \sqrt{\frac{t_i-t}{t_j-t}} & i < j \\ \\ A_s^\ell &=& \left(a_{ij}^\ell\right)_{i,j=1,2,\ldots,\ell} & \text{where} \left\{ \begin{array}{ll} a_{ii} &=& 1 \\ \\ a_{ij} &=& \rho_{i+s-1,j+s-1} & i < j \end{array} \right. \end{array}.$$

Then the (k+1) fold compound option can be found to be:

$$C_{1} = V.N_{k+1} \left(a_{1}, a_{2}, ..., a_{k+1}; A_{1}^{k+1} \right)$$

$$- \sum_{m=1}^{k+1} K_{m} e^{-r(t_{m}-t)} .N_{m} \left(b_{1}, b_{2}, ..., b_{m}; A_{1}^{m} \right).$$

Proof

Since C_1 is a call option, the following PDE holds for C_1 :

$$\frac{\partial C_1}{\partial t} = r.C_1 - r.V.\frac{\partial C_1}{\partial V} - \frac{1}{2}\sigma^2.V^2.\frac{\partial^2 C_1}{\partial V^2},$$

with $C_1(V, t_1) = \max(0, C_2(V, t_1) - K_1)$ as boundary condition.

Making use of the results in appendix B, the PDE for C_1 can be transformed into a diffusion equation:

$$\frac{\partial \tilde{x}}{\partial s} = \frac{\partial^2 \tilde{x}}{\partial v^2},$$

with adjusted boundary conditions for the variables p and s

$$\tilde{x}(p,0) = \begin{cases} C_2(V,t_1) - K_1 & \text{if } V \ge \overline{V}_1 \\ 0 & \text{if } V < \overline{V}_1 \end{cases}$$

 $(\overline{V}_1 \text{ being the solution of } C_2 - K_1 = 0.)$ (see [7] for a proof of the monotonicity of C with respect to V

or

or
$$\tilde{x}(p,0) = \begin{cases} V.N_k(d_2, d_3, ..., d_{k+1}; A_2^k) \\ -\sum_{m=2}^{k+1} K_m.e^{-r(t_m - t_1)}.N_{m-1}(f_2, f_3, ..., f_m; A_2^{m-1}) - K_1 & \text{if } p \ge 0 \\ 0 & \text{if } p < 0. \end{cases}$$

By introducing the notations

$$d_{\ell} = f_{\ell} + \sigma \sqrt{t_{\ell} - t_1} \qquad \ell = 2, 3, ..., k + 1$$

$$f_{\ell} = \frac{ln \frac{V}{V_{\ell}} + (r - \frac{\sigma^2}{2})(t_{\ell} - t_1)}{\sigma \sqrt{t_{\ell} - t_1}} \quad \ell = 2, 3, ..., k + 1$$

and for $\ell = 2, ..., k$ the matrices

$$A_2^{\ell} = \left(a_{ij}^{\ell}\right)_{i,j=1,2,\dots,\ell} \qquad \text{where} \begin{cases} a_{ii} = 1 \\ a_{ij} = a_{ji} = \rho_{i+1,j+1}|_{t_1} & i < j \\ \\ = \sqrt{\frac{t_{i+1} - t_1}{t_{j+1} - t_1}} \end{cases}$$

we find the following expression for V at $t = t_1$:

$$V = \overline{V}_1 \exp\left(\frac{\frac{1}{2}\sigma^2}{r - \frac{1}{2}\sigma^2}p\right),\,$$

which will be used later on.

Using a Green's function as delta-function, the expression for $\tilde{x}(p,s)$ can be written as:

$$\tilde{x}(p,s) = \int_{-\infty}^{+\infty} \tilde{x}(p',0).G(p-p',s) dp'$$
with $G(p-p',s) = \frac{1}{\sqrt{4\pi s}}.e^{-\frac{(p-p')^2}{4s}},$
or
$$\tilde{x}(p,s) = \int_{0}^{+\infty} \overline{V}_{1}.exp\left(\frac{\frac{1}{2}\sigma^2}{r-\frac{1}{2}\sigma^2}p'\right).N_{k}(d_{2},...,d_{k+1};A_{2}^{k}).\frac{1}{\sqrt{4\pi s}}.exp\left(-\frac{(p-p')^2}{4s}\right) dp'$$

$$-\int_{0}^{+\infty} \sum_{m=2}^{k+1} K_{m}.e^{-r(t_{m}-t_{1})}.N_{m-1}(f_{2},...,f_{m};A_{2}^{m-1}).\frac{1}{\sqrt{4\pi s}}.exp\left(-\frac{(p-p')^2}{4s}\right) dp'$$

$$-\int_{0}^{+\infty} K_{1}.\frac{1}{\sqrt{4\pi s}}.exp\left(-\frac{(p-p')^2}{4s}\right) dp'.$$

A substitution of $b = \frac{p'-p}{\sqrt{2s}}$ in all of the three integrals and a second substitution $b' = b - \sigma \sqrt{t_1 - t}$ in the first integral, lead to the following expression:

$$\tilde{x}(p,s) = \int_{-d_1}^{+\infty} V.e^{r(t_1-t)}.N_k(d_2^*,...,d_{k+1}^*;A_2^k).\frac{1}{\sqrt{2\pi}}.exp\left(-\frac{b'^2}{2}\right)db'$$

$$-\int_{-f_1}^{+\infty} \sum_{m=2}^{k+1} K_m \cdot e^{-r(t_m - t_1)} \cdot N_{m-1}(f_2^*, ..., f_m^*; A_2^{m-1}) \cdot \frac{1}{\sqrt{2\pi}} \cdot exp\left(-\frac{b^2}{2}\right) db$$
$$-\int_{-f_1}^{+\infty} K_1 \cdot \frac{1}{\sqrt{2\pi}} \cdot exp\left(-\frac{b^2}{2}\right) db.$$

In these calculations the new integration boundaries can be found as:

$$-f_1 = -\frac{\ln \frac{V}{\overline{V}_1} + \left(r - \frac{1}{2}\sigma^2\right)(t_1 - t)}{\sigma\sqrt{t_1 - t}}$$
$$-d_1 = -(f_1 + \sigma\sqrt{t_1 - t}),$$

while for $\ell = 2, 3, ..., k+1$ we get:

$$f_{\ell} = \frac{ln \frac{\overline{V}_{1}}{\overline{V}_{\ell}} + \frac{\frac{1}{2}\sigma^{2}}{r - \frac{1}{2}\sigma^{2}} p' + \left(r - \frac{1}{2}\sigma^{2}\right) (t_{\ell} - t_{1})}{\sigma \sqrt{t_{\ell} - t_{1}}}$$

$$f_{\ell}^{*} = \frac{b_{\ell} + \rho_{1\ell} b}{\sqrt{1 - \rho_{1\ell}^{2}}} \quad \text{if } \rho_{1\ell} = \sqrt{\frac{t_{1} - t}{t_{\ell} - t}}$$

and

$$d_{\ell} = \frac{ln \frac{\overline{V}_{1}}{\overline{V}_{\ell}} + \frac{\frac{1}{2}\sigma^{2}}{r - \frac{1}{2}\sigma^{2}} p' + \left(r + \frac{1}{2}\sigma^{2}\right) (t_{\ell} - t_{1})}{\sigma \sqrt{t_{\ell} - t_{1}}}$$

$$d_{\ell}^{*} = \frac{a_{\ell} + \rho_{1\ell} b'}{\sqrt{1 - \rho_{1\ell}^{2}}}.$$

An application of theorem A.1 in the first integral leads to the final expression for $\tilde{x}(p,s)$:

$$\tilde{x}(p,s) = V.e^{r(t_1-t)}.N_{k+1}(d_1, a_2, ..., a_{k+1}; A_1^{k+1})$$

$$-\sum_{m=2}^{k+1} K_m.e^{-r(t_m-t_1)}.N_m(f_1, b_2, ..., b_m; A_1^m) - K_1.N_1(f_1),$$

where the correlation matrices can be written as:

$$A_1^{k+1} = (a_{ij})_{i,j=1,2,\cdots,k+1}$$
 with $\begin{cases} a_{ii} = 1 \\ a_{ij} = a_{ji} = \rho_{ij} \end{cases}$ if $i < j$

and for
$$m = 2, 3, \dots, k+1$$
:
$$A_1^m = (a_{ij})_{i,j=1,2,\dots,m} \quad \text{with } \begin{cases} a_{ii} = 1 \\ a_{ij} = a_{ji} = \rho_{ij} \end{cases} \quad \text{if } i < j .$$

The final form for the (k+1) fold compound call option therefore equals

$$C_1(V,t) = V.N_{k+1}(a_1, a_2, ..., a_{k+1}; A_1^{k+1})$$

$$-\sum_{m=2}^{k+1} K_m.e^{-r(t_m-t)}.N_m(b_1, b_2, ..., b_m; A_1^m) - K_1.e^{-r(t_1-t)}.N_1(b_1),$$

with

$$a_{\ell} = b_{\ell} + \sigma \sqrt{t_{\ell} - t} \qquad \qquad \ell = 1, 2, ..., k + 1$$

$$b_{\ell} = \frac{\ln \frac{V}{\overline{V_{\ell}}} + (r - \frac{\sigma^2}{2})(t_{\ell} - t)}{\sigma \sqrt{t_{\ell} - t}} \qquad \qquad \ell = 1, 2, ..., k + 1$$

$$\overline{V}_{\ell} \qquad \text{determined by } C_{\ell+1}(V, t_{\ell}) = K_{\ell} \qquad \qquad \ell = 1, 2, ..., k$$

$$\overline{V}_{k+1} = M \qquad \qquad i < j \qquad \qquad i < j$$

$$A_{1}^{\ell} = \left(a_{ij}^{\ell}\right)_{i,j=1,2,...,\ell} \qquad \qquad \text{where } \begin{cases} a_{ii} = 1 \\ a_{ij} = a_{ji} = \rho_{ij} & i < j \end{cases} .$$

3 Conclusions

The notion of a n-fold compound option is introduced as a generalization of the compound option by Geske [2]. The closed-form analytic expression for this n-fold compound option is proved by induction and by using some interesting results on the relationship between (k+1)-variate normal distributions and k-variate normal distributions.

A A useful relationship between the k + 1 th multivariate distribution function and the k th multivariate distribution function

In theorem A.1 we will need the following two lemmas about a matrix and its inverse. Both of the lemmas can be proved in a straightforward way.

Lemma A.1 If

$$A = \begin{bmatrix} 1 & \frac{-a_{12}}{\sqrt{1 - a_{12}^2}} & \cdots & \frac{-a_{1k}}{\sqrt{1 - a_{1k}^2}} \\ 0 & \frac{1}{\sqrt{1 - a_{12}^2}} & O \\ \vdots & & \ddots & \\ 0 & O & \frac{1}{\sqrt{1 - a_{1k}^2}} \end{bmatrix}$$

then

$$A^{-1} = \begin{bmatrix} 1 & a_{12} & \cdots & a_{1k} \\ \hline 0 & \sqrt{1 - a_{12}^2} & & O \\ \vdots & & \ddots & \\ 0 & O & & \sqrt{1 - a_{1k}^2} \end{bmatrix}$$

by the use of the principle of partitioning.

Lemma A.2

If
$$A = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & & & \\ \vdots & B^{-1} & \end{bmatrix}$$
, then $A^{-1} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & & & \\ \vdots & & B & \end{bmatrix}$.

Let N_k be the k-variate normal distribution function and N_{k-1} the (k-1)-variate normal distribution function, the following expression can be determined between N_k and N_{k-1} .

Theorem A.1

$$\int_{-m_1}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x_1^2} N_{k-1} \left(\frac{m_2 + \rho_{12} x_1}{\sqrt{1 - \rho_{12}^2}}, \cdots, \frac{m_k + \rho_{1k} x_1}{\sqrt{1 - \rho_{1k}^2}}; B \right) dx_1 = N_k(m_1, \cdots, m_k; C),$$

with $C = (c_{ij})_{i,j=1,\cdots,k}$ a symmetric matrix with

$$\begin{cases}
c_{11} = 1 \\
c_{1j} = \rho_{1j} \\
c_{ij} = \rho_{1i} \rho_{1j} + \sqrt{(1 - \rho_{1i}^2)(1 - \rho_{1j}^2)} b_{i-1,j-1}
\end{cases}$$

and where for convenience we put

$$N_0 = 1$$
.

Proof by induction

For k = 1 the result is straightforward.

For the second part of the proof we first rewrite the integral as

with
$$P(x_1, \dots, x_k) = \begin{bmatrix} x_1 & x_2 & \cdots & x_k \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & & & \\ \vdots & & B^{-1} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_k \end{bmatrix}$$
.

Making use of the substitution

$$y_j = x_j \sqrt{1 - \rho_{1j}^2} + \rho_{1j} x_1$$
 for $j = 2, 3, \dots, k$

we can rewrite this expression as

$$= \int_{-m_1}^{+\infty} \cdots \int_{-m_k}^{+\infty} \frac{1}{\sqrt{(2\pi)^k \det B \prod_{j=2}^k (1-\rho_{1j}^2)}} \cdot e^{-\frac{1}{2}P^*(x_1, y_2, \dots, y_k)} dx_1 dy_2 \cdots dy_k.$$

In this formula, we introduced the matrix:

$$P^{*}(x_{1}, y_{2}, \dots, y_{k}) = \begin{bmatrix} x_{1} & y_{2} & \cdots & y_{k} \end{bmatrix} . D . \begin{bmatrix} \frac{1}{0} & 0 & \cdots & 0 \\ \vdots & B^{-1} & \\ 0 & & \end{bmatrix} . D^{t} . \begin{bmatrix} x_{1} & y_{2} & \\ \vdots & y_{k} & \\ \end{bmatrix},$$

where

$$D = \begin{bmatrix} 1 & \frac{-\rho_{12}}{\sqrt{1 - \rho_{12}^2}} & \cdots & \frac{-\rho_{1k}}{\sqrt{1 - \rho_{1k}^2}} \\ 0 & \frac{1}{\sqrt{1 - \rho_{12}^2}} & O \\ \vdots & & \ddots & \\ 0 & O & \frac{1}{\sqrt{1 - \rho_{1k}^2}} \end{bmatrix}.$$

Since we want to express the k-variate integration by means of a k-variate normal CDF, we now have to determine the correlation matrix C with:

$$C^{-1} = D. \begin{bmatrix} 1 & 0 & \cdots & 0 \\ \hline 0 & & & \\ \vdots & & B^{-1} & \\ 0 & & & \end{bmatrix} .D^t.$$

An application of lemma A.1 and A.2 leads to

$$C = \begin{bmatrix} \frac{1}{\rho_{12}} & 0 & \cdots & 0 \\ \hline \rho_{12} & \sqrt{1 - \rho_{12}^2} & O \\ \vdots & & \ddots & \\ \rho_{1k} & O & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot \begin{bmatrix} \frac{1}{0} & 0 & \cdots & 0 \\ \hline 0 & & & \\ \vdots & & & \\ 0 & & & B \end{bmatrix} \\ \cdot \begin{bmatrix} \frac{1}{\rho_{12}} & \cdots & \rho_{1k} \\ \hline 0 & \sqrt{1 - \rho_{12}^2} & O \\ \vdots & & \ddots & \\ 0 & O & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix},$$

or

$$C = \begin{bmatrix} 1 & \rho_{12} & \cdots & \rho_{1k} \\ \hline \rho_{12} & & \\ \vdots & & \mathbf{F} \\ \hline \rho_{1k} & & \end{bmatrix},$$

with F obtained by partitioning as

$$\begin{bmatrix} \rho_{12} \\ \vdots \\ \rho_{1k} \end{bmatrix} \cdot \begin{bmatrix} \rho_{12} & \cdots & \rho_{1k} \end{bmatrix} + \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \ddots & \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{12}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \sqrt{1 - \rho_{1k}^2} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt{1 - \rho_{1k}^2} & & \mathbf{O} \\ & \mathbf{O} & & \mathbf{O} \end{bmatrix} \cdot B \cdot \begin{bmatrix} \sqrt$$

Hence this results in:

$$C = (c_{ij})_{i,j=1,\dots,k} \quad \text{with} \quad \begin{cases} c_{11} = 1 \\ c_{1j} = \rho_{1j} \\ c_{ij} = \rho_{1i} \rho_{1j} + \sqrt{(1 - \rho_{1i}^2)(1 - \rho_{1j}^2)} b_{i-1,j-1} \end{cases}$$

Now, since $P^*(x_1, y_2, \dots, y_k)$ can be written as

$$P^*(x_1, y_2, \dots, y_k) = \begin{bmatrix} x_1 & y_2 & \cdots & y_k \end{bmatrix} . C^{-1} . \begin{bmatrix} x_1 \\ y_2 \\ \vdots \\ y_k \end{bmatrix}$$

and

$$\det C = \prod_{j=2}^{k} (1 - \rho_{1j}^2). \det B$$

for the integral we find

$$\int_{-m_1}^{+\infty} \dots \int_{-m_k}^{+\infty} \frac{1}{\sqrt{(2\pi)^k \det B \prod_{j=2}^k (1-\rho_{1j}^2)}} e^{-\frac{1}{2}P^*(x_1, y_2, \dots, y_k)} dx_1 dy_2 \dots dy_k$$

$$= \int_{-m_1}^{+\infty} \dots \int_{-m_k}^{+\infty} \frac{1}{\sqrt{(2\pi)^k \det C}} e^{-\frac{1}{2}P^*(x_1, y_2, \dots, y_k)} dx_1 dy_2 \dots dy_k$$

or

$$=N_k(m_1,\ldots,m_k;C),$$

which completes the proof.

B From PDE to diffusion equation

Consider the PDE:

$$\frac{\partial C_i}{\partial t} = r.C_i - r.V.\frac{\partial C_i}{\partial V} - \frac{1}{2}\sigma^2.V^2.\frac{\partial^2 C_i}{\partial V^2},$$

with boundary condition at time t_i given by

$$C_i(V, t_i) = \max(0, C_{i+1}(V, t_i) - K_i).$$

Let \overline{V}_i be defined as the value for which $C_{i+1}(\overline{V}_i, t_i) = K_i$.

Making use of some substitutions, we can rewrite this PDE as a diffusion equation.

Indeed, first we choose w as

$$w = ln \frac{V}{\overline{V}_i},$$

and we define the function x(w,t) as

$$x(w,t) = e^{r(t_i-t)}.C_i(V,t)$$

$$= e^{r(t_i-t)}.C_i(\overline{V}_i.e^w,t).$$

Secondly, we rescale the independent variables as

$$\begin{cases} w' = \frac{r - \frac{1}{2}\sigma^2}{\frac{1}{2}\sigma^2}w \\ s = \frac{\left(r - \frac{1}{2}\sigma^2\right)^2}{\frac{1}{2}\sigma^2}(t_i - t), \end{cases}$$

we define the function $\hat{x}(w',s)$

$$\hat{x}(w', s) = x(w, t).$$

With

$$p = w' + s,$$

finally we rewrite the dependent variable as

$$\tilde{x}(p,s) = \hat{x}(w',s).$$

Then it follows in a straightforward way that this last function satisfies the diffusion equation

$$\frac{\partial \tilde{x}}{\partial s} = \frac{\partial^2 \tilde{x}}{\partial v^2}.$$

See [5] for more details.

C Green's function

Consider Green's function

$$G(z-z',t') = \frac{1}{\sqrt{4\pi t'}} e^{-\frac{(z-z')^2}{4t'}},$$

which clearly satisfies the diffusion equation

$$\frac{\partial G}{\partial t'} = \frac{\partial^2 G}{\partial z^2}.$$

Note that G behaves like a delta-function in t' = 0:

* If
$$z \neq z'$$
 and $t' \rightarrow 0$ then $G(z-z',t') \rightarrow 0$

* If
$$z = z'$$
 and $t' \to 0$ then $G(z - z', t') \to \infty$

*

$$\int_{-\infty}^{+\infty} G(z - z', t') dz' = -\int_{+\infty}^{-\infty} \frac{1}{\sqrt{2\pi}} \exp(-\frac{q^2}{2}) dq$$
$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \exp(-\frac{q^2}{2}) dq$$
$$= 1$$

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