

### Black-Scholes Model Analytics

- Model Dynamics:

The Black-Scholes model assumes the following SDE under the Risk-Neutral measure:

$$dS_t = rS_t dt + \sigma S_t dW_t$$

Where:

- $S_t$  is the underlying asset spot price
- $r$  is the zero-coupon rate
- $\sigma$  is the underlying asset volatility
- $W_t$  is a standard Brownian motion

This SDE admits a unique solution:

$$S_t = S_0 \times e^{\left(r - \frac{\sigma^2}{2}\right) \times t + \sigma W_t}; \forall t \geq 0$$

- Black-Scholes Vanilla Price:

Let us assume a Vanilla Call paying the following cash-flow at  $T > 0$ :  $MAX(S_T - K; 0)$

Its price  $C_0$  under the Risk-Neutral measure is:  $e^{-rT} \times E[MAX(S_T - K; 0)]$ , where  $E[-]$  is the expectation operator

The expected value is analytically computed under Black-Scholes, and hence, the price  $C_0$ :

$$C_0 = e^{-rT} (F_T N(d_1) - K N(d_2))$$

$$F_t = E(S_t) = S_0 \times e^{rt} \text{ (The underlying asset } T\text{-forward price)}$$

$$d_1 = \frac{\ln(F_T) + \sigma^2 T / 2}{\sigma \sqrt{T}}$$

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

Using the Call-Put parity, the Vanilla Put price  $P_0$  is:

$$P_0 = e^{-rT} (-F_T N(-d_1) + K N(-d_2))$$

- Black-Scholes Digital Price:

Let us assume a Digital Call paying the following cash-flow at  $T > 0$ :  $1_{S_T > K}$

Its price  $C_0$  under the Risk-Neutral measure is:  $e^{-rT} \times E[1_{S_T > K}] = e^{-rT} \times P(S_T > K)$

The probability value is analytically computed under Black-Scholes, and hence, the price  $C_0$ :

$$C_0 = e^{-rT} N(d_2)$$

Similarly, the Digital Put price  $P_0$  is:

$$P_0 = e^{-rT} N(-d_2)$$

- Black-Scholes Asian Price: *Moment matching method*

Let us assume an Arithmetic Asian Call paying the following cash-flow at  $T > 0$ :

$$\text{MAX} \left( \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i} \times S_{t_i} - K ; 0 \right)$$

Where:  $(\alpha_{t_i})_i$  are the normalized weights

By approximating the arithmetic average with a log-normal distribution, Asian options can be priced via closed formulas in the Black-Scholes framework

Since:

$$\begin{aligned} m_1 &= E \left( \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i} \times S_{t_i} \right) \\ &= \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i} E(S_{t_i}) \\ &= \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i} F_{t_i} \end{aligned}$$

And:

$$\begin{aligned} m_2 &= E \left( \left( \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i} \times S_{t_i} \right)^2 \right) \\ &= \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i}^2 E(S_{t_i}^2) + 2 \times \sum_{t_i < t_j} \alpha_{t_i} \alpha_{t_j} E(S_{t_i} S_{t_j}) \\ &= \sum_{t_i \in \text{Fixing Dates}} \alpha_{t_i}^2 F_{t_i}^2 e^{\sigma^2 t_i} + 2 \times \sum_{t_i < t_j} \alpha_{t_i} \alpha_{t_j} F_{t_i} F_{t_j} e^{\sigma^2 t_i} \\ &= \sum_{t_i \in \text{Fixing Dates}} \left[ \alpha_{t_i} F_{t_i} e^{\sigma^2 t_i} \times \left( \alpha_{t_i} F_{t_i} + 2 \times \sum_{t_i < t_j} \alpha_{t_j} F_{t_j} \right) \right] \\ &= \sum_{t_i \in \text{Fixing Dates}} \left[ \alpha_{t_i} F_{t_i} e^{\sigma^2 t_i} \times \left( 2 \times \sum_{t_i \leq t_j} \alpha_{t_j} F_{t_j} - \alpha_{t_i} F_{t_i} \right) \right] \end{aligned}$$

Then, Asian Black-Scholes price formula is similar to the Vanilla's one with:

$$\begin{aligned} d_1 &= \frac{\ln\left(\frac{m_1}{K}\right) + \frac{\ln\left(\frac{m_2}{m_1^2}\right)}{2}}{\sqrt{\ln\left(\frac{m_2}{m_1^2}\right)}} = \frac{\ln\left(\frac{\sqrt{m_2}}{K}\right)}{\sqrt{\ln\left(\frac{m_2}{m_1^2}\right)}} \\ d_2 &= d_1 - \sqrt{\ln\left(\frac{m_2}{m_1^2}\right)} \end{aligned}$$

- Black-Scholes Basket Price: *Moment matching method*

Let us assume a Basket Call paying the following cash-flow at  $T > 0$ :

$$MAX\left(\sum_{i \in \text{Basket}} \alpha_i \times S_T^i - K; 0\right)$$

Where:  $(\alpha_i)_i$  are the normalized weights and  $(S^i)_i$  are  $\rho$  – Correlated

Under the Black-Scholes model:  $S_t^i = S_0^i \times e^{\left(r - \frac{\sigma_i^2}{2}\right) \times t + \sigma_i W_t^i}$ , where:  $d < W^i, W^j >_t = \rho_{i,j} dt$

By approximating the basket value with a log-normal distribution, Basket options can be priced via closed formulas in the Black-Scholes framework

Since:

$$m_1 = E\left(\sum_{i \in \text{Basket}} \alpha_i \times S_T^i\right) = \sum_{i \in \text{Basket}} \alpha_i \times E(S_T^i) = \sum_{i \in \text{Basket}} \alpha_i \times F_T^i$$

And:

$$\begin{aligned} m_2 &= E\left(\left(\sum_{i \in \text{Basket}} \alpha_i \times S_T^i\right)^2\right) \\ &= \sum_{i \in \text{Basket}} \sum_{j \in \text{Basket}} \alpha_i \times \alpha_j \times E(S_T^i S_T^j) \\ &= \sum_{i \in \text{Basket}} \sum_{j \in \text{Basket}} \alpha_i \alpha_j F_T^i F_T^j e^{\sigma_i \sigma_j \rho_{i,j} T} \end{aligned}$$

Then, Basket Black-Scholes price formula is similar to the Vanilla's one with:

$$\begin{aligned} d_1 &= \frac{\ln\left(\frac{m_1}{K}\right) + \frac{\ln\left(\frac{m_2}{m_1^2}\right)}{2}}{\sqrt{\ln\left(\frac{m_2}{m_1^2}\right)}} = \frac{\ln\left(\frac{\sqrt{m_2}}{K}\right)}{\sqrt{\ln\left(\frac{m_2}{m_1^2}\right)}} \\ d_2 &= d_1 - \sqrt{\ln\left(\frac{m_2}{m_1^2}\right)} \end{aligned}$$

- Black-Scholes Spread Price: *Kirk's approximation & Margrabe's formula*

Let us assume a Spread Call paying the following cash-flow at  $T > 0$ :  $MAX(S_T^1 - S_T^2 - K; 0)$

This payoff can also be written as:  $MAX\left(S_T^1 - S_T^{2Adj}; 0\right)$ , where:  $S_T^{2Adj} = S_T^2 + K$

In this case, under the absence of arbitrage assumption:  $S_t^{2Adj} = S_t^2 + K e^{-r(T-t)}$

As a consequence:

$$\begin{aligned} dS_t^{2Adj} &= dS_t^2 + rKe^{-r(T-t)}dt \\ &= r(S_t + Ke^{-r(T-t)})dt + \sigma_2 S_t^2 dW_t^2 \\ &= rS_t^{2Adj}dt + \sigma_2^{Adj}(t)S_t^{2Adj}dW_t^2 \end{aligned}$$

Where:  $\sigma_2^{Adj}(t) = \frac{S_t^2}{S_t^2 + Ke^{-r(T-t)}} \sigma_2$

The Kirk's approximation stands on freezing the  $\sigma_2^{Adj}(t)$  at  $t = 0$

Therefore,  $S_t^{2Adj}$  will become a geometric Brownian motion:

$$dS_t^{2Adj} = rS_t^{2Adj}dt + \sigma_2^{Adj}S_t^{2Adj}dW_t^2$$

Where:

$$\sigma_2^{Adj} = \frac{S_0^2}{S_0^2 + Ke^{-rT}} \sigma_2$$

Finally, the Margrabe's formula gives the following price  $C_0$ :

$$C_0 = S_0^1 N(d_1) - (S_0^2 + Ke^{-rT}) N(d_2)$$

$$\sigma = \sqrt{\sigma_1^2 + \sigma_2^{Adj2} - 2\sigma_1\sigma_2^{Adj}\rho_{1,2}}$$

$$d_1 = \frac{\ln\left(\frac{S_0^1}{S_0^2 + Ke^{-rT}}\right) + \sigma^2 T/2}{\sigma\sqrt{T}}$$

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

Similarly, the Spread Put price  $P_0$  is:  $C_0 = -S_0^1 N(-d_1) + (S_0^2 + Ke^{-rT}) N(-d_2)$

- Monte-Carlo Simulation:

- Single Asset:

The Black-Scholes spot price solution is directly used for simulations:

$$S_{n+1} = S_n \times e^{\left(r - \frac{\sigma^2}{2}\right) \times (t_{n+1} - t_n) + \sigma(W_{n+1} - W_n)}$$

Where:  $(W_{n+1} - W_n)$  follows a normal distribution  $N(\mu = 0, \sigma = \sqrt{t_{n+1} - t_n})$

Therefore:

$$S_{n+1} = S_n \times e^{\left(r - \frac{\sigma^2}{2}\right) \times (t_{n+1} - t_n) + \sigma\sqrt{t_{n+1} - t_n}G_{n+1}}$$

Where:  $G_{n+1}$  follows a normal distribution  $N(\mu = 0, \sigma = 1)$

- Multi Asset:

Let us assume  $N$   $\rho$  – Correlated assets. Under the Black-Scholes model:

$$S_t^i = S_0 \times e^{\left(r - \frac{\sigma_i^2}{2}\right) \times t + \sigma_i W_t^i}; \forall t \geq 0 \forall i \in [1, N]$$

Where:  $d < W^i, W^j >_t = \rho_{i,j} dt$

The Cholesky decomposition of the correlation matrix  $\rho = (\rho_{i,j})_{1 \leq i,j \leq N}$  is used to simulate the  $N$  assets independently. To do so, the correlation matrix must be positive-definite, and hence, the values  $\rho_{i,j}$  must be within the following interval:  $] -\frac{1}{N-1}, 1[$

Therefore, a new vector of independent standard Brownian motions  $(B_i)_{1 \leq i \leq N}$  is implied such as:  $W_t = LB_t$ , where:  $L$  is the Cholesky lower triangular matrix ( $\rho = LL^T$ )

Consequently:  $S_t^i = S_0 \times e^{\left(r - \frac{\sigma_i^2}{2}\right) \times t + \sigma L_i B_t}$ ;  $\forall t \geq 0 \forall i \in [1, N]$

Where:  $L_i$  is the  $i^{th}$  line of  $L$

Finally:  $S_{n+1}^i = S_n^i \times e^{\left(r - \frac{\sigma_i^2}{2}\right) \times (t_{n+1} - t_n) + \sigma_i \sqrt{t_{n+1} - t_n} L_i G_{n+1}}$

Where:  $G_{n+1}$  is an  $N$  – Vector of independent normal distributions  $N(\mu = 0, \sigma = 1)$