The intuition behind Black's formula

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1 Introduction

In 1973, Black and Scholes released their seminal paper on pricing European stock options. Black would extend their results in a 1976 paper applying the Black-Scholes pricing PDE methodology to pricing European options on commodity futures. However, solving the Black-Scholes PDE by means of transformation to the heat equation proffers no intuition on the form of the pricing formula. An alternate approach using the risk-neutral pricing method allows for a simpler, more direct derivation of the formula, and allows one to reexpress the formula in terms of risk-neutral conditional probabilities that provide an intuitive understanding of option price dynamics. In this short article, we derive the Black formula from the risk-neutral pricing formula, and use our derivation to rewrite the formula in a way that isolates the volatility premium, the portion of the option value due to the total volatility of the underlying, defined as $\sigma \sqrt{T-t}$.

2 A conditional expectation

Black's model for futures prices under the risk-neutral measure $\mathbb Q$ is

$$dF(t) = \sigma F(t)dW_t \tag{2.1}$$

Here W_t is a Wiener process under \mathbb{Q} , and we know that $W_t \sim \mathcal{N}(0, t)$. By the risk-neutral pricing formula, we can write the time t value C(F, t) of a European call option on F as the conditional expectation

$$C(F, t) = \mathbb{E}[D(t, T) \max\{F(T) - K, 0\} \mid \mathcal{F}_t]$$

Here F = F(t) is the current futures price, T is the time of option expiration, and K is the fixed strike price of the option. D(t, T) is a deterministic discount factor, which Black specifies as $e^{-r(T-t)}$, with r being a known, constant riskless discount rate. \mathcal{F}_t is the natural filtration of the process F(t). By the definition, we can rewrite the risk-neutral conditional expectation as

$$C(F, t) = D(t, T) \int_{-\infty}^{\infty} \max\{f - K, 0\} \varphi_{\tau}(f, T \mid \mathcal{F}_t) df = D(t, T) \int_{K}^{\infty} (f - K) \varphi_{\tau}(f, T \mid \mathcal{F}_t) df$$

Here $\varphi_{\tau}(F, T \mid \mathcal{F}_t)$ is the risk-neutral probability density for F(T) conditioned on the natural filtration \mathcal{F}_t , where $\tau \geq T \geq t$ is a delivery date. For brevity, we drop the explicit dependence on \mathcal{F}_t , T and use the simpler notation $\widetilde{\varphi}_{\tau}(F)$ instead. Dividing by D(t, T) and expanding the remaining integral, we have

$$\frac{1}{D(t, T)}C(F, t) = \int_{K}^{\infty} (f - K)\widetilde{\varphi}_{\tau}(f)df = \int_{K}^{\infty} f\widetilde{\varphi}_{\tau}(f)df - K \int_{K}^{\infty} \widetilde{\varphi}_{\tau}(f)df
= \int_{K}^{\infty} f\widetilde{\varphi}_{\tau}(f)df - K\mathbb{Q}(F(T) \ge K \mid \mathcal{F}_{t})$$
(2.2)

We see that the expectation can be written as the difference between a partial conditional expectation and K times a risk-neutral probability conditioned on the natural filtration \mathcal{F}_t . To simplify our computation of the partial expectation, we recall that the solution to (2.1) is given by the equation

$$F(t) = F_0 \exp\left(-\frac{1}{2}\sigma^2 t + \sigma W_t\right)$$

We want to find F(T) conditioned on \mathcal{F}_t . We set W_t to 0, and by the properties of the Wiener process, we write $W_{t \dots T}$ as $\sqrt{T-t} Z$, where $Z \sim \mathcal{N}(0, 1)$ under \mathbb{Q} . Substituting F for F(t), we have that

$$F(T) \mid \mathcal{F}_t = F \exp\left[-\frac{1}{2}\sigma^2(T-t) + \sigma\sqrt{T-t} Z\right]$$
 (2.3)

For notational simplicity, we set $\theta = T - t$. To simplify the partial expectation, we make note that F(T) conditioned on \mathcal{F}_t is a function of a standard normal random variable Z. Therefore, using (2.3), we can rewrite the partial expectation in terms of the standard normal density function $\phi(z)$ instead to get

$$\int_{K}^{\infty} f \widetilde{\varphi}_{\tau}(f) df = F \int_{\zeta}^{\infty} e^{-\frac{1}{2}\sigma^{2}\theta + \sigma\sqrt{\theta}z} \phi(z) dz = F \int_{\zeta}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\sigma^{2}\theta + \sigma\sqrt{\theta}z - \frac{1}{2}z^{2}} dz$$
 (2.4)

The last step follows from the definition of the standard normal density function $\phi(z)$, given below as

$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$$

Since we are changing our density function and also the random variable we are integrating, we need to change our threshold from K to ζ . We find ζ by solving the equation

$$K = F \exp\left(-\frac{1}{2}\sigma^2\theta + \sigma\sqrt{\theta} \zeta\right)$$

This relates ζ to K through the relation between $F(T) \mid \mathcal{F}_t$ and Z, and we see that ζ is

$$\zeta = \frac{\log(K/F) + \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}} \tag{2.5}$$

Here log is the natural logarithm. From (2.4), we see that the exponentiated polynomial is a square, so

$$\int_{K}^{\infty} f \widetilde{\varphi}_{\tau}(f) df = F \int_{\zeta}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(z - \sigma\sqrt{\theta})^{2}} dz = F \mathbb{Q}(\chi \ge \zeta)$$

Here we have $\chi \sim \mathcal{N}(\sigma\sqrt{\theta}, 1)$, as we see that the function in our integral is another normal density. Since $\chi = Z + \sigma\sqrt{\theta}$, we can rewrite $\mathbb{Q}(\chi \geq \zeta)$ in terms of Z, and substituting for ζ the result in (2.5), we have

$$\mathbb{Q}(\chi \ge \zeta) = \mathbb{Q}\left(Z + \sigma\sqrt{\theta} \ge \frac{\log(K/F) + \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) = \mathbb{Q}\left(Z \ge \frac{\log(K/F) - \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) \\
= \mathbb{Q}\left(Z \le \frac{\log(F/K) + \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) = \Phi\left(\frac{\log(F/K) + \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) \tag{2.6}$$

Here the second to last step follows from the symmetry of Z around 0, and the function $\Phi(z)$ is the standard normal cdf. Using our result from (2.6), and replacing θ with T-t, we have that

$$\int_{K}^{\infty} f \widetilde{\varphi}_{\tau}(f) df = F \Phi \left(\frac{\log(F/K) + \frac{1}{2} \sigma^{2} (T - t)}{\sigma \sqrt{T - t}} \right)$$
(2.7)

This is the first normal cdf term in Black's formula. Moving on to the risk-neutral conditional probability $\mathbb{Q}(F(T) \geq K \mid \mathcal{F}_t)$ we defined in (2.2), we again recall our expression for $F(T) \mid \mathcal{F}_t$ in (2.3). Substituting θ for T - t as before, we can rewrite $\mathbb{Q}(F(T) \geq K \mid \mathcal{F}_t)$ in terms of Z, where again $Z \sim \mathcal{N}(0, 1)$, as

$$\mathbb{Q}(F(T) \ge K \mid \mathcal{F}_t) = \mathbb{Q}\left(Fe^{-\frac{1}{2}\sigma^2\theta + \sigma\sqrt{\theta}Z} \ge K\right) = \mathbb{Q}\left(Z \ge \frac{\log(K/F) + \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) \\
= \mathbb{Q}\left(Z \le \frac{\log(F/K) - \frac{1}{2}\sigma^2\theta}{\sigma\sqrt{\theta}}\right) = \Phi\left(\frac{\log(F/K) - \frac{1}{2}\sigma^2(T - t)}{\sigma\sqrt{T - t}}\right) \tag{2.8}$$

Again we replace θ with T-t, and finally can complete our derivation. Substituting our results from (2.7) and (2.8) back into (2.2) and multiplying by $D(t, T) = e^{-r(T-t)}$, we get exactly Black's formula, given by

$$C(F, t) = D(t, T) \left[F\Phi\left(\frac{\log(F/K) + \frac{1}{2}\sigma^2(T - t)}{\sigma\sqrt{T - t}}\right) - K\Phi\left(\frac{\log(F/K) - \frac{1}{2}\sigma^2(T - t)}{\sigma\sqrt{T - t}}\right) \right]$$
(2.9)

3 The volatility premium

The main benefit of deriving Black's formula from the risk-neutral pricing formula is that one can re-express the normal cdf terms as risk-neutral probabilities in terms of F(T), which is the inverse of what was done when deriving the formula. Since we have already shown the relationship between the second normal cdf term and its corresponding \mathbb{Q} probability in (2.8), we work on the first normal cdf term. From (2.3), we have

$$Z = \frac{\log(F(T)/F) + \frac{1}{2}\sigma^2(T-t)}{\sigma\sqrt{T-t}} \mid \mathcal{F}_t \sim \mathcal{N}(0, 1)$$
(3.1)

Interchanging T-t and θ as usual, we can re-express the first normal cdf term in (2.9) using (3.1) as

$$\Phi\left(\frac{\log(F/K) + \frac{1}{2}\sigma^{2}(T - t)}{\sigma\sqrt{T - t}}\right) = \mathbb{Q}\left(Z \le \frac{\log(F/K) + \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}}\right) = \mathbb{Q}\left(Z \ge \frac{\log(K/F) - \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}}\right)$$

$$= \mathbb{Q}\left(\frac{\log(F(T)/F) + \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}} \ge \frac{\log(K/F) - \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}} \mid \mathcal{F}_{t}\right)$$

$$= \mathbb{Q}\left(F(T) \ge Ke^{-\sigma^{2}(T - t)} \mid \mathcal{F}_{t}\right)$$
(3.2)

Substituting our results from (2.8) and (3.2) into the Black formula given in (2.9), we have

$$C(F, t) = D(t, T) \left[F \mathbb{Q} \left(F(T) \ge K e^{-\sigma^2(T-t)} \mid \mathcal{F}_t \right) - K \mathbb{Q}(F(T) \ge K \mid \mathcal{F}_t) \right]$$
(3.3)

The intuition here is that under the Black-Scholes model, one can express the value of a European call option as the discounted difference between the current value of the underlying, here the futures price F, and the strike K, each weighted by a risk-neutral probability of exercise. One can observe that the first exercise probability $\mathbb{Q}(F(T) \ge Ke^{-\sigma^2(T-t)})$ has the strike weighted by the total variance $\sigma^2(T-t)$, such that ceteris paribus, if σ increases, C(F, t) increases. This expresses the fact that as volatility increases, ceteris paribus, the value of an option increases. We can quantify this by separating out the volatility premium¹, as

$$\mathbb{Q}\left(F(T) \ge Ke^{-\sigma^{2}(T-t)} \mid \mathcal{F}_{t}\right) = \Phi\left(\frac{\log(F/K) + \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}}\right) = \mathbb{Q}\left(Z \le -\zeta + \sigma\sqrt{\theta}\right) = \int_{-\infty}^{-\zeta + \sigma\sqrt{\theta}} \phi(z)dz$$
$$\mathbb{Q}\left(F(T) \ge K \mid \mathcal{F}_{t}\right) = \Phi\left(\frac{\log(F/K) - \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}}\right) = \mathbb{Q}(Z \le -\zeta) = \int_{-\infty}^{-\zeta} \phi(z)dz$$

Here we have ζ as defined in (2.5), $\theta = T - t$, and $\phi(z)$ is the normal pdf. We can thus see that

$$\mathbb{Q}\left(F(T) \ge Ke^{-\sigma^2(T-t)} \mid \mathcal{F}_t\right) = \mathbb{Q}(F(T) \ge K \mid \mathcal{F}_t) + \int_{-\zeta}^{-\zeta + \sigma\sqrt{T-t}} \phi(z)dz$$

We can therefore define the volatility premium as

$$\lambda_{\sigma}(F, \theta) = \int_{-\zeta}^{-\zeta + \sigma\sqrt{\theta}} \phi(z)dz, \text{ where } -\zeta = \frac{\log(F/K) - \frac{1}{2}\sigma^{2}\theta}{\sigma\sqrt{\theta}}$$
 (3.4)

Here we emphasize the time-dependent parameters only, as all other parameters are constants. We know that $\lambda_{\sigma} \geq 0$ as by definition of a density function, $\phi(z) \geq 0$, and because $\sigma > 0$ and $\theta \geq 0$. Using our new results in (3.4), we can now rewrite (3.3) with only $\mathbb{Q}(F(T) \geq K \mid \mathcal{F}_t)$ while incorporating λ_{σ} to arrive at

$$C(F, t) = D(t, T) \left[\mathbb{Q}(F(T) \ge K \mid \mathcal{F}_t)(F - K) + F \int_{-\zeta}^{-\zeta + \sigma\sqrt{T - t}} \phi(z) dz \right]$$
(3.5)

We have now expressed C(F, t) as a sum of the difference F - K between the underlying and the strike weighted by the risk-neutral exercise probability $\mathbb{Q}(F(T) \geq K)$ and the volatility premium weighted by F.

¹Here we use the term 'volatility' to refer to the total volatility, i.e. the instantaneous diffusion coefficient σ multiplied by $\sqrt{T-t}$, the square root of the time to maturity. The option time premium is implicitly included in this quantity.

4 Analyzing the vol premium

From our result in (3.5), it is implied that the value of European call options struck at the money is the discounted volatility premium weighted by F = K. We make this explicit below by writing

$$C(K, t) = D(t, T)K \int_{-\zeta'}^{-\zeta' + \sigma\sqrt{T - t}} \phi(z)dz = D(t, T)K \int_{-\zeta'}^{\zeta'} \phi(z)dz, \text{ where } -\zeta' = -\frac{1}{2}\sigma\sqrt{T - t}$$

However, since the greatest difference between the price of a European call option under the Black model and its intrinsic value is greatest at the money, one might be interested in how λ_{σ} changes as we vary the time t futures price F. Because we already showed that the value of at the money Black call is completely determined by F and λ_{σ} , our intuition tells us that a plot of λ_{σ} against F is most likely concave, with one maximum at F = K. Taking the partial derivative $\partial \lambda_{\sigma}/\partial F$, we thus have

$$\frac{\partial \lambda_{\sigma}}{\partial F} = \frac{\partial}{\partial F} \int_{-\zeta}^{-\zeta + \sigma\sqrt{\theta}} \phi(z) dz = \frac{\partial}{\partial F} \left[\Phi \left(-\zeta + \sigma\sqrt{\theta} \right) - \Phi(-\zeta) \right] = \phi \left(-\zeta + \sigma\sqrt{\theta} \right) \frac{1}{F\sigma\sqrt{\theta}} - \phi(-\zeta) \frac{1}{F\sigma\sqrt{\theta}} + \phi(-\zeta) \frac{1}{F\sigma\sqrt{\theta}$$

Factoring the expression above and rearranging terms, we simply have

$$\frac{\partial \lambda_{\sigma}}{\partial F} = \frac{1}{F\sigma\sqrt{\theta}} \left[\phi \left(-\zeta + \sigma\sqrt{\theta} \right) - \phi(-\zeta) \right] \tag{4.1}$$

We consider the cases F < K, F = K, and F > K to roughly understand how λ_{σ} changes as F changes.

F < K: In this case, assuming $\sigma \sqrt{\theta} > 0$, $\phi\left(-\zeta + \sigma\sqrt{\theta}\right) > \phi(-\zeta)$. We can thus conclude that

$$\frac{\partial \lambda_{\sigma}}{\partial F} \bigg|_{F < K} > 0$$

F=K: In this case, we see that $-\zeta=-\zeta'=-\frac{1}{2}\sigma\sqrt{\theta}$, so $-\zeta'+\sigma\sqrt{\theta}=\zeta'$. Due to the symmetry of $\phi(z)$ around 0, we have that $\phi(\zeta')-\phi(-\zeta')=0$, so we conclude that

$$\left. \frac{\partial \lambda_{\sigma}}{\partial F} \right|_{F=K} = 0$$

F > K: In this case, assuming $\sigma\sqrt{\theta} > 0$, $\phi\left(-\zeta + \sigma\sqrt{\theta}\right) < \phi(-\zeta)$. We can thus conclude that

$$\left. \frac{\partial \lambda_{\sigma}}{\partial F} \right|_{F>K} < 0$$

This confirms our intuition that λ_{σ} is greatest when F = K. Plots of λ_{σ} over F and $\partial \lambda_{\sigma}/\partial F$ over F in Figures 1 and 2 show more complex dynamics, but still confirm our results. Interestingly, because of the lognormal dynamics assumed in the Black model, the plots of λ_{σ} and $\partial \lambda_{\sigma}/\partial F$ over F are positively skewed.

5 Conclusion

In this article, we derived the Black formula by means of conditional expectation, which allowed us to reexpress the formula in terms of risk-neutral exercise probabilities and isolate the volatility premium embedded in an option's price. We were able to avoid constructing the risk-free portfolio and solving the resulting PDE as outlined in the original Black-Scholes (1973) and Black (1976) papers. The probabilistic approach may be preferred for its financial intuition over the PDE approach, which requires a change of variable into dimensionless groups, thus temporarily causing the equation to lose all financial meaning.

6 Figures

All figures were plotted in R.

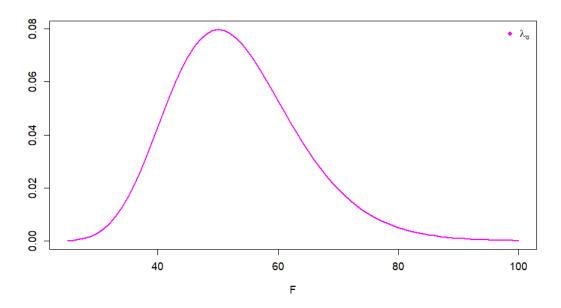


Figure 1: Plot of λ_{σ} over F with arbitrary parameters $\sigma=0.2,\,K=50,\,T-t=1.$

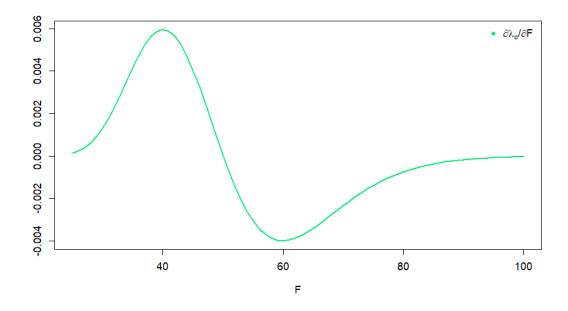


Figure 2: Plot of $\partial \lambda_{\sigma}/\partial F$ over F with arbitrary parameters $\sigma=0.2,\,K=50,\,T-t=1.$