

# European Option Pricing

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September 22, 2019

## Abstract

There is an explicit formula for pricing European options using cumulants.

An European option pays some function of the underlying instrument value at expiration. Let  $F$  be the (random) value of an instrument at option expiration and  $g$  be the payoff function. The forward value of the option is  $E[g(F)]$ . Any piecewise linear, continuous function can be approximated by a cash position, a forward contract and a linear combination of puts and calls.

## Puts and Calls

A put has payoff  $p(x) = \max\{k - x, 0\} = (k - x)^+$  and a call has payoff  $c(x) = (x - k)^+$ . Since  $c(x) - p(x) = x - k$  we can calculate the call value using the put value and a forward contract having the same strike.

If  $F$  is lognormal then the standard Black-Scholes/Merton theory can be used to value puts and calls. If  $F$  is a perturbation of a lognormal random variable there is an explicit formula for computing the option value using the cumulants of the perturbation.

Valuation formulas depend on the cumulative distribution function of  $F$ . The forward value of a put option is the expected value of the payoff

$$\begin{aligned} E[(k - F)^+] &= E[(k - F)1(k - F \geq 0)] \\ &= E[(k - F)1(F \leq k)] \\ &= kP(F \leq k) - E[F1(F \leq k)] \\ &= kP(F \leq k) - E[F]E[1(F \leq k)F/E[F]] \\ &= kP(F \leq k) - E[F]P^*(F \leq k) \end{aligned}$$

where  $P^*$  is the measure defined by  $dP^*/dP = F/E[F]$ . I.e.,  $E^*[X] = E[XF/E[F]]$ . Similarly, the value of a call is  $E[(F - k)^+] = E[F]P^*(F \geq k) - kP(F \geq k)$ .

*Exercise.* Derive this formula.

## Cumulant

The *cumulant* of a random variable  $X$  is  $\kappa^X(s) = \log E[\exp(sX)]$  and the *cumulants* are the coefficients in the power series expansion  $\kappa^X(s) = \kappa(s) = \sum_{n \geq 0} \kappa_n s^n / n!$ .

Note  $\kappa(0) = 0$ ,  $\kappa'(0) = \kappa_1 = E[X]$ , and  $\kappa''(0) = \kappa_2 = \text{Var } X$ .

*Exercise.* Prove this.

*Exercise.* If  $c$  is constant show  $\kappa_n^{cX} = c^n \kappa_n$ .

*Exercise.* If  $X$  and  $Y$  are independent show  $\kappa_n^{X+Y} = \kappa_n^X + \kappa_n^Y$ .

Define  $X = z(F) = (\kappa(s) + \log F/f)/s$ , so  $F = f \exp(sX - \kappa(s))$ . Note  $E[F] = f$  and  $\text{Var}(\log F) = s^2 \text{Var}(X)$ .

*Exercise.* Prove this.

The Black model has  $X$  standard normal and  $s = \sigma\sqrt{t}$ . In this case the cumulant of  $X$  is  $s^2/2$  so all cumulants are zero except the second  $\kappa_2 = 1$ .

*Exercise.* Use  $E \exp(N) = \exp(E[N] + \text{Var}(N)/2)$  for any normally distributed random variable  $N$  to prove this.

We can and do assume  $X$  has mean 0 and variance 1 as in the Black model.

*Exercise.* Show  $f \exp(s(\sigma X + \mu)) - \kappa^{\sigma X + \mu}(s) = f \exp(s^* X - \kappa^X(s^*))$  where  $s^* = \sigma s$ .

*Exercise.* Show if  $X$  is Poisson with parameter  $\lambda$ , i.e.,  $P(X = k) = e^{-\lambda} \lambda^k / k!$ ,  $k = 0, 1, \dots$ , then  $\kappa_n = \lambda$  for all  $n$ .

Note  $F/E[F] = e^{sX - \kappa(s)}$  and define  $P^*$  as above by  $dP^*/dP = e^{sX - \kappa(s)}$ .

*Exercise.* If  $X^*$  is the random variable defined by  $P(X^* \leq x) = P^*(X \leq x)$  show the cumulant of  $X^*$  is  $\kappa^*(u) = \kappa(u + s) - \kappa(s)$ .

*Exercise.* Show  $(\kappa^*)^{(n)}(0) = \kappa^n(s)$  for derivatives  $n > 0$ .

*Exercise.* Show the cumulants of  $X^*$  are  $\kappa_n^* = \sum_{k=0}^{\infty} \kappa_{n+k} s^k / k!$ .

In particular,  $E[X^*] = \kappa'(s)$  and  $\text{Var}(X^*) = \kappa''(s)$ .

## Bell polynomials

The (complete) Bell polynomials  $B_n(\kappa_1, \dots, \kappa_n)$  are defined by  $B_0 = 1$  and

$$B_{n+1}(\kappa_1, \dots, \kappa_{n+1}) = \sum_{k=0}^n \binom{n}{k} B_{n-k}(\kappa_1, \dots, \kappa_{n-k}) \kappa_{k+1}$$

for  $n > 0$ . They satisfy

$$e^{\sum_{n>0} \kappa_n s^n / n!} = \sum_{n \geq 0} B_n(\kappa_1, \dots, \kappa_n) s^n / n!$$

Differentiation both sides with respect to  $s$  and equating terms of equal power gives the recursive definition.

Note  $B_1(\kappa_1) = \kappa_1$  and  $B_2(\kappa_1, \kappa_2) = \kappa_1^2 + \kappa_2$ .

The *reduced Bell polynomials*,  $b_n(\kappa_1, \dots, \kappa_n) = B_n(\kappa_1, \dots, \kappa_n)/n!$  satisfy the recursion

$$b_{n+1}(\kappa_1, \dots, \kappa_{n+1}) = \frac{1}{n+1} \sum_{k=0}^n b_{n-k}(\kappa_1, \dots, \kappa_{n-k}) \kappa_{k+1} / k!$$

## Hermite polynomials

The (probabalists') Hermite polynomials are defined by  $H_0(x) = 1$ ,  $H_1(x) = x$ , and  $H_{n+1}(x) = xH_n(x) - nH_{n-1}(x)$  for  $n \geq 1$ . They satisfy

$$\frac{d^n}{dx^n} e^{-x^2/2}(x) = (-1)^n H_n(x) e^{-x^2/2}$$

## Fourier transform

The Fourier transform of  $\psi$  is  $\hat{\psi}(u) = E[\exp(-iuX)]$ . It is a general property of Fourier transforms that  $\hat{\psi}'(u) = iu\hat{\psi}(u)$ .

*Exercise.* Prove this.

Let  $\phi(x) = \exp(-x^2/2)/\sqrt{2\pi}$  be the standard normal density function and let  $\psi(x)$  be the probability density function of  $X$ . Note  $\hat{\phi}(u) = \phi(u)$ .

## Cumulative distribuiton

This section shows how to compute the cumulative distribution of a perturbation of a standard normal random variable.

$$\begin{aligned}
\hat{\psi}(u) &= E[\exp(-iuX)] \\
&= \exp(\kappa(-iu)) \\
&= \exp\left(\sum_{n>0} \kappa_n(-iu)^n/n!\right) \\
&= \exp(-u^2/2) \exp\left(\sum_{n\geq 3} \kappa_n(-iu)^n/n!\right) \\
&= \hat{\phi}(u) \exp\left(\sum_{n\geq 3} \kappa_n(-iu)^n/n!\right) \\
&= \hat{\phi}(u) \left(\sum_{n\geq 0} B_n(0, 0, \kappa_3, \dots, \kappa_n)(-iu)^n/n!\right) \\
&= \hat{\phi}(u) \left(1 + \sum_{n\geq 3} B_n(0, 0, \kappa_3, \dots, \kappa_n)(-iu)^n/n!\right) \\
&= \hat{\phi}(u) + \sum_{n\geq 3} B_n(\kappa_1, \dots, \kappa_n)(-1)^n \widehat{\phi^{(n)}}(u)/n!
\end{aligned}$$

Taking inverse Fourier transforms yields

$$\psi(x) = \phi(x) + \sum_{n\geq 3} B_n(0, 0, \kappa_3, \dots, \kappa_n)(-1)^n \phi^{(n)}(x)/n!.$$

Integrating gives

$$\begin{aligned}
\Psi(x) &= \Phi(x) + \sum_{n\geq 3} B_n(0, 0, \kappa_3, \dots, \kappa_n)(-1)^n \phi^{(n-1)}(x)/n! \\
&= \Phi(x) - \phi(x) \sum_{n\geq 3} B_n(0, 0, \kappa_3, \dots, \kappa_n) H_{n-1}(x)/n!
\end{aligned}$$

The value of a put is  $E[(k-F)^+] = k\Psi(z) - f\Psi^*(z)$  where  $z = (\kappa(s) + \log(f/F))/s$  and  $\Phi^*$  is the formula above with  $\kappa_n$  replaced with  $\kappa_n^*$ .

## Remarks

The Hermite polynomials can be expressed using Bell polynomials,  $H_n(x) = B_n(x, 1, 0, \dots, 0)$ .

The first seven Bell polynomials with  $k_1 = k_2 = 0$  are

$$\begin{aligned}
B_1(0) &= 0 \\
B_2(0, 0) &= 0 \\
B_3(0, 0, \kappa_3) &= \kappa_3 \\
B_4(0, 0, \kappa_3, \kappa_4) &= \kappa_4 \\
B_5(0, 0, \kappa_3, \kappa_4, \kappa_5) &= \kappa_5 \\
B_6(0, 0, \kappa_3, \kappa_4, \kappa_5, \kappa_6) &= 10\kappa_3^2 + \kappa_6 \\
B_7(0, 0, \kappa_3, \kappa_4, \kappa_5, \kappa_6, \kappa_7) &= 35\kappa_3\kappa_4 + \kappa_7
\end{aligned}$$

If  $f: \mathbf{R} \rightarrow \mathbf{R}$  has a piecewise continuous second derivative, then

$$f(x) = f(a) + f'(a)(x - a) + \int_{-\infty}^a (k - x)^+ f''(k) dk + \int_a^{\infty} (x - k)^+ f''(k) dk.$$

### Remarks

If the payoff function has jumps, digital options can be used to replicate it.