

5.5. Pricing Stock Options

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Normal
Distribution

Brownian
motion

Hitting
Time &
Maximum
Variable

Variations
on
Brownian
motion

Pricing
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Some Terminologies 套利

- ▶ In finance, **arbitrage** is the activity of buying shares or currency in one financial market and selling it at a profit in another.
 - ▶ **Options** 期权 are financial instruments that convey the right, but not the obligation, to engage in a future transaction on some underlying security. For example, buying a **call option** provides the right to buy a specified quantity of a security at a set strike price at some time (exercise date) on or before expiration, while buying a **put option** provides the right to sell. In this section, we discuss only call option.
- ▶ Option, costing c (the price of the option) per share, give us the option for purchasing shares of the stock at time t for the fixed price of K per share.

A Simple Example in Options Pricing

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- ▶ Consider one stock at two times: present time $t = 0$ and future time $t = 1$;
- ▶ Suppose that the price of the stock at present time is $X_0 = \$100$, and suppose we know that in the future $X_1 = \$200$ or $X_1 = \$50$ per share.
- ▶ Assume that the price of the option is c , which provides the right to buy a share of the stock at the fixed price \$150/per unit share at time $t = 1$.
- ▶ If the stock rises to \$200 then you would exercise the option at time 1 and realize a gain of $\$200 - \$150 = \$50$ for each option.
行使\$150/股的期权, 并以\$200/股卖出
- ▶ On the other hand, if the price of the stock is \$50, then the option is worthless at time 1.
放弃行使\$150/股的期权

Cont'd

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- ▶ At time 0, if we ^{portfolio} purchase x shares of stock at price \$100 and y shares of option at price c , then the original cost is $(x, y \text{ 可以为负, 表示卖出})$

$$\underbrace{100x}_{\text{卖(买)股票的钱}} + \underbrace{yc}_{\text{买(卖)期权的钱}}; \quad (x, y \text{ 符号通常相反: 可理解为对冲})$$

- ▶ and the value at time 1 is

$$200x + y(200 - 150) = 200x + 50y$$

if the price is \$200, or

$$50x$$

if the price is \$50.

- ▶ If we choose y such that

$$200x + 50y = 50x,$$

i.e. $y = -3x$ (Note that buy x shares \equiv sell $-x$ shares). Thus, with $y = -3x$, the value of our holding at time 1 is

$$\text{value} = 50x,$$

whatever price of the stock at time 1.

- Further, we gain on the transaction is

$$50x - (100x + yc) = 3cx - 50x = (3c - 50)x.$$

(可根据 c 选择 x 的正负)

- Thus, if $3c = 50$, then the gain is 0; otherwise, we can **guarantee a positive gain** (no matter what the price of the stock at time 1).

For example, if $c = 10$, we purchase $x = -2$ shares (i.e. sell 2 shares) of stock and purchase $y = -3 * (-2) = 6$ shares of option at time 0, then the gain is $(3 * 10 - 50) * (-2) = 40$;

if $c = 20$, we purchase $x = 2$ shares of stocks and purchase $y = -3 * 2 = -6$ shares (i.e. sell 6 shares) of option, then the gain is $(3 * 20 - 50) * 2 = 20$.

- A sure win betting scheme is called an **arbitrage**.

Thus, as we have just seen, the only option cost c that does not result in an arbitrage is $c = 50/3$.

The Black-Scholes option pricing formula

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The option pricing problem

- ▶ Price process of a stock Y_t , $0 \leq t \leq T$.
- ▶ We know the present price $Y_0 = y_0$.
- ▶ Note that for any future price Y_t , its present value is $e^{-rt} Y_t$, where r is the discount factor (fixed interest rate).
- ▶ Suppose c is the cost of an option to purchase one share at time T at the fixed price K (call).
pay at time 0
strike price
maturity date
- ▶ We want to determine value of c for which there is no betting strategy that leads to a sure win.

Pricing of a call option

- ▶ Stock price follows a Geometric BM:

$$Y_t = Y_0 e^{X_t} = Y_0 e^{mt + \sigma B_t}, \quad t \geq 0.$$

可以用 martingale 使结论更完备

- ▶ We assume the risk-neutral condition:

$$m + \frac{1}{2}\sigma^2 = r.$$

- ▶ Consider the wager of purchasing an option. At maturity time T ,

$$\begin{aligned} \text{the worth of option at time } T &= \begin{cases} Y_T - K, & \text{if } Y_T \geq K, \\ 0, & \text{if } Y_T < K. \end{cases} \\ &= (Y_T - K)^+. \end{aligned}$$

T 时刻 \$1 在 0 时刻的价值

- ▶ Hence, the present value of the worth of the option is $e^{-rT}(Y_T - K)^+$. In order for purchasing the option to have expected return 0, we must have that

$$E[e^{-rT}(Y_T - K)^+] = c. \quad (*)$$

The solution

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- To solve Equation (*), notice that $m + \frac{\sigma^2}{2} = r$, and

$$Y_T = Y_0 e^{mT + \sigma B_T},$$

$$mT + \sigma B_T \sim \mathcal{N}(mT, \sigma^2 T), \quad \text{with density f.}$$

$$\frac{1}{\sqrt{2\pi\sigma^2 T}} e^{-\frac{(y-mT)^2}{2\sigma^2 T}}, \quad y \in \mathbb{R}$$

$$= E[(Y_0 e^{X_T} - K)^+], \quad \text{其中 } X_T = mT + \sigma B_T \sim \mathcal{N}(mT, \sigma^2 T)$$

- we have

$$\begin{aligned} ce^{rT} &= E[(Y_T - K)^+] = \int_{-\infty}^{\infty} (y_0 e^y - K)^+ \frac{1}{\sqrt{2\pi\sigma^2 T}} e^{-\frac{(y-mT)^2}{2\sigma^2 T}} dy \\ &= y_0 e^{rT} \Phi(\sigma\sqrt{T} - a) - K\Phi(-a), \end{aligned}$$

where

[see details below]

$$a = \frac{1}{\sigma\sqrt{T}} \left(\log \frac{K}{y_0} - mT \right).$$

Note that $(y_0 e^y - K \geq 0 \Leftrightarrow y \geq \log \frac{K}{y_0})$, we have

$$\begin{aligned} ce^{rT} &= \int_{\log \frac{K}{y_0}}^{\infty} (y_0 e^y - K) \frac{1}{\sqrt{2\pi\sigma^2 T}} e^{-\frac{(y-mT)^2}{2\sigma^2 T}} dy \\ &\quad \left(\text{Let } w = \frac{y-mT}{\sigma\sqrt{T}}, y = mT + \sigma\sqrt{T}w, \text{ denote } a = \frac{\log \frac{K}{y_0} - mT}{\sigma\sqrt{T}} \right) \\ &= \int_a^{\infty} y_0 e^{mT + \sigma\sqrt{T}w} \frac{1}{\sqrt{2\pi}} e^{-\frac{w^2}{2}} dw - \int_a^{\infty} K \frac{1}{\sqrt{2\pi}} e^{-\frac{w^2}{2}} dw. \end{aligned}$$

The second integral equals $K\Phi(-a)$ and the first integral is

$$\begin{aligned} &y_0 e^{mT} \int_a^{\infty} e^{\frac{\sigma^2}{2} T} \frac{1}{\sqrt{2\pi}} e^{-\frac{(w-\sigma\sqrt{T})^2}{2}} dw \\ &= y_0 e^{(m+\frac{\sigma^2}{2})T} P\{N(\sigma\sqrt{T}, 1) > a\} = y_0 e^{(m+\frac{\sigma^2}{2})T} (1 - \Phi(a - \sigma\sqrt{T})) \\ &= y_0 e^{\alpha T} \Phi(\sigma\sqrt{T} - a). \end{aligned}$$

So

$$ce^{rT} = y_0 e^{(m+\frac{\sigma^2}{2})T} \Phi(\sigma\sqrt{T} - a) - K\Phi(-a).$$

The solution (cont'd)

Hence we have

$$c = \underbrace{y_0}_{\text{initial price}} \Phi(\sigma\sqrt{T} + b) - Ke^{-rT} \Phi(b), \quad (BS)$$

where

$$b = -a = \frac{mT - \log \frac{K}{y_0}}{\sigma\sqrt{T}} = \frac{rT - \frac{\sigma^2 T}{2} - \log \frac{K}{y_0}}{\sigma\sqrt{T}}.$$

若 $K=y_0, r=0$, 则 $c=y_0\Phi(\sigma\sqrt{T}+b)-y_0\Phi(b)$, $b=\frac{mT}{\sigma\sqrt{T}}=\frac{m}{\sigma}\sqrt{T}$

- ▶ Equation (BS) is known as the **Black-Scholes option cost valuation**.
- ▶ The value of c depends on the initial price of the stock y_0 , the option exercise time t , the option exercise price K , the discount factor r , and the value σ^2 , but not m . 因为 risk neutral condition, m 被表示为 σ 和 r 可以用 B-S equation 估计 σ^2 (implied volatility, can be computed in real time)
- ▶ If the option itself can be traded, then the formula of Equation (BS) can be used to set its price in such a way so that no arbitrage is possible.
- ▶ If at time s the price of the stock is $Y_s = x_s$, then the price of a (T, K) option ($s < T$)—that is, an option to purchase one unit of the stock at time t for a price K —should be set by replacing T by $T - s$ and y_0 by Y_s in Equation (BS).

Estimation the volatility parameter σ^2

Question: if we observe $\{Y_t\}$, a BM with drift coefficient m and variance parameter σ^2 , during the time interval $t \in [0, T]$, how can we estimate σ^2 ?

One solution:

- ▶ divide $(0, T]$ into N bins of binwidth h :

$$[0, T] = (0, h] + (h, 2h] + \cdots + ((N-1)h, Nh] .$$

- ▶ the N increments $W_i = Y_{ih} - Y_{(i-1)h}$, $1 \leq i \leq N$ are i.i.d. $\mathcal{N}(\underbrace{mh}, \underbrace{\sigma^2 h})$.

- ▶ One natural estimator of the variance $\sigma^2 h$ is the sample variance

$$S^2 = \frac{1}{N-1} \sum_{i=1}^N (W_i - \overline{W})^2, \quad \overline{W} = \frac{1}{N} \sum_{i=1}^N W_i ;$$

- ▶ Since $\frac{(N-1)S^2}{\sigma^2 h} \sim \chi_{N-1}^2$, $E\chi_k^2 = k$ and $\text{var}\chi_k^2 = 2k$, it follows that

$$E\left(\frac{S^2}{h}\right) = \sigma^2, \quad \text{var}\left(\frac{S^2}{h}\right) = \frac{2\sigma^4}{N-1} \rightarrow 0.$$

- ▶ So a “good” estimator of σ^2 is $\hat{\sigma}^2 = \frac{S^2}{h}$.