Chapter 7

Option Pricing

7.1 Discrete Time

In the next section we will discuss the Black–Scholes formula. To prepare for that, we will consider the much simpler problem of pricing options when there are a finite number of time periods and two possible outcomes at each stage. The restriction to two outcomes is not as bad as one might think. One justification for this is that we are looking at the process on a very slow time scale, so at most one interesting event happens (or not) per time period. We begin by considering a very simple special case.



Example 7.1 (Two-period binary tree). Suppose that a stock price starts at 100 at time 0. At time 1 (one day or one month or one year later) it will either be worth 120 or 90. If the stock is worth 120 at time 1, then it might be worth 140 or 115 at time 2. If the price is 90 at time 1, then the possibilities at time 2 are 120 and 80. Suppose now that you are offered a **European call option** with **strike price** 100 and **expiry** 2. This means you have an option to buy the stock (but not an obligation to do so) for 100 at time 2, i.e., after seeing

the outcome of the first and second stages. If the stock price is 80, you will not exercise the option to purchase the stock and your profit will be 0. In the other cases you will choose to buy the stock at 100 and then immediately sell it at X_2 to get a payoff of $X_2 - 100$ where X_2 is the stock price at time 2. Combining the two cases we can write the payoff in general as $(X_2 - 100)^+$, where $z^+ = \max\{z, 0\}$ denotes the positive part of z. Our problem is to figure out what is the right price for this option.

At first glance this may seem impossible since we have not assigned probabilities to the various events. However, it is a miracle of "pricing by the absence of arbitrage" that in this case we do not have to assign probabilities to the events to compute the price. To explain this we start by considering a small piece of the tree. When $X_1 = 90$, X_2 will be 120 ("up") or 80 ("down") for a profit of 30 or a loss of 10, respectively. If we pay c for the option, then when X_2 is up we make a profit of 20 - c, but when it is down we make -c. The last two sentences are summarized in the following table

$$\begin{array}{ccc} {\rm stock} & {\rm option} \\ {\rm up} & 30 & 20-c \\ {\rm down} & -10 & -c \end{array}$$

Suppose we buy x units of the stock and y units of the option, where negative numbers indicate that we sold instead of bought. One possible strategy is to choose x and y so that the outcome is the same if the stock goes up or down:

$$30x + (20 - c)y = -10x + (-c)y$$

Solving, we have 40x + 20y = 0 or y = -2x. Plugging this choice of y into the last equation shows that our profit will be (-10 + 2c)x. If c > 5, then we can make a large profit with no risk by buying large amounts of the stock and selling twice as many options. Of course, if c < 5, we can make a large profit by doing the reverse. Thus, in this case the only sensible price for the option is 5.

A scheme that makes money without any possibility of a loss is called an **arbitrage opportunity**. It is reasonable to think that these will not exist in financial markets (or at least be short-lived) since if and when they exist people take advantage of them and the opportunity goes away. Using our new terminology we can say that the only price for the option which is consistent with absence of arbitrage is c = 5, so that must be the price of the option (at time 1 when $X_1 = 90$).

Before we try to tackle the whole tree to figure out the price of the option at time 0, it is useful to look at things in a different way. Generalizing our example, let $a_{i,j}$ be the profit for the *i*th security when the *j*th outcome occurs.

Theorem 7.1. Exactly one of the following holds:

(i) There is a betting scheme $x = (x_1, x_2, \ldots, x_n)$ so that $\sum_{i=1}^m x_i a_{i,j} \ge 0$ for each j and $\sum_{i=1}^m x_i a_{i,k} > 0$ for some k.

(ii) There is a probability vector $p = (p_1, p_2, ..., p_n)$ with $p_j > 0$ so that $\sum_{i=1}^n a_{i,j} p_j = 0$ for all *i*.

Here a vector x satisfying (i) is an arbitrage opportunity. We never lose any money but for at least one outcome we gain a positive amount. Turning to (ii), the vector p is called a martingale measure since if the probability of the *j*th outcome is p_j , then the expected change in the price of the *i*th stock is equal to 0. Combining the two interpretations we can restate Theorem 1 as:

Theorem 7.2. There is no arbitrage if and only if there is a strictly positive probability vector so that all the stock prices are martingale.

Why is this true? One direction is easy. If (i) is true, then for any strictly positive probability vector $\sum_{i=1}^{m} \sum_{j=1}^{n} x_i a_{i,j} p_j > 0$, so (ii) is false.

Suppose now that (i) is false. The linear combinations $\sum_{i=1}^{m} x_i a_{i,j}$ when viewed as vectors indexed by j form a linear subspace of n-dimensional Euclidean space. Call it \mathcal{L} . If (i) is false, this subspace intersects the positive orthant $\mathcal{O} = \{y : y_j \ge 0 \text{ for all } j\}$ only at the origin. By linear algebra we know that \mathcal{L} can be extended to an n-1 dimensional subspace \mathcal{H} that only intersects \mathcal{O} at the origin.

Since \mathcal{H} has dimension n-1, it can be written as $\mathcal{H} = \{y : \sum_{j=1}^{n} y_j p_j = 0\}$. Since for each fixed *i* the vector $a_{i,j}$ is in $\mathcal{L} \subset \mathcal{H}$, (ii) holds. To see that all the $p_j > 0$ we leave it to the reader to check that if not, there would be a non-zero vector in \mathcal{O} that would be in \mathcal{H} .

To apply Theorem 1 to our simplified example we begin by noting that in this case $a_{i,j}$ is given by

$$j = 1 \quad j = 2$$

stock $i = 1 \quad 30 \quad -10$
option $i = 2 \quad 20 - c \quad -c$

By Theorem 2 if there is no arbitrage, then there must be an assignment of probabilities p_i so that

$$30p_1 - 10p_2 = 0 \qquad (20 - c)p_1 + (-c)p_2 = 0$$

From the first equation we conclude that $p_1 = 1/4$ and $p_2 = 3/4$. Rewriting the second we have

$$c = 20p_1 = 20 \cdot (1/4) = 5$$

To generalize from the last calculation to finish our example we note that the equation $30p_1 - 10p_2 = 0$ says that under p_j the stock price is a martingale (i.e., the average value of the change in price is 0), while $c = 20p_1 + 0p_2$ says that the price of the option is then the expected value under the martingale probabilities. Using these ideas we can quickly complete the computations in our example. When $X_1 = 120$ the two possible scenarios lead to a change of +20 or -5, so the relative probabilities of these two events should be 1/5 and 4/5. When $X_0 = 100$ the possible price changes on the first step are +20 and -10, so their relative probabilities are 1/3 and 2/3. Drawing a picture of the possibilities, we have



so the value of the option is

$$\frac{1}{15} \cdot 40 + \frac{4}{15} \cdot 15 + \frac{1}{6} \cdot 20 = \frac{80 + 120 + 100}{30} = 10$$

The last derivation may seem a little devious, so we will now give a second derivation of the price of the option. In the scenario described above, our investor has four possible actions:

 A_0 . Put \$1 in the bank and end up with \$1 in all possible scenarios.

 A_1 . Buy one share of stock at time 0 and sell it at time 1.

 A_2 . Buy one share at time 1 if the stock is at 120, and sell it at time 2.

 A_3 . Buy one share at time 1 if the stock is at 90, and sell it at time 2.

These actions produce the following payoffs in the indicated outcomes

time 1	time 2	A_0	A_1	A_2	A_3	option
120	140	1	20	20	0	40
120	115	1	20	-5	0	15
90	120	1	-10	0	30	20
90	80	1	-10	0	-10	0

Noting that the payoffs from the four actions are themselves vectors in fourdimensional space, it is natural to think that by using a linear combination of these actions we can reproduce the option exactly. To find the coefficients we write four equations in four unknowns,

$$z_0 + 20z_1 + 20z_2 = 40$$

$$z_0 + 20z_1 - 5z_2 = 15$$

$$z_0 - 10z_1 + 30z_3 = 20$$

$$z_0 - 10z_1 - 10z_3 = 0$$
(7.1)

Subtracting the second equation from the first and the fourth from the third gives $25z_2 = 25$ and $40z_3 = 20$ so $z_2 = 1$ and $z_3 = 1/2$. Pugging in these values, we have two equations in two unknowns:

$$z_0 + 20z_1 = 20$$
 $z_0 - 10z_1 = 5$

Taking differences, we conclude $30z_1 = 15$, so $z_1 = 1/2$ and $z_0 = 10$.

The reader may have already noticed that $z_0 = 10$ is the option price. This is no accident. What we have shown is that with \$10 cash we can buy and sell shares of stock to produce the outcome of the option in all cases. In the terminology of Wall Street, $z_1 = 1/2$, $z_2 = 1$, $z_3 = 1/2$ is a **hedging strategy** that allows us to **replicate the option**. Once we can do this it follows that the fair price must be \$10. To do this note that if we could sell it for \$12 then we can take \$10 of the cash to replicate the option and have a sure profit of \$2.

7.2 Continuous Time

To do option pricing in continuous time we need a model of the stock price, and for this we have to first explain **Brownian motion.** Let X_1, X_2, \ldots be independent and take the values 1 and -1 with probability 1/2 each. EX = 0and $EX^2 = 1$ so if we let $S_n = X_1 + \cdots + X_n$ then S_n/\sqrt{n} converges to χ a standard normal distribution. Intuitively, Brownian motion is what results when we look not only at time n but also at how the process got there. To be precise, we let $t \ge 0$ and consider $S_{[nt]}/\sqrt{n}$ where [nt] is the largest integer $\le nt$. In words we multiply n by t and then round down to the nearest whole number. When n=1000 the picture looks like:



Figure 7.1: Simulation of Brownian motion.

To understand the nature of the limit process we note that

$$\frac{S_{[nt]}}{\sqrt{n}} = \frac{S_{[nt]}}{\sqrt{[nt]}} \cdot \frac{\sqrt{[nt]}}{\sqrt{n}}$$

The first term approaches a standard normal distribution and the second \sqrt{t} so S_n/\sqrt{n} converges to $\sqrt{t}\chi$, a normal with mean 0 and variance t. Repeating the reasoning in the last paragraph we can see that if s < t then $(S_{[nt]} - S_{[ns]})/\sqrt{n}$ converges to a normal with mean zero and variance t - s. Noting that $(S_{[nt]} - S_{[ns]})$ is independent of $S_{[ns]}$ suggests the following definition of the limiting process which we call **Brownian motion**.

- B_t has a normal distribution with mean 0 and variance t
- If $0 < t_1 < \ldots < t_n$ then $B_{t_1}, B_{t_2} B_{t_1}, \ldots B_{t_n} B_{t_{n-1}}$ are independent

In modeling stock prices it is natural to assume that the daily percentage changes in the price are independent. For this reason and the mundane fact that stock prices must be > 0 we model the stock as what is called **geometric Brownian motion**.

$$X_t = X_0 \cdot \exp(\mu t + \sigma B_t) \tag{7.2}$$

 μ is the exponential growth rate of the stock, and σ is its volatility. In writing the model we have assumed that the growth rate and volatility of the stock are constant. If we also assume that the interest rate r is constant, then the discounted stock price is

$$e^{-rt}X_t = X_0 \cdot \exp((\mu - r)t + \sigma B_t)$$

Here we have to multiply by e^{-rt} , since \$1 at time t has the same value as e^{-rt} dollars today.

Our problem is to determine the fair price of a European call option $(X_t - K)^+$ with strike price K and expiry t. Extrapolating wildly from Theorem 2, we can say that any consistent set of prices must come from a martingale measure. This implies

$$\mu = r - \sigma^2 / 2 \tag{7.3}$$

To compute the value of the call option, we need to compute its value in the model in (7.2) for this special value of μ . Using the fact that $\log(X_t/X_0)$ has a normal($\mu t, \sigma^2 t$) distribution, one can show

Black–Scholes formula. The price of the European call option $(X_T - K)^+$ is given by

$$X_0 \Phi(\sigma \sqrt{t} - \alpha) - e^{-rt} K \Phi(-\alpha)$$

where Φ is the distribution function of a standard normal and

$$\alpha = \{\log(K/X_0 e^{\mu t})\} / \sigma \sqrt{t}$$

To try to come to grips with this ugly formula note that $K/X_0e^{\mu t}$ is the ratio of the strike price to the expected value of the stock at time t under the martingale probabilities, while $\sigma\sqrt{t}$ is the standard deviation of $\log(X_t/X_0)$.

Example 7.2 (Microsoft call options). The February 23, 1998, *Wall Street Journal* listed the following prices for July call options on Microsoft stock.

On this date Microsoft stock was trading at 815/8, while the annual interest rate was about 4% per year. Should you buy the call option with strike 80?

Solution. The answer to this question will depend on your opinion of the volatility of the market over the period. Suppose that we follow a traditional rule of thumb and decide that $\sigma = 0.3$; i.e., over a one-year period a stock's price might change by about 30% of its current value. In this case the drift rate for the martingale measure is

$$\mu = r - \sigma^2/2 = .04 - (.09)/2 = .04 - .045 = -.005$$

and so the log ratio is

$$\log(K/X_0 e^{\mu t}) = \log(80/(81.625e^{-.005(5/12)})) = \log(80/81.455) = -.018026$$

Five months corresponds to t = 5/12, so the standard deviation

$$\sigma\sqrt{t} = .3\sqrt{5/12} = .19364$$

and $\alpha = -.018026/.19364 = -.09309$. Plugging in now, we have a price of

$$81.625\Phi(.19365 + .09309) - e^{-.04(5/12)}80\Phi(.09309)$$

= $81.625\Phi(.28674) - 78.678\Phi(.09309)$
= $81.625(.6128) - 78.678(.5371) = 50.02 - 42.25 = 7.76$

This is somewhat lower than the price quoted in the paper. There are two reasons for this. First, the options listed in the *Wall Street Journal* are **American call options**. The holder has the right to exercise at any time during the life of the option. Since one can ignore the additional freedom to exercise early, American options are at least as valuable as their European counterparts. Second, and perhaps more importantly, we have not spent much effort on our estimate of r and σ . None the less as the next example shows the predictions of the formula are in rough agreement with the observed proces.

Example 7.3 (Intel call options). Again consulting the *Wall Street Journal* for February 23, 1998, we find the following prices listed for July call options on Intel stock, which was trading at $94 \ 3/16$.

strike	70	75	80	85	90	95	100	105
price	26	22	18	$14\frac{1}{2}$	$11\frac{3}{8}$	$8\frac{3}{4}$	$6\frac{1}{2}$	$4\frac{3}{8}$
formula	25.65	21.16	17.01	$13.\bar{5}9$	10.11	7.11	$5.\overline{3}9$	4.13

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