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# Stochastic Calculus and the Nobel Prize Winning Black-Scholes Equation

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The 1997 Nobel Prize in Economics went to Robert Merton and Myron Scholes for their revolutionary Black-Scholes differential equation for the value of financial instruments—termed a *stochastic* differential equation because it includes a random element.

To get started, let's toss a fair coin  $n$  times and let  $X_i = +1$  if the  $i^{\text{th}}$  toss is heads or  $-1$  if the  $i^{\text{th}}$  toss is tails, each with probability  $1/2$ . Each  $X_i$  is called a *random variable* because it takes on certain values with certain probabilities. It has mean 0, which is computed by taking each value multiplied by its probability and adding them up:

$$(1/2)(-1) + (1/2)(1) = 0.$$

It has variance 1, which is computed as the mean of the square of the difference from the mean, which in this case is always 1. The standard deviation, the square root of the variance, is also 1, and gives an estimate of how far  $X_i$  will deviate from its mean on average. The  $X_i$  are called independent because no toss is affected by another. In particular, the mean of  $X_i X_j$  is 0; if  $X_i$  is positive,  $X_j$  is equally likely to be positive or negative. The sum

$$f(n) = X_1 + \dots + X_n$$

has mean 0, variance  $n$ , and standard deviation  $\sigma = \sqrt{n}$ . To compute the variance, just note that the square

$$(X_1 + \dots + X_n)^2 = X_1^2 + \dots + X_n^2 + \text{cross terms}$$

has mean  $n$  because each  $X_i^2$  has mean 1 and the cross terms have mean 0. The fact that the mean of  $f(n)$  is 0 implies that if you toss a coin  $n$  times, the average number of heads is  $n/2$ . The fact that the standard deviation is  $\sqrt{n}$  means that in practice the deviation from  $n/2$  should be on the order of  $\sqrt{n}$ .

You could take a random walk on the line by tossing a coin every second and taking a unit step forward or backward according to whether the coin came up heads or tails. The random function  $f(n)$  would give your position after  $n$  seconds.

To get a continuous limit for  $0 \leq t \leq 1$  of random walks with rapid small steps, you could try considering

$$X_1 \Delta t + \dots + X_n \Delta t$$

with  $\Delta t = 1/n$  and mean 0, but this has standard deviation  $\sqrt{n}/n = 1/\sqrt{n}$ , which goes to 0 in the limit, and you end up just standing at the origin. The reason is that independent identical random variables with mean 0 tend to cancel when you add them up. In the stochastic calculus, this can be summarized by saying that

$$(1) \quad X_t dt = 0$$

because you always get 0 when you integrate or take limits of sums. If you replace  $\Delta t$  by some function  $a$  of  $\Delta t$  and consider

$$X_1 a + \dots + X_n a,$$

then the standard deviation is  $a\sqrt{n}$ , which is 1 if  $a = 1/\sqrt{n} = \sqrt{\Delta t}$ . Therefore

$$X_1 \sqrt{\Delta t} + \dots + X_n \sqrt{\Delta t}$$

does have a non-zero limit, with mean 0 and standard deviation 1 at time  $t = 1$ , called the *Wiener process* or *Brownian motion*  $z$ . This process is a solution to the stochastic differential equation

$$dz = X_t \sqrt{dt}.$$

At time  $t$ ,  $z(t)$  has mean 0 and standard deviation  $\sqrt{t}$ . More generally you can consider a *generalized Wiener process*  $x$  satisfying the differential equation

$$dx = a dt + b dz,$$

with solution  $x = at + bz$ . The mean, due to the deterministic component  $at$ , is  $at$ , while the standard deviation, due to the stochastic term  $bz$ , is  $b\sqrt{t}$ . Still more generally you can consider an *Ito process*

$$(2) \quad dx = a(x,t) dt + b(x,t) dz,$$

which can be hard to solve explicitly.

Stochastic calculus appears much trickier than ordinary calculus because  $dz^2$  is on the order of  $dt$  and hence it is not negligible the way that  $dt^2$  is. What makes it all manageable is *Ito's Lemma*, which in abbreviated form just says that

$$(3) \quad dz^2 = dt.$$

The essence of the proof is that

$$dz^2 = X_t^2 dt = 1 dt + (X_t^2 - 1) dt = dt .$$

The first equality is definition, the second is trivial. To understand the third, note that  $X_t^2 - 1$  is random variable with mean 0, so that  $(X_t^2 - 1) dt = 0$  as in (1). Here is the associated stochastic chain rule, also called Ito's Lemma:

**Ito's Lemma.** *Consider an Ito process (2) and let  $y = f(x)$  be a twice differentiable function of  $x$ . Then*

$$(4) \quad dy = (f'a + f''b^2/2) dt + f'b dz .$$

To see how (4) follows from (3), start with the second order Taylor series for  $y$ :

$$dy = f'(x) dx + \frac{1}{2}f''(x)dx^2 .$$

Note that the  $dx^2$  term is not negligible; indeed, by (2) and (3),  $dx^2 = b^2 dz^2 = b^2 dt$ . Equation (4) now follows from (2). The interesting feature is the appearance of the second derivative  $f''$  because  $dz^2 = dt$ .

Before applying stochastic calculus to stocks, recall how money grows in the bank at a risk-free rate  $r$ , which governs the relative growth rate of the balance  $B$ :

$$(5) \quad \frac{dB}{B} = r dt .$$

For the price  $S$  of a stock, in addition to a nonrandom growth rate  $\mu$ , one considers a random component, a multiple of Brownian motion:

$$(6) \quad \frac{dS}{S} = \mu dt + \sigma dz .$$

These two coefficients, the mean growth rate  $\mu$  and the so-called volatility  $\sigma$  are considered the two most important characteristics of a stock. In general, higher growth rate entails higher volatility and risk. Probably the most important principle from investment mathematics, called diversification, mandates buying many uncorrelated stocks with high  $\mu$  and  $\sigma$  with the expectation that their random fluctuations will tend to cancel and thus entail much less risk than any of them individually.

The hard part of investment analysis comes in treating more complicated financial instruments. A *call* is the right to buy for example 100 shares of Sears at \$95/share six months from now. The challenge facing Black, Scholes, and Merton was to figure out what such a call should be worth. The value  $C(S,t)$  of such a call varies in time and depends on how the price of the stock varies. Even though the current price of Sears is \$91, the call option is worth something, because the price may go above \$95. If the price stays at \$91, the value of the call will gradually decay over the six months to 0, but if the price rises, the value of the call may rise. It will never fall below 0, because it is just an *option* to buy, not an *obligation* to buy.

The key to evaluating the call is to note that it can be instantaneously replicated by some linear combination  $G = uS - vB$  of buying the stock and borrowing money from the bank. The call should have the same price as  $uS - vB$ . If, for example, the call had a higher price, one could go into the business of selling calls, buying replications, and making a risk-free profit. The opportunity for such “arbitrage” keeps market prices coherent. So the difficult problem of pricing the call seems to be reduced to the easier problem of pricing stocks.

The difficulty is that the coefficients  $u$  and  $v$  vary in time, depending in part on the price of the stock. Such “dynamic arbitrage” was a revolutionary idea. It means that instead of classical probability theory, you really need the random or stochastic calculus and differential equations we introduced above.

To replicate the call, the evolving linear combination  $G = uS - vB$  must satisfy certain conditions. First of all, you need to borrow more money to buy more stock, i.e., funds to increase  $u$  must come from corresponding increases in  $v$ , so that  $S du = B dv$ . Therefore,

$$dG = u dS - v dB + S du - B dv = u dS - v dB = (u\mu S - vrB) dt + u\sigma S dz$$

by (5) and (6). Meanwhile, by Ito’s Lemma (4),

$$dC = \left( \frac{\partial C}{\partial t} + \frac{\partial C}{\partial S} \mu S + \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 \right) dt + \frac{\partial C}{\partial S} \sigma S dz ,$$

with the extra term  $\partial C/\partial t$  because here  $C(S,t)$  also depends explicitly on  $t$ . For  $G$  to replicate  $C$ ,  $dG$  must equal  $dC$ . Equality of the  $dz$  terms means that  $u = \partial C/\partial S$ . Consequently,  $vB = uS - G = (\partial C/\partial S)S - C$ . Equality of the  $dt$  terms means that

$$\frac{\partial C}{\partial S} \mu S - \left( \frac{\partial C}{\partial S} S - C \right) r = \frac{\partial C}{\partial t} + \frac{\partial C}{\partial S} \mu S + \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 .$$

Canceling the  $\mu S$  terms yields the celebrated Black-Scholes differential equation for the value of the call option:

$$(7) \quad \frac{\partial C}{\partial t} + \frac{\partial C}{\partial S} r S + \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 = r C .$$

Here again the interesting feature is the appearance of the second derivative of  $C$ , multiplied by the volatility  $\sigma$ . By great good fortune, it happens that for  $r$  and  $\sigma$  constant, this differential equation has an exact, analytic solution, although the formula is a bit complicated (google “Black-Scholes” and see for yourself). It was discovered because it is essentially the same as the solution to the heat equation in physics. The main drawback is that the volatility  $\sigma$  is hard to estimate. For variable interest rates  $r$ , relatively easy to estimate by the prices of short- and long-term bonds, one can solve the differential equation numerically.

Merton’s landmark paper after Black and Scholes appeared in 1973. In 1994 Merton, Scholes, and others started a hedge fund, Long-Term Capital Management (LTCM), which was soon earning 40% a year. In 1997 Merton and Scholes won the Nobel Prize in Economics for their work (and Black received posthumous recognition). The very next year the LTCM fund crashed, losing \$4.6 billion. In an extraordinary move, the Federal Reserve intervened to rescue the fund and prevent international financial repercussions.

**Frank Morgan** works in minimal surfaces and has published six books, including two recent undergraduate texts on real analysis, *Calculus Lite*, and *The Math Chat Book*, based on his live, call-in TV show and column at MathChat.org. Inaugural winner of the MAA Haimo teaching award and founder of the NSF SMALL Undergraduate Research Project, he is Atwell Professor of Mathematics at Williams College. This article began as a talk at a special “Stochastic Fantastic Day,” which his chair Tom Garrity organized to give his colleagues a chance to explore a compelling but unfamiliar topic and enjoy dinner at his home afterwards. Morgan then tried it out on his Spring 2008 Investment Mathematics class, whom he thanks for many helpful comments.

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### **Brief Descriptive Summary**

The 1997 Nobel Prize in economics went to Robert Merton and Myron Scholes for their revolutionary Black-Scholes differential equation for the value of financial instruments. Unlike standard, deterministic differential equations, the Black-Scholes equation is a *stochastic* differential equation, including an element of randomness.

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