

Brownian Motion and Stochastic Differential Equations

Math 425

1 Brownian Motion

Mathematically Brownian motion, B_t $0 \leq t \leq T$, is a set of random variables, one for each value of the real variable t in the interval $[0, T]$. This collection has the following properties:

- B_t is continuous in the parameter t , with $B_0 = 0$.
- For each t , B_t is normally distributed with expected value 0 and variance t , and they are independent of each other.
- For each t and s the random variables $B_{t+s} - B_s$ and B_s are independent. Moreover $B_{t+s} - B_s$ has variance t .

The above states some properties, which Brownian motion must possess. However, just because we want something with certain properties does not guarantee that such a thing exists.

It can be shown that Brownian motion does indeed exist, and section 5.9 of *The Mathematics of Finance Modeling and Hedging* by Stampfli and Goodman indicates one way to construct a Brownian motion.

There is one important fact about Brownian motion, which is needed in order to understand why the process

$$S_t = e^{\sigma B_t} e^{(\mu - \sigma^2/2)t} \quad (1)$$

satisfies the stochastic differential equation

$$dS = \mu S dt + \sigma S dB. \quad (2)$$

The crucial fact about Brownian motion, which we need is

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

Equation (3) says two things. First $(dB)^2$ is deterministic, it is not random, and its magnitude is dt . That is, the amount of change in $(dB)^2$ caused by a change dt in the parameter is equal to dt .

To partially justify this statement we compute the expected value of $(B_{t+\Delta} - B_t)^2$.

$$\mathbb{E} \left[(B_{t+\Delta} - B_t)^2 \right] = \text{Var} [(B_{t+\Delta} - B_t)] \quad (4)$$

$$= \Delta. \quad (5)$$

2 Ito's Lemma

For a function $f(x, y)$ of the variables x and y it is not at all hard to justify that the equation below is correct to first order terms.

$$df = \frac{\partial f}{\partial x} dx + \frac{\partial f}{\partial y} dy. \quad (6)$$

However, what if we have a function f which depends not only on a real variable t , but also on a stochastic process such as Brownian motion. Suppose that $f = f(t, B_t)$, where B_t denotes Brownian motion. One is tempted to write as before that

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial B_t} dB_t. \quad (7)$$

However, in this case we would be badly mistaken. To see that this is so, we expand df using Taylor's formula; this time keep the terms involving the second derivatives of f

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial B_t} dB_t + \frac{1}{2} \frac{\partial^2 f}{\partial t^2} (dt)^2 + \frac{\partial^2 f}{\partial t \partial B_t} dt dB_t + \frac{1}{2} \frac{\partial^2 f}{\partial B_t^2} (dB_t)^2 + \text{higher order terms.} \quad (8)$$

We discard all terms involving dt to a power higher than 1. Note that the term $dt dB_t$ has magnitude $(dt)^{3/2}$. This leaves the following expression for df :

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial B_t} dB_t + \frac{1}{2} \frac{\partial^2 f}{\partial B_t^2} (dB_t)^2. \quad (9)$$

We next use the fact that $(dB_t)^2 = dt$ and write

$$df = \left(\frac{\partial f}{\partial t} + \frac{1}{2} \frac{\partial^2 f}{\partial B_t^2} \right) dt + \frac{\partial f}{\partial B_t} dB_t. \quad (10)$$

Equation (10) is called Ito's lemma, and gives us the correct expression for calculating differentials of composite functions which depend on Brownian processes.

3 Applications of Ito's Lemma

Let $f(B_t) = B_t^2$. Then Ito's lemma gives

$$d(B_t^2) = dt + 2B_t dB_t$$

This formula leads to the following integration formula

$$\begin{aligned} \int_{t_0}^t B_\tau dB_\tau &= \frac{1}{2} \left(\int_{t_0}^t d(B_\tau^2) - \int_{t_0}^t d\tau \right) \\ &= \frac{B_t^2 - B_{t_0}^2}{2} - \frac{t - t_0}{2}. \end{aligned}$$

Contrast this formula with the normal version

$$\int x dx = \frac{x^2}{2}.$$

Verify that $S_t = e^{\sigma B_t} e^{(\mu - \sigma^2/2)t}$ satisfies the stochastic differential equation

$$dS = \mu S dt + \sigma S dB_t. \quad (11)$$