

# Section 12 (Stat 171)

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- All the section materials (handouts & solutions) can be found either on Canvas or [here](#).

## 1 The Ito Integral

### 1.1 Definition

The Ito integral is defined as

$$\int_0^t X_s dB_s = \lim_{n \rightarrow \infty} \sum_{k=1}^n X_{t_{k-1}} (B_{t_k} - B_{t_{k-1}}), \quad (1)$$

for ever-finer partitions  $0 = t_0 < t_1 < \dots < t_{n-1} < t_n = t$ .

**Requirements:**

1.  $\int_0^t E(X_s^2) ds < \infty$

2.  $X_t$  does not depend on  $\{B_s : s > t\}$  but can depend on previous  $s \{B_s : s \leq t\}$  ( $X_t$  is adapted to brownian motion)
3.  $\lim_{n \rightarrow \infty} \sum_{k=1}^n X_{t_{k-1}}(B_{t_k} - B_{t_{k-1}})$  converges in a mean square sense to  $\int_0^t X_s dB_s$

## 1.2 Ito's Lemma

### Ito's Lemma

Ito's lemma gives us a term for  $dg(B_t)$  using the Taylor Expansion formula. Let  $g$  be a real valued, twice differentiable function. Then Ito's lemma is as follows:

$$g(B_t) - g(B_0) = \int_0^t g'(B_s)dB_s + \frac{1}{2} \int_0^t g''(B_s)ds$$

In shorthand differential form:

$$dg(B_t) = g'(B_t)dB_t + \frac{1}{2}g''(B_t)dt$$

### (Extension of) Ito's Lemma

Let  $g(t, x)$  be a real-valued function whose second-order partial derivatives are continuous. Then,

$$g(t, B_t) - g(0, B_0) = \int_0^t \left( \frac{\partial}{\partial t}g(s, B_s) + \frac{1}{2} + \frac{\partial^2}{\partial x^2}g(s, B_s) \right) ds + \int_0^t \frac{\partial}{\partial x}g(s, B_s)dB_s. \tag{2}$$

In shorthand differential form,

$$dg = \left( \frac{\partial g}{\partial t} + \frac{1}{2} \frac{\partial^2 g}{\partial x^2} \right) dt + \frac{\partial g}{\partial x} dB_t. \tag{3}$$

## 2 Stochastic Differential Equations

Recall the traditional differential equation:

- $\frac{dx}{dt} = f(t, x)$
- $X(0) = X_0$

We commonly see stochastic differential equations used when describing **diffusion**. The diffusion SDE is described as follows:

$$dX_t = \alpha(t, X_t)dt + \beta(t, X_t)dB_s$$

Where  $\alpha(t, X_t)$  is a drift coefficient and  $\beta(t, X_t)$  is the diffusion coefficient. We can rewrite this in integral form:

$$X_t - X_0 = \int_0^t \alpha(s, X_s)ds + \int_0^t \beta(s, X_s)ds$$

Notice this form is somewhat similar to Ito's Lemma. Let us introduce **Ito's Lemma for diffusion**:

Let  $g(t,x)$  be a real-valued function w/ continuous second-order derivatives. Let  $X_t$  be an Ito process such that:

$$dX_t = \alpha(t, X_t)dt + \beta(t, X_t)dB_t$$

As before. Then we have:

$$g(t, X_t) - g(0, X_0) = \int_0^t \left( \frac{\partial g}{\partial t} + \frac{\partial g}{\partial x} \alpha(s, X_s) + \frac{1}{2} \frac{\partial^2 g}{\partial x^2} \beta^2(s, X_s) \right) ds + \int_0^t \frac{\partial g}{\partial x} \beta(s, X_s) dB_s. \quad (4)$$

In shorthand differential form:

$$dg = \left( \frac{\partial g}{\partial t} + \frac{\partial g}{\partial x} \alpha(s, X_s) + \frac{1}{2} \frac{\partial^2 g}{\partial x^2} \beta^2(s, X_s) \right) dt + \frac{\partial g}{\partial x} \beta(t, X_t) dB_t.$$

## Problems

1. Use Ito's Lemma to evaluate  $d(B_t^4)$  and  $E(B_t^4)$ .

2. Use Ito's Lemma to show that  $E[B_t^k] = \frac{k(k-1)}{2} \int_0^t E[B_s^{k-2}] ds$ , for  $k \geq 2$ .

**3.** (a) Evaluate  $d(tB_t^2)$ .

(b) Derive a martingale that is a fourth-degree polynomial function of Brownian motion. Hint: use the result in (a) and Problem 1.

4. Show that if a function  $f(t, x)$  satisfies the partial differential equation

$$\frac{\partial f}{\partial t} + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} = 0$$

then  $f(t, B_t)$  is a martingale.