Stochastic Calculus, Courant Institute, Fall 2014

http://www.math.nyu.edu/faculty/goodman/teaching/StochCalc2014/index.html Always check the class message board before doing any work on the assignment.

Assignment 5, due end of semester

Corrections (check the class message board): (none yet.)

- 1. (A generalization of the Ito isometry formula) Here is a handy to calculate some things about Ito integrals
 - (a) Suppose f_t and g_t are non-anticipating functions, and the corresponding Ito integrals are

$$\begin{aligned} X_t &= \int_0^t f_s dW_s \\ Y_t &= \int_0^t g_s dW_s \;. \end{aligned}$$

Show that

$$\operatorname{cov}(X_t, Y_t) = \operatorname{E}[X_t Y_t] = \int_0^t \operatorname{E}[f_s g_s] \, ds \; .$$

Notice that if $g_t = f_t$, this becomes the formula from the Week 5 notes. The argument can be similar to the argument for the Ito isometry formula in the notes.

- (b) Suppose $f_t = t^2$ and $g_t = 1$. The notes for Week 5 show that $X_t \sim \mathcal{N}(0, t^5/5)$. Clearly $Y_t = W_t$. Compute the covariance of X_t and W_t using the result of part (a).
- (c) Since (X_t, W_t) is a bivariate normal whose variance/covariance structure you know, you can compute the conditional variance $\operatorname{var}(X_t|W_t)$. Use this result to show that X_t is not a function of W_t . This is an example of a general phenomenon, that the value of an Ito integral depends on the whole path $W_{[0,t]}$, not just the endpoint W_t .
- 2. (quadratic variation) Let X_t be a stochastic process of the form

$$X_t = \int_0^t f_s \, dW_s \;. \tag{1}$$

The quadratic variation is $[X]_t$ defined by

$$[X]_t = \lim_{m \to \infty} \sum_{t_j < t} \left(X_{t_{j+1}} - X_{t_j} \right)^2 \,. \tag{2}$$

This formula uses the definitions of the Week 5 notes. In particular, $\Delta t = 2^{-m}$ and $t_j = j\Delta t$. This exercise is a sequence of steps leading to the formula

$$[X]_t = \int_0^t f_s^2 ds \,. \tag{3}$$

This formula, like Ito's lemma next week, is a version of $(dW)^2 = dt$. The Ito integral (1) may be written informally as $dX_t = f_t dW_t$. Slightly more formally, if Δt is small, then $\Delta X = X_{t+\Delta t} - X_t \approx f_t \Delta W_t$. A term on the sum on the right in (2) is $(\Delta X_{t_j})^2 \approx f_{t_j}^2 (\Delta W_j)^2$. The informal derivation of (3) is

$$[X]_t = \int_0^t (dX_s)^2 = \int_0^t f_s^2 (dW_s)^2 = \int_0^t f_s^2 ds \; .$$

If you take the differential of both sides, you see that $(dX_t)^2 = f_s^2 (dW_t)^2 = f_s^2 dt$. The theorem related to this stuff is that: (i) the limit (2) exists, and (ii) it satisfies the integral formula (3). During this exercise, assume that for s > t,

$$\mathbf{E}\left[\left(f_{s}-f_{t}\right)^{2}\mid\mathcal{F}_{t}\right]\leq C(s-t)\;.$$
(4)

This hypothesis was used to show the convergence of Riemann sums to the Ito integral.

This exercise is a long sequence of steps of the kind in this week's material. Quadratic variation is important, but this exercise is here largely to demonstrate different ways to use the ideas. We assume that $\operatorname{E}[f_t^2] \leq F_2$ and $\operatorname{E}[f_t^4] \leq F_4$ for all t. Assume that f_t is a continuous function of t (unnecessary in the one place it is used).

- (a) Show that if X_t is not given by (1), but is a smooth (differentiable, or twice differentiable, say) function of t, then the definition (2) gives $[X]_t = 0.$
- (b) Prove the following variant of the Borel Cantelli lemma. Suppose A_m is a sequence of random variables. If

$$\sum_{m=1}^{\infty} \mathrm{E}[|A_m|] < \infty , \qquad (5)$$

then $A_m \to 0$ as $m \to \infty$. Then show that $A_m \to 0$ as $m \to \infty$ under the hypothesis that

$$\sum_{m=1}^{\infty} \mathbf{E} \left[A_m^2 \right] < \infty . \tag{6}$$

Hint for the second part: define $B_m = A_m^2$ and use the first part for B_m .

(c) The following notations simplify the calculations below. "Bin" number j is $B_j = [t_j, t_{j+1}]$. A term in the quadratic variation approximation sum is

$$\left(X_{t_{j+1}} - X_{t_j}\right)^2 = \left(\int_{B_j} f_t \, dW_t\right)^2 \, .$$

For $t \in B_j$ we make the approximation $f_t \approx f_{t_j}$. Show, in this notation, that

$$\int_{B_j} f_t \, dW_t = U_j + V_j \; ,$$

with

$$U_j = f_{t_j} \Delta W_j$$
, and $V_j = \int_{B_j} (f_t - f_{t_j}) dW_t$.

This easy step and the next just establish notation.

(d) Show that

$$\sum_{t_j < t} \left(X_{t_{j+1}} - X_{t_j} \right)^2 = S_1 + S_2 + S_3 ,$$

where

$$S_{1} = \sum_{t_{j} < t} f_{t_{j}}^{2} \Delta W_{j}^{2}$$

$$S_{2} = 2 \sum_{t_{j} < t} U_{j} V_{j}$$

$$S_{3} = \sum_{t_{j} < t} V_{j}^{2} .$$

$$(7)$$

(e) Show that $S_{2,m} \to 0$ as $m \to \infty$. We write $S_{2,m}$ instead of our usual S_2 only in formulas that involve a sum over m. Hint: First

$$\mathbf{E}[S_2] \le \sum_{t_j < t} \mathbf{E}[|U_j| |V_j|]$$

then Cauchy Schwarz, then $E[U_j] \leq F_2 \Delta t$, then

$$\mathbf{E}\big[V_j^2\big] \le C \int_0^{\Delta t} s \, ds \; .$$

The last step uses the Ito isometry formula and (4). You learn from this that $E[|U_j| |V_j|] \leq C\Delta t^{3/2}$.

(f) Show that $S_{3,m} \to 0$ as $m \to \infty$.

(g) Write $S_1 = S_4 + S_5$, where

$$S_4 = \sum_{t_j < t} f_{t_j}^2 \Delta t$$
$$S_5 = \sum_{t_j < t} f_{t_j}^2 \left(\Delta W_j^2 - \Delta t \right) \ .$$

Show that

$$S_4 \to \int_0^t f_s^2 \, ds \quad , \quad \text{ as } \quad m \to \infty \; .$$

Hint: this is a theorem of ordinary calculus.

- (h) Show that $S_{5,m} \to 0$ as $m \to \infty$. Hint: compute $\mathbb{E}[S_5^2]$. Show that the off diagonal terms vanish. Show that the diagonal terms are bounded by $2F_4\Delta t^2$.
- 3. Write a formula using differentials that expresses the statement that in a small increment of time dt, a process X_t has $E[dX | \mathcal{F}_t] = -\gamma X_t dt$, and $var(dX_t | \mathcal{F}_t) = \mu X_t dt$. Use dW_t to represent a random variable independent of \mathcal{F}_t that has mean zero and variance dt. Do not try to solve your stochastic differential equation.
- 4. (An Ito Leibnitz product formula) Suppose $X_t = f(W_t, t)$ and $Y_t = g(W_t, t)$. Show that

$$d(X_t Y_t) = (dX_t)Y_t + X_t dY_t + (dX_t dY_t) ,$$
(8)

where $(dX_t dY_t)$ is what you get by multiplying the Ito's lemma expressions for dX and dY, then keeping only the dW_t^2 part, then writing $dW_t^2 = dt$. Do this by applying Ito's lemma to the function h(w,t) = f(w,t)g(w,t)and $Z_t = h(W_t,t)$. Show that this works in the example $X_t = W_t^2$, $Y_t = W_t^3$. Compute your answer both directly as $d(W_t^5)$, and indirectly using (8).