Stochastic Calculus, Fall 2002 (http://www.math.nyu.edu/faculty/goodman/teaching/StochCalc/)

## Assignment 8.

Given November 7, due November 14 Last revised November 8.

Focus: Ito calculus

**1.** We will calculate the first four Picard iterates for the SDE  $dX_t = X_t dW_t$ , with  $X_0 = 1$ . These are given by  $X_t^{(0)} = 1$  for all  $t \ge 0$ , and

$$X_t^{(k+1)} = 1 + \int_0^t X_s^{(k)} dW_s \; .$$

This is partly to provide practice using the Ito calculus on the various stochastic integrals that come up.

- **a.** Compute  $X_t^{(1)}$ .
- **b.** Compute  $X_t^{(2)}$ . It will be simpler to compute

$$Y_t^{(2)} = X_t^{(2)} - X_t^{(1)} = \int_0^t (X_s^{(1)} - X_s^{(0)}) dW_s = \int_0^t Y_s^{(1)} dW_s \,.$$

- c. The exact solution of the SDE is  $X_t = e^{W_t}e^{-t/2}$ . Expand  $e^{W_t}$  and  $e^{-t/2}$  is Taylor series in  $W_t$  and t respectively, keeping terms up to and including O(t). Multiply these to get a short time approximation to  $X_t$  up to O(t). Show that this agrees with your answer to part b.
- **d.** Calculate  $Y_t^{(3)} = \int_0^t Y_s^{(2)} dW_s$ . To simplify the answer, you need a relationship between  $\int_0^t W_s ds$  and  $\int_0^t s dW_s$ .
- **e.** Calculate  $Y_t^{(4)} = \int_0^t Y_s^{(3)} dW_s$ . Combine with previous results to get  $X_t^{(4)}$ .
- **f.** Extend the calculation of part c to get an approximation containing all terms up to and including  $O(t^2)$ . Check that this agrees with your answer to part e.
- **2. a.** Using our bounds for  $X^{(k+1)} X^k$  from Lecture 7, paragraph 1.5 with the constant in (3), show that  $X_t X_t^{(k)} = O(t^{(k+1)/2})$ . For this you will need to know that

$$\sum_{j=k}^{\infty} z^{j} = \frac{z^{k}}{1-z} = O(z^{k}) \ , \ \text{as } z \to 0.$$

**b.** Justify the approximation  $X_{t+\Delta t} = X_t + \sigma(X_t)\Delta W_{\Delta t} + O(\Delta t)$ . Here,  $\Delta W_{\Delta t} = W_{t+\Delta t} - W_t$ . This uses part a and facts about Ito integrals. Hint: it might be easier notationally if you replace t by 0 and  $\Delta t$  by t.

**c.** Justify the approximation

$$X_{t+\Delta t} = X_t + \sigma(X_t, t) \Delta W_{\Delta t} + a(X_t, t) \Delta t + \frac{1}{2} \sigma(X, t) \partial_X \sigma(X_t, t) (\Delta W_{\Delta t}^2 - t) + O(t^{3/2})$$

This requires the approximation in part b and the technique behind it.

- **3.** Suppose  $dX_t = a(X_t, t)dt + \sigma(X_t, t)dW_t$  and  $f(x, t) = E_{x,t}[V(X_T)]$ . Use the fact that  $f_t = f(X_t, t)$  is a martingale and Ito's lemma for solutions of SDEs to find a partial differential equation  $\partial_t f + \cdots = 0$ . This is another in our collection of useful backward equations.
- 4. Derive the Black Scholes formula for European option pricing. For this, we need the cumulative distribution of a standard normal random variable. This is, for  $Z \sim \mathcal{N}(0, 1)$ ,

$$N(z) = P(Z \le z) = \frac{1}{\sqrt{2\pi}} \int_{y=-\infty}^{z} e^{-y^2/2} dy$$

Many integrals involving gaussians can be expressed in terms of N. For example, if  $X \sim \mathcal{N}(\mu, \sigma^2)$ , then  $P(X \leq \mu + z\sigma) = N(z)$ . In earlier assignments, we have used the fact that  $E[e^X]$  can be calculated by writing the integral, completing the square in the exponent, and recognizing this as another gaussian integral.

- **a.** A geometric Brownian motion satisfies the SDE  $dS_t = rSdt + \sigma S_t dW_t$ . Note that what we usually call  $\sigma$  is now  $\sigma S_t$ . Compute  $Y_t$ , the ordinary calculus guess at the solution (if the solution if  $W_t$  were a differentiable function of t). Use the Ito calculus to find and verify the correct solution in the form  $S_t = A(t)Y_t$ .
- **b.** Find an expression for  $S_T$  in terms of  $S_t$  and  $W_T W_t$ . This is just a line of algebra.
- c. For  $f(S,t) = E[V(S_T) | \mathcal{F}_t]$ , write the backward equation for f. This asks you to translate the general result in question 3 into a more specific equation that holds for geometric Brownian motion. The resulting equation is called the "Black Scholes" equation.
- **d.** For any final payout function V(s), write a formula for f(S,t) as the expected value of some function of a gaussian random variable whose mean and variance depend on t in a simple way. This is an application of the result of part b and the definition of f. It does not use the backward equation.
- e. For  $V(s) = \max(s K, 0) = (s K)_+$ , compute the integral in part d explicitly in terms of the N function. This is the "Black Scholes formula".
- f. Verify by direct differentiation that the formula satisfies the backward equation.
- 5. Use Ito's lemma to show that each of the following is a martingale. Comment on the difficulty of doing it this way or directly, as in an earlier assignment.

**a.** 
$$X_t = e^{k^2 t/2} \sin(kW_t)$$

**b.** 
$$Y_t = W_t^4 - 6 \int_0^t W_s^2 ds$$