

Derivative Securities, Courant Institute, Fall 2008

<http://www.math.nyu.edu/faculty/goodman/teaching/DerivSec08/index.html>

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Assignment 6, due October 22

Corrections: (Oct. 19: Problem 1 clarified – the U_k independent of each other, Problem 2 – assume $f(t)$ has finite variance for each t . Oct. 21: the hypotheses of Problem 2d clarified, the SDE in problem 4 corrected to put $S(t)$ in the noise term.)

This assignment is really Stochastic Calculus. Applications to finance are coming. It is very important to understand the two basics: (i) why dW^2 is important and in what sense $(dW)^2 = dt$, and (ii) the relation between partial differential equations (PDE) and stochastic differential equations (SDE).

(For Problem 1 and Problem 2) Suppose $f(t)$ is a random function of t – a Brownian motion or an Ornstein Uhlenbeck path or something like that. We say that $f(t)$ is *adapted* (or *non-anticipating*, which means almost the same thing) if the *future path*, $W(t+s) - W(t)$, for $s > 0$, is independent of all the values $f(s)$ for $s \leq t$. This is the same as saying that as much as $f(t)$ is influenced by values of W , it is influenced by values of $W(s)$ for $s \leq t$. For example,

$$f(t) = \int_0^t W(s) dW(s) = \frac{1}{2} (W(t)^2 - t) ,$$

where $f(t)$ depends only on $W(t)$, and $f(t) = X(t)$, the solution of $dX = -aXdt + \sigma dW$:

$$f(t) = e^{-at} X(0) + \sigma \int_0^t e^{-a(t-s)} dW(s) .$$

In the second example, $f(t)$ is not completely determined by W because it depends also on $X(0)$, which could be a different random variable. Also, $f(t)$ depends on the whole path $W(s)$ for $s \in [0, t]$, not just the value $W(t)$. In both cases, $f(t)$ has nothing to do with $W(s)$ for $s > t$.

1. Suppose F_k for $k \leq n$ are a family of random variables, and U_k are other random variables with U_k independent of F_1, \dots, F_k and independent of U_j for $j \neq k$. Suppose that $E[U_k] = 0$ and $E[U_k^2] = \sigma_k^2$. Let $A = \sum_{k=1}^n F_k U_k$.

Show that

$$E[A^2] = \sum_{k=1}^n \sigma_k^2 E[F_k^2] . \tag{1}$$

Relate this to the first formula on page 1 of Kohn's notes, Section 7, the *Ito isometry formula*.

2. (This is much like Kohn's discussion of discrete time rebalancing starting on page 3. It's an explanation of $dW^t = dt$.) Suppose $f(t)$ is a random adapted function. Consider the approximation

$$\int_0^T f(t) (dW(t))^2 \approx \sum_{t_k \leq T} f(t_k) (W(t_{k+1}) - W(t_k))^2,$$

where δt is a small time step and $t_k = k\delta t$. Mathematically, this means that the right side has a limit as $\delta t \rightarrow 0$ (with T fixed), and that limit is the definition of the left side. (Assume that $E[f(t)^2] < \infty$ for each t .)

- (a) Define

$$U_k = (W(t_{k+1}) - W(t_k))^2 - \delta t.$$

Calculate $E[U_k]$ and $E[U_k^2]$.

- (b) Define

$$A = \sum_{t_k \leq T} f(t_k) U_k,$$

so that

$$\sum_{t_k \leq T} f(t_k) (W(t_{k+1}) - W(t_k))^2 = \sum_{t_k \leq T} f(t_k) \delta t + A. \quad (2)$$

Explain why the hypotheses of Problem 1 apply to this A .

- (c) Use formula (1) to get a formula for $E[A^2]$.
 (d) Suppose that $E[f(t)] \leq M$ and $E[f(t)^2] \leq M$ for all $t \leq T$ and that $f(t)$ is a continuous function of t . Show that A is of the order of $\sqrt{\delta t}$, at least in the mean square sense, as $\delta t \rightarrow 0$.
 (e) Use this to argue that in the limit $\delta t \rightarrow 0$ we get

$$\int_0^T f(t) (dW(t))^2 = \int_0^T f(t) dt. \quad (3)$$

To summarize: The formula $(dW)^2 = dt$ is true only in the average sense that $E[(dW)^2] = dt$. However, the fluctuations are small enough that they cancel out in the sum (3).

- (f) (not action item) This exercise illustrates a feature of stochastic calculus. We prove things, such as $A \rightarrow 0$ as $\delta t \rightarrow 0$ by finding things we can calculate, here $E[A^2]$.

3. The Ornstein Uhlenbeck mean reverting process is

$$dX(t) = -aX(t)dt + \sigma dW(t). \quad (4)$$

- (a) Derive a formula for $X(T)$ in terms of $X(t)$ and the values of $W(s)$ for $t \leq s \leq T$. This formula will be similar to the one in the notes.

- (b) Use the formula from part (a) to evaluate

$$f(x, t) = E_{x,t} [X(T)^2] = E [X(T)^2 | X(t) = x] .$$

- (c) The general backward equation (not written explicitly in Kohn's notes) is

$$\partial_t f + \frac{1}{2} b^2(x, t) \partial_x^2 f + a(x, t) \partial_x f = 0 ,$$

where $f(x, t) = E_{x,t}[V(X(T))]$, $f(x, T) = V(x)$ (the *final conditions*, and $dX = a(X(t), t)dt + b(X(t), t)dW(t)$. Check that the result of part (b) satisfies the backward equation and the backward equation for the mean reverting process (4) and the correct final conditions.

- (d) Solve the backward equation for (4) that with final conditions $f(x, T) = x^4$. Hint: at each time, t , this is a polynomial of degree 4. You need only find the coefficients as functions of t . Check that this is true.
- (e) Evaluate $E[X(T)^4 | X(0) = 0]$ using the fact that $X(T)$ is a Gaussian with mean zero and known variance (in the notes) and the fact that if Y is normal with mean zero, then $E[Y^4] = 3\sigma_Y^4$, where $\sigma_Y^2 = \text{var}(Y)$.
4. A *volatility surface* is a function $\sigma(s, t)$. A volatility surface model of stock price movement is

$$dS(t) = \mu S(t)dt + \sigma(S(t), t) S(t) dW(t) .$$

- (a) Use the Δ -hedging argument of Black and Scholes outlined in Kohn's notes, Section 6 and 7 (not the binomial tree argument) to derive a PDE that prices European options for a volatility surface model.
- (b) Use the backward equation relationship to derive the risk neutral process (an SDE for $S(t)$) so that this price is $E_{s,t,RN}[V(S(T))]$.
- (c) The realized return variance for time T and time step δt is

$$R = \sum_{t_k \leq T} \left(\frac{S(t_{k+1}) - S(t_k)}{S(t_k)} \right)^2 .$$

Find a formula involving an integral for the limit of the realized variance as $\delta t \rightarrow 0$ in the stochastic volatility model.