

Exercises from
Arbitrage Theory in Continuous Time (3:rd ed)
Chapters 4-5

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1 Exercises

Exercise 1.1 Compute the stochastic differential dZ when

(a) $Z(t) = e^{\alpha t}$,

(b) $Z(t) = \int_0^t g(s)dW(s)$, where g is an adapted stochastic process.

(c) $Z(t) = e^{\alpha W(t)}$

(d) $Z(t) = e^{\alpha X(t)}$, where X has the stochastic differential

$$dX(t) = \mu dt + \sigma dW(t)$$

(μ and σ are constants).

(e) $Z(t) = X^2(t)$, where X has the stochastic differential

$$dX(t) = \alpha X(t)dt + \sigma X(t)dW(t).$$

Exercise 1.2 Compute the stochastic differential for Z when $Z(t) = \frac{1}{X(t)}$ and X has the stochastic differential

$$dX(t) = \alpha X(t)dt + \sigma X(t)dW(t).$$

By using the definition $Z = X^{-1}$ you can in fact express the right hand side of dZ entirely in terms of Z itself (rather than in terms of X). Thus Z satisfies a stochastic differential equation. Which one?

Exercise 1.3 Let $\sigma(t)$ be a given deterministic function of time and define the process X by

$$X(t) = \int_0^t \sigma(s) dW(s). \quad (1)$$

Show that the characteristic function of $X(t)$ (for a fixed t) is given by

$$E \left[e^{iuX(t)} \right] = \exp \left\{ -\frac{u^2}{2} \int_0^t \sigma^2(s) ds \right\}, \quad u \in R, \quad (2)$$

thus showing that $X(t)$ is normally distributed with zero mean and a variance given by

$$\text{Var}[X(t)] = \int_0^t \sigma^2(s) ds.$$

Exercise 1.4 Suppose that X has the stochastic differential

$$dX(t) = \alpha X(t) dt + \sigma(t) dW(t),$$

where α is a real number whereas $\sigma(t)$ is any stochastic process. Determine the function $m(t) = E[X(t)]$.

Exercise 1.5 Suppose that the process X has a stochastic differential

$$dX(t) = \mu(t) dt + \sigma(t) dW(t),$$

and that $\mu(t) \geq 0$ with probability one for all t . Show that this implies that X is a submartingale.

Exercise 1.6 A function $h(x_1, \dots, x_n)$ is said to be **harmonic** if it satisfies the condition

$$\sum_{i=1}^n \frac{\partial^2 h}{\partial x_i^2} = 0.$$

It is **subharmonic** if it satisfies the condition

$$\sum_{i=1}^n \frac{\partial^2 h}{\partial x_i^2} \geq 0.$$

Let W_1, \dots, W_n be independent standard Wiener processes, and define the process X by $X(t) = h(W_1(t), \dots, W_n(t))$. Show that X is a martingale (submartingale) if h is harmonic (subharmonic).

Exercise 1.7 The object of this exercise is to give an argument for the formal identity

$$dW_1 \cdot dW_2 = 0,$$

when W_1 and W_2 are independent Wiener processes. Let us therefore fix a time t , and divide the interval $[0, t]$ into equidistant points $0 = t_0 < t_1 < \dots < t_n = t$, where $t_i = \frac{i}{n} \cdot t$. We use the notation

$$\Delta W_i(t_k) = W_i(t_k) - W_i(t_{k-1}), \quad i = 1, 2.$$

Now define Q_n by

$$Q_n = \sum_{k=1}^n \Delta W_1(t_k) \cdot \Delta W_2(t_k).$$

Show that $Q_n \rightarrow 0$ in L^2 , i.e. show that

$$\begin{aligned} E[Q_n] &= 0, \\ \text{Var}[Q_n] &\rightarrow 0. \end{aligned}$$

Exercise 1.8 Let X and Y be given as the solutions to the following system of stochastic differential equations.

$$\begin{aligned} dX &= \alpha X dt - Y dW, & X(0) &= x_0, \\ dY &= \alpha Y dt + X dW, & Y(0) &= y_0. \end{aligned}$$

Note that the initial values x_0, y_0 are deterministic constants.

- (a) Prove that the process R defined by $R(t) = X^2(t) + Y^2(t)$ is deterministic.
- (b) Compute $E[X(t)]$.

Exercise 1.9 For a $n \times n$ matrix A , the **trace** of A is defined as

$$\text{tr}(A) = \sum_{i=1}^n A_{ii}.$$

- (a) If B is $n \times d$ and C is $d \times n$, then BC is $n \times n$. Show that

$$\text{tr}(BC) = \sum_{ij} B_{ij} C_{ji}.$$

- (b) With assumptions as above, show that

$$\text{tr}(BC) = \text{tr}(CB).$$

(c) Show that the multi dimensional Itô formula can be written

$$df = \left\{ \frac{\partial f}{\partial t} + \sum_{i=1}^n \mu_i \frac{\partial f}{\partial x_i} + \frac{1}{2} \text{tr} [\sigma^* H \sigma] \right\} dt + \sum_{i=1}^n \frac{\partial f}{\partial x_i} \sigma_i dW_i$$

where H denotes the Hessian matrix

$$H_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j}.$$

Exercise 1.10 Show that the scalar SDE

$$\begin{aligned} dX_t &= \alpha X_t dt + \sigma dW_t, \\ X_0 &= x_0, \end{aligned}$$

has the solution

$$X(t) = e^{\alpha t} \cdot x_0 + \sigma \int_0^t e^{\alpha(t-s)} dW_s, \quad (3)$$

by differentiating X as defined by eqn (3) and showing that X so defined actually satisfies the SDE.

Hint: Write eqn (3) as

$$X_t = Y_t + Z_t \cdot R_t,$$

where

$$\begin{aligned} Y_t &= e^{\alpha t} \cdot x_0, \\ Z_t &= e^{\alpha t} \cdot \sigma, \\ R_t &= \int_0^t e^{-\alpha s} dW_s, \end{aligned}$$

and first compute the differentials dZ , dY and dR . Then use the multidimensional Itô formula on the function $f(y, z, r) = y + z \cdot r$.

Exercise 1.11 Let A be an $n \times n$ matrix, and define the matrix exponential e^A by the series

$$e^A = \sum_{k=0}^{\infty} \frac{A^k}{k!}.$$

This series can be shown to converge uniformly.

(a) Show, by taking derivatives under the summation sign, that

$$\frac{de^{At}}{dt} = Ae^{At}.$$

(b) Show that

$$e^0 = I,$$

where 0 denotes the zero matrix, and I denotes the identity matrix.

(c) Convince yourself that if A and B commute, i.e. $AB = BA$, then

$$e^{A+B} = e^A \cdot e^B = e^B \cdot e^A.$$

Hint: Write the series expansion in detail.

(d) Show that e^A is invertible for every A , and that in fact

$$\left[e^A\right]^{-1} = e^{-A}.$$

(e) Show that for any A , t and s

$$e^{A(t+s)} = e^{At} \cdot e^{As}$$

(f) Show that

$$\left(e^A\right)^* = e^{A^*}$$

Exercise 1.12 Consider the n -dimensional linear SDE

$$\begin{cases} dX_t &= (AX_t + b_t) dt + \sigma_t dW_t, \\ X_0 &= x_0 \end{cases} \quad (4)$$

where A is an $n \times n$ matrix, b is an R^n -valued deterministic function (in column vector form), σ is a deterministic $n \times d$ deterministic matrix valued function, and W an d -dimensional Wiener process. Show that the solution of this equation is given by

$$X_t = e^{At} x_0 + \int_0^t e^{A(t-s)} b_s ds + \int_0^t e^{A(t-s)} \sigma_s dW_s. \quad (5)$$

Exercise 1.13 Consider again the linear SDE (4). Show that the expected value function

$$m(t) = E[X(t)]$$

, and the covariance matrix

$$C(t) = \{Cov(X_i(t), X_j(t))\}_{i,j}$$

are given by

$$\begin{aligned} m(t) &= e^{At} x_0 + \int_0^t e^{A(t-s)} b(s) ds, \\ C(t) &= \int_0^t e^{A(t-s)} \sigma(s) \sigma^*(s) e^{A^*(t-s)} ds, \end{aligned}$$

where $*$ denotes transpose.

Hint: Use the explicit solution above, and the fact that

$$C(t) = E [X_t X_t^*] - m(t)m^*(t).$$

Geometric Brownian motion (GBM) constitutes a class of processes which is closed under a number of nice operations. Here are some examples.

Exercise 1.14 Suppose that X satisfies the SDE

$$dX_t = \alpha X_t dt + \sigma X_t dW_t.$$

Now define Y by $Y_t = X_t^\beta$, where β is a real number. Then Y is also a GBM process. Compute dY and find out which SDE Y satisfies.

Exercise 1.15 Suppose that X satisfies the SDE

$$dX_t = \alpha X_t dt + \sigma X_t dW_t,$$

and Y satisfies

$$dY_t = \gamma Y_t dt + \delta Y_t dV_t,$$

where V is a Wiener process which is independent of W . Define Z by $Z = \frac{X}{Y}$ and derive an SDE for Z by computing dZ and substituting Z for $\frac{X}{Y}$ in the right hand side of dZ . If X is nominal income and Y describes inflation then Z describes real income.

Exercise 1.16 Suppose that X satisfies the SDE

$$dX_t = \alpha X_t dt + \sigma X_t dW_t,$$

and Y satisfies

$$dY_t = \gamma Y_t dt + \delta Y_t dW_t.$$

Note that now both X and Y are driven by the same Wiener process W . Define Z by $Z = \frac{X}{Y}$ and derive an SDE for Z .

Exercise 1.17 Suppose that X satisfies the SDE

$$dX_t = \alpha X_t dt + \sigma X_t dW_t,$$

and Y satisfies

$$dY_t = \gamma Y_t dt + \delta Y_t dV_t,$$

where V is a Wiener process which is independent of W . Define Z by $Z = X \cdot Y$ and derive an SDE for Z . If X describes the price process of, for example, IBM in US\$ and Y is the currency rate SEK/US\$ then Z describes the dynamics of the IBM stock expressed in SEK.

Exercise 1.18 Use a stochastic representation result in order to solve the following boundary value problem in the domain $[0, T] \times R$.

$$\begin{aligned}\frac{\partial F}{\partial t} + \mu x \frac{\partial F}{\partial x} + \frac{1}{2} \sigma^2 x^2 \frac{\partial^2 F}{\partial x^2} &= 0, \\ F(T, x) &= \ln(x^2).\end{aligned}$$

Here μ and σ are assumed to be known constants.

Exercise 1.19 Consider the following boundary value problem in the domain $[0, T] \times R$.

$$\begin{aligned}\frac{\partial F}{\partial t} + \mu(t, x) \frac{\partial F}{\partial x} + \frac{1}{2} \sigma^2(t, x) \frac{\partial^2 F}{\partial x^2} + k(t, x) &= 0, \\ F(T, x) &= \Phi(x).\end{aligned}$$

Here μ , σ , k and Φ are assumed to be known functions.

Prove that this problem has the stochastic representation formula

$$F(t, x) = E_{t,x} [\Phi(X_T)] + \int_t^T E_{t,x} [k(s, X_s)] ds,$$

where as usual X has the dynamics

$$\begin{aligned}dX_s &= \mu(s, X_s) ds + \sigma(s, X_s) dW_s, \\ X_t &= x.\end{aligned}$$

Hint: Define X as above, assume that F actually solves the PDE and consider the process $Z_s = F(s, X_s)$.

Exercise 1.20 Use the result of the previous exercise in order to solve

$$\begin{aligned}\frac{\partial F}{\partial t} + \frac{1}{2} x^2 \frac{\partial^2 F}{\partial x^2} + x &= 0, \\ F(T, x) &= \ln(x^2).\end{aligned}$$

Exercise 1.21 Consider the following boundary value problem in the domain $[0, T] \times R$.

$$\begin{aligned}\frac{\partial F}{\partial t} + \mu(t, x) \frac{\partial F}{\partial x} + \frac{1}{2} \sigma^2(t, x) \frac{\partial^2 F}{\partial x^2} + r(t, x) F &= 0, \\ F(T, x) &= \Phi(x).\end{aligned}$$

Here $\mu(t, x)$, $\sigma(t, x)$, $r(t, x)$ and $\Phi(x)$ are assumed to be known functions. Prove that this problem has a stochastic representation formula of the form

$$F(t, x) = E_{t,x} \left[\Phi(X_T) e^{\int_t^T r(s, X_s) ds} \right],$$

by considering the process $Z_s = F(s, X_s) \times \exp[\int_t^s r(u, X_u) du]$ on the time interval $[t, T]$.

Exercise 1.22 Solve the boundary value problem

$$\begin{aligned}\frac{\partial F}{\partial t}(t, x, y) + \frac{1}{2}\sigma^2 \frac{\partial^2 F}{\partial x^2}(t, x, y) + \frac{1}{2}\delta^2 \frac{\partial^2 F}{\partial y^2}(t, x, y) &= 0, \\ F(T, x, y) &= xy.\end{aligned}$$

Exercise 1.23 Go through the details in the derivation of the Kolmogorov forward equation.

Exercise 1.24 Consider the SDE

$$dX_t = \alpha dt + \sigma dW_t,$$

where α and σ are constants.

- (a) Compute the transition density $p(s, y; t, x)$, by solving the SDE.
- (b) Write down the Fokker-Planck equation for the transition density and check the equation is indeed satisfied by your answer in (a).

Exercise 1.25 Consider the standard GBM

$$dX_t = \alpha X_t dt + \sigma X_t dW_t$$

and use the representation

$$X_t = X_s \exp \left\{ \left[\alpha - \frac{1}{2}\sigma^2 \right] (t - s) + \sigma [W_t - W_s] \right\}$$

in order to derive the transition density $p(s, y; t, x)$ of GBM. Check that this density satisfies the Fokker-Planck equation.