
Appendix A: Malliavin Calculus on the Wiener Space

In this book we have, for several reasons, chosen to present the Malliavin calculus via chaos expansions. In the Brownian motion case this approach is basically equivalent to the construction given in the setting of the Hida white noise probability space $(\Omega, \mathcal{F}, \mathcal{P})$, where $\Omega = S'(\mathbb{R})$ is the Schwartz space of tempered distributions. In the Brownian case there is an alternative setting, namely the Wiener space $\Omega = C_0[0, T]$ of continuous functions $\omega : [0, T] \rightarrow \mathbb{R}$ with $\omega(0) = 0$. We now present this approach.

Malliavin calculus was originally introduced to study the regularity of the law of functionals of the Brownian motion, in particular, of the solution of stochastic differential equations driven by the Brownian noise [158].

Shortly, the idea is as follows. Let f be a smooth function on \mathbb{R}^d . The crucial idea for proving the regularity of the law of an \mathbb{R}^d -valued functional X of the Wiener process is to express the partial derivative of f at X as a derivative of the functional $f(X)$ with respect to a *new derivation* on the Wiener space. Based on some *integration by parts formula*, this derivation should exhibit the property of fulfilling the following relation:

$$E \left[\frac{\partial^{|\alpha|}}{\partial x^\alpha} f(X) \right] = E [f(X) L_\alpha(X)],$$

where $L_\alpha(X)$ is a functional of the Wiener process not depending on f and where $\frac{\partial^{|\alpha|}}{\partial x^\alpha}$ is the partial derivative of order $|\alpha| = \alpha_1 + \dots + \alpha_d$, $\alpha = (\alpha_1, \dots, \alpha_d)$. Provided $L_\alpha(X)$ is sufficiently integrable, the law of X should be smooth.

Hereafter we outline the classical presentation of the Malliavin derivative on the Wiener space. For further reading, we refer to, for example, [53, 169, 212].

A.1 Preliminary Basic Concepts

Let us first recall some basic concepts from classical analysis, see, for example, [79].

Definition A.1. Let U be an open subset of \mathbb{R}^n and let $f : U \rightarrow \mathbb{R}^m$.

(1) We say that f has a directional derivative at the point $x \in U$ in the direction $y \in \mathbb{R}^n$ if

$$D_y f(x) := \lim_{\varepsilon \rightarrow 0} \frac{f(x + \varepsilon y) - f(x)}{\varepsilon} = \frac{d}{d\varepsilon}[f(x + \varepsilon y)]|_{\varepsilon=0} \quad (\text{A.1})$$

exists. If this is the case we call the vector $D_y f(x) \in \mathbb{R}^m$ the directional derivative at x in the direction y . In particular, if we choose y to be the j th unit vector $e_j = (0, \dots, 1, \dots, 0)$, with 1 on j th place, we get

$$D_{e_j} f(x) = \frac{\partial f}{\partial x_j}(x),$$

the j th partial derivative of f .

(2) We say that f is differentiable at $x \in U$ if there exists a matrix $A \in \mathbb{R}^{m \times n}$ such that

$$\lim_{\substack{h \rightarrow 0 \\ h \in \mathbb{R}^n}} \frac{|f(x + h) - f(x) - Ah|}{|h|} = 0. \quad (\text{A.2})$$

If this is the case we call A the derivative of f at x and we write

$$A = f'(x).$$

Proposition A.2. The following relations between the two concepts hold true.

(1) If f is differentiable at $x \in U$, then f has a directional derivative in all directions $y \in \mathbb{R}^n$ and

$$D_y f(x) = f'(x)y = Ay. \quad (\text{A.3})$$

(2) Conversely, if f has a directional derivative at all $x \in U$ in all the directions $y = e_j$, $j = 1, \dots, n$, and all the partial derivatives

$$D_{e_j} f(x) = \frac{\partial f}{\partial x_j}(x)$$

are continuous functions of x , then f is differentiable at all $x \in U$ and

$$f'(x) = \left[\frac{\partial f_i}{\partial x_j}(x) \right]_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}} = A \in \mathbb{R}^{m \times n}, \quad (\text{A.4})$$

where f_i is component number i of f , that is, $f = (f_1, \dots, f_m)^T$.

We define similar operations in a more general context. First let us recall some basic concepts from functional analysis.

Definition A.3. Let X be a Banach space, that is, a complete, normed vector space over \mathbb{R} , and let $\|x\|$ denote the norm of the element $x \in X$. A linear functional on X is a linear map

$$T : X \rightarrow \mathbb{R}.$$

Recall that T is called linear if $T(ax + y) = aT(x) + T(y)$ for all $a \in \mathbb{R}$, $x, y \in X$. A linear functional T is called bounded (or continuous) if

$$\|T\| := \sup_{\|x\| \leq 1} |T(x)| < \infty$$

Sometimes we write $\langle T, x \rangle$ or Tx instead of $T(x)$ and call $\langle T, x \rangle$ “the action of T on x ”. The set of all bounded linear functionals is called the dual of X and is denoted by X^* . Equipped with the norm $\|\cdot\|$, the space X^* is a Banach space.

Example A.4. $X = \mathbb{R}^n$ with the Euclidean norm $|x| = \sqrt{x_1^2 + \dots + x_n^2}$ is a Banach space. In this case it is easy to see that we can identify X^* with \mathbb{R}^n .

Example A.5. Let $X = C_0([0, T])$ be the space of continuous real functions ω on $[0, T]$ such that $\omega(0) = 0$. Then

$$\|\omega\|_\infty := \sup_{t \in [0, T]} |\omega(t)|$$

is a norm on X called the *uniform norm*. With this norm, X is a Banach space and its dual X^* can be identified with the space $M([0, T])$ of all signed measures ν on $[0, T]$, with norm

$$\|\nu\| = \sup_{|f| \leq 1} \int_0^T f(t) d\nu(t) = |\nu|([0, T]).$$

Example A.6. Let $X = L^p([0, T]) = \{f : [0, T] \rightarrow \mathbb{R}; \int_0^T |f(t)|^p dt < \infty\}$ be equipped with the norm

$$\|f\|_p = \left[\int_0^T |f(t)|^p dt \right]^{1/p} \quad (1 \leq p < \infty).$$

Then X is a Banach space and its dual can be identified with $L^q([0, T])$, where

$$\frac{1}{p} + \frac{1}{q} = 1.$$

In particular, if $p = 2$, then $q = 2$, so $L^2([0, T])$ is its own dual.

We now extend the definitions of derivative and differentiability we had for \mathbb{R}^n to arbitrary Banach spaces.

Definition A.7. Let U be an open subset of a Banach space X and let f be a function from U into \mathbb{R}^m .

(1) We say that f has a directional derivative (or Gateaux derivative) $D_y f(x)$ at $x \in U$ in the direction $y \in X$ if

$$D_y f(x) := \frac{d}{d\varepsilon} [f(x + \varepsilon y)]_{\varepsilon=0} \in \mathbb{R}^m \quad (\text{A.5})$$

exists.

(2) We say that f is Fréchet-differentiable at $x \in U$, if there exists a bounded linear map

$$A : X \rightarrow \mathbb{R}^m,$$

that is, $A = (A_1, \dots, A_m)^T$, with $A_i \in X^*$ for $i = 1, \dots, m$, such that

$$\lim_{\substack{h \rightarrow 0 \\ h \in X}} \frac{|f(x + h) - f(x) - A(h)|}{\|h\|} = 0. \quad (\text{A.6})$$

We write

$$f'(x) = \begin{bmatrix} f'(x)_1 \\ \vdots \\ f'(x)_m \end{bmatrix} = A \in (X^*)^m. \quad (\text{A.7})$$

for the Fréchet derivative of f at x .

Similar to the Euclidean case (see Proposition A.2) we have the following result.

Proposition A.8.

(1) If f is Fréchet-differentiable at $x \in U \subset X$, then f has a directional derivative at x in all directions $y \in X$ and

$$D_y f(x) = \langle f'(x), y \rangle \in \mathbb{R}^m, \quad (\text{A.8})$$

where

$$\langle f'(x), y \rangle = (\langle f'(x)_1, y \rangle, \dots, \langle f'(x)_m, y \rangle)^T$$

is the m -vector whose i th component is the action of the i th component $f'(x)_i$ of $f'(x)$ on y .

(2) Conversely, if f has a directional derivative at all $x \in U$ in all directions $y \in X$ and the linear map

$$y \rightarrow D_y f(x), \quad y \in X$$

is continuous for all $x \in U$, then there exists an element $\nabla f(x) \in (X^*)^m$ such that

$$D_y f(x) = \langle \nabla f(x), y \rangle.$$

If this map $x \rightarrow \nabla f(x) \in (X^*)^m$ is continuous on U , then f is Fréchet differentiable and

$$f'(x) = \nabla f(x). \quad (\text{A.9})$$

A.2 Wiener Space, Cameron–Martin Space, and Stochastic Derivative

We now apply these operations to the Banach space $\Omega = C_0([0, T])$ considered in Example A.5 above. This space is called the *Wiener space*, because we can regard each path

$$t \rightarrow W(t, \omega)$$

of the Wiener process starting at 0 as an element ω of $C_0([0, T])$. Thus we may identify $W(t, \omega)$ with the value $\omega(t)$ at time t of an element $\omega \in C_0([0, T])$:

$$W(t, \omega) = \omega(t).$$

The space $\Omega = C_0([0, T])$ is naturally equipped with the Borel σ -algebra generated by the topology of the uniform norm. One can prove that this σ -algebra coincides with the σ -algebra generated by the cylinder sets (see, e.g., [36]). This measurable space is equipped with the probability measure P , which is given by the probability law of the Wiener process:

$$\begin{aligned} P\{W(t_1) \in F_1, \dots, W(t_k) \in F_k\} \\ = \int_{F_1 \times \dots \times F_k} \rho(t_1, x, x_1) \rho(t_2 - t_1, x, x_2) \cdots \rho(t_k - t_{k-1}, x_{k-1}, x_k) dx_1, \dots dx_k, \end{aligned}$$

where $F_i \subset \mathbb{R}$, $0 \leq t_1 < t_2 < \dots < t_k \leq T$, and

$$\rho(t, x, y) = (2\pi t)^{-1/2} \exp(-\frac{1}{2}|x - y|^2), \quad t \in [0, T], \quad x, y \in \mathbb{R}.$$

The measure P is called the *Wiener measure* on Ω .

Just as for Banach spaces, we now give the following definition.

Definition A.9. Let $F : \Omega \rightarrow \mathbb{R}$ be a random variable, choose $g \in L^2([0, T])$, and consider

$$\gamma(t) = \int_0^t g(s) ds \quad \in \Omega. \quad (\text{A.10})$$

Then we define the directional derivative of F at the point $\omega \in \Omega$ in direction $\gamma \in \Omega$ by

$$D_\gamma F(\omega) = \frac{d}{d\varepsilon} [F(\omega + \varepsilon\gamma)]|_{\varepsilon=0}, \quad (\text{A.11})$$

if the derivative exists in some sense (to be made precise later).

Note that we consider the derivative only in special directions, namely in the directions of elements γ of the form (A.10). The set of $\gamma \in \Omega$, which can be written on the form (A.10) for some $g \in L^2([0, T])$, is called the *Cameron–Martin space* and it is hereafter denoted by H . It turns out that it is difficult to obtain a tractable theory involving derivatives in all directions. However, the derivatives in the directions $\gamma \in H$ are sufficient for our purposes.

Definition A.10. Assume that $F : \Omega \rightarrow \mathbb{R}$ has a directional derivative in all directions γ of the form $\gamma \in H$ in the strong sense, that is,

$$\mathbf{D}_\gamma F(\omega) := \lim_{\varepsilon \rightarrow 0} \frac{F(\omega + \varepsilon\gamma) - F(\omega)}{\varepsilon} \quad (\text{A.12})$$

exists in $L^2(P)$. Assume in addition that there exists $\psi(t, \omega) \in L^2(P \times \lambda)$ such that

$$\mathbf{D}_\gamma F(\omega) = \int_0^T \psi(t, \omega) g(t) dt, \quad \text{for all } \gamma \in H. \quad (\text{A.13})$$

Then we say that F is differentiable and we set

$$\mathbf{D}_t F(\omega) := \psi(t, \omega). \quad (\text{A.14})$$

We call $\mathbf{D}_t F \in L^2(P \times \lambda)$ the stochastic derivative of F . The set of all differentiable random variables is denoted by $\mathcal{D}_{1,2}$.

Example A.11. Suppose $F = \int_0^T f(s) dW(s) = \int_0^T f(s) d\omega(s)$, where $f(s) \in L^2([0, T])$. Then if $\gamma \in H$, we have

$$\begin{aligned} F(\omega + \varepsilon\gamma) &= \int_0^T f(s)(d\omega(s) + \varepsilon d\gamma(s)) \\ &= \int_0^T f(s)d\omega(s) + \varepsilon \int_0^T f(s)g(s)ds, \end{aligned}$$

and hence

$$\frac{F(\omega + \varepsilon\gamma) - F(\omega)}{\varepsilon} = \int_0^T f(s)g(s)ds$$

for all $\varepsilon > 0$. Comparing with (A.13), we see that $F \in \mathcal{D}_{1,2}$ and

$$\mathbf{D}_t F(\omega) = f(t), \quad t \in [0, T], \omega \in \Omega. \quad (\text{A.15})$$

In particular, choosing

$$f(t) = \mathcal{X}_{[0, t_1]}(t)$$

we get

$$F = \int_0^T \mathcal{X}_{[0,t_1]}(s) dW(s) = W(t_1)$$

and hence

$$\mathbf{D}_t(W(t_1)) = \mathcal{X}_{[0,t_1]}(t). \quad (\text{A.16})$$

Let \mathbb{P} denote the family of all random variables $F : \Omega \rightarrow \mathbb{R}$ of the form

$$F = \varphi(\theta_1, \dots, \theta_n),$$

where $\varphi(x_1, \dots, x_n) = \sum_{\alpha} a_{\alpha} x^{\alpha}$, with $x^{\alpha} = x_1^{\alpha_1} \dots x_n^{\alpha_n}$ and $\alpha = (\alpha_1, \dots, \alpha_n)$,

is a polynomial and $\theta_i = \int_0^T f_i(t) dW(t)$ for some $f_i \in L^2([0, T])$, $i = 1, \dots, n$.

Such random variables are called *Wiener polynomials*. Note that \mathbb{P} is dense in $L^2(P)$.

Lemma A.12. Chain rule. *Let $F = \varphi(\theta_1, \dots, \theta_n) \in \mathbb{P}$. Then $F \in \mathcal{D}_{1,2}$ and*

$$\mathbf{D}_t F = \sum_{i=1}^n \frac{\partial \varphi}{\partial \theta_i}(\theta_1, \dots, \theta_n) \cdot f_i(t). \quad (\text{A.17})$$

Proof Let $\psi(t)$ denote the right-hand side of (A.17). Since

$$\sup_{s \in [0, T]} E[|W(s)|^N] < \infty \quad \text{for all } N \in \mathbb{N},$$

we see that

$$\begin{aligned} \frac{1}{\varepsilon} [F(\omega + \varepsilon \gamma) - F(\omega)] &= \frac{1}{\varepsilon} [\varphi(\theta_1 + \varepsilon(f_1, g), \dots, \theta_n + \varepsilon(f_n, g)) - \varphi(\theta_1, \dots, \theta_n)] \\ &\longrightarrow \sum_{i=1}^n \frac{\partial \varphi}{\partial \theta_i}(\theta_1, \dots, \theta_n) \cdot \mathbf{D}_{\gamma}(\theta_i), \quad \varepsilon \rightarrow 0, \end{aligned}$$

in $L^2(P)$. Hence F has a directional derivative in direction γ in the strong sense and by (A.15) we have

$$\mathbf{D}_{\gamma} F = \int_0^T \psi(t) g(t) dt.$$

By this we end the proof. \square

We now introduce the norm $\|\cdot\|_{1,2}$, on $\mathcal{D}_{1,2}$:

$$\|F\|_{1,2}^2 := \|F\|_{L^2(P)}^2 + \|\mathbf{D}_t F\|_{L^2(P \times \lambda)}^2, \quad F \in \mathcal{D}_{1,2}. \quad (\text{A.18})$$

Unfortunately, it is not clear if $\mathcal{D}_{1,2}$ is closed under this norm. To avoid this difficulty we work with the following family.

Definition A.13. We define $\mathbb{D}_{1,2}$ to be the closure of the family \mathbb{P} with respect to the norm $\|\cdot\|_{1,2}$.

Thus $\mathbb{D}_{1,2}$ consists of all $F \in L^2(P)$ such that there exists $F_n \in \mathbb{P}$ with the property that

$$F_n \longrightarrow F \quad \text{in } L^2(P) \quad \text{as } n \rightarrow \infty \quad (\text{A.19})$$

and

$$\{\mathbf{D}_t F_n\}_{n=1}^\infty \quad \text{is convergent in } L^2(P \times \lambda).$$

If this is the case, it is tempting to *define*

$$D_t F := \lim_{n \rightarrow \infty} \mathbf{D}_t F_n.$$

However, for this to work we need to know that this defines $D_t F$ *uniquely*. In other words, if there is another sequence $G_n \in \mathbb{P}$ such that

$$G_n \rightarrow F \quad \text{in } L^2(P) \text{ as } n \rightarrow \infty \quad (\text{A.20})$$

and

$$\{\mathbf{D}_t G_n\}_{n=1}^\infty \quad \text{is convergent in } L^2(P \times \lambda), \quad (\text{A.21})$$

does it follow that $\lim_{n \rightarrow \infty} \mathbf{D}_t F_n = \lim_{n \rightarrow \infty} \mathbf{D}_t G_n$?

By considering the difference $H_n = F_n - G_n$, we see that the answer to this question is positive, in view of the following theorem.

Theorem A.14. Closability of the derivative. *The operator \mathbf{D}_t is closable, that is, if the sequence $\{H_n\}_{n=1}^\infty \subset \mathbb{P}$ is such that*

$$H_n \rightarrow 0 \quad \text{in } L^2(P) \text{ as } n \rightarrow \infty \quad (\text{A.22})$$

and

$$\{\mathbf{D}_t H_n\}_{n=1}^\infty \quad \text{converges in } L^2(P \times \lambda) \text{ as } n \rightarrow \infty,$$

then

$$\lim_{n \rightarrow \infty} \mathbf{D}_t H_n = 0. \quad (\text{A.23})$$

The proof is based on the following useful result.

Lemma A.15. Integration by parts formula. *Suppose $F, \varphi \in \mathcal{D}_{1,2}$ and $\gamma \in H$ with $g \in L^2([0, T])$. Then*

$$E[\mathbf{D}_\gamma F \cdot \varphi] = E\left[F \cdot \varphi \cdot \int_0^T g(t)dW(t)\right] - E[F \cdot \mathbf{D}_\gamma \varphi]. \quad (\text{A.24})$$

Proof By the Cameron–Martin theorem (see, e.g., [159]) we have

$$\int_\Omega F(\omega + \varepsilon\gamma) \cdot \varphi(\omega) P(d\omega) = \int_\Omega F(\omega) \varphi(\omega - \varepsilon\gamma) Q(d\omega),$$

where

$$Q(d\omega) = \exp\left\{\varepsilon \int_0^T g(t)dW(t) - \frac{1}{2}\varepsilon^2 \int_0^T g^2(t)dt\right\} P(d\omega),$$

being $\omega(t) = W(t, \omega)$, $t \geq 0$, $\omega \in \Omega$, a Wiener process on the Wiener space $\Omega = C_0([0, T])$. This gives

$$\begin{aligned} E[\mathbf{D}_\gamma F \cdot \varphi] &= \int_{\Omega} \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} [F(\omega + \varepsilon\gamma) - F(\omega)] \cdot \varphi(\omega) P(d\omega) \\ &= \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_{\Omega} F(\omega + \varepsilon\gamma) \varphi(\omega) - F(\omega) \varphi(\omega) P(d\omega) \\ &= \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_{\Omega} F(\omega) \left[\varphi(\omega - \varepsilon\gamma) \exp \left\{ \varepsilon \int_0^T g(t) d\omega(t) \right. \right. \\ &\quad \left. \left. - \frac{1}{2} \varepsilon^2 \int_0^T g^2(t) dt \right\} - \varphi(\omega) \right] P(d\omega) \\ &= \int_{\Omega} F(\omega) \cdot \frac{d}{d\varepsilon} \left[\varphi(\omega - \varepsilon\gamma) \exp \left(\varepsilon \int_0^T g(t) d\omega(t) \right. \right. \\ &\quad \left. \left. - \frac{1}{2} \varepsilon^2 \int_0^T g^2(t) dt \right) \right]_{|\varepsilon=0} P(d\omega) \\ &= E \left[F \varphi \cdot \int_0^T g(t) dW(t) \right] - E[F \mathbf{D}_\gamma \varphi]. \end{aligned}$$

By this we end the proof. \square

Proof of Theorem A.14. By Lemma A.15 we get

$$E[\mathbf{D}_\gamma H_n \cdot \varphi] = E \left[H_n \varphi \cdot \int_0^T g dW \right] - E[H_n \cdot \mathbf{D}_\gamma \varphi] \longrightarrow 0, \quad n \rightarrow \infty,$$

for all $\varphi \in \mathbb{P}$. Since $\{\mathbf{D}_\gamma H_n\}_{n=1}^\infty$ converges in $L^2(P)$ and \mathbb{P} is dense in $L^2(P)$, we conclude that $\mathbf{D}_\gamma H_n \rightarrow 0$ in $L^2(P)$ as $n \rightarrow \infty$. Since this holds for all $\gamma \in H$, we obtain that $\mathbf{D}_t H_n \rightarrow 0$ in $L^2(P \times \lambda)$. \square

In view of Theorem A.14 and the discussion preceding it, we can now make the following unambiguous definition.

Definition A.16. Let $F \in \mathbb{D}_{1,2}$, so that there exists $\{F_n\}_{n=1}^\infty \subset \mathbb{P}$ such that

$$F_n \rightarrow F \quad \text{in } L^2(P)$$

and $\{\mathbf{D}_t F_n\}_{n=1}^\infty$ is convergent in $L^2(P \times \lambda)$. Then we define

$$D_t F = \lim_{n \rightarrow \infty} \mathbf{D}_t F_n \quad \text{in } L^2(P \times \lambda) \tag{A.25}$$

and

$$D_\gamma F = \int_0^T D_t F \cdot g(t) dt$$

for all $\gamma(t) = \int_0^t g(s) ds \in H$, with $g \in L^2([0, T])$. We call $D_t F$ the Malliavin derivative of F .

Remark A.17. Strictly speaking we now have two apparently different definitions of the derivative of F :

1. The stochastic derivative $\mathbf{D}_t F$ of $F \in \mathcal{D}_{1,2}$ given by Definition A.10.
2. The Malliavin derivative $D_t F$ of $F \in \mathbb{D}_{1,2}$ given by Definition A.16.

However, the next result shows that if $F \in \mathcal{D}_{1,2} \cap \mathbb{D}_{1,2}$, then the two derivatives coincide.

Lemma A.18. Let $F \in \mathcal{D}_{1,2} \cap \mathbb{D}_{1,2}$ and suppose that $\{F_n\}_{n=1}^\infty \subset \mathbb{P}$ has the properties

$$F_n \rightarrow F \quad \text{in } L^2(P) \quad \text{and} \quad \{\mathbf{D}_t F_n\}_{n=1}^\infty \quad \text{converges in } L^2(P \times \lambda). \quad (\text{A.26})$$

Then

$$\mathbf{D}_t F = \lim_{n \rightarrow \infty} \mathbf{D}_t F_n \quad \text{in } L^2(P \times \lambda). \quad (\text{A.27})$$

Hence

$$D_t F = \mathbf{D}_t F \quad \text{for } F \in \mathcal{D}_{1,2} \cap \mathbb{D}_{1,2}. \quad (\text{A.28})$$

Proof By (A.26) we get that $\{\mathbf{D}_\gamma F_n\}_{n=1}^\infty$ converges in $L^2(P)$ for each $\gamma(t) = \int_0^t g(s) ds$ with $g \in L^2([0, T])$. By Lemma A.15 and (A.26) we get

$$E[(\mathbf{D}_\gamma F_n - \mathbf{D}_\gamma F) \cdot \varphi] = E[(F_n - F) \cdot \varphi \cdot \int_0^t g dW] - E[(F_n - F) \cdot \mathbf{D}_\gamma \varphi] \rightarrow 0$$

for all $\varphi \in \mathbb{P}$. Hence $\mathbf{D}_\gamma F_n \rightarrow \mathbf{D}_\gamma F$ in $L^2(P)$ and (A.27) follows. \square

In view of Lemma A.18 we now use the same symbol $D_t F$ for the derivative and $D_\gamma F$ for the directional derivative of all the elements $F \in \mathcal{D}_{1,2} \cap \mathbb{D}_{1,2}$.

Remark A.19. Note that from the definition of $\mathbb{D}_{1,2}$ follows that, if $\{F_n\}_{n=1}^\infty \in \mathbb{D}_{1,2}$ with $F_n \rightarrow F$ in $L^2(P)$ and $\{D_t F_n\}_{n=1}^\infty$ converges in $L^2(P \times \lambda)$, then

$$F \in \mathbb{D}_{1,2} \quad \text{and} \quad D_t F = \lim_{n \rightarrow \infty} D_t F_n.$$

A.3 Malliavin Derivative via Chaos Expansions

Since an arbitrary $F \in L^2(P)$ can be represented by its chaos expansion

$$F = \sum_{n=0}^{\infty} I_n(f_n),$$

where $f_n \in \tilde{L}^2([0, T]^n)$ for all n , it is natural to ask if we can express the derivative of F (if it exists) by means of this. See Chap. 1 for the definition and properties of the Itô iterated integrals $I_n(f_n)$. Hereafter, we consider derivation according to Definition A.16 and Lemma A.18.

Let us first look at a special case.

Lemma A.20. *Suppose $F = I_n(f_n)$ for some $f_n \in \tilde{L}^2([0, T]^n)$. Then $F \in \mathbb{D}_{1,2}$ and*

$$D_t F = n I_{n-1}(f_n(\cdot, t)), \quad (\text{A.29})$$

where the notation $I_{n-1}(f_n(\cdot, t))$ means that the $(n-1)$ -iterated Ito integral is taken with respect to the $n-1$ first variables t_1, \dots, t_{n-1} of $f_n(t_1, \dots, t_{n-1}, t)$, that is, t is fixed and kept outside the integration.

Proof First consider the special case when

$$f_n = f^{\otimes n}$$

for some $f \in L^2([0, T])$, that is, when

$$f_n(t_1, \dots, t_n) = f(t_1) \dots f(t_n), \quad (t_1, \dots, t_n) \in [0, T]^n.$$

Then exploiting the definition and properties of Hermite polynomials h_n (see (1.15)), we have

$$I_n(f_n) = \|f\|^n h_n\left(\frac{\theta}{\|f\|}\right), \quad (\text{A.30})$$

where $\theta = \int_0^T f(t) dW(t)$. Moreover, by the chain rule (A.17) we have

$$D_t I_n(f_n) = \|f\|^n h'_n\left(\frac{\theta}{\|f\|}\right) \cdot \frac{f(t)}{\|f\|}.$$

Recall that a basic property of the Hermite polynomials is that

$$h'_n(x) = n h_{n-1}(x). \quad (\text{A.31})$$

This gives (A.29) in this case:

$$D_t I_n(f_n) = n \|f\|^{n-1} h_{n-1}\left(\frac{\theta}{\|f\|}\right) f(t) = n I_{n-1}(f^{\otimes(n-1)}) f(t) = n I_{n-1}(f_n(\cdot, t)).$$

Next, suppose f_n has the form

$$f_n = \xi_1^{\otimes \alpha_1} \widehat{\otimes} \xi_2^{\otimes \alpha_2} \widehat{\otimes} \cdots \widehat{\otimes} \xi_k^{\otimes \alpha_k}, \quad \alpha_1 + \cdots + \alpha_k = n, \quad (\text{A.32})$$

where $\widehat{\otimes}$ denotes symmetrized tensor product and $\{\xi_j\}_{j=1}^\infty$ is an orthonormal basis for $L^2([0, T])$. Then by an extension of (1.15) we have (see [120])

$$I_n(f_n) = h_{\alpha_1}(\theta_1) \cdots h_{\alpha_k}(\theta_k) \quad (\text{A.33})$$

with

$$\theta_j = \int_0^T \xi_j(t) dW(t)$$

and again (A.29) follows by the chain rule (A.17). Since any $f_n \in \tilde{L}^2([0, T]^n)$ can be approximated in $L^2([0, T]^n)$ by linear combinations of functions of the form given by (A.32), the general result follows. \square

Lemma A.21. *Let $\mathbb{P}_0 \subseteq \mathbb{P}$ denote the set of Wiener polynomials of the form*

$$p_k \left(\int_0^T \xi_1(t) dW(t), \dots, \int_0^T \xi_k(t) dW(t) \right),$$

where $p_k(x_1, \dots, x_k)$ is an arbitrary polynomial in k variables and $\{\xi_1, \xi_2, \dots\}$ is a given orthonormal basis for $L^2([0, T])$. Then \mathbb{P}_0 is dense in \mathbb{P} in the norm $\|\cdot\|_{1,2}$.

Proof Let $q := p \left(\int_0^T f_1(t) dW(t), \dots, \int_0^T f_k(t) dW(t) \right) \in \mathbb{P}$. We approximate q by

$$q^{(m)} := p \left(\sum_{j=0}^m (f_1, \xi_j)_{L^2([0, T])} \xi_j(t) dW(t), \dots, \sum_{j=0}^m (f_k, \xi_j)_{L^2([0, T])} \xi_j(t) dW(t) \right).$$

Then $q^{(m)} \rightarrow q$ in $L^2(P)$ and

$$D_t q^{(m)} = \sum_{i=1}^k \frac{\partial p}{\partial x_i} \cdot \sum_{j=1}^m (f_i, \xi_j)_{L^2([0, T])} \xi_j(t) \rightarrow \sum_{i=1}^k \frac{\partial p}{\partial x_i} \cdot f_i(t)$$

in $L^2(P \times \lambda)$ as $m \rightarrow \infty$. \square

Theorem A.22. Let $F = \sum_{n=0}^{\infty} I_n(f_n) \in L^2(P)$. Then $F \in \mathbb{D}_{1,2}$ if and only if

$$\sum_{n=1}^{\infty} n n! \|f_n\|_{L^2([0,T]^n)}^2 < \infty \quad (\text{A.34})$$

and if this is the case we have

$$D_t F = \sum_{n=1}^{\infty} n I_{n-1}(f_n(\cdot, t)). \quad (\text{A.35})$$

Proof Define $F_m = \sum_{n=0}^m I_n(f_n)$. Then $F_m \in \mathbb{D}_{1,2}$ and $F_m \rightarrow F$ in $L^2(P)$. Moreover, if $m > k$ we have

$$\begin{aligned} \|D_t F_m - D_t F_k\|_{L^2(P \times \lambda)}^2 &= \left\| \sum_{n=k+1}^m n I_{n-1}(f_n(\cdot, t)) \right\|_{L^2(P \times \lambda)}^2 \\ &= \int_0^T E \left[\left(\sum_{n=k+1}^m n I_{n-1}(f_n(\cdot, t)) \right)^2 \right] dt \\ &= \int_0^T \sum_{n=k+1}^m n^2(n-1)! \|f_n(\cdot, t)\|_{L^2([0,T]^{n-1})}^2 dt \\ &= \sum_{n=k+1}^m n n! \|f_n\|_{L^2([0,T]^n)}^2, \end{aligned} \quad (\text{A.36})$$

by Lemma A.20 and the orthogonality of the iterated Itô integrals (see (1.12)). Hence if (A.34) holds then $\{D_t F_n\}_{n=1}^{\infty}$ is convergent in $L^2(P \times \lambda)$ and hence $F \in \mathbb{D}_{1,2}$ and

$$D_t F = \lim_{m \rightarrow \infty} D_t F_m = \sum_{n=0}^{\infty} n I_{n-1}(f_n(\cdot, t)).$$

Conversely, if $F \in \mathbb{D}_{1,2}$ then, thanks to Lemma A.21, there exist polynomials $p_k(x_1, \dots, x_{n_k}) = \sum_{m_i: \sum m_i \leq k} a_{m_1, \dots, m_{n_k}} \prod_{i=1}^{n_k} h_{m_i}(x_i)$ of degree k for some $a_{m_1, \dots, m_{n_k}} \in \mathbb{R}$, such that if we put $F_k = p_k(\theta_1, \dots, \theta_{n_k})$ then $F_k \in \mathbb{P}$ and $F_k \rightarrow F$ in $L^2(P)$ and

$$D_t F_k \rightarrow D_t F \quad \text{in } L^2(P \times \lambda) \text{ as } k \rightarrow \infty.$$

By applying (A.33) we see that there exist $f_j^{(k)} \in \tilde{L}^2([0, T]^j)$, $1 \leq j \leq k$, such that

$$F_k = \sum_{j=0}^k I_j(f_j^{(k)}).$$

Since $F_k \rightarrow F$ in $L^2(P)$ we have

$$\sum_{j=0}^k j! \|f_j^{(k)} - f_j\|_{L^2([0,T]^j)}^2 \leq \|F_k - F\|_{L^2(P)}^2 \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

Therefore, $\|f_j^{(k)} - f_j\|_{L^2([0,T]^j)} \rightarrow 0$ as $k \rightarrow \infty$, for all j . This implies that

$$\|f_j^{(k)}\|_{L^2([0,T]^j)} \rightarrow \|f_j\|_{L^2([0,T]^j)} \quad \text{as } k \rightarrow \infty, \text{ for all } j. \quad (\text{A.37})$$

Similarly, since $D_t F_k \rightarrow D_t F$ in $L^2(P \times \lambda)$, we get by the Fatou lemma combined with the calculation, leading to (A.36) that

$$\begin{aligned} \sum_{j=0}^{\infty} j \cdot j! \|f_j\|_{L^2([0,T]^j)}^2 &= \sum_{j=0}^{\infty} \lim_{k \rightarrow \infty} (j \cdot j! \|f_j^{(k)}\|_{L^2([0,T]^j)}^2) \\ &\leq \liminf_{k \rightarrow \infty} \sum_{j=0}^{\infty} j \cdot j! \|f_j^{(k)}\|_{L^2([0,T]^j)}^2 \\ &= \lim_{k \rightarrow \infty} \|D_t F_k\|_{L^2(P \times \lambda)}^2 \\ &= \|D_t F\|_{L^2(P \times \lambda)}^2 < \infty, \end{aligned}$$

where we have put $f_j^{(k)} = 0$ for $j > k$. Hence (A.34) holds and the proof is complete. \square

Solutions

In this chapter we present a solution to the exercises marked with (*) in the book. The level of the exposition varies from fully detailed to just sketched.

Problems of Chap. 1

1.1 Solution

(a) Consider the following equalities:

$$\begin{aligned}\exp\left\{tx - \frac{t^2}{2}\right\} &= \exp\left\{\frac{1}{2}x^2\right\} \exp\left\{-\frac{1}{2}(x-t)^2\right\} \\ &= \exp\left\{\frac{1}{2}x^2\right\} \sum_{n=0}^{\infty} \frac{t^n}{n!} \frac{d^n}{dt^n} \exp\left\{-\frac{1}{2}(x-t)^2\right\}|_{t=0} \\ &= \exp\left\{\frac{1}{2}x^2\right\} \sum_{n=0}^{\infty} \left\{ \frac{(-1)^n t^n}{n!} \frac{d^n}{du^n} \exp\left\{-\frac{1}{2}u^2\right\} \right\}|_{u=x} \\ &= \exp\left\{\frac{1}{2}x^2\right\} \sum_{n=0}^{\infty} \frac{(-1)^n t^n}{n!} \frac{d^n}{dx^n} \exp\left\{-\frac{1}{2}x^2\right\} \\ &= \sum_{n=0}^{\infty} \frac{t^n}{n!} h_n(x),\end{aligned}$$

with the substitution $u = x - t$.

(b) Set $u = t\sqrt{\lambda}$. Using the result in (a), we get

$$\begin{aligned}\exp\left\{tx - \frac{t^2\lambda}{2}\right\} &= \exp\left\{u\frac{x}{\sqrt{\lambda}} - \frac{u^2}{2}\right\} \\ &= \sum_{n=0}^{\infty} \frac{u^n}{n!} h_n\left(\frac{x}{\sqrt{\lambda}}\right) \\ &= \sum_{n=0}^{\infty} \frac{t^n \lambda^{n/2}}{n!} h_n\left(\frac{x}{\sqrt{\lambda}}\right).\end{aligned}$$

(c) If we choose $x = \theta$, $\lambda = \|g\|^2$, and $t = 1$ in (b), we get

$$\exp \left\{ \int_0^T g dW - \frac{1}{2} \|g\|^2 \right\} = \sum_{n=0}^{\infty} \frac{\|g\|^n}{n!} h_n \left(\frac{\theta}{\|g\|} \right).$$

(d) In particular, if we choose $g(s) = \chi_{[0,t]}(s)$, $s \in [0, T]$, we get

$$\exp \left\{ W(t) - \frac{1}{2} t \right\} = \sum_{n=0}^{\infty} \frac{t^{n/2}}{n!} h_n \left(\frac{W(t)}{\sqrt{t}} \right). \quad \square$$

1.3 Solution

(a) $\xi = W(t) = \int_0^T \chi_{[0,t]}(s) dW(s)$, so $f_0 = 0$, $f_1 = \chi_{[0,t]}$, and $f_n = 0$ for $n \geq 2$.

(b) $\xi = \int_0^T g(s) dW(s)$, so $f_0 = 0$, $f_1 = g$, and $f_n = 0$ for $n \geq 2$.

(c) Since

$$\int_0^t \int_0^{t_2} 1 dW(t_1) dW(t_2) = \int_0^t W(t_2) dW(t_2) = \frac{1}{2} W^2(t) - \frac{1}{2} t,$$

we get that

$$\begin{aligned} W^2(t) &= t + 2 \int_0^t \int_0^{t_2} 1 dW(t_1) dW(t_2) \\ &= t + 2 \int_0^T \int_0^{t_2} \chi_{[0,t]}(t_1) \chi_{[0,t]}(t_2) dW(t_1) dW(t_2) = t + I_2[f_2]. \end{aligned}$$

Thus $f_0 = t$,

$$f_2(t_1, t_2) = \chi_{[0,t]}(t_1) \chi_{[0,t]}(t_2) =: \chi_{[0,t]}^{\otimes 2},$$

and $f_n = 0$ for $n \neq 2$.

(d) By Problem 1.1 (c) and (1.15), we have

$$\begin{aligned} \xi &= \exp \left\{ \int_0^T g(s) dW(s) \right\} \\ &= \exp \left\{ \frac{1}{2} \|g\|^2 \right\} \sum_{n=0}^{\infty} \frac{\|g\|^n}{n!} h_n \left(\frac{\theta}{\|g\|} \right) \\ &= \exp \left\{ \frac{1}{2} \|g\|^2 \right\} \sum_{n=0}^{\infty} J_n[g^{\otimes n}] = \sum_{n=0}^{\infty} \frac{1}{n!} \exp \left\{ \frac{1}{2} \|g\|^2 \right\} I_n[g^{\otimes n}]. \end{aligned}$$

Hence

$$f_n = \frac{1}{n!} \exp\left(\frac{1}{2}\|g\|^2\right) g^{\otimes n}, \quad n = 0, 1, 2, \dots,$$

where

$$g^{\otimes n}(x_1, \dots, x_n) := g(x_1)g(x_2)\cdots g(x_n).$$

(e) We have the following equalities:

$$\begin{aligned} \xi &= \int_0^T g(s)W(s) ds \\ &= \int_0^T g(s) \int_0^s 1 dW(t) ds \\ &= \int_0^T \int_t^T g(s) ds dW(t) \\ &= I_1(f_1), \end{aligned}$$

where $f_1(t) := \int_t^T g(s) ds$, $t \in [0, T]$. \square

1.4 Solution

(a) Since $\int_0^T W(t)dW(t) = \frac{1}{2}W^2(T) - \frac{1}{2}T$, we have

$$F = W^2(T) = T + 2 \int_0^T W(t)dW(t).$$

Hence $E[F] = T$ and $\varphi(t) = 2W(t)$, $t \in [0, T]$.

(b) Define $M(t) = \exp\{W(t) - \frac{1}{2}t\}$, $t \in [0, T]$. Then by the Itô formula

$$dM(t) = M(t)dW(t)$$

and therefore

$$M(T) = 1 + \int_0^T M(t)dW(t).$$

Moreover,

$$F = \exp\{W(T)\} = \exp\left\{\frac{T}{2}\right\} + \exp\left\{\frac{T}{2}\right\} \int_0^T \exp\left\{W(t) - \frac{1}{2}t\right\} dW(t).$$

Hence

$$E[F] = \exp\left\{\frac{T}{2}\right\} \quad \text{and} \quad \varphi(t) = \exp\left\{W(t) + \frac{T-t}{2}\right\}, \quad t \in [0, T].$$

(c) Integration by parts (application of the Itô formula) gives

$$F = \int_0^T W(t)dt = TW(T) - \int_0^T tdW(t) = \int_0^T (T-t)dW(t).$$

Hence, $E[F] = 0$ and $\varphi(t) = T - t$, $t \in [0, T]$.

(d) By the Itô formula

$$dW^3(t) = 3W^2(t)dW(t) + 3W(t)dt.$$

Hence

$$F = W^3(T) = 3 \int_0^T W^2(t)dW(t) + 3 \int_0^T W(t)dt.$$

Therefore, by (c) we get

$$E[F] = 0 \quad \text{and} \quad \varphi(t) = 3W^2(t) + 3T(1-t), \quad t \in [0, T].$$

(e) Put $X(t) = e^{\frac{1}{2}t}$, $Y(t) = \cos W(t)$, $N(t) = X(t)Y(t)$, $t \in [0, T]$. Then we have

$$\begin{aligned} dN(t) &= X(t)dY(t) + Y(t)dX(t) + dX(t)dY(t) \\ &= e^{\frac{1}{2}t}[-\sin W(t)dW(t) - \frac{1}{2}\cos W(t)dt] + \cos W(t)e^{\frac{1}{2}t}\frac{1}{2}dt \\ &= -e^{\frac{1}{2}t}\sin W(t)dW(t). \end{aligned}$$

Hence

$$e^{\frac{1}{2}T}\cos W(T) = 1 - \int_0^T e^{\frac{1}{2}t}\sin W(t)dW(t)$$

and also

$$F = \cos W(T) = e^{-\frac{1}{2}T} - e^{-\frac{1}{2}T} \int_0^T e^{\frac{1}{2}t}\sin W(t)dW(t).$$

Hence $E[F] = e^{-\frac{1}{2}T}$ and $\varphi(t) = -e^{\frac{1}{2}(t-T)}\sin W(t)$, $t \in [0, T]$. \square

1.5 Solution

(a) By Itô formula and Kolmogorov backward equation we have

$$\begin{aligned}
dY(t) &= \frac{\partial g}{\partial t}(t, X(t))dt + \frac{\partial g}{\partial x}(t, X(t))dX(t) + \frac{1}{2} \frac{\partial^2 g}{\partial x^2}(t, X(t))(dX(t))^2 \\
&= \frac{\partial}{\partial t}[P_{T-t}f(\xi)]_{|\xi=X(t)}dt + \sigma(X(t)) \frac{\partial}{\partial \xi}[P_{T-t}f(\xi)]_{|\xi=X(t)}dW(t) \\
&\quad + \{b(X(t)) \frac{\partial}{\partial \xi}[P_{T-t}f(\xi)]_{|\xi=X(t)} + \frac{1}{2}\sigma^2(X(t)) \frac{\partial^2}{\partial \xi^2}[P_{T-t}f(\xi)]_{|\xi=X(t)}\}dt \\
&= \frac{\partial}{\partial t}[P_{T-t}f(\xi)]_{|\xi=X(t)}dt + \sigma(X(t)) \frac{\partial}{\partial \xi}[P_{T-t}f(\xi)]_{|\xi=X(t)}dW(t) \\
&\quad + \frac{\partial}{\partial u}[P_u f(\xi)]_{|u=T-t}dt \\
&= \sigma(X(t)) \frac{\partial}{\partial \xi}[P_{T-t}f(\xi)]_{|\xi=X(t)}dW(t).
\end{aligned}$$

Hence

$$Y(T) = Y(0) + \int_0^T [\sigma(x) \frac{\partial}{\partial \xi} P_{T-t}f(\xi)]_{|\xi=X(t)} dW(t).$$

Since $Y(T) = g(T, X(T)) = [P_0 f(\xi)]_{|\xi=X(T)} = f(X(T))$ and $Y(0) = g(0, X(0)) = P_T f(X)$, (1.29) follows.

(b.1) If $F = W^2(T)$, we apply (a) to the case when $f(\xi) = \xi^2$ and $X(t) = x + W(t)$ (assuming $W(0) = 0$ as before). This gives

$$P_s f(\xi) = E^\xi[f(X(s))] = E^\xi[X^2(s)] = \xi^2 + s$$

and hence

$$E[F] = P_T f(x) = x^2 + T$$

and

$$\varphi(t) = [\frac{\partial}{\partial \xi}(\xi^2 + s)]_{|\xi=x+W(t)} = 2W(t) + 2x.$$

(b.2) If $F = W^3(T)$, we choose $f(\xi) = \xi^3$ and $X(t) = x + W(t)$ and get

$$P_s f(\xi) = E^\xi[X^3(s)] = \xi^3 + 3s\xi.$$

Hence

$$E[F] = P_T f(x) = x^3 + 3Tx$$

and

$$\varphi(t) = [\frac{\partial}{\partial \xi}(\xi^3 + 3(T-t)\xi)]_{|\xi=x+W(t)} = 3(x + W(t))^2 + 3(T-t).$$

(b.3) In this case $f(\xi) = \xi$, so

$$P_s f(\xi) = E^\xi[X(s)] = \xi e^{\rho s}$$

and so

$$E[F] = P_T f(x) = x e^{\rho T}$$

and

$$\begin{aligned}\varphi(t) &= [\alpha\xi \frac{\partial}{\partial\xi} (\xi e^{\rho(T-t)})]_{|\xi=X(t)} \\ &= \alpha X(t) \exp(\rho(T-t)) \\ &= \alpha x \exp\left\{\rho T - \frac{1}{2}\alpha^2 t + \alpha W(t)\right\}.\end{aligned}$$

(c) We proceed as in (a) and put

$$Y(t) = g(t, X(t)), \quad t \in [0, T], \quad \text{with } g(t, x) = P_{T-t}f(x)$$

and

$$dX(t) = b(X(t))dt + \sigma(X(t))dW(t); \quad X(0) = x \in \mathbb{R}^n,$$

where

$$b : \mathbb{R}^n \rightarrow \mathbb{R}^n, \quad \sigma : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times m} \quad \text{and} \quad W(t) = (W_1(t), \dots, W_m(t))$$

is the m -dimensional Wiener process. Then by Itô formula and (1.31), we have

$$\begin{aligned}dY(t) &= \frac{\partial g}{\partial t}(t, X(t))dt + \sum_{i=1}^n \frac{\partial g}{\partial x_i}(t, X(t))dX_i(t) \\ &\quad + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 g}{\partial x_i \partial x_j}(t, X(t))dX_i(t)dX_j(t) \\ &= \frac{\partial}{\partial t}[P_{T-t}f(\xi)]_{|\xi=X(t)}dt + [\sigma^T(\xi)\nabla_\xi(P_{T-t}f(\xi))]_{|\xi=X(t)}dW(t) \\ &\quad + [L_\xi(P_{T-t}f(\xi))]_{|\xi=X(t)}dt,\end{aligned}$$

where

$$L_\xi = \sum_{i=1}^n b_i(\xi) \frac{\partial}{\partial \xi_i} + \frac{1}{2} \sum_{i,j=1}^n (\sigma \sigma^T)_{ij}(\xi) \frac{\partial^2}{\partial \xi_i \partial \xi_j}$$

is the generator of the Itô diffusion $X(t)$, $t \geq 0$. So by the Kolmogorov backward equation we get

$$dY(t) = [\sigma^T(\xi)\nabla_\xi(P_{T-t}f(\xi))]_{|\xi=X(t)}dW(t)$$

and hence, as in (a),

$$Y(T) = f(X(T)) = P_T f(x) + \int_0^T [\sigma^T(\xi)\nabla_\xi(P_{T-t}f(\xi))]_{|\xi=X(t)}dW(t),$$

which gives, with $F = f(X(T))$,

$$E[F] = P_T f(x) \quad \text{and} \quad \varphi(t) = [\sigma^T(\xi)\nabla_\xi(P_{T-t}f(\xi))]_{|\xi=X(t)}, \quad t \in [0, T]. \quad \square$$

Problems of Chap. 2

2.4 Solution

(a) Since $W(t)$, $t \in [0, T]$, is \mathbb{F} -adapted, we have

$$\int_0^T W(t) \delta W(t) = \int_0^T W(t) dW(t) = \frac{1}{2} W^2(T) - \frac{1}{2} T.$$

(b) $\int_0^T \left(\int_0^T g(s) dW(s) \right) \delta W(t) = \int_0^T I_1[f_1(\cdot, t)] \delta W(t) = I_2[\tilde{f}_1]$, where $f_1(t_1, t) = g(t_1)$, $t \in [0, T]$. This gives

$$\tilde{f}_1(t_1, t) = \frac{1}{2} [g(t_1) + g(t)], \quad t \in [0, T],$$

and hence

$$\begin{aligned} I_2[\tilde{f}_1] &= 2 \int_0^T \int_0^{t_2} \tilde{f}_1(t_1, t_2) dW(t_1) dW(t_2) \\ &= \int_0^T \int_0^{t_2} g(t_1) dW(t_1) dW(t_2) + \int_0^T \int_0^{t_2} g(t_2) dW(t_1) dW(t_2) \quad (\text{S.1}) \\ &= \int_0^T \int_0^{t_2} g(t_1) dW(t_1) dW(t_2) + \int_0^T W(t_2) g(t_2) dW(t_2). \end{aligned}$$

Using integration by parts (i.e., the Itô formula), we see that

$$\begin{aligned} \left(\int_0^T g(t_1) dW(t_1) \right) W(T) &= \int_0^T \int_0^{t_2} g(t_1) dW(t_1) dW(t_2) \\ &\quad + \int_0^T g(t_2) W(t_2) dW(t_2) + \int_0^T g(t) dt. \quad (\text{S.2}) \end{aligned}$$

Combining (S.1) and (S.2), we get

$$\int_0^T \left(\int_0^T g(s) dW(s) \right) \delta W(t) = \left(\int_0^T g(t) dW(t) \right) W(T) - \int_0^T g(t) dt.$$

(c) By Problem 1.3 (c) we have

$$\int_0^T W^2(t_0) \delta W(t) = \int_0^T (t_0 + I_2[f_2(\cdot, t)]) \delta W(t),$$

where

$$f_2(t_1, t_2, t) = \chi_{[0, t_0]}(t_1) \chi_{[0, t_0]}(t_2), \quad t \in [0, T].$$

Now

$$\begin{aligned} \tilde{f}_2(t_1, t_2, t) &= \frac{1}{3} [f_2(t_1, t_2, t) + f_2(t, t_2, t_1) + f_2(t_1, t, t_2)] \\ &= \frac{1}{3} [\chi_{[0, t_0]}(t_1) \chi_{[0, t_0]}(t_2) + \chi_{[0, t_0]}(t) \chi_{[0, t_0]}(t_2) + \chi_{[0, t_0]}(t_1) \chi_{[0, t_0]}(t)] \\ &= \frac{1}{3} [\chi_{\{t_1, t_2 < t_0\}} + \chi_{\{t, t_2 < t_0\}} + \chi_{\{t_1, t < t_0\}}] \\ &= \chi_{\{t, t_1, t_2 < t_0\}} + \frac{1}{3} \chi_{\{t_1, t_2 < t_0 < t\}} + \frac{1}{3} \chi_{\{t, t_2 < t_0 < t_1\}} + \frac{1}{3} \chi_{\{t, t_1 < t_0 < t_2\}} \end{aligned}$$

and hence, using (1.15),

$$\begin{aligned} \int_0^T W^2(t_0) \delta W(t) &= t_0 W(T) + \int_0^T I_2[f_2(\cdot, t)] \delta W(t) \\ &= t_0 W(T) + I_3[\tilde{f}_2] = t_0 W(T) + 6J_3[\tilde{f}_2] \\ &= t_0 W(T) \\ &\quad + 6 \int_0^T \int_0^{t_3} \int_0^{t_2} \chi_{[0, t_0]}(t_1) \chi_{[0, t_0]}(t_2) \chi_{[0, t_0]}(t_3) dW(t_1) dW(t_2) dW(t_3) \\ &\quad + 6 \int_0^T \int_0^{t_3} \int_0^{t_2} \frac{1}{3} \chi_{\{t_1, t_2 < t_0 < t_3\}} dW(t_1) dW(t_2) dW(t_3) \\ &= t_0 W(T) + t_0^{3/2} h_3\left(\frac{W(t_0)}{\sqrt{t_0}}\right) \\ &\quad + 2 \int_{t_0}^T \int_0^{t_0} \int_0^{t_2} dW(t_1) dW(t_2) dW(t_3) \\ &= t_0 W(T) + t_0^{3/2} \left(\frac{W^3(t_0)}{t_0^{3/2}} - 3 \frac{W(t_0)}{\sqrt{t_0}}\right) \\ &\quad + 2 \int_{t_0}^T \left(\frac{1}{2} W^2(t_0) - \frac{1}{2} t_0\right) dW(t_3) \\ &= t_0 W(T) + W^3(t_0) - 3t_0 W(t_0) + (W^2(t_0) - t_0)(W(T) - W(t_0)) \\ &= W^2(t_0) W(T) - 2t_0 W(t_0). \end{aligned}$$

(d) By Problem 1.3 (d) and (1.15) we get

$$\begin{aligned} \int_0^T \exp(W(T)) \delta W(t) &= \int_0^T \sum_{n=0}^{\infty} \frac{1}{n!} e^{T/2} I_n[1] \delta W(t) \\ &= \sum_{n=0}^{\infty} \frac{1}{n!} e^{T/2} I_{n+1}[1] \\ &= e^{T/2} \sum_{n=0}^{\infty} \frac{1}{n!} T^{\frac{n+1}{2}} h_{n+1}\left(\frac{W(T)}{\sqrt{T}}\right). \end{aligned}$$

(e) Using Problem 1.3 (e) we have that

$$F = \int_0^T F \delta W(t) = I_1(f_1),$$

with $f_1(t) := \int_t^T g(s) ds$, $t \in [0, T]$. Hence

$$\int_0^T F \delta W(t) = I_2(\tilde{f}_1),$$

where

$$\tilde{f}_1(t_1, t_2) = \frac{1}{2} \left(\int_{t_1}^T g(s) ds + \int_{t_2}^T g(s) ds \right).$$

This gives

$$\begin{aligned} F &= \int_0^T F \delta W(t) \\ &= J_2 \left(\int_{t_1}^T g(s) ds \right) + J_2 \left(\int_{t_2}^T g(s) ds \right) \\ &= \int_0^T \int_0^{t_2} \left(\int_{t_1}^T g(s) ds \right) dW(t_1) dW(t_2) \\ &\quad + \int_0^T \left(\int_0^{t_2} 1 dW(t_1) \right) \left(\int_{t_2}^T g(s) ds \right) dW(t_2) \\ &= \int_0^T \left(\int_0^{t_2} g(s) W(s) ds \right) dW(t_2) + \int_0^T 2W(t_2) \left(\int_{t_2}^T g(s) ds \right) dW(t_2). \quad \square \end{aligned}$$

Problems of Chap. 3

3.2 Solution

(a) $D_t W(T) = \chi_{[0,T]}(t) = 1$, $t \in [0, T]$, by (3.8).

(b) By (3.8) we get

$$D_t \int_0^T s^2 dW(s) = t^2.$$

(c) By (3.2) we have

$$\begin{aligned} D_t \int_0^T \int_0^{t_2} \cos(t_1 + t_2) dW(t_1) dW(t_2) &= D_t \left(\frac{1}{2} I_2[\cos(t_1 + t_2)] \right) \\ &= \frac{1}{2} 2 I_1[\cos(\cdot + t)] \\ &= \int_0^T \cos(t_1 + t) dW(t_1). \end{aligned}$$

(d) By the chain rule, we get

$$\begin{aligned} D_t (3W(s_0)W^2(t_0) + \log(1 + W^2(s_0))) &= \left[3W^2(t_0) + \frac{2W(s_0)}{1 + W^2(s_0)} \right] \chi_{[0,s_0]}(t) \\ &\quad + 6W(s_0)W(t_0)\chi_{[0,t_0]}(t). \end{aligned}$$

(e) By Problem 2.4 (b) we have

$$\begin{aligned} D_t \int_0^T W(t_0) \delta W(t) &= D_t (W(t_0)W(T) - t_0) \\ &= W(t_0) \chi_{[0,T]}(t) + W(T) \chi_{[0,t_0]}(t) \\ &= W(t_0) + W(T) \chi_{[0,t_0]}(t). \quad \square \end{aligned}$$

3.3 Solution

(a) By Problem 1.3 (d) and (3.2), we have

$$\begin{aligned} D_t \exp \left\{ \int_0^T g(s) dW(s) \right\} &= D_t \sum_{n=0}^{\infty} I_n[f_n] = \sum_{n=1}^{\infty} n I_{n-1}[f_n(\cdot, t)] \\ &= \sum_{n=1}^{\infty} n \frac{1}{n!} \exp \left\{ \frac{1}{2} \|g\|^2 \right\} I_{n-1}[g(t_1) \dots g(t_{n-1}) g(t)] \\ &= g(t) \sum_{n=1}^{\infty} \frac{1}{(n-1)!} \exp \left\{ \frac{1}{2} \|g\|^2 \right\} I_{n-1}[g^{\otimes(n-1)}] \\ &= g(t) \exp \left\{ \int_0^T g(s) dW(s) \right\}. \end{aligned}$$

(b) The suggested chain rule and (3.8) give

$$\begin{aligned} D_t \exp \left\{ \int_0^T g(s) dW(s) \right\} &= \exp \left\{ \int_0^T g(s) dW(s) \right\} D_t \left(\int_0^T g(s) dW(s) \right) \\ &= g(t) \exp \left\{ \int_0^T g(s) dW(s) \right\}. \end{aligned}$$

(c) The points above together with Corollary 3.13 give

$$D_t \exp\{W(t_0)\} = \exp\{W(t_0)\} \chi_{[0,t_0]}(t). \quad \square$$

Problems of Chap. 4

4.1 Solution

(a) If $s > t$ we have

$$\begin{aligned} E_Q[\widetilde{W}(s)|\mathcal{F}_t] &= \frac{E[Z(T)\widetilde{W}(s)|\mathcal{F}_t]}{E[Z(T)|\mathcal{F}_t]} \\ &= \frac{E[Z(T)\widetilde{W}(s)|\mathcal{F}_t]}{Z(t)} \end{aligned} \tag{S.3}$$

$$\begin{aligned} &= Z^{-1}(t)E[E[Z(T)\widetilde{W}(s)|\mathcal{F}_s]|\mathcal{F}_t] \\ &= Z^{-1}(t)E[\widetilde{W}(s)E[Z(T)|\mathcal{F}_s]|\mathcal{F}_t] \\ &= Z^{-1}(t)E[\widetilde{W}(s)Z(s)|\mathcal{F}_t]. \end{aligned} \tag{S.4}$$

Applying Itô formula to $Y(t) := Z(t)\widetilde{W}(t)$, we get

$$\begin{aligned} dY(t) &= Z(t)d\widetilde{W}(t) + \widetilde{W}(t)dZ(t) + d\widetilde{W}(t)dZ(t) \\ &= Z(t)[\theta(t)dt + dW(t)] + \widetilde{W}(t)[- \theta(t)Z(t)dW(t)] - \theta(t)Z(t)dt \\ &= Z(t)[1 - \theta(t)\widetilde{W}(t)]dW(t), \end{aligned}$$

and hence $Y(t)$ is an \mathcal{F}_t -martingale (with respect to P). Therefore, by (S.3),

$$E_Q[\widetilde{W}(s)|\mathcal{F}_t] = Z^{-1}(t)E[Y(s)|\mathcal{F}_t] = Z^{-1}(t)Y(t) = \widetilde{W}(t).$$

(b) We apply the Girsanov theorem to the case with $\theta(t) = a$. Then X is a Wiener process with respect to the measure Q defined by

$$Q(d\omega) = Z(T, \omega)P(d\omega) \quad \text{on } \mathcal{F}_T,$$

where

$$Z(t) = \exp \left\{ -aW(t) - \frac{1}{2}a^2t \right\}, \quad 0 \leq t \leq T.$$

(c) In this case we have

$$\beta(t) = bY(t), \quad \alpha(t) = aY(t), \quad \gamma(t) = cY(t)$$

and hence we put

$$\theta = \frac{\beta(t) - \alpha(t)}{\gamma(t)} = \frac{b-a}{c}$$

and

$$Z(t) = \exp \left\{ -\theta W(t) - \frac{1}{2}\theta^2 t \right\}, \quad 0 \leq t \leq T.$$

Then

$$\widetilde{W}(t) := \theta t + W(t), \quad 0 \leq t \leq T,$$

is a Wiener process with respect to the measure Q defined by $Q(d\omega) = Z(T, \omega)P(d\omega)$ on \mathcal{F}_T and

$$dY(t) = bY(t)dt + cY(t)[d\widetilde{W}(t) - \theta dt] = aY(t)dt + cY(t)d\widetilde{W}(t). \quad \square$$

4.2 Solution

(a) $F = W(T)$ implies $D_t F = \chi_{[0,T]}(t) = 1$, for $t \in [0, T]$, and hence

$$E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) = \int_0^T 1 dW(t) = W(T) = F.$$

(b) $F = \int_0^T W(s)ds$ implies $D_t F = \int_0^T D_t W(s)ds = \int_0^T \chi_{[0,s]}(t)ds = \int_t^T ds = T-t$, which gives

$$\begin{aligned} E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) &= \int_0^T (T-t) dW(t) \\ &= \int_0^T W(s) dW(s) = F, \end{aligned}$$

using integration by parts.

(c) $F = W^2(T)$ implies $D_t F = 2W(T)D_t W(T) = 2W(T)$. Hence

$$\begin{aligned} E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) &= T + \int_0^T E[2W(T)|\mathcal{F}_t] dW(t) \\ &= T + 2 \int_0^T W(t) dW(t) \\ &= T + W^2(T) - T = W^2(T) = F. \end{aligned}$$

(d) $F = W^3(T)$ implies $D_t F = 3W^2(T)$. Hence, by Itô formula,

$$\begin{aligned}
E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) &= \int_0^T E[3W^2(T) | \mathcal{F}_t] dW(t) \\
&= 3 \int_0^T E[(W(T) - W(t))^2 + 2W(t)W(T) - W^2(t) | \mathcal{F}_t] dW(t) \\
&= 3 \int_0^T (T-t) dW(t) + 6 \int_0^T W^2(t) dW(t) - 3 \int_0^T W^2(t) dW(t) \\
&= 3 \int_0^T W^2(t) dW(t) - 3 \int_0^T W(t) dt = W^3(T).
\end{aligned}$$

(e) $F = \exp\{W(T)\}$ implies $D_t F = \exp\{W(T)\}$. Hence

$$\begin{aligned}
RHS &= E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) \\
&= e^{T/2} + \int_0^T E[\exp\{W(T)\} | \mathcal{F}_t] dW(t) \\
&= e^{T/2} + \int_0^T E[\exp\{W(T) - \frac{1}{2}T\} e^{T/2} | \mathcal{F}_t] dW(t) \\
&= e^{T/2} + \exp\{\frac{1}{2}T\} \int_0^T \exp\{W(t) - \frac{1}{2}t\} dW(t). \quad (\text{S.5})
\end{aligned}$$

Here we have used that

$$M(t) := \exp\left\{W(t) - \frac{1}{2}t\right\}$$

is a martingale. In fact, by Itô formula we have $dM(t) = M(t)dW(t)$. Combined with (S.5) this gives

$$RHS = \exp\{\frac{1}{2}T\} + \exp\{\frac{1}{2}T\}(M(T) - M(0)) = \exp W(T) = F.$$

(f) $F = (W(T) + T) \exp\{-W(T) - \frac{1}{2}T\}$ implies $D_t F = \exp\{-W(T) - \frac{1}{2}T\}[1 - W(T) - T]$. Note that

$$Y(t) := (W(t) + t)N(t), \quad \text{with} \quad N(t) = \exp\{-W(t) - \frac{1}{2}t\}$$

is a martingale, since

$$\begin{aligned} dY(t) &= (W(t) + t)N(t)(-dW(t)) + N(t)(dW(t) + dt) - N(t)dt \\ &= N(t)[1 - t - W(t)]dW(t). \end{aligned}$$

Therefore,

$$\begin{aligned} E[F] + \int_0^T E[D_t F | \mathcal{F}_t] dW(t) &= \int_0^T E[N(T)(1 - (W(T) + T)) | \mathcal{F}_t] dW(t) \\ &= \int_0^T N(t)(1 - (W(t) + t)) dW(t) \\ &= \int_0^T 1 dY(t) = Y(T) - Y(0) \\ &= (W(T) + T) \exp \left\{ -W(T) - \frac{1}{2}T \right\} = F. \quad \square \end{aligned}$$

4.3 Solution

(a) $\tilde{\varphi}(t) = E_Q[D_t F - F \int_t^T D_t \theta(s) d\tilde{W}(s) | \mathcal{F}_t]$. If $\theta(s)$, $t \in [0, T]$, is deterministic, then $D_t \theta = 0$ and hence

$$\begin{aligned} \tilde{\varphi}(t) &= E_Q[D_t F | \mathcal{F}_t] = E_Q[2W(T) | \mathcal{F}_t] \\ &= E_Q[2\tilde{W}(T) - 2 \int_0^T \theta(s) ds | \mathcal{F}_t] \\ &= 2\tilde{W}(t) - 2 \int_0^T \theta(s) ds \\ &= 2W(t) - 2 \int_t^T \theta(s) ds. \end{aligned}$$

(b) By application of the generalized Clark–Ocone formula, we have

$$\begin{aligned} \tilde{\varphi}(t) &= E_Q[D_t F | \mathcal{F}_t] \\ &= E_Q \left[\exp \left\{ \int_0^T \lambda(s) dW(s) \right\} \lambda(t) | \mathcal{F}_t \right] \\ &= \lambda(t) E_Q \left[\exp \left\{ \int_0^T \lambda(s) d\tilde{W}(s) - \int_0^T \lambda(s) \theta(s) ds \right\} | \mathcal{F}_t \right] \end{aligned}$$

$$\begin{aligned}
&= \lambda(t) \exp \left\{ \int_0^T \left(\frac{1}{2} \lambda^2(s) - \lambda(s) \theta(s) \right) ds \right\} E_Q \left[\exp \left\{ \int_0^T \lambda(s) d\widetilde{W}(s) \right. \right. \\
&\quad \left. \left. - \frac{1}{2} \int_0^T \lambda^2(s) ds \right\} | \mathcal{F}_t \right] \\
&= \lambda(t) \exp \left\{ \int_0^T \lambda(s) \left(\frac{1}{2} \lambda(s) - \theta(s) \right) ds \right\} \exp \left\{ \int_0^t \lambda(s) d\widetilde{W}(s) - \frac{1}{2} \int_0^t \lambda^2(s) ds \right\} \\
&= \lambda(t) \exp \left\{ \int_0^t \lambda(s) dW(s) + \int_t^T \lambda(s) \left(\frac{1}{2} \lambda(s) - \theta(s) \right) ds \right\}.
\end{aligned} \tag{S.6}$$

(c) By application of the generalized Clark–Ocone formula, we have

$$\begin{aligned}
\tilde{\varphi}(t) &= E_Q[D_t F - F \int_t^T D_t \theta(s) d\widetilde{W}(s) | \mathcal{F}_t] \\
&= E_Q[\lambda(t) F | \mathcal{F}_t] - E_Q[F \int_t^T d\widetilde{W}(s) | \mathcal{F}_t] \\
&=: A - B.
\end{aligned} \tag{S.7}$$

Now $\widetilde{W}(t) = W(t) + \int_0^t \theta(s) ds = W(t) + \int_0^t W(s) ds$ or

$$dW(t) + W(t) dt = d\widetilde{W}(t).$$

We solve this equation for $W(t)$ by multiplying by the “integrating factor” e^t and get

$$d(e^t W(t)) = e^t d\widetilde{W}(t).$$

Hence

$$W(u) = e^{-u} \int_0^u e^s d\widetilde{W}(s) \tag{S.8}$$

or

$$dW(u) = -e^{-u} \int_0^u e^s d\widetilde{W}(s) du + d\widetilde{W}(u). \tag{S.9}$$

Using (S.9) we may rewrite F as follows:

$$\begin{aligned} F &= \exp \left\{ \int_0^T \lambda(s) dW(s) \right\} \\ &= \exp \left\{ \int_0^T \lambda(s) d\widetilde{W}(s) - \int_0^T \lambda(u) e^{-u} \left(\int_0^u e^s d\widetilde{W}(s) \right) du \right\} \\ &= \exp \left\{ \int_0^T \lambda(s) d\widetilde{W}(s) - \int_0^T \left(\int_0^T \lambda(u) e^{-u} du \right) e^s d\widetilde{W}(s) \right\} \\ &= K(T) \exp \left\{ \frac{1}{2} \int_0^T \xi^2(s) ds \right\}, \end{aligned}$$

where

$$\xi(s) = \lambda(s) - e^s \int_s^T \lambda(u) e^{-u} du \quad (\text{S.10})$$

and

$$K(t) = \exp \left\{ \int_0^t \xi(s) d\widetilde{W}(s) - \frac{1}{2} \int_0^t \xi^2(s) ds \right\}, \quad 0 \leq t \leq T. \quad (\text{S.11})$$

Hence

$$A = E_Q[\lambda(t)F|\mathcal{F}_t] \quad (\text{S.12})$$

$$\begin{aligned} &= \lambda(t) \exp \left\{ \frac{1}{2} \int_0^T \xi^2(s) ds \right\} E[K(T)|\mathcal{F}_t] \\ &= \lambda(t) \exp \left\{ \frac{1}{2} \int_0^T \xi^2(s) ds \right\} K(t). \end{aligned} \quad (\text{S.13})$$

Moreover, if we put

$$H := \exp \left\{ \frac{1}{2} \int_0^T \xi^2(s) ds \right\}, \quad (\text{S.14})$$

we get

$$\begin{aligned} B &= E_Q[F(\widetilde{W}(T) - \widetilde{W}(t))|\mathcal{F}_t] \\ &= H E_Q[K(T)(\widetilde{W}(T) - \widetilde{W}(t))|\mathcal{F}_t] \end{aligned}$$

$$\begin{aligned}
&= H E_Q \left[K(t) \exp \left\{ \int_t^T \xi(s) d\widetilde{W}(s) - \frac{1}{2} \int_t^T \xi^2(s) ds \right\} (\widetilde{W}(T) - \widetilde{W}(t)) | \mathcal{F}_t \right] \\
&= H K(t) E_Q \left[\exp \left\{ \int_t^T \xi(s) d\widetilde{W}(s) - \frac{1}{2} \int_t^T \xi^2(s) ds \right\} (\widetilde{W}(T) - \widetilde{W}(t)) \right] \\
&= H K(t) E \left[\exp \left\{ \int_t^T \xi(s) dW(s) - \frac{1}{2} \int_t^T \xi^2(s) ds \right\} (W(T) - W(t)) \right]. \quad (\text{S.15})
\end{aligned}$$

This last expectation can be evaluated by using the Itô formula. Put

$$X(t) = \exp \left\{ \int_{t_0}^t \xi(s) dW(s) - \frac{1}{2} \int_{t_0}^t \xi^2(s) ds \right\}$$

and

$$Y(t) = X(t) (W(t) - W(t_0)).$$

Then

$$\begin{aligned}
dY(t) &= X(t) dW(t) + (W(t) - W(t_0)) dX(t) + dX(t) dW(t) \\
&= X(t) [1 + (W(t) - W(t_0)) \xi(t)] dW(t) + \xi(t) X(t) dt
\end{aligned}$$

and hence

$$\begin{aligned}
E[Y(T)] &= E[Y(t_0)] + E \left[\int_{t_0}^T \xi(s) X(s) ds \right] \\
&= \int_{t_0}^T \xi(s) E[X(s)] ds \quad (\text{S.16}) \\
&= \int_{t_0}^T \xi(s) ds. \quad (\text{S.17})
\end{aligned}$$

Combining (S.7) and (S.10)–(S.16), we conclude that

$$\begin{aligned}
\tilde{\varphi}(t) &= \lambda(t) HK(t) - H K(t) \int_t^T \xi(s) ds \\
&= \exp \left\{ \frac{1}{2} \int_0^T \xi^2(s) ds \right\} \exp \left\{ \int_0^t \xi(s) d\widetilde{W}(s) \right. \\
&\quad \left. - \frac{1}{2} \int_0^t \xi^2(s) ds \right\} \left[\lambda(t) - \int_t^T \xi(s) ds \right]. \quad \square
\end{aligned}$$

4.4 Solution

(a) Since $u = \frac{\mu - \rho}{\sigma}$ is constant we get using (4.27)

$$\theta_1(t) = e^{\rho t} \sigma^{-1} S^{-1}(t) E_Q[e^{-\rho T} D_t W(T) | \mathcal{F}_t] = e^{\rho(t-T)} \sigma^{-1} S^{-1}(t).$$

(b) Here $\mu = 0$, $\sigma(s) = c S^{-1}(s)$ and hence

$$u(s) = \frac{\mu - \rho}{\sigma} = -\frac{\rho}{c} S(s) = -\rho(W(s) + S(0)).$$

Hence

$$\int_t^T D_t u(s) d\widetilde{W}(s) = \rho [\widetilde{W}(t) - \widetilde{W}(T)].$$

Therefore,

$$B := E_Q \left[F \int_t^T D_t u(s) d\widetilde{W}(s) | \mathcal{F}_t \right] = \rho E_Q [e^{-\rho T} W(T) (\widetilde{W}(t) - \widetilde{W}(T)) | \mathcal{F}_t]. \quad (\text{S.18})$$

To proceed further, we need to express W in terms of \widetilde{W} : since

$$\widetilde{W}(t) = W(t) + \int_0^t u(s) ds = W(t) - \rho S(0)t - \rho \int_0^t W(s) ds,$$

we have

$$d\widetilde{W}(t) = dW(t) - \rho W(t) dt - \rho S(0) dt$$

or

$$e^{-\rho t} dW(t) - e^{-\rho t} \rho W(t) dt = e^{-\rho t} (d\widetilde{W}(t) + \rho S(0) dt)$$

or

$$d(e^{-\rho t} W(t)) = e^{-\rho t} d\widetilde{W}(t) + \rho e^{-\rho t} S(0) dt.$$

Hence

$$W(t) = S(0)[e^{\rho t} - 1] + e^{\rho t} \int_0^t e^{-\rho s} d\widetilde{W}(s). \quad (\text{S.19})$$

Substituting this in (S.18) we get

$$\begin{aligned} B &= \rho E_Q \left[\int_0^T e^{-\rho s} d\widetilde{W}(s) (\widetilde{W}(t) - \widetilde{W}(T)) | \mathcal{F}_t \right] \\ &= \rho E_Q \left[\int_0^t e^{-\rho s} d\widetilde{W}(s) (\widetilde{W}(t) - \widetilde{W}(T)) | \mathcal{F}_t \right] \end{aligned}$$

$$\begin{aligned}
& + \rho E_Q \left[\int_t^T e^{-\rho s} d\widetilde{W}(s) (\widetilde{W}(t) - \widetilde{W}(T)) | \mathcal{F}_t \right] \\
& = \rho E_Q \left[\int_t^T e^{-\rho s} d\widetilde{W}(s) (\widetilde{W}(t) - \widetilde{W}(T)) \right] \\
& = \rho \int_t^T e^{-\rho s} (-1) ds = e^{-\rho T} - e^{-\rho t}.
\end{aligned}$$

Hence

$$\begin{aligned}
\theta_1(t) &= e^{\rho t} c^{-1} (E_Q[D_t(e^{-\rho T} W(T)) | \mathcal{F}_t] - B) \\
&= e^{\rho t} c^{-1} (e^{-\rho T} - e^{-\rho T} + e^{-\rho t}) = c^{-1},
\end{aligned}$$

as expected.

(c) Here $\sigma = c S^{-1}(t)$ and hence

$$u(s) = \frac{\mu - \rho}{c} S(s) = \frac{\mu - \rho}{c} \left[e^{\mu s} S(0) + c \int_0^s e^{\mu(s-r)} dW(r) \right].$$

So

$$D_t u(s) = (\mu - \rho) e^{\mu(s-t)} \chi_{[0,s]}(t).$$

Hence

$$\begin{aligned}
\theta_1(t) &= e^{\rho t} c^{-1} E_Q[(D_t(e^{-\rho T} W(T)) - e^{-\rho T} W(T) \int_t^T D_t u(s) d\widetilde{W}(s)) | \mathcal{F}_t] \\
&= e^{\rho(t-T)} c^{-1} (1 - (\mu - \rho) E_Q[W(T) \int_t^T e^{\mu(s-t)} d\widetilde{W}(s) | \mathcal{F}_t]). \quad (\text{S.20})
\end{aligned}$$

Again we try to express W in terms of \widetilde{W} : since

$$\begin{aligned}
d\widetilde{W}(t) &= dW(t) + u(t) dt \\
&= dW(t) + \frac{\mu - \rho}{c} [e^{\mu t} S(0) + c \int_0^t e^{\mu(t-r)} dW(r)] dt,
\end{aligned}$$

we have

$$e^{-\mu t} d\widetilde{W}(t) = e^{-\mu t} dW(t) + \left[\frac{\mu - \rho}{c} S(0) + (\mu - \rho) \int_0^t e^{-\mu r} dW(r) \right] dt. \quad (\text{S.21})$$

If we put

$$X(t) = \int_0^t e^{-\mu r} dW(r), \quad \tilde{X}(t) = \int_0^t e^{-\mu r} d\tilde{W}(r),$$

(S.21) can be written as

$$d\tilde{X}(t) = dX(t) + \frac{\mu - \rho}{c} S(0) dt + (\mu - \rho) X(t) dt$$

or

$$d(e^{(\mu-\rho)t} X(t)) = e^{(\mu-\rho)t} d\tilde{X}(t) - \frac{\mu - \rho}{c} S(0) e^{(\mu-\rho)t} dt$$

or

$$\begin{aligned} X(t) &= e^{(\rho-\mu)t} \int_0^t e^{-\rho s} d\tilde{W}(s) - \frac{\mu - \rho}{c} S(0) e^{(\rho-\mu)t} \int_0^t e^{(\mu-\rho)s} ds \\ &= e^{(\rho-\mu)t} \int_0^t e^{-\rho s} d\tilde{W}(s) - \frac{S(0)}{c} [1 - e^{(\rho-\mu)t}]. \end{aligned}$$

From this we get

$$\begin{aligned} e^{-\mu t} dW(t) &= e^{(\rho-\mu)t} e^{-\rho t} d\tilde{W}(t) + (\rho - \mu) e^{(\rho-\mu)t} \left(\int_0^t e^{-\rho s} d\tilde{W}(s) \right) dt \\ &\quad + \frac{S(0)}{c} (\rho - \mu) e^{(\rho-\mu)t} dt \end{aligned}$$

or

$$dW(t) = d\tilde{W}(t) + (\rho - \mu) e^{\rho t} \left(\int_0^t e^{-\rho s} d\tilde{W}(s) \right) dt + \frac{S(0)}{c} (\rho - \mu) e^{\rho t} dt.$$

In particular,

$$W(T) = \tilde{W}(T) + (\rho - \mu) \int_0^T e^{\rho s} \left(\int_0^s e^{-\rho r} d\tilde{W}(r) \right) ds + \frac{S(0)}{\rho c} (\rho - \mu) (e^{\rho T} - 1). \quad (\text{S.22})$$

Substituted in (S.20) this gives

$$\begin{aligned} \theta_1(t) &= e^{\rho(t-T)} c^{-1} \left\{ 1 - (\mu - \rho) E_Q [\tilde{W}(T) \int_t^T e^{\mu(s-t)} d\tilde{W}(s) | \mathcal{F}_t] \right. \\ &\quad \left. + (\mu - \rho)^2 E_Q \left[\int_0^T e^{\rho s} \left(\int_0^s e^{-\rho r} d\tilde{W}(r) \right) ds \int_t^T e^{\mu(s-t)} d\tilde{W}(s) | \mathcal{F}_t \right] \right\} \end{aligned}$$

$$\begin{aligned}
&= e^{\rho(t-T)} c^{-1} \left\{ 1 - (\mu - \rho) \int_t^T e^{\mu(s-t)} ds \right. \\
&\quad \left. + (\mu - \rho)^2 \int_t^T e^{\rho s} E_Q \left[\left(\int_t^s e^{-\rho r} d\tilde{W}(r) \right) \left(\int_t^T e^{\mu(r-t)} d\tilde{W}(r) \right) | \mathcal{F}_t \right] ds \right\} \\
&= e^{\rho(t-T)} c^{-1} \left\{ 1 - \frac{\mu - \rho}{\mu} (e^{\mu(T-t)} - 1) + (\mu - \rho)^2 \int_t^T e^{\rho r} \left(\int_t^s e^{-\rho r} e^{\mu(r-t)} dr \right) ds \right\} \\
&= e^{\rho(t-T)} c^{-1} \left\{ 1 - \frac{\mu - \rho}{\rho} (e^{\rho(T-t)} - 1) \right\}. \quad \square
\end{aligned}$$

Problems of Chap. 5

5.5 Solution

(a) We have the following equations:

$$\begin{aligned}
\int_0^T W(T) \delta W(t) &= \int_0^T W(T) \diamond \dot{W}(t) dt \\
&= W(T) \diamond \int_0^T \dot{W}(t) dt \\
&= W(T) \diamond W(T) = W^2(T) - T,
\end{aligned}$$

by (5.65).

(b) We have the following equations:

$$\begin{aligned}
\int_0^T \left(\int_0^T g dW \right) \diamond \dot{W}(t) dt &= \left(\int_0^T g dW \right) \diamond \int_0^T \dot{W}(t) dt \\
&= \left(\int_0^T g dW \right) \diamond W(T) \\
&= \left(\int_0^T g dW \right) W(T) - \int_0^T g(s) ds,
\end{aligned}$$

by (5.62).

(c) We have the following equations:

$$\begin{aligned}
\int_0^T W^2(t_0) \delta W(t) &= \int_0^T (W^{\diamond 2}(t_0) + t_0) \delta W(t) \\
&= W^{\diamond 2}(t_0) \diamond W(T) + t_0 W(T) \\
&= W^{\diamond 2}(t_0) \diamond (W(T) - W(t_0)) \\
&\quad + W^{\diamond 2}(t_0) \diamond W(t_0) + t_0 W(T) \\
&= W^{\diamond 2}(t_0) (W(T) - W(t_0)) + W^{\diamond 3}(t_0) + t_0 W(T) \\
&= (W^2(t_0) - t_0) (W(T) - W(t_0)) + W^3(t_0) \\
&\quad - 3t_0 W(t_0) + t_0 W(T) \\
&= W^2(t_0) W(T) - 2t_0 W(t_0),
\end{aligned}$$

where we have used (5.40) and (5.65).

(d) We have the following equations:

$$\begin{aligned}
\int_0^T \exp(W(T)) \delta W(t) &= \exp(W(T)) \diamond \int_0^T \dot{W}(t) dt \\
&= \exp(W(T)) \diamond W(T) \\
&= \exp^{\diamond}(W(T) + \frac{1}{2}T) \diamond W(T) \\
&= \exp(\frac{1}{2}T) \sum_{n=0}^{\infty} \frac{1}{n!} W(T)^{\diamond(n+1)} \\
&= \exp(\frac{1}{2}T) \sum_{n=0}^{\infty} \frac{T^{\frac{n+1}{2}}}{n!} h_{n+1} \left(\frac{W(T)}{\sqrt{T}} \right). \quad \square
\end{aligned}$$

Problems of Chap. 6

6.4 Solution

Since $\dot{W}(s) = \sum_{i=1}^{\infty} e_i(s) H_{\epsilon^{(i)}}$, the expansion (6.8) for $D_t \dot{W}(s)$ is

$$\begin{aligned}
D_t \dot{W}(s) &= \sum_{i,k=1}^{\infty} e_i(s) e_k(t) H_{\epsilon^{(i)} - \epsilon^{(k)}} \chi_{\{i=k\}} \\
&= \sum_{i=1}^{\infty} e_i(s) e_i(t),
\end{aligned}$$

which is not convergent. Hence, for all $s \in \mathbb{R}$, we have $\dot{W}(s) \notin Dom(D_t)$. \square

Problems of Chap. 7

7.2 Solution

By Proposition 7.2 we have

$$g(e^Y) = \int_{\mathbb{R}} g(t^y) \frac{1}{\sqrt{2\pi v}} \exp^\diamond \left\{ -\frac{(y - Y)^{\diamond 2}}{2v} \right\} dy.$$

Substituting $e^y = z$ in the integral, we obtain

$$g(Z) = g(e^Y) = \int_0^\infty g(z) \frac{1}{\sqrt{2\pi v}} \exp^\diamond \left\{ -\frac{(\log z - \log Z)^{\diamond 2}}{2v} \right\} \frac{dz}{z}.$$

Hence the Donsker delta function of Z is

$$\delta_Z(z) = \frac{1}{\sqrt{2\pi v}} \frac{1}{z} \exp^\diamond \left\{ -\frac{(\log z - \log Z)^{\diamond 2}}{2v} \right\} \chi_{(0,\infty)}(z),$$

as claimed. \square

Problems of Chap. 8

8.2 Solution

$$(a) \int_0^T W(T) d^-W(t) = W(T) \int_0^T dW(t) = W^2(T)$$

(b) Consider the following equations:

$$\begin{aligned} & \int_0^T W(t) [W(T) - W(t)] d^-W(t) \\ &= \int_0^T W(t) W(T) d^-W(t) - \int_0^T W^2(t) d^-W(t) \\ &= W(T) \int_0^T W(t) dW(t) - \left[\frac{1}{3} W^3(T) - \int_0^T W(t) dt \right] \\ &= W(T) \left[\frac{1}{2} W^2(T) - \frac{1}{2} T \right] - \left[\frac{1}{3} W^3(T) - \int_0^T W(t) dt \right] \\ &= \frac{1}{6} W^3(T) - \frac{1}{2} T W(T) + \int_0^T W(t) dt. \end{aligned}$$

(c) Consider the following equations:

$$\begin{aligned} & \int_0^T \left(\int_0^T g(s) dW(s) \right) d^-W(t) \\ &= \int_0^T g(s) dW(s) \int_0^T d^-W(t) \\ &= W(T) \int_0^T g(s) dW(s). \quad \square \end{aligned}$$

Problems of Chap. 9

9.1 Solution

(a) By Theorem 9.4 we have

$$\begin{aligned}
 dY(t) &= 2Y(t)[\alpha(t)dt + \beta(t)dW(t)] + \beta^2(t)dt \\
 &\quad + \int_{\mathbb{R}_0} [(X(t) + \gamma(t, z))^2 - X^2(t) - 2X(t)\gamma(t, z)]\nu|(dz)dt \\
 &\quad + \int_{\mathbb{R}_0} [(X(t^-) + \gamma(t, z))^2 - X^2(t^-)]\tilde{N}(dt, dz) \\
 &= (2\alpha(t)Y(t) + \beta^2(t) + \int_{\mathbb{R}_0} \gamma^2(t, z)\nu(dz))dt + 2\beta(t)Y(t)dW(t) \\
 &\quad + \int_{\mathbb{R}_0} [2X(t^-)\gamma(t, z) + \gamma^2(t, z)]\tilde{N}(dt, dz).
 \end{aligned}$$

(b) By Theorem 9.4 we have

$$\begin{aligned}
 dY(t) &= Y(t)[\alpha(t)dt + \beta(t)dW(t)] + \frac{1}{2}Y(t)\beta^2(t)dt \\
 &\quad + \int_{\mathbb{R}_0} [\exp\{X(t) + \gamma(t, z)\} - \exp\{X(t)\} - \exp\{X(t)\}\gamma(t, z)]\nu(dz)dt \\
 &\quad + \int_{\mathbb{R}_0} [\exp\{X(t^-) + \gamma(t, z)\} - \exp\{X(t^-)\}]\tilde{N}(dt, dz) \\
 &= Y(t^-) \left[\left(\alpha(t) + \frac{1}{2}\beta^2(t) + \int_{\mathbb{R}_0} [\exp\{\gamma(t, z)\} - 1 - \gamma(t, z)]\nu(dz) \right) dt \right] \\
 &\quad + \beta(t)dW(t) + \int_{\mathbb{R}_0} [\exp\{\gamma(t, z)\} - 1]\tilde{N}(dt, dz).
 \end{aligned}$$

(c) By Theorem 9.4 we have

$$\begin{aligned}
 dY(t) &= -\sin X(t)[\alpha(t)dt + \beta(t)dW(t)] - \frac{1}{2}\cos X(t)\beta^2(t)dt \\
 &\quad + \int_{\mathbb{R}_0} [\cos(X(t) + \gamma(t, z)) - \cos X(t) + \sin X(t)\gamma(t, z)]\nu(dz)dt \\
 &\quad + \int_{\mathbb{R}_0} [\cos(X(t^-) + \gamma(t, z)) - \cos X(t^-)]\tilde{N}(dt, dz) \\
 &= \left[-\alpha(t)\sin X(t) - \frac{1}{2}\beta^2(t)\cos X(t) \right] \\
 &\quad + \cos X(t) \int_{\mathbb{R}_0} [\cos\gamma(t, z) - 1]\nu(dz) \\
 &\quad + \sin X(t) \int_{\mathbb{R}_0} [\gamma(t, z) - \sin\gamma(t, z)]\nu(dz)dt - \beta(t)\sin X(t)dW(t) \\
 &\quad + \int_{\mathbb{R}_0} [\cos X(t^-)(\cos\gamma(t, z) - 1) - \sin X(t^-)\sin\gamma(t, z)]\tilde{N}(dt, dz). \quad \square
 \end{aligned}$$

9.2 Solution

Applying Problem 9.1 (b) to the case

$$\alpha(t) = - \int_{\mathbb{R}_0} [e^{h(t)z} - 1 - h(t)z] \nu(dz),$$

$$\beta(t) = 0$$

$$\gamma(t, z) = h(t)z,$$

we obtain

$$\begin{aligned} dY(t) &= Y(t) \left\{ \left(- \int_{\mathbb{R}_0} [e^{h(t)z} - 1 - h(t)z] \nu(dz) + \int_{\mathbb{R}_0} [e^{h(t)z} - 1 - h(t)z] \nu(dz) \right) dt \right. \\ &\quad \left. + \int_{\mathbb{R}_0} [e^{h(t)z} - 1] \tilde{N}(dt, dz) \right\} \\ &= Y(t) \int_{\mathbb{R}_0} [e^{h(t)z} - 1] \tilde{N}(dt, dz). \quad \square \end{aligned}$$

9.6 Solution

(a) Since $d(t\eta(t)) = td\eta(t) + \eta(t)dt$, we have

$$\begin{aligned} F &= \int_0^T \eta(t) dt \\ &= T\eta(T) - \int_0^T t d\eta(t) \\ &= T \int_0^T \int_{\mathbb{R}_0} z \tilde{N}(dt, dz) - \int_0^T t \int_{\mathbb{R}_0} z \tilde{N}(dt, dz) \\ &= \int_0^T \int_{\mathbb{R}_0} (T-t)z \tilde{N}(dt, dz). \end{aligned}$$

Hence F is replicable, with replicating portfolio $\varphi(t) = T-t$, $t \in [0, T]$.

(c) Define

$$Y(t) = \exp \left\{ \eta(t) - \int_0^t \int_{\mathbb{R}_0} (e^z - 1 - z) \nu(dz) ds \right\}.$$

Then, by Problem 9.2 we have

$$dY(t) = Y(t^-) \int_{\mathbb{R}_0} (e^z - 1) \tilde{N}(dt, dz).$$

Hence

$$Y(T) = 1 + \int_0^T \int_{\mathbb{R}_0} Y(t^-) (e^z - 1) \tilde{N}(dt, dz).$$

Therefore,

$$\begin{aligned} e^{\eta(T)} &= Y(T)M \\ &= M + \int_0^T \int_{\mathbb{R}_0} M Y(t^-)(e^z - 1) \tilde{N}(dt, dz), \end{aligned}$$

where

$$M = \exp \left\{ T \int_{\mathbb{R}_0} (e^z - 1 - z) \nu(dz) \right\}.$$

Hence $F = \exp\{\eta(T)\}$ is not replicable unless $\nu(dz) = \lambda \delta_{z_0}(dz)$ is a point mass at some point $z_0 \neq 0$. In this case, the process $\eta(t)$, $t \in [0, T]$, corresponds to the compensated Poisson process with jump size z_0 and intensity $\lambda > 0$. \square

Problems of Chap. 10

10.1 Solution

(a) By the Itô formula we have

$$\begin{aligned} \eta^3(T) &= \int_0^T \int_{\mathbb{R}_0} [(\eta(t) + z)^3 - \eta^3(t) - 3\eta^2(t)z] \nu(dz) dt \\ &\quad + \int_0^T \int_{\mathbb{R}_0} [(\eta(t) + z)^3 - \eta^3(t)] \tilde{N}(dt, dz) \\ &= \int_0^T \int_{\mathbb{R}_0} [3\eta(t)z^2 + z^3] \nu(dz) dt \\ &\quad + \int_0^T \int_{\mathbb{R}_0} [3\eta^2(t)z + 3\eta(t)z^2 + z^3] \tilde{N}(dt, dz) \\ &= \int_0^T \int_{\mathbb{R}_0} z^3 \nu(dz) dt + \int_0^T \int_{\mathbb{R}_0} 3\eta(t)z^2 \nu(dz) dt \\ &\quad + \int_0^T \int_{\mathbb{R}_0} z^3 \tilde{N}(dz, dt) + \int_0^T \int_{\mathbb{R}_0} 3\eta(t)z^2 \tilde{N}(dt, dz) \\ &\quad + \int_0^T \int_{\mathbb{R}_0} 3\eta^2(t)z \tilde{N}(dt, dz) \\ &= \int_0^T \int_{\mathbb{R}_0} z^3 \nu(dz) dt + \int_0^T \int_{\mathbb{R}_0} 3 \left(\int_0^t \int_{\mathbb{R}_0} z_1 \tilde{N}(ds, dz_1) \right) z^2 \nu(dz) dt \\ &\quad + \int_0^T \int_{\mathbb{R}_0} z^3 \tilde{N}(dt, dz) + \int_0^T \int_{\mathbb{R}_0} \left(\int_0^t \int_{\mathbb{R}_0} 3z_1 \tilde{N}(dt, dz_1) \right) z^2 \tilde{N}(dt, dz) \\ &\quad + \int_0^T \int_{\mathbb{R}_0} 3\eta^2(t)z \tilde{N}(dt, dz). \end{aligned}$$

Now we also have

$$\begin{aligned} \int_0^T \int_{\mathbb{R}_0} 3\eta^2(t)z\tilde{N}(dt, dz) &= \int_0^T \int_{\mathbb{R}_0} 3\left(t \int_{\mathbb{R}_0} \zeta^2 \nu(d\zeta)\right) z\tilde{N}(dt, dz) \\ &\quad + \int_0^T \int_{\mathbb{R}_0} 3\left(\int_0^t \int_{\mathbb{R}_0} z_1^2 \tilde{N}(dt, dz_1)\right) z\tilde{N}(dt, dz) \\ &\quad + \int_0^T \int_{\mathbb{R}_0} 6\left(\int_0^t \int_{\mathbb{R}_0} \int_0^{t_2} \int_{\mathbb{R}_0} z_1 z_2 \tilde{N}(dt_1, dz_1) \tilde{N}(dt_2, dz_2)\right) z\tilde{N}(dt, dz) \\ &= I_1(3t_1 z_1 \int_{\mathbb{R}_0} \zeta^2 \nu(d\zeta)) + J_2(3z_1^2 z_2) + J_3(6z_1 z_2 z_3). \end{aligned}$$

Summing up we get

$$\begin{aligned} \eta^3(T) &= Tm_3 + I_1(3Tm_2 z_1 + z_1^3) + J_2(3z_1 z_2^2 + 3z_1^2 z_2) + J_3(6z_1 z_2 z_3) \\ &= Tm_3 + I_1(3Tm_2 z_1 + z_1^3) + I_2\left(\frac{3}{2}(z_1 z_2^2 + z_1^2 z_2)\right) + I_3(z_1 z_2 z_3), \end{aligned}$$

where $m_i = \int_{\mathbb{R}_0} \zeta^i \nu(d\zeta)$, $i = 2, 3, \dots$

(b) By Example 10.4 we have that

$$F_0 := \exp \left\{ \int_0^T \int_{\mathbb{R}_0} z\tilde{N}(dt, dz) - \int_0^T \int_{\mathbb{R}_0} (e^z - 1 - z)\nu(dz)dt \right\}$$

has chaos expansion $F_0 = \sum_{n=0}^{\infty} I_n(f_n)$ given by (10.7):

$$f_n = \frac{1}{n!} (e^z - 1)^{\otimes n}(t_1, z_1, \dots, t_n, z_n).$$

It follows that F has the expansion

$$F = \sum_{n=0}^{\infty} I_n(Kf_n) \quad \text{where} \quad K := \exp \left\{ T \int_{\mathbb{R}_0} (e^z - 1 - z)\nu(dz) \right\}.$$

(c) We have the following equalities:

$$F = \int_0^T g(s)d\eta(s) = \int_0^T \int_{\mathbb{R}_0} g(s)z\tilde{N}(ds, dz) = I_1(f_1),$$

where $f_1(s, z) = g(s)z$, $s \in [0, T]$, $z \in \mathbb{R}_0$.

(d) We have the following equalities:

$$\begin{aligned} F &= \int_0^T g(s)\eta(s) = \int_0^T g(s) \int_0^s \int_{\mathbb{R}_0} z\tilde{N}(ds, dz) \\ &= \int_0^T \int_{\mathbb{R}_0} \left(\int_t^T g(s)ds \right) z\tilde{N}(dt, dz) = I_1(f_1) \end{aligned}$$

with $f_1(t, z) = z \int_t^T g(s)ds$, $t \in [0, T]$, $z \in \mathbb{R}_0$. \square

Problems of Chap. 11

11.2 Solution

(a) Since

$$\int_0^T g(s)d\eta(s) = \int_0^T \int_{\mathbb{R}_0} g(s)z \tilde{N}(dt, dz) = I_1(g(t_1)z_1),$$

we get

$$\begin{aligned} & \int_0^T \left(\int_0^T g(s)d\eta(s) \right) f(t)\delta\eta(t) = \int_0^T \int_{\mathbb{R}_0} I_1(g(t_1)z_1)f(t_z)z_2 \tilde{N}(dt_2, dz_2) \\ &= I_2\left(\frac{1}{2}(g(t_1)f(t_2) + g(t_2)f(t_1))z_1z_2\right) \\ &= \int_0^T \int_{\mathbb{R}_0} \left(\text{int}_0^{t_2^-} \int_{\mathbb{R}_0} \frac{1}{2}(g(t_1)f(t_2) + g(t_2)f(t_1))z_1z_2 \tilde{N}(dt_1, dz_1) \right) \tilde{N}(dt_2, dz_2) \\ &= \int_0^T \int_{\mathbb{R}_0} \left[f(t_2) \int_0^{t_2^-} g(t_1)d\eta(t_1) + g(t_2) \int_0^{t_2^-} f(t_1)d\eta(t_1) \right] z_2 \tilde{N}(dt_2, dz_2) \\ &= \int_0^T \left(\int_0^{t_2^-} g(t_1)d\eta(t_1) \right) f(t_2)d\eta(t_2) + \int_0^T \left(\int_0^{t_2^-} f(t_1)d\eta(t_1) \right) g(t_2)d\eta(t_2). \end{aligned} \tag{S.23}$$

(b) Using the computation in (S.23), but with f and g interchanged, we get

$$\int_0^T \left(\int_0^T f(t)d\eta(t) \right) g(s)\delta\eta(s) = I_2\left(\frac{1}{2}(f(t_1)g(t_2) + f(t_2)g(t_1))z_1z_2\right),$$

which is the same as we obtained in (a).

(c) This is a direct consequence of (a) and (b). \square

Problems of Chap. 12

12.1 Solution

(a) By Problem 10.1 we have the expansion

$$\eta^3(T) = Tm_3 + I_1(3Tm_2z_1 + z_1^3) + I_2\left(\frac{3}{2}(z_1z_2^2 + z_1^2z_2)\right) + I_3(z_1z_2z_3),$$

where $m_i = \int_{\mathbb{R}_0} \zeta^i \nu(d\zeta)$, $i = 1, 2, \dots$. This gives

$$\begin{aligned} D_{t,z}\eta^3(T) &= 3Tm_2z + z^3 + 3I_1(z_1z^2 + z_1^2z) + 3I_2(z_1z_2z) \\ &= 3Tm_2z + z^3 + 3z^2\eta(T) + 3zI_1(z_1^2) + 3zI_2(z_1z_2). \end{aligned}$$

If we use that

$$\eta^2(T) = Tm_2 + I_1(z^2) + I_2(z_1 z_2)$$

(see Example 12.9), we can see that the above expression can be written

$$D_{t,z}\eta^3(T) = 3\eta^2(T)z + 3\eta(T)z^2 + z^3.$$

(b) By Problem 10.1 we have the expansion

$$e^{\eta(T)} = \sum_{n=0}^{\infty} I_n(g_n),$$

where

$$\begin{aligned} g_n &= K f_n \quad \text{with} \quad f_n = \frac{1}{n!}(e^z - 1)^{\otimes n}, \quad n = 1, 2, \dots \\ \text{and} \quad K &= \exp \left\{ T \int_{\mathbb{R}_0} (e^z - 1 - z) \nu(dz) \right\}. \end{aligned}$$

This gives

$$\begin{aligned} D_{t,z} e^{\eta(T)} &= \sum_{n=1}^{\infty} n I_{n-1}(g_n(\cdot, t.z)) \\ &= \sum_{n=1}^{\infty} n I_{n-1} \left(\frac{K}{n!} (e^z - 1)^{\otimes n-1} \right) (e^z - 1) \\ &= \sum_{n=1}^{\infty} I_{n-1} \left(\frac{K}{(n-1)!} (e^z - 1)^{\otimes n-1} \right) (e^z - 1) \\ &= e^{\eta(T)} (e^z - 1). \quad \square \end{aligned}$$

12.2 Solution

The direct application of Theorem 12.8 yields

- (a) $D_{t,z}\eta^3(T) = (\eta(T) + z)^3 - \eta^3(T) = 3\eta^2(T)z + 3\eta(T)z^2 + z^3.$
- (b) $D_{t,z}e^{\eta(T)} = e^{\eta(T)+z} - e^{\eta(T)} = e^{\eta(T)}(e^z - 1).$

Compare with the solution of Problem 12.1. \square

Problems of Chap. 13

13.1 Solution

Recall that

$$\eta(t) = m_2 \sum_{i=1}^{\infty} \left(\int_0^t e_i(s) ds \right) K_{\varepsilon^{(i,1)}}, \quad t \in \mathbb{R},$$

and

$$\dot{\eta}(t) = m_2 \sum_{i=1}^{\infty} e_i(t) K_{\varepsilon^{(i,1)}}, \quad t \in \mathbb{R},$$

where $\varepsilon^{(i,j)} = \varepsilon^{(\kappa(i,j))}$. Hence

$$\frac{\eta(t+h) - \eta(t)}{h} - \dot{\eta}(t) = m_2 \sum_{i=1}^{\infty} \left(\frac{1}{h} \int_t^{t+h} [e_i(s) - e_i(t)] ds \right) K_{\varepsilon^{(i,1)}}.$$

By (13.11) we have

$$\kappa(i, 1) = 1 + \frac{i(i-1)}{2}.$$

Therefore, if we put

$$a_i(h) := \frac{1}{h} \int_t^{t+h} [e_i(s) - e_i(t)] ds,$$

we have

$$\begin{aligned} \left\| \frac{\eta(t+h) - \eta(t)}{h} - \dot{\eta}(t) \right\|_{-q}^2 &= m_2^2 \sum_{i=1}^{\infty} |a_i(h)|^2 \varepsilon^{(i,1)}! (2\mathbb{N})^{-q\varepsilon^{(i,1)}} \\ &= m_2^2 \sum_{i=1}^{\infty} |a_i(h)|^2 (2\kappa(i, 1))^{-q} \\ &= m_2^2 \sum_{i=1}^{\infty} |a_i(h)|^2 (2 + i(i-1))^{-q} \end{aligned}$$

by (13.25). Since

$$\sup_{t \in \mathbb{R}} |e_k(t)| = \mathcal{O}(k^{-1/2})$$

(see [105]), we see that

$$\sup \{|a_i(h)| \mid h \in [0, 1], i = 1, 2, \dots\} < \infty.$$

Moreover, since

$$a_i(h) \rightarrow 0, \quad h \rightarrow 0 \quad (i = 1, 2, \dots),$$

we can conclude that

$$\left\| \frac{\eta(t+h) - \eta(t)}{h} - \dot{\eta}(t) \right\|_{-q}^2 \rightarrow 0, \quad h \rightarrow 0,$$

for all $q \geq 1$, by bounded convergence. This implies that

$$\frac{d}{dt} \eta(t) = \dot{\eta}(t) \quad \text{in } (\mathcal{S})^*. \quad \square$$

Problems of Chap. 15

15.1 Solution

(a) By Lemma 15.5 we have

$$\begin{aligned}
X(t) &= \int_0^t \int_{\mathbb{R}_0} \eta(T)z \tilde{N}(\delta s, dz) \\
&= \int_0^t \int_{\mathbb{R}_0} (\eta(T)z + D_{t^+, z}\eta(T)z) \tilde{N}(\delta s, dz) - \int_0^t \int_{\mathbb{R}_0} z^2 \tilde{N}(ds, dz) \\
&= \int_0^t \int_{\mathbb{R}_0} \eta(T)z \tilde{N}(d^-s, dz) - \int_0^T \int_{\mathbb{R}_0} z^2 \nu(dz)ds - \int_0^t \int_{\mathbb{R}_0} z^2 \tilde{N}(ds, dz) \\
&= \eta(t)\eta(T) - \int_0^t \int_{\mathbb{R}_0} z^2 N(ds, dz), \quad t \in [0, T].
\end{aligned}$$

(b) Since

$$D_{t^+, z}\eta(t) = D_{t^+, z} \left(\int_0^t \int_{\mathbb{R}_0} z^2 N(ds, dz) \right) = 0,$$

by applications of the chain rule we get that $D_{t^+, z}X(t) = \eta(t)z$.

(c) By (15.8), we have that

$$\theta(t, z) := S_{t, z}\gamma(t, z) = zS_{t, z}\eta(T) = z(\eta(T) - z).$$

(d) First note that by the above we have

$$\begin{aligned}
A(t, z) &:= D_{t^+, z} \left(2X(t^-)z(\eta(T) - z) + z^2(\eta(T) - z)^2 \right) \\
&= 2X(t^-)z^2 + z(\eta(T) - z)2\eta(t^-)z + 2\eta(t^-)z^3 \\
&\quad + z^2(\eta^2(T) - (\eta(T) - z)^2) \\
&= 2X(t^-)z^2 + 2\eta(t^-)\eta(T)z^2 + 2\eta(T)z^3 - z^4.
\end{aligned}$$

Therefore, the Itô formula for Skorohod integrals gives

$$\begin{aligned}
\delta X^2(t) &= \int_{\mathbb{R}_0} \left[(X(t^-) + \theta(t, z))^2 - X^2(t^-) + A(t, z) \right] \tilde{N}(\delta t, dz) \\
&\quad + \int_{\mathbb{R}_0} \left[(X(t) + \theta(t, z))^2 - X^2(t) - 2X(t)\theta(t, z) - A(t, z) \right. \\
&\quad \quad \left. - f'(X(t))D_{t^+, z}\theta(t, z) \right] \nu(dz)dt \\
&= \int_{\mathbb{R}_0} \left[2X(t^-)z(\eta(T) - z) + z^2(\eta(T) - z)^2 + A(t, z) \right] \tilde{N}(\delta t, dz) \\
&\quad + \int_{\mathbb{R}_0} \left[z^2(\eta(T) - z)^2 + A(t, z) - 2X(t)z^2 \right] \nu(dz)dt. \quad \square
\end{aligned}$$

Problems of Chap. 16

16.1 Solution

(a) By the chain rule and (16.5) we have

$$D_t F_\pi(T) = D_t \frac{1}{X_\pi(T)} = -\frac{1}{X_\pi^2(T)} D_t X_\pi(T) = -\frac{1}{X_\pi(T)} \sigma(t) \pi(t)$$

and

$$\begin{aligned} D_{t,z} F_\pi(T) &= \frac{1}{X_\pi(T) + D_{t,z} X_\pi(T)} - \frac{1}{X_\pi(T)} \\ &= \frac{1}{X_\pi(T)(1 + \pi(t)\theta(t, z))} - \frac{1}{X_\pi(T)} \\ &= \frac{-\pi(t)\theta(t, z)}{X_\pi(T)(1 + \pi(t)\theta(t, z))}. \end{aligned}$$

(b) The equation for the optimal deterministic portfolio π is obtained by choosing $\mathcal{E}_t = \mathcal{F}_0$ for all $t \in [0, T]$ in (16.12). This gives

$$\mu(s) - \rho(s) - 2\sigma^2(s)\pi(s) - \int_{\mathbb{R}_0} \frac{\pi(s)\theta^2(s, z)(2 + \pi(s)\theta(s, z))}{(1 + \pi(s)\theta(s, z))^2} \nu(dz) = 0. \quad \square$$

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Notation and Symbols

Numbers

\mathbb{N}	The natural numbers
\mathbb{Z}	The integer numbers
\mathbb{Q}	The rational numbers
\mathbb{R}	The real numbers
\mathbb{R}_0	p. 162
\mathbb{C}	The complex numbers
$\mathbb{C}^{\mathbb{N}}$	The set of all sequences of complex numbers
T	p. 7
\mathcal{J}	p. 66
$(2\mathbb{N})^\alpha$	p. 68
$\mathbb{K}_q(R)$	p. 74
$\mathbb{C}_c^{\mathbb{N}}$	p. 75
m_2	p. 217

Measures

P	pp. 7, 64, 215, 238
P^W	pp. 197, 238
$P^{\widetilde{N}}$	pp. 197, 238
$\lambda = dt$	Lebesgue measure
ν	p. 162
ρ	p. 217

Operations

$f \otimes g$	p. 11
$f \hat{\otimes} g$	p. 11
$W^{\otimes(n+1)}$	p. 13
$\langle \omega, \phi \rangle$	p. 64
$X \diamond Y$	pp. 70, 221

Spaces and Norms

(Ω, \mathcal{F}, P)	pp. 7, 64, 161, 197, 215, 238
S_n	p. 8
G_n	p. 178
$C_0([0, T])$	pp. 27, 355
$L^2([0, T]^n)$	p. 8
$\tilde{L}^2([0, T]^n)$	p. 8
$\tilde{L}^2((\lambda \times \nu)^n)$	p. 178
$L^2(S_n)$	p. 8
$L^2(G_n)$	p. 178
$L^2((\lambda \times \nu)^n)$	p. 177
$L^2(([0, T] \times \mathbb{R}_0)^n)$	p. 177
$L^2(P)$	p. 9
$L^2(\mathcal{F}_T, P)$	p. 168
$L^2(P \times \lambda)$	p. 22
$L^2(P \times \lambda \times \nu)$	p. 188
$L^2(S)$	p. 210
$Dom(\delta)$	pp. 20, 183
$\mathbb{D}_{1,2}$	pp. 28, 187, 360
$Dom(D_t)$	p. 89
$\mathcal{S} = \mathcal{S}(\mathbb{R}^d),$	p. 63
$\ \cdot\ _{K,\alpha},$	p. 63
$\mathcal{S}' = \mathcal{S}'(\mathbb{R}^d)$	p. 64
$(\mathcal{S})_k$	p. 69
$\ f\ _k^2$	p. 69
(\mathcal{S})	pp. 69, 219
$(\mathcal{S})_{-q}$	p. 69
$\ F\ _{-q}^2$	p. 69
$(\mathcal{S})^*$	pp. 69, 219
\mathcal{G}_λ	pp. 77, 229
\mathcal{G}	pp. 78, 229
\mathcal{G}^*	pp. 78, 229
\mathbb{D}_0	p. 140
$\mathbb{D}_{1,2}^\mathcal{E}$	p. 190
$\tilde{\mathbb{D}}_{1,2}$	p. 207
$\mathbb{D}_{1,2}^W$	p. 240
\mathcal{M}	p. 268
$\mathbb{M}_{1,2}$	p. 268
$\mathbb{D}_{1,p}$	p. 341
$\mathbb{D}_{1,\infty}$	p. 341
$\mathbb{D}_{k,p}$	p. 343
\mathbb{D}_∞	p. 343
$\mathcal{D}_{1,2}$	p. 358
\mathbb{P}	p. 359

Filtrations and σ -Algebras

\mathbb{F}	pp. 7, 163
\mathcal{F}_t	p. 7
\mathcal{F}_G	p. 30
\mathcal{F}_T^W	p. 197
$\mathcal{F}_{\widetilde{T}}^N$	p. 197
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\mathcal{G}_t	p. 144
\mathbb{E}	p. 200
\mathcal{E}_t	p. 200
\mathbb{H}	p. 302
\mathcal{H}_t	p. 305

Functions, Random Variables, and Transforms

\tilde{f}	pp. 8, 20, 177
h_n	p. 10
e_k	p. 66
$\chi = \chi_A(x) = \chi_{\{x \in A\}}$	p. 15
H_α	p. 66
$\mathcal{H}X(\cdot) = \tilde{X}(\cdot)$	p. 74
f^\diamond	p. 76
$\delta_Y(\cdot)$	pp. 114, 122, 244
$\Delta\eta(t)$	p. 162
\mathbb{H}_α	p. 197
l_m	p. 217
p_j	p. 217
$\kappa(i, j)$	p. 217
$\delta_{\kappa(i, j)}$	p. 217
K_α	p. 218
$L_t(x)$	p. 250

Processes and Fields

$W = W(t) = W(t, \omega)$	p. 7
$w(\phi, \omega) = w_\phi(\omega)$	p. 64
$\dot{W}(t)$	p. 70
$\eta(t) = \eta(t, \omega)$	p. 161
$N = N(dt, dz)$	p. 162
$\tilde{N} = \tilde{N}(dt, dz)$	p. 163
$\dot{\eta}(t)$	p. 220
$\dot{\tilde{N}}(t, z)$	p. 220

Integrals and Differentials

$J_n(f)$	pp. 8, 178
$I_n(g)$	pp. 10, 178
$\delta(u)$	pp. 20, 183
$\tilde{N}(\delta t, dz)$	p. 183
$\delta\eta(t)$	pp. 20, 184
$d^-W(s)$	p. 134
$\tilde{N}(d^-t, dz)$	p. 267

Derivatives

$\partial^\alpha,$	p. 64
$D_t F$	pp. 28, 88, 89, 360
$D_\gamma F$	pp. 87, 357
$\mathbf{D}_\gamma F$	p. 358
$\mathbf{D}_t F$	p. 358
$D_{t,z} F$	pp. 188, 230
$D_{t^+}\varphi(t)$	p. 140
$D_{t^+,z}\theta(t, z)$	p. 268
$\mathcal{D}_t F$	p. 240
D_{t_1, \dots, t_j}^j	p. 343
$D_y f$	pp. 354, 356

Admissible Controls

\mathcal{A}	p. 170
$\mathcal{A}_{\mathbb{F}}$	pp. 131, 301,
$\mathcal{A}_{\mathbb{G}}$	pp. 145, 132
$\mathcal{A}_{\mathbb{G}, \mathcal{Q}}$	p. 150
$\mathcal{A}_{\mathbb{E}}$	pp. 200, 278, 291, 296
$\mathcal{A}_{\mathbb{H}}$	pp. 305, 311, 322

Notations

M^T	Transpose of a matrix M
$P \sim Q$	Measure P is equivalent to measure Q
$E[F]$	(generalized) Expectation w.r.t. measure P
$E_Q[F]$	Expectation w.r.t. measure Q
$E[F \mathcal{F}_t]$	(generalized) Conditional expectation
càdlàg	Right continuous with left limits
càglàd	Left continuous with right limits
a.a., a.e., a.s.	Almost all, almost everywhere, almost surely
s.t.	Such that
w.r.t.	With respect to
SDE	Stochastic differential equation
BSDE	Backward stochastic differential equation
$:=$	Equal to by definition

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 \mathcal{S} -transform, 76
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