

Mathematics for New Technologies in Finance

Solution sheet 4

Exercise 4.1 (Signatures) Through this exercise, we let $E = \mathbb{R}^d$, J an interval on $\mathbb{R}$, and denote $\mathbf{Sig}_J: \mathcal{C}_0^1(J, E) \rightarrow \mathbf{T}(E)$ the signature map such that for all $X \in \mathcal{C}_0^1(J, E)$ and we let $\mathbf{Sig}_J^{(M)}$ denote the truncated signature map up to order M : $\mathbf{Sig}_J^{(M)}(X) = (1, \mathbf{s}_1, \dots, \mathbf{s}_M) \in \mathbf{T}^{(M)}(E)$. Let $X \in \mathcal{C}_0^1([0, s], E)$ and $Y \in \mathcal{C}_0^1([s, t], E)$.

(a) Let $X_t = t\mathbf{x} \in \mathbb{R}^d$ for all $t \in [0, 1]$. Calculate $\mathbf{Sig}_{[0,1]}(X)$.

(b) Let $X \in \mathcal{C}_0^1([0, T], E)$ and $X_0 = 0$. Prove that

$$\mathbf{Sig}_{[0,1]}(X)_{1,2} + \mathbf{Sig}_{[0,1]}(X)_{2,1} = \mathbf{Sig}_{[0,1]}(X)_1 \cdot \mathbf{Sig}_{[0,1]}(X)_2.$$

(c) Let $X \in \mathcal{C}_0^1([0, 1], \mathbb{R})$ s.t. $X_t = \sin(t)$ for all $t \in [0, 1]$. Calculate $\mathbf{Sig}_{[0,1]}^{(2)}(X)$ i.e. the signatures of X up to order 2.

(d) Let $X \in \mathcal{C}_0^1([0, 1], \mathbb{R}^2)$ s.t. $X_t = (t, \sin(t))$ for all $t \in [0, 1]$. Calculate $\mathbf{Sig}_{[0,1]}^{(2)}(X)$ i.e. the signatures of X up to order 2.

(e) Let $X \in \mathcal{C}_0^1([0, 1], \mathbb{R})$ and $n \in \mathbb{N}$. Calculate $\int_0^1 t^n dX_t$ when

1. $X_t = t$
2. $X_t = \sin(t)$

Solution 4.1

(a)

$$\mathbf{Sig}_{[0,1]}(X) = (1, \mathbf{x}, \frac{\mathbf{x}^{\otimes 2}}{2!}, \dots). \quad (1)$$

(b) By integration by part, we directly show the equality

$$\int_0^1 u_t^{(1)} du_t^{(2)} + \int_0^1 u_t^{(2)} du_t^{(1)} = \int_0^1 d(u^{(1)} \cdot u^{(2)})_t = u_1^{(1)} \cdot u_1^{(2)} \quad (2)$$

(c)

$$\left(1, \sin(1), \int_0^1 \sin(t) \cos(t) dt\right) \quad (3)$$

(d)

$$\left(1, 1, \sin(1), \frac{1}{2}, \int_0^1 \sin(t) dt, \int_0^1 t \cos(t) dt, \int_0^1 \sin(t) \cos(t) dt\right) \quad (4)$$

(e) 1.

$$\frac{t^{n+1}}{n+1} \Big|_0^1 \quad (5)$$

2.

$$\begin{aligned}
\int_0^1 t^n d\sin(t) &= \sin(t)t^n \Big|_0^1 + \int_0^1 nt^{n-1} d\cos(t) \\
&= \sin(t)t^n \Big|_0^1 + \int_0^1 nt^{n-1} d\cos(t) \\
&= \sin(t)t^n \Big|_0^1 + n\cos(t)t^{n-1} \Big|_0^1 - \int_0^1 n(n-1)t^{n-2} d\sin(t) \\
&= \dots
\end{aligned} \tag{6}$$

Exercise 4.2 (Ito's formula) Let W be a Brownian motion on $[0, \infty)$ and define

$$Q^n(W) = \sum_{i=1}^n (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2.$$

(a) Prove that $Q^n(W)$ converges to 1 in $\mathbb{L}^2$. Does the same property hold for a smooth ($\mathcal{C}^1$) function? What does this imply on the regularity of the paths of a Brownian motion W ?

(b) Prove the following convergence in $\mathbb{L}^2$ sense

$$\lim_{n \rightarrow \infty} \sum_{i=1}^n W_{\frac{i-1}{n}} (W_{\frac{i}{n}} - W_{\frac{i-1}{n}}) = \frac{W_1^2 - 1}{2}$$

(c) Prove that if f is smooth and bounded

$$f(W_t) = f(0) + \int_0^t f'(W_s) dW_s + \int_0^t \frac{f''(W_s)}{2} ds.$$

Solution 4.2

(a)

$$\begin{aligned}
\mathbb{E} \left[\left(\sum_{i=1}^n (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2 - 1 \right)^2 \right] &= \text{Var} \left(\sum_{i=1}^n (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2 \right) \\
&= \sum_{i=1}^n \text{Var}((W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2) \\
&= n(\mathbb{E}(W_{\frac{1}{n}}^4) - \frac{1}{n^2}) \\
&= n\left(\frac{3}{n^2} - \frac{1}{n^2}\right) \rightarrow 0 \quad \text{as } n \rightarrow \infty
\end{aligned}$$

The same property does not hold true for any $\mathcal{C}^1$ process X , since the element $Q_n(X)$ always converges to 0 (pathwise) as $n \rightarrow \infty$. This is a consequence of the fact that a $\mathcal{C}^1$ function is in particular of bounded variation on any interval as such $[0, 1]$. A notable consequence is that the Brownian motion is not BV (and therefore not $\mathcal{C}^1$) on any interval $[0, T]$.

(b) We notice first that

$$2W_{\frac{i-1}{n}}(W_{\frac{i}{n}} - W_{\frac{i-1}{n}}) = (W_{\frac{i}{n}} + W_{\frac{i-1}{n}})(W_{\frac{i}{n}} - W_{\frac{i-1}{n}}) - (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2 = W_{\frac{i}{n}}^2 - W_{\frac{i-1}{n}}^2 - (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2.$$

- (c) Ito's formula, here presented in its integral version, follows naturally from the properties of the Brownian motion outlined in the two previous points. In particular, point (b) implies directly that

$$W_1^2 = 1 + \int_0^1 2W_s dW_s,$$

which is Ito's formula for $f = (\cdot)^2$ and $t = 1$. The extension to any smooth f is a bit technical, but straightforward. Details can be found, for instance, in Theorem 3.3 of [1].

Exercise 4.3 (Black-Scholes model) Let $\sigma > 0$, $X_t = X_0 \exp\{\sigma W_t - \frac{\sigma^2 t}{2}\}$.

- (a) Prove that X is a solution of

$$dX_t = \sigma X_t dW_t.$$

- (b) Let $K > 0$, calculate

$$C_0 = \mathbb{E}[(X_T - K)_+].$$

- (c) Let $K > 0$, calculate

$$\frac{\partial}{\partial X_0} \mathbb{E}[(X_T - K)_+].$$

Solution 4.3

- (a) One can directly check that

$$\begin{aligned} dX_t &= d[X_0 \exp(\sigma W_t - \frac{\sigma^2 t}{2})] = X_0 d[\exp(\sigma W_t - \frac{\sigma^2 t}{2})] \\ &= X_0 \exp(\sigma W_t - \frac{\sigma^2 t}{2}) (\sigma dW_t + \frac{1}{2} \partial_t^2 W_t dt - \frac{\sigma^2 dt}{2}) \\ &= \sigma X_t dW_t. \end{aligned}$$

- (b) Applying Black-Scholes formula, we have

$$C_0 = X_0 \Phi(d_1) - K \Phi(d_2) \tag{7}$$

where

$$d_1 = \frac{\log(\frac{X_0}{K}) + \frac{\sigma^2}{2} T}{\sigma \sqrt{T}}, \quad d_2 = d_1 - \sigma \sqrt{T}.$$

Proving Black-Scholes formula is elementary, but requires some calculations. One has to notice first that $\sigma W_T - \frac{\sigma^2 T}{2} \sim \mathcal{N}\left(-\frac{\sigma^2 T}{2}, \sigma^2 T\right)$. Notably, X_T is a log-normal random variable. A similar variable admits an explicit probability density function $f_{X_T} = f_T$. Hence, we may rewrite

$$\mathbb{E}[(X_T - K)_+] = \int_{\mathbb{R}} (x - K)_+ f_T(x) dx = \int_K^\infty (x - K) f_T(x) dx.$$

The rest of the calculations is left as an exercise. It may be useful to recall that, if Φ is the cumulative distribution function of a standard Gaussian random variable, $\Phi(-x) = 1 - \Phi(x)$ for any $x \in \mathbb{R}$.

- (c)

$$\frac{\partial}{\partial X_0} \mathbb{E}[(X_T - K)_+] = \frac{\partial}{\partial X_0} (X_0 \Phi(d_1) - K \Phi(d_2)) = \phi(d_1)$$

Exercise 4.4 (Options' pricing in Black-Scholes and Heston models)

- (a) Code option pricing and simulation for European call options and Digital call options in a Black-Scholes model. See exercise notebook 1.
- (b) Compare simulations based on BS and Heston model. See exercise notebook 1.

Solution 4.4

- (a) See solution notebook 1.
- (b) See solution notebook 1.

References

- [1] Ioannis Karatzas and Steven E. Shreve. *Brownian Motion and Stochastic Calculus*. Springer, 1998.