# Mathematics for New Technologies in Finance

## Solution sheet 4

**Exercise 4.1 (Brownian motion)** For all  $N \in \mathbb{N}$ , we split [0, 1] into N many intervals with equal length  $\delta = \frac{1}{N}$ . We define a discrete stochastic processes  $W^N$  at  $\{0, \delta, 2\delta, \ldots, 1\}$  s.t.  $W_0^N = 0$  and

$$W_{n\delta+\delta}^N - W_{n\delta}^N = X_n, \quad \forall n = 1, \dots, N,$$

where  $(X_n)$  are i.i.d. random variable s.t.  $\mathbb{P}(X_n = \sqrt{\delta}) = \mathbb{P}(X_n = -\sqrt{\delta}) = \frac{1}{2}$ .  $W^N$  is a symmetric random walk with walking size  $\sqrt{\delta}$ .

(a) Prove that at  $t = n\delta$ .

$$\mathbb{E}[W_t^N] = 0, \quad \operatorname{Var}[W_t^N] = t. \tag{1}$$

(b) Prove that at each fixed  $t = n\delta$ , as  $n \to \infty$ .

$$W_t^N \to \mathcal{N}(0, t).$$
 (2)

- (c) What is the definition of Brownian motion?
- (d) Prove that Brownian motion is  $\frac{1}{2} \epsilon$  Hölder continuous a.s. for all  $\epsilon \in (0, \frac{1}{2})$ . (Hint: Borel-Cantelli Lemma)

#### Solution 4.1

(a)

$$\mathbb{E}[W_t^N] = \mathbb{E}[\sum_{i=1}^n X_i] = 0.$$
(3)

By independence

$$\operatorname{Var}[W_t^N] = \operatorname{Var}(\sum_{i=1}^n X_i) = \sum_{i=1}^n \operatorname{Var}(X_i) = n\delta = t.$$
 (4)

(b) By CLT

$$\lim_{n \to \infty} \frac{\sum_{i=1}^{n} X_i}{\sqrt{n}} \sim \mathcal{N}(0, t) \tag{5}$$

- (c) A equivalent definition to the standard one is the following: Let  $H = L^2((0, \infty), \mathcal{B}((0, \infty)), \lambda)$ a Hilbert space where  $\lambda$  is the Lesbesgue meansure. Denote by W the H-isonormal Gaussian process and let  $B_t = W(\mathbb{1}_{(0,t]})$ . Then  $(B_t)$  has the law as the Brownian motion and the a.s. continuous path can be proven by the standard Kolmogorov's continuity theorem.
- (d) Recall the Kolmogorov's continuity theorem (an application of Borel-Cantelli). If a real-valued stochastic process  $(X_t)$  has the property that for all K > 0, there exists positive  $\alpha, \beta, C$  such that for all  $0 \le t \le t + h \le K$ ,

$$\mathbb{E}[|X_{t+h} - X_t|^{\alpha}] \le Ch^{1+\beta}.$$
(6)

Then a.s.  $(X_t)$  is  $\gamma$ -Hölder-continuous for all  $\gamma < \frac{\beta}{\alpha}$ . Since for all  $k \in \mathbb{N}$  there exists C > 0 s.t.

$$\mathbb{E}[|B_{t+h} - B_t|^{2k}] \le Ch^{1+k}.$$
(7)

Therefore Brownian motion is  $\frac{1}{2} - \epsilon$  Hölder continuous a.s. for all  $\epsilon \in (0, \frac{1}{2})$ .

Updated: March 29, 2023

1/3

$$Q^{n}(W) = \sum_{i=1}^{n} (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^{2}.$$
(8)

- (a) Prove that  $Q^n(W)$  converges to 1 in  $L^2$
- (b) Prove the following convergence in  $L^2$  sense

$$\lim_{n \to \infty} \sum_{i=1}^{n} W_{\frac{i-1}{n}} (W_{\frac{i}{n}} - W_{\frac{i-1}{n}}) = \frac{W_1^2 - 1}{2}$$
(9)

(c) Prove that if f is smooth and bounded

$$f(W_t) = f(0) + \int_0^t f'(W_s) dW_s + \int_0^t \frac{f''(W_s)}{2} ds.$$
 (10)

#### Solution 4.2

(a)

$$\begin{split} \mathbb{E}\Big[\Big(\sum_{i=1}^{n} (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2 - 1\Big)^2\Big] &= \operatorname{Var}(\sum_{i=1}^{n} (W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2) \\ &= \sum_{i=1}^{n} \operatorname{Var}((W_{\frac{i}{n}} - W_{\frac{i-1}{n}})^2) \\ &= n(\mathbb{E}(W_{\frac{1}{n}}^4) - \frac{1}{n^2}) \\ &= n(\frac{3}{n^2} - \frac{1}{n^2}) \to 0 \quad \text{ as } n \to \infty \end{split}$$

(b) Combining (a) and the fact that

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

(c) By Ito's formula

$$df(W_t) = f'(W_t)dW_t + \frac{1}{2}f''(W_t)dt.$$

Then taking integral of both sides gives us the result.

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

Updated: March 29, 2023

### (a) Check by Ito's formula

(b)

$$C_0 = X_0 \Phi(d_1) - K \Phi(d_2) \tag{11}$$

where

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

We show details in exercise class.

(c)

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

We show details in exercise class.