

## Stochastic Calculus for Finance II Continuous-Time Models Chapter 3 Exercise

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✓ **Exercise 3.3 (Normal kurtosis).** The *kurtosis* of a random variable is defined to be the ratio of its fourth central moment to the square of its variance. For a normal random variable, the kurtosis is 3. This fact was used to obtain (3.4.7). This exercise verifies this fact.

Let  $X$  be a normal random variable with mean  $\mu$ , so that  $X - \mu$  has mean zero. Let the variance of  $X$ , which is also the variance of  $X - \mu$ , be  $\sigma^2$ . In (3.2.13), we computed the moment-generating function of  $X - \mu$  to be  $\varphi(u) = \mathbb{E}e^{u(X-\mu)} = e^{\frac{1}{2}u^2\sigma^2}$ , where  $u$  is a real variable. Differentiating this function with respect to  $u$ , we obtain

$$\varphi'(u) = \mathbb{E} \left[ (X - \mu)e^{u(X-\mu)} \right] = \sigma^2 u e^{\frac{1}{2}\sigma^2 u^2}$$

and, in particular,  $\varphi'(0) = \mathbb{E}(X - \mu) = 0$ . Differentiating again, we obtain

$$\varphi''(u) = \mathbb{E} \left[ (X - \mu)^2 e^{u(X-\mu)} \right] = (\sigma^2 + \sigma^4 u^2) e^{\frac{1}{2}\sigma^2 u^2}$$

and, in particular,  $\varphi''(0) = \mathbb{E}[(X - \mu)^2] = \sigma^2$ . Differentiate two more times and obtain the normal kurtosis formula  $\mathbb{E}[(X - \mu)^4] = 3\sigma^4$ .

**Ans.**

$$\varphi'''(u) = \mathbb{E}[(X - \mu)^3 e^{u(X-\mu)}] = (3\sigma^4 u + \sigma^6 u^3) e^{\frac{1}{2}\sigma^2 u^2}$$

$$\varphi''''(u) = \mathbb{E}[(X - \mu)^4 e^{u(X-\mu)}] = (3\sigma^4 + 3\sigma^6 u^2) e^{\frac{1}{2}\sigma^2 u^2} + \sigma^2 u (3\sigma^4 u + \sigma^6 u^3) e^{\frac{1}{2}\sigma^2 u^2}$$

$$\therefore \varphi''''(0) = \mathbb{E}[(X - \mu)^4 e^{0(X-\mu)}] = \mathbb{E}[(X - \mu)^4] = 3\sigma^4$$

Exercise 3.6 Let  $W(t)$  be a Brownian motion and let  $\mathcal{F}(t), t \geq 0$ , be an associated filtration.

(i) For  $\mu \in \mathbb{R}$ , consider the *Brownian motion with drift*  $\mu$ :

$$X(t) = \mu t + W(t).$$

Show that for any Borel-measurable function  $f(y)$ , and for any  $0 \leq s < t$ , the function

$$g(x) = \frac{1}{\sqrt{2\pi(t-s)}} \int_{-\infty}^{\infty} f(y) \exp \left\{ -\frac{(y-x-\mu(t-s))^2}{2(t-s)} \right\} dy$$

satisfies  $\mathbb{E}[f(X(t)) | \mathcal{F}(s)] = g(X(s))$ , and hence  $X$  has the Markov property. We may rewrite  $g(x)$  as  $g(x) = \int_{-\infty}^{\infty} f(y) p(\tau, x, y) dy$ , where  $\tau = t - s$  and

$$p(\tau, x, y) = \frac{1}{\sqrt{2\pi\tau}} \exp \left\{ -\frac{(y-x-\mu\tau)^2}{2\tau} \right\}$$

is the *transition density* for Brownian motion with drift  $\mu$ .

(ii) For  $\nu \in \mathbb{R}$  and  $\sigma > 0$ , consider the *geometric Brownian motion*

$$S(t) = S(0)e^{\sigma W(t) + \nu t}.$$

Set  $\tau = t - s$  and

$$p(\tau, x, y) = \frac{1}{\sigma y \sqrt{2\pi\tau}} \exp \left\{ -\frac{(\log \frac{y}{x} - \nu\tau)^2}{2\sigma^2\tau} \right\}.$$

Show that for any Borel-measurable function  $f(y)$  and for any  $0 \leq s < t$  the function  $g(x) = \int_0^{\infty} h(y) p(\tau, x, y) dy$  satisfies  $\mathbb{E}[f(S(t)) | \mathcal{F}(s)] = g(S(s))$  and hence  $S$  has the Markov property and  $p(\tau, x, y)$  is its transition density.

**Ans.**

(i)

$$\mathbb{E}[f(X(t)) | \mathcal{F}_s] = \mathbb{E}[f(\mu t + W(t)) | \mathcal{F}_s]$$

$$= \mathbb{E}[f(\mu(t-s) + W(t) - W(s) + X(s)) | \mathcal{F}_s]$$

$$\because \mu(t-s) + W(t) - W(s) \sim N(\mu(t-s), t-s)$$

$$\text{Let } y = \mu(t-s) + W(t) - W(s) + X(s)$$

$$y - X(s) \sim N(\mu(t-s), t-s)$$

$$\therefore \mathbb{E}[f(X(t)) | \mathcal{F}_s] = \int_{-\infty}^{\infty} f(y) p(\tau = t-s, X(s), y) dy$$

$$= \frac{1}{\sqrt{2\pi(t-s)}} \int_{-\infty}^{\infty} f(y) \exp \left\{ -\frac{(y-X(s)-\mu(t-s))^2}{2(t-s)} \right\} dy$$

$$= g(X(s))$$

$\therefore X$  has the Markov property.

(ii)

$$\because \log\left(\frac{S(t)}{S(s)}\right) = \sigma(W(t) - W(s)) + \nu(t - s), W(t) - W(s) \sim N(0, t - s)$$

$$\therefore \log\left(\frac{S(t)}{S(s)}\right) \sim N(\nu\tau, \sigma\tau), \tau = t - s$$

$$\text{Let } y = S(t), z = \log\left(\frac{y}{S(s)}\right) \Rightarrow \log\left(\frac{y}{S(s)}\right) \sim N(\nu\tau, \tau), dz = \frac{1}{y} dy$$

$$\therefore E[f(S(t)) | F_s] = \int_{-\infty}^{\infty} f(y) p(\tau, S(s), y) \frac{1}{y} dy$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sigma y \sqrt{2\pi\tau}} f(y) \exp\left\{-\frac{(\log(\frac{y}{S(s)}) - \nu\tau)^2}{2\sigma^2\tau}\right\} dy = g(S(s)) \text{ where } \tau = t - s$$

$\therefore S$  has the Markov property.

## 補充

### Exercise 1.

Let  $c > 0$  be a constant. Prove that

(1)  $X_t := W_{t+c} - W_c$  is a Brownian motion.

(2)  $X_t := \frac{1}{c}W_{c^2t}$  is a Brownian motion.

**Ans.**

(1)

$$1. X_0 = W_{0+c} - W_c = 0$$

$$2. X_{t+s} - X_t = (W_{t+s+c} - W_c) - (W_{t+c} - W_c) = W_{t+s+c} - W_{t+c} \sim N(0, s)$$

$$3. \forall t_1 \leq t_2 \leq t_3 \leq t_4$$

$$X_{t_4} - X_{t_3} = W_{t_4+c} - W_{t_3+c}, X_{t_2} - X_{t_1} = W_{t_2+c} - W_{t_1+c}$$

$\because t_1 + c \leq t_2 + c \leq t_3 + c \leq t_4 + c, W_t$  is a B.M

$$\therefore W_{t_4+c} - W_{t_3+c} \perp W_{t_2+c} - W_{t_1+c} \Rightarrow X_{t_4} - X_{t_3} \perp X_{t_2} - X_{t_1}$$

By 1.2.3 and definition 3.3.1,  $X_t$  is a B.M.

(2)

$$1. X_0 = \frac{1}{c}W_{c^2 \cdot 0} = \frac{1}{c}W_0 = 0$$

$$2. X_{t+s} - X_t = \frac{1}{c}(W_{c^2(t+s)} - W_{c^2t})$$

$$\because (W_{c^2(t+s)} - W_{c^2t}) \sim N(0, c^2(t+s) - c^2t) = N(0, c^2s) \therefore X_{t+s} - X_t \sim N(0, s)$$

$$3. \forall t_1 \leq t_2 \leq t_3 \leq t_4$$

$$X_{t_4} - X_{t_3} = \frac{1}{c}(W_{c^2t_4} - W_{c^2t_3}), X_{t_2} - X_{t_1} = \frac{1}{c}(W_{c^2t_2} - W_{c^2t_1})$$

$\because c^2t_1 \leq c^2t_2 \leq c^2t_3 \leq c^2t_4, W_t$  is a B.M

$$\therefore (W_{c^2t_4} - W_{c^2t_3}) \perp (W_{c^2t_2} - W_{c^2t_1})$$

$$\forall x, y \in \mathbb{R}, P(X_{t_4} - X_{t_3} \leq x; X_{t_2} - X_{t_1} \leq y) = P\left(\frac{1}{c}(W_{c^2t_4} - W_{c^2t_3}) \leq x; \frac{1}{c}(W_{c^2t_2} - W_{c^2t_1}) \leq y\right)$$

$$= P((W_{c^2t_4} - W_{c^2t_3}) \leq cx; (W_{c^2t_2} - W_{c^2t_1}) \leq cy)$$

$$= P((W_{c^2t_4} - W_{c^2t_3}) \leq cx)P((W_{c^2t_2} - W_{c^2t_1}) \leq cy) \quad \because (W_{c^2t_4} - W_{c^2t_3}) \perp (W_{c^2t_2} - W_{c^2t_1})$$

$$= P\left(\frac{1}{c}(W_{c^2t_4} - W_{c^2t_3}) \leq x\right)P\left(\frac{1}{c}(W_{c^2t_2} - W_{c^2t_1}) \leq y\right)$$

$$= P(X_{t_4} - X_{t_3} \leq x)P(X_{t_2} - X_{t_1} \leq y)$$

$$\therefore X_{t_4} - X_{t_3} \perp X_{t_2} - X_{t_1}$$

By 1.2.3 and definition 3.3.1,  $X_t$  is a B.M.

## Exercise2.

Check whether the following processes are martingales.

$$(1) X_t = W_t + 4t.$$

$$(2) X_t = W_t^2.$$

$$(3) X_t = W_t^3 - 3tW_t.$$

**Ans.**

(1)

$$E[X_t | F_s] = E[W_t + 4t | F_s], t > s$$

$$= E[(W_t - W_s) + 4(t-s) + X_s | F_s]$$

$$\because (W_t - W_s) \text{ and } (t-s) \text{ are independent of } F_s$$

$$= E[(W_t - W_s)] + 4(t-s) + X_s$$

$$= 4(t-s) + X_s > X_s$$

$\therefore X_t$  is not a martingale.

(2)

$$E[X_t | F_s] = E[(W_s + (W_t - W_s))^2 | F_s], t > s$$

$$= E[W_s^2 + 2W_s(W_t - W_s) + (W_t - W_s)^2 | F_s]$$

$$\because (W_t - W_s)^2 \text{ and } (W_t - W_s) \text{ are independent of } F_s$$

$$= X_s + 2W_s E[W_t - W_s] + E[(W_t - W_s)^2]$$

$$(\because (W_t - W_s) \sim N(0, t-s) \therefore E[(W_t - W_s)^2] = \text{var}[(W_t - W_s)] = t-s)$$

$$= (t+s) + X_s > X_s$$

$\therefore X_t$  is not a martingale.

(3)

$$E[X_t | F_s] = E[((W_t - W_s) + W_s)^3 - 3(t-s + s)((W_t - W_s) + W_s) | F_s], t > s$$

$$\text{Let } A = (W_t - W_s), \tau = t-s$$

$$E[X_t | F_s] = E[A^3 + 3A^2W_s + 3A(W_s^2) - 3\tau A - 3\tau W_s - 3sA + W_s^3 - 3sW_s | F_s]$$

$$\because \tau, A^3, A^2 \text{ and } A \text{ are independent of } F_s$$

$$= E[A^3] + 3E[A^2]W_s + 3E[A](W_s^2 - \tau - s) - 3\tau W_s + X_s = X_s$$

$\therefore X_t$  is a martingale.

$$\text{Hint: } A \sim N(0, t-s), \phi(u) = E[e^{uA}] = e^{\frac{1}{2}u^2(t-s)}, \phi'''(0) = E[A^3] = 0$$