

## Homework 2

1. Consider the **Bessel process** which satisfies

$$dR_t = \frac{2\delta - 1}{R_t} dt + dW_t$$

for  $\delta \geq 0, R_0 > 0$ . Using Ito's lemma, compute the SDE satisfied by  $Z_t = R_t^2$ .

**Solution.**

$$\begin{aligned} dZ_t &= 2R_t dR_t + \frac{1}{2} \cdot 2dt \\ &= 2[(2\delta - 1)dt + R_t dW_t] + dt \\ &= (4\delta - 1)dt + 2\sqrt{Z_t} dW_t. \end{aligned}$$

2. Consider a process  $X_t$  satisfying the SDE

$$dX_t = X_t^2 dW_t.$$

Compute the SDE for  $R_t = 1/X_t$  in terms of  $R_t$ .  $X$  is a rare example of a process which is driftless but it not an  $\mathcal{F}^W$ -martingale (in fact it can be shown that  $\mathbb{E}(X_t|X_s) < X_s$ , see FM04 for details).

**Solution:**

$$dR_t = -\frac{1}{X_t^2} dX_t + \frac{1}{2} \frac{2}{X_t^3} X_t^4 dt = -\frac{1}{X_t^2} X_t^2 dW_t + \frac{1}{2} \frac{2}{X_t^3} X_t^4 dt = -dW_t + \frac{1}{R_t} dt$$

3. Apply Ito's lemma to  $(1 - t/T)W_t$ , and integrate the resulting equation from  $t = 0$  to  $t = T$ . Use this to compute the distribution of  $\frac{1}{T} \int_0^T W_t dt$ .

**Solution.** Let  $f(x, t) = (1 - t/T)x$ . Then applying Ito's lemma to  $f(W_t, t)$ , we see that

$$df(W_t, t) = -\frac{1}{T} W_t dt + (1 - \frac{t}{T}) dW_t.$$

Integrating from 0 to  $T$ , we see that

$$f(W_T, T) - f(W_0, 0) = 0 = -\frac{1}{T} \int_0^T W_t dt + \int_0^T (1 - \frac{t}{T}) dW_t$$

so we see that the average of  $W$  over the interval  $[0, T]$  is given by  $\int_0^T (1 - \frac{t}{T}) dW_t$ . Moreover, since this is a stochastic integral of the form  $\int_0^T \phi(t) dW_t$ , where  $\phi$  is non-random,  $\int_0^T (1 - \frac{t}{T}) dW_t \sim N(0, \int_0^T \phi(t)^2 dt)$  (see part of lecture notes on the Ornstein-Uhlenbeck process) and when you evaluate the integral here, one finds that  $\int_0^T \phi(t)^2 dt = \frac{1}{3}T$ . This means that  $\text{Var}(\frac{1}{T} \int_{t=0}^T W_t dt)$ , i.e. the variance of the average of  $W$  over  $[0, T]$  is one-third the variance of  $W_T$  itself (which we know is  $T$ ).

4. Consider the following SDE

$$dR_t = (\frac{1}{R_t} - \frac{R_t}{1-t}) dt + dW_t$$

for  $t < 1$  with  $R_0 > 0$  (you may assume that  $R_t > 0$  for  $t < 1$ ). Compute an SDE for  $Y_t = R_t^2$ .  $R$  is known as the **Brownian excursion process**, which is Brownian motion conditioned to return to zero for the first time at time 1.

**Solution.** Let  $Y_t = R_t^2$ . Then

$$\begin{aligned} dY_t &= 2R_t dR_t + \frac{1}{2} \cdot 2dt = 2R_t((\frac{1}{R_t} - \frac{R_t}{1-t})dt + dW_t) + dt \\ &= 2R_t dW_t + 3dt - \frac{2R_t^2}{1-t} dt \\ &= (3 - \frac{2Y_t}{1-t})dt + 2\sqrt{Y_t} dW_t. \end{aligned}$$

5. Consider the following SDE

$$dS_t = \delta(\beta S_t + 1 - \beta)dW_t.$$

for  $\delta > 0$ . Derive the SDE satisfied by  $X_t = \beta S_t + 1 - \beta$ .  $S$  is known as a **displaced-diffusion** process.

**Solution.**

$$dX_t = d(\beta S_t + 1 - \beta) = \delta\beta X_t dW_t.$$

and we note that  $X$  is Geometric Brownian motion.  $S$  is often used to approximate the **CEV process**  $dS_t = \delta S_t^\beta dW_t$  for  $\beta \in (0, 1)$  when  $S_0 = 1$ , since  $S^\beta$  and  $\beta S + 1 - \beta$  have the same slope and value at  $S = 1$ .