

## Home Exercises for Chapter 5 in Klebaner's Book

Throughout this set of exercises  $B = \{B(t)\}_{t \geq 0}$  denotes Brownian motion.

**Task 1.** Find a diffusion type SDE that has a well-defined and unique strong solution, but that does not satisfy the conditions of Theorem 5.4 or Theorem 5.5 in Klebaner's book.

**Task 2.** Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be a two times continuously differentiable and strictly increasing function. Find a diffusion type SDE

$$dX(t) = \mu(X(t)) dt + \sigma(X(t)) dB(t) \quad \text{for } t > 0, \quad X(0) = f(0),$$

that has solution  $X(t) = f(B(t))$ .

**Task 3.** Show by means of direct calculation and/or inspection (not using Theorem 5.6 in Klebaner's book) that the solution given by Equation 5.13 in Klebaner's book to the Langevin equation 5.12 in Klebaner's book is a Markov process.

**Task 4.** Consider a stochastic processes  $\{X(t)\}_{t \geq 0}$  that is adapted to a filtration  $\{\mathcal{F}_t\}_{t \geq 0}$  and satisfies

$$\mathbf{E}\{|X(t)|\} < \infty \quad \text{for } t \geq 0 \quad \text{and} \quad \mathbf{E}\{X(t) | X(s)\} = X(s) \quad \text{for } 0 \leq s \leq t.$$

Which is the most restrictive of the further requirements that  $X$  is a martingale and that  $X$  is a Markov process?

**Task 5.** The solution  $\{X(t)\}_{t \geq 0}$  to the Langevin equation 5.12 in Klebaner's book is a Markov process with transition probability density function

$$p(t, x, y) = \frac{d}{dy} \mathbf{P}\{X(t+s) \leq y | X(s) = x\} = \frac{\sqrt{\alpha}}{\sqrt{\pi(1 - e^{-2\alpha t})} \sigma} \exp\left\{-\frac{\alpha(y - x e^{-\alpha t})^2}{\sigma^2(1 - e^{-2\alpha t})}\right\}$$

for  $s \geq 0, t > 0$  and  $x, y \in \mathbb{R}$ . (This can be verified, e.g., by means of use of Equation 5.13 in Klebaner's book.) Suppose that we know the value of the parameter  $\alpha > 0$ , but that we want to do a maximum likelihood estimation of the value of the parameter  $\sigma > 0$  using observations  $\{x_i\}_{i=0}^n$  of the process values  $\{X(i)\}_{i=0}^n$ . Find that maximum likelihood estimator.

**Task 1**

The simplest examples are ODE like  $dX(t) = -X(t)^2 dt$ ,  $X(0) = 1$  with solution  $X(t) = (1+t)^{-1}$ .  
A truly SDE example would be  $dX(t) = -X(t)^3 dt + dB(t)$ .

**Task 2**

$X(t) = f(B(t)) \Rightarrow dX(t) = f'(B(t))dB(t) + \frac{1}{2}f''(B(t))dt$   
 $= \frac{1}{2}f''(f^{-1}(X(t)))dt + f'(f^{-1}(X(t)))dB(t)$  so  $\mu(x) = \frac{1}{2}f''(f^{-1}(x))$ ,  $\sigma(x) = f'(f^{-1}(x))$

**Task 3**

$X(t) = e^{-\alpha t} (x_0 + \int_0^t e^{\alpha r} dB(r)) = e^{-\alpha(t-s)} e^{-\alpha s} (x_0 + \int_0^s e^{\alpha r} dB(r))$   
 $+ e^{-\alpha t} \int_s^t e^{\alpha r} dB(r) = e^{-\alpha(t-s)} X(s) + e^{-\alpha t} \int_s^t e^{\alpha r} dB(r)$ .

**Task 4**

$P(X(t) \in A | \mathcal{F}_s^X) = P(X(t) \in A | X(s)) \Rightarrow E(X(t) | \mathcal{F}_s^X) = E(X(t) | X(s))$   
 $= X(s)$  while  $E(X(t) | \mathcal{F}_s^X) = X(s) = E(X(t) | X(s))$  does not imply Markov so martingale is the less demanding additional assumption and Markov most restrictive.

**Task 5**

$$f_{X(1), \dots, X(n)}(x_1, \dots, x_n) = \prod_{i=1}^n p(1, x_{i-1}, x_i)$$

$$= \prod_{i=1}^n \frac{\sqrt{\alpha}}{\sqrt{\pi(1-e^{-2\alpha})} \sigma} \exp\left(-\frac{\alpha(x_i - e^{-\alpha} x_{i-1})^2}{\sigma^2(1-e^{-2\alpha})}\right)$$

$$\stackrel{\text{wrt } \sigma}{=} \frac{\alpha^{n/2}}{\pi^{n/2}(1-e^{-2\alpha})^{n/2} \sigma^n} \exp\left(-\frac{\alpha}{\sigma^2} \sum_{i=1}^n \frac{(x_i - e^{-\alpha} x_{i-1})^2}{(1-e^{-2\alpha})}\right)$$

with derivatives wrt  $\sigma$  given by

$$-\frac{n}{\sigma} + \frac{2\alpha}{\sigma^3} \sum_{i=1}^n \frac{(x_i - e^{-\alpha} x_{i-1})^2}{(1-e^{-2\alpha})} \quad \text{which is zero for}$$

$$\sigma = \hat{\sigma} = \sqrt{\frac{2\alpha}{n} \sum_{i=1}^n \frac{(x_i - e^{-\alpha} x_{i-1})^2}{(1-e^{-2\alpha})}}$$