

# Numerical Stochastic Differential Equations

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1. A stochastic process  $\{W(t), t \geq 0\}$  is a Wiener process if
  - 1)  $W(0) = 0$ , w.p. 1
  - 2)  $\{W(t), t \geq 0\}$  has independent and stationary increments.
  - 3)  $W(t) \sim N(0, \sigma^2 t)$ ,  $\forall t > 0$  (usually  $\sigma = 1$ )Show that the process  $X(t) = tW(\frac{1}{t})$  for  $t > 0$  and  $X(0) = 0$  is a Wiener process.

Since  $X(0) = 0$ , condition 1 is satisfied.

To investigate condition 2, take  $t_1 < t_2$ :

$$X(t_2) - X(t_1) = t_2 W(\frac{1}{t_2}) - t_1 W(\frac{1}{t_1})$$

Since  $W(t)$  is a Wiener process, so is  $W(\frac{1}{t})$  and it inherits the independent and stationary increments properties. Therefore  $X(t)$  has increments that do not depend on the past, but only on the difference  $t_2 - t_1$ , thus condition 2 is satisfied.

In order to satisfy condition 3, two statements need to be proved:

- 1)  $\mathbb{E}[X(t)] = 0$
- 2)  $Var[X(t)] = \sigma^2 t$

Since  $W(\frac{1}{t})$  is Wiener process,  $W(\frac{1}{t}) \sim N(0, \frac{\sigma^2}{t}) \Rightarrow \mathbb{E}[W(\frac{1}{t})] = 0, Var[W(\frac{1}{t})] = \frac{\sigma^2}{t}$ .

$$\mathbb{E}[X(t)] = \mathbb{E}[tW(\frac{1}{t})] = t\mathbb{E}[W(\frac{1}{t})] = 0$$

$$Var[X_t] = Var[tW(\frac{1}{t})] = t^2 Var[W(\frac{1}{t})] = t^2 \frac{\sigma^2}{t} = \sigma^2 t$$

Thus condition 3 is also satisfied, therefore  $X(t)$  is indeed a Wiener process.

2. Show that

$$\Phi_t = \cos(aW_t + b)$$

is a solution of the Itô SDE

$$d\Phi_t = -\frac{1}{2}a^2\Phi_t dt - a\sqrt{1 - \Phi_t^2}dW_t, \Phi_0 = \cos(b)$$

Itô's Differential Rule:

$$X_t = W_t \Rightarrow dX_t = dW_t, g = 1$$

Let  $\phi(x) = \cos(ax + b)$ :

$$\frac{\partial \phi}{\partial t} = 0, \frac{\partial \phi}{\partial x} = -a \sin(ax + b), \frac{\partial^2 \phi}{\partial x^2} = -a^2 \cos(ax + b)$$

Since  $d\Phi_t \stackrel{\text{Itô}}{=} \frac{\partial \phi}{\partial t} dt + \frac{\partial \phi}{\partial x} dX_t + \frac{1}{2}g^2 \frac{\partial^2 \phi}{\partial x^2} dt$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} -a \sin(aX_t + b) dX_t - \frac{1}{2}a^2 \cos(aX_t + b) dt$$

Substitute back  $X_t = W_t$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} -a \sin(aW_t + b) dW_t - \frac{1}{2}a^2 \cos(aW_t + b) dt$$

Since  $\sin x = \sqrt{1 - \cos^2 x}$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} -a\sqrt{1 - \cos^2(aW_t + b)}dW_t - \frac{1}{2}a^2\cos(aW_t + b)dt$$

Since  $\Phi_t = \cos(aW_t + b)$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} -\frac{1}{2}a^2\Phi_t dt - a\sqrt{1 - \Phi_t^2}dW_t$$

Since  $W_t$  - Wiener process  $\Rightarrow W_0 = 0$ :

$$\Phi_0 = \cos(aW_0 + b) = \cos(b)$$

3. Derive the Itô SDE for the process  $\Phi_t = W_t^4$ .

Itô's Differential Rule:

$$X_t = W_t \Rightarrow dX_t = dW_t, g = 1$$

Let  $\phi(x) = x^4$ :

$$\frac{\partial \phi}{\partial t} = 0, \frac{\partial \phi}{\partial x} = 4x^3, \frac{\partial^2 \phi}{\partial x^2} = 12x^2$$

Since  $d\Phi_t \stackrel{\text{Itô}}{=} \frac{\partial \phi}{\partial t} dt + \frac{\partial \phi}{\partial x} dX_t + \frac{1}{2}g^2 \frac{\partial^2 \phi}{\partial x^2} dt$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} 4X_t^3 dX_t + 6X_t^2 dt$$

Substitute back  $X_t = W_t$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} 6W_t^2 dt + 4W_t^3 dW_t$$

Since  $\Phi_t = W_t^4 \Rightarrow W_t = \Phi_t^{\frac{1}{4}}$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} 6\Phi_t^{\frac{1}{2}} dt + 4\Phi_t^{\frac{3}{4}} dW_t$$

Since  $W_t$  - Wiener process  $\Rightarrow W_0 = 0$ :

$$\Phi_0 = W_0^4 = 0$$

4. The stochastic process  $X_t$  satisfies the stochastic differential equation

$$dX_t = -\gamma X_t dt + X_t^3 dW_t$$

1) Calculate the differential equations for  $\mathbb{E}[X_t]$  and  $\mathbb{E}[X_t^2]$ .

2) Derive the differential equation for the variance  $V(X_t)$  and solve it.

1.1) Calculate the differential equation for  $\mathbb{E}[X_t]$ .

$$d\mathbb{E}[X_t] = \mathbb{E}[dX_t] = \mathbb{E}[-\gamma X_t dt + X_t^3 dW_t]$$

Since  $\mathbb{E}[dW_t] = 0$ :

$$d\mathbb{E}[X_t] = \mathbb{E}[-\gamma X_t dt] = -\gamma \mathbb{E}[X_t] dt$$

Differentiating with respect to  $t$ :

$$\frac{\partial}{\partial t} \mathbb{E}[X_t] = -\gamma \mathbb{E}[X_t]$$

1.2) Calculate the differential equation for  $\mathbb{E}[X_t^2]$ .

Itô's Differential Rule:

$$dX_t = -\gamma X_t dt + X_t^3 dW_t, g = X_t^3$$

Let  $\phi(x) = x^2$ :

$$\frac{\partial \phi}{\partial t} = 0, \frac{\partial \phi}{\partial x} = 2x, \frac{\partial^2 \phi}{\partial x^2} = 2$$

Since  $d\Phi_t \stackrel{\text{Itô}}{=} \frac{\partial \phi}{\partial t} dt + \frac{\partial \phi}{\partial x} dX_t + \frac{1}{2}g^2 \frac{\partial^2 \phi}{\partial x^2} dt$ :

$$d\Phi_t \stackrel{\text{Itô}}{=} 2dX_t + X_t^6 dt$$

Substituting back  $\Phi_t = X_t^2$  and  $dX_t = -\gamma X_t dt + X_t^3 dW_t$

$$dX_t^2 \stackrel{\text{Itô}}{=} -2\gamma X_t^2 dt - 2X_t^4 dW_t + X_t^6 dt$$

$$d\mathbb{E}[X_t^2] = \mathbb{E}[dX_t^2] = \mathbb{E}[-2\gamma X_t^2 dt - 2X_t^4 dW_t + X_t^6 dt]$$

Since  $\mathbb{E}[dW_t] = 0$ :

$$d\mathbb{E}[X_t^2] = \mathbb{E}[-2\gamma X_t^2 dt + X_t^6] = -2\gamma \mathbb{E}[X_t^2] dt + \mathbb{E}[X_t^6] dt$$

Differentiating with respect to  $t$ :

$$\frac{\partial}{\partial t} \mathbb{E}[X_t^2] = -2\gamma \mathbb{E}[X_t^2] + \mathbb{E}[X_t^6]$$

2.1) Derive the differential equation for the variance  $V(X_t)$ .

$$V(X_t) = \mathbb{E}[X_t^2] - (\mathbb{E}[X_t])^2$$

Differentiating with respect to  $t$ :

$$\frac{\partial}{\partial t} V(X_t) = \frac{\partial}{\partial t} (\mathbb{E}[X_t^2] - (\mathbb{E}[X_t])^2)$$

Applying chain rule:

$$\frac{\partial}{\partial t} V(X_t) = \frac{\partial}{\partial t} \mathbb{E}[X_t^2] - 2\mathbb{E}[X_t] \frac{\partial}{\partial t} \mathbb{E}[X_t]$$

Substituting with previously derived  $\frac{\partial}{\partial t} \mathbb{E}[X_t^2] = -2\gamma \mathbb{E}[X_t^2] + \mathbb{E}[X_t^6]$  and  $\frac{\partial}{\partial t} \mathbb{E}[X_t] = -\gamma \mathbb{E}[X_t]$ :

$$\frac{\partial}{\partial t} V(X_t) = -2\gamma \mathbb{E}[X_t^2] + \mathbb{E}[X_t^6] + 2\gamma (\mathbb{E}[X_t])^2$$

Substituting back  $V(X_t) = \mathbb{E}[X_t^2] - (\mathbb{E}[X_t])^2$ :

$$\frac{\partial}{\partial t} V(X_t) = -2\gamma V(X_t) + \mathbb{E}[X_t^6]$$

2.2) Solve the differential equation for the variance  $V(X_t)$ .

Let  $k = \mathbb{E}[X_t^6]$ :

$$\frac{\partial}{\partial t} V(X_t) = -2\gamma V(X_t) + k$$

Rearrange the terms:

$$\frac{\partial}{\partial t} V(X_t) + 2\gamma V(X_t) = k$$

Multiplying by  $e^{2\gamma t}$ :

$$e^{2\gamma t} \frac{\partial}{\partial t} V(X_t) + 2\gamma V(X_t) e^{2\gamma t} = k e^{2\gamma t}$$

Since  $\frac{\partial}{\partial t} (e^{2\gamma t} V(X_t)) = e^{2\gamma t} \frac{\partial}{\partial t} V(X_t) + 2\gamma V(X_t) e^{2\gamma t}$ :

$$\frac{\partial}{\partial t} (e^{2\gamma t} V(X_t)) = k e^{2\gamma t}$$

Integrating with respect to  $t$ :

$$e^{2\gamma t} V(X_t) = \frac{k}{2\gamma} e^{2\gamma t} + C$$

Dividing by  $e^{2\gamma t}$ :

$$V(X_t) = \frac{k}{2\gamma} + C e^{-2\gamma t}$$

Since  $V(X_0) = \frac{k}{2\gamma} + C \Rightarrow C = V(X_0) - \frac{k}{2\gamma}$ :

$$V(X_t) = \frac{k}{2\gamma} + (V(X_0) - \frac{k}{2\gamma}) e^{-2\gamma t}$$

5. Consider the deterministic differential equation for the process  $x(t)$ :

$$\frac{\partial x}{\partial t} = ax^2 - bx^3 + c \sin x, x(0) = 1, t \geq 0$$

where  $a, b, c$  are constants. Now assume that the parameter  $a$  is not fixed, but subject to fluctuations in time. So we replace  $a$  by the stochastic quantity  $A_t$  that can be described as stochastic process:

$$A_t = a + \xi(t),$$

where  $\xi(t)$  is a zero mean Gaussian white noise process. The fluctuations in  $A_t$  will make the trajectory of  $x(t)$  also noisy.

Derive the Itô SDE for the stochastic process  $X_t$ .

White noise process  $N_t$  in ODE = Stratonovitz SDE:

$$\frac{\partial x}{\partial t} \stackrel{\text{Str}}{=} (a + N_t)X_t^2 - bX_t^3 + c \sin(X_t) = aX_t^2 - bX_t^3 + c \sin(X_t) + X_t^2 N_t, X_0 = 1$$

Multiplying both sides by  $dt$  ( $N_t dt = W_t$ ):

$$dX_t \stackrel{\text{Str}}{=} (aX_t^2 - bX_t^3 + c \sin(X_t))dt + X_t^2 dW_t, X_0 = 1$$

Since  $g = X_t^2$ :

$$\frac{\partial g}{\partial x} = 2X_t$$

Itô correction term:

$$\frac{1}{2}g \frac{\partial g}{\partial x} dt = X_t^3 dt$$

Itô SDE = Stratonovitz SDE + Itô correction term:

$$dX_t \stackrel{\text{Itô}}{=} (aX_t^2 - bX_t^3 + c \sin(X_t) + X_t^3)dt + X_t^2 dW_t, X_0 = 1$$

6. Show by means of a counter example that  $\mathbb{E}[\int_{t_0}^t G_s dW_s]$  in the notes does not always hold for Stratonovich integrals.

Take Stratonovitz integral:

$$\int_{t_0}^t W_s dW_s \stackrel{\text{Str}}{=} \frac{W_s^2}{2} \Big|_{t_0}^t = \frac{W_t^2}{2} - \frac{W_{t_0}^2}{2}$$

Since  $\mathbb{E}[W_t^2] = t$ :

$$\mathbb{E}[\int_{t_0}^t W_s dW_s] \stackrel{\text{Str}}{=} \mathbb{E}[\frac{W_t^2}{2} - \frac{W_{t_0}^2}{2}] = \frac{1}{2}(t - t_0)$$

7. Derive the Stratonovitz SDE for the process  $\Phi_t = W_t^4$ .

Itô SDE:

$$d\Phi_t \stackrel{\text{Itô}}{=} 6\Phi_t^{\frac{1}{2}} dt + 4\Phi_t^{\frac{3}{4}} dW_t$$

Since  $g = 4\Phi_t^{\frac{3}{4}}$ :

$$\frac{\partial g}{\partial \phi} = 3\Phi_t^{-\frac{1}{4}}$$

Itô correction term:

$$\frac{1}{2}g \frac{\partial g}{\partial \phi} dt = 6\Phi_t^{\frac{1}{2}} dt$$

Stratonovitz SDE = Itô SDE - Itô correction term:

$$d\Phi_t \stackrel{\text{Str}}{=} 4\Phi_t^{\frac{3}{4}} dW_t$$

8. Consider the Itô SDE as a particle model for dissolved matter in a river:

$$dX_t = Udt + g dW_t, X_0 = 0, t \geq 0$$

$$dM_t = -aM_t dt, M_0 = 1, t \geq 0,$$

where  $X_t$  is the position of a particle and  $M_t$  its mass.  $U$  is the velocity of the water, and  $g$  and  $a$  are positive constants. Because of a chemical reaction, the dissolved matter slowly decays, and the mass of the particles decreases in time. Derive the partial differential equation for the probability  $p(x, m, t)$  at time  $t$  to find a particle at position  $x$ , with mass  $m$ .

$$\begin{aligned} d \begin{pmatrix} X_t \\ M_t \end{pmatrix} &\stackrel{\text{Itô}}{=} \begin{pmatrix} U \\ -aM_t \end{pmatrix} dt + \begin{pmatrix} g \\ 0 \end{pmatrix} dW_t, \begin{pmatrix} X_0 \\ M_0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ \Rightarrow \vec{f}(X_t, M_t, t) &= \begin{pmatrix} U \\ -aM_t \end{pmatrix}, \vec{g}(X_t, M_t, t) = \begin{pmatrix} g \\ 0 \end{pmatrix} \end{aligned}$$

Fokker-Planck equation  $\frac{\partial p}{\partial t} = -\sum \frac{\partial(f_i p)}{\partial x_i} + \frac{1}{2} \sum \sum \frac{\partial^2((gg^T)_{ij} p)}{\partial x_i \partial x_j}, t \geq t_0$ :

$$\begin{aligned} \frac{\partial p}{\partial t} &= -\frac{\partial}{\partial x}(Up) - \frac{\partial}{\partial m}(-amp) + \frac{1}{2} \frac{\partial^2}{\partial x^2}(g^2 p) = -U \frac{\partial p}{\partial x} + a \frac{\partial mp}{\partial m} + \frac{g^2}{2} \frac{\partial^2 p}{\partial x^2} \\ p(x, m, 0) &= \delta(x - x_0) \delta(m - m_0) = \delta(x) \delta(m - 1) \end{aligned}$$