

Stochastic Differential equations:

$$dX_t = b(X_t, t) \cdot dt + \sigma(X_t, t) \cdot dW_t$$

where W_t is the standard Wiener process

Example 1: Consider

$$dX_t = \alpha X_t dt + \beta X_t dW_t, \quad X_0 \text{ is given.}$$

Take $g(x, t) = \ln x$ and derive the SDE for $Y_t = g(X_t)$.

By Itô Formula:

$$\begin{aligned} dY_t &= \frac{\partial g}{\partial t} \cdot dt + \frac{\partial g}{\partial x} \cdot dX_t + \frac{1}{2} \frac{\partial^2 g}{\partial x^2} \cdot (dX_t)^2 \\ &= 0 + \frac{1}{X_t} \cdot (\alpha X_t dt + \beta X_t dW_t) + \frac{1}{2} \cdot \frac{-1}{X_t^2} \cdot [\beta^2 X_t^2 dt] \\ &= [\alpha - \frac{1}{2} \beta^2] dt + \beta dW_t \end{aligned}$$

Implying that $Y_t = Y_0 + (\alpha - \frac{1}{2} \beta^2)t + \beta W_t$ which in turn

implies $X_t = X_0 \cdot \exp[(\alpha - \frac{1}{2} \beta^2)t + \beta W_t]$.

Exercise: - Simulate this and argue about the role of α and β when you think of X_t as the value of an investment X_0 after t -time on an asset with uncertain return.

- Rigorously argue about different limits of X_t
 - in those cases $\alpha \geq \frac{1}{2} \beta^2$ by using

for the same ...

Lemma: (The law of iterated logarithm)

$$\limsup_{t \rightarrow \infty} \frac{W_t}{\sqrt{2t \cdot \log \log t}} = 1 \text{ a.s.}$$

Example 2: Consider $X_t = W_t$ and $g(t, x) = e^{ix} = \cos(x) + i \sin(x)$

if $Y_t = g(X_t) = \cos(W_t) + i \sin(W_t) \Rightarrow |Y_t| \equiv 1$

$$dY_t = -\sin(W_t) dW_t - \frac{1}{2} \cos(W_t) dt + i \left[\cos(W_t) dW_t - \frac{1}{2} \sin(W_t) dt \right]$$

$$dY_t = -\frac{1}{2} [\cos(W_t) + i \sin(W_t)] dt + [-\sin(W_t) + i \cos(W_t)] dW_t$$

$$dY_t = -\frac{1}{2} Y_t dt + i \overline{Y_t} dW_t \rightarrow \text{"Brownian motion on unit circle"}$$

Existence and Uniqueness of solutions:

Recall the two famous non-examples in ODEs:

1) non-existence: $\frac{dx_t}{dt} = x_t^2$, $x_0 = 1 \Rightarrow x_t = \frac{1}{1-t}$, $0 \leq t < 1$.

and as $t \rightarrow 1^-$, $x_t \rightarrow \infty$ (or doesn't exist for $t \geq 1$).

Issue: $x \mapsto x^2$ grows too fast (faster than $x \mapsto |x|$)

2) non-uniqueness: $\frac{dx_t}{dt} = 3x_t^{2/3}$, $x_0 = 0 \Rightarrow$

for any $a \geq 0$: $x_t = \begin{cases} 0 & \text{for } t \leq a \\ (t-a)^3 & \text{for } t > a \end{cases}$ is a solution! (it's not unique).

Issue: $x \mapsto x^3$ is not Lipschitz continuous

"Strong solutions" of SDEs:

Theorem 5.2.1. (Existence and uniqueness theorem for stochastic differential equations).

Let $T > 0$ and $b(\cdot, \cdot): [0, T] \times \mathbf{R}^n \rightarrow \mathbf{R}^n, \sigma(\cdot, \cdot): [0, T] \times \mathbf{R}^n \rightarrow \mathbf{R}^{n \times m}$ be measurable functions satisfying

$$|b(t, x)| + |\sigma(t, x)| \leq C(1 + |x|); \quad x \in \mathbf{R}^n, t \in [0, T] \quad (5.2.1)$$

for some constant C , (where $|\sigma|^2 = \sum |\sigma_{ij}|^2$) and such that

$$|b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq D|x - y|; \quad x, y \in \mathbf{R}^n, t \in [0, T] \quad (5.2.2)$$

for some constant D . Let Z be a random variable which is independent of the σ -algebra $\mathcal{F}_\infty^{(m)}$ generated by $B_s(\cdot), s \geq 0$ and such that

$$E[|Z|^2] < \infty.$$

Then the stochastic differential equation

$$dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t, \quad 0 \leq t \leq T, X_0 = Z \quad (5.2.3)$$

has a unique t -continuous solution $X_t(\omega)$ with the property that

$$X_t(\omega) \text{ is adapted to the filtration } \mathcal{F}_t^Z \text{ generated by } Z \text{ and } B_s(\cdot); s \leq t \quad (5.2.4)$$

and

$$E\left[\int_0^T |X_t|^2 dt\right] < \infty. \quad (5.2.5)$$

Note: Here uniqueness means if $X_t(\omega)$ and $\hat{X}_t(\omega)$ are two such solutions then $X_t(\omega) = \hat{X}_t(\omega) \quad \forall t \leq T, \text{ a.s.}$

Proof: (Uniqueness)

$$\text{suppose } \begin{cases} dX_t = b(X_t, t)dt + \sigma(X_t, t)dW_t, & X_0 = Z \\ d\hat{X}_t = b(\hat{X}_t, t)dt + \sigma(\hat{X}_t, t)dW_t, & \hat{X}_0 = \hat{Z} \end{cases}$$

Now, let's bound

$$\mathbb{E} [|X_t - \hat{X}_t|^2] \leq 3 \mathbb{E} [|Z - \hat{Z}|^2] + 3 \mathbb{E} \left[\left| \int_0^t b(X_s, s) - b(\hat{X}_s, s) ds \right|^2 \right] \\ + 3 \mathbb{E} \left[\left| \int_0^t [\sigma(X_s, s) - \sigma(\hat{X}_s, s)] dW_s \right|^2 \right]$$

by $\begin{cases} \text{Itô Isometry,} \\ \int_0^t a_s ds \leq t \int_0^t a_s^2 ds \\ \uparrow \\ (b_1 + \dots + b_n)^2 \leq n(b_1^2 + \dots + b_n^2) \end{cases}$

$$\leq 3 \mathbb{E} [|Z - \hat{Z}|^2] + 3t \mathbb{E} \left[\int_0^t |b(X_s, s) - b(\hat{X}_s, s)|^2 ds \right] \\ + 3 \mathbb{E} \left[\int_0^t |\sigma(X_s, s) - \sigma(\hat{X}_s, s)|^2 ds \right]$$

by Lipschitz assumption

Tonelli/Fubini

$$\leq 3 \mathbb{E} [|Z - \hat{Z}|^2] + 3(1+t) D^2 \mathbb{E} \left[\int_0^t |X_s - \hat{X}_s|^2 ds \right] \\ = 3 \mathbb{E} [|Z - \hat{Z}|^2] + 3(1+t) D^2 \int_0^t \mathbb{E} [|X_s - \hat{X}_s|^2] ds$$

Define $V_t := 3 \mathbb{E} [|X_t - \hat{X}_t|^2]$, $V_0 = 3 \mathbb{E} [|Z - \hat{Z}|^2]$, then

$$V_t \leq V_0 + (1+t) D^2 \int_0^t V_s ds, \quad 0 \leq t \leq T$$

So by Gronwall lemma: $\Rightarrow V_t \leq V_0 \exp((1+t) D^2 t)$, $0 \leq t \leq T$

Now, if $Z = \hat{Z}$ then $V_0 = 0 \Rightarrow V_t = 0, \forall t \geq 0 \Rightarrow$

Q is rational numbers $\Rightarrow \mathbb{P} \left[\bigcap_{t \in Q \cap [0, T]} \{ |X_t(\omega) - \hat{X}_t(\omega)| = 0 \} \right] = 1$

$t \mapsto |X_t(\omega) - \hat{X}_t(\omega)|$ is continuous $\Rightarrow \mathbb{P} \left[\bigcap_{t \in [0, T]} \{ |X_t(\omega) - \hat{X}_t(\omega)| = 0 \} \right] = 1$

$\Rightarrow X_t = \hat{X}_t \quad \forall t \leq T, \text{ a.s.}$

} general argument about const. path processes.

- For existence, read page 70 in Øksendal.

Discussion on the growth condition for SDE:

- ... that is uniform in t :

supp. $F(x,t)$ has global Lipschitz property ...

$$\exists k > 0 : |F(x,t) - F(x',t)| \leq k|x - x'| \quad \forall x, x', \forall t \in [0, T]$$

then
$$|F(x,t) - F(0,t)| \leq k|x| \quad \text{or}$$
$$|F(x,t)| \leq k|x| + \sup_{t \in [0, T]} |F(0,t)|$$

So, if $|F(0,t)|$ is bounded on $t \in [0, T]$ by $M > 0$ then

$$|F(x,t)| \leq \max(k, M)(1 + |x|) \quad \text{"the growth condition"}$$

Therefore, we can replace the growth condition in this theorem by

$$\exists M > 0 : \sup_{t \in [0, T]} (|b(0,t)| + |\sigma(0,t)|) < M.$$

"weak solution" of SDE :

- x_t is a "strong solution" of $dx_t = b(x_t, t)dt + \sigma(x_t, t)dW_t$
where x_t is $\tilde{F}_t^{z, W}$ -adapted for the given Wiener process W_t on
a given $(\Omega, \mathcal{F}, P)$.

- Now, only given functions $b(x, t)$ and $\sigma(x, t)$, we may find
 $(\tilde{x}_t, \tilde{W}_t, \{\mathcal{H}_t\})$ s.t. $d\tilde{x}_t = b(\tilde{x}_t, t)dt + \sigma(\tilde{x}_t, t)d\tilde{W}_t$
and such that $\tilde{x}_t$ is $\mathcal{H}_t$ -adapted and $\tilde{W}_t$ is a
 $\mathcal{H}_t$ -Wiener process on some $(\Omega, \mathcal{H}, P)$; i.e. $\tilde{W}_t$

is a Wiener process and $\tilde{W}_t$ is a $\mathcal{H}_t$ -martingale.

Then, we call $(\tilde{x}_t, \tilde{W}_t, \{\mathcal{H}_t\})$ a "weak solution".

Lemma: Any weak or strong solution under the same conditions of theorem is weakly unique; meaning that $\tilde{X}_t, \hat{X}_t$ have the same Law for all t , (or have identical finite-dimensional distributions).

Fact: 1- Any strong solution is a weak solution!
 2- There might exist weak solution where there is no strong solution. ex: $dX_t = \text{sgn}(X_t) dW_t$, $X_0 = 0$,
 with $\text{sgn}(x) = \begin{cases} +1 & x \geq 0 \\ -1 & x < 0 \end{cases}$. (see page 73 in Øksendal)
 $\rightarrow$ Note that sgn is not Lipschitz Continuous!

Example: $dX_t = A X_t dt + B dW_t$, where X_0 given, W_t is n -dim.

"Linear time-invariant LTI system"

Use Itô formula: $g(x, t) = \exp(-At) x : \mathbb{R}^n \times [0, \infty) \rightarrow \mathbb{R}^n$

$$Y_t = g(X_t, t) = \exp(-At) X_t \Rightarrow Y_0 = X_0$$

$$\text{and } dY_t = \left[-A \exp(-At) X_t + \frac{1}{2} x_0 \right] dt + \exp(-At) dX_t$$

$$\therefore \exp(-At) [-A X_t + dX_t] = \exp(-At) B dW_t$$

$$dY_t = \text{emil} \dots$$

$$\Rightarrow Y_t = Y_0 + \int_0^t \exp(-As) B dw_s$$

$$\Rightarrow X_t = \exp(At) X_0 + \int_0^t \exp(A(t-s)) B dw_s !$$

Exercise: solve $dX_t = (A X_t + H_t) dt + B dw_t$.

$\underbrace{H_t}_{\text{deterministic}}$

n-dim Ito Formula (compact form) : $Y_t = g(X_t, t)$

with $dX_t = b(X_t, t) dt + \sigma(X_t, t) dw_t$. Then

$$\begin{aligned} dY_t &= \frac{\partial g}{\partial t} \cdot dt + \nabla_x g^T dX_t + \frac{1}{2} dX_t^T \nabla_x^2 g dX_t \\ &= \left[\frac{\partial g}{\partial t} + \nabla_x g^T b + \frac{1}{2} \text{Tr}(\nabla_x^2 g dX_t dX_t^T) \right] dt + \nabla_x g^T \sigma dw_t \\ &= \left[\frac{\partial g}{\partial t} + \nabla_x g^T b + \frac{1}{2} \text{Tr}(\nabla_x^2 g \underbrace{\sigma dw_t dw_t^T \sigma^T}_{= I dt}) \right] dt + \nabla_x g^T \sigma dw_t \\ &= \left[\frac{\partial g}{\partial t} + \nabla_x g^T b + \frac{1}{2} \text{Tr}(\nabla_x^2 g \sigma \sigma^T) \right] dt + \nabla_x g^T \sigma dw_t . \end{aligned}$$
