

# Stochastic Differential Equations (SDE)

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The power of action.

SDE :  $dX_t = \beta(t, X_t) dt + \sigma(t, X_t) dW_t$

- $\beta(t, x)$ ,  $\sigma(t, x)$  are the "drift", "diffusion"
- $X_0 = x_0$  : initial condition
- solutions of such SDE's are called Ito diffusion.

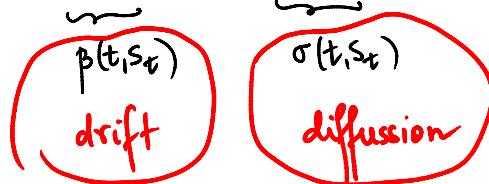
Problem : find  $X_T$  such that :

$$\begin{cases} X_0 = x_0 \\ X_T = x_0 + \int_0^T \beta(t, X_t) dt + \int_0^T \sigma(t, X_t) dW_t \end{cases}$$

• for suitable drift  $\beta(t, x)$ , and diffusion  $\sigma(t, x)$  coefficients we would like to answer questions concerning the existence, the uniqueness, and various other properties of the solution

example : • Generalized Geometric Brownian Motion :

SDE :  $ds_t = \underbrace{\alpha_s s_t}_\text{drift} dt + \underbrace{\sigma_s s_t}_\text{diffusion} dW_t$



$$\Rightarrow S_t = S_0 \exp \left\{ \int_0^t \sigma_s dW_s + \int_0^t \alpha_s - \frac{1}{2} \sigma_s^2 ds \right\}$$

is the solution of the above SDE

$t_0, t_1, \dots, t_n, t_{n+1}, \dots$

(can be verified using Ito formula)

• for the generalized GBM :  $\begin{cases} \beta(t, x) = \alpha_t \cdot x \\ \sigma(t, x) = \sigma_t \cdot x \end{cases}$  linear functions of  $x$

Linear Systems :  $\begin{cases} \beta(t, x) = A_t x + a_t \\ \sigma(t, x) = B_t x + b_t \end{cases}$

$$dX_t = (A_t X_t + a_t) dt + (B_t X_t + b_t) dW_t$$

→ this eq can also be solved explicitly

→ solution is a Gaussian process if  $X_0$  ~ normal distr. and  $X_0$  is independent of  $W$

→ recall the "Discounting" technique + Itô's lemma

$$D_t = \exp \left\{ - \int_0^t A_s ds \right\} : dD_t = - A_t D_t dt$$

$$\begin{aligned} d(D_t X_t) &= D_t dX_t + X_t dD_t + \underbrace{dX_t \cdot dD_t}_{=0} \\ &= D_t (A_t X_t + a_t) dt + D_t (B_t X_t + b_t) dW_t - A_t D_t X_t dt \\ &= D_t a_t dt + D_t (B_t X_t + b_t) dW_t \end{aligned}$$

① • in particular if  $B_t \equiv 0$  we can write the solution as:

$$D_t X_t = X_0 + \int_0^t D_s a_s ds + \int_0^t D_s b_s dW_s$$

$$\Rightarrow X_t = \frac{X_0}{D_t} + \frac{1}{D_t} \int_0^t D_u a_u du + \frac{1}{D_t} \int_0^t D_u b_u dW_u$$

- assuming  $X_0 \sim N(m, \sigma^2)$  independent of  $W$

$$\left\{
 \begin{aligned}
 & \cdot E X_t = \frac{1}{D_t} \left[ m + \int_0^t D_u a_u du \right] \\
 & \cdot \text{Cov}(X_s, X_t) = E[(X_t - EX_t) \cdot (X_s - EX_s)] = \\
 & = E \left[ \frac{1}{D_t} \left( X_0 - m + \int_0^t D_u b_u dW_u \right) \cdot \frac{1}{D_s} \left( X_0 - m + \int_0^s D_u b_u dW_u \right) \right] \\
 & = \frac{1}{D_t D_s} \text{Var}(X_0) + \frac{1}{D_t D_s} \left( \underbrace{E \left( \int_0^{s \wedge t} D_u b_u dW_u \right)^2}_{= \int_0^{s \wedge t} D_u^2 b_u^2 du} + \underbrace{E \left( \int_{s \wedge t}^t D_u b_u dW_u \right)}_{= 0} \right) \\
 & = \frac{1}{D_t D_s} \left[ \sigma^2 + \int_0^{s \wedge t} D_u^2 b_u^2 du \right]
 \end{aligned}
 \right.$$

- ② for  $A_t \equiv a_t \equiv b_t = 0 \rightarrow dX_t = B_t X_t dW_t$

solution of this eq is given by :

$$X_t = X_0 \exp \left\{ \int_0^t B_s dW_s - \frac{1}{2} \int_0^t B_s^2 ds \right\}$$

- ③ for  $A_t \equiv -\alpha < 0, b_t = b > 0, a_t = B_t = 0$

(Langevin equation)

$$dX_t = -\alpha X_t dt + b dW_t$$

this equation leads to the so called

Ornstein-Uhlenbeck process (Brownian Motion with proportional restoring force)

$$\rightsquigarrow d(e^{xt} X_t) = b e^{xt} dW_t$$

③ Hull-White interest rate model:

$$dR_t = (a_t - b_t R_t) dt + \sigma_t d\tilde{W}_t$$

( $\tilde{W}_t$ : Brownian motion under the risk neutral measure  $\tilde{\mathbb{P}}$ )

$a_t, b_t, \sigma_t$  are deterministic positive functions

$$\begin{cases} \beta(t, r) = a_t - b_t \cdot r \\ \sigma(t, r) = \sigma_t \end{cases}$$

• initial condition  $R_0 = r$

• to solve this eq, we can use the "discounting" once again!

$$D_t = e^{+\int_0^t b_s ds} : dD_t = b_t D_t dt$$

$$\begin{aligned} d(D_t R_t) &= D_t dR_t + R_t dD_t + \underbrace{dR_t \cdot dt}_{=0} \\ &= D_t (a_t - b_t R_t) dt + D_t \sigma_t d\tilde{W}_t + D_t b_t R_t dt \\ &= D_t a_t dt + D_t \sigma_t d\tilde{W}_t \end{aligned}$$

$$\Rightarrow D_t R_t = R_0 + \int_0^t D_u a_u du + \int_0^t D_u \sigma_u dW_u$$

⇒  $R_t$  ... interest rate function

④ Box - Ingersoll - Ross interest rate model

$$dR_t = (a - bR_t)dt + \sigma \sqrt{R_t} d\tilde{W}_t ; \quad R_0 = r$$

$a, b, \sigma$  : positive constants

- although there is no formula for  $R_t$ , there is a unique solution nonetheless !!

$\Rightarrow R_t$  can be approximated by Monte Carlo simulation

What does "suitable" conditions on  $\beta(t,x)$ ,  $\sigma(t,x)$  mean ?

Theorem: Suppose  $\beta(t,x)$ ,  $\sigma(t,x)$  satisfy the "Lipschitz" and "linear growth" conditions.

- (1) •  $|\beta(t,x) - \beta(t,y)| + |\sigma(t,x) - \sigma(t,y)| \leq k|x-y| \quad (\forall)x,y$
- (2) •  $|\beta(t,x)| + |\sigma(t,x)| \leq k(1+|x|) \quad (\forall)x$

for some  $k, K > 0$ . Then there exists a unique process  $X$  that satisfies :

$$X_t = X_0 + \int_0^t \beta(s, X_s) ds + \int_0^t \sigma(s, X_s) dW_s$$

•  $X_t$  has continuous paths, is adapted to the filtration  $\{\mathcal{F}_t\}$  of the Brownian Motion  $W$

- Condition (2) ensures the solution exists

(i.e. it does not "explode" :  $|X_t(\omega)| \rightarrow \infty$  in a finite time)

- Condition (1) guarantees uniqueness of the solution

Example :

- $dX_t = X_t^2 dt$ ,  $X_0 = 1$

$\rightsquigarrow$  corresponds to a drift :  $p(x) = x^2$   
(which does not satisfy condition (2))

$\rightsquigarrow$  the solution :  $X_t = \frac{1}{1-t}$  :  $0 \leq t < 1$

$\lim_{t \rightarrow 1^-} X_t = \infty$  :  $X_t$  explodes in finite time !!

- $dX_t = 3X_t^{2/3} dt$ ;  $X_0 = 0$

• this equation has more than 1 solution :

$$\rightsquigarrow X_t = \begin{cases} 0 & \text{for } t \leq a \\ (t-a)^3 & \text{for } t > a \end{cases} \quad \text{where } a > 0.$$

• in this case :  $b(x) = 3x^{2/3}$  does not satisfy the Lipschitz condition (1) at  $x=0$ .