

Consider the SDE

(6)

$$dS(t) = \mu(t) S(t) dt + \sigma(t) S(t) dW(t)$$

$$S(0) = S_0$$

$$\text{where } |\mu(t)| + |\sigma(t)| \leq K$$

$$\text{Here } b(t, x) = \mu(t)x \text{ and } \sigma(t, x) = \sigma(t)x$$

Both the conditions for uniqueness and existence are satisfied. Now by Ito-formula.

$$d(\ln S(t)) = \frac{1}{S(t)} dS(t) + \left(-\frac{1}{2S^2(t)}\right) dS(t) dS(t).$$

$$= \frac{1}{S(t)} \left[\mu(t) S(t) dt + \sigma(t) S(t) dW(t) \right]$$

$$- \frac{1}{2S^2(t)} \sigma^2(t) S^2(t) dt$$

$$= \sigma(t) dW(t) + \left(\mu(t) - \frac{1}{2} \sigma^2(t) \right) dt$$

$$\text{Thus } \ln S(T) - \ln S(0) = \int_0^T \sigma(t) dW(t) + \int_0^T \left(\mu(t) - \frac{1}{2} \sigma^2(t) \right) dt$$

$$\Rightarrow S(T) = S_0 \exp \left\{ \int_0^T \sigma(t) dW(t) + \int_0^T \left(\mu(t) - \frac{1}{2} \sigma^2(t) \right) dt \right\}.$$

consider a first order ordinary differential equation

$$\frac{dx(t)}{dt} = f(t)x(t) + g(t)$$

$x(0) = x$, where $f(t)$ is a continuous function

To solve this equation we multiply both side by Integrating factor $e^{-\int_0^t f(s) ds}$, we get

$$e^{-\int_0^t f(s) ds} \left[\frac{dx(t)}{dt} - f(t)x(t) \right] = g(t) e^{-\int_0^t f(s) ds}$$

Let $h(t) = e^{-\int_0^t f(s) ds}$, then

$$\frac{d}{dt}(h(t)x(t)) = h(t)g(t)$$

The solution of the above equation is

$$h(t)x(t) = x + \int_0^t h(s)g(s)ds.$$

$$\text{on } x(t) = x[h(t)]^{-1} + [h(t)]^{-1} \int_0^t h(s)g(s)ds.$$

By a Linear SDE, we mean a SDE of the form

$$\begin{cases} dx(t) = (\phi(t)x(t) + \theta(t))dw(t) + (f(t)x(t) + g(t))dt \\ x(0) = x. \end{cases}$$

$$H(t) = e^{-Y(t)}.$$

$$Y(t) = \int_0^t \phi(s)dw(s) + \int_0^t f(s)ds - \frac{1}{2} \int_0^t \phi^2(s)ds.$$

$$\text{Now (i) } d(H(t)x(t)) = x(t)dH(t) + H(t)dx(t) + dH(t)dx(t).$$

$$dH(t) = -e^{-Y(t)}dY(t) + \frac{1}{2}e^{-Y(t)}dY(t)dY(t).$$

$$= -H(t)[f(t)dt + \phi(t)dw(t) - \frac{1}{2}\phi^2(t)dt] \\ + \frac{1}{2}H(t)\phi^2(t)dt.$$

$$= -H(t)f(t)dt - H(t)\phi(t)dw(t) + H(t)\phi^2(t)dt.$$

$$\text{Thus } dH(t)dx(t) = -H(t)\phi(t)[\phi(t)x(t) + \theta(t)]dt.$$

Therefore we get.

$$d(x(t)H(t)) = H(t)[\cancel{-x(t)f(t)dt} - \cancel{x(t)\phi(t)dw(t)} + \cancel{x(t)\phi^2(t)dt} \\ + (\cancel{\phi(t)x(t)} + \theta(t))dw(t) + (\cancel{f(t)x(t)} + g(t))dt \\ - \cancel{\phi^2(t)x(t)dt} - \phi(t)\theta(t)dt].$$

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$$d(x(t)H(t)) = H(t) [\theta(t)dW(t) + g(t)dt - \theta(t)\phi(t)dt].$$

$$\Rightarrow H(t)x(t) = x + \int_0^t H(s)\theta(s)dW(s) + \int_0^t H(s)\{g(s) - \theta(s)\phi(s)\}ds$$

$$\text{Thus } x(t) = x e^{\gamma(t)} + \int_0^t e^{\gamma(t)-\gamma(s)} \theta(s) dW(s) + \int_0^t e^{\gamma(t)-\gamma(s)} \{g(s) - \theta(s)\phi(s)\} ds.$$

Example:-
$$\begin{cases} dx(t) = \mu x(t)dt + \sigma dW(t). \\ x(0) = x_0. \end{cases}$$

$$\text{Thus } f(t) = \mu, g(t) = 0, \phi(t) = 0, \theta(t) = \sigma$$

$$\text{Hence } \gamma(t) = \mu t.$$

Therefore:-
$$x(t) = x_0 e^{\mu t} + \int_0^t e^{\mu(t-s)} \sigma dW(s).$$

Example:- $P(x_t = \gamma_t) = 1$ but $P(x(t) = \gamma(t) : \forall t \geq 0) = 0.$

consider a positive random variable Z with a continuous distribution. put $x(t) \equiv 0$ and let

$$\gamma_t = \begin{cases} 0 & ; t \neq Z \\ 1 & ; t = Z \end{cases}$$

$$\text{we have } P(x(t) = \gamma(t)) = P(Z \neq t) = 1$$

$$\text{But } P(x(t) = \gamma(t) : \forall t \in [0, T]) = 0.$$

Let $\Omega = [0, T]$ and Z is a random variable uniformly distributed over the interval $[0, T]$.

$$\text{Define } \gamma_t(\omega) = \begin{cases} 1 & \text{if } t = \omega \\ 0 & \text{if } t \neq \omega. \end{cases}$$

(9)

consider the SDE

$$dx(t) = b(t, x(t))dt + \sigma(t, x(t))dw(t)$$

let $h(y)$ be a Borel-measurable function. Define

$$g(t, x) = \mathbb{E}[h(x(\tau)) | x(t) = x] := \mathbb{E}^{t, x}[h(x(\tau))]$$

Assume that $\mathbb{E}[h(x(\tau)) | x(t) = x] < \infty \forall t \in [0, T] \& x \in \mathbb{R}$

Theorem:- Let $x(u)$, $u \geq 0$, be a solution to the SDE with initial condition given at 0. Then for $0 \leq t \leq T$

$$\mathbb{E}[h(x(\tau)) | \mathcal{F}(t)] = g(t, x(t)).$$

Corollary:- solutions to stochastic differential equations are Markov processes

The following theorem relates SDE's and PDE's

Theorem:- (Feynman-Kac). Consider the stochastic differential equation

$$dx(t) = b(t, x(t))dt + \sigma(t, x(t))dw(t).$$

Let $h(y)$ be a Borel-measurable function. Define the function

$$g(t, x) = \mathbb{E}^{t, x}[h(x(\tau))]$$

Then $g(t, x)$ satisfies the partial differential equation

$$g_t(t, x) + b(t, x)g_x(t, x) + \frac{1}{2}\sigma^2(t, x)g_{xx}(t, x) = 0.$$

and the terminal condition

$$g(\tau, x) = h(x) \forall x.$$

Proof:- claim $g(t, x(t))$, $0 \leq t \leq T$ is a martingale.

By previous theorem

$$g(t, x(t)) = \mathbb{E}[h(x(T)) | \mathcal{F}(t)]$$

Thus for $s < t$

$$\begin{aligned} \mathbb{E}[g(t, x(t)) | \mathcal{F}(s)] &= \mathbb{E}[\mathbb{E}[h(x(T)) | \mathcal{F}(t)] | \mathcal{F}(s)] \\ &= \mathbb{E}[h(x(T)) | \mathcal{F}(s)] = g(s, x(s)). \end{aligned}$$

Hence the claim. Now

$$\begin{aligned} d(g(t, x(t))) &= g_t(t, x(t)) dt + g_x(t, x(t)) dx(t) \\ &\quad + \frac{1}{2} g_{xx}(t, x(t)) dx(t) dx(t). \\ &= g_t(t, x(t)) dt + g_x(t, x(t)) b(t, x(t)) dt + g_x(t, x(t)) \sigma(t, x(t)) dw(t) \\ &\quad + \frac{1}{2} g_{xx}(t, x(t)) \sigma^2(t, x(t)) dt \\ &= [g_t(t, x(t)) + b(t, x(t)) g_x(t, x(t)) + \frac{1}{2} g_{xx}(t, x(t)) \sigma^2(t, x(t))] dt \\ &\quad + g_x(t, x(t)) \sigma(t, x(t)) dw(t). \end{aligned}$$

since $g(t, x(t))$ is a martingale so the dt term must be equal to 0. Thus we must have

$$g_t(t, x) + g_x(t, x) b(t, x) + \frac{1}{2} \sigma^2(t, x) g_{xx}(t, x) = 0.$$

Theorem:- consider the stochastic differential equation

$$dx(t) = b(t, x(t)) dt + \sigma(t, x(t)) dw(t).$$

Let $h(y)$ be a Borel-measurable function. Define

$$f(t, x) = \mathbb{E}^{t, x}[e^{r(T-t)} h(x(T))]$$

where r is a constant. Then $f(t, x)$ satisfies the partial differential equation $f_t(t, x) + b(t, x) f_x(t, x) + \frac{1}{2} \sigma^2(t, x) f_{xx}(t, x) = r f(t, x)$ with $f(T, x) = h(x)$.