

Ito Calc & SDE ...

Initial example of stock price -

$$S_t = S_0 e^{(r - \frac{1}{2}\sigma^2)t + \sigma W_t}$$

where W_t = Brownian motion.

* Model for stock w/ ~~independent~~ forcing independent on disjoint intervals.

SDE:

$$dS_t = r S_t dt + \sigma S_t dW_t.$$

Eg - suppose the stock pays $\frac{\delta}{m} S_t$

i.e. a dividend @ rated its present value

m times per year for large m say $m \geq 12$

then the SDE is

$$dS_t = r S_t dt - \delta S_t dt + \sigma S_t dW_t$$

↳

$$\cancel{S_t = S_0 e^{(r-\delta)t + \sigma W_t}} \quad S_t = S_0 e^{(r-\delta - \frac{1}{2}\sigma^2)t + \sigma W_t}$$

Solving SDE -

First, recall solving ODEs - find $x(t)$ for the o.d.e.

$$\frac{dx}{dt} = p(t) x(t) + q(t) ; x(0) = x_0$$

Integrating factor $g(t) = e^{-\int_0^t p(s) ds}$

then find - (write $p_t = p(t)$ etc)

$$\frac{d}{dt}(xg) = \left(\frac{dx}{dt}\right)g + \left(\frac{dg}{dt}\right)x$$

$$= (p_t x_t + q_t) g_t + x_t g_t (-p_t) = q_t g_t$$

\therefore Fundamental theorem of Calc \Rightarrow

$$x_t g_t = x_0 g_0 + \int_0^t g(s) q_s ds$$

~~$$x_t = x_0 + \int_0^t q_s e^{-\int_0^s p(u) du} ds$$~~

$$\hookrightarrow x_t = x_0 e^{\int_0^t p_u du} + \int_0^t q_s e^{\int_s^t p_u du} ds$$

Now instead of deterministic perturbation

- let us add nasty perturbation

- additively.

$$dx = p_t x_t dt + q_t dW_t.$$

Same integrating factor $g_t = e^{-\int_0^t p_s ds}$

$$d(X_t g_t) = (dx_t) g_t + x_t dg_t$$

$$= g_t (p_t x_t dt + q_t dW_t) + x_t g_t (-p_t) dt$$

$$= g_t q_t dW_t$$

$$\Leftrightarrow X_t g_t = X_0 + \int_0^t q_s g_s dW_s$$

$$X_t = X_0 e^{\int_0^t p_s ds} + e^{\int_0^t p_s ds} \int_0^t q_s g_s dW_s$$

Note $E \int_0^t q_s g_s dW_s = 0 \rightarrow E \int_0^t q_s g_s dW_s = 0$

$$\therefore E X_t = X_0 e^{\int_0^t p_s ds}$$

~~Important~~

Eg Ornstein Uhlenbeck

Security having value μ
Model for a noisy ~~process~~ ~ or time dependent variance

$$dX_t = \mu(k(X_0 - X_t)) dt + \sigma dW_t$$

$k, \mu, \sigma > 0$

try integrating factor ~~e~~ e^{kt}

$$\begin{aligned} d(X_t e^{kt}) &= e^{kt} dX_t + k e^{kt} X_t dt \\ &= (\cancel{\mu k} dt + \cancel{k} dW_t) e^{kt} \end{aligned}$$

$$\hookrightarrow \cancel{X_t e^{kt}} = X_0 + \int_0^t \cancel{\mu k} ds$$

$$X_t e^{kt} = X_0 + \mu \int_0^t e^{ks} ds + \sigma \int_0^t e^{ks} dW_s$$

$$X_t = X_0 e^{-kt} + \mu \{e^{-kt} - 1\} e^{kt} + \sigma \int_0^t e^{-k(t-s)} dW_s$$

$$= \underbrace{(X_0 - 1)}_{\text{initial position}} e^{-kt} + \underbrace{\mu}_{\text{attraction point}} + \underbrace{\sigma \int_0^t e^{-k(t-s)} dW_s}_{\substack{\text{as } t \rightarrow s \\ \text{the noise at timesegment } s \text{ is } dW_s}}$$

decays exponentially

a problem for Finance = value X_t may be < 0 !

HESTON MODEL ~

MODEL for stochastic variance

V_t = variance @ time t

the stock price is

$$dS_t = r S_t dt + \sqrt{V_t} S_t dW_t^{(1)}$$

suppose V_t satisfies

$$dV_t = \kappa (\theta - V_t) dt + \tau \sqrt{V_t} dW_t^{(2)}$$

$W^{(1)}, W^{(2)}$ are correlated B.M.

$$\lim_{t \rightarrow 0} \frac{1}{t} \mathbb{E} (W_t^{(2)} \cdot W_t^{(1)}) = \rho \equiv \text{correlation}$$

→ this is also written as

$$\mathbb{E} (dW^{(1)} dW^{(2)}) = \rho dt.$$

* Variance is always positive.

* $\rho \equiv \text{correlation} \rightarrow$ shocks to volatility are correlated to shocks in stock prices

→ Sharp increase in stock price corresponds to sharp increase in volatility.

* No closed form sol \square but we can write down sol \square in terms of P.D.E.