

Assignment 5
Girsanov's Theorem & Solving SDEs

1. Solve the following SDE:

$$dx_t = (\beta - \log x_t) x_t dt + \sigma x_t dB_t,$$

$$X(0) = x_0, \quad \beta \text{ \& \; } \sigma \text{ are constants.}$$

→ let's try transforming the SDE using Itô's lemma
We assume $x_t > 0$ for all t because $\log x_t$ must be defined.

let $Y_t = \log x_t$, by Itô's lemma

$$dY_t = 0 \cdot dt + \frac{1}{x_t} dx_t - \frac{1}{2} \frac{1}{(x_t)^2} \cdot dx_t \cdot dx_t \quad - (1)$$

but given,

$$dx_t = (\beta - \log x_t) x_t dt + \sigma x_t dB_t$$

$$\frac{dx_t}{x_t} = (\beta - \log x_t) dt + \sigma dB_t$$

$$\& \ dx_t \cdot dx_t = \sigma^2 (x_t)^2 dt \quad (\because dt \cdot dt = 0, dB_t \cdot dt = 0)$$

$$\therefore \frac{dx_t \cdot dx_t}{(x_t)^2} = \sigma^2 dt$$

\therefore eqⁿ (1) becomes

$$dY_t = (\beta - \log x_t) dt + \sigma dB_t - \frac{1}{2} \sigma^2 dt$$

$$\therefore dy_t = (\beta - \frac{1}{2}\sigma^2 - y_t)dt + \sigma dB_t$$

but this e_t^n is in the form of Ornstein-Uhlenbeck process

$$[dx_t = \theta(\mu - x_t)dt + \sigma dw_t]$$

Solution to it is $x_t = x_0 e^{-\theta t} + \mu(1 - e^{-\theta t}) + \frac{\sigma}{\sqrt{2\theta}} W_{1 - e^{-2\theta t}}$

$$\therefore Y_t = e^{-t} Y_0 + (\beta - \frac{1}{2}\sigma^2)(1 - e^{-t})$$

$$+ \sigma \int_0^t e^{-(t-s)} dB_s$$

but $Y_t = \log X_t$

$$\therefore X_t = e^{Y_t}$$

$$\therefore X_t = \exp \left\{ e^{-t} \log X_0 + (1 - e^{-t}) \left(\beta - \frac{1}{2}\sigma^2 \right) + \sigma \int_0^t e^{-(t-s)} dB_s \right\}$$

2. Let $B_t = (B_t^1, B_t^2, \dots, B_t^d)$ be a d -dimensional Brownian motion and

$$R_t = \|B_t\| = \sqrt{(B_t^1)^2 + \dots + (B_t^d)^2}.$$

What is the distribution of the process

$$X_t = \sum_{i=1}^d \int_0^t \frac{B_s^i}{R_s} dB_s^i$$

→ The integrand in each term of the sum, $\frac{B_s^i}{R_s}$, represents the i -th component of

the brownian motion divided by the Radial distance from the origin in d -dimensional space. This can be seen as the projection of the d -dimensional Brownian motion on the unit sphere scaled by R_s .

* Quadratic variation of X_t .

$$\langle X_t, X_t \rangle = \left\langle \sum_{i=1}^d \int_0^t \frac{B_s^i}{R_s} dB_s^i, \sum_{i=1}^d \int_0^t \frac{B_s^i}{R_s} dB_s^i \right\rangle$$

Since the integrals are stochastic integrals with respect to orthogonal components (independent B_s^i 's) of the brownian motion, the quadratic variation add up to

$$\langle X, X \rangle_t = \sum_{i=1}^d \left\langle \int_0^t \frac{B_s^i}{R_s} dB_s^i, \int_0^t \frac{B_s^i}{R_s} dB_s^i \right\rangle$$

$$\langle X, X \rangle_t = \sum_{i=1}^d \int_0^t \left(\frac{B_s^i}{R_s} \right)^2 ds = \int_0^t \sum_{i=1}^d \left(\frac{B_s^i}{R_s} \right)^2 ds$$

Notice that $\left(\frac{B_s^i}{R_s} \right)^2$ is the squared component of the unit vector pointing in the direction of B_s .

The sum of the squares of these components is 1 (radius of \uparrow unit sphere)
d-dimensional

So,

$$\langle X, X \rangle = \int_0^t 1 ds = t \quad \text{--- (1)}$$

Also: $X_0 = 0$ as $B_0 = 0 \forall i$ s --- (2)

each term $\left(\frac{B_s^i}{R_s} \right)$ is continuous (as brownian motion) & X_t is a continuous function of brownian motions. Hence it is also continuous. X_t has continuous paths by construction (3)

Also each term in $\sum_{i=1}^d \int_0^t \left(\frac{B_s^i}{R_s} \right)^2 ds$ is

an ~~ito~~ (stochastic) integral which are

martingales (Property) --- (4)

From (1), (2), (3), (4)

X_t is a continuous local martingale starting at 0 with quadratic variation t , by Levy's theorem, X_t is a standard Brownian motion.

Therefore distribution of X_t is Normal i.e. $X_t \sim N(0, t)$.

3. Find a stochastic differential equation that R_t satisfies.

→ To find a stochastic differential eqⁿ that $R_t = \|B_t\| = \sqrt{(B_t^1)^2 + \dots + (B_t^d)^2}$ satisfies, where,

$B_t = (B_t^1, B_t^2, \dots, B_t^d)$ is a d -dimensional Brownian motion, we can use Itô's formula. The process R_t can be thought as the radial part of a multidimensional Brownian motion, and its dynamics can be analyzed by considering the square root of the sum of squares of the components of B_t .

1) We define $R_t = f(B_t)$ where $f(x) = \sqrt{x_1^2 + \dots + x_d^2}$

Using Itô formula for multidimensional case on $f(x_t)$

$$df(x_t) = \sum_{i=1}^d \frac{\partial f}{\partial x_i} dx_t^i + \frac{1}{2} \sum_{i,j=1}^d \frac{\partial^2 f}{\partial x_i \partial x_j} d\langle x^i, x^j \rangle_t$$

For R_t , $x_t^i = B_t^i$ & $d\langle B^i, B^j \rangle_t = \delta_{ij} dt$
(where δ_{ij} is the Kronecker delta).

2) The first & second partial derivatives of f are

$$\frac{\partial f}{\partial x_i} = \frac{x_i}{\sqrt{x_1^2 + \dots + x_d^2}} = \frac{B_t^i}{R_t}$$

$$\frac{\partial^2 f}{\partial x_i^2} = \frac{(R_t)^2 - (B_t^i)^2}{(R_t)^3}$$

The mixed partial derivatives for $i \neq j$

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = - \frac{B_t^i B_t^j}{R_t^3}$$

3. Applying itô's formula

$$dR_t = \sum_{i=1}^d \frac{B_t^i}{R_t} dB_t^i + \frac{1}{2} \sum_{i=1}^d \frac{(R_t)^2 - (B_t^i)^2}{(R_t)^3} dt,$$

As $\sum_{i=1}^d (B_t^i)^2 = R_t^2$ (Radius of d -dimensional Sphere)

$$\therefore dR_t = \sum_{i=1}^d \frac{B_t^i}{R_t} dB_t^i + \frac{1}{2} \left(d - \frac{R_t^2}{R_t^2} \right) \frac{dt}{R_t}$$

$$\therefore dR_t = \sum_{i=1}^d \frac{B_t^i}{R_t} dB_t^i + \frac{d-1}{2} \frac{dR}{R_t}$$

Since $\sum_{i=1}^d \frac{B_t^i}{R_t} dB_t^i$ can be seen as the stochastic

integral of a unit vector in the direction of B_t (normalized by R_t) with respect to B_t , it can be represented by as a standard brownian motion.

$$\therefore dR_t = dW_t + \frac{d-1}{2} \cdot \frac{dt}{R_t}$$

This SDE reflects the behaviour of the radial part of a d -dimensional Brownian motion.

The term $\frac{d-1}{2} \cdot \frac{dt}{R_t}$ arises from the curvature

effects in higher dimensions, effectively a mean-reverting force proportional to the inverse of the distance from origin.