



The Moment for Some quotient Stochastic Differential Equation with Application

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ABSTRACT

In this article we use the product stochastic differential equations in order to study the solution for some quotient stochastic differential equation by using itô's formula, Then we find their moments (mean, variance and the k.th moments). Also we gave some examples to explain the method.

1. Introduction

The main definition of the Stochastic differential equations (simply SDE's) is that differential equations in which one or more of its terms are stochastic (random) processes, for which their solutions may be stochastic process, [Arnold, 1974. The Stochastic differential equations (SDEs) used in many field of science such as biology, chemistry, climatology, mechanics, physics, economics and finance. Many researcheres have given their contribution in these field (Akinbo B.J., et, al. (2015)), Guangqiang LAN et, al. (2014) derive the new sufficient conditions of existence, moment of the solution of stochastic differential equation, J.C. Jimenez (2015) uses the explicit formulas for the mean and variance of the solutions of linear stochastic differential equation, Platens [5] study the strong and weak approximation methods for the numerical methods to get the solution of stochastic differential equations, Nayak and Chakraverty [6] worked on numerical solution of fuzzy stochastic differential equation. Christios H.skiadas, [7] Study the exact solution of stochastic differential equation (Gomertz, Generalized logistic and revised exponential. Akinbo B.J. et al [2] study numerical solution of stochastic differential equation, and so on.

In this paper we study some form of stochastic differential equation as a quotient stochastic differential equation, then we explain how to apply itô's integral formula to find the solution of those equations and we find the moments of their solutions.

2. Preliminaries and method

Definition 1 :(random variable) [5]

A random variable is a mapping or a function from the sample space Ω onto the real line \mathbb{R} , (i.e. $X: \Omega \rightarrow \mathbb{R}$)

Definition 2 :(Expectation) or (mean) of a random variable:[5]

Let X is a random variable defined on the probability space (Ω, \mathcal{F}, P) , then the expected values or the mean of X is:

$$E(X) = \mu = \sum_i x_i p(x_i).$$

That is the average of X over the entire probability space.

For a random variable continuous over \mathbb{R} :

$$E(X) = \int_{-\infty}^{\infty} xf(x)dx$$

Definition 3: (Variance):[5]

The Variance is a measure of the spread of data about the mean μ $\text{Var}(X) = E((X - \mu)^2) = E(X^2) - (E(X))^2$

Definition 4: The kth -order moment:[5]

The k th -order moment of a continuous random variable is defined by:

$$E(\mathbf{X}^k) = \int_{-\infty}^{\infty} \mathbf{x}^k f(\mathbf{x}) d\mathbf{x}$$

Where $f(x)$ is the probability density function

Or $E(\mathbf{x}^k) = \sum_i \mathbf{x}_i^k p(\mathbf{x}_i)$; (For discrete time and $p(\mathbf{x}_i)$ is probability mass function)

Definition 5: (stochastic process) [1]

A stochastic process is a family of random variables denoted by $\{x(t), t \in T\}$ where t is time parameter and $T \in R$.

Definition 6: (Wiener process) [1]

A wiener process (Brownian motion) over $[0, T]$ denoted by $\{w(t)\}$ is a continuous-time stochastic process satisfying:

- 1: $W(0) = 0$
- 2: For all $t, s \geq 0$, $W(t) - W(s)$ is normally distributed with mean zero and variance $|t - s|$.
- 3: The increment's $W(t) - W(s)$ and $W(v) - W(u)$ are independent.

Definition 7 : (Itô –formula)[1]

Let $X(\cdot)$ be a real-valued stochastic process which satisfying

$$x(a) = x(b) + \int_a^b F dt + \int_a^b G dw \dots (1)$$

For some $G \in L^2(0, T)$, $F \in L^1(0, T)$ and $0 < a < b < T$. Then we say that $X(\cdot)$ has a stochastic differential equation

$$dX = Fdt + Gdw; \text{ for } 0 < t < T \dots (2)$$

Remark. [3]

$L^1[0, T]$, $L^2[0, T]$ denotes the space of all real-valued, adaptive processes $\{x_t\}$, $\{y_t\}$ respectively, such that

$$E\left(\int_0^T |x_t| dt\right) < \infty$$

$$E\left(\int_0^T |y_t| dt\right) < \infty$$

If $u : R \times [0, T] \rightarrow R$ is continuous and their first and second partial derivative for t exist and are continuous.

If we take $Y(t) = u(x(t), t)$, then we have the following Itô formula:

$$dY = \left(\frac{\partial u}{\partial t} + F \frac{\partial u}{\partial x} + \frac{1}{2} \frac{\partial^2 u}{\partial x^2} G^2\right) dt + \frac{\partial u}{\partial x} G dw$$

Theorem : (1) [1]

Let $u(x) = x^m$, $m = 0, 1, 2, \dots$ then

$$d(x^m) = mx^{m-1} dx + \frac{1}{2} m(m-1) x^{m-2} G^2 dt \dots (3)$$

see [1]

Lemma (1): [2]

Let w_t is a Brownian motion then, by Itô's Formula, we have:

$$(dw_t)^2 = dt, (dt)(dw_t) = 0 \text{ and } (dt)^2 = 0. \dots (4)$$

Lemma(2): [4]

Suppose $\{w_t\}$ is a Brownian motion then, by using Ito's Formula, we get:

$$dw_t^2 = 2w_t dw_t + dt, (dw_t)^2 = dt, dt dw_t = 0 \text{ and } dw_t^3 = 3w_t^2 dw_t + 3w_t dt \dots (5)$$

Theorem (2): (Itô product rule) [1]

Let $dx_i = \alpha_i(t)dt + \beta_i(t)dW(t)$; ($i = 1, 2$)

($0 \leq t \leq T$): $\alpha_i(t) \in L^1(0, T)$, $\beta_i(t) \in L^2(0, T)$. Then

$$d(x_1(t)x_2(t)) = X_1(t)dX_2(t) + X_2(t)dX_1(t) + \beta_1(t)\beta_2(t)dt.$$

Let $\alpha_i(t) = \alpha_i$; $\beta_i(t) = \beta_i$ independent of t , where $i = 1, 2$

$$\text{Therefore } d(x_1(t)x_2(t)) = X_1(t)dX_2(t) + X_2(t)dX_1(t) + \beta_1\beta_2 dt \dots (6)$$

3: Propositions:

a: The quotient stochastic differential equation of the first order:

let x_1 and x_2 are two stochastic processes and time independent. Then by using Ito Formula we have:

$$d\left(\frac{x_1}{x_2}\right) = \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) dt + \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw \dots (7)$$

Proof: from equation (6) then we have

$$d\left(\frac{x_1}{x_2}\right) = x_1 d\left(\frac{1}{x_2}\right) + \frac{1}{x_2} dx_1 + G_1 G_2 dt \dots (8)$$

let $z = \frac{1}{x_2}$ and $dz = d\left(\frac{1}{x_2}\right)$ from equation (2) then

$$\begin{aligned} dx_1 &= F_1 dt + G_1 dw; dz = F_2 dt + G_2 dw \\ d\left(\frac{x_1}{x_2}\right) &= d(x_1 z) = x_1 dz + z dx_1 + G_1 G_2 dt = \\ &= x_1 (F_2 dt + G_2 dw) + z (F_1 dt + G_1 dw) + G_1 G_2 dt = \\ &= x_1 F_2 dt + x_1 G_2 dw + z F_1 dt + z G_1 dw + G_1 G_2 dt = \\ &= (x_1 F_2 + z F_1 + G_1 G_2) dt + (x_1 G_2 + z G_1) dw \end{aligned}$$

Since $z = \frac{1}{x_2}$ then we have

$$d(x_1 z) = d\left(\frac{x_1}{x_2}\right) = \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) dt + \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw$$

Then from equation (7) we have:

$$\begin{aligned} \frac{x_1(t)}{x_2(t)} &= \frac{x_1(0)}{x_2(0)} + \int_0^t \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) ds + \\ &+ \int_0^t \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw_s \dots (9) \end{aligned}$$

b: The quotient stochastic differential equation of degree two (i.e. $d\left(\frac{x_1}{x_2}\right)^2$):

$$\begin{aligned} d\left(\frac{x_1}{x_2}\right)^2 &= \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) dt + \\ &+ \left(\frac{2x_1 G_1}{x_2^2} + \frac{2x_1^2 G_2}{x_2}\right) dw \dots (10) \end{aligned}$$

proof:

let $z = \frac{1}{x_2}$, $z^2 = \frac{1}{x_2^2}$ and $dz^2 = d\left(\frac{1}{x_2^2}\right)$. From (2) then we

$$\begin{aligned} \text{have } dx_1 &= F_1 dt + G_1 dw; dz = F_2 dt + G_2 dw \\ d\left(\frac{x_1}{x_2}\right)^2 &= d(x_1^2 z^2) = 2x_1 z^2 dx_1 + 2x_1^2 z dz + dx_1^2 dz^2 \end{aligned}$$

By using theorem(1), we have

$$dx_1^2 = 2x_1 dx_1 + G_1^2 dt; dz^2 = 2z dz + G_2^2 dt$$

Then,

$$\begin{aligned} d\left(\frac{x_1}{x_2}\right)^2 &= d(x_1^2 z^2) = 2x_1 z^2 dx_1 + 2x_1^2 z dz + \\ &+ dx_1^2 dz^2 = 2x_1 z^2 (F_1 dt + G_1 dw) + 2x_1^2 z (F_2 dt + \\ &+ G_2 dw) + (2x_1 dx_1 + G_1^2 dt)(2z dz + G_2^2 dt) = \\ &= 2x_1 z^2 F_1 dt + 2x_1 z^2 G_1 dw + 2x_1^2 z F_2 dt + \\ &+ 2x_1^2 z G_2 dw + \\ &+ (2x_1 (F_1 dt + G_1 dw) + G_1^2 dt)(2z (F_2 dt + G_2 dw) + \\ &+ G_2^2 dt) = (2x_1 z^2 F_1 + 2x_1^2 z F_2) dt + (2x_1 z^2 G_1 + \\ &+ 2x_1^2 z G_2) dw + (2x_1 F_1 dt + 2x_1 G_1 dw + G_1^2 dt)(2z F_2 dt + \\ &+ 2z G_2 dw + G_2^2 dt) = (2x_1 z^2 F_1 + 2x_1^2 z F_2) dt + \\ &+ (2x_1 z^2 G_1 + 2x_1^2 z G_2) dw + 4x_1 z F_1 F_2 (dt)^2 + \\ &+ 4x_1 z F_1 G_2 dw dt + 2x_1 z F_1 G_2^2 (dt)^2 + \\ &+ 4x_1 z F_2 G_1 dw dt + 4x_1 z G_1 G_2 (dw)^2 + \\ &+ 2x_1 G_1 G_2^2 dw dt + 2z F_2 G_1^2 (dt)^2 + 2z G_2 G_1^2 dw dt + \\ &+ G_1^2 G_2^2 (dt)^2 \end{aligned}$$

From equation (4) and (5) then we have

$$d(x_1^2 z^2) = (2x_1 z^2 F_1 + 2x_1^2 z F_2)dt + (2x_1 z^2 G_1 + 2x_1^2 z G_2)dw + 4x_1 z G_1 G_2 dt = (2x_1 z^2 F_1 + 2x_1^2 z F_2 + 4x_1 z G_1 G_2)dt + (2x_1 z^2 G_1 + 2x_1^2 z G_2)dw$$

since $z = \frac{1}{x_2}$ then $d\left(\frac{x_1}{x_2}\right)^2 = \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right)dt + \left(\frac{2x_1 G_1}{x_2^2} + \frac{2x_1^2 G_2}{x_2}\right)dw$

The integral of (10) is.

$$\frac{x_1^2(t)}{x_2^2(t)} = \frac{x_1^2(0)}{x_2^2(0)} + \int_0^t \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) ds + \int_0^t \left(\frac{2x_1 G_1}{x_2^2} + \frac{2x_1^2 G_2}{x_2}\right) dw_s \dots(11)$$

C: The quotient stochastic differential equation in the general form :

$$d\left(\frac{x_1}{x_2}\right)^m = \left(\frac{mx_1^{m-1}F_1}{x_2^m} + \frac{mx_1^m F_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right)dt + \left(\frac{mx_1^{m-1}G_1}{x_2^m} + \frac{mx_1^m G_2}{x_2^{m-1}}\right)dw \dots(12)$$

Proof:

let $z = \frac{1}{x_2}$, $z^m = \frac{1}{x_2^m}$ and $dz^m = d\frac{1}{x_2^m}$. from (2) then

we have $dx_1 = F_1 dt + G_1 dw$; $dz = F_2 dt + G_2 dw$

$$d\left(\frac{x_1}{x_2}\right)^m = d(x_1^m z^m) =$$

$$mx_1^{m-1} z^m dx_1 + mx_1^m z_1^{m-1} dz + dx_1^m dz^m = mx_1^{m-1} z^m (F_1 dt + G_1 dw) + mx_1^m z_1^{m-1} (F_2 dt + G_2 dw) + dx_1^m dz^m$$

From Theorem (1), we have $dx_1^m = mx_1^{m-1} dx_1 + \frac{1}{2} m(m-1)x_1^{m-2} G_1^2 dt$

and $dz^m = mz^{m-1} dz + m(m-1)z^{m-2} dt$

Then,

$$d\left(\frac{x_1}{x_2}\right)^m = mx_1^{m-1} z^m F_1 dt + mx_1^{m-1} z^m G_1 dw + mx_1^m z_1^{m-1} F_2 dt + mx_1^m z_1^{m-1} G_2 dw + \left(mx_1^{m-1} dx_1 + \frac{1}{2} m(m-1)x_1^{m-2} G_1^2 dt\right) (mz^{m-1} dz + m(m-1)z^{m-2} dt) = mx_1^{m-1} z^m F_1 dt + mx_1^{m-1} z^m G_1 dw + mx_1^m z_1^{m-1} F_2 dt + mx_1^m z_1^{m-1} G_2 dw + \left(mx_1^{m-1} (F_1 dt + G_1 dw) + \frac{1}{2} m(m-1)x_1^{m-2} G_1^2 dt\right) (mz^{m-1} (F_2 dt + G_2 dw) + m(m-1)z^{m-2} dt)$$

From equation (4) and equation (5), we have

$$d\left(\frac{x_1}{x_2}\right)^m = mx_1^{m-1} z^m F_1 dt + mx_1^{m-1} z^m G_1 dw + mx_1^m z_1^{m-1} F_2 dt + mx_1^m z_1^{m-1} G_2 dw + m^2 x_1^{m-1} z^{m-1} G_1 G_2 dt = (mx_1^{m-1} z^m F_1 + mx_1^m z_1^{m-1} F_2 + m^2 x_1^{m-1} z^{m-1} G_1 G_2)dt + (mx_1^{m-1} z^m G_1 + mx_1^m z_1^{m-1} G_2)dw$$

Since $z = \frac{1}{x_2}$ then:

$$d\left(\frac{x_1}{x_2}\right)^m = \left(\frac{mx_1^{m-1} F_1}{x_2^m} + \frac{mx_1^m F_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right)dt + \left(\frac{mx_1^{m-1} G_1}{x_2^m} + \frac{mx_1^m G_2}{x_2^{m-1}}\right)dw$$

Or equivalently by integrated $d\left(\frac{x_1}{x_2}\right)^m$, we have

$$\frac{x_1^m(t)}{x_2^m(t)} = \frac{x_1^m(0)}{x_2^m(0)} + \int_0^t \left(\frac{mx_1^{m-1} F_1}{x_2^m} + \frac{mx_1^m F_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right) ds + \int_0^t \left(\frac{mx_1^{m-1} G_1}{x_2^m} + \frac{mx_1^m G_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right) dw_s \dots(13)$$

4. The moment

In this paragraph we find the moments to the solution of the Quotient stochastic differential equation (Mean, Variance and the k-moment) by using the above proposition:

Let we have

$$d\left(\frac{x_1}{x_2}\right) = \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right)dt + \left(x_1 G_2 + \frac{G_1}{x_2}\right)dw$$

Then the mean of $\left(\frac{x_1}{x_2}\right)$ is

$$E\left(\frac{x_1(t)}{x_2(t)}\right) = E\left(\frac{x_1(0)}{x_2(0)}\right) + E\left(\int_0^t \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) ds\right) + E\left(\int_0^t \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw_s\right) = \frac{x_1(0)}{x_2(0)} + E\left(\int_0^t \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) ds\right) \dots(14)$$

And from equation (11) we can find the expected value to $\left(\frac{x_1}{x_2}\right)^2$ as:

$$E\left(\frac{x_1^2(t)}{x_2^2(t)}\right) = E\left(\frac{x_1^2(0)}{x_2^2(0)}\right) + E\left(\int_0^t \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) ds\right) + E\left(\int_0^t \left(\frac{2x_1 G_1}{x_2^2} + \frac{2x_1^2 G_2}{x_2}\right) dw_s\right) = \frac{x_1^2(0)}{x_2^2(0)} + E\left(\int_0^t \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) ds\right) \dots(15)$$

Then, $Var(X) = E(X^2) - (E(X))^2$, where $X = \frac{x_1}{x_2}$

The moment for general form (equation (13)) or the k'th moment is:

$$E\left(\frac{x_1^m(t)}{x_2^m(t)}\right) = E\left(\frac{x_1^m(0)}{x_2^m(0)}\right) + E\left(\int_0^t \left(\frac{mx_1^{m-1} F_1}{x_2^m} + \frac{mx_1^m F_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right) ds\right) + E\left(\int_0^t \left(\frac{mx_1^{m-1} G_1}{x_2^m} + \frac{mx_1^m G_2}{x_2^{m-1}}\right) dw_s\right) = \frac{x_1^m(0)}{x_2^m(0)} + E\left(\int_0^t \left(\frac{mx_1^{m-1} F_1}{x_2^m} + \frac{mx_1^m F_2}{x_2^{m-1}} + \frac{m^2 x_1^{m-1} G_1 G_2}{x_2^{m-1}}\right) ds\right) \dots(16)$$

Example: (1)

Suppose $d\left(\frac{x_1}{x_2}\right) = \left(\frac{s}{r}\right)dw$ or we can write it as

$$dx_1 = s dw \text{ and } dx_2 = r dw$$

s and r are constants, also let $x_1 = t^2 + 1$, $x_2 = t$, where t is a scalar (t ≠ time). By using Itô's formula find $E\left(\frac{x_1}{x_2}\right)$ and $var\left(\frac{x_1}{x_2}\right)$.

Solution: To find the mean of $\left(\frac{x_1(t)}{x_2(t)}\right)$:

Let $z = \frac{1}{x_2}$ then $dz = d\left(\frac{1}{x_2}\right)$, by equation (2) we get

$$dx_1 = F_1 dt + G_1 dw; dz = F_2 dt + G_2 dw$$

Then

$$d(x_1 z) = (x_1 F_2 + z F_1 + G_1 G_2)dt + (x_1 G_2 + z G_1)dw = (0 + 0 + sr)dt + (sz + rx_1)dw = sr dt + (sz + rx_1)dw$$

So

$$\int_0^t d(x_1 z) = \int_0^t sr ds + \int_0^t (sz + rx_1)dw$$

$$\frac{x_1(t)}{x_2(t)} = \frac{x_1(0)}{x_2(0)} + \int_0^t sr ds + \int_0^t \left(\frac{s}{x_2} + rx_1\right)dw$$

Then the expected value (mean) of $\left(\frac{x_1(t)}{x_2(t)}\right)$ is:

$$E\left(\frac{x_1(t)}{x_2(t)}\right) = \frac{x_1(0)}{x_2(0)} + E\left(\int_0^t sr ds\right) + E\left(\int_0^t \left(\frac{s}{x_2} + rx_1\right)dw\right) = \frac{x_1(0)}{x_2(0)} + \int_0^t E(sr)ds = \frac{x_1(0)}{x_2(0)} + srt. \text{ Where } x_2(0) \neq 0$$

The variance: First we need to find $E\left(\frac{x_1}{x_2}\right)^2$.

Let $z = \frac{1}{x_2}$, $z^2 = \frac{1}{x_2^2}$ and $dz^2 = d\frac{1}{x_2^2}$
 $d(x_1^2 z^2) = (2x_1 z^2 F_1 + 2x_1^2 z F_2 + 4x_1 z G_1 G_2)dt + (2x_1 z^2 G_1 + 2x_1^2 z G_2)dw$

$\int_0^t d(x_1^2 z^2) = \int_0^t (2x_1 z^2 F_1 + 2x_1^2 z F_2 + 4x_1 z G_1 G_2)ds + \int_0^t (2x_1 z^2 G_1 + 2x_1^2 z G_2)dw_s$
 $x_1^2(t)z^2(t) = x_1^2(0)z^2(0) + \int_0^t 4sr x_1 z ds + \int_0^t (2sx_1 z^2 + rx_1^2 z)dw_s$

Where $z = \frac{1}{x_2}$ and $z^2 = \frac{1}{x_2^2}$ then

$E\left(\frac{x_1^2(t)}{x_2^2(t)}\right) = \frac{x_1^2(0)}{x_2^2(0)} + E\int_0^t 4sr \frac{x_1}{x_2} ds = \frac{x_1^2(0)}{x_2^2(0)} + 4sr \int_0^t E\left(\frac{x_1}{x_2}\right) ds$

$= \frac{x_1^2(0)}{x_2^2(0)} + 4sr \int_0^t \left(\frac{x_1(0)}{x_2(0)} + srt\right) ds = \frac{x_1^2(0)}{x_2^2(0)} + 4sr \left(\frac{x_1(0)}{x_2(0)}t + \frac{1}{2}srt^2\right) = \frac{x_1^2(0)}{x_2^2(0)} + 4sr \frac{x_1(0)}{x_2(0)}t + 2s^2 r^2 t^2$

$var\left(\frac{x_1}{x_2}\right) = E\left(\frac{x_1^2}{x_2^2}\right) - \left(E\left(\frac{x_1}{x_2}\right)\right)^2$

$var\left(\frac{x_1}{x_2}\right) = \left\{\frac{x_1^2(0)}{x_2^2(0)} + 4sr \frac{x_1(0)}{x_2(0)}t + 2s^2 r^2 t^2\right\} - \left\{\frac{x_1(0)}{x_2(0)} + srt\right\}^2 = \frac{2sr x_1(0)}{x_2(0)} + s^2 r^2 t^2$

In the same way we can find the higher moment.

Example: (2)

Suppose $dX_i = F_i dt + G_i dw$, $i=1,2$, $X_i = \frac{x_i}{x_2}$ and let $dx_1 = t^3 dt + 2tdw$ and $dx_2 = t^2 dt + 4tdw$, where $x_1 = t^2 + 1$ and $x_2 = t$ where t is a scalar ($t \neq$ time). Then by using Itô's formula find $E\left(\frac{x_1}{x_2}\right)$, $var\left(\frac{x_1}{x_2}\right)$, where $x_1(0) = 0$, $x_2(0) \neq 0$.

Solution: we have

$d\left(\frac{x_1}{x_2}\right) = \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) dt + \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw = \left((t^2 + 1)t^2 + \frac{t^3}{t} + 8t^2\right) dt + \left((t^2 + 1)(4t) + \frac{2t}{t}\right) dw = (t^4 + t^2 + t^2 + 8t^2) dt + (4t^3 + 4t + 2) dw = (t^4 + 10t^2) dt + (4t^3 + 4t + 2) dw$

Then from equation (9), we have

$\frac{x_1(t)}{x_2(t)} = \frac{x_1(0)}{x_2(0)} + \int_0^t (s^4 + 10s^2) ds + \int_0^t (4s^3 + 4s + 2) dw_s$

Since $x_1(0) = 0$, then by taking the expectation for both sides, we get

$E\left(\frac{x_1(t)}{x_2(t)}\right) = E\left(\int_0^t (s^4 + 10s^2) ds\right) + E\left(\int_0^t (4s^3 + 4s + 2) dw_s\right) = E\left(\int_0^t (s^4 + 10s^2) ds\right) = \frac{1}{5}t^5 + \frac{10}{3}t^3$

To find the variance of $\left(\frac{x_1}{x_2}\right)$ we need $E\left(\left(\frac{x_1}{x_2}\right)^2\right)$

From equation (10), we have:

$d\left(\frac{x_1}{x_2}\right)^2 = \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) dt + \left(\frac{2x_1 G_1}{x_2} + \frac{2x_1^2 G_2}{x_2}\right) dw = \left(\frac{2t^2(0)}{4t^2} + \frac{2t^4(2t)}{2t} + \frac{4t^2(0)(2t)}{2t}\right) dt + \left(\frac{2t^2(0)}{4t^2} + \frac{2t^4(2t)}{2t}\right) dw = \frac{1}{2} dt + 2t^4 dw$

$= \left(\frac{2(t^2+1)t^3}{t^2} + \frac{4(t^2+1)^2 t^2}{t} + \frac{4(t^2+1)(2t)(4t)}{t}\right) dt + \left(\frac{2(t^2+1)2t}{t^2} + \frac{4(t^2+1)(4t)}{t}\right) = (2t^3 + 2t + 4t^5 + 8t^3 + 4t + 32t^2 + 32t) dt + \left(4t + \frac{4}{t} + 16t^2 + 16\right) dw = (4t^5 + 10t^3 + 38t) dt + \left(16t^2 + 4t + \frac{4}{t} + 16\right) dw$

So that:

$E\left(\frac{x_1(t)}{x_2(t)}\right)^2 = \left(\frac{x_1(0)}{x_2(0)}\right)^2 + \int_0^t (4s^5 + 10s^3 + 38s) ds + \int_0^t \left(16s^2 + 4s + \frac{4}{s} + 16\right) dw_s$

By taking the expectation for both side (since $x_1(0) = 0$), then

$E\left(\left(\frac{x_1(t)}{x_2(t)}\right)^2\right) = E\left(\int_0^t (4s^5 + 10s^3 + 38s) ds\right) + E\left(\int_0^t \left(16s^2 + 4s + \frac{4}{s} + 16\right) dw_s\right)$

$E\left(\left(\frac{x_1(t)}{x_2(t)}\right)^2\right) = E\left(\int_0^t (4s^5 + 10s^3 + 38s) ds\right) = E\left(\frac{3}{2}t^6 + \frac{5}{2}t^4 + 19t^2\right) = \frac{3}{2}t^6 + \frac{5}{2}t^4 + 19t^2$

$Var\left(\frac{x_1}{x_2}\right) = E\left(\left(\frac{x_1}{x_2}\right)^2\right) - \left(E\left(\frac{x_1}{x_2}\right)\right)^2$

Then

$Var\left(\frac{x_1}{x_2}\right) = \left(\frac{3}{2}t^6 + \frac{5}{2}t^4 + 19t^2\right) - \left(\frac{1}{5}t^5 + \frac{10}{3}t^3\right)^2 = \frac{5}{2}t^4 + 19t^2 - \frac{1}{25}t^{10} - \frac{5}{3}t^8 - \frac{891}{18}t^6$

Example: (3)

Suppose $dx_1 = dt$ and $dx_2 = 2tdw$, where $x_1 = t^2$ and $x_2 = 2t$, where t is a scalar ($t \neq$ time). Then by using Itô's formula find $E\left(\frac{x_1}{x_2}\right)$ and $var\left(\frac{x_1}{x_2}\right)$, Where $x_1(0) = 0$, $x_2(0) \neq 0$.

Solution: from equation (7) then

$d\left(\frac{x_1}{x_2}\right) = \left(x_1 F_2 + \frac{F_1}{x_2} + G_1 G_2\right) dt + \left(x_1 G_2 + \frac{G_1}{x_2}\right) dw = \left(t^2(0) + \frac{1}{2t} + 0(2t)\right) dt + \left(t^2(2t) + \frac{1}{2t}(0)\right) dw = \frac{1}{2t} dt + 2t^3 dw$

$\frac{x_1(t)}{x_2(t)} = \frac{x_1(0)}{x_2(0)} + \int_0^t \frac{1}{2s} ds + \int_0^t 2s^3 dw_s = \int_0^t \frac{1}{2s} ds + \int_0^t 2s^3 dw_s$

$E\left(\frac{x_1(t)}{x_2(t)}\right) = E\left(\int_0^t \frac{1}{2s} ds\right) + E\left(\int_0^t 2s^3 dw_s\right) = E\left(\int_0^t \frac{1}{2s} ds\right) = \frac{1}{2} \ln(t)$

$E\left(\int_0^t \frac{1}{2s} ds\right) = \frac{1}{2} \ln(t)$

To find the variance of $\left(\frac{x_1}{x_2}\right)$, we need $E\left(\left(\frac{x_1}{x_2}\right)^2\right)$

From equation (10) we have:

$d\left(\frac{x_1}{x_2}\right)^2 = \left(\frac{2x_1 F_1}{x_2^2} + \frac{2x_1^2 F_2}{x_2} + \frac{4x_1 G_1 G_2}{x_2}\right) dt + \left(\frac{2x_1 G_1}{x_2} + \frac{2x_1^2 G_2}{x_2}\right) dw = \left(\frac{2t^2}{4t^2} + \frac{2t^4(0)}{2t} + \frac{4t^2(0)(2t)}{2t}\right) dt + \left(\frac{2t^2(0)}{4t^2} + \frac{2t^4(2t)}{2t}\right) dw = \frac{1}{2} dt + 2t^4 dw$

Then,

$E\left(\frac{x_1(t)}{x_2(t)}\right)^2 = \left(\frac{x_1(0)}{x_2(0)}\right)^2 + \int_0^t \frac{1}{2} ds + \int_0^t 2s^4 dw_s = \int_0^t \frac{1}{2} ds + \int_0^t 2s^4 dw_s$

And then,

$E\left(\left(\frac{x_1(t)}{x_2(t)}\right)^2\right) = E\left(\int_0^t \frac{1}{2} ds\right) = \frac{1}{2} t$

So that

$$\text{Var}\left(\frac{x_1}{x_2}\right) = E\left(\left(\frac{x_1}{x_2}\right)^2\right) - \left(E\left(\frac{x_1}{x_2}\right)\right)^2 = \frac{1}{2}t - \left(\frac{1}{2}\ln(t)\right)^2 = \frac{1}{2}t - \frac{1}{4}(\ln(t))^2 = \frac{1}{2}t - \frac{1}{2}\ln(t)$$

For the higher moment we can use the same way.

5. Conclusion

In this paper, we showed using Itô's formula that the quotient stochastic differential equations can be found by the same method for product stochastic

differential equations with some attention when we using their theorem's. (That is, Itô's formula is valid for rational form of the functions $u(x(t), t)$ of the variables x_1 and x_2). Also we find the moments to the solution of the Quotient stochastic differential equation by using the above proposition with some examples.

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العزوم لبعض المعادلات التفاضلية التصادفية الكسرية مع التطبيق

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الملخص

في هذا البحث سوف نستخدم صيغة المعادلات التفاضلية التصادفية في حالة الضرب لدراسة وإيجاد حل بعض المعادلات التفاضلية التصادفية الكسرية باستخدام صيغة Itô التكاملية، ثم نحاول إيجاد العزوم لها المتمثلة (المتوسط (mean)، التباين variance والعزم-k (k'th moment)، و قدمنا بعض الأمثلة لتوضيح الطريقة.