

Stochastic Differential Equation

- ▶ Introduce uncertainties by introducing an additive white noise term, i.e.,

$$dX(t) = b(t, X(t))dt + dW(t)$$

where $b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ is a continuous function.

The term $dW(t)$ is called as white noise and its integral is Brownian Motion $W(t)$



Now we are moving into the stochastic differential equations. Introduce uncertainties by introducing an additive white noise term that is a dX_t is equal to $b(t, X_t) dt + dW_t$ where b is a real valued continuous function from $[0, T] \times \mathbb{R}$. The term dW_t is called as a white noise and its integral is Brownian motion W_t . Here the above equation is also known as stochastic differential equation or SDE the meaning of which would be more clear after the introduction of a stochastic integral concept.

Stochastic Differential Equation...

- ▶ Note that $\{X(t), t \geq 0\}$ is a stochastic process. The integral form is

$$X(t) = X(0) + \int_0^t b(s, X(s))ds + W(t)$$

a stochastic integral equation.

- ▶ In general, if $b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}, \sigma : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ are two suitable functions, then an integral equation of the form

$$X(t) = X(0) + \int_0^t b(s, X(s))ds + \int_0^t \sigma(s, X(s))dW(s) \quad (2)$$

is called a stochastic integral equation.

- ▶ It is defined by integration of a stochastic process with respect to Brownian motion.



Note that X of t is a stochastic process. The integral form of the differential equation is the X of stochastic differential equation is X of t is equal to X of 0 plus integration 0 to t of b of s of X of s ds plus Wt is a stochastic integral equation.

Stochastic Differential Equation...

- ▶ Equation (2) can be written as

$$dX(t) = b(t, X(t))dt + \sigma(t, X(t))dW(t), \quad (0 \leq t \leq T) \quad (3)$$

where $b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ and $\sigma : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ are two given functions.

- ▶ The equation (3) is referred to as a stochastic differential equation.
- ▶ The interpretation of (3) tells us that the change $dX(t) = X(t + \Delta t) - X(t)$ is caused by a change dt of time, with factor $b(t, X(t))$ in combination with a change $dW(t) = W(t + \Delta t) - W(t)$ of Brownian motion with factor $\sigma(t, X(t))$.



In general if b and Σ are the two suitable functions then the integral equation of the form X of t is equal to X of 0 plus integration 0 to t function b division with respect to s plus 0 to t Σ of X of s dW_s . In the equation two, the first equation – the first integral is different from the second integral. The second integral is integration with respect to the Brownian motion sample path W_s . This integral equation is defined by the integration of stochastic process with respect to the Brownian motion. So the equation two that is nothing but this equation, equation two can be written as dX_t is equal to B of $t, X_t dt$ plus σ of t of $X_t dW_t$ where t lies between 0 to T where b and Σ are two given functions. This the equation 3 is referred to as a stochastic differential equation. The interpretation of equation 3 tells us that the change that is d of X_t that is nothing but the X of t plus Δt minus X_t is caused by a change dT of time with the factor b of t, X_t in combination with the change dW_t that is nothing but W of t plus Δt minus W_t of Brownian motion with the factor Σ of t, X_t .

The Brownian motion is adopted to the natural filtration. So the unknown is in the Σ as well as b and the increment of Brownian motion therefore this equation is called stochastic differential equation.

Strong Solution

- ▶ Let (Ω, \mathcal{F}, P) be a probability space, and $\{W(t), t \geq 0\}$ be a Brownian motion defined on it.
- ▶ An adapted process $X(t)$ satisfying (2) is said to be a strong solution (uniquely) if $\{X(t)\}$ and $\{Y(t)\}$ are two solutions on the same probability space satisfying (2) then

$$P\{X(t) = Y(t) \forall t\} = 1$$
- ▶ A strong solution is an explicit function f such that $X(t) = f(t, W(s), s < t)$.

Now we are going to discuss there are two types of solutions for the stochastic differential equation. The first type is called a strong solution. The second type is called weak solution. So we are going to discuss the strong solution first. Let us Σ sorry let Ω, \mathcal{F}, P be the probability space and W_t be a Brownian motion defined on it. The adapted process X of t satisfying the equation two that's a stochastic differential equation is said to be a strong solution uniquely if X of t and the W_t are the two solutions on the same probability space satisfying the stochastic differential equation two then the probability of X of t is equal to Y of t for all t that will be 1 then X_t is called a strong solution and it is also a unique solution. That means if you have another solution Y of t then probability of X of t is equal to Y of t for all t will be 1. In

general a strong solution is an explicit function F such that X of t is a function of t , W_s where s is less than t . One can write the solution in that explicit function F of t with the Brownian motion. Then this solution is called the strong solution.

Weak Solution

- ▶ Both strong and weak solutions require the existence of the process $\{X(t), t \geq 0\}$ that solves the integral equation version of the SDE.
- ▶ The difference between the two lies in the underlying probability space (Ω, \mathcal{F}, P) .
- ▶ A weak solution consists of a probability space and the process that satisfies the integral equation, while a strong solution is a process that satisfies the equation and is defined on a given probability space.
- ▶ When no known explicit solution exists for a given SDE, then we can approximate it by a numerical solution, replacing differentials by differences.



Hence, approximate solution method is similar to numerical integration.

Now we are going to discuss what is the weak solution of stochastic differential equation. Weak solution both strong and weak solutions require the existence of the process X_t that solve the integral equation version of the SDE. The difference between the two lies in the underlying probability space. You have each solution consists of a probability space and the process that satisfies the integral equation. While a strong solution is the process that satisfies the equation and is defined on a given probability space. When no explicit solution exists for a given SDE then we can approximate it by the numerical solution replacing differentials by differences. Hence, approximate solution method is similar to the numerical integration.

So with this we have discussed the strong solution and weak solutions of stochastic differential equations.

Existence and Uniqueness Solution

- ▶ Now, we discuss the existence of strong solution.
- ▶ Suppose $b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}, \sigma : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ are continuous and satisfying Lipschitz condition

$$|b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| < k |x - y|$$

and $X(0)$ and $W(t)$ are independent random variables.

- ▶ Then the equation (2) has a unique solution.



In this course we are interested to find the strong solution not the weak solution. When the above results hold good we say that the quadratic variation accumulated by the Brownian motion over the interval $0, T$ is T in mean square and this. Equation 2 can be written as dx_t is equal to $b(t, x_t) dt$ plus $\sigma(t, x_t) dW_t$ where t lies between 0 to T where b and σ are two given functions.

Now we discuss the simple examples for the stochastic differential equation. Consider this stochastic differential equation dx_t is equal to $xt dt$ with x_0 is equal to 1 . here $b(t, x)$ is equal to 0 and $\sigma(t, x)$ is equal to x . You can verify the Lipschitz condition for this b is equal to 0 and σ is equal to x hence the strong solution exists. Obtaining the strong solution will be explained in the further lectures.

Example 3

Let $S(t)$ be the stock price at time t . Consider the SDE

$$dS(t) = \mu S(t)dt + \sigma S(t)dW(t), \quad S(0) \text{ is known}$$

where μ is the constant growth rate of the stock and σ is the volatility.

Here $b(t, x) = \mu x$ and $\sigma(t, x) = \sigma x$. Since μ and σ are constants, Lipschitz condition is satisfied.

Hence, the strong solution exists. Obtaining the strong solution will be explained in further lectures.



We will see one more example for the stochastic differential equation. Here s of t be the stock price at time t and the corresponding stochastic differential equation for this example is a d of st is equal to μ times s of tdt plus σ of s of $tdwt$ with s of 0 is known. Here μ is a constant growth rate of the stock and σ is the volatility when you compare it to standard as stochastic differential equation we get a b of t, x is equal to μ of x and σ t, x is same as σ x . Since μ and σ are constants the Lipschitz condition is satisfied. Hence the strong solution exists and this example also how to find the solution that will be discussed in the further lectures.

References

- ▶ S. E. Shreve. Stochastic Calculus for Finance, Vol. I & Vol. II, Springer, 2004.
- ▶ Thomas Mikosch Elementary Stochastic Calculus with Finance in view, World Scientific, 2006.
- ▶ Suresh Chandra, S. Dharmaraja, Aparna Mehra, R. Khemchandani, "Financial Mathematics: An Introduction", Narosa Publication House, 2012.



Now we are going to discuss the existence and uniqueness solution that is basically a strong solution. Now we discuss the existence of strong solution. Suppose b is a continuous function. Similarly σ is the continuous function. Satisfying the Lipschitz condition the absolute of difference of b of t,x minus b of t,y plus in the absolute σ t,x minus σ t,y if this summation is less than k times absolute of x minus y where k is a positive constant and also the initial distribution x_0 and w_t are independent random variables then we can say the solution is going to exist that will be unique also. So whenever the Lipschitz conditions satisfied with the two continuous function b and σ for a positive constant k along with x_0 and w_t are independent random variables. If both the conditions are satisfied by any stochastic differential equations then we can conclude it has the unique and it has the existence of strong solution as well as it will be unique.

Ito-Picard Iteration

- ▶ Note that, the existence and uniqueness follow very closely the standard Picard's method for constructing solutions of ODE.
- ▶ Using the Ito-Picard iteration and $X(0) = x_0$, we get for $n = 1, 2, \dots$

$$X_{n+1}(t) = x_0 + \int_0^t b(s, X_n(s)) ds + \int_0^t \sigma(s, X_n(s)) dW(s)$$

- ▶ Remark that, the iterations are well defined.
- ▶ By the convergence of iteration scheme, we finally obtain

$$X(t) = \lim_{n \rightarrow \infty} X_n(t)$$



This is similar to the existence and uniqueness solution of a ODE the only difference is it does not have the term and the Sigma term so it has only the first term which is less than a times absolute of x minus y that's Lipschitz condition for ODE. So here also the same thing along with the continuous function Sigma. If this condition is satisfied along with this condition x_0 and w_t are independent random variables then the given SDE as unique solution have the existence of a strong solution and that will give unique. Note that the existence and uniqueness follow very closely the standard Picard's method for constructing solutions of ODE. You know the Picard iteration for a ODE, Ordinary Differential Equation and this iteration is called Ito-Picard iteration so using Ito-Picard iteration X_n is equal to x_0 we get for n is equal to 1, 2, 3. X_{n+1} will be X_n plus the integration plus the another integration.

That means with the initial value X_n you can find for n is equal to 1 you can find for n is equal to 0 you will find the x_1 of t first using X_n . Then for n is equal to 1 you will get x of 2 x suffix 2 of t and recursively you can get the x_{n+1} of t for every n as n tends to infinity you can get the x of t .

So remark that the iterations are well defined because it satisfies the Lipschitz conditions as well as X_n and the w_t are independent random variables the solution is going to be exist as well as it will be unique and these iterations are well defined by the convergence of iteration scheme we finally obtain x of t is the limit n tends to infinity x_n of t . for every n it is a random variable so this random variable converges to the random variable x of t . So this we are showing through the Ito-Picard iteration. This Ito-Picard iteration is similar to the Picard iteration of ordinary differential equation.

Example 2

Consider the stochastic differential equation

$$dX(t) = X(t)dW(t), \text{ with } X(0) = 1.$$

Here, $b(t, x) = 0$ and $\sigma(t, x) = x$. Hence, Lipschitz condition is satisfied.

Hence, the strong solution exists. Obtaining the strong solution will be explained in further lectures.



Now we discuss the simple examples for the stochastic differential equation. Consider the stochastic differential equation dX_t is equal to $X_t dW_t$ with X_0 is equal to 1. here b of t, x is equal to 0 and σ for t, x is equal to X . You can verify the Lipschitz condition for this b is equal to 0 and σ is equal to x hence the strong solution exists. Obtaining the strong solution will be explained in the further lectures.

We will see one more example for the stochastic differential equation. Here S_t be the stock price at time t and the corresponding stochastic differential equation for this example is dS_t is equal to $\mu S_t dt + \sigma S_t dW_t$ with S_0 is known. Here μ is a constant growth rate of the stock and σ is the volatility. When you compare it to standard stochastic differential equation we get a b of t, x is equal to μx and σ of t, x is same as σx . Since μ and σ are constants the Lipschitz condition is satisfied. Hence the strong solution exists and this example also how to find the solution that will be discussed in the further lectures.

Here is the list of books for the reference.

In this lecture we have discussed stochastic differential equation. For that we have discussed the variations of a real valued function starting with the first order variation. Pth order variation. Then followed by that we have discussed the variations of Brownian motion starting with the first order variation, quadratic variation and Pth order variation also. Then we have discussed that stochastic differential equation by adding white noise term in the ordinary differential equation. Then we have discussed the equivalent as stochastic integral equations and also we have discussed the strong and weak solutions and finally we have given existence of – existence as well as a uniqueness of a strong solution. And finally we have discussed Ito-Picard alternation methods.