

Stochastic Differential Equation

Consider $H(t) = \int_{s=0}^t \mu(s) ds$. Moreover partition the interval $[0, t]$ into $0 = t_0 < t_1 < \dots < t_n = t$ and define $H_n(t) = \sum_{i=1}^n \mu(t_i^*) [t_i - t_{i-1}]$, where $t_i^* \in [t_{i-1}, t_i]$.

Now by Riemann Integral we know that $H(t) = \lim_{n \rightarrow \infty} H_n(t)$, where $\delta_n = \max_{1 \leq i \leq n} |t_i - t_{i-1}|$.

Moreover $I(t) = \int_{s=0}^t \sigma(s) dF(s)$.

Using the fundamentals mentioned above we can easily obtain the following which is

$$\frac{I_n(t+\varepsilon) - I_n(t)}{\varepsilon} = \sigma(t) \left\{ \frac{F(t+\varepsilon) - F(t)}{\varepsilon} \right\} \text{ and also define } \frac{dX(t)}{dt} = \mu(t)dt + \sigma(t)dF(t)$$

Its Calculus

Without any loss of generality assume $X(0) = 0$ and define

$$X(t) = \int_0^t \mu\{X(s), s\} ds + \int_0^t \sigma\{X(s), s\} dZ(s) \text{ where } Z(\cdot) \text{ is the standard Brownian motion.}$$

Furthermore we may write $dX(t) = \mu\{X(t), t\}dt + \sigma\{X(t), t\}dZ(t)$.

Using some calculation we can show that the BS model differential equations is

$$\frac{1}{2} \sigma^2(X, t) \frac{\partial^2 P}{\partial x^2} + \mu(x, t) \frac{\partial P}{\partial x} = - \frac{\partial P}{\partial t}$$

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Black-Scholes Model

Important : In Black-Scholes model the main assumption is the fact that $\ln(P) \sim$ Brownian Motion process, i.e., $P(t) \sim$ log normal distribution, where $P(t)$ is the price of stocks.

The main parameters for the BS model are:

1. S_0 = Stock price and this is known
2. K = strike price and this is known
3. r = interest rate and this is known
4. T = time period and this is known
5. σ = volatility and this is unknown

One should remember that the main component in the model is $\sigma^2 t$, but the bottle neck is the fact that σ which is the volatility is stochastic.

Two key assumptions is Black Scholes Model

1. Log normal prices
2. Volatility constant i.e., it is independent of time

Now recall that $\int_a^b f(x) dx = F(b) - F(a)$ and consider that

$$I = \int_0^a \frac{2}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx, \text{ if } W = \frac{x}{\sqrt{t}} \text{ which implies that } dw = dx/\sqrt{t}$$

$$\begin{aligned} \therefore I &= \int_0^{a/\sqrt{t}} \frac{2}{\sqrt{2\pi}} e^{-w^2/2} dw \\ \therefore \frac{\partial}{\partial t} \left[\int_0^{a/\sqrt{t}} \frac{2}{\sqrt{2\pi}} e^{-w^2/2} dx \right] &= \frac{\partial}{\partial t} [H(a/\sqrt{t})] = H' \left(\frac{a}{\sqrt{t}} \right) \frac{\partial}{\partial t} (a/\sqrt{t}) \\ &= \left\{ \frac{2}{\sqrt{2\pi}} e^{-a^2/2t} \right\} a - \frac{1}{2} t^{-3/2} \end{aligned}$$

$$\therefore g(t) = \begin{cases} \frac{a}{\sqrt{2\pi t^3}} e^{-\frac{a^2}{2t}} & t > 0 \\ 0 & \text{otherwise} \end{cases}, \text{ assuming } \sigma^2 = 1$$

Now generally if σ^2 is there, then we have the following

$$g(t) = \begin{cases} \frac{a}{\sqrt{2\pi\sigma^2 t^3}} \exp \left\{ -\frac{(a-\mu t)^2}{2\sigma^2 t} \right\} & t > 0 \\ 0 & \text{otherwise} \end{cases}$$

NOTE

- If there is no drift then the expected time to reach a is ∞ .
- If there is drift then the expected time to reach a is finite

Suppose at $t = 0$ the process is at the origin and let $p(k, n)$ denote the probability of being K step up after n trials

$\therefore P(k, n) = P_{0k}^n = P_{0, k-1}^{n-1} P_{k-1, k} + P_{0, k+1}^{n-1} P_{k+1, k}$ which is easily understood from **Figure 10.13**

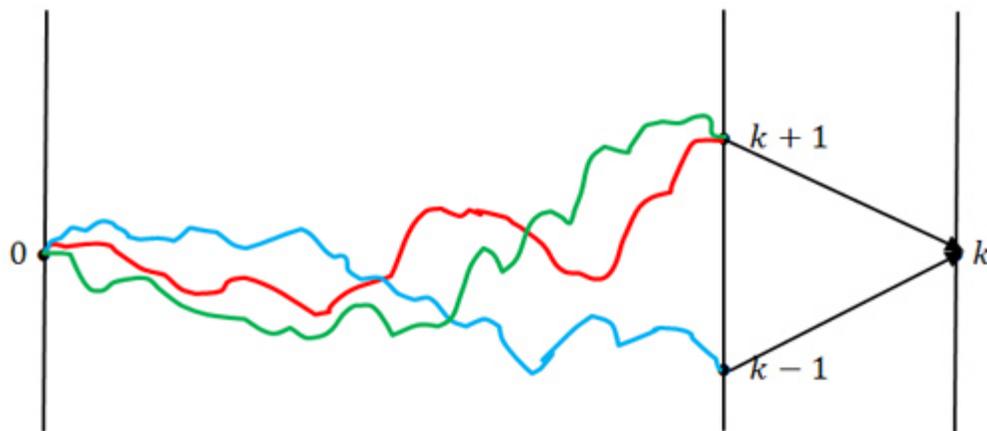


Figure 10.13: Movement of stock price according which can be considered as random

Thus:

$$p(K, n) = \frac{1}{2} p(K-1, n-1) + \frac{1}{2} p(K+1, n-1)$$

Let the time between step be Δ and let the jump size be η . Here $\frac{\eta^2}{\Delta}$ which is a constant is equal to σ^2 .

Hence we have $p \frac{(x, t+\Delta) - p(x, t)}{\Delta} = \frac{1}{2} \left[\frac{p(x+n, t) - 2p(x, t) + p(x-n, t)}{\eta^2} \right]$, i.e.,

$$p \frac{(x, t+\Delta) - p(x, t)}{\Delta} = \frac{1}{2} \sigma^2 \left[\frac{p(x+n, t) - 2p(x, t) + p(x-n, t)}{\eta^2} \right]$$

First set $x = k\eta$ and $n\Delta = t$ and let us also consider $\Delta \rightarrow 0$. Then we can easily prove that

$$\frac{\partial p}{\partial t} = \frac{1}{2} \sigma^2 \frac{\partial^2 p}{\partial x^2}$$

Using first order expansion we have

$$\begin{aligned}
 C(S, t) \times (1 + r\Delta) &= pC(S + \eta S, t + \Delta) + (1 - p)C(S - \eta S, t + \Delta) \\
 &= \frac{1}{2} \{C(S + \eta S, t + \Delta) + C(S - \eta S, t + \Delta)\} \\
 &\quad + \frac{r\Delta}{2\eta} \{C(S + \eta S, t + \Delta) - C(S - \eta S, t + \Delta)\} \\
 C(S, t) \times (1 + r\Delta) - C(S, t + \Delta) &= \frac{1}{2} \{C(S + \eta S, t + \Delta) - 2C(S, t + \Delta) + C(S - \eta S, t + \Delta)\} \\
 &\quad + \frac{r\Delta}{2\eta} \{C(S + \eta S, t + \Delta) - C(S - \eta S, t + \Delta)\}
 \end{aligned}$$

Let us now first divide both sides of the equations with Δ and take limits as $\Delta \rightarrow 0$. Furthermore we also divide and simultaneously multiply the first term and the second term on the right hand side of equality by $(\eta S)^2$ and $2\eta S$ respectively. This results in the following form which is as follows

$$\frac{1}{2} \frac{\partial^2 C(S, t)}{\partial S^2} \sigma^2 S^2 + rS \frac{\partial C(S, t)}{\partial S} + \frac{\partial C(S, t)}{\partial t} - rC(S, t) = 0$$

This is the fundamental **partial differential equation** (PDE) for pricing derivatives under the underlying assumption that logarithmic price of the underlying financial asset has **Brownian motion**.

In case one is interested to understand how the theoretical relationship between $C(S, t)$ and S varies then Figure gives a fair idea how that behaves.

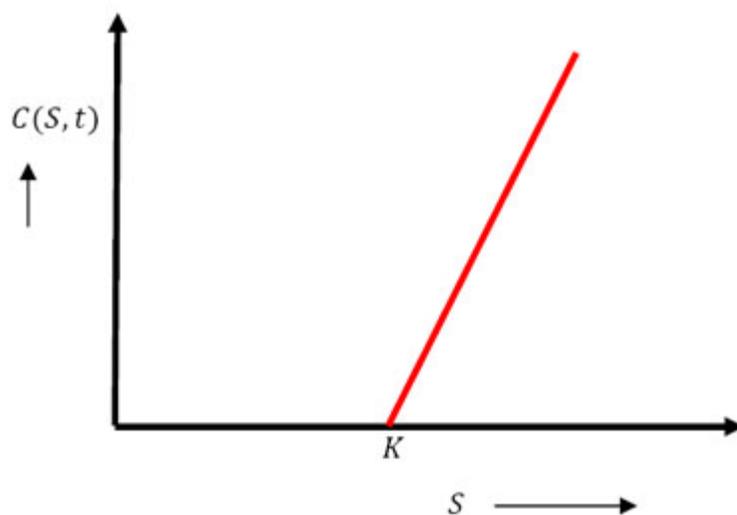


Figure 10.12: Illustration for PDE for pricing derivatives

Recall from Markov Chain Theory that $P_{ij}^n = \sum_{k=0}^{\infty} P_{ik}^{n-1} P_{kj}$. Now consider a simple random walk with equal probability of a step up and step down i.e., $\frac{1}{2}$

Now let us extend these concepts just discussed for the case when we have the figure given as

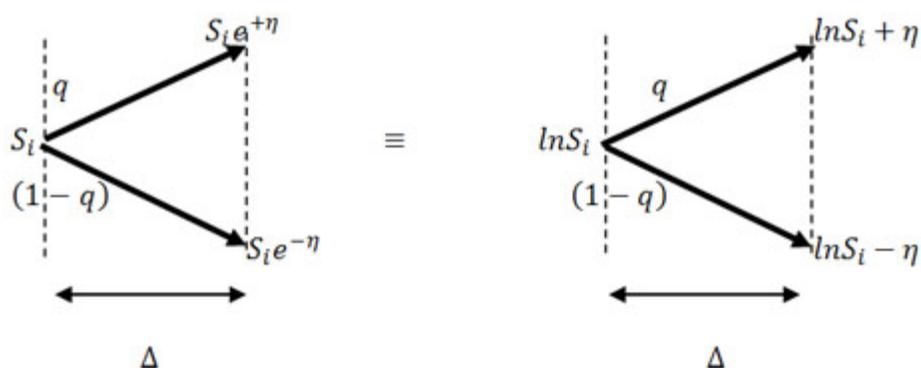


Figure 10.11: A stock and its increase and decrease of price

For the Figure we select Δ and η to be small, such that $\frac{\eta^2}{\Delta} = \sigma^2$ (a constant), and this σ^2 is variance per unit time. Hence it turns out that this **random walk** converges to a **Brownian motion** as $\Delta \rightarrow 0$.

Thus: $S_t e^{r\Delta} = S_t p e^{\eta} + S_t (1-p) e^{-\eta}$ where η and r have their usual meaning.

Using Taylor series expansion we get $r\Delta = p \times \eta + (1-p) \times (-\eta) = 2p\eta - \eta$.

Which results in

$$p = \frac{r\Delta}{2\eta} + \frac{1}{2}$$

Now when time is taken as 1 we have the following $C_i = e^{-r} [p \times C_{i+1}^{n+1} + (1-p) \times C_{i-1}^{n+1}]$, as true. In the general case when the time period is Δ , the equation takes the following form which is

$$C_i = e^{-r\Delta} [p \times C_{i+1}^{n+1} + (1-p) \times C_{i-1}^{n+1}].$$

Thus we have the following

$$(S_i + \xi^* C_i^n) e^r = S_i e^{\eta} + \xi^* C_{i+1}^{n+1} = S_i + \xi^* C_i^{n+1} = S_i e^{-\eta} + \xi^* C_{i-1}^{n+1}$$

Now if we have only $i+1$ and $i-1$ movement then the equation reduces to

$$(S_i + \xi^* C_i^n) e^r = S_i e^{\eta} + \xi^* C_{i+1}^{n+1} = S_i e^{-\eta} + \xi^* C_{i-1}^{n+1}$$

Thus we should have an unique value of p , say p^* such that the following holds true:

$S_i e^r = S_i p e^{\eta} + S_i (1-p) e^{-\eta}$. In fact it can be easily proved for the case when we have only two states, i.e., j and $i-1$, then $p^* = \frac{e^r - e^{-\eta}}{e^{\eta} - e^{-\eta}}$.

Furthermore combining we can easily see that

$$(S_i e^{\eta} + \xi^* C_{i+1}^{n+1}) \times p^* + (S_i e^{-\eta} + \xi^* C_{i-1}^{n+1}) \times (1-p^*)$$

Thus we have

$\xi^* C_i^n e^r = \xi^* \{p^* \times C_{i+1}^{n+1} + (1-p^*) \times C_{i-1}^{n+1}\}$ and as $\xi^* \neq 0$ one easily obtains

$$C_i^n = e^{-r} \{p^* \times C_{i+1}^{n+1} + (1-p^*) \times C_{i-1}^{n+1}\}$$

In general $p^* \neq q$ and it reflects the risk premium example. In case $p^* = q$, then we operate in the risk free environment and in that case p^* is termed as the **risk-neutral probability measure**. In general using this value of p^* (whether risk free or not is immaterial) we can obtain the discounted expected value of the future claim.

The Lecture Contains:

- ☰ Black-Scholes Model
- ☰ Two key assumptions is Black Scholes Model
- ☰ Stochastic Differential Equation
- ☰ Its Calculus

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