

Quantitative Finance
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Module – 06

Lecture - 34

So, continue our discussion about Ito's Lemma, venna process, generalized venna process. Based on that, we proceeded and found out that stock prices can be simply modeled using very simple concept of geometric prominent motion and continuing that if we find out. So, generally the geometric prominent motion would consist of two terms; one is left; one on the right hand side; would basically, be the mean drift rate which will be given by mu and the second term would be the volatility which is the external noise. So, if we expand the equation in the discrete form, this is what we have.

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Ito Process for Stock Prices

The discrete version of the model is:

$$\Delta S/S = \mu \Delta t + \sigma \varepsilon (\Delta t)^{1/2}$$

where:

- $\Delta S = S(t=t_2) - S(t=t_1)$, such that $t_2 - t_1 = \Delta t$
- μ = expected rate of return per unit time for the stock. [Note: for Δt small we assume μ to be constant].
- σ = volatility of the stock. [Note: for Δt small we assume σ to be constant].
- $\varepsilon \sim N(0, 1)$

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ΔS by S or ΔS by S would basically, be equal to the drift rate plus the volatility, multiplied by the venna process, where ΔS is basically spot price or stock price difference between t_2 to t_1 , t_1 to t_0 , t_3 , t_2 , t_1 so on and so forth. Here μ is the expected rate of return per unit time for the stock. Note for Δt small t , we assume it to be constant. For other cases, it would be dependent on t , but in general we consider μ is not dependent on time. This is a very simplistic assumption. If we consider the

generalized venna process, we saw x was dependent on x and t , which is basically μ would have been dependent on S and t ; that means, the mean value would depend from where the spot stock price is as of now and how far you will go.

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Ito Process for Stock Prices

The discrete version of the model is:

$$\Delta S/S = \mu \cdot \Delta t + \sigma \cdot \epsilon \cdot (\Delta t)^{1/2}$$

where:

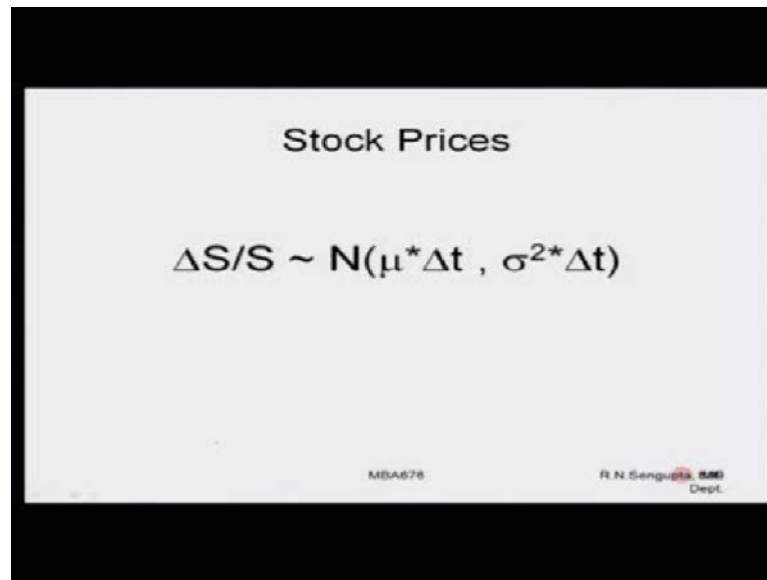
- $\Delta S = S(t=t_2) - S(t=t_1)$, such that $t_2 - t_1 = \Delta t$
- μ = expected rate of return per unit time for the stock. [Note: for Δt small we assume μ to be constant].
- σ = volatility of the stock. [Note: for Δt small we assume σ to be constant].
- $\epsilon \sim N(0, 1)$

Handwritten notes: $\mu \neq f(t)$, $\mu(x, t)$, $\sigma(x, t)$, $\mu(S, t)$, $\sigma(S, t)$, $\epsilon(S, t)$

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Sigma is the volatility of the stock. Note for again, for small delta time t volatility is independent of time. Again if you go back, we did consider b was dependent on x and t . If you consider this, sigma would be dependent on s and t and also finally, we did mention and this we will stick throughout the assumption, is that the epsilon would basically have a normal distribution with 0 mean time on standard deviation.

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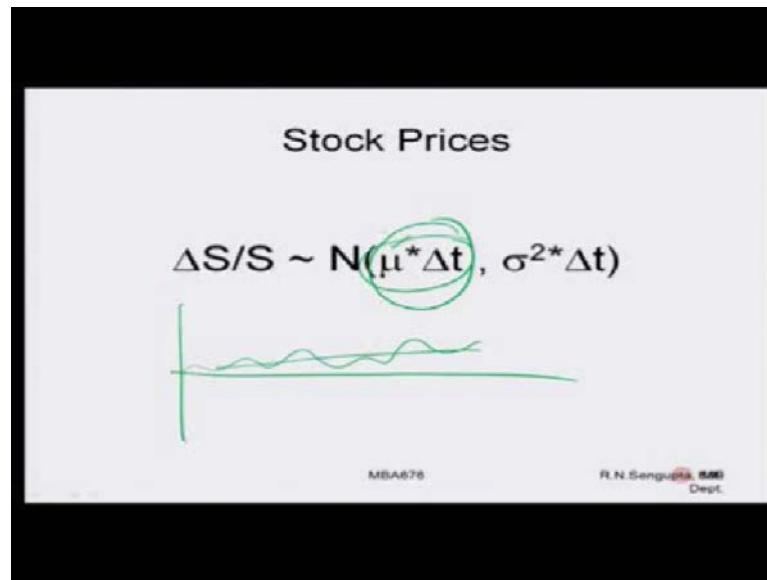
Stock Prices

$$\Delta S/S \sim N(\mu * \Delta t, \sigma^2 * \Delta t)$$

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So, now finally, once we are able to derive we will see that the rate of change of the stock price with respect to its present value would be a normal distribution. Here is what is the interesting part; this normal distribution mean and variances are not constant; they are changing with respect to time. So, the first term which is the mu value would be, the normal value would be, for the normal distribution the expected value is mu into delta t, which is the drift rate multiplied by the rate of change of time which you are considering, and in the other case, if you consider you will basically have the variability of the volatility given by sigma square into delta t, which means that if I am able to draw the functional form of delta s into s, the mean value would change.

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So, the rate of change per unit time is this, and if I basically consider delta t time t this is the total rate of change; total value change in the drift. If I consider sigma square is the white noise then for the time difference of delta t, from where we are and where you want to measure the volatility; it would be given by sigma square into delta t. Now pause; let us pause here and note. If you remember I did I have been mentioning in the last class time and again few times that the volatility is dependent on time either to a scale of delta t or to a scale of square root of delta t. Now, this is what is; if it is variance, it is dependent on delta t, hence variance could be additive. If it is basically standard deviation, it would be dependent on that function by square root of delta t, hence it is not additive.

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Monte Carlo Simulation

We can sample random paths for the stock price by sampling values for ε

Suppose $\mu = 0.14$, $\sigma = 0.20$, and $\delta t = 0.01$, then

$$\delta S = 0.0014 S + 0.02 S \varepsilon$$

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We can sample random parts for the stock market as I was discussing, considering that the values of mu is 0.14, sigma is 0.20 and delta t is given by 0.01, whatever unit of time. So, if we find out; if you take S down so, obviously, S would not be there, S would not be there. So, mu into delta t which is, this comes out to be 0.0014 and the second term was sigma into the epsilon into the square root of delta t. So, this you have multiplied by square root and it would be given by the case, multiplied by standard deviation of the epsilon.

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Monte Carlo Simulation

We can sample random paths for the stock price by sampling values for ε

Suppose $\mu = 0.14$, $\sigma = 0.20$, and $\delta t = 0.01$, then

$$\delta S = 0.0014 S + 0.02 S \varepsilon$$

Handwritten notes:
Under S in the first term: S
Under 0.02 in the second term: $\sigma \sqrt{\delta t}$
Under ε in the second term: ε

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So, what you are doing is b into z ; so, b is basically what you have is σ into the value of ϵ into square root of t ; this you know; this you know. This is generated and you can find out.

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Period	Stock Price at Start of Period	Random Sample for ϵ	Change in Stock Price, ΔS
0	20.000	0.52	0.236
1	20.236	1.44	0.611
2	20.847	-0.86	-0.329
3	20.518	1.46	0.628
4	21.146	-0.69	-0.262

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So, we have a taken a sample shot at time t is equal to 0; the stock price is 20. So, I randomly sample from ϵ 0 and I take a time, whatever the time zero zero 1.11 whatever, and change in the stock prices given. So, what I do is that given this price using this formula of μ into dt plus σ into ϵ into square root of Δt ; this you consider as constant with respect to time; this you consider constant with respect to time; find it out from the historical values. Δt can be whatever the time difference is. Here in this case, is 1 minus 0, 2 minus 1, is just a theoretical example. So, once you are able to do that you generate, and generate from a standard normal deviate. Once you find out, you find out the change of the price is, because on the right hand side what we actually are finding out.

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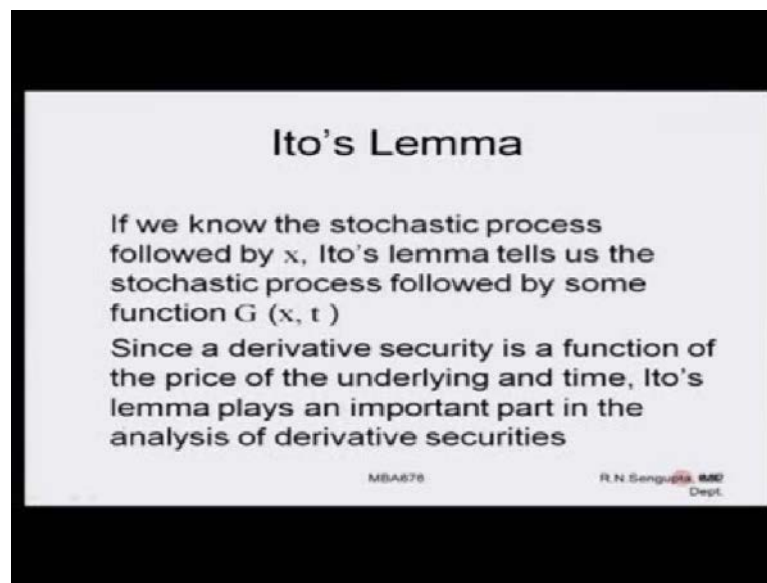
Monte Carlo Simulation – One Path

Period	Stock Price at Start of Period	Random Sample for ϵ	Change in Stock Price, ΔS
0	20.000	0.52	0.236
1	20.236	1.44	0.611
2	20.847	-0.86	-0.329
3	20.518	1.46	0.628
4	21.146	-0.69	-0.262

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So, this would be multiplied by s ; this would be multiplied by s ; this is Δs . So, once Δs is found out, this plus this would give you this. Then again generate, you would get this value. This plus this would give you this and you basically, bootstrap by side. That is 20.000 plus 0.236 would give you 20.236. Then again 20.236 plus 0.611 would give you 20.847. Again 20.847, now there is not a plus, because the random number generated ϵ is not now negative. So, 20 plus 20.847 minus 0.329 would now give you a new value of the stock point price, which is 20.518 and so on and so forth. So, these values which generate are dependent on the random sample, which you have generated from the ϵ . So, if you are able to generate another set; obviously, these values would change. So, what I meant was that if you do it a long time, infinite number of times, then the expected value of the change of x or Δx , t minus Δx 0 or x t and x 0 would be 0, because whatever fluctuation you have they would always add up to 0 in the long run.

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Ito's Lemma

If we know the stochastic process followed by x , Ito's lemma tells us the stochastic process followed by some function $G(x, t)$

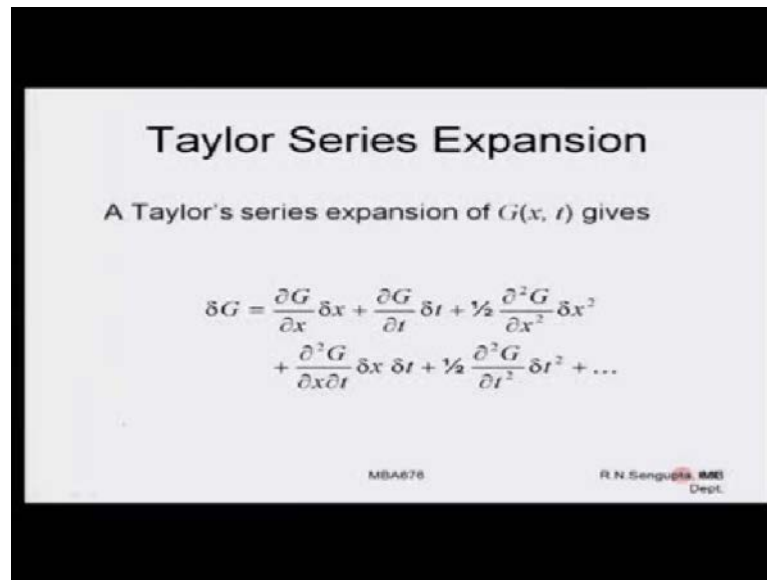
Since a derivative security is a function of the price of the underlying and time, Ito's lemma plays an important part in the analysis of derivative securities

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So, Ito's Lemma, if you know the stochastic process followed by x and which is the random variable, this Ito's Lemma tells us that the stochastic process would follow, will be followed by function, by some functional form of $g(x, t)$. Since a derivative security is a function of the price of the underlying security and the time, Ito's Lemma plays a vital role and an important role in trying to analyze the price of this derivative, which means that given the concept that they are Brownian motion and given the Ito's Lemma, it gives you of two information that how the rate of change of that function is happening with respect to the underlying the stock. It will give you a lot of information that how Ito's Lemma can be formulated in a very simple way to find out the price of that derivative, given the spot prices of the stocks are available to you.

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Taylor Series Expansion

A Taylor's series expansion of $G(x, t)$ gives

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} \delta x^2 + \frac{\partial^2 G}{\partial x \partial t} \delta x \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial t^2} \delta t^2 + \dots$$

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Now, consider a Taylor series expansion, in a very simple one of $G(x, t)$, where x is the place where you are and t is the time for that particular random variable. So, if we expand this, which is very simple. It is basically two variables. So, basically what I have; this is the first differentiate with respect to the first variable dx ; second differentiate, the first differential with respect to the second variable. So, if I go to the second derivative, second derivative would have basically three terms. The second derivative with respect to x twice; second derivative with respect to t twice and second derivative with respect to x , then t and then second derivative of respect to t and then x . So, what we would be doing is this.

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Taylor Series Expansion

A Taylor's series expansion of $G(x, t)$ gives

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} \delta x^2 + \frac{\partial^2 G}{\partial x \partial t} \delta x \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial t^2} \delta t^2 + \dots$$

$\frac{\partial^2 G}{\partial x \partial t} = \frac{\partial^2 G}{\partial t \partial x}$

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So, if you consider the function to be symmetric, they are equal. If they are equal, they are multiplied twice, hence there is no half or two terms here. So, this is the term and then if you consider then in the cubic form it will basically expand. So, cubic form what do we have? You have $\delta^2 G \delta x^3$ plus $\delta^2 G \delta t^3$ and you will take the values of x and t in this proportions. You will differentiate with x . Then differentiate twice with t or else you differentiate with t and then differentiate twice with x and basically have a combination of that and proceed. So, this is basically a simple multivariate Taylor series expansion.

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Ignoring Terms of Higher Order Than δt

In ordinary calculus we have

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t$$

In stochastic calculus this becomes

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} \delta x^2$$

because δx has a component which is of order $\sqrt{\delta t}$

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So, in ordinary calculus we have, if we are able to find out considering that we are able to ignore the third and higher terms. So, this is the first term. First term, because they are only first differentiate; obviously, there would be second term, second term differentiate. So, now if we bring into the picture and considering that the terms of delta t as t tends to 0 can be ignored. If you go back to the slides it will make sense.

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Taylor Series Expansion

A Taylor's series expansion of $G(x, t)$ gives

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} \delta x^2 + \frac{\partial^2 G}{\partial x \partial t} \delta x \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial t^2} \delta t^2 + \dots$$

$\frac{\partial^2 G}{\partial x \partial t} = \frac{\partial^2 G}{\partial t \partial x}$

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This cannot be ignored. This cannot be ignored, because delta t has some value, but now consider this term. What you are doing? You are trying to find out the differential of that particular function, twice with respect to t and then multiplying by a very small term, multiplied twice, which is here. So, obviously, it can be safely assumed to be 0 and also the interesting fact is that we would be differentiating the function with respect to two times. One with respect to delta x which is not that small and again, if you differentiate with respect to t, the smaller value; it will now tend towards 0, plus the fact it is being multiplied by delta t. So, technically even though theoretically, it may not be right, but practically, we will assume these values are almost tending to 0; that means, this last term and the second last term. So, if I highlight it, this is the values. These two terms are basically ignored.

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Ignoring Terms of Higher Order Than δt

In ordinary calculus we have

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t$$

In stochastic calculus this becomes

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} \delta x^2$$

because δx has a component which is of order $\sqrt{\delta t}$

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So, if you do that you will find out the question, is the first terms which are here and the second term is basically, coming from the second differentiate of the x term; nothing to do with t, because if you note; delta x is a component which is at order of square root of delta t, hence it can be ignored.

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Substituting for δx

Suppose ~~$\mu(S, t)$~~ $\sigma(S, t)$

$$dx = a(x, t)dt + b(x, t)dz$$

so that

$$\delta x = a \delta t + b \epsilon \sqrt{\delta t}$$

Then ignoring terms of higher order than δt

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} b^2 \epsilon^2 \delta t$$

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Now, suppose the rates which is mu, which we did consider that it is constant with respect to time if delta t is almost tending to 0 or it also said that sigma is a constant, if you consider rate with respect to t, where delta t is tending to 0; then if you are able to

expand that, then we had the simple equation where factor of time and x are not going to come, but if you consider the factors of time and space are going to come; equation is this. If you are, if we bring here this as mu S,t and this is sigma S, t, then they finding the equation finding the Taylor series expansion may be tricky, even though it may give you better results, but that would not be what the overall effort you are going to spend.

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Substituting for δx

Suppose $\mu(S, t)$ $\sigma(S, t)$

$dx = a(x, t)dt + b(x, t)dz$

so that

$$\delta x = a \delta t + b \epsilon \sqrt{\delta t}$$

Then ignoring terms of higher order than δt

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} b^2 \epsilon^2 \delta t$$

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So, now, the final equation which we have considering that a or mu, b or sigma, are not dependent on where you are and not dependent on time, the final equation, ignoring any second terms of delta t, square root or delta t, square root squares; actual equation is this, which means the partial differentiate equation, using the Taylor series expansion is given by this term.

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The $\epsilon^2 \Delta t$ term

Since $\epsilon \sim \phi(0,1)$ $E(\epsilon) = 0$ ✓

$E(\epsilon^2) - [E(\epsilon)]^2 = 1$

$E(\epsilon^2) = 1$ ✓

It follows that $E(\epsilon^2 \delta t) = \delta t$ ✓

The variance of δt is proportional to δt^2 and can be ignored. Hence

$$\delta G = \frac{\partial G}{\partial x} \delta x + \frac{\partial G}{\partial t} \delta t + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} b^2 \delta t$$

$\frac{\partial^2 G}{\partial x \partial t} \delta x \delta t$ X
 $\frac{\partial^2 G}{\partial t^2} \delta t^2$ X
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So, now, as epsilon we have considered is basically, standard normal deviate with 0 mean and one standard deviation, hence the mean value is 0, and if we consider the variance would also come up to 1 as you already considered. So, it follows that if we find out the expected value of epsilon square delta t, it would come out to delta t, because the respect of these two assumptions, which you have already noted down. The variance of delta t is proportional to delta t square and can be ignored, hence the final equation as we mentioned is this. This is with respect to x first differential. This is with respect to t first differential and this is respect to x, but the fact that higher term of this can be ignored and higher terms of this can be ignored. This is the case where you differentiate between x and then t or t to x and here, you are differentiating twice with t.

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Taking Limits

Taking limits $dG = \frac{\partial G}{\partial x} dx + \frac{\partial G}{\partial t} dt + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} b^2 dt$

Substituting $dx = a dt + b dz$

We obtain $dG = \left(\frac{\partial G}{\partial x} a + \frac{\partial G}{\partial t} + \frac{1}{2} \frac{\partial^2 G}{\partial x^2} b^2 \right) dt + \frac{\partial G}{\partial x} b dz$

This is Ito's Lemma

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If you take the limits, we obtain the final formula which is this, which is basically, the Ito's Lemma. So, now, what we need to do is that given the spot price is s , you are going to find out a derivative which is a functional spot price, depending on the values of μ and σ we can find out what is the actual price of that derivative.

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Application of Ito's Lemma to a Stock Price Process

The stock price process is $dS = \mu S dt + \sigma S dz$

For a function G of S and t

$dG = \left(\frac{\partial G}{\partial S} \mu S + \frac{\partial G}{\partial t} + \frac{1}{2} \frac{\partial^2 G}{\partial S^2} \sigma^2 S^2 \right) dt + \frac{\partial G}{\partial S} \sigma S dz$

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So, applications of Ito's Lemma towards stock price process stocks, price and its process; the stock price, as we already know is given by this drift term, volatility term. This is the vennis process; this s is the spot price. Now, if you bring it into picture for a functional

z , which is basically of s and t ; not x and t ; s is the spot price. The actual equation basically becomes this. So, now, you will basically have the drift rate. You will have the volatility rate here and you will basically, have the squared terms of the volatility depending on what is the stock price, but remember one thing; the fact that we have assumed a and b are independent of x and t or μ and σ are independent of s and t , even though are logical to serve and solve; in practical sense, they may not be true in the sense, that the assumptions we are taking as the μ on σ to be independent are not true.

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Examples

1. The forward price of a stock for a contract maturing at time T

$G = S e^{r(T-t)}$ $G = S e^{rfT}$
 $dG = (\mu - r)G dt + \sigma G dz$

2. $G = \ln S$

$dG = \left(\mu - \frac{\sigma^2}{2} \right) dt + \sigma dz$

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Now, if you consider the forward price of a stock for a contract, maturing at time t , is given by this; we already know. So, because it is American one and if you consider it is e to the power $r f$, which is a risk free interest rate into t into $s 0$. If you consider this G is a function of this, then if you solve it, the fraction becomes like this, which is rate of change of G is now being given by two terms, which is the excess rate of the particular stock over the risk free interest rate. If you remember we have considered that in some of the optimization problem and the volatility. So, now, it consists of two terms; excess rate over the risk free interest rate and the volatility rate and what are the other factors. One would be the time difference, which is Δt or t and one would be the simple venna process, which you have assumed underlying assumptions in the model.

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Examples

1. The forward price of a stock for a contract maturing at time T

$$G = S e^{r(T-t)}$$

$$dG = (\mu - r)G dt + \sigma G dz$$
2. $G = \ln S$

$$dG = \left(\mu - \frac{\sigma^2}{2} \right) dt + \sigma dz$$

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Now, if G is a function, if you solve it the overall function comes out to be this. Again, this is the average rate, but the average rate would now decrease by the term of sigma square by 2, with sigma is basically, the volatility and in the same way, the second term which is the white noise, comes continues to be remained as sigma.

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Black-Scholes Model

As $\frac{\Delta S}{S} = \mu \Delta t + \sigma \Delta z$, hence we have:

$$d \log_e S = (\mu - \sigma^2/2) dt + \sigma dz$$
 where we have assumed the stock prices are lognormally distributed, thus:

$$\log_e S_T \sim N[\log_e S_0 + (\mu - \sigma^2/2)T, \sigma^2 T]$$

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Now, consider that if we had the equation as given, the simple one, which is true; μ is the drift rate; sigma is the volatility or the white noise and on the left hand side, we have we have the rate of change, which is delta s into x, divided by s. Hence, if you are able to

solve it, it comes out to be this, where we have assumed that the stock prices are log normally distributed. If it is normally distributed the prices are given by this, where the log of the spot price as of today, would basically, be a normal distribution with two parameters, which is the mean and the standard deviation. So, the standard deviation, the variance continues to be sigma square into t; t is the time difference from where, you are and where you are trying to measure, and sigma square is the volatility per unit time or volatility is just constant, if delta t is very small. Delta t and this capital t are again, different; let me again mention. This is what we are considering and the mean value would basically, depend on the mean value as of now, plus a constant term and where is this term coming from?

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Examples

1. The forward price of a stock for a contract maturing at time T

$$G = S e^{r(T-t)} \quad G = S e^{rf}$$

$$dG = (\mu - r)G dt + \sigma G dz$$

2. $G = \ln S$

$$dG = \left(\mu - \frac{\sigma^2}{2} \right) dt + \sigma dz$$

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This was what we have considered in the last case; just the last case here; multiply it by again capital t. Again, capital t is given by this formula.

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**Assumptions for Black-Scholes
Merton model**

- 1) **The stock pays no dividends during the option's life:** Most companies pay dividends to their share holders. A common way of adjusting the model for this situation is to subtract the discounted value of a future dividend from the stock price.
- 2) **European exercise terms are used:** European exercise terms dictate that the option can only be exercised on the expiration date. American exercise term allow the option to be exercised at any time during the life of the option, making American options more valuable due to their greater flexibility. This limitation is not a major concern because very few calls are ever exercised before the last few days of their life. This is true because when you exercise a call early, you forfeit the remaining time value on the call.

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So, assumptions of the Black-Scholes; we have been able to at least, get a feel that how the stock prices using Ito's Lemma, the venna process, the generalized venna process and the assumptions will consider about the volatility and the drift rate. So, based on the fact we have been able to simply, derive and understand the Black-Scholes model, but let us first, also try to analyze what are the important assumptions based on this model, was developed. The assumptions are very simple.

The stock pays no dividends during the option's life. If you remember we did consider this time and again, that all this is no dividend; that means, the value of I_0 or any independent payment, which is being made and they are being calculated as of now, would not be considered. Most companies pay dividends to their shareholders. A common way of adjusting the model for this situation is to subtract the discounted value of the future dividend from the stock price as of now, which is S_0 minus I_0 , but we are not going to do that in the simple calculation of the Black-Scholes model.

Second one is European exercise terms are used; it means, if you remember the time to expiration for the European one was exactly fixed at time t is equal to capital t . Those are not American options. It says that European exercise terms dictates that the option can only be exercised on the expiration date. American exercise terms allows the options to be exercised on any time during that period of time before it expires. So, this simulation is not a major concern, but it can be solved accordingly.

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Assumptions for Black-Scholes Merton model

3) **Markets are efficient:** This assumption suggests that people cannot consistently predict the direction of the market or an individual stock. The market operates continuously with share prices following a continuous Itô process. To understand what a continuous Itô process is, you must first know that a Markov process is "one where the observation in time period t depends only on the preceding observation." An Itô process is simply a Markov process in continuous time. If you were to draw a continuous process you would do so without picking the pen up from the piece of paper.

4) **No commissions are charged:** Usually market participants do have to pay a commission to buy or sell options. Even floor traders pay some kind of fee, but it is usually very small. The fees that individual investor's pay is more substantial and can often distort the output of the model.

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We will also consider markets are efficient; that means, the demand and supply of the prices are such that it would basically, be able to give the perfect information of the spot prices are being dictated. There is no commission charge and also we will consider there are no taxes; that means, if you are selling and buying a particular option or a stock in the market, you are using the broker self. You will consider that you are not either paying revenue amount to the broker for buying and selling this different types of stock and option.

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Assumptions for Black-Scholes Merton model

5) **Interest rates remain constant and known:** The Black and Scholes model uses the risk-free rate to represent this constant and known rate. In reality there is no such thing as the risk-free rate, but the discount rate on Government Treasury Bills with 30 days left until maturity is usually used to represent it. During periods of rapidly changing interest rates, these 30 day rates are often subject to change, thereby violating one of the assumptions of the model.

6) **Returns are lognormally distributed:** This assumption suggests, returns on the underlying stock are normally distributed, which is reasonable for most assets that offer options.

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We will finally, consider the interest rates remain constant; that means, the μ value or the σ value which are constant, will always continue to be constant with respect to time. This, I had mentioned time and again when you are deriving, and the returns are log normal distributed, such that based on the fact, the derivation which you did for the using the Ito's Lemma for the Black-Scholes model, continues to hold true.

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Black-Scholes Merton differential equation

$$\frac{\partial f}{\partial t} + r_f S \frac{\partial f}{\partial S} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 f}{\partial S^2} = r_f f$$

Where:

- f : Value of the Forward contract
- S : Stock price
- r_f : Risk free interest rate
- t : time
- σ : Volatility of the underlying Asset

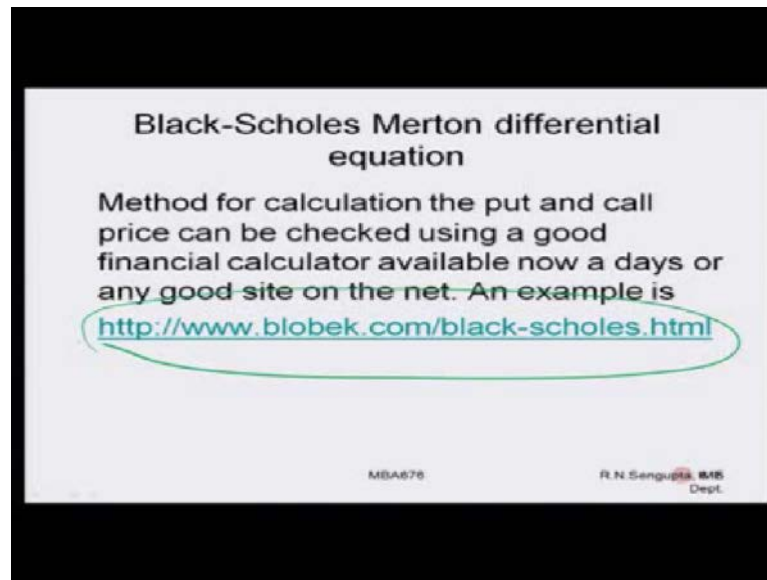
The model describes the evolution of the price of an option over time, as a function of time and the value of the underlying. The assumption of normality is very important as it leads to underpricing in non-gaussian situations. Must be used with caution.

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Handwritten notes: $G(S, t)$ and μ, r_f

So, finally, we wrap up this class. We will take another two minutes. The final equations for the Black-Scholes model would be given by this, where f is basically, the corresponding value of the forward contract, which we are discussing in the capital f or the small f , which is dependent of risk free interest rate and the spot price and the differential equation, which we have already got. So, G was basically the function which was dependent of s and t and based on that, we found out and μ was the rate of change of the function, and r_f was the risk free interest rate. So, we are basically trying to find out with respect to how big or small of the value of the rate of change of the function is with respect to r_f . Based on that we will try to invest into that derivative. So, s is the stock price; r_f is the risk free interest rate; t is the time and the volatility of the underlying asset is given, which is the actual stock we are trying to buy and so, remember one thing; that r_f is constant as μ and σ are constant.

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So, method for calculation for the put and call option can be found out using any financial calculators which you have, and a good site for an example in the net, can be found out here. So, we will continue our discussion with other models later on and then, come back and trying to explain that how the Black-Scholes model can be utilized in trying to find out the stock the prices of that options based on the spots, which are available.

Thank you very much.