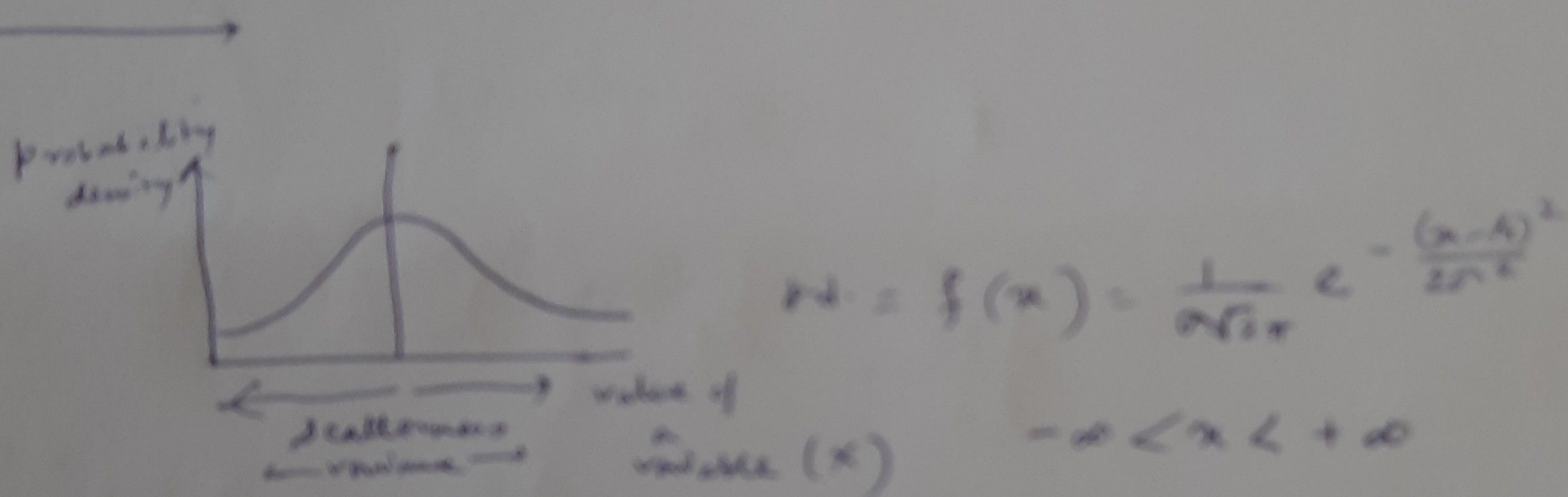


3

If we assume stock price = 100 INR (₹)
the simulation of the stock price will be

| Stock price | ' ϵ ' (random sample) | ΔS (change in stock price) |
|-------------|--------------------------------|---------------------------------------|
| 100.00 | 0.52 | 2.45 |
| 102.45 | 0.44 | 6.43 |
| 108.88 | -0.86 | -3.58 |
| 105.30 | 1.46 | 6.70 |
| ... | ... | ... |



$$\int_{-\infty}^{+\infty} f(x) dx = 1 \quad \text{(Total area under curve)}$$

when we are generating ' ϵ ' we have to draw random numbers b/w 0 and 1 first and then inverse cumulative normal distrib. afterward.

ϵ follows a p.d.f of a standard normal distrib.

$$\epsilon \sim N(0, 1)$$

Mean 0 and variance 1

(4)

Application of Itô's Lemma:-

we saw $S_T = S_0 e^{rT}$ (under constant drift assumption)

$\therefore F_0 = S_0 e^{rT}$ (r is risk free interest rate)
 F is forward ~~rate~~ price at $t=0$)

$S_0 =$ Spot price.

$T =$ time to maturity.

If we take gap/interval of time
 b/w t and T then

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

$$\frac{\partial F}{\partial S} = e^{r(T-t)} \quad \frac{\partial^2 F}{\partial S^2} = 0$$

$$\frac{\partial F}{\partial r} = -r S e^{r(T-t)}$$

\therefore Using Itô's Lemma,

$$dF = \left[e^{r(T-t)} \mu S - r S e^{-r(T-t)} \right] dt + e^{r(T-t)} \sigma S dz$$

If we use $F = S e^{r(T-t)}$,

$$dF = (\mu - r) F dt + \sigma F dz$$

F (forward price) follows a Geometric Brownian motion with $(\mu - r)$, σ .

Process of a stock price:-

A stock price follows a generalized Wiener process.

i.e. it has a constant drift rate and a constant variance rate.

However, this type of model fails to capture a key aspect of stock price movement.

people look at return rather than absolute value. (think of GDP growth rate!)

If we assume the drift rate is constant then

$$\Delta S = \mu S \Delta t \quad (\text{where } \mu \text{ is the constant drift})$$

here we believe there is no volatility term.

As $\Delta t \rightarrow 0$ then

$$\frac{dS}{S} = \mu dt$$

or % change in $S = \mu dt$ (if multiplied by 100)

$$\therefore \int \frac{dS}{S} = \mu \int dt$$

$$\ln S = \mu t + \text{constant of integration}$$

$$\text{or } S_T = S_0 e^{\mu T} \quad (\text{between } t=0 \text{ and } T)$$

But in reality stock price exhibits volatility

$$\therefore dS = \mu S dt + \alpha S dz$$

where α is volatility (sp) factor

$$\therefore \frac{dS}{S} = \mu dt + \alpha dz$$

$\left\{ \begin{array}{l} z \text{ follows Wiener process} \\ \therefore \Delta z = \epsilon \sqrt{\Delta t} \end{array} \right.$

(2)

Monte Carlo (named after city of Casino)

$$\frac{ds}{s} = \mu dt + \sigma dz$$

For a ~~small~~ time interval Δt the change in stock price ΔS will be

$$\frac{\Delta S}{S} = \mu \Delta t + \sigma \Delta z$$

Let's assume $\begin{cases} \mu = 0.15 \\ \sigma = 0.3 \end{cases}$ (From past data of stock we calculate historical figures)
 \downarrow
 Expressed in annual % form.

$$\therefore \frac{\Delta S}{S} = 0.15 \Delta t + 0.30 \epsilon \sqrt{\Delta t}$$

$\epsilon \sim N(0,1)$, Let's assume a time interval of one week

one week = 0.0192 year.

$$\therefore \Delta t = 0.0192.$$

$$\begin{aligned} \therefore \Delta S &= (S \times 0.15 \times 0.0192) + (S \times 0.30 \times \sqrt{0.0192}) \epsilon \\ &= (0.00288)S + (0.0416)S \epsilon \end{aligned}$$

Now, ϵ 's value can be attained/simulated from repeated sampling.

In 'excel' we can artificially generate ' ϵ ' values -

$$= \text{NORMSINV}(\text{RAND}())$$

→ 0.52
 → 0.44
 → -0.88
 → 1.46