

Unit-3

* σ-field: F (a algebra) is a family of subset of Ω (sample space) which satisfies following properties:

- ① $\emptyset \in F$
- ② If $A \in F$ then $A^c \in F$
- ③ If A_1, A_2, \dots are in F and is countable sequence, then $\bigcup_{i \in \mathbb{N}} A_i \mid \sum_{i=1}^n A_i \in F$

e.g. A coin is tossed 3 times

$$\Omega = \{ HHH, HHT, HTH, HTT, THH, THT, TTH, TTT \}$$

$$A_1 = \{ HHH, HHT, HTH, HTT \} \quad (\text{first coin is head})$$

$$A_2 = \{ THH, THT, TTH, TTT \} \quad (\text{" " " tail})$$

$$F_1 = \{ \emptyset, \Omega \}$$

$$F_2 = \{ \emptyset, \Omega, A_1, A_2 \}$$

$$A_{11} = \{ HHH, HHT \}$$

$$A_{12} = \{ HTH, HTT \}$$

$$F_3 = \{ \emptyset, \Omega, A_{11}, A_{12}, A_2, A_{11} \cup A_2, A_{11} \cup A_{12} \}$$

~~66₂~~
~~X₂~~
~~66₂~~
~~1265x6₃~~
~~265x2~~

$$A_{111} = \sum HHH \Sigma$$

$$A_{112} = \sum HHHT \Sigma$$

$$\begin{aligned}
 F_4 &= \left\{ \emptyset, \Omega, A_{111}, A_{112}, A_{12}^c, \right. \\
 &\quad (A_{111} \cup A_{112} \cup A_{12} = A_{11}^c), \\
 &\quad (A_{111} \cup A_{112} \cup A_{12} = A_{12}^c), \\
 &\quad \left. (A_{111} \cup A_{112} \cup A_{12} = A_2^c) \right\} \\
 &\xrightarrow{* \text{ Confirm}} (A_{111} \cup A_{112} \cup A_{12} = A_{12}^c)
 \end{aligned}$$

~~① A die is thrown once such that it contains 3 disjoint sets with σ -field on 3 disjoint subsets,~~
~~How many σ -fields can we generate.~~

Avg

Stochastic Process
 \Rightarrow let (Ω, \mathcal{F}, P) where P is the probability field
 measure defined on \mathcal{F} be a given probability space,
 a collection of random variables $\{X_t, t \in T\}$
 T = index set / time index defined on the
 probability space (Ω, \mathcal{F}, P)
 is called a Stochastic process.

The stochastic process is function of two variables which are independent.

$$\{x_t(\omega), t \in T, \omega \in \Omega\}$$

$$X: T \times \Omega \rightarrow \mathbb{R}$$

$x(\cdot, \omega) \rightarrow$ t is varying, fixed ω ,
 $x(t, \omega)$ the trajectory be called sample path

↳ fixed t in a random variable.

let $\{x_t, t \in T\}$ be a given stochastic process
the set $\{t \in T\}$ is called a parameter space.
over index set the collection of all possible values of
 $x_t + t \in T$ is called state space.

→ Discrete time Discrete Space

→ Continuous time Discrete Space

→ Discrete time Continuous Space

→ Cont. time, " "

whenever discrete state space/parameter space is finite or
countably infinite then it is said to have discrete
nature. When it is represented on ~~number~~ real line
it is continuous.

→ independent increment

for all n and $t_1 < t_2 < \dots < t_n$

the random variable

$$X(t_2) - X(t_1), X(t_3) - X(t_2), \dots, X(t_n) - X(t_{n-1})$$

are independent R.V. Then the process is said to ~~be~~ have independent increment.

⇒ strict sense stationary process (Strong Stationary process)

The stochastic process $\{X(t) : t \geq 0\}$ is a SSSP if for arbitrary $0 < t_1 < t_2 < \dots < t_n$

the finite dimensional random vector

$$(X(t_1), X(t_2), \dots, X(t_n))$$

$$(X(t_1+h), X(t_2+h), \dots, X(t_n+h))$$
 have

same joint distribution for all $h > 0$

and all t_1, t_2, \dots, t_n

\Rightarrow Wide sense stationary process

The SP $\{x(t), t \in \mathbb{R}\}$ is WSSP if it satisfies the following ~~known~~ conditions.

- (i) $E(x(t)) = \mu(t)$ is independent of t
- (ii) $\text{Cov}(x(t), x(s))$ depends only on the time diff. $|t-s|$ for all t and s
- (iii) $E(x^2(t)) < \infty$ (~~has~~ finite second order moment)

A WSSP is also called covariance stationary or weak stationary or second order stationary

$$x(t) = A \cos \omega t + B \sin \omega t$$

where A and B are uncorrelated R.V with

$$\begin{cases} E(A) = E(B) = 0 \\ \text{Var}(A) = \text{Var}(B) = 1 \end{cases}$$

ω is +ve constant

$$\begin{aligned} E(x(t)) &= \cos \omega t E(A) + \sin \omega t E(B) \\ E(x(t)) &= 0 \quad \checkmark \text{ independent of } t \end{aligned}$$

$$\text{Cov}(x_t, x_s) = E(x_t \cdot x_s)$$

$$= E\{(A \cos \omega t + B \sin \omega t) \cdot (A \cos \omega s + B \sin \omega s)\}$$

$$(A \cos \omega s + B \sin \omega s)$$

$$= \cos \omega t \cos \omega s E(A^2) + \sin \omega t \sin \omega s E(B^2)$$

$$+ \sin(\omega t + \omega s) E(AB)$$

$$= \cos(\omega t - \omega s), \quad t > s$$

$$= \cos(\omega(t-s))$$

↳ func. of $t-s$

(iii) $E(x^2) - E(x)^2 = \text{Var}(x)$

↓
∅

$$E(x^2) = \text{Var}(x)$$

$$E(x^2(t)) = \text{Var}(x(t)) < \infty$$

Conditional Expectation

= conditional distribution of X given

$$F(X=x|Y=y) = P\{X \leq x | Y = y\}$$

\Rightarrow Conditional expression of X

$$E(X|Y=y) = \int_{-\infty}^{\infty} x dF(x|Y=y)$$

If X and Y are R. V, then $E(X|Y)$

$$E(X) = \sum_y E(X|Y=y) \cdot P(Y=y)$$

{from Brown Notes}

for ball pen

2 refills are selected at random from a box containing 3 blue, 2 red and 3 green refills

$x = \text{no. of blue refill selected}$

$y = \text{no. of red " " }$

(a) Find joint prob. distribution.

(b) The ~~prob. of $(x,y) \in A$~~

(b) $P\{(x,y); (x,y) \in A\}$ where $A: x+y \leq 1$

(c) Expected value $E(x|y=1)$

(d) Conditional dist. of x given $y=1$

Ans. (a) $R_x = \{0, 1, 2\}$

$$R_y = \{0, 1, 2\}$$

$x \setminus y$	0	1	2	$n(x)$
0	$3/28$	$6/28$	$1/28$	$10/28$
1	$9/28$	$6/28$	0	$15/28$
2	$3/28$	0	0	$3/28$
$n(y)$	$15/28$	$12/28$	$1/28$	1

(b) $A = \{(0,0), (0,1), (1,0)\}$

$$P\{(x,y) \in A\} = P(0,0) + P(0,1) + P(1,0) \\ = \frac{18}{28}$$

(d) $P(x|y=1) = \overbrace{P(x|y=1)}^{\text{mass function}}$

$$= \frac{P(x,y)}{P(y)} = \frac{P(x,y=1)}{P(y=1)} = \frac{28}{12} P(x,1)$$

$$b(x|y=1) = \frac{28}{12} P(x,1)$$

$$b(0|y=1) = \frac{28}{12} P(0,1) = \frac{6}{12} = \frac{1}{2}$$

$$b(1|y=1) = \frac{28}{12} P(1,1) = \frac{6}{12} = \frac{1}{2}$$

$$b(2|y=1) = \frac{28}{12} P(2,1) = 0$$

$$(c) E(x|y=1) = \sum_x x b(x|y=1)$$

$$= 0 \cdot b(0|y=1) + 1 \cdot b(1|y=1) + 2 \cdot b(2|y=1)$$

$$\boxed{E(x|y=1) = \frac{1}{2}}$$

Thm. Let a stochastic process $\{S_k\}; k=0, 1, 2, \dots$

be a asymmetric random walk. Then,

(i) ~~for each k~~ : $E(S_k) = 0; \text{Var}(S_k) = k$

(ii) ~~it has~~ independent increment

(iii) ~~it has~~ stationary

(iv) It is a markov process

Proof (i)

$$S_k = \sum_{j=1}^k X_j$$

$$E(S_k) = E(\sum X_j) = \sum E(X_j)$$

$$\because E(X_j) = 0 ; \text{Var}(X_j) = 1$$

$$\boxed{E(S_k) = 0}$$

$$\begin{aligned}\text{Var}(S_k) &= \text{Var}(\sum_{j=1}^n X_j) \\ &= \sum_{j=1}^k \text{Var} X_j \\ &= k\end{aligned}$$

(ii) we choose an arbitrary integer n ,
and then choose non-negative integers

then $S_{k_{i+1}} - S_{k_i} = \sum_{j=k_i+1}^{k_{i+1}} X_j$

Then

$$S_{k_{i+1}} - S_{k_i} = \sum_{k_i+1}^{k_{i+1}} X_j$$

Since X_j are iid R.V having
bernoulli distribution

$\therefore S_{k_i} - S_{k_0}, S_{k_2} - S_{k_1}, \dots, S_{k_n} - S_{k_0}$
are mutually independent

Hence the SP $\sum S_k$; $k=0, 1, 2, \dots, n$ has independent increment.

Class NotesProof :Since for every t

$$S(t) = S(0)e^{H(t)} \quad \text{--- (A)}$$

$$H(t) = \mu(t) + \sigma H(T)$$

is normally distributed

$$\mu(t) \sim N(\mu t, \sigma^2 t)$$

$$\begin{aligned} E(S(t)) &= E(S(0)e^{H(t)}) \\ &= S(0) \otimes E(e^{H(t)}) \\ &= S(0) e^{(\mu + \frac{1}{2}\sigma^2)t} \\ &= S(0) \exp[(\mu + \sigma^2/2)t] \end{aligned}$$

$$\begin{aligned} \text{Var}(S(t)) &= E(S(t)^2) - [E(S(t))]^2 \\ &= E(S^2(0) e^{2H(t)}) - E[S(t)]^2 \\ &= S^2(0) \otimes E(e^{2H(t)}) - " \end{aligned}$$

$$\boxed{\begin{aligned} E(e^{H(t)}) &= e^{(\mu t + \frac{1}{2}\sigma^2 t \otimes^2)} \\ &= S^2(0) e^{2\mu t + 2\sigma^2 t} - S^2(0) e^{(2\mu t + \sigma^2 t)} \end{aligned}}$$

$$\text{Var}(S(t)) = [S(0) \otimes e^{(\mu + \sigma^2/2)t}]^2 (\exp(\sigma^2 t) - 1)$$

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Filtration :

let ω be the sample space and

$F_0 = \{\emptyset, \Omega\}$ then a filtration

in discrete time is an increasing

sequence of sigma-fields $F_0 \subset F_1 \subset F_2 \subset \dots \subset F_n$
of sigma field per time instance.

Eg : $\Omega = \{HHH, HHT, THH, TTH, \cancel{HTH}, TTT, THT, HTT, HTH\}$

$$F_0 = \{\emptyset, \Omega\}$$

$$F_1 = \{\emptyset, E_H, E_T, \Omega\}$$

$$F_2 = \{\emptyset, E_{H_1}, E_{H_2}, E_T, E_H, E_H \vee E_T,$$

$$E_{H_2} \vee E_T, \Omega\}$$

\hookrightarrow check from book

$$F_0 \subset F_1 \subset F_2$$

Let ω be the sample space and T be a fixed positive number. and assume that

for $t \in [0, T]$ there is sigma field F_t

Assume further that if $s \leq t$ then

every set of F_s is in F_t . then the collection
of sigma field $\{F_t ; 0 \leq t \leq T\}$

is called a filtration in continuous time.

Sigma field generated by SP

Let $\{Y(t), t \in T\}$ be a given SP then sigma field generated by this SP is the smallest sigma field containing all sets in the form $\{\omega : \text{the sample path } Y(t), t \in T, \text{ belong to } C\}$.

for all suitable sets C of the functions are T .

X_t is F_t -measurable let F_t is a sigma field (a field) then the random variable X_t is F_t -measurable if every ?

$\Rightarrow A = R.V$ is F_t -measurable if and only if the info in F_t is sufficient to determine the value of X_t .

Adopted Process

A sequence of RV X_1, X_2, \dots are said to be adopted to a filtration

f_1, f_2, \dots if X_n is F_n -measurable for each $n = 1, 2, \dots$

Martingales

↓ field

Let (Ω, \mathcal{F}, P) be a probability space and let $\{X_n, n=0, 1, 2, \dots\}$ be a SP and let $\{\mathcal{F}_n, n=0, 1, 2, \dots\}$ be the filtration.

So the SP $\{X_n, n=0, \dots\}$ is said to be a martingale corresponding to the filtration if it satisfies the following condition for every n .

- (i) $E(X_n)$ exists. (and finite)
- (ii) Each X_n is \mathcal{F}_n -measurable $\forall n$
- (iii) $\forall n$, $E(X_{n+1} | \mathcal{F}_n)$ must be equal to X_n

$$E(E(X|Y)) = E(X)$$

$E(X_{n+1} | \mathcal{F}_n) > X_n$ then it is

Sub-martingale

if $E(X_{n+1} | \mathcal{F}_n) < X_n \Rightarrow$ Super-martingale

Ex Let x_1, x_2, \dots be a seq. of iid R.V each taking values +1 and -1 with equal probability.

Let us define $s_0 = 0$, $s_n = \sum_{i=1}^n x_i$, $n = 1, 2, \dots$

This ~~discrete~~ SP is simple random walk

$\{s_n\}$

Solⁿ

$$\begin{aligned} E(s_{n+1} | x_n) &= E(s_n + x_{n+1} | x_n, x_{n-1}, \dots, x_1) \\ &= s_n + E(x_{n+1} | x_n, \dots) \\ &= s_n + E(\overset{\circ}{x}_{n+1}) \end{aligned}$$

$\left\{ x_n \text{ is iid} \right\}$

$$\boxed{E(s_{n+1} | x_n) = s_n}$$

$\therefore s_n$ is martingale

Q

Consider a symmetric walk
 $\{S_n, n=0, 1, 2, \dots\}$ which is
 a martingale with respect to
 a filtration $\{F_n, n=0, 1, 2, \dots\}$
 where $F_0 = \{\emptyset, \Omega\}$ and
 $F_n = \mathcal{F} \cap \{x_1, x_2, \dots, x_n\}; n \geq 1$

verify if $\{S_n^2, n=0, 1, 2, \dots\}$ is a
 martingale w.r.t. filtration F_n

Sol $E(S_n^2) = (x_1 + x_2 + x_3 + \dots + x_n)^2$

$$\begin{aligned} E(S_n^2) &= E(x_1^2 + x_2^2 + \dots) \\ &= E(x_1^2) + E(x_2^2) + \dots \\ &\quad + 2 \sum E(x_i x_j), \quad i \neq j \end{aligned}$$

$$E(S_n^2) \Rightarrow \text{exists } S_n^2 = \{ \dots \}$$

$$E(S_{n+1}^2 | F_n) = E((S_n + x_{n+1})^2 | F_n)$$

$$= E(S_n^2 | F_n) + 2 E(S_n x_{n+1} | F_n)$$

$$+ E(x_{n+1}^2 | F_n)$$

$$= S_n^2 + 2 S_n E(x_{n+1} | F_n)$$

$$+ E(x_{n+1}^2)$$

$$E(S_{n+1}^2 | F_n) = 1 + S_n^2 \neq S_n^2$$

not martingale

01/04/20

Hö - doeblin formula for B.M (first)

$\{N(t), T > \bar{\omega}\}$ is a h with parameter 1. Decide
 poisson process
 the nature if martingale or not.

Ans

$$[w(t) - w(s)]$$

\mathcal{F}_t $F_n \rightarrow$ filtration

- 1) for every n , $E(x_n)$ exists
- 2) Each x_n is F_n - measurable
- 3) For every n , $E(x_{n+1}/F_n) = x_n$

$$1) x_n = x(t)$$

$E(x(t))$ is F_t - measurable

x_t

1)

2) Each $x(t)$ is F_t - measurable

3) for every $0 < s < t$ $E(x(t)/F_s) = x(s)$

Let $f(x, t)$ is a function of $x \neq t$ have continuous partial derivative of at least second order. s.t. $\{w(t), w(0) \in t > 0\}$ be a process then, $\frac{d}{dt} f(x)$

$$df(t, w(t)) = \int_0^t b_t(t, w(t)) dt$$

$$+ \int_x(t, w(t)) d(w(t))$$

$$+ \frac{1}{2} \int_{xx}(t, w(t)) dt$$

-③

where $x = w(t)$

$$\int_0^T df(t, w(t)) = \int_0^T b_t(t, w(t)) dt + \int_0^T \int_x(t, w(t)) dw(t)$$

$$+ \frac{1}{2} \int_0^T \int_{xx}(t, w(t)) dt$$

↓

$$f(T, w(T)) - f(0, w(0)) = " "$$

$$f(t, x) = x^2/2, \quad x = w(t)$$

$$b_t = 0$$

$$b_{xx} = 1$$

$$b_x = x$$

$$\frac{\omega^2(T)}{2} - \sigma = \int_0^T dt + \int_0^T \omega(s) d(\omega(s)) + \frac{1}{2} \int_0^T 1 \cdot dt$$

$$\frac{\omega^2(T)}{2} \approx$$

$$\boxed{\int_0^T \omega(s) d(\omega(s)) = \frac{\omega^2(T) - T}{2}}$$

Stochastic differential equation

$$\frac{dX(t)}{dt} = b(t, X(t)), \quad t \in [0, T] \\ X(0) = x_0$$

$$b : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$$

and b is a continuous func.

then this diff. eqn has a solution :

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

* provided the ~~Lipschitz's~~ condition by the function

$$b : \exists K > 0 \text{ s.t.}$$

$$|b(t, x) - b(t, y)| \leq K|x - y| \quad \forall t \in [0, T] \text{ s.t. } x, y \in \mathbb{R}$$

Now suppose $X(0)$ or b is random then solution is not unique, rather it will depend upon $w \in \Omega$ and it will be different random differential eqn.

Adding an uncertainty

$$\frac{dX(t)}{dt} = b(t, X(t)) + a(t, X(t)) \frac{dw(t)}{dt} - 0 \quad (0 \leq t \leq T)$$

$$dX(t) = b(t, X(t)) dt + a(t, X(t)) dw(t) \quad -(2)$$

$$b(t, X(t)) : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$$

$$a(t, X(t)) : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$$

b and a are two given functions

eq (2) is stochastic diff. eqn

$$\int_0^T dX(t) = \int_0^T b(t, w(t)) dt + \int_0^T a(t, w(t)) dw(t)$$

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$$x(T) - x(0) =$$

- (3)

Diffusion process

is a solution of stochastic diff. eqn (strong or weak) is called diffusion

existence theorem

Let $E(x^2(0)) < \infty$ and $x(0)$ be independent of $\{w(t), t \geq 0\}$ for $t \in [0, T]$ and $x, y \in \mathbb{R}^n$. $b(t, x), a(t, x)$ be continuous function and satisfies the lipchitz condition w.r.t 2nd variable

$$|b(t, x) - b(t, y)| + |a(t, x) - a(t, y)|$$

$$\leq \kappa k|x - y|$$

$$+ t \in [0, T]$$

and K is a constant. Then the stochastic diff. eqn has a unique strong solution

$$\{x(t); T \geq t \geq 0\}$$

04/04/24

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

$$X(t) - X(0) = \int_0^t b(u, X(u))du + \int_0^t \alpha(u, X(u))dW(u)$$

Q $\sin(\omega(t))$, $d\int_0^t b(u, \omega(u))du = \int_0^t dt + \int_0^t b_x d\omega(u)$
 $+ \frac{1}{2} \int_0^t b_{xx} du$

A let $b(t, x) = \sin x$

when $x = \omega(t)$

$$\int_0^t dt = 0$$

$$\int_0^t x = \cos x$$

$$\int_0^t x dx = -\sin x$$

Substituting in eco-doubtate formula,

$$d\sin(\omega(t)) = -\frac{1}{2} \sin(\omega(t))dt + \cos(\omega(t)) + d(\omega(t))$$

Q SDE : $w^3(t)$

$$b(t, x) = x^3, x = \omega(t)$$

$$\int_0^t dt = 0$$

$$\int_0^t x = 3x^2$$

$$\int_0^t x dx = 6x^3$$

$$dw^3(t) = 3\omega(t)dt + 3w^3(t) + d(\omega(t))$$

SDE of GBM

let $S_{\text{GBM}}(t)$ be the stock price at time t and μ ($-\infty < \mu < \infty$) is the constant growth rate and $\sigma > 0$ be the volatility.

then SDE

$$dS(t) = \mu S(t) dt + \sigma S(t) dW(t) \quad (1)$$

$S(0)$ is known

* $S(t) = S(0) e^H$

$$H = \mu t + \sigma W(t)$$

$$\ln(S(t)) = \ln(S(0)) + \mu t + \sigma W(t)$$

$$\frac{d(S(t))}{S(t)} = 0 + \mu dt + \sigma dW(t)$$

$$d(S(t)) = \mu S(t) dt + \sigma S(t) dW(t)$$

let $S(t) = f(t, \omega(t))$

$$df(t, \omega(t)) = \left(f_t + \frac{1}{2} f_{xx} \right) dt + f_x d(\omega(t)) \quad (2)$$

$$f_t + \frac{1}{2} f_{xx} = \mu f = \mu b \quad (3)$$

$$f_x = \sigma f(t) = \sigma b \quad (4)$$

$$\Rightarrow b(t, x) = A(t) e^{\alpha x} \quad \text{--- (5)}$$

$$(A'(t) + \frac{1}{2} \alpha^2 A(t)) e^{\alpha x} = \mu A(t) e^{\alpha x}$$

$$A'(t) e^{\alpha x} = (\mu - \frac{\alpha^2}{2}) A(t) e^{\alpha x}$$

$$A(t) = C e^{(\mu - \frac{\alpha^2}{2}) t} \quad \text{--- (6)}$$

arbitrary constant

Substituting value of $A(t)$ from (6) and (5)

$$b(t, x) = C \cdot e^{(\mu - \frac{\alpha^2}{2}) t} \cdot e^{\alpha x}$$

at $t = 0$)

$$C = b(0, x) = S(0)$$

put $x = \omega t$)

$$b(t, x) = S(0) e^{(\mu - \frac{\alpha^2}{2}) t + \alpha x}$$

$$S(t) = b(t, \omega(t)) = S(0) e^{(\mu - \frac{\alpha^2}{2}) t + \alpha \omega(t)}$$

$$(1) E(S(t)) = S(0) e^{\mu t}$$

$$(2) E(S^2(t)) = S^2(0) e^{(2\mu t + \alpha^2)t}$$

$$(3) \text{Var}(S(t)) = E(S^2(t)) - [E(S(t))]^2$$

$$= S^2(0) e^{2\mu t} [e^{\alpha t} - 1]$$

Proof

$$\begin{aligned}
 E(S(t)) &= E(S(0) \cdot e^{(\mu - \sigma^2/2)t + \sigma \omega(t)}) \\
 &= S(0) E\left(e^{(\mu - \sigma^2/2)t + \sigma \omega(t)}\right) \\
 &\quad \xrightarrow{\omega \sim N((\mu - \sigma^2/2)t, \sigma^2 t)} \\
 E(e^x) &= e^{(\mu - \sigma^2/2)t + \frac{1}{2}\sigma^2 t} \\
 &= e^{\mu t}
 \end{aligned}$$

$E(S(t)) = S(0) e^{\mu t}$

① $\{S(t), t \geq 0\}$ follows GBM with
 drift $(\mu) = 0.12$ per year
 and $\sigma = 0.24$ per annum

$$S(0) = \text{Rs } 40$$

what is the prob. that an european call option
 having 4 yrs to exercise with strike price
 Rs 42 will be exercised.

Ans.

$$\mu = 0.12$$

$$\sigma = 0.24$$

$$S(0) = \text{Rs } 40$$

$$T = 4 \text{ yrs}$$

$$K = \text{Rs } 42$$

$$P(S(t) > 42) = P(S(4) > 42)$$

$$\begin{aligned}
 & P\left(\frac{S(4)}{S(0)} > \frac{42}{50}\right) \\
 &= P\left(\ln\left(\frac{S(4)}{S(0)}\right) > \ln\left(\frac{42}{50}\right)\right) \\
 &= P\left(\frac{\ln(S(4)/S(0)) - 0.48}{0.48} > \frac{\ln(42/50) - 0.48}{0.48}\right) \underbrace{(X - \mu)/\sigma}_{\sim N(0, 1)} \\
 &= P(Z > -0.8983)
 \end{aligned}$$

- Q A stock price is currently \$50.
 $\mu = 18\%$ per annum (expected return)
 $\sigma = 30\%$. What is prob. distribution for
 Stock price in 2 years ($T = 2$ yrs).
 Calculate the mean and standard deviation of
 the distribution and determine 95%
 Confidence Interval.

if not mentioned then SDE

Solⁿ

As nothing is mentioned it is SDE

SDE follows log-normal distribution

so $\ln(S(t))$ follows normal dist.

$$\ln(S(t)) \sim N\left(\mu - \frac{\sigma^2}{2}T + \ln S(0), \sigma \sqrt{T}\right)$$

$$\sim N\left(\ln S(0) + \left(0.18 - \frac{0.09}{2}\right)T, 0.3 \times \sqrt{2}\right)$$

$$\sim N(4.18, 0.42)$$

$$\begin{aligned} E(S(t)) &= S(0) e^{\mu t} \\ &= S(0) e^{0.18 \times 2} \\ &= 71.67 \end{aligned}$$

~~$$S.D = \sqrt{\text{Var}(S(t))} = S(0) e^{\mu t} (e^{\sigma^2 T} - 1)^{1/2}$$~~

~~$$S.D = 31.83$$~~

$$x \in (9.28, 134.06)$$

$$\begin{cases} \frac{x-\mu}{\sigma} = Z_\alpha \\ x \in (\mu \pm Z_\alpha \sigma) \\ \underbrace{(\mu \pm 1.96 \sigma)}_{\hookrightarrow 95\%} \end{cases}$$

for 95%.

$$\ln(S(2)) \in \left(4.18 - 1.96 \times 0.4^2, 4.18 + 1.96 \times 0.4 \right)$$

$$\ln(S(2)) \in (e^{3.36}, 5)$$

$$S(2) \in (e^{3.36}, e^5)$$

$$S(2) \in (28.79, 148.41)$$

this will become $S(0)$
if we calculate after
1 year from now.

O A S.P $S(t)$

{ Solved
in book }

$$dS(t) = aS(t)dt + bS(t)dW(t)$$

when a and b are constant

$$(i) \sqrt{S(t)}$$

$$(ii) \ln(S(t))$$

Ans. $\int I(t, x) = \sqrt{S(t)}$

$$\sqrt{S(t)} = \sqrt{x}$$

$$x = S(t) \quad \textcircled{2}$$

$$\begin{aligned} \int t &= 0 \\ \int x &= \frac{1}{2\sqrt{x}} \end{aligned} \quad \left. \begin{aligned} \int_{xx} &= \frac{-1}{4x^{3/2}} \end{aligned} \right] - \textcircled{3}$$

$$df(t, x) = f_t dt + f_x dS(t) + \frac{1}{2} f_{xx} dS(t) dS(t)$$

—④

Substitute in ④ from ② and ③,

$$\begin{aligned} df(t, x) &= \sigma + \frac{1}{2\sqrt{S(t)}} (a S(t) dt + b S(t) dw(t)) \\ &\quad + \sigma \left(-\frac{1}{4(S(t))^{3/2}} \right) b^2 S^2(t) dt \\ &= dt \left[\frac{a}{2} \frac{S(t)}{\sqrt{S(t)}} - \frac{b^2}{4} \frac{S^2(t)}{S(t)^{3/2}} \right] \\ &\quad + \frac{b}{2} \sqrt{S(t)} dw(t) \end{aligned}$$

$$d(\sqrt{S(t)}) = \frac{\sqrt{S(t)}}{2} \left(a - \frac{b^2}{2} \right) dt + \underbrace{\frac{b}{2}}_{\text{missing}}$$

$$\# d(S(t)) = \mu S(t) dt + \sigma S(t) dw(t)$$

$t \in [0, T]$

$\beta(t)$ be a risk free asset

$$d\beta(t) = \beta(t) \circ r dt$$

$$\beta(t) = \beta(0) e^{rt}$$

when r is the risk-free rate

At time t we take a portfolio which contains $a(t)$ shares and $b(t)$ risk free assets.

$$V_p(t) = a(t) \cdot S(t) + b(t) \cdot B(t) \quad \text{--- (3)}$$

$$dV_p(t) = a(t) dS(t) + b(t) dB(t) \quad \text{--- (4)}$$

$$t \in [0, T]$$

discounted stock price of one ~~share~~ share of stock is

$$S(t) = e^{-rt} S(t)$$

Applying ITO doblet formula of version 2

$$d(S(t)) = -r e^{-rt} S(t) dt + e^{-rt} d(S(t))$$

$$= -r e^{-rt} S(t) dt + e^{-rt} \left[\mu S(t) dt + \right.$$

$$\left. a S(t) dw(t) \right]$$

$$= e^{-rt} S(t) (\mu - r) dt$$

$$+ e^{-rt} a S(t) dw(t)$$

$$= \tilde{S}(t) \left[(\mu - r) dt + a dw(t) \right]$$

$$= \alpha \tilde{S}(t) d\tilde{w}(t) ; d\tilde{w}(t) = \left(\frac{\mu - r}{\sigma}\right) dt + dw(t)$$

$$\Rightarrow \tilde{w}(t) = \frac{\mu - r}{\sigma} t + w(t)$$

$(\mu - r)$ is called risk premium

$\frac{\mu - r}{\sigma}$ is risk premium per unit of risk
and is called Market price of risk.