



Subject: Financial Engineering 1

Chapter:

**Category: Assignment 1
Solutions**

Answer 1 –

- i) Given: Derivative price = 0.1448,
payoff = 1 at time 2 if $S_2 < S_0$

$$A = \begin{cases} 0.761, S_t = uS_{t-1} \\ 1.522, S_t = dS_{t-1} \end{cases}$$

Calculating value of p:

derivative pays after 2 downward moves

$$0.1448 = A_1^2 \cdot (1-p)^2$$

$$0.1448 / (1.522^2) = (1-p)^2$$

$$p = 0.749983$$

- ii) **Calculating value of q:**

$A_1 = 0.716$ for an upward move

$$A_1 = \exp(-r) \cdot (q/p)$$

$$0.716 = \exp(-0.05) \cdot q / 0.749983$$

$$q = 0.56452$$

- iii) **Derivative price at time 0, if payoff = 1 for $S_2 > S_0$, using q**

If candidates are assuming $ud = 1$, which is generally accepted assumption,

$$\text{price} = 0.6^2 \cdot e^{(-0.05 \times 2)} \cdot 1$$

$$\text{price} = 0.3257$$

[3]

Or

If the candidates assume specifically for recombining binomial tree and use the condition $ud = 1$. This gives payoffs for upper and middle nodes.

$$\text{price} = 0.6^2 \cdot \exp(-0.05 \times 2) + 2 \times 0.6 \times 0.4 \exp(-0.05 \times 2)$$

$$\text{price} = 0.7601$$

[3]

Answer 2 –

i) Let $Z \sim N(0, 1)$. Then, since $W(t) \sim N(0, t)$ we get [0.5]
 $E[|W(t)|] = E[|\sqrt{t}Z(t)|]$ [1]
 $= \sqrt{t}E[|Z(t)|]$ [1]
 $= \sqrt{t} \cdot 2 \int_0^{\infty} \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}} z \, dz$ (from standard normal integration, since the absolute is used, integration is from 0 to infinity multiplied by 2 times) [2]
 $= \sqrt{t} \cdot 2 \int_0^{\infty} \frac{e^{-w}}{\sqrt{2\pi}} dw$ (by $w = z^2/2$, $dw = z \, dz$) [2]
 $= \frac{\sqrt{t} \cdot 2}{\sqrt{2\pi}} * [e^{-w}]_{\infty}^0$ [1]
 $= \frac{\sqrt{2t}}{\sqrt{\pi}}$ [0.5]

ii) ABC Ltd's price is higher than XYZ Ltd's price at the end of the day would be equivalent to:

$$X(t) > 0$$

If $X(t)$ is an Arithmetic Brownian motion process

Then

$$E[X_T] = x + \mu T$$

Where,

$$x = X(3/4) = 40.25 - 39.75 = 0.50$$

$$\mu = 0$$

$$T = (1 - 3/4)$$

$$\text{Var}[X_T] = \sigma^2 T = 0.3695/4 = 0.092375$$

$$P(X(1) \geq 0) = P\{(X(1) - E[X_T]) / \sqrt{\text{Var}[X_T]} \geq (0 - 0.5) / \sqrt{0.092375}\}$$

$$= 1 - P(Z \leq -1.6451) = 0.95$$

Answer 3-

i) a) Risk neutral probability measure –

Q is the risk-neutral probability measure, which gives the risk-neutral probability q to an upward move in prices and $1-q$ to a downward move.

The risk-neutral probabilities ensure that the underlying security yields an expected return equal to the risk-free rate

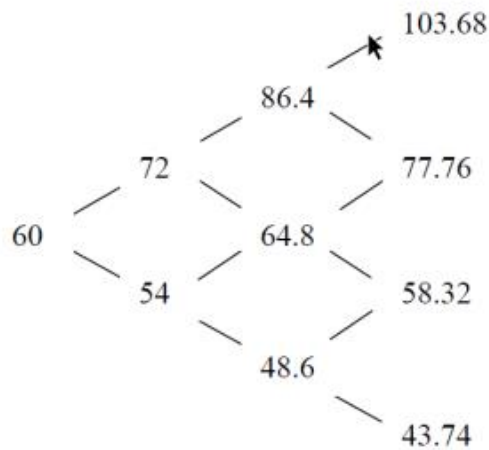
Under the probability measure Q , investors are neutral with regard to risk, they require no additional return for taking on more risk.

b) Recombining binary tree A recombining binomial tree (or binomial lattice) is one in which values of u and d , and consequently the risk-neutral probabilities, are the same in all states.

With such models:

- the volume of computation required is greatly reduced
- N_t , the number of up-steps up to time t , has a binomial distribution with parameters t and q

ii)



$$C_{uuu} = (u^3 S_0 - K)^+;$$

$$C_{uud} = (u^2 d S_0 - K)^+;$$

$$C_{udd} = (u d^2 S_0 - K)^+;$$

$$C_{ddd} = (d^3 S_0 - K)^+.$$

and in the lowest state,

$$C_2(dd) = \frac{1}{1+r} [q C_{udd} + (1-q) C_{ddd}],$$

where the risk-neutral probability of an upward move is

$$q = \frac{(1+r) - d}{u - d}.$$

At time 1, we get in the upper state,

$$C_1(u) = \frac{1}{1+r} [q C_2(uu) + (1-q) C_2(ud)],$$

and in the lower state,

$$C_1(d) = \frac{1}{1+r} [q C_2(ud) + (1-q) C_2(dd)].$$

At time 0,

$$C_0 = \frac{1}{1+r} [q C_1(u) + (1-q) C_1(d)].$$

Answer 4

i)

$$0.7610 = e^{-r}q/p \quad \text{if } S_1 = S_0u$$

$$1.5220 = e^{-r}(1-q)/(1-p) \quad \text{if } S_1 = S_0d$$

Solving the 2 equations with 2 unknown gives

$$\begin{aligned} e^{-r} \cdot 0.7610 \cdot p &= (1 - 1.5220(1-p)) \cdot e^{-r} \\ &= 1 - 1.5220 \cdot e^{-r} + 1.5220 \cdot p \cdot e^{-r} \end{aligned}$$

$$1.5220 \cdot e^{-r} - 1 = p \cdot e^{-r} \cdot (1.5220 - 0.7610)$$

This gives $p = 0.7748$ and $q = 0.63236$.

ii)

$$(e^{-r} - d)/(u - d) = q$$

Using the definition for a recombining model ; $u \times d = 1$

Rearranging the terms gives;

$$(e^{-r} - d)/(1/d - d) = q$$

$$(e^{-r} - d)d/(1 - d^2) = q$$

$$(e^{-r}d - d^2) = q - d^2q$$

$$d^2 \times (1 - q) - d \times e^{-r} + q = 0$$

Solving the quadratic equation; Roots are:

$$= -b \pm \sqrt{b^2 - 4ac} / 4a$$

$$= e^{-r} \pm \sqrt{e^{-2r} - 4 \cdot (1-q) \cdot q} / 4 \cdot (1-q)$$

Substituting values of r and q gives

$$= (1.072508 + 0.469412)/0.74 \quad \text{or} \quad (1.072508 - 0.469412)/0.74$$

$$d = 2.0970 \quad \text{or} \quad 0.8202$$

since ' d ' has to be less than 1

$$d = 0.8202$$

$$u = 1.2192$$

$$\text{iii) In this case price} = 0.63^2 \times \exp(-0.07 \times 2) + 2 \times 0.63 \times 0.37 \times \exp(-0.07 \times 2)$$

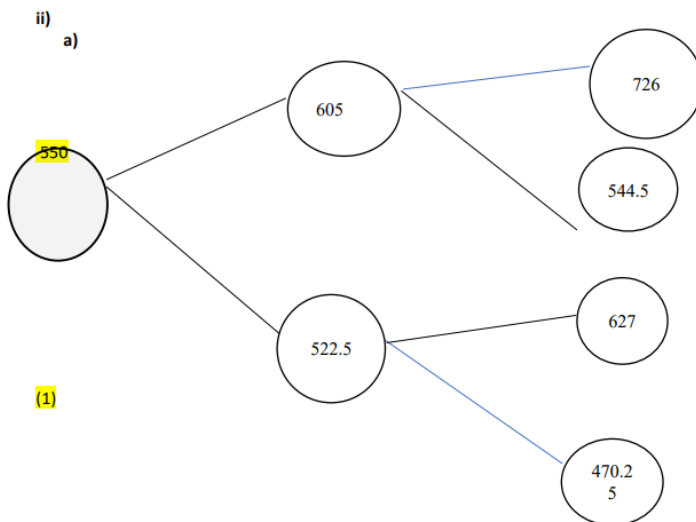
$$= 0.3476 + 0.4042$$

$$= 0.7519$$

Answer 5

i) The recombining binomial tree is assumed to have the size of up-steps and down-steps to be same under all states and across all time intervals. It means $u_t(j)=u$ and $d_t(t)=d$ for all times t and states j , also $d < e^r < u$. Hence the risk neutral probability 'q' is also constant for all times and in all states.

The main advantage of recombining binomial tree is that it greatly reduces the computational time by reducing the number of states.



The risk-neutral probability for each state

$$q_1 = (e^{0.02} - 0.95) / (1.10 - 0.95) = 0.4680 \text{ (time 1)}$$

$$q_2 = (e^{0.03} - 0.90) / (1.20 - 0.90) = 0.4349 \text{ (time 2)}$$

Possible Payoffs after six months are

$$p_{uu} = \text{Max}(500 - 726, 0) = 0$$

$$p_{ud} = \text{Max}(500 - 544.5, 0) = 0$$

$$p_{du} = \text{Max}(500 - 627, 0) = 0$$

$$p_{dd} = \text{Max}(500 - 470.25) = 29.75$$

The value of the option at first 3 months if upside

$$V_1(1) = e^{-0.03} (0.4348 * 0 + 0.5652 * 0) = 0$$

For downside

$$V_1(2) = e^{-0.03} (0.4348 * 0 + 0.5652 * 29.75) = 16.3178$$

At time 0 the value of the put option will be

$$V_0 = e^{-0.02} (0.4680 * 0 + 0.5320 * 16.3178) \\ = 8.5092$$

(1)
[Max 6]

b) For the first 3 months, the payoff if upside

$$\text{Max}(500 - 3/6 * 605, 0) = 197.5$$

And if downside

$$\text{Max}(500 - 3/6 * 522.5, 0) = 238.75$$

(1)

The value of the option at first 3 months [since the final payoffs are same as ii) a)]

$$V_1(1) = 0 \text{ and } V_1(2) = 16.3178$$

It is optimal to exercise the American put option early in this case as the payoffs are high compared to the value of put option at first 3 months (1)

Hence, the value of the put option assuming it is exercised at the first 3 months

$$V_0 = e^{-0.02} (0.4680 * (197.5 - 0) + 0.5320 * (238.75 - 16.3178)) \\ = 206.5905 \quad (2)$$

At expiry the American put option value is same as the European put option derived in ii) a) which is = 8.5092

Hence, the total value of American put is $206.5905 + 8.5092 = 215.0997$

Answer 6

Let $dX_t = A_t dt + B_t dZ_t$,

Where, $A_t = \alpha \mu (T-t)$, $B_t = \sigma \sqrt{(T-t)}$ Eq 1

$$dF = \frac{\partial f}{\partial x} B_t dZ_t + \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial x} A_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} B_t^2 \right) dt \quad (\text{Ito's lemma})$$

$$= -f B_t dZ_t + f \frac{\partial m}{\partial t} (T-t) dt - f A_t dt + \frac{1}{2} f B_t^2 dt$$

[Since $\frac{\partial f}{\partial x} = -e^{(m(T-t)-x)}$ and $\frac{\partial f}{\partial t} = \frac{\partial m}{\partial t} (T-t) * e^{(m(T-t)-x)}$
(using chain rule) and $\frac{\partial^2 f}{\partial x^2} = e^{(m(T-t)-x)}$]

$$dF = f \left(\frac{\partial m}{\partial t} (T-t) - A_t + \frac{1}{2} B_t^2 \right) dt - f B_t dZ_t$$

For f to be a martingale, $\frac{\partial m}{\partial t} (T-t) - A_t + \frac{1}{2} B_t^2 = 0$

$$\text{Thus, } \frac{\partial m}{\partial t} (T-t) = A_t - \frac{1}{2} B_t^2$$

Substituting Eq 1 above gives $\frac{\partial m}{\partial t} = \alpha \mu - \frac{1}{2} \sigma^2$

Answer7

i) Given

$$F(t, x) = e^{-t} x^2$$

$$\frac{df}{dt} = -e^{-t} x^2 = -f$$

$$\frac{df}{dx} = 2e^{-t} x$$

$$\frac{df}{dx^2} = 2e^{-t}$$

$$Y_t = f(t, X_t)$$

Applying Ito's lemma

$$\begin{aligned} dY_t &= \frac{df}{dt} dt + \frac{df}{dx} dX_t + \frac{1}{2} \frac{d^2 f}{dx^2} \sigma^2 X_t^2 dt \\ &= -f dt + 2e^{-t} X_t dX_t + e^{-t} \sigma^2 X_t^2 dt \\ &= -Y_t dt + 2e^{-t} X_t^2 \frac{dX_t}{X_t} + e^{-t} \sigma^2 X_t^2 dt \\ &= -Y_t dt + 2Y_t [0.25 dt + \sigma dW_t] + \sigma^2 Y_t dt \end{aligned}$$

$$\begin{aligned} \frac{dY_t}{Y_t} &= [2 * (0.25) - 1 + \sigma^2] dt + 2\sigma dW_t \\ &= [\sigma^2 - 0.5] dt + 2\sigma dW_t \end{aligned}$$

Therefore

$$dY_t = [\sigma^2 - 0.5] Y_t dt + 2\sigma Y_t dW_t$$

ii) The process is martingale if drift is zero. This means $\sigma^2 - 0.5 = 0$
i.e. $\sigma^2 = 0.5$

Answer 8

The required probability is the probability of the stock price being greater than Rs. 258 in 6 months' time.

The stock price follows Geometric Brownian motion

$$\text{i.e. } S_t = S_0 \exp\left(\left(\mu - \frac{\sigma^2}{2}\right)t + \sigma W_t\right)$$

Therefore $\ln(S_t)$ follows normal distribution with mean $\ln(S_0) + (\mu - \sigma^2/2)t$ and variance $(\sigma^2)t$

$$\begin{aligned} \text{Implies } \ln(S_t) \text{ follows } &\varphi\left(\ln 254 + (0.16 - 0.35^2/2) \cdot 0.5, 0.35 \cdot 0.5^{1/2}\right) \\ &= \varphi(5.59, 0.247) \end{aligned}$$

This means $[\ln(S_t) - \ln(S_0)] / (\sigma \sqrt{t})$ follows standard normal distribution.

Hence the probability that stock price will be higher than the strike price of Rs. 258 in 6 months' time =

$$1 - N(5.55 - 5.59) / 0.247 = 1 - N(-0.1364) = 0.5542.$$

The put option is exercised if the stock price is less than Rs. 258 in 6 months' time.

$$\text{The probability of this} = 1 - 0.5542 = 0.4457$$

Answer 9

i) The given relationship can be written as:

$$S_t = S_0 e^{\mu t + \sigma B_t}$$

Since S_t is a function of standard Brownian motion, B_t , applying Ito's Lemma, the SDE for the underlying stochastic process becomes:

$$dB_t = 0 \times dt + 1 \times dB_t$$

Let $G(t, B_t) = S_t = S_0 e^{\mu t + \sigma B_t}$, then

$$dG/dt = \mu S_0 e^{\mu t + \sigma B_t} = \mu S_t$$

$$dG/dB_t = \sigma S_0 e^{\mu t + \sigma B_t} = \sigma S_t$$

$$d^2G/dB_t^2 = \sigma^2 S_0 e^{\mu t + \sigma B_t} = \sigma^2 S_t$$

Hence, using Ito's Lemma from Page 46 in the Tables we have:

$$dG = [0 \times \sigma S_t + \frac{1}{2} \times 1^2 \times \sigma^2 S_t + \mu S_t] dt + 1 \times \sigma S_t dB_t$$

$$\text{i.e. } dS_t = (\mu + \frac{1}{2} \sigma^2) S_t dt + \sigma S_t dB_t$$

Thus,

$$dS_t/S_t = \sigma dB_t + (\mu + \frac{1}{2} \sigma^2) dt$$

$$\text{So, } c_1 = \sigma \text{ and } c_2 = \mu + \frac{1}{2} \sigma^2$$

ii)

The expected value of S_t is:

$$E[S_t] = E[S_0 e^{\mu t + \sigma B_t}] = S_0 e^{\mu t} E[e^{\sigma B_t}]$$

Since $B_t \sim N(0,1)$, its MGF is $E[e^{\theta B_t}] = e^{\frac{1}{2} \theta^2 t}$

$$\text{So, } E[S_t] = S_0 e^{\mu t} \times e^{\frac{1}{2} \sigma^2 t} = S_0 e^{\mu t + \frac{1}{2} \sigma^2 t}$$

The variance of S_t is:

$$\begin{aligned} \text{Var}[S_t] &= E[S_t^2] - (E[S_t])^2 \\ &= E[S_0^2 e^{2\mu t + 2\sigma B_t}] - (S_0 e^{\mu t + \frac{1}{2} \sigma^2 t})^2 \\ &= S_0^2 e^{2\mu t} E[e^{2\sigma B_t}] - S_0^2 e^{2\mu t + \sigma^2 t} \\ &= S_0^2 e^{2\mu t + 2\sigma^2 t} - S_0^2 e^{2\mu t + \sigma^2 t} \\ &= S_0^2 e^{2\mu t} (e^{2\sigma^2 t} - e^{\sigma^2 t}) \end{aligned}$$

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iii)

$$\text{Cov}[S_{t1}, S_{t2}] = E[S_{t1}, S_{t2}] - E[S_{t1}] E[S_{t2}]$$

From above,

$$E[S_{t1}] = S_0 e^{\mu t1 + \frac{1}{2} \sigma^2 t1} \text{ and } E[S_{t2}] = S_0 e^{\mu t2 + \frac{1}{2} \sigma^2 t2}$$

The expected value of the product is:

$$\begin{aligned} E[S_{t1}, S_{t2}] &= E[S_0 \exp(\mu t1 + \sigma B_{t1}) S_0 \exp(\mu t2 + \sigma B_{t2})] \\ &= S_0^2 e^{\mu(t1 + t2)} E[\exp(\sigma B_{t1} + \sigma B_{t2})] \end{aligned}$$

To evaluate this we need to split B_{t2} into two independent components:

$$B_{t2} = B_{t1} + (B_{t2} - B_{t1}) \text{ where } B_{t2} - B_{t1} \sim N(0, t2 - t1)$$

Hence,

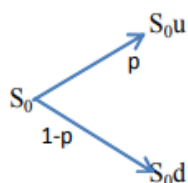
$$\begin{aligned} E[S_{t1}, S_{t2}] &= S_0^2 e^{\mu(t1 + t2)} E[\exp(\sigma B_{t1} + \sigma \{B_{t1} + (B_{t2} - B_{t1})\})] \\ &= S_0^2 e^{\mu(t1 + t2)} E[\exp(2\sigma B_{t1} + \sigma \{B_{t2} - B_{t1}\})] \\ &= S_0^2 e^{\mu(t1 + t2)} E[\exp(2\sigma B_{t1})] E[\exp\{\sigma \{B_{t2} - B_{t1}\}\}] \\ &= S_0^2 e^{\mu(t1 + t2)} \exp(2\sigma^2 t1) \exp\left[\frac{1}{2} \sigma^2 (t2 - t1)\right] \\ &= S_0^2 e^{\mu(t1 + t2)} \exp\left(\frac{3}{2} \sigma^2 t1 + \frac{1}{2} \sigma^2 t2\right) \end{aligned}$$

Putting all the equations together:

$$\begin{aligned} \text{Cov}[S_{t1}, S_{t2}] &= S_0^2 e^{\mu(t1 + t2)} \exp\left(\frac{3}{2} \sigma^2 t1 + \frac{1}{2} \sigma^2 t2\right) - S_0 e^{\mu t1 + \frac{1}{2} \sigma^2 t1} \cdot S_0 e^{\mu t2 + \frac{1}{2} \sigma^2 t2} \\ &= S_0^2 e^{\mu(t1 + t2)} \left(\exp\left(\frac{3}{2} \sigma^2 t1\right) - \exp\left(\frac{1}{2} \sigma^2 t1\right)\right) \exp\left(\frac{1}{2} \sigma^2 t2\right) \end{aligned}$$

Answer 10

i) Setting up the commodity tree using u for up move and d for down move, p is up-step probability:



Where p is the up probability and (1-p) the down probability

Then $E(C_t) = S_0[pu + (1-p)d]$, and

$$\begin{aligned}\text{Var}(C_t) &= E(C_t^2) - E(C_t)^2 \\ &= S_0^2 [pu^2 + (1-p)d^2] - S_0^2 [pu + (1-p)d]^2 \\ &= S_0^2 [pu^2 + (1-p)d^2 - (pu + (1-p)d)^2] \\ &= S_0^2 [p(1-p)u^2 + p(1-p)d^2 - 2p(1-p)] \quad (\because d = 1/u) \\ &= S_0^2 p(1-p)(u-d)^2\end{aligned}$$

Equating moments:

$$S_0 e^{rt} = S_0 [pu + (1-p)d] \quad \text{_____ (A)}$$

$$\text{And } \sigma^2 S_0^2 t = S_0^2 p(1-p)(u-d)^2 \quad \text{_____ (B)}$$

From (A) we get

$$p = \frac{e^{rt} - d}{u - d} \quad \text{_____ (C)}$$

Substituting p into equation (B), we get

$$\begin{aligned}\sigma^2 t &= \frac{e^{rt} - d}{u - d} \left(1 - \frac{e^{rt} - d}{u - d}\right) (u - d)^2 \\ &= - (e^{rt} - d)(e^{rt} - u) = (u + d) e^{rt} - (1 + e^{2rt})\end{aligned}$$

Putting $d = 1/u$, and multiplying through by u we get

$$u^2 e^{rt} - u(1 + e^{2rt} + \sigma^2 t) + e^{rt} = 0$$

This is a quadratic in u which can be solved in the usual way.

ii)

a) $\sigma = 0.15$, $t = 0.25 \Rightarrow u = \exp(.15 \cdot \sqrt{.25}) = \exp(.075) = 1.077884$, $d = 1/u = .92774$

The tree is

t=0	t=.25	t=.5	t=.75	
			100.186	Node A
		92.947		
	86.231		86.232	Node B
80		80.001		
	74.22		74.22	Node C
		68.857		
			63.882	Node D

b) $r=0$, we have $p = \frac{e^{rt} - d}{u - d} = \frac{(1 - .927744)}{(1.077884 - .927744)} = .48126$

Discounting back the final payoff at $t=.75$ to $t=0$ along the tree using p and $(1-p)$, we get

$t=0$	$t=.25$	$t=.5$	$t=.75$	
			20.186	Node A
		12.948		
	7.787		6.232	Node B
4.496		2.999		
	1.443		0	Node C
		0		
			0	Node D

Hence value of the call option is 4.496.

c) The lookback call pays the difference between the minimum value and the final value.
Notate paths by U for up and D for down, in order

We get the payoffs

UUU	$(100.186 - 80) = 20.186$	Node A
UDU	$(86.232 - 80) = 6.232$	Node B
UUD	$(86.232 - 80) = 6.232$	Node B
UDD	$(74.22 - 74.22) = 0$	Node C
DUU	$(86.232 - 74.22) = 12.012$	Node B
DUD	$(74.22 - 74.22) = 0$	Node C
DDU	$(74.22 - 68.857) = 5.363$	Node C
DDD	$(63.882 - 63.882) = 0$	Node D

The lookback payoffs are, for each successful path (i.e. with a non- zero result)

Probabilities of arriving at each node are:

Node A= $p^3 = .11147$

Node B= $p^2(1-p) = .12015$

Node C= $p(1-p)^2 = .12950$

Node D= $p(1-p)^3 = .13959$

Hence the tree value of lookback option is:

$$(.11147 * 20.186) + (.12015 * [6.232 + 6.232 + 12.012]) + (.12950 * 5.363)$$

$$= 5.8854$$

Answer 11

Given $Z(t)$ is standard Brownian

$$\begin{aligned} \text{a. } dU(t) &= 2dZ(t) - 0 \\ &= 0dt + 2dZ(t). \end{aligned}$$

Thus, the stochastic process $\{U(t)\}$ has zero drift.

$$\begin{aligned} \text{b. } dV(t) &= d[Z(t)]^2 - dt. \\ d[Z(t)]^2 &= 2Z(t)dZ(t) + 2/2 [dZ(t)]^2 \\ &= 2Z(t)dZ(t) + dt \text{ by the multiplication rule} \end{aligned}$$

Thus, $dV(t) = 2Z(t)dZ(t)$. The stochastic process $\{V(t)\}$ has zero drift.

$$\begin{aligned} \text{c. } dW(t) &= d[t^2 Z(t)] - 2t Z(t)dt \\ \text{Because } d[t^2 Z(t)] &= t^2 dZ(t) + 2tZ(t)dt, \text{ we have} \\ dW(t) &= t^2 dZ(t). \end{aligned}$$

Thus The process $\{W(t)\}$ has zero drift

Answer 12

$$\text{i. Let } f = f(S_t, t) = S^k$$

$$\frac{\partial f}{\partial S} = kS^{k-1}, \quad \frac{\partial^2 f}{\partial S^2} = k(k-1)S^{k-2}, \quad \frac{\partial f}{\partial t} = 0$$

$$\begin{aligned} \text{Using Ito calculus, } df &= kS^{k-1}dS_t + \frac{1}{2} k(k-1)S^{k-2}(dS_t)^2 \\ &= f[k\mu + \frac{1}{2}k(k-1)\sigma^2]dt + f[k\sigma]dW_t \end{aligned}$$

Hence $f = S^k$ follows Geometric Brownian motion, with drift $\mu' = k\mu + \frac{1}{2}k(k-1)\sigma^2$ and volatility $\sigma' = k\sigma$.

$$\text{Hence } f_t = f_0 e^{(\mu' - \frac{1}{2}\sigma'^2)t + \sigma'W_t}$$

$$\text{This means } \frac{f}{f_0} \sim \text{lognormal}((\mu' - \frac{1}{2}\sigma'^2)t, \sigma'^2 t)$$

The mean and variance of a log normal distribution with parameters μ and σ^2

$$\text{Is given by } e^{\mu + \frac{1}{2}\sigma^2} \text{ and } (e^{\sigma^2} - 1)e^{2\mu + \sigma^2}$$

$$\text{This means } E(f) = f_0 e^{\mu' t} \text{ and } V(f) = f_0^2 e^{2\mu' t} (e^{\sigma'^2 t} - 1)$$

ii. Let $f = f(t, S_t) = e^{-rt} S_t$

$$\frac{\partial f}{\partial t} = -r e^{-rt} S_t, \quad \frac{\partial f}{\partial S} = e^{-rt}, \quad \frac{\partial^2 f}{\partial S^2} = 0$$

$$\begin{aligned} \text{By Ito calculus, } df &= -r e^{-rt} S_t dt + e^{-rt} dS_t \\ &= (\mu - r) f dt + \sigma f dW_t \quad (\text{on substituting the expression for } dS_t) \end{aligned}$$

which is a martingale if and only if $\mu = r$

iii. Combining results from part (a) and (b), we need, for discounted S^k to be a martingale,
 $k\mu + \frac{1}{2}k(k-1)\sigma^2 = r$

For given values of r , σ and k , one can solve for the value of μ for which discounted S^k will be a martingale.