

There are 12 problems and 10 points each.

1. (Property of Brownian Bridge)

Let $B_t^0 = \{B_t, 0 \leq t \leq 1 | B_1 = 0\}$ be a Brownian bridge, and define $dX_t = -\frac{X_t}{1-t}dt + dB_t$, with $X_0 = 0$.

a) Show that $\{B_t^0, 0 \leq t \leq 1\}$ and $\{X_t, 0 \leq t \leq 1\}$ have the same distribution.

b) Show that $Y_t = (1+t)B_{\frac{t}{1+t}}^0$ is a standard Brownian motion.

sol:

(a)

(1) $B_t^0 = B_t - tB_1, \because B_t \sim N(0, t), B_1 \sim N(0, 1), tB_1 \sim N(0, t^2) \therefore$ linear combination of normal distribution is still normal. Compute mean and variance.

$$EB_t^0 = 0,$$

$$EB_s^0 B_t^0 = s(1-t) \text{ if } s < t \Rightarrow \text{Var} B_t^0 = t(1-t)$$

$$B_t^0 \sim N(0, t(1-t)) \quad \forall 0 \leq t \leq 1.$$

(2) According to the solution of OU process, $X_t = (1-t) \int_0^t \frac{1}{1-s} dB_s$

X_t is the linear combination of N.D., therefore it is still a N.D. Compute mean and variance

$$X_t = 0,$$

$$\text{If } s \leq t : EX_s X_t$$

$$= (1-s)(1-t) E[\int_0^s \frac{1}{1-u} dB_u \int_0^t \frac{1}{1-v} dB_v]$$

$$= (1-s)(1-t) E[\int_0^s \frac{1}{1-u} dB_u (\int_0^s \frac{1}{1-v} dB_v + \int_s^t \frac{1}{1-v} dB_v)]$$

$$= (1-s)(1-t) E[(\int_0^s \frac{1}{1-u} dB_u)^2] \text{ (independent increment)}$$

$$= (1-s)(1-t) E[\int_0^s (\frac{1}{1-u})^2 ds] \text{ (Ito's isometry)}$$

$$= (1-s)(1-t) \frac{s}{1-s} = s(1-t) \Rightarrow \text{Var} X_t = t(1-t)$$

$$X_t \sim N(0, t(1-t))$$

(b)

$\because B_{\frac{t}{1+t}} \sim N(0, \frac{t}{1+t}), B_{\frac{s}{1+s}} \sim N(0, \frac{s}{1+s}), B_1 \sim N(0, 1) \therefore Y_t \sim \text{N.D.}$

$$EX_t = E(1+t)(B_{\frac{t}{1+t}} - \frac{t}{1+t} B_1) = 0$$

If $0 \leq s < t \leq 1$ ($0 \leq \frac{s}{1+s} < \frac{t}{1+t} \leq 1$):

$$EX_s X_t = (1+s)(1+t)[EB_{\frac{s}{1+s}} B_{\frac{t}{1+t}} - \frac{t}{1+t} EB_{\frac{s}{1+s}} B_1 - \frac{s}{1+s} EB_{\frac{t}{1+t}} B_1 + \frac{s}{1+s} \frac{t}{1+t} EB_1^2] = s \Rightarrow$$

$$\text{Var} Y_t = t$$

$$Y_t \sim N(0, t)$$

$$(1) Y_0 = 0$$

(2) Independent increment

$$\forall 0 = t_0 < t_1 < \dots < t_n = 1 : Y_{t_1} - Y_{t_0}, Y_{t_2} - Y_{t_1}, \dots, Y_{t_n} - Y_{t_{n-1}}$$

$$\text{Cov}(Y_{t_i} - Y_{t_{i-1}}, Y_{t_j} - Y_{t_{j-1}}) = 0$$

(3) stationary increment

$$E(Y_t - Y_s) = 0$$

$$\text{Var}(Y_t - Y_s) = \text{Var}(Y_t) + \text{Var}(Y_s) - 2\text{Cov}(Y_t, Y_s) = t + s - 2s = t - s$$

$$Y_t - Y_s \sim N(0, t - s)$$

$$(4) Y_t \sim N(0, t)$$

(5) B_t is continuous in $t \Rightarrow Y_t$ is also continuous in t

From (1)~(5), Y_t is a SBM.

2. a) Solve the following stochastic differential equation

$$dX_t = (-\alpha X_t + \beta)dt + \sigma dB_t, \quad \text{where } X_0 = x_0 \text{ and } \alpha > 0.$$

b) Verify the solution can be written as

$$X_t = e^{-\alpha t} \left(x_0 + \frac{\beta}{\alpha} (e^{\alpha t} - 1) + \sigma \int_0^t e^{\alpha s} dB_s \right).$$

c) Show that X_t converges in distribution, as $t \rightarrow \infty$, and find the limiting distribution.

d) Find the covariance $\text{Cov}(X_s, X_t)$ for $s < t$.

sol:

(a)

$$dX_t + \alpha X_t dt = \beta dt + \sigma dB_t$$

multiply $e^{\alpha t}$ for both sides

$$d(e^{\alpha t} X_t) = \beta e^{\alpha t} dt + \sigma e^{\alpha t} dB_t$$

$$\Rightarrow e^{\alpha s} X_s \Big|_0^t = \beta \int_0^t e^{\alpha s} ds + \sigma \int_0^t e^{\alpha s} dB_s$$

$$\Rightarrow X_t = e^{-\alpha t} \left(x_0 + \frac{\beta}{\alpha} (e^{\alpha t} - 1) + \sigma \int_0^t e^{\alpha s} dB_s \right)$$

(b)

$$\frac{\partial X_t}{\partial t} = -\alpha X_t + \beta$$

$$\frac{\partial X_t}{\partial B_t} = \sigma$$

$$\frac{\partial^2 X_t}{\partial B_t^2} = 0$$

By Ito's formula:

$$dX_t = (-\alpha X_t + \beta)dt + \sigma dB_t$$

(c)

$$(1) EX_t = x_0 e^{-\alpha t} + \frac{\beta}{\alpha} (1 - e^{-\alpha t}) \rightarrow \frac{\beta}{\alpha} \text{ as } t \rightarrow \infty$$

$$(2) \text{Var} X_t = \sigma^2 E \int_0^t e^{-2\alpha(t-s)} ds = \frac{\sigma^2}{2\alpha} (1 - e^{-2\alpha t}) \rightarrow \frac{\sigma^2}{2\alpha} \text{ as } t \rightarrow \infty$$

Therefore $X_t \xrightarrow{\mathcal{D}} N\left(\frac{\beta}{\alpha}, \frac{\sigma^2}{2\alpha}\right)$ as $t \rightarrow \infty$

$$(d) \text{Cov}(X_s, X_t) = \sigma^2 e^{-\alpha(t+s)} \text{Cov}\left(\int_0^s e^{\alpha u} dB_u, \int_0^t e^{\alpha u} dB_u\right)$$

$$= \sigma^2 e^{-\alpha(t+s)} E \int_0^{s \wedge t} e^{2\alpha u} du$$

$$= \frac{\sigma^2}{2\alpha} e^{-\alpha(t+s)} (e^{2\alpha(s \wedge t)} - 1)$$

3. a) Give a precise description of the simplest Girsanov theorem.
 b) Let L be the line given by the equation $y = a + bt$ with $a > 0$. Define

$$\tau_L = \inf\{t : B_t = a + bt\}.$$

Use the simplest Girsanov theorem (Brownian motion with drift) to derive the probability density function $f_{\tau_L}(t)$ of τ_L .

sol:

(a)

If the process $\{B_t\}$ is a P-BM and Q is the measure on $C[0, T]$ induced by the process $X_t = B_t + \mu t$, then every bounded Borel measurable function W on the space $C[0, T]$ satisfies

$$E_Q(W) = E_P(W M_T),$$

where M_t is the P-mtgle defined by $M_t = \exp(\mu B_t - \frac{\mu^2 t}{2})$

(b)

$$\begin{aligned}
\tau_L &= \inf\{t : B_t = a + bt\}, \tau_a = \inf\{t : X_t = a\}, \text{ where } X_t = B_t - bt \\
P(\tau_L \leq t) &= Q(\tau_a \leq t) = E_Q(1(\tau_a \leq t)) = E_P(1(\tau_a \leq t)M_T) \text{ (simplest Girsanov theorem)} \\
\text{where } M_t &= \exp\{-bB_T - \frac{b^2}{2}T\} \\
&= E_P(1(\tau_a \leq t)M_{t \wedge \tau_a}) \text{ (}\{\tau_a \leq t\} \text{ is } \mathcal{F}_{t \wedge \tau_a} \text{ measurable)} \\
&= E_P(1(\tau_a \leq t) \exp(-ab - \frac{b^2}{2}\tau_a)) \\
&= \int_0^t \exp(-ab - \frac{b^2}{2}s) \frac{a}{s^{3/2}} \phi(\frac{a}{\sqrt{s}}) ds \\
f_{\tau_a}(t) &= \frac{\partial}{\partial t} P(\tau_L \leq t) = \frac{a}{t^{3/2}} \phi(\frac{a+bt}{\sqrt{t}}) \text{ for } t \geq 0.
\end{aligned}$$

4. a) State the Martingale representation theorem.

b) State the Feynman-Kac representation theorem for Brownian motion.

Sol:

(a) X_t is an $\{\mathcal{F}_t\}$ -martingale, where $\{\mathcal{F}_t\}$ is the standard Brownian filtration, if $\exists a$ and T , such that $E(X_T^2) < \infty$,

and $\exists! \phi(s, \omega) \in \mathcal{H}^2[0, T]$, such that $X_t = \int_0^t \phi(s, \omega) dB_s, \forall 0 \leq t \leq T$.

(b) Suppose that the function $q : R \rightarrow R$ is bounded. Consider

$$\begin{cases} u_t(t, x) &= \frac{1}{2}u_{xx}(t, x) + q(x)u(t, x) \\ u(0, x) &= f(x) \end{cases}$$

where $f : R \rightarrow R$ is also bounded.

If $u(t, x)$ is the unique bounded solution of the function, then

$$u(t, x) = E[f(x + B_t) \exp(\int_0^t q(x + B_s) ds)]$$

5. The Black-Scholes model is assumed to be

$$dS_t = \mu S_t dt + \sigma S_t dB_t,$$

$$d\beta_t = r\beta_t dt.$$

a) By using arbitrage theory to show the procedure of deriving the Black-Scholes PDE for European call option.

b) Show how to transform this PDE into a heat equation.

6. (continued)

- c) Use Fourier transform technique to derive the solution of the above heat equation.
 d) Derive Black-Scholes formula for European call option via Feynman-Kac representation.

Sol:

(a) Hedging price $Y_t = \beta_t B_t + \gamma_t S_t$.

Ito's formula:

$$dY = \left(\frac{\partial Y}{\partial t} + \mu S \frac{\partial Y}{\partial S} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 Y}{\partial S^2} \right) dt + \sigma S \frac{\partial Y}{\partial S} dW_t$$

And

$$\begin{aligned} dY &= \beta_t dB_t + \gamma_t dS_t \\ &= (r\beta_t B_t + \mu\gamma_t S_t) dt + \sigma\gamma_t S_t dW_t \end{aligned}$$

Matching coefficient:

$$\begin{cases} \gamma_t &= \frac{\partial Y}{\partial S} \\ \beta_t &= \frac{1}{rB_t} \left(\frac{\partial Y}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 Y}{\partial S^2} \right) \end{cases}$$

$$\therefore Y_t = \frac{1}{rB_t} \left(\frac{\partial Y}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 Y}{\partial S^2} \right) B_t + \frac{\partial Y}{\partial S} S_t$$

$$\Rightarrow \begin{cases} \frac{\partial Y}{\partial t}(T, S_t) + \frac{1}{2} \sigma^2(T, S_t) S_t^2 \frac{\partial^2 Y}{\partial S^2}(T, S_t) = 0 \\ Y(T, S) = f(T, S) = (S - K)^+ \end{cases}$$

(b) Let

$$\begin{cases} \theta = \sigma^2(T - t) \\ z = \ln S + \left(r - \frac{\sigma^2}{2}\right)(T - t) \end{cases}$$

Set $u(\theta, z) = e^{r(T-t)Y(t, S_t)}$, then

$$\begin{cases} \frac{\partial u}{\partial \theta} = \frac{1}{2} \frac{\partial^2 u}{\partial z^2} \\ u(0, z) = f(z) = (e^z - K)^+ \end{cases}$$

(c) Take Fourier transform:

$$\begin{cases} \frac{\partial}{\partial t} \hat{u}(w, \theta) = \frac{1}{2}(iw)^2 \hat{u}(w, \theta) = -\frac{1}{2} \hat{u}(w, \theta) \\ \hat{u}(0, w) = \hat{f}(w) \end{cases}$$

$$\Rightarrow \hat{u}(\theta, w) = \hat{f}(w) e^{-\frac{1}{2} w^2 \theta}$$

Inverse Fourier transform:

$$u(\theta, z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \hat{f}(w) e^{-\frac{1}{2} w^2 \theta} e^{iwz} dw$$

By transform of Gaussian, we have $\mathcal{F}(e^{-\frac{z^2}{2\theta}})(w) = \sqrt{\theta} e^{-\frac{1}{2} w^2 \theta}$
 $\Rightarrow e^{-\frac{z^2}{2\theta}} = \mathcal{F}\left(\frac{1}{\sqrt{\theta}} e^{-\frac{z^2}{2\theta}}\right)(w) \equiv \mathcal{F}(g_{\theta}(z))(w)$

Therefore,

$$\begin{aligned} u(\theta, z) &= f * g_{\theta}(z) \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(y) g_{\theta}(z - y) dy \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(y) \frac{1}{\sqrt{\theta}} e^{-\frac{(z-y)^2}{2\theta}} dy \\ &= \frac{1}{\sqrt{2\pi\theta}} \int_{-\infty}^{\infty} f(y) e^{-\frac{(z-y)^2}{2\theta}} dy \\ &= \frac{1}{\sqrt{2\pi\theta}} \int_{-\infty}^{\infty} (e^y - K)^+ e^{-\frac{(z-y)^2}{2\theta}} dy \end{aligned}$$

(d) $y = W_{\theta} + z \Rightarrow dy = dW_{\theta}$, where $\{W_{\theta}\}$ is the standard Brownian motion.

$$\begin{aligned} E(e^{W_{\theta}+z} - K)^+ &= E(e^z e^{\sqrt{\theta} W_1} - K)^+ \\ &= E(e^{z+\frac{\theta}{2}} e^{\sqrt{\theta} W_1 - \frac{\theta}{2}} - K)^+ \\ &= e^{z+\frac{\theta}{2}} \Phi\left(\frac{z-\ln K+\theta}{\sqrt{\theta}}\right) - K e^{z+\frac{\theta}{2}} \Phi\left(\frac{z-\ln K}{\sqrt{\theta}}\right) \\ &= S_0 \Phi\left(\frac{\ln(\frac{S_0}{K})+(r+\frac{\sigma^2}{2})(T-t)}{\sigma\sqrt{T-t}}\right) - K e^{-rT} \Phi\left(\frac{\ln(\frac{S_0}{K})+(r-\frac{\sigma^2}{2})(T-t)}{\sigma\sqrt{T-t}}\right) \end{aligned}$$

7. Derive the pricing formula for European put option by using the technique of change of numeraire and Girsanov theorem.

$$\text{payoff } X = (K - S(T))^+$$

$$p = \tilde{E}[e^{-rT} (K - S(T))^+]$$

$$= \tilde{E}[e^{-rT} (K - S(T)) I(K \geq S(T))]$$

$$= e^{-rT} K \tilde{P}(K \geq S(T)) - e^{-rT} \tilde{E}[S(T) I(K \geq S(T))]$$

$$= (1) - (2)$$

$$\begin{aligned}
K &\geq S_0 e^{(r-\frac{\sigma^2}{2})T + \sigma\sqrt{T}Z} \Leftrightarrow z \leq -\frac{1}{\sigma\sqrt{T}}[\ln(\frac{S_0}{K}) + (r-\frac{\sigma^2}{2})T] \equiv -d_2, \\
\text{where } d_2 &= \frac{1}{\sigma\sqrt{T}}[\ln(\frac{S_0}{K}) + (r-\frac{\sigma^2}{2})T] \\
(1) &= K e^{-rT} \tilde{P}(z \leq -d_2) = K e^{-rT} \Phi(-d_2) \\
(2) &= S_0 \tilde{E}[e^{-rT} \frac{S(T)}{S(0)} I(K \geq S(T))] \\
(\text{Let } \frac{d\tilde{P}}{d\tilde{P}} &= e^{-rT} \frac{S(T)}{S(0)} = e^{\sigma\tilde{W}_T - \frac{\sigma^2}{2}T} \\
\text{Let } \tilde{W}_t &= \tilde{W}_t - \sigma t \\
k &\geq S_0 e^{(r-\frac{\sigma^2}{2})T + \sigma\sqrt{T}Z} = S_0 e^{(r+\frac{\sigma^2}{2})T + \sigma\tilde{W}_T} \\
&\Rightarrow z \leq -d_1 \\
&= S_0 \tilde{P}(z \leq -d_1) = S_0 \Phi(-d_1) \text{ Therefore, } p = K e^{-rT} \Phi(-d_2) - S_0 \Phi(-d_1)
\end{aligned}$$

8. Barrier option pricing:

Find the arbitrage price of a contingent claim that pays M if the stock price gets to a level K or higher during the time period $[0, T]$ and that pays zero otherwise. In other words, find a formula for the arbitrage price of the claim

$$X = M I_{\{\sup_{t \in [0, T]} S_t > K\}}.$$

sol:

$$\begin{aligned}
\sup_{t \in [0, T]} S_t > K &\Leftrightarrow \sup_{t \in [0, T]} B_t + \frac{1}{\sigma}(r - \frac{\sigma^2}{2})t \geq \frac{1}{\sigma} \ln \frac{K}{S_0} \\
\text{Let } a &= \frac{1}{\sigma} \ln \frac{K}{S_0}, b = -\frac{1}{\sigma}(r - \frac{\sigma^2}{2})t \\
\tau &= \inf\{B_t = a + bt\} \\
f_\tau &= \frac{a}{\sqrt{2\pi t^3}} e^{-\frac{(a+bt)^2}{2t}} \\
u_0 &= e^{-rT} \tilde{E}[M I(\sup_{t \in [0, T]} S_t > K)] = e^{-rT} M \tilde{P}(\sup_{t \in [0, T]} S_t > K) \\
&= e^{-rT} M P(\tau \leq T) \quad (\text{Law}(\mu + \sigma B_t, t \leq T | \tilde{P}_T) = \text{Law}(r + \sigma B_t, t \leq T | P_T)) \\
&= e^{-rT} M P(\tau_L \leq T) \\
&= e^{-rT} M [1 - \Phi(\frac{\ln \frac{K}{S_0} - (r - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}) + \frac{K}{S_0} e^{\frac{2}{\sigma^2}(r - \frac{\sigma^2}{2})T} \Phi(\frac{-\ln \frac{K}{S_0} - (r - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}})]
\end{aligned}$$

9. a) State the Poisson process martingale property theorem.

b) Let $N(t)$ be a Poisson process with intensity λ . Show that

$$\exp\{\ln(1-u)N(t) + u\lambda t\}, \quad 0 < u < 1,$$

is a martingale.

Sol:

(a) Let $N(t)$ be a Poisson process with intensity λ . Then $M(t) = N(t) - \lambda t$ is a martingale.

(b)

$$\begin{aligned} E[\exp\{\ln(1-u)N(t+s)\}|\mathcal{F}_t] &= E[\exp\{\ln(1-u)N(t) + \ln(1-u)(N(t+s) - N(t))\}|\mathcal{F}_t] \\ &= \exp\{\ln(1-u)N(t)\}E[\exp\{\ln(1-u)(N(t+s) - N(t))\}|\mathcal{F}_t] \\ &= \exp\{\ln(1-u)N(t)\}E[\exp\{\ln(1-u)(N(t+s) - N(t))\}] \\ &= \exp\{\ln(1-u)N(t)\} \exp\{-u\lambda s\} \end{aligned}$$

Therefore, $E[\exp\{\ln(1-u)N(t+s) + u\lambda(t+s)\}|\mathcal{F}_t] = \exp\{\ln(1-u)N(t) + u\lambda t\}$

10. Consider the stock and bond model given by

$$dS_t = \mu(t, S_t)dt + \sigma(t, S_t)dB_t \quad \text{and} \quad d\beta_t = r(t, S_t)\beta_t dt,$$

where all of the model coefficients $\mu(t, S_t)$, $\sigma(t, S_t)$, and $r(t, S_t)$ are given by explicit functions of the current time and the current stock price. Use the coefficient matching method to show that arbitrage price at time t of a European option with terminal time T and payout $h(S_T)$ is given by $f(t, S_t)$, where $f(t, S_t)$ is the solution of the terminal value problem

$$\begin{aligned} f_t(t, S_t) &= -\frac{1}{2}\sigma^2(t, x)f_{xx}(t, x) - r(t, x)xf_x(t, x) + r(t, x)f(t, x), \\ f(T, x) &= h(x). \end{aligned}$$

Sol:

(a) For European call option, the payout is $h(S_T) = (S_T - K)^+$.

Consider the replicating portfolio at time t , V_t

Let a_t : the number of units of stock.

b_t : the number of units of the bond.

Total value of the portfolio at time t is

$$V_t = a_t S_t + b_t \beta_t$$

We require that the restructuring of the portfolio be self-financing.

So, the requirement is

$$dV_t = a_t dS_t + b_t d\beta_t$$

Hence put (1) and (2) into (4)

$$\begin{aligned} dV_t &= \mu dt + \sigma(t, S_t)dB_t \\ &= [a_t u(t, S_t) + b_t r(t, S_t)\beta_t]dt + a_t \sigma(t, S_t)dB_t \end{aligned}$$

$\therefore V_t = f(t, S_t)$ and Ito formula for geometric Brownian Motion. \Rightarrow

$$\begin{aligned} dV_t &= f_t(t, S_t)dt + (1/2)f_{xx}(t, S_t)dS_t dS_t + f_x(t, S_t)dS_t \\ &= [f_t(t, S_t) + (1/2)f_{xx}(t, S_t)\sigma^2(t, S_t) + f_x(t, S_t)\mu(t, S_t)]dt + f_x(t, S_t)\sigma(t, S_t)dB_t \end{aligned}$$

Compare (1) and (2) \Rightarrow

$$\begin{aligned} a_t &= f_x(t, S_t) \\ b_t &= \frac{1}{r(t, S_t)\beta_t}(f_t(t, S_t) + 1/2f_{xx}(t, S_t)\sigma^2(t, S_t)) \\ \therefore V_t &= f(t, S_t) \\ &= a_t S_t + b_t \beta_t \\ &= f_x(t, S_t)S_t + \frac{1}{r(t, S_t)\beta_t}(f_t(t, S_t) + 1/2f_{xx}(t, S_t)\sigma^2(t, S_t)) \end{aligned}$$

The arbitrage price at time t of European option with terminal time T is $f(t, x)$ which is the solution of the terminal value problem.

$$\begin{cases} f_t(t, x) &= -\frac{1}{2}\sigma^2(t, x)f_{xx}(t, x) - r(t, x)xf_x(t, x) + r(t, x)f(t, x) \\ f(T, x) &= h(x) \end{cases}$$

(b)from (a)

$$\begin{aligned} a_t &= f_x(t, S_t) \\ b_t &= \frac{1}{r(t, S_t)\beta_t}(f_t(t, S_t) + 1/2f_{xx}(t, S_t)\sigma^2(t, S_t)) \\ V_t &= f(t, S_t) = a_t S_t + b_t \beta_t \end{aligned}$$

Hence they replicate $h(S_T)$.

11. The price of a security U is defined as the product of two asset prices R and S , i.e., $U_t = R_t S_t$. Suppose R and S are Ito processes given by

$$\begin{aligned} dR_t &= 2(100 - R_t)dt + S_t dB_t, \\ dS_t &= R_t S_t dt + S_t dB_t. \end{aligned}$$

Show that $\{tU_t, t \geq 0\}$ is also an Ito process, and find its expected instantaneous rate of return at time 0 if $R_0 = 0.15$ and $S_0 = 35$. Sol:

(a).

$$\begin{aligned} dU_t &= U_R dR + U_S dS + U_{RS} dR dS \\ &= S_t(2(100 - R_t)dt + S_t dB_t) + R_t(R_t S_t dt + S_t dB_t) + S_t^2 dt \\ &= (200S_t - 2S_t R_t + R_t^2 S_t + S_t^2)dt + (S_t^2 + R_t S_t)dB_t \end{aligned}$$

Let $f = tU_t$, by Ito formula

$$\begin{aligned} dtU_t &= f_t dt + f_U dU + \frac{1}{2} f_{UU} (dU)^2 \\ &= U_t dt + t dU_t \\ &= U_t dt + (200tS_t - 2tU_t + tU_t R_t + tS_t^2)dt + (tS_t^2 + tR_t S_t)dB_t \end{aligned}$$

(b).

$$\begin{aligned} E\left(\frac{dtU_t}{dt}\right) &= U_t dt + 200tS_t - 2tU_t + tU_t R_t + tS_t^2 \\ &= U_0 + 0 \quad (t \rightarrow \infty) \\ &= R_0 S_0 \\ &= 0.15 \cdot 35 \\ &= 5.25 \end{aligned}$$

12. Assume that X follows a geometric Brownian motion with drift α and volatility σ . The economy is risk-neutral, and the risk-free rate of interest is r . A machine prints a certificate worth $X(t)$ at random time t generated by a Poisson arrival process with intensity λ . What is the value of the machine? Hint: By how much should the asset value change at the time the certificate is printed? Assume that V is linear in X .

sol:

$dX = \alpha X dt + \sigma X dW$ To find the value V of the machine, which pays a cash flow of $X(t)$ determined by the Poisson process:

$$V = V(X)$$

By Ito's lemma:

$$\begin{aligned} dV &= V_X dX + \frac{1}{2} V_{XX} (dX)^2 \\ &= [\alpha X V_X + \frac{1}{2} \sigma^2 X^2 V_{XX}] dt + \sigma X V_X dW \end{aligned}$$

$$\text{Expected cash gain (ECG)} = E[dV] = [\alpha X V_X + \frac{1}{2} \sigma^2 X^2 V_{XX}] dt$$

$$\text{Expected cash flow (ECF)} = X \lambda dt \quad (\text{a cash flow paid with probability } \lambda dt)$$

$$\text{Total Revenue (TR)} = \text{ECG} + \text{ECF} = [\alpha X V_X + \frac{1}{2} \sigma^2 X^2 V_{XX} + X \lambda] dt$$

Arbitrage theory:

$$[\alpha X V_X + \frac{1}{2} \sigma^2 X^2 V_{XX} + X \lambda] dt = r V dt$$

$$\Rightarrow \alpha X V_X + \frac{1}{2} \sigma^2 X^2 V_{XX} + X \lambda = rV$$

$$\text{Let } V = AX + B, V_X = A, V_{XX} = 0$$

$$r(AX + B) = \alpha X A + 0 + X \lambda$$

$$\Rightarrow rA = \alpha A + \lambda rB = 0$$

$$A = \frac{\lambda}{r-\alpha}, B = 0$$

$$\text{Hence, } V = \frac{\lambda X}{r-\alpha}$$