

**MATH7091 – Assignment 4**  
**Joel Thomas 44793203**

**1.**

If  $\{X_t\}_{t \in [0, T]}$  is a  $(\mathbb{P}, \{\mathcal{F}_t\}_{t \in [0, T]})$ -martingale, then  $\forall 0 \leq s \leq t \leq T$ :

$$\mathbb{E}_{\mathbb{P}}[X_t | \mathcal{F}_s] = X_s$$

Let  $X_{t+dt} = X_t + dX_t$ , we want  $\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t$ :

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = \mathbb{E}_{\mathbb{P}}[X_t + dX_t | \mathcal{F}_t]$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = \mathbb{E}_{\mathbb{P}}[X_t + \mu_t dt + \sigma_t dW_t | \mathcal{F}_t]$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = \mathbb{E}_{\mathbb{P}}[X_t | \mathcal{F}_t] + \mathbb{E}_{\mathbb{P}}[\mu_t dt | \mathcal{F}_t] + \mathbb{E}_{\mathbb{P}}[\sigma_t dW_t | \mathcal{F}_t], \quad (\text{linearity})$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t + \mu_t dt + \sigma_t \mathbb{E}_{\mathbb{P}}[dW_t | \mathcal{F}_t], \quad (X_t, \mu_t, dt, \sigma_t \text{ are } \mathcal{F}_t\text{-measurable})$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t + \mu_t dt + \sigma_t \mathbb{E}_{\mathbb{P}}[dW_t], \quad (\text{BM has independent increments})$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t + \mu_t dt + \sigma_t \cdot 0, \quad (dW_t \sim N(0, dt))$$

$$\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t + \mu_t dt$$

Thus,  $\mathbb{E}_{\mathbb{P}}[X_{t+dt} | \mathcal{F}_t] = X_t$  if and only if  $\mu_t = 0 \forall t \in [0, T]$  almost surely.

**2.**

**a)**

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

$$dB_t = r_t B_t dt$$

Need to first find dynamics of  $\left\{\frac{S_t}{B_t}\right\}_{t \in [0, T]}$  under  $\mathbb{P}$  before performing change of measure using Girsanov's theorem.

First find dynamics of  $\left\{\frac{1}{B_t}\right\}_{t \in [0, T]}$ :

$$f(x) = \frac{1}{x}, \quad f'(x) = -\frac{1}{x^2}, \quad f''(x) = \frac{2}{x^3}$$

$$\therefore d\left(\frac{1}{B_t}\right) = \frac{\partial f}{\partial x}(B_t)dB_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(B_t)(dB_t)^2$$

$$d\left(\frac{1}{B_t}\right) = -\frac{1}{B_t^2}(r_t B_t dt) + \frac{1}{2} \cdot \frac{2}{B_t^3}(r_t B_t dt)^2$$

$$d\left(\frac{1}{B_t}\right) = -\frac{r_t}{B_t} dt + \frac{1}{2} \cdot \frac{2}{B_t^3} \cdot 0, \quad (\text{by Ito rules})$$

$$d\left(\frac{1}{B_t}\right) = -\frac{r_t}{B_t} dt$$

Using Ito's product rule:

$$d\left(\frac{S_t}{B_t}\right) = S_t d\left(\frac{1}{B_t}\right) + \frac{1}{B_t} dS_t + (dS_t) \left(d\left(\frac{1}{B_t}\right)\right)$$

$$d\left(\frac{S_t}{B_t}\right) = S_t \left(-\frac{r_t}{B_t} dt\right) + \frac{1}{B_t} (\mu S_t dt + \sigma S_t dW_t) + (\mu S_t dt + \sigma S_t dW_t) \left(-\frac{r_t}{B_t} dt\right)$$

$$d\left(\frac{S_t}{B_t}\right) = -r_t \frac{S_t}{B_t} dt + \mu \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} dW_t + 0, \quad (\text{by Ito rules})$$

$$d\left(\frac{S_t}{B_t}\right) = (\mu - r_t) \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} dW_t$$

Thus,  $\left\{\frac{S_t}{B_t}\right\}_{t \in [0, T]}$  is clearly not a martingale under  $\mathbb{P}$  because the  $dt$ -term has a non-zero coefficient. Taking  $\alpha_t = \frac{\mu - r_t}{\sigma}$

in Girsanov's theorem, we have:

$$\tilde{W}_t = W_t + \int_0^t \alpha_u du$$

$$\rightarrow d\tilde{W}_t = dW_t + \alpha_t dt$$

Therefore, under  $\mathbb{Q}$ , we have:

$$d\left(\frac{S_t}{B_t}\right) = (\mu - r_t) \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} dW_t$$

$$d\left(\frac{S_t}{B_t}\right) = (\mu - r_t) \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} (d\tilde{W}_t - \alpha_t dt)$$

$$d\left(\frac{S_t}{B_t}\right) = (\mu - r_t) \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} d\tilde{W}_t - \sigma \left(\frac{\mu - r_t}{\sigma}\right) \frac{S_t}{B_t} dt$$

$$d\left(\frac{S_t}{B_t}\right) = (\mu - r_t) \frac{S_t}{B_t} dt - (\mu - r_t) \frac{S_t}{B_t} dt + \sigma \frac{S_t}{B_t} d\tilde{W}_t$$

$$d\left(\frac{S_t}{B_t}\right) = \sigma \frac{S_t}{B_t} d\tilde{W}_t$$

Thus,  $\left\{\frac{S_t}{B_t}\right\}_{t \in [0, T]}$  is clearly a martingale under  $\mathbb{Q}$  because the  $dt$ -term has a zero coefficient.

**b)**

Since  $dB_t = r_t B_t dt$ , we have under both  $\mathbb{P}$  and  $\mathbb{Q}$ :

$$B_t = e^{\int_0^t r_u du}$$

Using the following facts:

- Fact 1 – by the Martingale Representation theorem,  $\exists$  a replicating portfolio  $\{\Theta_t\}_{t \in [0, T]}$  under the risk-neutral measure  $\mathbb{Q}$  whose value process  $\{V_t\}_{t \in [0, T]}$  replicates the financial contract payoff  $\{C_t\}_{t \in [0, T]}$ .
- Fact 2 – from a),  $\left\{\frac{S_t}{B_t}\right\} = \left\{e^{-\int_0^t r_u du} S_t\right\}$  is a  $(\mathbb{Q}, \{\mathcal{F}_t\}_{t \in [0, T]})$ -martingale so the discounted price of any portfolio of  $(S, B)$  is a  $(\mathbb{Q}, \{\mathcal{F}_t\}_{t \in [0, T]})$ -martingale.
- Fact 3 – it can be shown that  $\mathbb{P}$  and  $\mathbb{Q}$  from Girsanov's theorem are equivalent.

Then  $\exists$  a replicating portfolio under  $\mathbb{Q}$  whose discounted value  $\left\{\frac{V_t}{B_t}\right\} = \left\{e^{-\int_0^t r_u du} V_t\right\}$  is a  $\mathbb{Q}$ -martingale:

$$e^{-\int_0^t r_u du} C_t = e^{-\int_0^t r_u du} V_t$$

$$e^{-\int_0^t r_u du} C_t = \mathbb{E}_{\mathbb{Q}} \left[ e^{-\int_0^T r_u du} V_T \middle| \mathcal{F}_t \right], \quad \left( \left\{ e^{-\int_0^t r_u du} V_t \right\} \text{ is a } \mathbb{Q}\text{-martingale} \right)$$

$$e^{-\int_0^t r_u du} C_t = \mathbb{E}_{\mathbb{Q}} \left[ e^{-\int_0^T r_u du} C_T \middle| \mathcal{F}_t \right], \quad (V_T = C_T \text{ at maturity})$$

$$\rightarrow C_t = e^{\int_0^t r_u du} \mathbb{E}_{\mathbb{Q}} \left[ e^{-\int_0^T r_u du} C_T \middle| \mathcal{F}_t \right]$$

$$C_t = \mathbb{E}_{\mathbb{Q}} \left[ e^{\int_0^t r_u du} e^{-\int_0^T r_u du} C_T \middle| \mathcal{F}_t \right], \quad \left( e^{\int_0^t r_u du} \text{ is } \mathcal{F}_t\text{-measurable} \right)$$

$$C_t = \mathbb{E}_{\mathbb{Q}} \left[ e^{-\int_t^T r_u du} C_T \middle| \mathcal{F}_t \right]$$

3.

a)

Let  $C_t$  denote the option price at time  $t$  with the payoff at maturity given by  $C_T = (Y_T - K)^+$ . We have:

$$C_t = e^{-R_{disc}(T-t)} \mathbb{E}_{\mathbb{Q}}[C_T | \mathcal{F}_t]$$

$$C_t = e^{-R_{disc}(T-t)} \mathbb{E}_{\mathbb{Q}}[(Y_T - K)^+ | \mathcal{F}_t]$$

$$C_t = e^{-R_{disc}(T-t)} \mathbb{E}_{\mathbb{Q}}[(Y_T - K) \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t]$$

$$C_t = e^{-R_{disc}(T-t)} \left[ \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] - \mathbb{E}_{\mathbb{Q}}[K \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] \right], \quad (\text{linearity})$$

$$C_t = e^{-R_{disc}(T-t)} \left[ \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] - K \mathbb{E}_{\mathbb{Q}}[\mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] \right]$$

$$C_t = e^{-R_{disc}(T-t)} \left[ \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] - K \mathbb{Q}(Y_T > K | \mathcal{F}_t) \right], \quad (\mathbb{E}_{\mathbb{Q}}[\mathbb{I}_A] = \mathbb{Q}(A))$$

Let 1) =  $\mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t]$  and 2) =  $\mathbb{Q}(Y_T > K | \mathcal{F}_t)$ . Solving for 2) first, under  $\mathbb{Q}$  and conditional on  $\mathcal{F}_t$ , we have:

$$\log(Y_T) \sim N \left( \log(Y_t) + \left( R_{grow} - \frac{s^2}{2} \right) (T - t), s^2 (T - t) \right)$$

$$\therefore \mathbb{Q}(Y_T > K | \mathcal{F}_t)$$

$$= \mathbb{Q}(\log(Y_T) > \log(K) | \mathcal{F}_t)$$

$$= \mathbb{Q} \left( \frac{\log(Y_T) - \log(Y_t) - \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{\sqrt{s^2 (T - t)}} > \frac{\log(K) - \log(Y_t) - \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{\sqrt{s^2 (T - t)}} \right),$$

(normalising to  $Z \sim N(0,1)$ )

$$= \mathbb{Q} \left( \frac{\log \left( \frac{Y_T}{Y_t} \right) - \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{s \sqrt{T - t}} > \frac{\log \left( \frac{K}{Y_t} \right) - \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{s \sqrt{T - t}} \right)$$

$$= N \left( - \frac{\log \left( \frac{K}{Y_t} \right) - \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{s \sqrt{T - t}} \right), \quad (\text{by symmetry of the normal distribution, } 1 - N(x) = N(-x))$$

$$\begin{aligned}
&= N \left( \frac{\log \left( \left( \frac{K}{Y_t} \right)^{-1} \right) + \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{s\sqrt{T - t}} \right) \\
&= N \left( \frac{\log \left( \frac{Y_t}{K} \right) + \left( R_{grow} - \frac{s^2}{2} \right) (T - t)}{s\sqrt{T - t}} \right) \\
&= N \left( \frac{\log \left( \frac{Y_t}{K} \right) + \left( R_{grow} + \frac{s^2}{2} \right) (T - t)}{s\sqrt{T - t}} - s\sqrt{T - t} \right) \\
&= N(d_1 - s\sqrt{T - t}) \\
&= N(d_2)
\end{aligned}$$

Now solving for 1):

$$\begin{aligned}
&\mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] \\
&= e^{R_{grow}(T-t)} Y_t \mathbb{E}_{\mathbb{Q}} \left[ \frac{Y_T}{e^{R_{grow}(T-t)} Y_t} \mathbb{I}_{\{Y_T > K\}} \middle| \mathcal{F}_t \right], \quad (e^{R_{grow}(T-t)} Y_t \text{ is } \mathcal{F}_t\text{-measurable}) \\
&= F_t \mathbb{E}_{\mathbb{Q}} \left[ \frac{Y_T}{F_t} \mathbb{I}_{\{Y_T > K\}} \middle| \mathcal{F}_t \right], \quad (F_t = e^{R_{grow}(T-t)} Y_t) \\
&= F_t \mathbb{E}_{\mathbb{Q}} \left[ e^{\log(\frac{Y_T}{F_t})} \mathbb{I}_{\{\log(\frac{Y_T}{F_t}) > \log(\frac{K}{F_t})\}} \middle| \mathcal{F}_t \right]
\end{aligned}$$

Since under  $\mathbb{Q}$  conditional on  $\mathcal{F}_t$ :

$$\begin{aligned}
\log(Y_T) &\sim N \left( \log(Y_t) + \left( R_{grow} - \frac{s^2}{2} \right) (T - t), s^2(T - t) \right) \\
\log(Y_T) &\sim N \left( \log(e^{R_{grow}(T-t)} Y_t) - \frac{s^2}{2} (T - t), s^2(T - t) \right) \\
\rightarrow \log(Y_T) - \log(e^{R_{grow}(T-t)} Y_t) &\sim N \left( -\frac{s^2}{2} (T - t), s^2(T - t) \right)
\end{aligned}$$

$$\therefore \log\left(\frac{Y_T}{F_t}\right) \sim N\left(-\frac{s^2}{2}(T-t), s^2(T-t)\right)$$

Let  $y = \log\left(\frac{Y_T}{F_t}\right)$ . Continuing:

$$\begin{aligned} & \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] \\ &= F_t \int_{\log(\frac{K}{F_t})}^{\infty} e^y \cdot \frac{1}{\sqrt{2\pi} \cdot s\sqrt{T-t}} e^{-\frac{1}{2} \frac{\left(y + \frac{s^2}{2}(T-t)\right)^2}{s^2(T-t)}} dy \\ &= F_t \int_{\log(\frac{K}{F_t})}^{\infty} \frac{1}{\sqrt{2\pi} \cdot s\sqrt{T-t}} e^{-\frac{1}{2} \frac{\left(y + \frac{s^2}{2}(T-t)\right)^2}{s^2(T-t)} + y} dy \end{aligned}$$

Observe that:

$$\begin{aligned} & -\frac{1}{2} \frac{\left(y + \frac{s^2}{2}(T-t)\right)^2}{s^2(T-t)} + y \\ &= \frac{-\left(y^2 + ys^2(T-t) + \frac{s^4}{4}(T-t)^2\right) + 2ys^2(T-t)}{2s^2(T-t)} \\ &= \frac{-\left(y^2 - ys^2(T-t) + \frac{s^4}{4}(T-t)^2\right)}{2s^2(T-t)} \\ &= -\frac{1}{2} \frac{\left(y - \frac{s^2}{2}(T-t)\right)^2}{s^2(T-t)} \end{aligned}$$

Continuing:

$$\begin{aligned} & \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] \\ &= F_t \int_{\log(\frac{K}{F_t})}^{\infty} \frac{1}{\sqrt{2\pi} \cdot s\sqrt{T-t}} e^{-\frac{1}{2} \frac{\left(y - \frac{s^2}{2}(T-t)\right)^2}{s^2(T-t)}} dy \end{aligned}$$

Apply change of variables, let  $z = \frac{y - \frac{s^2}{2}(T-t)}{s\sqrt{T-t}}$ :

$$\frac{dz}{dy} = \frac{1}{s\sqrt{T-t}} \rightarrow dy = s\sqrt{T-t} dz$$

$$y = \infty \rightarrow z = \infty$$

$$y = \log\left(\frac{K}{F_t}\right) \rightarrow z = \frac{\log\left(\frac{K}{F_t}\right) - \frac{s^2}{2}(T-t)}{s\sqrt{T-t}}$$

Continuing:

$$\mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t]$$

$$= F_t \int_{\frac{\log\left(\frac{K}{F_t}\right) - \frac{s^2}{2}(T-t)}^{\infty} \frac{1}{\sqrt{2\pi} \cdot s\sqrt{T-t}} e^{-\frac{1}{2}z^2} s\sqrt{T-t} dz$$

$$= F_t \int_{\frac{\log\left(\frac{K}{F_t}\right) - \frac{s^2}{2}(T-t)}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz$$

$$= F_t N\left(-\frac{\log\left(\frac{K}{F_t}\right) - \frac{s^2}{2}(T-t)}{s\sqrt{T-t}}\right), \quad (\text{by symmetry of the normal distribution, } 1 - N(x) = N(-x))$$

$$= F_t N\left(\frac{\log\left(\left(\frac{K}{e^{R_{grow}(T-t)} Y_t}\right)^{-1}\right) + \frac{s^2}{2}(T-t)}{s\sqrt{T-t}}\right)$$

$$= F_t N\left(\frac{\log\left(\frac{e^{R_{grow}(T-t)} Y_t}{K}\right) + \frac{s^2}{2}(T-t)}{s\sqrt{T-t}}\right)$$

$$= F_t N\left(\frac{\log\left(\frac{Y_t}{K}\right) + \left(R_{grow} + \frac{s^2}{2}\right)(T-t)}{s\sqrt{T-t}}\right)$$

$$= F_t N(d_1)$$

Combining results from 1) and 2):

$$C_t = e^{-R_{disc}(T-t)} \mathbb{E}_{\mathbb{Q}}[(Y_T - K)^+ | \mathcal{F}_t]$$

$$C_t = e^{-R_{disc}(T-t)} [\mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T > K\}} | \mathcal{F}_t] - K \mathbb{Q}(Y_T > K | \mathcal{F}_t)]$$

$$C_t = e^{-R_{disc}(T-t)} [F_t N(d_1) - KN(d_2)]$$

$$C_t = C^{BS}(Y_t, t, K, T, R_{grow}, R_{disc}, s)$$

**b)**

**i)**

$Y_T \sim N(\mu, \sigma^2)$ ,  $\mu \neq 0$ ,  $\sigma \neq 1$  in general.

$$C_T = (K - Y_T)^+$$

$$\therefore C_t = e^{-r(T-t)} \mathbb{E}_{\mathbb{Q}}[C_T | \mathcal{F}_t]$$

$$C_t = e^{-r(T-t)} \mathbb{E}_{\mathbb{Q}}[(K - Y_T)^+ | \mathcal{F}_t]$$

$$C_t = e^{-r(T-t)} \mathbb{E}_{\mathbb{Q}}[(K - Y_T) \mathbb{I}_{\{K > Y_T\}} | \mathcal{F}_t]$$

$$C_t = e^{-r(T-t)} [\mathbb{E}_{\mathbb{Q}}[K \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t] - \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t]], \quad (\text{linearity})$$

$$C_t = e^{-r(T-t)} [K \mathbb{E}_{\mathbb{Q}}[\mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t] - \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t]]$$

$$C_t = e^{-r(T-t)} [K \mathbb{Q}(Y_T < K | \mathcal{F}_t) - \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t]], \quad (\mathbb{E}_{\mathbb{Q}}[\mathbb{I}_A] = \mathbb{Q}(A))$$

Let 1) =  $\mathbb{Q}(Y_T < K | \mathcal{F}_t)$  and 2) =  $\mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t]$ . Solving for 1) first, under  $\mathbb{Q}$  and conditional on  $\mathcal{F}_t$ , we have:

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

$$\therefore \mathbb{Q}(Y_T < K | \mathcal{F}_t) = \mathbb{Q}\left(\frac{Y_T - \mu}{\sigma} < \frac{K - \mu}{\sigma} \middle| \mathcal{F}_t\right), \quad (\text{normalising to } Z \sim N(0,1))$$

$$\mathbb{Q}(Y_T < K | \mathcal{F}_t) = N\left(\frac{K - \mu}{\sigma}\right)$$



Now solving for 2):

$$\begin{aligned}
& \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t] \\
&= \int_{-\infty}^K y \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy \\
&= \int_{-\infty}^K (y - \mu) \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy + \int_{-\infty}^K \mu \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy \\
&= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^K \frac{y - \mu}{\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy + \mu \int_{-\infty}^K \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy
\end{aligned}$$

Observe that:

$$\frac{d}{dy} \left( -e^{-\frac{1(y-\mu)^2}{2\sigma^2}} \right) = \frac{y - \mu}{\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}}$$

Continuing:

$$\begin{aligned}
& \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t] \\
&= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^K \frac{d}{dy} \left( -e^{-\frac{1(y-\mu)^2}{2\sigma^2}} \right) dy + \mu \int_{-\infty}^K \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1(y-\mu)^2}{2\sigma^2}} dy \\
&= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^K \frac{d}{dy} \left( -e^{-\frac{1(y-\mu)^2}{2\sigma^2}} \right) dy + \mu \int_{-\infty}^{\frac{K-\mu}{\sigma}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz, \\
&\quad \left( \text{apply change of variables to second integral, let } z = \frac{y - \mu}{\sigma} \right) \\
&= \frac{1}{\sqrt{2\pi}} \left[ -e^{-\frac{1(y-\mu)^2}{2\sigma^2}} \right]_{y=-\infty}^{y=K} + \mu N\left(\frac{K - \mu}{\sigma}\right) \\
&= \frac{1}{\sqrt{2\pi}} \left( -e^{-\frac{1(K-\mu)^2}{2\sigma^2}} + 0 \right) + \mu N\left(\frac{K - \mu}{\sigma}\right) \\
&= -\frac{1}{\sqrt{2\pi}} e^{-\frac{1(K-\mu)^2}{2\sigma^2}} + \mu N\left(\frac{K - \mu}{\sigma}\right)
\end{aligned}$$

Combining expressions 1) and 2):

$$C_t = e^{-r(T-t)} \mathbb{E}_{\mathbb{Q}}[C_T | \mathcal{F}_t]$$

$$C_t = e^{-r(T-t)} \left[ K \mathbb{Q}(Y_T < K | \mathcal{F}_t) - \mathbb{E}_{\mathbb{Q}}[Y_T \mathbb{I}_{\{Y_T < K\}} | \mathcal{F}_t] \right]$$

$$C_t = e^{-r(T-t)} \left[ KN \left( \frac{K - \mu}{\sigma} \right) - \left( -\frac{1}{\sqrt{2\pi}} e^{-\frac{1(K-\mu)^2}{2\sigma^2}} + \mu N \left( \frac{K - \mu}{\sigma} \right) \right) \right]$$

$$C_t = e^{-r(T-t)} \left[ (K - \mu) N \left( \frac{K - \mu}{\sigma} \right) + \frac{1}{\sqrt{2\pi}} e^{-\frac{1(K-\mu)^2}{2\sigma^2}} \right]$$

$$\therefore C_0 = e^{-rT} \left[ (K - \mu) N \left( \frac{K - \mu}{\sigma} \right) + \frac{1}{\sqrt{2\pi}} e^{-\frac{1(K-\mu)^2}{2\sigma^2}} \right]$$

ii)

We wish to show that  $X_T \sim N(\mu, \sigma^2)$  for some  $\mu \neq 0, \sigma^2 \neq 1$  in general. We also know that  $\{\tilde{W}_t\}_{t \in [0, T]}$  is a  $\mathbb{Q}$ -BM.

$$dX_t = (\alpha - \beta X_t) dt + \gamma d\tilde{W}_t$$

Note that this SDE has the same form as the Vasicek interest rate SDE which was covered in Tutorial 8 Q7. Hence, citing the result from Tutorial 8 Q7.d) but replacing  $r_t, r_0, \sigma, dW_s$  appearing in the Vasicek interest rate SDE by  $X_t, X_0, \gamma, d\tilde{W}_s$  appearing in the SDE for  $X_t$ , the solution for  $X_t$  is given by:

$$X_t = e^{-\beta t} X_0 + \frac{\alpha}{\beta} (1 - e^{-\beta t}) + \gamma e^{-\beta t} \int_0^t e^{\beta s} d\tilde{W}_s$$

Which is of the form *constant + constant  $\times$  Ito integral*. We wish to find the distribution of the Ito integral:

$$\int_0^t e^{\beta s} d\tilde{W}_s = \lim_{n \rightarrow \infty} \sum_{i=1}^n e^{\beta t_i} (\tilde{W}_{t_{i+1}} - \tilde{W}_{t_i}), \quad (\text{by definition of the Ito integral})$$

Note that each:

$$\tilde{W}_{t_{i+1}} - \tilde{W}_{t_i} \sim N(0, t_{i+1} - t_i)$$

$$\rightarrow e^{\beta t_i} (\tilde{W}_{t_{i+1}} - \tilde{W}_{t_i}) \sim N(0, (t_{i+1} - t_i) e^{2\beta t_i})$$

Since each  $(\tilde{W}_{t_{i+1}} - \tilde{W}_{t_i})$  is an independent increment, using the fact that the probabilistic limit of a sum of independent normal random variables is normally distributed, we arrive at:

$$\int_0^t e^{\beta s} d\tilde{W}_s \sim N(\mu, \sigma^2)$$

We just need to find the mean and variance parameters:

$$\mathbb{E}_{\mathbb{Q}} \left[ \int_0^t e^{\beta s} d\tilde{W}_s \right] = 0, \text{ (using the property that } \mathbb{E}[\cdot] \text{ of any Ito integral is 0, see L6.32)}$$

$$\rightarrow \mathbb{E}_{\mathbb{Q}} \left[ \int_0^t e^{\beta s} d\tilde{W}_s \right]^2 = 0$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \mathbb{E}_{\mathbb{Q}} \left[ \left( \int_0^t e^{\beta s} d\tilde{W}_s \right)^2 \right] - \mathbb{E}_{\mathbb{Q}} \left[ \int_0^t e^{\beta s} d\tilde{W}_s \right]^2$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \mathbb{E}_{\mathbb{Q}} \left[ \left( \int_0^t e^{\beta s} d\tilde{W}_s \right)^2 \right]$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \mathbb{E}_{\mathbb{Q}} \left[ \int_0^t (e^{\beta s})^2 ds \right], \quad (\text{by Ito isometry})$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \mathbb{E}_{\mathbb{Q}} \left[ \int_0^t e^{2\beta s} ds \right]$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \int_0^t e^{2\beta s} ds$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \frac{1}{2\beta} [e^{2\beta s}]_{s=0}^{s=t}$$

$$\text{Var}_{\mathbb{Q}} \left( \int_0^t e^{\beta s} d\tilde{W}_s \right) = \frac{1}{2\beta} (e^{2\beta t} - 1)$$

$$\therefore \int_0^t e^{\beta s} d\tilde{W}_s \sim N \left( 0, \frac{1}{2\beta} (e^{2\beta t} - 1) \right)$$

$$\rightarrow \gamma e^{-\beta t} \int_0^t e^{\beta s} d\tilde{W}_s \sim N\left(0, \gamma^2 e^{-2\beta t} \cdot \frac{1}{2\beta} (e^{2\beta t} - 1)\right)$$

$$\gamma e^{-\beta t} \int_0^t e^{\beta s} d\tilde{W}_s \sim N\left(0, \frac{\gamma^2}{2\beta} (1 - e^{-2\beta t})\right)$$

$$\rightarrow X_t = e^{-\beta t} X_0 + \frac{\alpha}{\beta} (1 - e^{-\beta t}) + \gamma e^{-\beta t} \int_0^t e^{\beta s} d\tilde{W}_s$$

$$\sim N\left(e^{-\beta t} X_0 + \frac{\alpha}{\beta} (1 - e^{-\beta t}), \frac{\gamma^2}{2\beta} (1 - e^{-2\beta t})\right)$$

Now using result from **i)**, redefine  $C_T = (K - X_T)^+$

$$\rightarrow C_0 = e^{-rT} \left[ (K - \mu) N\left(\frac{K - \mu}{\sigma}\right) + \frac{1}{\sqrt{2\pi}} e^{-\frac{1(K-\mu)^2}{2\sigma^2}} \right]$$

$$\mu = e^{-\beta t} X_0 + \frac{\alpha}{\beta} (1 - e^{-\beta t})$$

$$\sigma^2 = \frac{\gamma^2}{2\beta} (1 - e^{-2\beta t})$$

**4.**

**a)**

Under  $\mathbb{P}$ :

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

$$dB_t = r B_t dt$$

Want  $\left\{\frac{B_t}{S_t}\right\}_{t \in [0, T]}$  to be a martingale, first need to find dynamics of  $\left\{\frac{1}{S_t}\right\}_{t \in [0, T]}$ :

$$f(x) = \frac{1}{x}, \quad f'(x) = -\frac{1}{x^2}, \quad f''(x) = \frac{2}{x^3}$$

$$\therefore d\left(\frac{1}{S_t}\right) = \frac{\partial f}{\partial x}(S_t) dS_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(S_t) (dS_t)^2$$

$$d\left(\frac{1}{S_t}\right) = -\frac{1}{S_t^2}(\mu S_t dt + \sigma S_t dW_t) + \frac{1}{2} \cdot \frac{2}{S_t^3}(\mu S_t dt + \sigma S_t dW_t)^2$$

$$d\left(\frac{1}{S_t}\right) = -\frac{\mu}{S_t}dt - \frac{\sigma}{S_t}dW_t + \frac{1}{S_t^3} \cdot (\sigma^2 S_t^2 (dW_t)^2), \quad (\text{by Ito rules})$$

$$d\left(\frac{1}{S_t}\right) = -\frac{\mu}{S_t}dt - \frac{\sigma}{S_t}dW_t + \frac{\sigma^2}{S_t}dt$$

$$d\left(\frac{1}{S_t}\right) = (\sigma^2 - \mu)\frac{1}{S_t}dt - \frac{\sigma}{S_t}dW_t$$

Using Ito's product rule:

$$d\left(\frac{B_t}{S_t}\right) = B_t d\left(\frac{1}{S_t}\right) + \frac{1}{S_t}dB_t + (dB_t)\left(d\left(\frac{1}{S_t}\right)\right)$$

$$d\left(\frac{B_t}{S_t}\right) = B_t \left( (\sigma^2 - \mu)\frac{1}{S_t}dt - \frac{\sigma}{S_t}dW_t \right) + \frac{1}{S_t}(rB_t dt) + (rB_t dt) \left( (\sigma^2 - \mu)\frac{1}{S_t}dt - \frac{\sigma}{S_t}dW_t \right)$$

$$d\left(\frac{B_t}{S_t}\right) = (\sigma^2 - \mu)\frac{B_t}{S_t}dt - \sigma\frac{B_t}{S_t}dW_t + r\frac{B_t}{S_t}dt + 0, \quad (\text{by Ito rules})$$

$$d\left(\frac{B_t}{S_t}\right) = (r + \sigma^2 - \mu)\frac{B_t}{S_t}dt - \sigma\frac{B_t}{S_t}dW_t$$

Observe that the coefficient of the  $dt$ -term is non-zero so  $\left\{\frac{B_t}{S_t}\right\}_{t \in [0, T]}$  is not a martingale under  $\mathbb{P}$ .

Taking  $\alpha_t = \alpha = -\frac{r + \sigma^2 - \mu}{\sigma}$  in Girsanov's theorem:

$$\tilde{W}_t^S = W_t + \int_0^t \alpha_u du$$

$$\rightarrow d\tilde{W}_t^S = dW_t + \alpha_t dt$$

$$d\tilde{W}_t^S = dW_t + \alpha dt$$

$$d\tilde{W}_t^S = dW_t - \left(\frac{r + \sigma^2 - \mu}{\sigma}\right)dt$$

Therefore, under  $\mathbb{Q}^S$ , we have:

$$d\left(\frac{S_t}{B_t}\right) = (r + \sigma^2 - \mu) \frac{B_t}{S_t} dt - \sigma \frac{B_t}{S_t} dW_t$$

$$d\left(\frac{S_t}{B_t}\right) = (r + \sigma^2 - \mu) \frac{B_t}{S_t} dt - \sigma \frac{B_t}{S_t} \left( d\tilde{W}_t^S + \left( \frac{r + \sigma^2 - \mu}{\sigma} \right) dt \right)$$

$$d\left(\frac{S_t}{B_t}\right) = (r + \sigma^2 - \mu) \frac{B_t}{S_t} dt - \sigma \frac{B_t}{S_t} d\tilde{W}_t^S - (r + \sigma^2 - \mu) \frac{B_t}{S_t} dt$$

$$d\left(\frac{S_t}{B_t}\right) = -\sigma \frac{B_t}{S_t} d\tilde{W}_t^S$$

Observe that the coefficient of the  $dt$ -term is zero so  $\left\{ \frac{B_t}{S_t} \right\}_{t \in [0, T]}$  is clearly a martingale under  $\mathbb{Q}^S$ . Under  $\mathbb{Q}^S$ , we also have:

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

$$dS_t = \mu S_t dt + \sigma S_t \left( d\tilde{W}_t^S + \left( \frac{r + \sigma^2 - \mu}{\sigma} \right) dt \right)$$

$$dS_t = \mu S_t dt + \sigma S_t d\tilde{W}_t^S + (r + \sigma^2 - \mu) S_t dt$$

$$dS_t = (r + \sigma^2) S_t dt + \sigma S_t d\tilde{W}_t^S$$

$$dB_t = r B_t dt, \quad (\text{no change})$$

**b)**

$$V_t = \Theta_t \cdot X_t$$

$$V_t = (a_t, b_t) \cdot (S_t, B_t)$$

$$V_t = a_t S_t + b_t B_t$$

Under  $\mathbb{P}$ , we have:

$$dV_t = a_t dS_t + b_t dB_t$$

Using Ito's product rule:

$$d\left(\frac{V_t}{S_t}\right) = V_t d\left(\frac{1}{S_t}\right) + \frac{1}{S_t} dV_t + (dV_t) \left(d\left(\frac{1}{S_t}\right)\right)$$

$$\begin{aligned} d\left(\frac{V_t}{S_t}\right) &= (a_t S_t + b_t B_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right) + \frac{1}{S_t} (a_t dS_t + b_t dB_t) \\ &\quad + (a_t dS_t + b_t dB_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right) \end{aligned}$$

Let 1) =  $(a_t S_t + b_t B_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right)$ , 2) =  $\frac{1}{S_t} (a_t dS_t + b_t dB_t)$ , 3) =  $(a_t dS_t + b_t dB_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right)$ . Evaluating 1) first:

$$\begin{aligned} &(a_t S_t + b_t B_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right) \\ &= (\sigma^2 - \mu) a_t dt - \sigma a_t dW_t + (\sigma^2 - \mu) b_t \frac{B_t}{S_t} dt - \sigma b_t \frac{B_t}{S_t} dW_t \\ &= (\sigma^2 - \mu) \left( a_t + b_t \frac{B_t}{S_t} \right) dt - \sigma \left( a_t + b_t \frac{B_t}{S_t} \right) dW_t \end{aligned}$$

Now evaluating 2):

$$\begin{aligned} &\frac{1}{S_t} (a_t dS_t + b_t dB_t) \\ &= \frac{a_t}{S_t} dS_t + \frac{b_t}{S_t} dB_t \\ &= \frac{a_t}{S_t} (\mu S_t dt + \sigma S_t dW_t) + \frac{b_t}{S_t} (r B_t dt) \\ &= \mu a_t dt + \sigma a_t dW_t + r b_t \frac{B_t}{S_t} dt \\ &= \left( \mu a_t + r b_t \frac{B_t}{S_t} \right) dt + \sigma a_t dW_t \end{aligned}$$

Finally evaluating 3):

$$\begin{aligned}
& (a_t dS_t + b_t dB_t) \left( (\sigma^2 - \mu) \frac{1}{S_t} dt - \frac{\sigma}{S_t} dW_t \right) \\
&= (\sigma^2 - \mu) a_t \frac{1}{S_t} (dS_t)(dt) - \sigma a_t \frac{1}{S_t} (dS_t)(dW_t) + (\sigma^2 - \mu) b_t \frac{1}{S_t} (dB_t)(dt) - \sigma b_t \frac{1}{S_t} (dB_t)(dW_t) \\
&= 0 - \sigma a_t \frac{1}{S_t} (\mu S_t dt + \sigma S_t dW_t)(dW_t) + 0 - 0, \quad (\text{by Ito rules}) \\
&= -\sigma a_t \frac{1}{S_t} (\sigma S_t (dW_t)^2) \\
&= -\sigma a_t \frac{1}{S_t} (\sigma S_t dt) \\
&= -\sigma^2 a_t dt
\end{aligned}$$

Combining all three expressions for 1),2) and 3):

$$\begin{aligned}
d\left(\frac{V_t}{S_t}\right) &= (\sigma^2 - \mu) \left(a_t + b_t \frac{B_t}{S_t}\right) dt - \sigma \left(a_t + b_t \frac{B_t}{S_t}\right) dW_t + \left(\mu a_t + r b_t \frac{B_t}{S_t}\right) dt + \sigma a_t dW_t - \sigma^2 a_t dt \\
d\left(\frac{V_t}{S_t}\right) &= \left(\sigma^2 a_t + \sigma^2 b_t \frac{B_t}{S_t} - \mu a_t - \mu b_t \frac{B_t}{S_t} + \mu a_t + r b_t \frac{B_t}{S_t} - \sigma^2 a_t\right) dt + \left(-\sigma a_t - \sigma b_t \frac{B_t}{S_t} + \sigma a_t\right) dW_t \\
d\left(\frac{V_t}{S_t}\right) &= (r + \sigma^2 - \mu) b_t \frac{B_t}{S_t} dt - \sigma b_t \frac{B_t}{S_t} dW_t \\
d\left(\frac{V_t}{S_t}\right) &= b_t \left( (r + \sigma^2 - \mu) \frac{B_t}{S_t} dt - \sigma \frac{B_t}{S_t} dW_t \right) \\
d\left(\frac{V_t}{S_t}\right) &= b_t d\left(\frac{B_t}{S_t}\right)
\end{aligned}$$

Under  $\mathbb{Q}^S$ :

$$\begin{aligned}
d\left(\frac{V_t}{S_t}\right) &= b_t d\left(\frac{B_t}{S_t}\right) \\
d\left(\frac{V_t}{S_t}\right) &= -\sigma b_t \frac{B_t}{S_t} d\tilde{W}_t^S
\end{aligned}$$



$\rightarrow \left\{ \frac{V_t}{S_t} \right\}_{t \in [0, T]}$  is also a  $\mathbb{Q}^S$ -martingale. Recall from **a)** that under  $\mathbb{Q}^S$ , we have:

$$dS_t = (r + \sigma^2)S_t dt + \sigma S_t d\tilde{W}_t^S$$

$$dB_t = rB_t dt$$

$\therefore dV_t = a_t dS_t + b_t dB_t = \Theta_t \cdot dX_t^S$  still holds and hence the portfolio  $\Theta$  still self-finances under the  $\mathbb{P} \rightarrow \mathbb{Q}^S$  measure change. Note that the proportional drift rate for the GBM  $\{S_t\}_{t \in [0, T]}$  has changed from  $\mu$  to  $r + \sigma^2$  but volatility stays the same.

**c)**

Using the following facts:

- Fact 1 – by the Martingale Representation theorem,  $\exists$  a replicating portfolio  $\{\Theta_t\}_{t \in [0, T]}$  under the risk-neutral measure  $\mathbb{Q}^S$  whose value process  $\{V_t\}_{t \in [0, T]}$  replicates the financial contract payoff  $\{C_t\}_{t \in [0, T]}$ .
- Fact 2 – from **a)**,  $\left\{ \frac{B_t}{S_t} \right\}$  is a  $(\mathbb{Q}^S, \{\mathcal{F}_t\}_{t \in [0, T]})$ -martingale so the  $S$ -normalised price of any portfolio of  $(S, B)$  is a  $(\mathbb{Q}, \{\mathcal{F}_t\}_{t \in [0, T]})$ -martingale.
- Fact 3 – it can be shown that  $\mathbb{P}$  and  $\mathbb{Q}$  from Girsanov's theorem are equivalent.

Then  $\exists$  a replicating portfolio under  $\mathbb{Q}^S$  whose  $S$ -normalised value  $\left\{ \frac{V_t}{S_t} \right\}$  is a  $\mathbb{Q}^S$ -martingale:

$$\frac{C_t}{S_t} = \frac{V_t}{S_t}$$

$$\frac{C_t}{S_t} = \mathbb{E}_{\mathbb{Q}^S} \left[ \frac{V_T}{S_T} \middle| \mathcal{F}_t \right]$$

$$\frac{C_t}{S_t} = \mathbb{E}_{\mathbb{Q}^S} \left[ \frac{C_T}{S_T} \middle| \mathcal{F}_t \right]$$

$$\therefore C_t = S_t \mathbb{E}_{\mathbb{Q}^S} \left[ \frac{C_T}{S_T} \middle| \mathcal{F}_t \right]$$

**d)**

$$C_T = S_T \log(S_T)$$

$$\rightarrow \frac{C_T}{S_T} = \log(S_T)$$

$$\therefore C_0 = S_0 \mathbb{E}_{\mathbb{Q}^S} \left[ \frac{C_T}{S_T} \middle| \mathcal{F}_0 \right]$$

$$C_0 = S_0 \mathbb{E}_{\mathbb{Q}^S} [\log(S_T) | \mathcal{F}_0]$$

Recall that under  $\mathbb{Q}^S$ :

$$dS_t = (r + \sigma^2)S_t dt + \sigma S_t d\tilde{W}_t^S$$

So  $\{S_t\}_{t \in [0, T]}$  is still a GBM under  $\mathbb{Q}^S$  with drift rate proportional to  $r + \sigma^2$  and the same volatility. Hence, under  $\mathbb{Q}^S$  and conditional on  $\mathcal{F}_0$ , we have:

$$\log(S_T) \sim N \left( \log(S_0) + \left( r + \frac{\sigma^2}{2} \right) T, \sigma^2 T \right)$$

So  $\mathbb{E}_{\mathbb{Q}^S} [\log(S_T) | \mathcal{F}_0]$  is simply the mean of the normally distributed random variable  $\log(S_T)$ :

$$\rightarrow C_0 = S_0 \mathbb{E}_{\mathbb{Q}^S} [\log(S_T) | \mathcal{F}_0]$$

$$C_0 = \left( \log(S_0) + \left( r + \frac{\sigma^2}{2} \right) T \right) S_0$$