

**Exercise 1:**

The objective of the exercise is to check that the following fact is true:

**Fact 1.** If a process  $Y(t)$  is a martingale under  $\mathbb{Q}$  and  $\eta_t = \frac{d\mathbb{Q}}{d\mathbb{P}}$ , then the process  $M(t) = Y(t)\eta_t$  is a martingale under  $\mathbb{P}$ .

We will focus on the case where both  $Y(t)$  and  $\eta(t)$  are modelled as diffusions processes with respective dynamics

$$dY(t) = f(t, Y(t))dt + g(t, Y(t))dX(t)$$

and

$$\frac{d\eta(t)}{\eta(t)} = -\theta(t)dX(t)$$

where  $X(t)$  is a standard Brownian motion under the  $\mathbb{P}$  measure.

**Questions -**

- (i). Knowing that  $Y(t)$  is a martingale under  $\mathbb{Q}^\theta$ , express the drift function  $f(\cdot)$  in terms of the diffusion function  $g(\cdot)$  and of the process  $\theta(t)$ .
- (ii). Apply the Itô product rule to show that  $M(t) = Y(t)\eta_t$  is a martingale under  $\mathbb{P}$ .

**Exercise 2: (Optional)**

Derive formula (25) on slide 80

$$C(t) = B(t, U)N[d_1(B(t, U), t, T)] - KB(t, T)N[d_2(B(t, U), t, T)] \quad (1)$$

where

$$\begin{aligned} d_1(b, t, T) &= \frac{\ln\left(\frac{b}{K}\right) - \ln B(t, T) + \frac{1}{2}v_U(t, T)}{v_U(t, T)} \\ d_2(b, t, T) &= d_1 - v_U(t, T) \\ v_U^2(t, T) &= \int_t^T (b(s, U) - b(s, T))^2 ds \end{aligned}$$

Start from the forward asset pricing formula given in equation (24), on slide 79,

$$C(t) = B(t, T) \mathbf{E}^{\mathbb{P}^T} [(F_B(T, T, U) - K)^+ | \mathcal{F}_t] \quad (2)$$

where the dynamics of the forward price  $F_B(t, T, U)$  is given in equations (22) and (23) on slide 78.

**Hints:**

1. you could use an approach similar to the derivation of the Black-Scholes formula presented in Section 3.3 of Lecture 3.3 (slides 63-75);
2. Note that the random variable  $Y(T) = \int_t^T (b(s, U) - b(s, T)) dX^T(s)$  is Normally distributed with mean 0 and variance  $v_U^2(t, T)$ .

## Solutions

### 1- Exercise 1

(i). By Girsanov, the  $\mathbb{Q}$ -Brownian motion  $X^{\mathbb{Q}}(t)$  is defined as

$$X_t^{\mathbb{Q}} = X_t + \int_0^t \theta(s) ds, \quad t \in [0, T] \quad (3)$$

Hence, the dynamics of  $Y(t)$  under the  $\mathbb{Q}$ -measure is given by

$$\begin{aligned} dY(t) &= f(t, Y(t))dt + g(t, Y(t))dX(t) \\ &= f(t, Y(t))dt + g(t, Y(t)) \left( dX^{\mathbb{Q}}(t) - \theta(t)dt \right) \\ &= [f(t, Y(t)) - g(t, Y(t))\theta(t)] dt + g(t, Y(t))dX^{\mathbb{Q}}(t) \end{aligned}$$

For  $Y(t)$  to be a  $\mathbb{Q}$ -martingale, we need it to be driftless, which implies that

$$f(t, Y(t)) - g(t, Y(t))\theta(t) = 0$$

Therefore, we can express the drift function  $f(t, Y(t))$  in terms of the diffusion function  $g(\cdot)$  and of the process  $\theta(t)$  as

$$f(t, Y(t)) = g(t, Y(t))\theta(t)$$

(ii). We apply the Itô product rule to derive the dynamics of  $M(t)$  under  $\mathbb{P}$ :

$$\begin{aligned} dM(t) &= d(Y_t \eta_t) \\ &= dY_t \cdot \eta_t + Y_t \cdot d\eta_t - \theta(t)\eta(t)g(t, Y(t))dt \\ &= (f(t, Y(t))dt + g(t, Y(t))dX(t)) \eta(t) \\ &\quad - \theta(t)\eta(t)Y(t)dX(t) - \theta(t)\eta(t)g(t, Y(t))dt \\ &= (f(t, Y(t))\eta(t) - \theta(t)\eta(t)g(t, Y(t))) dt \\ &\quad + \eta(t) [g(t, Y(t)) - \theta(t)Y(t)] dX(t) \end{aligned}$$

Now,  $Y(t)$  is  $\mathbb{Q}$ -martingale, which implies that  $f(t, Y(t)) = g(t, Y(t))\theta(t)$ . Substituting in the previous equation, we find that

$$dM(t) = (g(t, Y(t)) - \theta(t)Y(t)) \eta(t) dX(t)$$

The dynamics of  $M(t)$  is driftless. Therefore  $M(t)$  is a martingale under  $\mathbb{P}$ .

## 2- Derivation of Formula (24) on Slide 79

We start from the forward asset pricing formula

$$C(t) = B(t, T) \mathbf{E}^{\mathbb{P}_T} [(F_B(T, T, U) - K)^+ | \mathcal{F}_t] \quad (4)$$

which we rewrite in the (now) usual way as

$$\begin{aligned} C(t) &= B(t, T) \left( \mathbf{E}^{\mathbb{P}_T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \right. \\ &\quad \left. - K \mathbf{E}^{\mathbb{P}_T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \right) \end{aligned} \quad (5)$$

**Step 1: Evaluating the Second Expectation:  $\mathbf{E}^{\mathbb{P}_T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t]$**

$$\begin{aligned} &\mathbf{E}^{\mathbb{P}_T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\ &= \mathbb{P}_T [F_B(T, T, U) \geq K | \mathcal{F}_t] \\ &= \mathbb{P}_T \left[ F_B(t, T, U) \exp \left\{ \int_t^T (b(s, U) - b(s, T)) dX^T(s) - \frac{1}{2} v_U^2(t, T) \right\} \geq K \right] \\ &= \mathbb{P}_T \left[ \int_t^T (b(s, U) - b(s, T)) dX^T(s) \geq \ln \frac{K}{F_B(t, T, U)} + \frac{1}{2} v_U^2(t, T) \right] \end{aligned}$$

Note that the random variable  $Y(T) = \int_t^T (b(s, U) - b(s, T)) dX^T(s)$  is Normally distributed with mean 0 and variance  $v_U^2(t, T)$ .

We can then define a standard Normal random variable  $Z \sim N(0, 1)$  as  $Z = \frac{Y}{v_U(t, T)}$  and express the expectation as

$$\begin{aligned} &\mathbf{E}^{\mathbb{P}_T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\ &= \mathbb{P}_T \left[ Z \geq \frac{\ln \frac{K}{F_B(t, T, U)} + \frac{1}{2} v_U^2(t, T)}{v_U(t, T)} \right] \end{aligned}$$

By symmetry of the Normal distribution, we conclude that:

$$\begin{aligned} &\mathbf{E}^{\mathbb{P}_T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\ &= \mathbb{P}_T \left[ Z \leq \frac{\ln \frac{F_B(t, T, U)}{K} - \frac{1}{2} v_U^2(t, T)}{v_U(t, T)} \right] \\ &= N[d_2(B(t, U), t, T)] \end{aligned}$$

where

$$d_2(b, t, T) = \frac{\ln\left(\frac{b}{K}\right) - \ln B(t, T) - \frac{1}{2}v_U(t, T)}{v_U(t, T)}$$

**Step 2: Evaluating the First Expectation:**  $\mathbf{E}^{\mathbb{P}^T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t]$

$$\begin{aligned} & \mathbf{E}^{\mathbb{P}^T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\ = & F_B(t, T, U) \mathbf{E}^{\mathbb{P}^T} \left[ \exp \left\{ \int_t^T (b(s, U) - b(s, T)) dX^T(s) - \frac{1}{2}v_U^2(t, T) \right\} \mathbf{1}_{\{F_B(T, T, U) \geq K\}} \right] \end{aligned}$$

The process

$$\Lambda(t) = \exp \left\{ \int_t^T (b(s, U) - b(s, T)) dX^T(s) - \frac{1}{2}v_U^2(t, T) \right\}$$

is an exponential martingale which we can use to define a new measure  $\mathbb{Z}$  via the Radon-Nikod'ym derivative:

$$\frac{d\mathbb{Z}}{d\mathbb{P}^T} = \Lambda(t) \tag{6}$$

Under the  $\mathbb{Z}$ -measure,

$$X^{\mathbb{Z}}(t) = X^T(t) - \int_t^T (b(s, U) - b(s, T)) ds$$

is a standard Brownian motion and

$$F_B(T, T, U) = F_B(t, T, U) \exp \left\{ \int_t^T (b(s, U) - b(s, T)) dX^T(s) + \frac{1}{2}v_U^2(t, T) \right\}$$

Therefore,

$$\begin{aligned} & \mathbf{E}^{\mathbb{P}^T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\ = & F_B(t, T, U) \mathbf{E}^{\mathbb{Z}} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}}] \\ = & F_B(t, T, U) \mathbb{Z} [F_B(T, T, U) \geq K] \\ = & F_B(t, T, U) \mathbb{Z} \left[ \int_t^T (b(s, U) - b(s, T)) dX^{\mathbb{Z}}(s) \geq \ln \frac{K}{F_B(t, T, U)} - \frac{1}{2}v_U^2(t, T) \right] \end{aligned}$$

After a few additional manipulations similar to what was done in **Step 1**, we obtain:

$$\begin{aligned}
& \mathbf{E}^{\mathbb{P}^T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \\
&= F_B(t, T, U) \mathbb{Z} \left[ Z \leq \frac{\ln \frac{F_B(t, T, U)}{K} + \frac{1}{2} v_U^2(t, T)}{v_U(t, T)} \right] \\
&= F_B(t, T, U) N [d_1(B(t, U), t, T)]
\end{aligned}$$

where

$$d_1(b, t, T) = \frac{\ln \left( \frac{b}{K} \right) - \ln B(t, T) + \frac{1}{2} v_U(t, T)}{v_U(t, T)}$$

### Step 3: Concluding

Putting it all together,

$$\begin{aligned}
C(t) &= B(t, T) \left( \mathbf{E}^{\mathbb{P}^T} [F_B(T, T, U) \mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] - K \mathbf{E}^{\mathbb{P}^T} [\mathbf{1}_{\{F_B(T, T, U) \geq K\}} | \mathcal{F}_t] \right) \\
&= B(t, T) F_B(t, T, U) N [d_1(B(t, U), t, T)] - K N [d_2(B(t, U), t, T)] \\
&= B(t, U) N [d_1(B(t, U), t, T)] - K N [d_2(B(t, U), t, T)]
\end{aligned}$$

where in the last line, we have used the fact that  $F_B(t, T, U) = \frac{B(t, U)}{B(t, T)}$  (see equation (19) on slide 74).