

# QF620 Additional Examples

## Session 7: Equivalent Martingale Measure

### 1 Examples

1. Consider the following stochastic differential equation

$$dX_t = 0.16dt + 4.563dW_t,$$

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion. Is there a probability measure  $\tilde{\mathbb{P}}$  under which the drift rate of  $X_t$  is  $-0.16$  instead?

2. Consider the following stochastic differential equation

$$dY_t = \sigma dW_t,$$

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion. Express the stochastic differential equation of  $Y_t$  using a  $\tilde{\mathbb{P}}$ -Brownian motion so that it drifts at the risk-free rate  $r$ .

3. In the real-world probability measure  $\mathbb{P}$ , a process  $Z_t$  follows the stochastic differential equation

$$dZ_t = \mu Z_t dt + \sigma Z_t dW_t,$$

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion. Find a way to express  $Z_t$  in another measure  $\tilde{\mathbb{P}}$  such that the drift of  $Z_t$  under  $\tilde{\mathbb{P}}$  is  $\nu dt$  instead of  $\mu dt$ .

4. In the real-world probability measure  $\mathbb{P}$ , the stock price process  $S_t$  follows the stochastic differential equation

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

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion. What is the stochastic differential equation followed by the forward price process  $F_t$  in the real-world measure?

5. Show that by changing measure to the risk-neutral measure  $\mathbb{Q}$ , the forward price process becomes a martingale.
6. You have \$10 and will be tossing a coin twice. Each time you get a head, *your wealth increases by \$2*, and each time you get a tail, *your wealth decreases by \$2*. You believe that the coin is fair, and hence the probability of obtaining a tail or a head is the same at  $\frac{1}{2}$ . Your opponent believes that the coin is biased and that the probability of getting a head is  $\frac{1}{4}$ , while the probability of getting a tail is  $\frac{3}{4}$ . Let  $X_2$  denote the value of your wealth after two tosses. Let  $\mathbb{Q}$  denote your probability measure and let  $\mathbb{P}$  denote your opponent's probability measure. What is your expectation of your wealth after two tosses, i.e.  $\mathbb{E}^{\mathbb{Q}}[X_2]$ ? What is your opponent's expectation of your wealth after two tosses, i.e.  $\mathbb{E}^{\mathbb{P}}[X_2]$ ?

7. In the same setting as the previous question, evaluate

$$\mathbb{E}^{\mathbb{Q}} \left[ X_2 \frac{d\mathbb{P}}{d\mathbb{Q}} \right] \quad \text{and} \quad \mathbb{E}^{\mathbb{P}} \left[ X_2 \frac{d\mathbb{Q}}{d\mathbb{P}} \right].$$

8. Let  $\mathbb{P}$  denote the real-world probability measure, and  $W_t$  denote a  $\mathbb{P}$ -Brownian motion. A stock price process follows the stochastic differential equation

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

In the market, there is also a risk-free bond whose price process follows the differential equation

$$dB_t = rB_t dt.$$

Let  $\mathbb{Q}$  denote the risk-neutral probability measure. What is  $\mathbb{E}^{\mathbb{P}}[S_T]$  and  $\mathbb{E}^{\mathbb{Q}}[S_T]$ ?

9. Let  $\mathbb{P}$  denote the real-world probability measure and  $\mathbb{Q}$  denote the risk-neutral probability measure. Let  $W_t$  be a  $\mathbb{P}$ -Brownian motion and  $W_t^B$  be a  $\mathbb{Q}$ -Brownian motion. We have

$$\begin{cases} dS_t = \mu S_t dt + \sigma S_t dW_t \\ dS_t = r S_t dt + \sigma S_t dW_t^B. \end{cases}$$

Show that  $\mathbb{E}^{\mathbb{P}} \left[ S_T \frac{d\mathbb{Q}}{d\mathbb{P}} \right] = \mathbb{E}^{\mathbb{Q}}[S_T] = S_0 e^{rT}$ .

10. In a similar setting as the previous question, now show that  $\mathbb{E}^{\mathbb{Q}} \left[ S_T \frac{d\mathbb{P}}{d\mathbb{Q}} \right] = \mathbb{E}^{\mathbb{P}}[S_T] = S_0 e^{\mu T}$ .

11. Consider the following stochastic differential equation

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

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion in the real-world probability measure. Determine the probability of the event  $\{S_T > K\}$  under the real-world probability measure.

12. Consider two tradable assets in the financial market:

$$\begin{cases} dS_t = \mu S_t dt + \sigma S_t dW_t \\ dB_t = r B_t dt \end{cases}$$

where  $W_t$  is a  $\mathbb{P}$ -Brownian motion in the real-world probability measure. What is the probability of the event  $\{S_T > K\}$  under the risk-neutral probability measure  $\mathbb{Q}^*$  where the numeraire is the risk-free bond?

13. A European cash-or-nothing digital option pays

$$\mathbb{1}_{K_1 < S_T < K_2} = \begin{cases} \$1, & K_1 < S_T < K_2 \\ \$0, & \text{otherwise} \end{cases}$$

on the expiry date  $T$ . Derive a valuation formula for this option.

14. Under the risk-neutral measure  $\mathbb{Q}^*$  associated to the risk-free bond as the choice of numeraire, the stock price follows the following stochastic differential equation

$$dS_t = r S_t dt + \sigma S_t dW_t^*,$$

where  $W_t^*$  is a  $\mathbb{Q}^*$ -Brownian motion. Derive the valuation formula for an option paying  $(S_T^2 - K)^+$  on expiry date  $T$ .

15. **Discussion** We know that under the risk-neutral measure, all asset ratios are martingales, as long as the expectation is taken under the risk-neutral measure associated with the numeraire. So for instance, we have worked out in Week 5 that  $\frac{B_t}{S_t}$  is a martingale as long as we take the expectation under the risk-neutral measure associated with the stock numeraire, i.e.  $\mathbb{Q}^S$ . To this end, we have shown, using Itô's Formula, that with  $Y_t = \frac{B_t}{S_t} = f(B_t, S_t)$ , we have

$$\begin{aligned} dY_t &= (r - \mu + \sigma^2)Y_t dt - \sigma Y_t dW_t \\ &= -\sigma Y_t \left( dW_t - \frac{r - \mu + \sigma^2}{\sigma} dt \right) \\ &= -\sigma Y_t dW_t^S, \end{aligned}$$

where  $W_t^S$  is a  $\mathbb{Q}^S$ -Brownian motion, and the measure  $\mathbb{Q}^S$  is related to the real-world measure  $\mathbb{P}$  by the Radon-Nikodym derivative

$$\frac{d\mathbb{Q}^S}{d\mathbb{P}} = \exp \left[ -\frac{1}{2}\kappa^2 T - \kappa W_T \right], \quad \kappa = -\frac{r - \mu + \sigma^2}{\sigma},$$

and

$$dW_t^S = dW_t - \frac{r - \mu + \sigma^2}{\sigma} dt.$$

So under  $\mathbb{Q}^S$  measure, the stock price follows the following stochastic differential equation

$$\begin{aligned} dS_t &= \mu S_t dt + \sigma S_t dW_t \\ &= \mu S_t dt + \sigma S_t \left( dW_t^S + \frac{r - \mu + \sigma^2}{\sigma} dt \right) \\ &= (r + \sigma^2) S_t dt + \sigma S_t dW_t^S. \end{aligned}$$

This sde can be solved (Itô's Formula to  $f(S_t) = \log S_t$ ) and the solution is given by

$$S_T = S_0 e^{\left(r + \frac{\sigma^2}{2}\right)T + \sigma W_T^S}.$$

The expectation under the  $\mathbb{Q}^S$  measure is given by

$$\begin{aligned} \mathbb{E}^{\mathbb{Q}^S} [S_T] &= \mathbb{E}^{\mathbb{Q}^S} \left[ S_0 e^{\left(r + \frac{\sigma^2}{2}\right)T + \sigma W_T^S} \right] \\ &= S_0 e^{(r + \sigma^2)T}. \end{aligned}$$

Under the risk-neutral measure  $\mathbb{Q}^*$  associated to the risk-free bond  $B_t$  as the numeraire, the stock price follows the following stochastic differential equation

$$dS_t = r S_t dt + \sigma S_t dW_t^*.$$

The solution is given by

$$S_T = S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^*}.$$

The Radon-Nikodym  $\frac{d\mathbb{Q}^S}{d\mathbb{Q}^*}$  is given by

$$\frac{d\mathbb{Q}^S}{d\mathbb{Q}^*} = \frac{S_T/S_0}{B_T/B_0} = \frac{S_0 \exp \left[ \left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^* \right] / S_0}{B_0 e^{rT} / B_0} = \exp \left( -\frac{1}{2}\sigma^2 T + \sigma W_T^* \right).$$

If we take the expectation under the  $\mathbb{Q}^S$  measure, we see that we obtain

$$\begin{aligned}
\mathbb{E}^{\mathbb{Q}^S} [S_T] &= \mathbb{E}^{\mathbb{Q}^*} \left[ S_T \times \frac{d\mathbb{Q}^S}{d\mathbb{Q}^*} \right] = \mathbb{E}^{\mathbb{Q}^*} \left[ S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^*} \times e^{-\frac{1}{2}\sigma^2 T + \sigma W_T^*} \right] \\
&= S_0 e^{(r - \sigma^2)T} \mathbb{E}^{\mathbb{Q}^S} \left[ e^{2\sigma W_T^*} \right] \\
&= S_0 e^{(r - \sigma^2)T} e^{\frac{4\sigma^2 T}{2}} \\
&= S_0 e^{(r + \sigma^2)T}.
\end{aligned}$$

This is consistent with what we would expect to see.  $\triangleleft$

## 2 Suggested Solutions

1. Note that

$$\begin{aligned} dX_t &= 0.16dt + 4.563dW_t \\ &= -0.16dt + 0.32dt + 4.563dW_t \\ &= -0.16dt + 4.563 \left( dW_t + \frac{0.32}{4.563}dt \right) \end{aligned}$$

By Girsanov's Theorem, under the Radon-Nikodym derivative

$$\frac{d\tilde{\mathbb{P}}}{d\mathbb{P}} = \exp \left( -\kappa W_T - \frac{1}{2}\kappa^2 T \right) \quad \text{where} \quad \kappa = \frac{0.32}{4.563},$$

there exists a probability measure  $\tilde{\mathbb{P}}$  which is equivalent to  $\mathbb{P}$  and  $\tilde{W}_t$  is a  $\tilde{\mathbb{P}}$ -Brownian motion, and we have

$$\tilde{W}_t = W_t + \kappa t \quad \Rightarrow \quad d\tilde{W}_t = dW_t + \kappa dt.$$

So

$$dX_t = -0.16dt + 4.563d\tilde{W}_t. \quad \triangleleft$$

2. We proceed as follow:

$$\begin{aligned} dY_t &= \sigma dW_t = rdt - rdt + \sigma dW_t \\ &= rdt + \sigma \left( dW_t - \frac{r}{\sigma}dt \right). \end{aligned}$$

By Girsanov's Theorem, under the Radon-Nikodym derivative

$$\frac{d\tilde{\mathbb{P}}}{d\mathbb{P}} = \exp \left( -\kappa W_T - \frac{1}{2}\kappa^2 T \right) \quad \text{where} \quad \kappa = -\frac{r}{\sigma},$$

there exists a probability measure  $\tilde{\mathbb{P}}$  which is equivalent to  $\mathbb{P}$  and  $\tilde{W}_t$  is a  $\tilde{\mathbb{P}}$ -Brownian motion, and we have

$$\tilde{W}_t = W_t + \kappa t \quad \Rightarrow \quad d\tilde{W}_t = dW_t + \kappa dt.$$

So

$$\begin{aligned} dY_t &= rdt + \sigma \left( dW_t - \frac{r}{\sigma}dt \right) \\ &= rdt + \sigma d\tilde{W}_t. \quad \triangleleft \end{aligned}$$

3. First we write

$$\begin{aligned} dZ_t &= \mu Z_t dt + \sigma Z_t dW_t \\ &= \nu Z_t dt - \nu Z_t dt + \mu Z_t dt + \sigma Z_t dW_t \\ &= \nu Z_t dt + \sigma Z_t \left( dW_t + \frac{\mu - \nu}{\sigma}dt \right). \end{aligned}$$

By Girsanov's Theorem, under the Radon-Nikodym derivative

$$\frac{d\tilde{\mathbb{P}}}{d\mathbb{P}} = \exp \left( -\kappa W_T - \frac{1}{2}\kappa^2 T \right) \quad \text{where} \quad \kappa = \frac{\mu - \nu}{\sigma},$$

there exists a probability measure  $\tilde{\mathbb{P}}$  which is equivalent to  $\mathbb{P}$  and  $\tilde{W}_t$  is a  $\tilde{\mathbb{P}}$ -Brownian motion, and we have

$$\begin{aligned} dZ_t &= \nu Z_t dt + \sigma Z_t \left( dW_t + \frac{\mu - \nu}{\sigma} dt \right) \\ &= \nu Z_t dt + \sigma Z_t d\tilde{W}_t. \quad \triangleleft \end{aligned}$$

4. The forward price is defined as  $F_t = S_t e^{r(T-t)} = f(t, S_t)$ , where  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(t, x) = x e^{r(T-t)}$ . The partial derivatives are given by

$$\frac{\partial f}{\partial t}(t, S_t) = -rF_t, \quad \frac{\partial f}{\partial x}(t, S_t) = e^{r(T-t)}, \quad \frac{\partial^2 f}{\partial x^2}(t, S_t) = 0.$$

Applying Itô's Formula, we obtain

$$\begin{aligned} dF_t &= \frac{\partial f}{\partial t}(t, S_t)dt + \frac{\partial f}{\partial x}(t, S_t)dS_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, S_t)(dS_t)^2 \\ &= -rF_t dt + e^{r(T-t)}(\mu S_t dt + \sigma S_t dW_t) \\ &= (\mu - r)F_t dt + \sigma F_t dW_t. \quad \triangleleft \end{aligned}$$

5. As we've discussed, in the risk-neutral measure, the stock price follows

$$dS_t = rS_t dt + \sigma S_t dW_t^B,$$

where  $W_t^B$  is a  $\mathbb{Q}$ -Brownian motion. The measure change is achieved via the Radon-Nikodym derivative

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = \exp \left( -\kappa W_T - \frac{1}{2} \kappa^2 T \right) \quad \text{where} \quad \kappa = \frac{\mu - r}{\sigma},$$

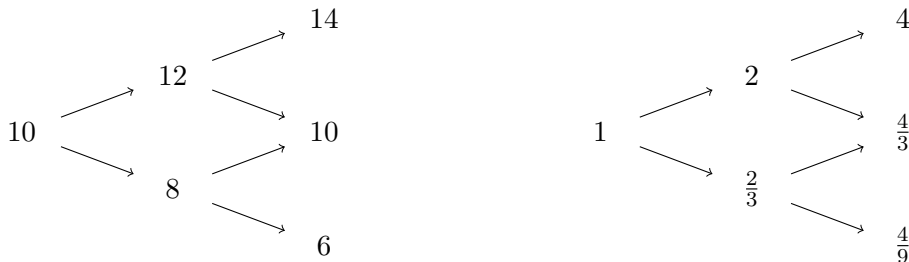
and  $W_T$  is a  $\mathbb{P}$ -Brownian motion. This allows us to define

$$W_t^B = W_t + \kappa t \quad \Rightarrow \quad dW_t^B = dW_t + \kappa dt$$

as a  $\mathbb{Q}$ -Brownian motion. Substituting, we have

$$\begin{aligned} dF_t &= (\mu - r)F_t dt + \sigma F_t dW_t \\ &= (\mu - r)F_t dt + \sigma F_t \left( dW_t^B - \frac{\mu - r}{\sigma} dt \right) \\ &= (\mu - r)F_t dt - (\mu - r)F_t dt + \sigma F_t dW_t^B \\ &= \sigma F_t dW_t^B. \quad \triangleleft \end{aligned}$$

6. The following binomial trees sketch the wealth process  $X$  and the Radon-Nikodym derivative process  $\frac{d\mathbb{Q}}{d\mathbb{P}}$ :



Taking expectations, we have

$$\begin{aligned}\mathbb{E}^{\mathbb{Q}}[X_2] &= \frac{1}{2} \cdot \frac{1}{2} \times 14 + 2 \times \frac{1}{2} \cdot \frac{1}{2} \times 10 + \frac{1}{2} \cdot \frac{1}{2} \times 6 = 10 \\ \mathbb{E}^{\mathbb{P}}[X_2] &= \frac{1}{4} \cdot \frac{1}{4} \times 14 + 2 \times \frac{1}{4} \cdot \frac{3}{4} \times 10 + \frac{3}{4} \cdot \frac{3}{4} \times 6 = 8. \quad \triangleleft\end{aligned}$$

7. Radon-Nikodym derivative allows us to evaluate expectation with a change of measure, so that  $\mathbb{E}^{\mathbb{P}}[X_2] = \mathbb{E}^{\mathbb{Q}}\left[X_2 \frac{d\mathbb{P}}{d\mathbb{Q}}\right]$  and  $\mathbb{E}^{\mathbb{Q}}[X_2] = \mathbb{E}^{\mathbb{P}}\left[X_2 \frac{d\mathbb{Q}}{d\mathbb{P}}\right]$ . This can be readily verified as follow:

$$\begin{aligned}\mathbb{E}^{\mathbb{Q}}\left[X_2 \frac{d\mathbb{P}}{d\mathbb{Q}}\right] &= \frac{1}{2} \cdot \frac{1}{2} \times 14 \times \frac{1}{4} + 2 \times \frac{1}{2} \cdot \frac{1}{2} \times 10 \times \frac{3}{4} + \frac{1}{2} \cdot \frac{1}{2} \times 6 \times \frac{9}{4} = 8 \\ \mathbb{E}^{\mathbb{P}}\left[X_2 \frac{d\mathbb{Q}}{d\mathbb{P}}\right] &= \frac{1}{4} \cdot \frac{1}{4} \times 14 \times 4 + 2 \times \frac{1}{4} \cdot \frac{3}{4} \times 10 \times \frac{4}{3} + \frac{3}{4} \cdot \frac{3}{4} \times 6 \times \frac{4}{9} = 10. \quad \triangleleft\end{aligned}$$

8. Solving the stochastic differential equation of the stock price process directly under the real-world probability measure by applying Itô's formula to the function  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(x) = \log(x)$ , with  $X_t = \log S_t = f(S_t)$ , we have

$$S_T = S_0 \exp \left[ \left( \mu - \frac{\sigma^2}{2} \right) T + \sigma W_T \right].$$

Taking expectation under the  $\mathbb{P}$  measure, we obtain

$$\mathbb{E}^{\mathbb{P}}[S_T] = S_0 e^{\mu T}.$$

Under the risk-neutral measure  $\mathbb{Q}$ , the price process  $\frac{S_t}{B_t}$  is a martingale. We are able to identify the Radon-Nikodym derivative that allows us to attain this measure change

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = \exp \left( -\kappa W_T - \frac{1}{2} \kappa^2 T \right) \quad \text{where} \quad \kappa = \frac{\mu - r}{\sigma},$$

and a  $\mathbb{Q}$ -Brownian motion  $W_t^B$ , where

$$dW_t^B = dW_t + \frac{\mu - r}{\sigma} dt.$$

Substituting, we obtain

$$\begin{aligned}dS_t &= \mu S_t dt + \sigma S_t dW_t \\ &= \mu S_t dt + \sigma S_t \left( dW_t^B - \frac{\mu - r}{\sigma} dt \right) \\ &= r S_t dt + \sigma S_t dW_t^B.\end{aligned}$$

Solving this stochastic differential equation (again by applying Itô's Formula to  $X_t = \log S_t$ ), we obtain

$$S_T = S_0 \exp \left[ \left( r - \frac{\sigma^2}{2} \right) T + \sigma W_T^B \right].$$

Taking expectation under the  $\mathbb{Q}$  measure, we obtain

$$\mathbb{E}^{\mathbb{Q}}[S_T] = S_0 e^{rT}. \quad \triangleleft$$

9. The 2<sup>nd</sup> inequality,  $\mathbb{E}^{\mathbb{Q}}[S_T] = S_0 e^{rT}$ , should be clear. To show that  $\mathbb{E}^{\mathbb{P}} \left[ S_T \frac{d\mathbb{Q}}{d\mathbb{P}} \right]$  yields the same result, first we note that

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = e^{-\kappa W_T - \frac{1}{2}\kappa^2 T} = e^{-\frac{\mu-r}{\sigma} W_T - \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T}.$$

We then proceed as follow (remember  $W_t$  is a  $\mathbb{P}$ -Brownian motion):

$$\begin{aligned} \mathbb{E}^{\mathbb{P}} \left[ S_T \frac{d\mathbb{Q}}{d\mathbb{P}} \right] &= \mathbb{E}^{\mathbb{P}} \left[ \overbrace{S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T + \sigma W_T}}^{S_T} \overbrace{e^{-\frac{\mu-r}{\sigma} W_T - \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T}}^{\frac{d\mathbb{Q}}{d\mathbb{P}}} \right] \\ &= S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \mathbb{E}^{\mathbb{P}} \left[ e^{\left(\sigma - \frac{\mu-r}{\sigma}\right) W_T} \right] \\ &= S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} e^{\frac{\left(\sigma - \frac{\mu-r}{\sigma}\right)^2 T}{2}} \\ &= S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} e^{\frac{\left(\sigma^2 - 2(\mu-r) + \left(\frac{\mu-r}{\sigma}\right)^2\right)T}{2}} \\ &= S_0 e^{rT}. \quad \triangleleft \end{aligned}$$

10. Same as before, the challenge is to show that  $\mathbb{E}^{\mathbb{Q}} \left[ S_T \frac{d\mathbb{P}}{d\mathbb{Q}} \right] = S_0 e^{\mu T}$ . Note the Radon-Nikodym derivative

$$\begin{aligned} \frac{d\mathbb{Q}}{d\mathbb{P}} &= e^{-\kappa W_T - \frac{1}{2}\kappa^2 T} = e^{-\frac{\mu-r}{\sigma} W_T - \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \\ \Rightarrow \frac{d\mathbb{P}}{d\mathbb{Q}} &= e^{\frac{\mu-r}{\sigma} W_T + \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T}. \end{aligned}$$

Also, note the relationship between the  $\mathbb{P}$ - and  $\mathbb{Q}$ -Brownian motion:

$$W_t^B = W_t + \kappa t = W_t + \frac{\mu-r}{\sigma} t \quad \Rightarrow \quad W_t = W_t^B - \frac{\mu-r}{\sigma} t,$$

and hence the Radon-Nikodym derivative can be written (in terms of  $W_t^B$ ) as

$$\begin{aligned} \frac{d\mathbb{P}}{d\mathbb{Q}} &= e^{\frac{\mu-r}{\sigma} W_T + \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \\ &= e^{\frac{\mu-r}{\sigma} (W_T^B - \frac{\mu-r}{\sigma} T) + \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \\ &= e^{\frac{\mu-r}{\sigma} W_T^B - \left(\frac{\mu-r}{\sigma}\right)^2 T + \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \\ &= e^{\frac{\mu-r}{\sigma} W_T^B - \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \end{aligned}$$

We proceed to evaluate the expectation (remember  $W_t^B$  is a  $\mathbb{Q}$ -Brownian motion):

$$\begin{aligned} \mathbb{E}^{\mathbb{Q}} \left[ S_T \frac{d\mathbb{P}}{d\mathbb{Q}} \right] &= \mathbb{E}^{\mathbb{Q}} \left[ \overbrace{S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^B}}^{S_T} \overbrace{e^{\frac{\mu-r}{\sigma} W_T^B - \frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T}}^{\frac{d\mathbb{P}}{d\mathbb{Q}}} \right] \\ &= S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} \mathbb{E}^{\mathbb{Q}} \left[ e^{\left(\sigma + \frac{\mu-r}{\sigma}\right) W_T^B} \right] \\ &= S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} e^{\frac{\left(\sigma + \frac{\mu-r}{\sigma}\right)^2 T}{2}} \\ &= S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T} e^{-\frac{1}{2}\left(\frac{\mu-r}{\sigma}\right)^2 T} e^{\frac{\left(\sigma^2 + 2(\mu-r) + \left(\frac{\mu-r}{\sigma}\right)^2\right)T}{2}} \\ &= S_0 e^{\mu T}. \quad \triangleleft \end{aligned}$$



11. The solution is given by

$$S_T = S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T + \sigma W_T}$$

The probability of the event  $\{S_T > K\}$  under the real-world probability measure  $\mathbb{P}$  is therefore given by

$$\begin{aligned} S_0 e^{\left(\mu - \frac{\sigma^2}{2}\right)T + \sigma W_T} &> K \\ \Rightarrow x &> \frac{\log \frac{K}{S_0} - \left(\mu - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} = x^*. \end{aligned}$$

So

$$\begin{aligned} \mathbb{E}^{\mathbb{P}} [\mathbb{1}_{S_T > K}] &= \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\ &= \Phi(\infty) - \Phi(x^*) \\ &= 1 - \Phi(x^*) \\ &= \Phi(-x^*) \quad \left( \because 1 - \Phi(x) = \Phi(-x) \right) \\ &= \Phi\left(\frac{\log \frac{S_0}{K} + \left(\mu - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}\right). \quad \triangleleft \end{aligned}$$

12. Under the probability measure  $\mathbb{Q}^*$ , the stock price follows the stochastic differential equation (show it)

$$dS_t = rS_t dt + \sigma S_t dW_t^*.$$

The solution to this stochastic differential equation is given by

$$S_T = S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^*}.$$

To determine the probability of the event  $\{S_T > K\}$  under  $\mathbb{Q}^*$ , the inequality to satisfy is

$$\begin{aligned} S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T} &> K \\ \Rightarrow x &> \frac{\log \frac{K}{S_0} - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} = x^*. \end{aligned}$$

So

$$\begin{aligned} \mathbb{E}^{\mathbb{Q}^*} [\mathbb{1}_{S_T > K}] &= \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\ &= \Phi(\infty) - \Phi(x^*) \\ &= 1 - \Phi(x^*) \\ &= \Phi(-x^*) \quad \left( \because 1 - \Phi(x) = \Phi(-x) \right) \\ &= \Phi\left(\frac{\log \frac{S_0}{K} + \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}\right). \quad \triangleleft \end{aligned}$$

13.  $S_t$  follows the stochastic differential equation

$$dS_t = rS_t dt + \sigma S_t dW_t^*$$

where  $W_t^*$  is a  $\mathbb{Q}^*$ -Brownian motion. The solution is given by

$$S_T = S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_T^*}.$$

Let  $V_t$  denote the value of this digital option at time  $t$ , the ratio  $\frac{V_t}{B_t}$  is a martingale under the  $\mathbb{Q}^*$  measure. We have

$$\begin{aligned} \frac{V_0}{B_0} &= \mathbb{E}^{\mathbb{Q}^*} \left[ \frac{V_T}{B_T} \right] \\ V_0 &= e^{-rT} \mathbb{E}^{\mathbb{Q}^*} [\mathbb{1}_{K_1 < S_T < K_2}] \end{aligned}$$

The inequalities to satisfy are:

$$\begin{aligned} K_1 &< S_T < K_2 \\ K_1 &< S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma \sqrt{T}x} < K_2 \end{aligned}$$

The left-hand inequality is

$$\begin{aligned} K_1 &< S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma \sqrt{T}x} \\ \Rightarrow x_L^* &= \frac{\log \frac{K_1}{S_0} - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma \sqrt{T}} < x. \end{aligned}$$

The right-hand inequality is

$$\begin{aligned} S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma \sqrt{T}x} &< K_2 \\ \Rightarrow x &< \frac{\log \frac{K_2}{S_0} - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma \sqrt{T}} = x_H^*. \end{aligned}$$

And so we can evaluate the expectation

$$\begin{aligned} V_0 &= e^{-rT} \mathbb{E}^{\mathbb{Q}^*} [\mathbb{1}_{K_1 < S_T < K_2}] \\ &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \mathbb{1}_{K_1 < S_T < K_2} e^{-\frac{x^2}{2}} dx \\ &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x_L^*}^{x_H^*} e^{-\frac{x^2}{2}} dx \\ &= e^{-rT} [\Phi(x_H^*) - \Phi(x_L^*)]. \quad \triangleleft \end{aligned}$$

14. Solving the SDE

$$dS_t = rS_t dt + \sigma S_t dW_t,$$

we obtain

$$S_T = S_0 \exp \left[ \left( r - \frac{1}{2} \sigma^2 \right) T + \sigma W_T \right],$$

and hence

$$S_T^2 = S_0^2 \exp \left[ (2r - \sigma^2) T + 2\sigma W_T \right].$$

The option can be valued as follow:

$$V = e^{-rT} \mathbb{E} [(S_T^2 - K)^+] = \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} [S_0^2 e^{(2r-\sigma^2)T+2\sigma\sqrt{T}x} - K] e^{-\frac{x^2}{2}} dx,$$

where  $x^*$  is given by

$$S_0^2 e^{(2r-\sigma^2)T+2\sigma\sqrt{T}x^*} - K > 0$$

$$x^* > \frac{\log\left(\frac{K}{S_0^2}\right) - (2r - \sigma^2)T}{2\sigma\sqrt{T}}.$$

Carrying on with the integration by completing the square for the first part, we obtain

$$\begin{aligned} V &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} S_0^2 e^{(2r-\sigma^2)T} e^{2\sigma^2 T} e^{-\frac{(x-2\sigma\sqrt{T})^2}{2}} dx - \frac{K e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\ &= S_0^2 e^{(r+\sigma^2)T} \Phi\left(\frac{\log\frac{S_0^2}{K} + (2r + 3\sigma^2)T}{2\sigma\sqrt{T}}\right) - K e^{-rT} \Phi\left(\frac{\log\frac{S_0^2}{K} + (2r - \sigma^2)T}{2\sigma\sqrt{T}}\right). \quad \triangleleft \end{aligned}$$