

## Mathematical Finance Solutions Sheet 4

### Solution 4-1

Writing  $X_t := (\mu - r)t + \sigma W_t + J_t$ , we see that

$$dS_t = S_{t-} dX_t.$$

Hence,  $S$  is the stochastic exponential given by

$$\begin{aligned} S_t &= S_0 \mathcal{E}(X)_t \\ &= S_0 \exp\left(X_t - \frac{1}{2}[X]_t\right) \prod_{0 < s \leq t} (1 + \Delta X_s) \exp\left(-\Delta X_s + \frac{1}{2}(\Delta X_s)^2\right) \end{aligned}$$

with  $\Delta X_s = X_s - X_{s-}$  and

$$[X]_t = \langle X^c, X^c \rangle_t + \sum_{0 < s \leq t} (\Delta X_s)^2,$$

where  $X^c$  denotes the continuous local martingale part of  $X$ . Therefore,

$$\begin{aligned} &\exp\left(X_t - \frac{1}{2}[X]_t\right) \prod_{0 < s \leq t} \exp\left(-\Delta X_s + \frac{1}{2}(\Delta X_s)^2\right) \\ &= \exp\left(\left(\mu - r - \frac{1}{2}\sigma^2\right)t + \sigma W_t\right). \end{aligned}$$

On the other hand,

$$\prod_{0 < s \leq t} (1 + \Delta X_s) = \prod_{i=1}^{N_t} Y_i = \exp\left(\sum_{i=1}^{N_t} \log(Y_i)\right).$$

Then  $\tilde{J}_t := \sum_{i=1}^{N_t} \log(Y_i)$  is a compound Poisson process so that

$$\tilde{X}_t = \left(\mu - r - \frac{1}{2}\sigma^2\right)t + \sigma W_t + \tilde{J}_t$$

is a Levy process.

### Solution 4-2

a)  $d\langle B, W \rangle_t = \rho dt$  because  $d\langle W, W' \rangle_t = 0$  and

$$\begin{aligned} \langle S, Y \rangle_t &= \left\langle \int_0^t \sigma(u, S_u, Y_u) dW_u, \int_0^t a(u, Y_u) dB_u \right\rangle_t \\ &= \int_0^t \sigma(u, S_u, Y_u) a(u, Y_u) d\langle W, B \rangle_u = \int_0^t \sigma(u, S_u, Y_u) a(u, Y_u) \rho du. \end{aligned}$$

- b) Following the lecture notes  $Q \sim P$  has a density process  $Z^Q$  which has a continuous version. Defining  $L^Q$  by

$$L^Q = \int \frac{1}{Z^Q} dZ^Q$$

we have  $Z^Q = Z_0^Q \mathcal{E}(L^Q)$ . By a general version of Kunita-Watanabe decomposition  $L^Q$  is in our case given by

$$L^Q = \int \gamma^Q \sigma dW + N^Q$$

with  $N^Q \in \mathcal{M}_{0,loc}(P)$  and  $\langle N^Q, \int \sigma dW \rangle = 0$ . By Bayes' rule  $Q$  is an ELMM for  $S$  iff  $Z^Q S \in \mathcal{M}_{loc}(P)$ . Thus by the product rule we obtain

$$\begin{aligned} d(Z_t^Q S_t) &= Z_t^Q \sigma_t dW_t + S_t dZ_t^Q + Z_t^Q \left( \mu_t dt + d \left\langle L^Q, \int \sigma dW \right\rangle_t \right) \\ &= Z_t^Q \sigma_t dW_t + S_t dZ_t^Q + Z_t^Q (\mu_t dt + \gamma^Q \sigma^2 dt), \end{aligned}$$

yielding  $Z^Q S \in \mathcal{M}_{loc}(P)$  if and only if  $\gamma^Q = -\frac{\mu_t}{\sigma_t^2}$ . Therefore the equivalent local martingale measures  $Q$  are parametrized via

$$\frac{Z^Q}{Z_0^Q} = \mathcal{E} \left( - \int \frac{\mu}{\sigma} dW + N^Q \right).$$

Since the filtration is generated by  $(W, W')$  we can apply the martingale representation theorem to write  $N^Q$  as

$$N^Q = \int \psi dW + \int \nu dW',$$

where  $\psi$  and  $\nu$  are some predictable processes. As  $\langle N^Q, \int \sigma dW \rangle = 0$ , it follows that  $\psi = 0$  such that we finally obtain

$$\frac{Z^Q}{Z_0^Q} = \mathcal{E} \left( - \int \lambda dW + \int \nu dW' \right)$$

where  $\lambda = \mu/\sigma$  and  $\nu$  is some predictable process.

- c) We want to change the measure, so we must assume that  $\nu$  is sufficiently nice, for instance that it satisfies Novikov's condition

$$E \left[ e^{\frac{1}{2} \int_0^T \nu_t^2 dt} \right] < \infty.$$

By Girsanov,  $(W^Q, W'^Q)$ , defined by  $W^Q = W + \int \lambda dt$  and  $W'^Q = W' - \int \nu dt$  is a 2-dimensional  $Q$ -Brownian motion. Plugging this into the SDE's

$$dS = \mu dt + \sigma(dW^Q - \lambda dt) = (\mu - \lambda\sigma)dt + \sigma dW^Q$$

and

$$\begin{aligned} dY &= bdt + a\rho(dW^Q - \lambda dt) + a\sqrt{1 - \rho^2}(dW'^Q + \nu dt) \\ &= (b + a(\sqrt{1 - \rho^2}\nu - \rho\lambda))dt + adB^Q \end{aligned}$$

for the  $Q$ -Brownian motion  $B^Q = \rho W_t^Q + \sqrt{1 - \rho^2} W_t'^Q$ .

**Solution 4-3**

a) To show convexity of  $\mathcal{C}(x)$ , let  $X^1, X^2 \in \mathcal{C}(x)$  and  $\lambda \in ]0, 1[$ . Then,

$$\begin{aligned} \sup_{Y \in \mathcal{D}} E[Y(\lambda X^1 + (1 - \lambda)X^2)] &\leq \lambda \sup_{Y \in \mathcal{D}} E[YX^1] + (1 - \lambda) \sup_{Y \in \mathcal{D}} E[YX^2] \\ &\leq \lambda x + (1 - \lambda)x = x, \end{aligned}$$

and so  $\lambda X^1 + (1 - \lambda)X^2 \in \mathcal{C}(x)$ . To show closedness in  $L_+^0(\mathcal{F}_T)$ , let  $(X^n)_{n \in \mathbb{N}}$  be a sequence in  $\mathcal{C}(x)$  converging in probability to some  $X \in L_+^0(\mathcal{F}_T)$ . Then a subsequence, called again  $(X^n)_{n \in \mathbb{N}}$ , converges  $P$ -a.s. to  $X$ . Hence, for each  $Y \in \mathcal{D}$ , by Fatou's lemma,

$$E[YX] \leq \liminf_{n \rightarrow \infty} E[YX^n] \leq x.$$

Since  $\mathcal{D}$  contains strictly positive random variables (e.g. the densities of equivalent  $\sigma$ -martingale measures), this implies that  $X \in L^0(\mathcal{F}_T)$ . Now, taking the supremum over  $\mathcal{D}$  gives  $X \in \mathcal{C}(x)$ .

b) It suffices to show that  $E[U(\cdot)]$  is continuous from below, i.e., that  $EU(X^n) \uparrow EU(X)$  as  $X^n \uparrow X$  in  $P$  in  $\mathcal{C}(x)$ . Without loss of generality, we may assume that  $U^-(X^n) \in L^1(P)$  for all  $n \in \mathbb{N}$ . By (a),  $\mathcal{C}(x)$  is convex and closed, and since  $P \ll\ll Q$  and

$$\lim_{n \rightarrow \infty} \sup_{X_T \in \mathcal{C}(x)} Q(|X_T| > n) \leq \lim_{n \rightarrow \infty} \sup_{H \in \mathcal{A}} Q(|(H \cdot S)_T| > n) = 0,$$

it is also bounded. So, we may apply Komlos' lemma (see e.g. W. Schahermayer, *Optimal Investment in Incomplete Financial Markets*, Lemma 3.3.): there exists a sequence  $(\tilde{X}^n)_{n \in \mathbb{N}}$  with  $\tilde{X}^n \in \text{conv}(X^n, X^{n+1}, \dots)$  such that  $\tilde{X}^n$  converges  $P$ -a.s., and a fortiori in probability, to some  $\mathcal{F}_T$ -measurable random variable  $\hat{X}$  taking values in  $[0, \infty]$ . By convexity of  $\mathcal{C}(x)$ ,  $\tilde{X}^n \in \mathcal{C}(x)$  for all  $n \in \mathbb{N}$ , and by closedness of  $\mathcal{C}(x)$  in  $L_+^0(\mathcal{F}_T)$ ,  $\hat{X} \in \mathcal{C}(x)$ .

Now, uniform integrability of  $U^+(\mathcal{C}(x))$  (which follows from the assumption  $u(x) < \infty$ ) gives

$$\lim_{n \rightarrow \infty} E[U^+(\tilde{X}^n)] = E[U^+(\hat{X})],$$

while Fatou's lemma gives

$$\liminf_{n \rightarrow \infty} E[U^-(\tilde{X}^n)] \geq E[U^-(\hat{X})].$$

Moreover, since  $E[U(\cdot)]$  is concave and increasing and  $(X^n)_{n \in \mathbb{N}}$  is increasing, we have

$$\begin{aligned} E[U(\tilde{X}^n)] &= E[U(\lambda_1^n X^n + \lambda_2^n X^{n+1} \dots)] \geq \lambda_1^n E[U(X^n)] + \lambda_2^n E[U(X^{n+1})] \dots \\ &\geq \inf_{m \geq n} E[U(X^m)] = E[U(X^n)], \quad n \in \mathbb{N}. \end{aligned}$$

Putting everything together gives

$$E[U(\hat{X})] \geq \liminf_{n \rightarrow \infty} E[U(\tilde{X}^n)] \geq \liminf_{n \rightarrow \infty} E[U(X^n)] = E[U(X)].$$

In particular, if we choose  $X^n \in \{X \in \mathcal{C}(x) : 1/n \leq u(x) - E[U(X)] \leq 1/(n - 1)\}$ ,  $n \in \mathbb{N}$ , we see that  $u(x)$  is attained for some  $\hat{X} \in \mathcal{C}(x)$ .

- c) Choose  $r > 1$  small enough that  $p := \frac{1}{r-b} > 1$ . We show that  $U^+(\mathcal{C}(x))$  is bounded in  $L^r(P)$  and hence uniformly integrable. To this end, by the assumption on  $U^+$ , it suffices to show that

$$\sup_{X \in \mathcal{C}(x)} E[(X^b)^r] = \sup_{X \in \mathcal{C}(x)} E[X^{1/p}] < \infty.$$

Let  $q > 1$  be the exponent conjugate to  $p$ . Then, by Hölder's inequality, by the hint of part (a), and by the fact that  $\left(\frac{dQ}{dP}\right)^{-1}$  has moments of all orders, for each  $X \in \mathcal{C}(x)$ ,

$$\begin{aligned} E[X^{1/p}] &= E \left[ \left( X \frac{dQ}{dP} \right)^{1/p} \left( \frac{dQ}{dP} \right)^{-1/p} \right] \leq \left( E \left[ X \frac{dQ}{dP} \right] \right)^{1/p} \left( E \left[ \left( \frac{dQ}{dP} \right)^{-q/p} \right] \right)^{1/q} \\ &\leq x^{1/p} \left( E \left[ \left( \frac{dQ}{dP} \right)^{-q/p} \right] \right)^{1/q} < \infty. \end{aligned}$$

Taking the supremum over  $\mathcal{C}(x)$  establishes the claim.

#### Solution 4-4

Denote by  $Z = (Z_t)_{t \in [0, T]}$  the density process of  $Q$  with respect to  $P$ .

- a) The second claim follows immediately from the first claim together with the fact that  $yZ_T = y \frac{dQ}{dP} \in \mathcal{D}(y)$  and the fact that the function  $V$  is decreasing. So it remains to show the first claim. Seeking a contradiction, suppose there exists  $z \in \mathcal{D}(y)$  such that  $A := \{z > yZ_T\}$  has  $P[A] > 0$ . Set  $a = Q[A] > 0$  and define the  $Q$ -martingale  $M = (M_t)_{t \in [0, T]}$  by  $M_t := E_Q[1_A | \mathcal{F}_t]$ . Then  $M$  is nonnegative and  $M_0 = a$  by the fact that  $F_0$  is  $P$ -trivial. Under  $Q$ , there exists admissible  $H$  such that  $M = a + H \cdot S$ . It follows that  $M_T \in \mathcal{C}(a)$ , i.e.,  $\frac{1}{a}M_T \in \mathcal{C}(1)$ . Now, on the one hand, by the definition of  $\mathcal{D}(y)$ ,

$$E\left[\frac{1}{a}M_T z\right] \leq y \text{ i.e. } E[M_T z] \leq ay.$$

On the other hand,

$$E[M_T Z_T] = E_Q[M_T] = M_0 = a.$$

Thus, we arrive at the contradiction

$$0 \geq E[M_T(z - yZ_T)] = E[1_{\{z > yZ_T\}}(z - yZ_T)] > 0.$$

- b) Note that  $0 \leq y_0 < \infty$  and  $v(y) < \infty$  on  $]y_0, \infty[$ . Moreover, recall that the function  $V$  is strictly decreasing, strictly convex and continuous on  $]0, \infty[$ .

First, define the function  $g : ]y_0, \infty[ \rightarrow ]-\infty, 0]$  by

$$g(s) = E[Z_T V'(sZ_T)].$$

This is well defined as  $Z_T \geq 0$   $P$ -a.s. and  $V' < 0$ . Moreover, it is increasing as  $V'$  is increasing. Thus if  $g(s_0) > -\infty$  for some  $s_0 > y_0$ , it follows by dominated convergence that it is continuous on  $[s_0, \infty)$ .

Next, for  $y_1, y_2 \in ]y_0, \infty[$ ,  $y_1 < y_2$ , the fundamental theorem of calculus gives

$$V(y_1 Z_T) - V(y_2 Z_T) = \int_{y_1}^{y_2} Z_T V'(sZ_T) ds. \quad (1)$$

Now, the left hand side of (1) is integrable by assumption. Thus, the right hand side is so, too, and since  $V' < 0$ , the integrand on the right hand side is strictly negative, and Fubini's theorem gives

$$v(y_2) - v(y_1) = \int_{y_1}^{y_2} g(s) ds.$$

In particular, the function  $g$  is finite a.e. on  $]y_0, \infty[$ , and thus continuous and finite on  $]y_0, \infty[$ . Now the claim follows from the fundamental theorem of calculus.

c) First, by the hint of part (a) of the previous exercise  $X \in \mathcal{C}(x)$  if and only if

$$\sup_{Y \in \mathcal{D}} E[XY] \leq x.$$

By part (a) of this exercise, this is equivalent to

$$\mathbb{E}[XZ_T] \leq x.$$

Now, by part (b) and the choice of  $y_x$ ,

$$\mathbb{E}[\widehat{X}Z_T] = E[-V'(y_x Z_T)Z_T] = -v'(y_x) = x, \text{ i.e. } U'(\widehat{X}) = y_x Z_T. \quad (2)$$

We see that  $\widehat{X} \in \mathcal{C}(x)$ .

Next, fix  $X \in \mathcal{C}(x)$ . We may assume without loss of generality that  $E[U(X)] > -\infty$ . By the fact that  $\widehat{X} > 0$   $P$ -a.s. and  $U$  is in  $C^1(]0, \infty[)$  and strictly concave on  $]0, \infty[$ ,

$$U(X) - U(\widehat{X}) \leq U'(\widehat{X})(X - \widehat{X}),$$

where the equality is strict on  $\{X \neq \widehat{X}\}$ . Taking expectations, (2) yields

$$\mathbb{E}[U(X) - U(\widehat{X})] \leq \mathbb{E}[U'(\widehat{X})(X - \widehat{X})] = y_x \mathbb{E}[Z_T(X - \widehat{X})] \leq 0.$$

If  $X = \widehat{X}$   $P$ -a.s., then both inequalities are trivially equalities, and if  $P[X \neq \widehat{X}] > 0$ , then the first inequality is strict.

#### Solution 4-5

The discounted stock price process  $S^1$  satisfies the SDE

$$dS_t^1 = S_t^1((\mu - r)dt + \sigma dW_t) = S_t^1 \sigma(\lambda dt + dW_t),$$

where  $\lambda := \frac{\mu - r}{\sigma}$  denotes the market price of risk. By the previous round exercise 3-1, there exists a unique equivalent martingale measure  $Q \sim P$  on  $\mathcal{F}_T$  given by

$$\frac{dQ}{dP} = \mathcal{E}(-\lambda W)_T.$$

Moreover, elementary analysis gives  $V(y) = \frac{1-\gamma}{\gamma} y^{-\frac{\gamma}{1-\gamma}}$  and  $V'(y) = -y^{-\frac{1}{1-\gamma}}$ .

a) Fix  $y > 0$ . Then by Exercise 4-4 (a) and the fact that  $\mathcal{E}(aW)$  is a  $P$ -martingale for all  $a \in \mathbb{R}$ ,

$$\begin{aligned} v(y) &= E \left[ \frac{1-\gamma}{\gamma} y^{-\frac{\gamma}{1-\gamma}} (\mathcal{E}(-\lambda W)_T)^{-\frac{\gamma}{1-\gamma}} \right] \\ &= \frac{1-\gamma}{\gamma} y^{-\frac{\gamma}{1-\gamma}} E \left[ \exp \left( \frac{\lambda \gamma}{1-\gamma} W_T + \frac{1}{2} \frac{\lambda^2 \gamma}{1-\gamma} T \right) \right] \\ &= \frac{1-\gamma}{\gamma} y^{-\frac{\gamma}{1-\gamma}} \exp \left( \frac{1}{2} \frac{\lambda^2 \gamma}{(1-\gamma)^2} T \right) E \left[ \mathcal{E} \left( \frac{\lambda \gamma}{1-\gamma} W \right)_T \right] \\ &= \frac{1-\gamma}{\gamma} y^{-\frac{\gamma}{1-\gamma}} \exp \left( \frac{1}{2} \frac{\lambda^2 \gamma}{(1-\gamma)^2} T \right) < \infty. \end{aligned}$$

b) First, note that  $v(y) < \infty$  for some  $y \in ]0, \infty[$  implies that

$$u(x) \leq v(y) + vx < \infty, \quad x \in ]0, \infty[.$$

Next, fix  $x > 0$ . Then by Exercise 4-4 (b) and part (a),

$$\begin{aligned} \widehat{X}_x &= -V' \left( y_x \frac{dQ}{dP} \right) = y_x^{-\frac{1}{1-\gamma}} (\mathcal{E}(-\lambda W)_T)^{-\frac{1}{1-\gamma}} \\ &= -v'(y_x) \exp \left( -\frac{1}{2} \frac{\lambda^2 \gamma}{(1-\gamma)^2} T \right) \exp \left( \frac{\lambda}{1-\gamma} W_T + \frac{1}{2} \frac{\lambda^2}{1-\gamma} T \right) \\ &= x \exp \left( \frac{\lambda}{1-\gamma} (W_T + \lambda T) - \frac{1}{2} \frac{\lambda^2}{(1-\gamma)^2} T \right) \\ &= x \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_T. \end{aligned}$$

c) Fix  $x > 0$ . By the definition of the stochastic exponential,

$$\begin{aligned} \widehat{X} &= x \left( 1 + \int_0^T \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_t \frac{\lambda}{1-\gamma} dR_t \right) \\ &= x + \int_0^T x \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_t \frac{\lambda}{1-\gamma} \frac{1}{\sigma S_t} dS_t. \end{aligned}$$

This gives the first claim. Using again that  $\mathcal{E}(aW)$  is a  $P$ -martingale for all  $a \in \mathbb{R}$  gives

$$\begin{aligned} u(x) &= \mathbb{E} \left[ U(\widehat{X}_x) \right] = \frac{x^\gamma}{\gamma} \mathbb{E} \left[ \left( \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_T \right)^\gamma \right] \\ &= \frac{x^\gamma}{\gamma} \mathbb{E} \left[ \exp \left( \frac{\lambda \gamma}{1-\gamma} (W_T + \lambda T) - \frac{1}{2} \frac{\lambda^2 \gamma}{(1-\gamma)^2} T \right) \right] \\ &= \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\lambda^2 \gamma}{1-\gamma} T \right) \mathbb{E} \left[ \mathcal{E} \left( \frac{\lambda \gamma}{1-\gamma} W \right)_T \right] \\ &= \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\lambda^2 \gamma}{1-\gamma} T \right). \end{aligned}$$

This establishes the second claim.