

# Mathematical Finance

## Exercise Sheet 12

Submit by 12:00 on Wednesday, December 18 via the course homepage.

**Exercise 12.1** (*Some properties of  $u$* ) Let  $U : (0, \infty) \rightarrow \mathbb{R}$  be a concave and increasing function. Define the function  $u : (0, \infty) \rightarrow (-\infty, +\infty]$  by

$$u(x) := \sup_{V \in \mathcal{V}(x)} E[U(V_T)],$$

where  $\mathcal{V}(x) := \{x + G(\vartheta) : \vartheta \in \Theta_{\text{adm}}^x\}$ .

- (a) Show that  $u$  is concave and increasing.
- (b) If additionally  $u(x_0) < \infty$  for some  $x_0 > 0$ , show that  $u(x) < \infty$  for all  $x > 0$ .

### Solution 12.1

- (a) We first prove that  $u$  is concave. Let  $x, y \in (0, \infty)$  and  $\lambda \in (0, 1)$  be fixed. We need to show that

$$u(\lambda x + (1 - \lambda)y) \geq \lambda u(x) + (1 - \lambda)u(y).$$

First note that if either  $u(x)$  or  $u(y)$  is  $-\infty$ , then the inequality holds trivially. So assume that  $u(x), u(y) > -\infty$ . Take  $x + G(\vartheta^x) \in \mathcal{V}(x)$  and  $y + G(\vartheta^y) \in \mathcal{V}(y)$  with  $U(x + G(\vartheta^x))^-$  and  $U(y + G(\vartheta^y))^-$  both in  $L^1$ . Then

$$\lambda(x + G(\vartheta^x)) + (1 - \lambda)(y + G(\vartheta^y)) = \lambda x + (1 - \lambda)y + G(\lambda\vartheta^x + (1 - \lambda)\vartheta^y).$$

As  $U$  is concave, we have

$$\lambda U(x + G(\vartheta^x)) + (1 - \lambda)U(y + G(\vartheta^y)) \leq U(\lambda x + (1 - \lambda)y + G(\lambda\vartheta^x + (1 - \lambda)\vartheta^y)).$$

So also  $U(\lambda x + (1 - \lambda)y + G(\lambda\vartheta^x + (1 - \lambda)\vartheta^y))^- \in L^1$ . Furthermore, since  $\lambda\vartheta^x + (1 - \lambda)\vartheta^y \in \Theta_{\text{adm}}^{\lambda x + (1 - \lambda)y}$ , we can take expectations, which yields

$$\lambda E[U(x + G(\vartheta^x))] + (1 - \lambda)E[U(y + G(\vartheta^y))] \leq u(\lambda x + (1 - \lambda)y).$$

Finally, taking the supremum over all  $x + G(\vartheta^x) \in \mathcal{V}(x)$  and  $y + G(\vartheta^y) \in \mathcal{V}(y)$  with integrable negative parts gives the required inequality.

It remains to prove that  $u$  is increasing. This follows from the fact that  $\Theta_{\text{adm}}^x \subseteq \Theta_{\text{adm}}^y$  for  $0 < x < y$ . Indeed, for  $x + G(\vartheta^x) \in \mathcal{V}(x)$  so that  $\vartheta^x \in \Theta_{\text{adm}}^x$ , we have  $y + G(\vartheta^x) \in \mathcal{V}(y)$ , and as  $U$  is increasing, this implies

$$E\left[U\left(x + G(\vartheta^x)\right)\right] \leq E\left[U\left(y + G(\vartheta^x)\right)\right] \leq u(y).$$

Taking the supremum over all  $\vartheta^x \in \Theta_{\text{adm}}^x$  gives  $u(x) \leq u(y)$ , completing the proof.

- (b) As  $u$  is increasing, we know that  $u(x) < \infty$  for all  $x < x_0$ . It thus remains to show that  $u(x) < \infty$  for all  $x > x_0$ . By choosing  $\lambda \in (0, 1)$  small enough, we can find  $y \in (0, x_0)$  such that

$$x_0 = \lambda x + (1 - \lambda)y.$$

By concavity of  $u$ , we have

$$\lambda u(x) + (1 - \lambda)u(y) \leq u(x_0) < \infty,$$

which gives the result because  $u(y) \leq u(x_0) < \infty$  and  $u(y) \geq U(y) > -\infty$ .

**Exercise 12.2** (*Utility in a market with arbitrage*) Consider a general market with finite time horizon  $T$ . Let  $U : (0, \infty) \rightarrow \mathbb{R}$  be an increasing and concave utility function. Suppose that  $U$  is unbounded from above and that either the market admits a 0-admissible arbitrage opportunity, or we are in finite discrete time and the market admits an (admissible) arbitrage opportunity. Show that in both cases, we have  $u \equiv \infty$ .

Without imposing that  $U$  is unbounded from above, what can you say about the relationship between  $u(x)$  and  $U(x)$  as  $x \rightarrow \infty$ ?

**Solution 12.2** By assumption, there exists  $\vartheta \in \Theta_{\text{adm}}$  such that  $G_T(\vartheta) \geq 0$   $P$ -a.s. and  $P[G_T(\vartheta) > 0] > 0$ . By Exercise 4.2, we may assume that  $\vartheta$  is 0-admissible, and so also  $n\vartheta$  is 0-admissible for each  $n \in \mathbb{N}$ . It follows that  $x + nG_T(\vartheta) \in \mathcal{V}(x)$  for every  $x > 0$  and  $n \in \mathbb{N}$ . So setting  $A := \{G_T(\vartheta) > 0\}$ , we have that for all  $x > 0$  and  $n \in \mathbb{N}$ ,

$$u(x) \geq E\left[U\left(x + nG_T(\vartheta)\right)\right] = E\left[U\left(x + nG_T(\vartheta)\right)\mathbf{1}_A\right] + E\left[U(x)\mathbf{1}_{A^c}\right].$$

As  $U$  is increasing, we can let  $n \rightarrow \infty$  and apply the monotone convergence theorem to get that for all  $x > 0$ ,

$$u(x) \geq E\left[U(\infty)\mathbf{1}_A\right] + E\left[U(x)\mathbf{1}_{A^c}\right].$$

Note that  $U$  is increasing gives that the limit  $U(\infty) := \lim_{x \rightarrow \infty} U(x) \in \mathbb{R} \cup \{\infty\}$  exists. Since  $U$  is unbounded from above we have  $U(\infty) = \infty$ , and as  $P[A] > 0$ , we can conclude that  $u \equiv \infty$ , as required.

Now suppose that  $U$  is not necessarily unbounded from above. We still have

$$u(x) \geq E[U(\infty)\mathbf{1}_A] + E[U(x)\mathbf{1}_{A^c}] = U(\infty)P[A] + U(x)P[A^c].$$

Also, by the definition of  $u$ ,  $u(x) \leq U(\infty)$  as  $U$  is increasing. So for each  $x > 0$ ,

$$U(\infty)P[A] + U(x)P[A^c] \leq u(x) \leq U(\infty).$$

Letting  $x \rightarrow \infty$  in the above gives  $u(\infty) = U(\infty)$ . This completes the problem.

**Exercise 12.3** (*Utility in a complete market*) Consider a financial market modelled by an  $\mathbb{R}^d$ -valued semimartingale  $S$  satisfying NFLVR. Let  $U : (0, \infty) \rightarrow \mathbb{R}$  be a utility function such that  $u(x) < \infty$  for some (and hence for all)  $x \in (0, \infty)$ . Assume that the market is complete in the sense that there exists a unique E $\sigma$ MM  $Q$  on  $\mathcal{F}_T$ . Assume furthermore that  $\mathcal{F}_0$  is trivial.

- (a) Show that  $h \leq z \frac{dQ}{dP}$   $P$ -a.s. for all  $h \in \mathcal{D}(z)$ , and deduce that

$$j(z) = E\left[J\left(z \frac{dQ}{dP}\right)\right].$$

- (b) Let  $z_0 := \inf\{z > 0 : j(z) < \infty\}$ . Show that the function  $j$  defined in the lecture notes is in  $C^1((z_0, \infty); \mathbb{R})$  and satisfies

$$j'(z) = E\left[\frac{dQ}{dP} J'\left(z \frac{dQ}{dP}\right)\right], \quad z \in (z_0, \infty).$$

- (c) Set  $x_0 := \lim_{z \downarrow z_0} (-j'(z))$  and fix  $x \in (0, x_0)$ . Let  $z_x \in (z_0, \infty)$  be the unique number such that  $-j'(z_x) = x$ . Show that  $f^* := I\left(z_x \frac{dQ}{dP}\right)$  is the unique solution to the primal problem

$$u(x) = \sup_{f \in \mathcal{C}(x)} E[U(f)].$$

### Solution 12.3

For notational convenience, we denote by  $Z^Q = (Z_t^Q)_{0 \leq t \leq T}$  the density process of  $Q$  with respect to  $P$ , so that  $Z_T^Q = \frac{dQ}{dP}$ .

- (a) Recall that in general, a payoff  $H \in L_+^0(\mathcal{F}_T)$  is attainable if and only if the supremum

$$\sup_{Q^0 \in \mathbb{P}_{e,\sigma}} E_{Q^0}[H]$$

is finite and attained at some  $Q^* \in \mathbb{P}_{e,\sigma}$ . In our setting,  $\mathbb{P}_{e,\sigma}$  is the singleton set  $\{Q\}$ , so that a payoff  $H \in L_+^0(\mathcal{F}_T)$  is attainable if and only if  $E_Q[H] < \infty$ , i.e. if and only if  $H \in L_+^1(Q, \mathcal{F}_T)$ .

Now we recall that

$$\mathcal{D}(z) := \{h \in L_+^0(\mathcal{F}_T) : \exists Z \in \mathcal{Z}(z) \text{ with } h \leq Z_T\}.$$

So take  $h \in \mathcal{D}(z)$  and suppose for contradiction that we do not have  $h \leq zZ_T^Q$   $P$ -a.s. Then setting  $A := \{h > zZ_T^Q\}$ , we have  $P[A] > 0$ . Now define the process  $M = (M_t)_{0 \leq t \leq T}$  by

$$M_t := E_Q[\mathbf{1}_A \mid \mathcal{F}_t].$$

Then  $M$  is a nonnegative  $Q$ -martingale with  $M_0 = Q[A] > 0$  because  $Q \approx P$ . Since  $E_Q[M_T] \leq 1 < \infty$ , it follows that  $M_T \in L_+^0(\mathcal{F}_T)$  is attainable so that there exists some  $\vartheta \in \Theta_{\text{adm}}$  with

$$M = M_0 + G(\vartheta).$$

Since  $M$  is nonnegative, we must have  $\vartheta \in \Theta_{\text{adm}}^{M_0}$  and hence  $M \in \mathcal{V}(M_0)$ .

Now, since  $h \in \mathcal{D}(z)$ , there exists  $Z \in \mathcal{Z}(z)$  such that  $h \leq Z_T$ . By the definition of  $\mathcal{Z}(z)$ , the product  $ZM$  is a  $P$ -supermartingale. We thus have

$$E[hM_T] \leq E[Z_T M_T] \leq E[Z_0 M_0] = zM_0.$$

Also, we have  $E[zZ_T^Q M_T] = E_Q[zM_T] = zM_0$ , and thus

$$E\left[(h - zZ_T^Q) M_T\right] \leq 0.$$

But recalling  $M_T = \mathbf{1}_A$  and  $P[A] > 0$  gives

$$E\left[(h - zZ_T^Q) M_T\right] > 0,$$

which gives a contradiction. Hence we must have  $h \leq zZ_T^Q$   $P$ -a.s., as required. In particular, as any  $Z_T \in \mathcal{D}(z)$  for  $Z \in \mathcal{Z}(z)$ , this gives  $Z_T \leq zZ_T^Q$  for any  $Z \in \mathcal{Z}(z)$ .

It remains to show  $j(z) = E[J(zZ_T^Q)]$ . First we recall that

$$j(z) := \inf_{Z \in \mathcal{Z}(z)} E[J(Z_T)].$$

For each  $Z \in \mathcal{Z}(z)$  we have  $Z_T \leq zZ_T^Q$ . As  $J$  is decreasing, we have

$$J(Z_T) \geq J(zZ_T^Q),$$

and thus

$$E[J(Z_T)] \geq E\left[J(zZ_T^Q)\right].$$

Taking the infimum over all  $Z \in \mathcal{Z}(z)$  gives

$$j(z) \geq E\left[J(zZ_T^Q)\right].$$

As  $zZ_T^Q \in \mathcal{Z}(z)$ , this concludes the proof.

- (b) Note that  $0 \leq z_0 < \infty$  by Theorem 12.4, and also by Theorem 12.4, we have that  $j(z) < \infty$  for  $z \in (z_0, \infty)$ .

Now recall that  $J$  is in  $C^1$  and strictly decreasing. We can thus define the function  $g : (z_0, \infty) \rightarrow [-\infty, 0]$  by

$$g(s) := E[Z_T^Q J'(sZ_T^Q)].$$

Moreover, as  $J$  is also strictly convex,  $J'$  is increasing, and thus  $g$  is also increasing since  $Z_T^Q > 0$ . As  $g$  is negative-valued, it follows from the dominated convergence theorem that if  $g(s_0) > -\infty$  for some  $s_0 > z_0$ , we have that  $g$  is continuous on  $(s_0, \infty)$ .

Next, since  $\frac{d}{ds} J(sZ_T^Q) = Z_T^Q J'(sZ_T^Q)$  by the chain rule, we have by the fundamental theorem of calculus that for  $z_0 < z_1 < z_2 < \infty$ ,

$$J(z_2 Z_T^Q) - J(z_1 Z_T^Q) = \int_{z_1}^{z_2} Z_T^Q J'(sZ_T^Q) ds.$$

By part (a), we know that  $j(z) = E[J(zZ_T^Q)]$ . Thus taking expectations of both sides in the above gives

$$j(z_2) - j(z_1) = E \left[ \int_{z_1}^{z_2} Z_T^Q J'(sZ_T^Q) ds \right] = \int_{z_1}^{z_2} E[Z_T^Q J'(sZ_T^Q)] ds = \int_{z_1}^{z_2} g(s) ds,$$

where the second step uses the Fubini–Tonelli theorem, keeping in mind that the integrand is strictly negative.

Note that by the definition of  $z_0$ , we have that  $j(z_2) - j(z_1)$  is finite, and thus the function  $g$  is finite a.e. on  $(z_0, \infty)$ . From the above, we can conclude that  $g$  is continuous and finite on  $(z_0, \infty)$ . By dividing by  $z_2 - z_1$  and letting  $z_2 \rightarrow z_1$ , we get that

$$j'(z) = E[Z_T^Q J'(zZ_T^Q)] = g(z)$$

as required. Now since  $g$  is continuous on  $(z_0, \infty)$ , we have  $j \in C^1((z_0, \infty); \mathbb{R})$ , completing the proof.

- (c) Before establishing that  $f^*$  is a solution to the primal problem, we first need to check that  $f^* \in \mathcal{C}(x)$ . To this end, recall that  $f \in \mathcal{C}(x)$  if and only if

$$\sup_{h \in \mathcal{D}(1)} E[fh] \leq x.$$

By part (a), this is equivalent to

$$E[fZ_T^Q] \leq x.$$

Now by the definition of  $f^*$  and  $I$ , we have

$$E[f^* Z_T^Q] = E[I(z_x Z_T^Q) Z_T^Q] = E[-J'(z_x Z_T^Q) Z_T^Q].$$

Moreover, by part (b), we have  $E[Z_T^Q J'(z_x Z_T^Q)] = j'(z_x)$ , and since  $-j'(z_x) = x$  by definition of  $z_x$ , we have

$$E[f^* Z_T^Q] = x$$

and thus in particular  $f^* \in \mathcal{C}(x)$ , as required.

Next, we establish that  $f^*$  is a solution to the primal problem. So fix  $f \in \mathcal{C}(x)$ . We need to show that  $E[U(f^*)] \geq E[U(f)]$ . We may thus assume without loss of generality that  $E[U(f)] > -\infty$ . Now since  $U$  is in  $C^1$  and strictly concave on  $(0, \infty)$ , and since  $f^* > 0$   $P$ -a.s., we have

$$U(f) - U(f^*) \leq U'(f^*)(f - f^*),$$

with strict inequality on the event  $\{f \neq f^*\}$ . Now note that

$$U'(f^*) = U'(I(z_x Z_T^Q)) = z_x Z_T^Q.$$

Thus taking expectations of the above inequality yields

$$E[U(f) - U(f^*)] \leq E[z_x Z_T^Q (f - f^*)],$$

and since  $E[Z_T^Q f^*] = x$  and  $E[Z_T^Q f] \leq x$  and  $z_x > 0$ , we have

$$E[U(f) - U(f^*)] \leq 0,$$

and the inequality is strict when  $P[f \neq f^*] > 0$ . It follows immediately that  $f^*$  is the unique solution to the primal problem. This completes the proof.

**Exercise 12.4** (*The Merton problem*) Consider the Black–Scholes market given by

$$\begin{aligned} d\tilde{S}_t^0 &= r\tilde{S}_t^0 dt, & \tilde{S}_0^0 &= 1, \\ d\tilde{S}_t^1 &= \tilde{S}_t^1(\mu dt + \sigma dW_t), & \tilde{S}_0^1 &= s > 0. \end{aligned}$$

Let  $U : (0, \infty) \rightarrow \mathbb{R}$  be defined by  $U(x) = \frac{1}{\gamma}x^\gamma$ , where  $\gamma \in (-\infty, 1) \setminus \{0\}$ . We consider the *Merton problem* of maximising expected utility from final wealth (in units of  $\tilde{S}_0^0$ ).

(a) Show that for  $z > 0$ ,

$$j(z) = \frac{1 - \gamma}{\gamma} z^{-\frac{\gamma}{1-\gamma}} \exp\left(\frac{1}{2} \frac{\gamma}{(1-\gamma)^2} \frac{(\mu - r)^2 T}{\sigma^2}\right).$$

(b) Show that the unique solution to the primal problem

$$u(x) = \sup_{f \in \mathcal{C}(x)} E[U(f)], \quad x \in (0, \infty),$$

is given by  $f_x^* := x\mathcal{E}(\frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R)_T$ , where the process  $R = (R_t)_{0 \leq t \leq T}$  is defined by  $R_t = W_t + \frac{\mu-r}{\sigma}t$ .

(c) Deduce that  $f_x^* = V_T(x, \vartheta^x)$ , where the integrand  $\vartheta^x = (\vartheta_t^x)_{0 \leq t \leq T}$  is given by

$$\vartheta_t^x = \frac{x}{S_t^1} \frac{1}{1-\gamma} \frac{\mu-r}{\sigma^2} \mathcal{E} \left( \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R \right)_t, \quad x \in (0, \infty),$$

and show that

$$u(x) = \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\gamma}{1-\gamma} \frac{(\mu-r)^2}{\sigma^2} T \right), \quad x \in (0, \infty).$$

(d) For any  $x$ -admissible  $\vartheta$  with  $V(x, \vartheta) > 0$ , denote by

$$\pi_t := \frac{\vartheta_t S_t^1}{V_t(x, \vartheta)}$$

the fraction of wealth that is invested in the stock. Show that the optimal strategy  $\vartheta^x$  is given by the *Merton proportion*

$$\pi_t^* = \frac{1}{1-\gamma} \frac{\mu-r}{\sigma^2}.$$

### Solution 12.4

(a) In the Black–Scholes model, there exists a unique EMM  $Q$ , and thus Exercise 12.3(a) is applicable. We hence have

$$j(z) = E \left[ J \left( z \frac{dQ}{dP} \right) \right].$$

To compute this, we start by writing

$$J(y) = \sup_{x>0} (U(x) - xy) = \sup_{x>0} \left( \frac{1}{\gamma} x^\gamma - xy \right).$$

Taking the derivative of  $\frac{1}{\gamma} x^\gamma - xy$  with respect to  $x$  and setting it equal to zero, we get  $x = y^{\frac{1}{\gamma-1}}$ , and hence

$$J(y) = \frac{1}{\gamma} y^{\frac{\gamma}{\gamma-1}} - y^{\frac{\gamma}{\gamma-1}} = \frac{1-\gamma}{\gamma} y^{\frac{\gamma}{\gamma-1}}.$$

We also recall that in the Black–Scholes model,

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

where  $\lambda := \frac{\mu-r}{\sigma}$ . So we have

$$\begin{aligned}
 j(z) &= E \left[ J \left( z \frac{dQ}{dP} \right) \right] \\
 &= \frac{1-\gamma}{\gamma} z^{\frac{\gamma}{\gamma-1}} E \left[ \mathcal{E}(-\lambda W)_T^{\frac{\gamma}{\gamma-1}} \right] \\
 &= \frac{1-\gamma}{\gamma} z^{\frac{\gamma}{\gamma-1}} E \left[ \exp \left( \frac{\lambda\gamma}{1-\gamma} W_T - \frac{1}{2} \frac{\lambda^2\gamma}{\gamma-1} T \right) \right] \\
 &= \frac{1-\gamma}{\gamma} z^{\frac{\gamma}{\gamma-1}} \exp \left( -\frac{1}{2} \frac{\lambda^2\gamma}{\gamma-1} T \right) E \left[ \mathcal{E} \left( \frac{\lambda\gamma}{1-\gamma} W \right)_T \right] \\
 &= \frac{1-\gamma}{\gamma} z^{\frac{\gamma}{\gamma-1}} \exp \left( \frac{1}{2} \frac{\lambda^2\gamma}{(1-\gamma)^2} T \right),
 \end{aligned}$$

where in the last step we use that  $\mathcal{E}(aW)$  is a  $P$ -martingale for each  $a \in \mathbb{R}$ . Substituting  $\lambda = \frac{\mu-r}{\sigma}$  then gives the result.

(b) First, note that  $j(z) < \infty$  for some  $z \in (0, \infty)$  implies that

$$u(x) \leq j(z) + zx < \infty, \quad x \in (0, \infty).$$

We computed  $J(z) = \frac{1-\gamma}{\gamma} z^{-\frac{\gamma}{1-\gamma}}$ , and hence  $J'(z) = -z^{-\frac{1}{1-\gamma}}$ . Now fix  $x > 0$ . With the same notation as in Exercise 12.3, we have

$$\begin{aligned}
 f_x^* &= -J' \left( z_x \frac{dQ}{dP} \right) = z_x^{-\frac{1}{1-\gamma}} (\mathcal{E}(-\lambda W)_T)^{-\frac{1}{1-\gamma}} \\
 &= -j'(z_x) \exp \left( -\frac{1}{2} \frac{\lambda^2\gamma}{(1-\gamma)^2} T \right) \exp \left( \frac{\lambda}{1-\gamma} W_T + \frac{1}{2} \frac{\lambda^2}{1-\gamma} T \right) \\
 &= x \exp \left( \frac{\lambda}{1-\gamma} (W_T + \lambda T) - \frac{1}{2} \frac{\lambda^2}{(1-\gamma)^2} T \right) \\
 &= x \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_T.
 \end{aligned}$$

This completes the proof.

(c) Fix  $x > 0$ . By the definition of the stochastic exponential and using that  $\lambda = \frac{\mu-r}{\sigma}$ , we have

$$\begin{aligned}
 f_x^* &= x \left( 1 + \int_0^T \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_t \frac{\lambda}{1-\gamma} dR_t \right) \\
 &= x + \int_0^T x \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_t \frac{\lambda}{1-\gamma} \frac{1}{\sigma S_t^1} dS_t^1 \\
 &= x + \int_0^T x \mathcal{E} \left( \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R \right)_t \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} \frac{1}{\sigma S_t^1} dS_t^1 \\
 &= x + \int_0^T \frac{x}{S_t^1} \frac{1}{1-\gamma} \frac{\mu-r}{\sigma^2} \mathcal{E} \left( \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R \right)_t dS_t^1.
 \end{aligned}$$



This gives the first claim. Now using again that  $\mathcal{E}(aW)$  is a  $P$ -martingale for all  $a \in \mathbb{R}$  and that  $\lambda = \frac{\mu-r}{\sigma}$ , we have

$$\begin{aligned} u(x) &= E[U(f_x^*)] = \frac{x^\gamma}{\gamma} E \left[ \left( \mathcal{E} \left( \frac{\lambda}{1-\gamma} R \right)_T \right)^\gamma \right] \\ &= \frac{x^\gamma}{\gamma} E \left[ \exp \left( \frac{\lambda\gamma}{1-\gamma} (W_T + \lambda T) - \frac{1}{2} \frac{\lambda^2\gamma}{(1-\gamma)^2} T \right) \right] \\ &= \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\lambda^2\gamma}{1-\gamma} T \right) E \left[ \mathcal{E} \left( \frac{\lambda\gamma}{1-\gamma} W \right)_T \right] \\ &= \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\lambda^2\gamma}{1-\gamma} T \right) \\ &= \frac{x^\gamma}{\gamma} \exp \left( \frac{1}{2} \frac{\gamma}{1-\gamma} \frac{(\mu-r)^2}{\sigma^2} T \right). \end{aligned}$$

This completes the proof.

(d) By part (b) and since  $\lambda = \frac{\mu-r}{\sigma}$ , we have

$$V_t(x, \vartheta^x) = x \mathcal{E} \left( \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R \right)_t,$$

and by part (c), we have

$$\vartheta_t^x = \frac{x}{S_t^1} \frac{1}{1-\gamma} \frac{\mu-r}{\sigma^2} \mathcal{E} \left( \frac{1}{1-\gamma} \frac{\mu-r}{\sigma} R \right)_t.$$

Therefore, we obtain directly that

$$\pi_t^* := \frac{\vartheta_t^x S_t^1}{V_t(x, \vartheta^x)} = \frac{1}{1-\gamma} \frac{\mu-r}{\sigma^2}.$$

This completes the proof.

**Exercise 12.5** ( $\frac{d\hat{P}}{dP}$  has moments of all orders) Let  $S$  be a continuous real-valued semimartingale satisfying the structure condition (SC), i.e. there exist a continuous local martingale  $M$  null at zero and a predictable process  $\lambda$  such that

$$S = S_0 + M + \int \lambda d\langle M \rangle,$$

and with the mean-variance tradeoff process  $K = \int \lambda^2 d\langle M \rangle$  bounded. Now define  $\hat{Z} := \mathcal{E}(-\lambda \bullet M)$  and  $\frac{d\hat{P}}{dP} := \hat{Z}_T$ .

(a) Show that  $\hat{P} \in \mathbb{P}_{e,\text{loc}}(S)$ .

- (b) Show that both  $\frac{d\hat{P}}{dP}$  and  $\frac{dP}{d\hat{P}}$  have moments of all orders.

### Solution 12.5

- (a) We need to show that  $\hat{P}$  is an equivalent probability measure, and that  $S$  is a  $\hat{P}$ -local martingale. To this end, first note that since  $K$  is bounded, we have that

$$E \left[ \exp \left( \frac{1}{2} \langle -\lambda \bullet M \rangle_T \right) \right] = E \left[ \exp \left( \frac{1}{2} K_T \right) \right] < \infty.$$

So by Novikov's condition, we can conclude that  $\hat{Z}$  is a martingale. As  $\hat{Z}$  is strictly positive, it follows that  $\hat{P}$  is an equivalent probability measure. It now remains to show that  $S$  is a  $\hat{P}$ -local martingale. To this end, we first apply the stochastic product rule to  $\hat{Z}S$  and write

$$d(\hat{Z}S) = \hat{Z} dS + S d\hat{Z} + d\langle \hat{Z}, S \rangle.$$

Then we use that  $S$  satisfies (SC) and that

$$d\hat{Z} = d\mathcal{E}(-\lambda \bullet M) = \mathcal{E}(-\lambda \bullet M) d(-\lambda \bullet M) = -\lambda \mathcal{E}(-\lambda \bullet M) dM = -\lambda \hat{Z} dM$$

to compute

$$\begin{aligned} d(\hat{Z}S) &= \hat{Z} dM + \hat{Z} \lambda d\langle M \rangle - S \lambda \hat{Z} dM - \lambda \hat{Z} d\langle M \rangle \\ &= (\hat{Z} - S \lambda \hat{Z}) dM. \end{aligned}$$

As  $\hat{Z}$ ,  $S$  and  $M$  are continuous, it follows that  $\hat{Z}S$  is a  $P$ -local martingale, so that  $S$  is a  $\hat{P}$ -local martingale, and hence  $\hat{P} \in \mathbb{P}_{e,loc}$ , as required.

- (b) We compute, for any  $p \in \mathbb{R}$ ,

$$\begin{aligned} \hat{Z}_T^p &= \exp \left( -p\lambda \bullet M_T - \frac{1}{2} p^2 \lambda^2 \bullet \langle M \rangle_T \right) \\ &= \exp \left( -p\lambda \bullet M_T - \frac{1}{2} p^2 \lambda^2 \bullet \langle M \rangle_T \right) \exp \left( \frac{1}{2} (p^2 - p) \lambda^2 \bullet \langle M \rangle_T \right) \\ &= \mathcal{E}(-p\lambda \bullet M)_T \exp \left( (p^2 - p) K_T \right). \end{aligned}$$

So letting  $C < \infty$  be a bound on  $K$ , we can write

$$E[\hat{Z}_T^p] \leq E[\mathcal{E}(-p\lambda \bullet M)_T] \exp(C|p^2 - p|) \leq \exp(C|p^2 - p|) < \infty,$$

since  $\mathcal{E}(-p\lambda \bullet M)$  is a supermartingale. As  $Z_T = \frac{d\hat{P}}{dP}$  and  $Z_T^{-1} = \frac{dP}{d\hat{P}}$ , this completes the proof.