

# Mathematical Finance

## Exercise Sheet 11

*Submit by 12:00 on Wednesday, December 11 via the course homepage.*

**Exercise 11.1** (*Equivalence of (NA)*) Show that  $S$  satisfies (NA) if and only if 0 is maximal in  $\mathcal{G}_{\text{adm}}$ .

**Solution 11.1** We have that  $S$  satisfies (NA) if and only if  $\mathcal{G}_{\text{adm}} \cap L_+^0 = \{0\}$ , and 0 is maximal in  $\mathcal{G}_{\text{adm}}$  if and only if whenever  $g \in \mathcal{G}_{\text{adm}}$  satisfies  $g \geq 0$  (i.e.  $g \in L_+^0$ ), we have  $g = 0$ . It is clear that these two statements are equivalent, thus completing the proof.

**Exercise 11.2** (*Discrete time: all elements are maximal*) Fix a finite time horizon  $T \in \mathbb{N}$ , and let  $S = (S_k)_{k=0,1,\dots,T}$  be a discrete-time process. Let  $\Theta$  denote the space of all predictable processes. Show that if  $S$  satisfies (NA), then neither  $\mathcal{G}_{\text{adm}}$  nor  $G_T(\Theta)$  contain any non-maximal element.

Can you also show the result without using Theorem 1.2?

**Solution 11.2** As  $\mathcal{G}_{\text{adm}} \subseteq G_T(\Theta)$ , it suffices to prove that every element of  $G_T(\Theta)$  is maximal. So take  $G_T(\vartheta) \in G_T(\Theta)$ , and suppose  $G_T(\vartheta') \in G_T(\Theta)$  such that  $G_T(\vartheta) \leq G_T(\vartheta')$ . Then we have

$$G_T(\vartheta' - \vartheta) = G_T(\vartheta') - G_T(\vartheta) \geq 0.$$

Since  $\vartheta' - \vartheta \in \Theta$ , we have  $G_T(\vartheta' - \vartheta) \in G_T(\Theta) \cap L_+^0$ . As  $S$  satisfies (NA), Theorem 1.2 implies that  $G_T(\Theta) \cap L_+^0 = \{0\}$ , and hence  $G_T(\vartheta' - \vartheta) = 0$ . It follows immediately that  $G_T(\vartheta) = G_T(\vartheta')$ , so that  $G_T(\vartheta)$  is maximal in  $G_T(\Theta)$ , as required.

Another way to get the result is as follows. Since  $S$  satisfies (NA), Corollary 1.3 implies that there is an equivalent martingale measure  $Q$  for  $S$ . Now take  $G_T(\vartheta) \in G_T(\Theta)$ , and suppose there exists  $G_T(\vartheta') \in G_T(\Theta)$  satisfying  $G_T(\vartheta') \geq G_T(\vartheta)$ . Then we have  $G_T(\vartheta' - \vartheta) \geq 0$ , and thus the  $Q$ -local martingale  $G(\vartheta' - \vartheta)$  is in fact a  $Q$ -martingale. This implies that  $E[G_T(\vartheta' - \vartheta)] = 0$ , and hence  $G_T(\vartheta' - \vartheta) = 0$   $Q$ -a.s., and thus also  $P$ -a.s. So we have  $G_T(\vartheta) = G_T(\vartheta')$ , and thus  $G_T(\vartheta)$  is maximal in  $G_T(\Theta)$ . This completes the proof.

**Exercise 11.3** (*Uniqueness*) Let  $S$  be an RCLL semimartingale satisfying (NFLVR) and let  $w$  be a feasible weight function. Suppose  $f \in L^0$  with  $|f| \leq w$ , and

assume  $\alpha = \beta$  are finite, where as usual  $\alpha := \inf \Gamma_+$  and  $\beta := \sup \Gamma_-$  are the super- and subreplicating prices for  $f$ .

- (a) Show that there exists a unique  $g \in \mathcal{G}_w$  such that  $f \leq \alpha + g$ .
- (b) Show directly (without using any results from the course) that  $g$  is maximal in  $\mathcal{G}_w$ .

### Solution 11.3

- (a) By Theorem 9.1, there exist  $g, g' \in \mathcal{G}_w$  with

$$\alpha + g \geq f \quad \text{and} \quad -\beta + g' \geq -f.$$

(Note this gives existence of  $g$ .) Adding these inequalities and remembering that  $\alpha = \beta$ , we get  $g + g' \geq 0$ . But also by Lemma 8.1(a), we have  $E_Q[g + g'] \leq 0$  for any  $Q \in \mathbb{P}_{e,\sigma}^w$ , so that  $g + g' = 0$ . We then have

$$\alpha + g \geq f \quad \text{and} \quad -\alpha - g \geq -f.$$

The second inequality above is equivalent to  $f \geq \alpha + g$ , and together with the first inequality above we get  $\alpha + g = f$ , so that  $g = f - \alpha$ . It follows immediately that  $g$  is unique, as required.

- (b) Assume  $g' \in \mathcal{G}_w$  with  $g' \geq g$ . Then we have  $f \leq \alpha + g \leq \alpha + g'$ , and by the uniqueness in part (a), we get  $g' = g$  so that  $g$  is maximal, as required.

**Exercise 11.4** (*Maximality in a larger set*) Let  $S$  be an RCLL semimartingale satisfying (NFLVR). Let  $w$  be a feasible weight function and fix  $g \in \mathcal{G}_w$ . Define the random variable  $w' := w + g^+$ , where  $g^+ := \max\{g, 0\}$ . Show that  $w'$  is a feasible weight function, and that if  $g$  is maximal in  $\mathcal{G}_w$  then  $g$  is also maximal in  $\mathcal{G}_{w'}$ .

**Solution 11.4** Since  $S$  satisfies (NFLVR), we know that  $\mathbb{P}_{e,\sigma}^w \neq \emptyset$ . So take  $Q \in \mathbb{P}_{e,\sigma}^w$ . By Lemma 8.1(a), we have  $E_Q[g] \leq 0$ , and hence  $g^+ \in L^1(Q)$ . Since  $w' \geq w$  and  $Q \in \mathbb{P}_{e,\sigma}^w \subseteq \mathbb{P}_{e,\sigma}$ , it follows that  $w'$  is a feasible weight function.

Now assume  $g$  is maximal in  $\mathcal{G}_w$ . Before showing  $g$  is maximal in  $\mathcal{G}_{w'}$ , we first check that  $\mathbb{P}_{e,\sigma}^{w'} = \mathbb{P}_{e,\sigma}^w$ . Since  $w' \geq w$ , clearly  $\mathbb{P}_{e,\sigma}^{w'} \subseteq \mathbb{P}_{e,\sigma}^w$ , and the reverse inclusion follows as  $g^+ \in L^1(Q)$  and  $Q \in \mathbb{P}_{e,\sigma}^w$  was chosen arbitrarily.

Now using  $w' \geq w$  and  $\mathbb{P}_{e,\sigma}^{w'} = \mathbb{P}_{e,\sigma}^w$ , we have  $\mathcal{G}_{w'} \supseteq \mathcal{G}_w$ , so that in particular  $g \in \mathcal{G}_{w'}$ . It remains to prove that  $g$  is also maximal in  $\mathcal{G}_{w'}$ . To this end, suppose  $\tilde{g} \in \mathcal{G}_{w'}$  with  $\tilde{g} \geq g$ . As  $g \in \mathcal{G}_w$ , there exists some  $a \geq 0$  such that  $g \geq -aw$ , and thus also  $\tilde{g} \geq -aw$ . Hence Lemma 8.1(d) gives  $\tilde{g} \in \mathcal{G}_{aw}^1 \subseteq \mathcal{G}_w$ . As  $\tilde{g} \in \mathcal{G}_w$  and  $g$  is maximal in  $\mathcal{G}_w$ , it follows that  $\tilde{g} = g$ , as claimed.

**Exercise 11.5** (*An example where  $\Gamma_+ \cap \Gamma_-$  is large*) Construct a model for a financial market and a payoff  $f$  such that  $\Gamma_+(f)$  and  $\Gamma_-(f)$  intersect in more than one point.

**Solution 11.5** Consider a process  $S = (S_0, S_1)$  in discrete time such that  $S_0 = 1$  and  $S_1$  satisfies  $P[S_1 = \frac{3}{2}] = P[S_1 = 2] = \frac{1}{2}$ . Note that for each  $b \in \mathbb{R}$  there is an element  $G_T(\vartheta_b) \in G_T(\Theta)$  satisfying  $P[G_T(\vartheta_b) = b] = P[G_T(\vartheta_b) = \frac{b}{2}] = \frac{1}{2}$ , and moreover every element  $G_T(\vartheta) \in G_T(\Theta)$  is of this form.

Now define the payoff  $f \equiv 0$ . A real number  $a \in \mathbb{R}$  is an element of  $\Gamma_+(f)$  if  $a + b \geq 0$  and  $a + \frac{b}{2} \geq 0$  for some  $b \in \mathbb{R}$ . Thus  $\Gamma_+ \supseteq [-\frac{b}{2}, \infty)$  for all  $b \geq 0$ , and hence  $\Gamma_+(f) = \mathbb{R}$ .

Similarly, a real number  $a \in \mathbb{R}$  is an element of  $\Gamma_-(f)$  if  $a \leq b$  and  $a \leq \frac{b}{2}$  for some  $b \in \mathbb{R}$ . Thus  $\Gamma_- \supseteq (-\infty, \frac{b}{2}]$  for all  $b \geq 0$ , and hence  $\Gamma_-(f) = \mathbb{R}$ , so that  $\Gamma_+(f) \cap \Gamma_-(f) = \mathbb{R}$ .