## Solutions to the Exercises of Section 2.5.

2.5.1. Let S be convex set in  $E_k$ , and let  $\overline{S}$  denote its closure. Let  $x \in \overline{S}$  and  $y \in \overline{S}$ , and let  $0 \le \lambda \le 1$ . We are to show that  $\lambda x + (1 - \lambda)y \in \overline{S}$ . Since  $x \in \overline{S}$ , we can find a sequence  $x_n \in S$  such that  $x_n \to x$  as  $n \to \infty$ . Similarly, we can find  $y_n \in S$  such that  $y_n \to y$  as  $n \to \infty$ . Since S is convex,  $\lambda x_n + (1 - \lambda)y_n \in S$ . Then, since  $\lambda x_n + (1 - \lambda)y_n \to \lambda x + (1 - \lambda)y$ , we have  $\lambda x + (1 - \lambda)y \in \overline{S}$ .

2.5.2. Let S be closed from below and bounded, and let  $\mathbf{p}^T = (p_1, \ldots, p_k)$  be an arbitrary prior distribution. Let  $\alpha_0 = \inf\{\alpha : \mathbf{p}^T \mathbf{x} \leq \alpha \text{ for some } x \in S\}$ . First we find  $\mathbf{x}_n \in S$  such that  $\mathbf{p}^T \mathbf{x}_n \leq \alpha_0 + (1/n)$ . Then since S is bounded, we may find a subsequence of the  $\mathbf{x}_n$  that converges to some  $\mathbf{x}$ . This  $\mathbf{x}$  must be in  $\overline{S}$ . Although this point has minimum Bayes risk,  $\mathbf{p}^T \mathbf{x} = \alpha_0$ ,  $\mathbf{x}$  might not be in S. However, consider the set  $S' = Q_{\mathbf{x}} \cap \overline{S}$ . All points  $\mathbf{y} \in S'$  have the same Bayes risk. (The Bayes risk cannot be any better since  $\mathbf{y} \in \overline{S}$ , and it cannot be any worse since  $\mathbf{y} \in Q_{\mathbf{x}}$ .) Moreover, S' is closed (the intersection of two closed sets), convex (the intersection of two convex sets), and nonempty ( $\mathbf{x} \in S'$ ). Hence, by Lemma 1, the lower boundary of S' is not empty,  $\lambda(Q_{\mathbf{x}} \cap \overline{S}) \neq \emptyset$ . However, the lower boundary of S' is contained in the lower boundary of S, because if  $\mathbf{y} \in \lambda(Q_{\mathbf{x}} \cap \overline{S})$ , then  $\{\mathbf{y}\} = Q_{\mathbf{y}} \cap (\overline{Q_{\mathbf{x}} \cap \overline{S}}) = Q_{\mathbf{y}} \cap Q_{\mathbf{x}} \cap \overline{S} = Q_{\mathbf{y}} \cap \overline{S}$ . Since S is closed from below, all points of  $\lambda(Q_{\mathbf{x}} \cap \overline{S})$  are in S and all have the minimum Bayes risk.

2.5.3. We are given that the risk set, S, is bounded from below and closed from below and that  $\delta_0$  is admissible. Let  $\mathbf{x} = (R(\theta_1, \delta_0), \dots, R(\theta_k, \delta_0))$ . We are to show that  $\mathbf{x} \in \lambda(S)$ .

Let  $S_1 = Q_{\mathbf{x}} \cap \overline{S}$ . Then  $S_1$  is nonempty  $(\mathbf{x} \in S_1)$ , convex (both  $Q_{\mathbf{x}}$  and  $\overline{S}$  are convex), and bounded from below (since S is bounded from below). Lemma 2.5.1 applies to give us that  $\lambda(S_1)$  is not empty. Let  $\mathbf{y} \in \lambda(S_1)$ . Then using the fact that  $S_1$  is closed,

$$\{\mathbf{y}\} = Q_{\mathbf{y}} \cap \overline{S}_1 = Q_{\mathbf{y}} \cap S_1 = Q_{\mathbf{y}} \cap Q_{\mathbf{x}} \cap \overline{S} = Q_{\mathbf{y}} \cap \overline{S},$$

since  $\mathbf{y} \in Q_{\mathbf{x}}$  implies that  $Q_{\mathbf{y}} \subseteq Q_{\mathbf{x}}$ . This shows that  $\mathbf{y} \in \lambda(S)$  and since S is closed from below,  $\mathbf{y} \in S$ . But since  $\delta_0$  is admissible,  $\mathbf{x}$  is the only point in  $Q_{\mathbf{x}} \cap S$ . That is,  $\mathbf{x} = \mathbf{y} \in \lambda(S)$ .

2.5.4. (a) The counterexample of Exercise 2.4.4 is also a counterexample here.

(b) Let  $S = \{(x, y) : x \leq 0, 0 < y \leq 1\} \cup \{(0, 0)\}$ . Then since  $\lambda(S) = \emptyset \subseteq S$ , S is closed from below. But  $\{(0, 0)\} \notin \lambda(S)$ , although the  $\delta$  corresponding to (0, 0) is admissible.