Functional Analysis I

D-MATH Prof. Marc Burger

## Exercise Sheet 11 - Solutions

1. Let  $\Gamma$  be a group with a normal subgroup  $N \lhd \Gamma$  such that N and  $\Gamma/N$  satisfy the conclusion of the Markov-Kakutani fixed point theorem. Show that it holds for  $\Gamma$  as well.

Solution: Let V be a topological vector space generated by a sufficient family of seminorms with a continuous action by the discrete group  $\Gamma$ . Suppose  $A \subset V$  is a convex, compact, non-empty and  $\Gamma$ -invariant subset. Define

$$A^N := \{ v \in A | gv = v \ \forall g \in N \}.$$

The MK-fixed point theorem says  $A^N \neq \emptyset$ . Note that for each  $v \in A^N$ ,  $g \in \Gamma$  and  $h \in N$  we have

$$h(gv) = g((g^{-1}hg)v) = gv$$

because N is normal in  $\Gamma$ . This says that  $A^N$  is  $\Gamma$ -invariant. By definition, the subgroup N acts trivially on  $A^N$ , so the  $\Gamma$ -action on  $A^N$  factors through  $\Gamma/N$ . Moreover,  $A^N$  is closed in A (because N acts by continuous automorphisms) and convex. So the MK-fixed point theorem says that there is  $v \in A^N$  such that gNv = g for all  $gN \in \Gamma/N$ . In particular, there is  $v \in A$  such that gv = v for all  $g \in \Gamma$ .

2. Show that if  $\Gamma$  has property (F) (Remark VI.24) then it satisfies the conclusions of the MK-fixed point theorem.

Solution: Let  $A \subset V$  be compact, convex,  $\Gamma$ -invariant, and non-empty. For each finite  $F \subset \Gamma$  we define the map  $T_F \colon A \to A$  by

$$T_F(v) := \frac{1}{|F|} \sum_{g \in F} gv$$

for all  $v \in A$ .

By property (F), there exists a sequence  $(F_n)_n$  of subset of  $\Gamma$  with

$$\frac{|\gamma F_n \triangle F_n|}{|F_n|} \to 0$$

as  $n \to \infty$ . Let  $v_0 \in A$ . There exists a subsequence of  $(T_{F_n}v_0)_n$  which converges to some  $v \in A$  because A is compact. Let  $|| \cdot ||$  be a continuous seminorm on V. Because A is compact there exists  $C \ge 0$  with  $||w|| \le C$  for all  $w \in A$ . For each  $\gamma \in \Gamma$  we have

$$\begin{split} ||\gamma v - v|| &= \left| \left| \lim_{n \to \infty} \frac{1}{|F_n|} \sum_{g \in F_n} \gamma g v_0 - g v_0 \right| \right| \\ &\leqslant \lim_{n \to \infty} \frac{1}{|F_n|} \sum_{g \in \gamma F_n \triangle F_n} ||g v_0|| \\ &\leqslant \lim_{n \to \infty} \frac{C |\gamma F_n \triangle F_n|}{|F_n|} = 0. \end{split}$$

Functional Analysis I

D-MATH Prof. Marc Burger

3. Determine the extreme points of  $M^1([0,1])$ .

Solution: Let  $\mu$  be an extreme point of  $M^1([0,1])$ . Suppose the support of  $\mu$  is not a point. There exists a decomposition  $[0,1] = [0,t) \sqcup [t,1]$  such that  $\mu|_{[0,t)} \neq 0$  and  $\mu|_{[t,1]} \neq 0$ . We can write

$$\mu = \mu([0,t))\frac{\mu|_{[0,t)}}{\mu([0,t)} + \mu([t,1])\frac{\mu|_{[t,1]}}{\mu([t,1])}.$$

This is a contradiction to  $\mu$  being an extreme point. Therefore the support of  $\mu$  is a point and  $\mu$  is a Dirac measure.

Let  $x \in [0, 1]$  and consider the Dirac measure  $\delta_x$  supported at x. Suppose there exist  $\mu_1, \mu_2 \in M^1([0, 1])$  and  $t_1, t_2 \in (0, 1)$  with  $t_1 + t_2 = 1$  and

$$\delta_x = t_1 \mu_1 + t_2 \mu_2.$$

Let  $U \subset [0,1]$  be an open subset such that  $x \notin U$ . Then

$$0 = \delta_x(U) = t_1 \mu_1(U) + t_2 \mu_2(U).$$

This implies  $\mu_1(U) = \mu_2(U) = 0$  because  $t_i \neq 0$ . This says  $\mu_1$  and  $\mu_2$  are supported at x and hence  $\mu_1 = \mu_2 = \delta_x$ .

4. Let  $\varphi: [0,1] \to [0,1], \varphi(x) := x^2$ ; determine all the  $\varphi$ -invariant probability measures.

Solution: Let  $\mu$  be a  $\varphi$ -invariant probability measure. For each  $t \in (0, 1)$  define  $I_t := (0, t)$ . Note that  $\varphi(I_t) = I_{t^2}$ , so we get

$$\mu(I_t) = \mu(I_{t^{2^n}})$$

for each  $n \ge 1$ . The dominated convergence theorem implies that, as  $n \to \infty$ , this equality converges to

$$\mu(I_t) = 0$$

for each  $t \in (0, 1)$ . Thus  $\mu = t_0 \delta_0 + t_1 \delta_1$  where  $\delta_0$  and  $\delta_1$  are the Dirac measures supported at 0 and 1 respectively and  $t_0 + t_1 = 1$ .

5. In Example VI. 26 justify why for  $\alpha \notin \mathbb{Q}/\mathbb{Z}$ ,  $\lambda$  is the unique  $T_{\alpha}$ -invariant probability measure. Solution: Let  $\mu$  be a  $T_{\alpha}$ -invariant probability measure. Define

$$T_m(f)(x) := \frac{1}{m} \sum_{k=0}^{m-1} f(x+k\alpha)$$

for each  $m \ge 1$  and  $f \in C(S^1)$ . Then

$$\int f d\mu = \int T_m f d\mu$$

for all  $f \in C(S^1)$  and  $m \ge 1$  because  $\mu$  is  $T_{\alpha}$ -invariant. Define  $\varphi_n(x) := e^{2\pi i n x}$  for each  $n \in \mathbb{Z}$ . Because  $\alpha \notin \mathbb{Q}/\mathbb{Z}$ , we have

$$T_m(\varphi_n)(x) = \frac{\varphi_n(x)(e^{2\pi i n\alpha m} - 1)}{m(e^{2\pi i n\alpha} - 1)}$$

D-MATH Prof. Marc Burger Functional Analysis I

for all  $n \neq 0$ . This implies  $T_m(\varphi_n) \to 0$  as  $m \to \infty$  and, therefore,

$$\int_{S^1} \varphi_n d\mu = 0$$

for  $n \neq 0$ . Let  $f \in C^{\infty}(S^1)$  then we can write for each  $x \in S^1$ 

$$f(x) = \sum_{n \in \mathbb{Z}} \hat{f}(n) \varphi_n(x).$$

Because f is smooth the Fourier series converges pointwise. Because  $S^1$  is compact we get that the Fourier series converges uniformly. Thus dominated convergence implies

$$\int_{S^1} f d\mu = \int_{S^1} \sum_{n \in \mathbb{Z}} \hat{f}(n) \varphi_n(x) d\mu(x) = \sum_{n \in \mathbb{Z}} \int_{S^1} \hat{f}(n) \varphi_n(x) d\mu(x) = \hat{f}(0) \int_{S^1} d\mu = \hat{f}(0).$$

The smooth functions are dense in the space of continuous functions (convolve with a smooth bump function), so this determines  $\mu = \lambda$ .

6. Let  $T \in SL_2(\mathbb{R})$  with real eigenvalues  $\{\lambda, \lambda^{-1}\}$  with  $\lambda > 1$ . Then T induces a homeomorphism of  $\mathbb{P}^1(\mathbb{R})$  still denoted T. Classify all T-invariant probability measures on  $\mathbb{P}^1(\mathbb{R})$ . Prove that there exists no probability measure on  $\mathbb{P}^1(\mathbb{R})$  which is invariant under  $SL_2(\mathbb{Z})$ .

Solution: There is  $Q \in SL_2(\mathbb{R})$  with  $QTQ^{-1} = \operatorname{diag}(\lambda^{-1}, \lambda)$ . We begin by determining the  $S := \operatorname{diag}(\lambda, \lambda^{-1})$ -invariant probability measures. We denote by  $\mathbb{P}^1(\mathbb{R})$  the set of lines in  $\mathbb{R}^2$ . This space is diffeomorphic to the quotient

$$(\mathbb{R}^2 \setminus \{0\}) / \mathbb{R}^* \xrightarrow{\sim} \mathbb{P}^1(\mathbb{R})$$

via the map which sends v to  $v\mathbb{R}$  (by definition). This diffemorphism provides the homogeneous coordinates for  $\mathbb{P}^1(\mathbb{R})$  where  $[x : y] \in \mathbb{P}^1(\mathbb{R})$  denotes the line  $(x, y)\mathbb{R}$  i.e. the line passing through (x, y). This space is compact because the inclusion  $S^1 \to \mathbb{R}^2 \setminus \{0\}$  induces a continuous surjection

$$S^1 \twoheadrightarrow \mathbb{R}^2 \setminus \{0\}$$

Let  $\mu$  be a S-invariant probability measure. For each  $0 \leq t < \infty$  we define the two subset

$$I_t^+ := \{ [1:s] | s \in [0,t) \}$$
$$I_t^- := \{ [-1:s] | s \in [0,t) \}.$$

We have  $S(I_t^{\pm}) = I_{\lambda^2 t}^{\pm}$  for all  $0 \leq t \leq \infty$ . This implies

$$\mu(I_t^{\pm}) = \mu(I_0^{\pm})$$

for all  $0 \le t \le \infty$ . In particular, the measure  $\mu$  is supported at the two points [1, 0] and [0, 1]. These two points are fixed by S, so all the S-invariant probability measures are probability measures supported at these two points.

Let  $\mu$  be a *T*-invariant probability measure. The pullback  $Q^*\mu$  is *S*-invariant, so it is supported at the two fixed points of *S*. This means  $\mu$  is supported at the two fixed points of *T* (the eigenvectors). Any probability measure supported at these two points is automatically *T*-invariant, so these are all the *T*-invariant probability measures.