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Probability Theory — Sets of admissible shifts of convex measures, by LAVRENTIN M. ARUTYUNYAN and EGOR D. KOSOV, communicated on 14 November 2014.

ABSTRACT. — We prove that for any convex probability measure on a linear space the set of its non-singular shifts is convex and the set of its equivalent shifts is a linear subspace.

KEY WORDS: Convex measure, logarithmically concave measure, admissible shift.

MATHEMATICS SUBJECT CLASSIFICATION: 28C20, 46G12, 60B11.

We study the sets of non-singular and equivalent shifts for convex measures on locally convex spaces. It is well known that for a Radon Gaussian measure (see [1]) these sets are linear and coincide with the Cameron–Martin space, for example, for the countable power of the standard Gaussian measure on the real line this is the usual l^2 . For a general measure, these sets need not be convex. We prove that for a Radon convex measure, the set of its equivalent shifts is always a linear space and the set of its non-singular shifts is always convex.

Let us recall some concepts and notation. A Borel probability measure μ on a locally convex space X is called Radon if for every Borel set A and for every $\varepsilon > 0$ there is a compact set $K \subset A$ such that $\mu(A \setminus K) < \varepsilon$. In addition to the Borel σ -field $\mathscr{B}(X)$ we shall also need the cylindric σ -field $\sigma(X)$ of X, i.e., the smallest σ -field with respect to which all continuous linear functionals on X are measurable. In the case of a separable Banach space the two σ -fields coincide.

Let μ and ν be two probability measures absolutely continuous with respect to a positive measure λ , i.e. $\mu = \varphi \cdot \lambda$, $\nu = \psi \cdot \lambda$. The number

$$H(\mu, \mathbf{v}) = \int \sqrt{\varphi \psi} \, d\lambda$$

is called the Hellinger integral of this pair of measures. It is independent of our choice of a measure λ and the following estimate is true (see [2, Theorem 4.7.37]):

$$2(1 - H(\mu, \nu)) \le \int |\varphi - \psi| \, d\lambda = \|\mu - \nu\| \le 2\sqrt{1 - H^2(\mu, \nu)},$$

where $\|\cdot\|$ denotes the total variation norm. The latter estimate shows that $\mu \perp \nu$ is equivalent to $H(\mu, \nu) = 0$, where \perp denotes the mutual singularity relation between measures. The equivalence of μ and ν is denoted by $\mu \sim \nu$.

A probability measure μ on the σ -field $\sigma(X)$ is called convex (or logarithmically concave) if

$$\mu(tA + (1-t)B) \ge (\mu(A))^t (\mu(B))^{1-t}$$

for all $t \in [0, 1]$ and $A, B \in \sigma(X)$. The convexity of a Radon probability measure is defined as the convexity of its restriction to the σ -field $\sigma(X)$. It is easily seen from the definition that the topological support of a Radon convex measure, i.e., the smallest closed set of full measure, is convex.

Another equivalent definition can be formulated in terms of finite-dimensional projections (see [5]). A probability measure on \mathbb{R}^n is called convex if it is defined by a density of the form e^{-V} with respect to Lebesgue measure on some affine subspace L, where $V: L \to (-\infty, \infty]$ is a convex function on this subspace. A Radon probability measure μ on a locally convex space X is convex if and only if for every continuous linear operator $P: X \to \mathbb{R}^n$ the measure $\mu \circ P^{-1}$ is convex on \mathbb{R}^n , where $\mu \circ P^{-1}(A) = \mu(P^{-1}(A))$. Let μ be a Borel measure on a locally convex space. Let μ_h be the shift of the

measure μ by the vector h, i.e., the measure defined by the formula

$$\mu_h(A) = \mu(A - h), \quad A \in \mathscr{B}(X).$$

Let $M(\mu)$ denote the set of all its non-singular shifts and let $M(\mu) \subset M(\mu)$ be the set of all its equivalent shifts, i.e.,

$$M(\mu) = \{h : \mu \not\perp \mu_h\},\ ilde{M}(\mu) = \{h : \mu \sim \mu_h\}.$$

In relation to these sets let us also mention the subspace of continuity $C(\mu)$ of μ consisting of all vectors h such that $\lim_{t\to 0} ||\mu - \mu_{th}|| = 0$. The subspace of quasiinvariance $Q(\mu)$ is the set of all vectors h such that $\mu_{th} \sim \mu$ for all t. Clearly, $Q(\mu) \subset \tilde{M}(\mu)$, but the inclusion may be strict. An important difference is that $Q(\mu)$ is always a linear subspace, moreover, $Q(\mu) \subset C(\mu)$ and $C(\mu)$ is also linear (see [3, Chapter 5], [4]). Some difference between these sets occurs already in the one-dimensional case: if $\mu = \sum_{n} 2^{-|n|-1} \delta_n$, where summation is taken over all integer numbers n and δ_n is Dirac's measure at n, then $C(\mu) = Q(\mu) = \{0\}$, $M(\mu) = M(\mu) = \mathbb{Z}$, for Lebesgue measure λ on the interval [0, 1] regarded as a measure on \mathbb{R} we have $C(\lambda) = \mathbb{R}$, $Q(\lambda) = \tilde{M}(\lambda) = \{0\}$, $M(\lambda) = (-1, 1)$.

In infinite dimensions, some more subtle phenomena take place (see [7], [12]).

Another important set is the set $D_C(\mu)$ of all vectors h such that the measure μ has a Skorohod derivative $d_h\mu$ along h. The latter is defined by means of the integration by parts formula

$$\int_X \partial_h f(x)\mu(dx) = -\int_X f(x) \, d_h\mu(dx)$$

for all functions f of the form $f(x) = \varphi(l_1(x), \dots, l_n(x))$, where $\varphi \in C_b^{\infty}(\mathbb{R}^n)$ and $l_i \in X^*$. The Skorohod derivative $d_h\mu$ can be equivalently defined as a limit of $(\mu_{th} - \mu)/t$ in the weak topology as $t \to 0$; its existence is equivalent to the estimate $\|\mu_{th} - \mu\| \leq C|t|$ with some constant C. The set $D_C(\mu)$ is a linear subspace of $C(\mu)$. For a convex measure μ given by a density on \mathbb{R}^n , the sets $C(\mu)$ and $D_C(\mu)$ coincide with \mathbb{R}^n (for $D_C(\mu)$ this is not trivial and was first shown by Krugova [8]). For any convex measure μ (in finite or infinite dimensions) there is the following useful inequality due Krugova [9]:

$$\|\mu_{th} - \mu\| \ge 2 - 2e^{-\frac{1}{2}\|d_h\mu\|},$$

which in the case where $d_h\mu$ does not exist asserts that μ and μ_h are mutually singular. Therefore, for any convex measure μ we have

$$M(\mu) \subset C(\mu) = D_C(\mu).$$

However, it can happen that $M(\mu)$ is not linear: for example, for Lebesgue measure λ on [0, 1] regarded as a measure on the real line we have $M(\lambda) = (-1, 1)$. More generally, if μ is an absolutely continuous convex measure on \mathbb{R}^d , then its topological support is a convex set U with a nonempty interior W and then $M(\mu) = W - W$. Our result says that $M(\mu)$ is a convex subset in $C(\mu)$ and that $\tilde{M}(\mu) = Q(\mu)$.

It is worth noting that in the general case $C(\mu)$ is complete with respect to the distance $d(a,b) = \sup_{|t| \le 1} ||\mu_{ta} - \mu_{tb}||$ that is consistent with the vector structure, but a similar distance $d_0(a,b) = ||\mu_a - \mu_b||$ may fail to be consistent with the vector structure even on $Q(\mu)$ (see [12]). However, it follows by the Krugova inequality that for a convex measure μ the distance d_0 defines the same topology as the distance d. Indeed, if $d_0(h_n, h) \to 0$, then by this inequality $||d_{h-h_n}\mu|| \to 0$, whence it follows that

$$\|\mu_{th} - \mu_{th_n}\| \le |t| \|d_{h-h_n}\mu\| \le \|d_{h-h_n}\mu\| \to 0$$

uniformly in $t \in [-1, 1]$.

It is readily verified that for any Radon measure μ the sets introduced above do not change if we consider μ on $\sigma(X)$ in place of $\mathscr{B}(X)$.

It is unknown whether $M(\mu) \neq \{0\}$ for any non Dirac convex measure.

THEOREM 1. Let μ be a convex Radon measure on a locally convex space X. Then the function $H(h) = H(\mu, \mu_h)$ is logarithmically concave, i.e.,

$$H(th + (1 - t)q) \ge H^{t}(h)H^{1-t}(q) \quad \forall t \in [0, 1].$$

It follows that $M(\mu)$ is a convex set.

PROOF. We first consider the case of a finite-dimensional space. In this case, the measure μ is absolutely continuous with respect to the standard Lebesgue measure on some affine subspace L + v, where L is a linear subspace and v is a vector. If h is not in L, then H(h) = 0 and the inequality is obvious. So, we assume that $h \in L$ and, passing to the subspace L, we can assume that μ has a density with

respect to Lebesgue measure on the whole space, i.e., $\mu = e^{-V} dx$, where V is a convex function. In this case we have

$$H(h) = \int e^{-(V(x)+V(x+h))/2} dx.$$

Let us recall the following Prékopa–Leindler inequality (see, e.g., [10], [11] or [2, Theorem 3.10.21]): if

$$f(tx + (1 - t)y) \ge (\varphi(x))^t (\psi(y))^{1-t}$$

then

$$\int f(x) \, dx \ge \left(\int \varphi(x) \, dx\right)^t \left(\int \psi(x) \, dx\right)^{1-t}.$$

Applying this inequality to the functions

$$f(x) = \varphi(x) = \psi(x) = e^{-(V(x) + V(x+h))/2}$$

we immediately obtain the desired assertion.

Let us proceed to the infinite-dimensional case. Let λ be a Radon probability measure such that $\mu = \varrho \cdot \lambda$, $\mu_h = \varrho_h \cdot \lambda$, e.g., $\lambda = (\mu + \mu_h)/2$. Note that the functions ϱ and ϱ_h can be chosen measurable with respect to the σ -field $\sigma(X)$ (see [1, Corollary A.3.13]). Then, there exists a σ -field generated by countably many continuous linear functionals $\{\ell_i\}$ such that ϱ_h and ϱ are measurable with respect to it (see [1, Lemma 2.1.2]). Let ϱ^n and ϱ_h^n be the conditional expectations of the functions ϱ and ϱ_h , correspondingly, with respect to the measure λ and the σ -field generated by the functionals $\{\ell_i\}_{i=1}^n$. Then the functions ϱ^n and ϱ_h^n converge in $L^1(\lambda)$ to the functions ϱ and ϱ_h , respectively. Therefore, we have convergence $\sqrt{\varrho^n} \to \sqrt{\varrho}, \sqrt{\varrho_h^n} \to \sqrt{\varrho_h}$ in $L^2(\lambda)$. Let us define continuous linear operators $P_n : X \to \mathbb{R}^n$ by the formula $x \mapsto (l_1(x), \ldots, l_n(x))$. Obviously,

$$H(h) = \lim_{n \to \infty} \int \sqrt{\varrho^n \varrho_h^n} \, d\lambda.$$

On the other hand,

$$\int \sqrt{\varrho^n \varrho_h^n} \, d\lambda = H(\mu \circ P_n^{-1}, \mu_h \circ P_n^{-1})$$

It follows from the finite-dimensional case that the function

$$H(\mu \circ P_n^{-1}, \mu_h \circ P_n^{-1}) = H(\mu \circ P_n^{-1}, (\mu \circ P_n^{-1})_{P_n(h)})$$

is logarithmically concave. Hence the same is true for the function H(h) that is a limit of logarithmically concave functions.

Let μ be a Radon measure on a locally convex space X, $h \in X$, $\ell \in X^*$, $\ell(h) = 1$ and $S = \ker \ell$. Then any vector x can be uniquely represented in the form $\alpha h + z$, $z \in S$. Let us consider the sections of the set A by the lines parallel to h:

$$A^{z} = \{ \alpha \in \mathbb{R} : z + \alpha h \in A \}, \quad A \in \mathscr{B}(X).$$

Let $\mu^{\{z\}}$ denote the conditional measures of the measure μ associated with ℓ (see [2, Chapter 10]), i.e.,

$$\mu(A) = \int_{S} \mu^{\{z\}}(A^{z})\mu \circ \pi^{-1}(dz),$$

where $\pi(x) = x - \ell(x)h$.

THEOREM 2. If μ is a convex Radon measure on a locally convex space X, then $\tilde{M}(\mu)$ is a linear subspace in X, hence $\tilde{M}(\mu) = Q(\mu)$.

PROOF. It is known that for $\mu \circ \pi^{-1}$ -a.e. *z* the measure $\mu^{\{z\}}$ is convex (see [6] or [3, Theorem 4.3.6]). Let $h \in \tilde{M}(\mu)$. Let us show that for $\mu \circ \pi^{-1}$ -a.e. point *z* in the subspace *S*, the measure $\mu^{\{z\}}$ on the real line has full support. Suppose the opposite, i.e., there is a compact set *B* such that $\mu \circ \pi^{-1}(B) > 0$ and for every $z \in B$ we have

$$\mu^{\{z\}}([a_z, b_z]) = 1,$$

where the interval $[a_z, b_z]$ is the support of the measure $\mu^{\{z\}}$ (possibly unbounded or degenerate, i.e., $a_z = b_z$). Let us note that for every $z \in B$

$$\mu^{\{z\}}((b_z - 1, b_z]) > 0.$$

Let us consider the set

$$A = \{x : x = \alpha h + z, z \in B, b_z < \alpha \le b_z + 1\}.$$

It is easy to see that $\mu(A) = 0$ and $\mu_h(A) > 0$. Indeed, for almost every z we have $(\mu_h)^{\{z\}} = (\mu^{\{z\}})_1$, which is an easy consequence of the a.e. uniqueness of conditional measures. Therefore, this contradicts the equivalence of the measures μ and μ_h and shows that the supports $[a_z, b_z]$ are unbounded from above; the unboundedness from below is proved similarly. So, $\mu \sim \mu_{th}$ for every t, because

$$(\mu_{th})^{\{z\}} = (\mu^{\{z\}})_t \sim \mu^{\{z\}},$$

i.e., $th \in \tilde{M}(\mu)$ for every $t \in \mathbb{R}$ and every $h \in \tilde{M}(\mu)$. The fact that $h + q \in \tilde{M}(\mu)$ for all $h, q \in \tilde{M}(\mu)$ follows immediately from the definition of the equivalence of measures.

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Lavrentin M. Arutyunyan Department of Mechanics and Mathematics Moscow State University Moscow, 119899, Russia lavrentin@yandex.ru

Egor D. Kosov Department of Mechanics and Mathematics Moscow State University Moscow, 119899, Russia ked_2006@mail.ru