

HARMONIC ANALYSIS ON CERTAIN VECTOR SPACES

BY

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1. Introduction. Let l denote the vector space of all sequences of real numbers with the topology of coordinate-wise convergence. For $0 < p < \infty$ let l_p denote the subset of l consisting of all sequences which have $\|x\| = \{\sum_{i=1}^{\infty} |x_i|^p\}^{1/p}$ finite where $x = (x_1, x_2, \dots)$. Thus, for $1 \leq p < \infty$, we have $\|\cdot\|$ as the usual norm for l_p , and for $0 < p < 1$ it is known that $\|\cdot\|^p$ generates a distance function under which l_p is a complete linear metric space. Our main efforts in this paper are to generalize Bochner's theorem and Levy's continuity theorem to these l_p spaces. For $p=2$ our work coincides with and simplifies some of the work of L. Gross in [3], [4], and [5], and is related to the work of V. Sazonov [8], [9]; Ju. V. Prohorov [7], [8]; and to that of N. N. Vakhania in [10], [11].

It should be mentioned that the case $p=2$ is brought into our work from two different points of view. One of these handles the situation for $0 \leq p \leq 2$ and the other for $2 \leq p < \infty$. In both instances they agree with the work of L. Gross mentioned above. The tool that allows us to simplify Gross' work is a "stochastic inner product", defined in §3, which allows the immediate application of usual measure theoretic manipulations. This idea of a stochastic inner product is suggested in the work of R. H. Cameron and R. E. Graves [1].

We will frequently think of the l_p spaces as being subsets of l and if $x \in l$ we define

$$\begin{aligned} P_N x &= (x_1, \dots, x_N, 0, 0, \dots), \\ Q_N x &= (0, \dots, 0, x_{N+1}, \dots), \\ \mathcal{P}_N x &= (0, \dots, 0, x_N, 0, \dots). \end{aligned}$$

This terminology is standard throughout the paper.

2. The Fourier transform (or characteristic functional) of a probability measure μ on the Borel subsets of a linear topological space X is the function $\phi(x)$ on X^* (the topological dual of X) such that

$$\phi(x) = \int_X \exp \{i(x, y)\} d\mu(y).$$

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The classical version of Bochner's theorem asserts that a function $\phi(x)$, $x \in R_n$, is the Fourier transform of some probability measure on the Borel subsets of R_n if and only if ϕ is positive definite, $\phi(0)=1$, and ϕ is continuous at $x=0$. In l_2 , and hence in any real separable Hilbert space, it is possible to introduce a topology τ (which is determined by certain compact operators) such that a function ϕ on l_2 is the Fourier transform of some probability measure on the Borel subsets of l_2 if and only if ϕ is positive definite, $\phi(0)=1$, and ϕ is continuous at zero in the τ -topology. Here we are, of course, identifying l_2^* and l_2 . The l_2 result is due independently to L. Gross [5] and to V. Sazonov's earlier work [9].

Before proving our analogue for Bochner's theorem for l_p , $0 < p \leq 2$, we first take a closer look at the situation in l_2 .

An operator on l_2 which is linear, symmetric, nonnegative, compact, and having finite trace will be called an S -operator. If T is an S -operator on l_2 then it is well known that T has the representation

$$(2.1) \quad Tx = \sum_{n=1}^{\infty} \lambda_n(x, e_n) e_n$$

where $\{e_n\}$ is some orthonormal subset of l_2 , $\lambda_n \geq 0$, and $\sum_{n=1}^{\infty} \lambda_n < \infty$. The S -operator T also has a representation as an infinite symmetric positive-definite matrix $T = \{t_{ik}\}$ where by positive-definite it is meant that $\sum_{i,k=1}^n t_{ik} x_i x_k \geq 0$ for any integer n and any $x \in R_n$. Furthermore, $t_{ik} = (Tf_i, f_k)$ where f_j is a sequence of all zeros and having 1 in the j th position and hence $\sum_{i=1}^{\infty} t_{ii} = \sum_{j=1}^{\infty} \lambda_j < \infty$ where the λ_j 's are as in (2.1). From the representation in (2.1) it is easy to verify that $(Tcx, cx)^{1/2} = |c|(Tx, x)^{1/2}$ for any real number c and $(T(x+y), (x+y))^{1/2} \leq (Tx, x)^{1/2} + (Ty, y)^{1/2}$. Thus $(Tx, x)^{1/2}$ is a seminorm on l_2 . The τ -topology on l_2 is the topology generated by taking as a subbase all translates of all sets of the form

$$\{x \in l_2 : (Tx, x) < r\}$$

where $r > 0$ and T is an S -operator.

As mentioned above, Gross [5] and Sazonov [9] have proved an analogue of Bochner's theorem which states that continuity in the τ -topology at the origin is necessary and sufficient for a positive-definite complex-valued function ϕ on l_2 with $\phi(0)=1$ to be the Fourier transform of a positive finite measure. The following lemma demonstrates that τ -continuity on l_2 is equivalent to ordinary continuity and continuity with respect to a certain directed set of distributions.

LEMMA 2.1. *Let $\phi(x)$ be a positive-definite functional defined on l_2 such that $\phi(0)=1$. Then $\phi(x)$ is τ -continuous at zero (and hence everywhere on l_2) if and only if $\phi(x)$ is continuous in the norm topology and*

$$(2.2) \quad \phi(0) = \lim_{N \rightarrow \infty} \int_{l_2} \phi(x) \lambda(\sigma_N, dx)$$

where $\lambda(\sigma_N, \cdot)$ denotes any Gaussian measure on $P_N(l_2)$ with density

$$\left[(2\pi)^N \prod_{j=1}^N \sigma_{N,j}^2 \right]^{-1/2} \cdot \exp \left\{ -\frac{1}{2} \sum_{j=1}^N \frac{x_j^2}{\sigma_{N,j}^2} \right\}$$

such that $\max_{1 \leq j \leq N} \sigma_{N,j}^2$ tends to zero as N tends to infinity.

Proof. Suppose ϕ is continuous on l_2 in the τ -topology. Then we have for each $\varepsilon > 0$ an S -operator such that $x \in E = \{x : \sum_{i,j=1}^{\infty} t_{ij} x_i x_j < 1\}$ implies $|1 - \phi(x)| < \varepsilon$. Thus if

$$A_N = \int_{l_2} [1 - \phi(x)] \lambda(\sigma_N, dx)$$

we find

$$\begin{aligned} |A_N| &\leq \varepsilon + 2 \int_{E^c} \lambda(\sigma_N, dx) \\ &\leq \varepsilon + 2 \int_{l_2} \sum_{i,j=1}^{\infty} t_{ij} x_i x_j \lambda(\sigma_N, dx) \\ &\leq \varepsilon + 2 \sum_{j=1}^N t_{jj} \sigma_{N,j}^2 \\ &\leq \varepsilon + 2 \max_{1 \leq j \leq N} \sigma_{N,j}^2 \sum_{j=1}^{\infty} t_{jj}. \end{aligned}$$

Hence $\limsup_N |A_N| \leq \varepsilon$ so actually A_N tends to zero, and τ -continuity of ϕ implies (2.2) holds.

Since ϕ is positive definite and norm continuous it follows that $\phi(P_k(\cdot))$ is a continuous, positive-definite function on l for $k=1, 2, \dots$. Thus there exists a probability measure μ on the Borel subsets of l whose finite dimensional distributions are determined by $\phi(P_k(\cdot))$, $k=1, 2, \dots$. Hence

$$A_N = \int_l \left[1 - \exp \left\{ -\frac{1}{2} \sum_{j=1}^N \sigma_{N,j}^2 x_j^2 \right\} \right] d\mu(x).$$

Now $\lim_N A_N = 0$ implies $\sum_{j=1}^N \sigma_{N,j}^2 x_j^2$ tends to zero in μ -measure as $\max_{1 \leq j \leq N} \sigma_{N,j}^2$ tends to zero. This implies $\sum_{j=1}^{\infty} x_j^2$ is finite for almost all $x \in l$ with regard to the measure μ so $\mu(l_2) = 1$. Hence if $\psi(x) = \int_{l_2} \{i(x, y)\} d\mu(y)$ we find $\psi(x) = \phi(x)$ on $P_k(l_2)$ for $k=1, 2, \dots$. Since both functions are continuous on l_2 and $\bigcup_{k=1}^{\infty} P_k l_2$ is dense in l_2 this implies $\psi(x) = \phi(x)$ on l_2 . Thus ϕ is the Fourier transform of μ and hence

$$\phi(x) = \int_{l_2} \exp \{i(x, y)\} d\mu(y).$$

Let $\varepsilon > 0$ be given and choose a compact set $K \subseteq l_2$ such that $\mu(l_2 - K) < \varepsilon/2$. Then

$$1 - \operatorname{Re} \phi(x) = \int_{l_2} [1 - \cos(x, y)] d\mu(y) \leq \frac{1}{2} \int_K (x, y)^2 d\mu(y) + \frac{\varepsilon}{2}$$

and if we denote by T the S -operator determined by the relation

$$(Tx, x) = \int_K (x, y)^2 d\mu(y)$$

it follows that $(Tx, x) < \varepsilon$ implies $1 - \operatorname{Re} \phi(x) < \varepsilon$. The τ -continuity of ϕ at zero now follows since

$$|1 - \phi(x)|^2 \leq 2(1 - \operatorname{Re} \phi(x)).$$

Thus continuity in the τ -topology for positive-definite continuous functions on l_2 is equivalent to continuity with respect to a certain directed set of Gaussian distributions. This motivates our next result which generalizes Bochner's theorem to l_p , $0 < p \leq 2$. In view of the previous lemma this agrees with the known results for $p=2$ in [5] and [9].

A function ϕ on l_p^* will be called sequentially weak-star continuous if for each sequence $\{x_n\}$ in l_p^* satisfying $\lim_n (y, x_n) = (y, x)$ for every $y \in l_p$ and some $x \in l_p^*$ we have $\lim_n \phi(x_n) = \phi(x)$.

THEOREM 2.1. *Let $0 < p \leq 2$ and suppose $\phi(x)$ is a function defined on l_p^* ⁽³⁾. Then ϕ is the Fourier transform of a probability measure on l_p if and only if ϕ is positive definite, $\phi(0)=1$, ϕ is sequentially weak-star continuous, and*

$$(2.3) \quad \phi(0) = \lim_{N \rightarrow \infty} \int_{l_p^*} \phi(x) \lambda_p(\varepsilon_N, dx).$$

Here $\lambda_p(\varepsilon_N, \cdot)$ denotes any stable distribution on $P_N(l_p^*)$ whose Fourier transform is

$$\phi(t_1, \dots, t_N) = \exp \left\{ - \sum_{j=1}^N \varepsilon_{N,j} |t_j|^p \right\}$$

where the $\varepsilon_{N,j}$'s are positive and $\max_{1 \leq j \leq N} \varepsilon_{N,j}$ tends to zero as N approaches infinity.

Proof. Let $B(\varepsilon_N) = \int_{l_p^*} \phi(x) \lambda_p(\varepsilon_N, dx)$. Since ϕ is positive definite, $\phi(0)=1$, and ϕ is continuous on l_p^* there exists a probability measure μ on l whose finite-dimensional distributions are determined by $\phi(P_K(\cdot))$ for $K=1, 2, \dots$. Therefore,

$$\begin{aligned} B(\varepsilon_N) &= \int_{P_N(l_p^*)} \phi(x) \lambda_p(\varepsilon_N, dx) \\ &= \int_{P_N(l_p^*)} \int_l \exp \{i(x, y)\} d\mu(y) \lambda_p(\varepsilon_N, dx) \\ &= \int_l \int_{P_N(l_p^*)} \exp \{i(x, y)\} \lambda_p(\varepsilon_N, dx) d\mu(y) \\ &= \int_l \exp \left\{ - \sum_{j=1}^N \varepsilon_{N,j} |y_j|^p \right\} d\mu(y). \end{aligned}$$

⁽³⁾ If $0 < p \leq 1$ then $l_p^* = l_\infty$, the space of bounded sequences, with the usual sup norm.

Since $B(\varepsilon_N)$ tends to $\phi(0)=1$ when $\max_{1 \leq j \leq N} \varepsilon_{N,j}$ tends to zero as N approaches infinity we find $\max_{1 \leq j \leq N} \varepsilon_{N,j} \sum_{j=1}^N |y_j|^p$ tends to zero in μ -measure. Thus $\sum_{j=1}^\infty |y_j|^p$ is finite with μ -measure one. In other words, $\mu(l_p^*)=1$. If

$$\psi(x) = \int_{l_p} \exp \{i(x, y)\} d\mu(y) \quad \text{for } x \in l_p^*$$

then $\psi(x)=\phi(x)$ on $\bigcup_{K=1}^\infty P_K(l_p^*)$ and since both ψ and ϕ are continuous on l_p^* it follows that ϕ actually equals ψ on l_p^* , and hence is the Fourier transform of μ .

On the other hand, if μ is a probability measure on l_p , then it is trivial to verify that

$$\phi(x) = \int_{l_p} \exp \{i(x, y)\} d\mu(y) \quad (x \in l_p^*)$$

is positive definite, continuous, and that $\phi(0)=1$. Further,

$$\begin{aligned} 1 \geq B(\varepsilon_N) &= \int_{l_p} \exp \left\{ - \sum_{j=1}^N \varepsilon_{N,j} |y_j|^p \right\} d\mu(y) \\ &\geq \int_{l_p} \exp \left\{ - \max_{1 \leq j \leq N} \varepsilon_{N,j} \sum_{j=1}^N |y_j|^p \right\} d\mu(y) \end{aligned}$$

and the last integral converges to one so $B(\varepsilon_N)$ tends to $\phi(0)=1$ in the manner indicated.

In our investigation of probability measures on l_p , $2 \leq p < \infty$, via Fourier transforms, we found the following concept useful. A family of probability measures $\{\mu_\alpha : \alpha \in A\}$ on l_p , $2 \leq p < \infty$, is a λ -family for some λ in the positive cone of $l_{p/2}^*$ if for every ε , $\delta > 0$ there is a sequence $\{\varepsilon_N\}$ such that

$$\mu_\alpha \left\{ y \in l_p : \sum_{i=N+1}^\infty \lambda_i y_i^2 < \delta \right\} > 1 - \varepsilon$$

implies

$$\mu_\alpha \left\{ y \in l_p : \sum_{i=N+1}^\infty |y_i|^p < \delta \right\} > 1 - (\varepsilon + \varepsilon_N)$$

where $\lim_N \varepsilon_N = 0$.

It is quite clear that any family of probability measures in l_2 is a λ -family for $\lambda=(1, 1, \dots)$.

We also need a generalization of the τ -topology to l_q ($1/q+1/p=1$). If $2 \leq p < \infty$ then a linear operator T from l_q into l_p is an S_p -operator if T can be represented as an infinite symmetric positive-definite matrix (t_{ij}) such that $\sum_{i=1}^\infty (t_{ii})^{p/2}$ is finite. Here, by positive-definite, we mean that $\sum_{i,j=1}^n t_{ij} x_i x_j \geq 0$ for all $x \in R_n$ and all integers n . The τ_p -topology $2 \leq p < \infty$, is generated by taking as a subbase all translates of all sets of the form $\{x \in l_p^* : (Tx, x) < r\}$ where $r > 0$ and T is an S_p operator. Thus the τ_2 -topology is the τ -topology.

The next theorem is a generalization of Prohorov's result [7] which handled the case $p=2$.

THEOREM 2.2. A family $\{\mu_\alpha : \alpha \in A\}$ of probability measures on l_p , $2 \leq p < \infty$, is conditionally compact if and only if:

(a) For every μ_α and $\varepsilon > 0$ there exists an S_p -operator $T_{\alpha,\varepsilon}$ such that for $x \in l_p^*$

$$\operatorname{Re} (1 - \phi(x, \mu_\alpha)) \leq (T_{\alpha,\varepsilon} x, x) + \varepsilon$$

where $\phi(\cdot, \mu_\alpha)$ is the Fourier transform of μ_α .

(b) The norms of the operators $T_{\alpha,\varepsilon}$ are uniformly bounded by a constant M_ε .

(c) $\lim_N \sup_\alpha \sum_{k=N}^\infty (t_{kk}^{(\alpha,\varepsilon)})^{p/2} = 0$.

(d) $\{\mu_\alpha : \alpha \in A\}$ is a λ -family for some λ in the positive cone of $l_{p/2}^*$.

The proof will depend on the following lemmas.

LEMMA 2.2. If $\{\mu_\alpha : \alpha \in A\}$ is a conditionally compact set of probability measures on l_p , $2 \leq p < \infty$, then $\{\mu_\alpha : \alpha \in A\}$ is a λ -family for any λ in the positive cone of $l_{p/2}^*$.

Proof. Let ε, δ be given and choose $\lambda \in l_{p/2}^*$ where $\lambda_k > 0$ for $k = 1, 2, \dots$. Since $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact there exists a compact set K of l_p such that $\mu_\alpha(K) > 1 - \varepsilon$ for each $\alpha \in A$. Thus there exists an N such that for each $\alpha \in A$

$$\mu_\alpha \left\{ x \in l_p : \sum_{j=N+1}^\infty |x_j|^p < \delta \right\} > 1 - \varepsilon,$$

so clearly $\{\mu_\alpha : \alpha \in A\}$ is a λ -family.

LEMMA 2.3. If $\{\mu_\alpha, \alpha \in A\}$ is conditionally compact then conditions (a), (b), (c), and (d) of Theorem 2.2 hold.

Proof. That (d) is true follows from Lemma 2.2. Let $\varepsilon > 0$ be given. Since $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact there exists a compact set K such that $\mu_\alpha(K) > 1 - \varepsilon/2$ for every $\alpha \in A$. Now

$$\begin{aligned} \operatorname{Re} (1 - \phi(x, \mu_\alpha)) &= \operatorname{Re} \left[\int_{l_p} [1 - \exp \{i(x, y)\}] d\mu_\alpha(y) \right] \\ &= \int_{l_p} [1 - \cos(x, y)] d\mu_\alpha(y) \\ &\leq \frac{1}{2} \int_K (x, y)^2 d\mu_\alpha(y) + \varepsilon. \end{aligned}$$

Let $T_{\alpha,\varepsilon} = (t_{ij}^{(\alpha,\varepsilon)})$ be the operator given by the matrix with $t_{ij}^{(\alpha,\varepsilon)} = \frac{1}{2} \int_K x_i x_j d\mu_\alpha(x)$. Clearly $T_{\alpha,\varepsilon}$ is positive definite and symmetric. Hence $T_{\alpha,\varepsilon}$ is an S_p -operator from l_p^* into l_p since

$$\begin{aligned} \sum_{i=1}^\infty (t_{ii}^{(\alpha,\varepsilon)})^{p/2} &= \sum_{i=1}^\infty \left(\frac{1}{2} \int_K x_i^2 d\mu_\alpha(x) \right)^{p/2} \leq \sum_{i=1}^\infty \int_K |x_i|^p d\mu_\alpha(x) \\ (2.4) \quad &= \int_K \|x\|^p d\mu_\alpha(x) \leq \sup_{x \in K} \|x\|^p \end{aligned}$$

and the last integral is finite because K is a compact subset of l_p . Hence (a) holds. Since the norm of $T_{\alpha,\varepsilon}$ is dominated by

$$\left[\sum_{i=1}^{\infty} (t_{ii}^{(\alpha,\varepsilon)})^{p/2} \right]^2 \leq \left[\sup_{x \in K} \|x\|^p \right]^2$$

we have (b) and (c) following from (2.4).

LEMMA 2.4. *If $\{\mu_\alpha : \alpha \in A\}$ is a λ -family of probability measures on l_p for some λ in the positive cone of $l_{p/2}^*$ then $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact if:*

(i) *For every integer N the Fourier transforms $\phi(P_N x, \mu_\alpha)$ are equicontinuous at zero in $P_N l_p^*$.*

(ii) $\lim_N \sup_\alpha \lim_k J_{N,k}[1 - \operatorname{Re} \phi(x, \mu_\alpha)] = 0$ where

$$J_{N,k}[\cdots] = \int_{(P_{N+k} - P_N)l_p^*} [\cdots] \lambda(N, k, dx)$$

and $\lambda(N, k, \cdot)$ is the Gaussian product measure on $(P_{N+k} - P_N)l_p^*$ with each coordinate x_i , $N+1 \leq i \leq N+k$, being Gaussian with mean zero and variance λ_i .

Proof. Since $\phi(\cdot, \mu_\alpha)$ is the Fourier transform of μ_α and $\lambda(N, k, \cdot)$ is symmetric about zero it follows that

$$J_{N,k}[1 - \operatorname{Re} \phi(x, \mu_\alpha)] = \int_{l_p} \left[1 - \exp \left\{ -\frac{1}{2} \sum_{i=N+1}^{N+k} \lambda_i x_i^2 \right\} \right] d\mu_\alpha(x).$$

Thus (ii) implies that

$$(2.5) \quad \lim_N \sup_\alpha \int_{l_p} \left[1 - \exp \left\{ -\frac{1}{2} \sum_{N+1}^{\infty} \lambda_i x_i^2 \right\} \right] d\mu_\alpha(x) = 0.$$

Let $0 < \delta < 1$, $\varepsilon > 0$ be given and let $E_N = \{x \in l_p : \sum_{N+1}^{\infty} \lambda_i x_i^2 < \delta/2\}$. Since $t/2 \leq 1 - e^{-t}$, $0 \leq t \leq 1$, it follows that

$$\begin{aligned} \mu_\alpha(E_N) &= 1 - \int_{E_N^c} d\mu_\alpha \geq 1 - \frac{8}{\delta} \int_{E_N^c} \left[1 - \exp \left\{ -\frac{1}{2} \sum_{i=N+1}^{\infty} \lambda_i x_i^2 \right\} \right] d\mu_\alpha(x) \\ &\geq 1 - \frac{8}{\delta} \int_{l_p} \left[1 - \exp \left\{ -\frac{1}{2} \sum_{N+1}^{\infty} \lambda_i x_i^2 \right\} \right] d\mu_\alpha(x) \geq 1 - \frac{\varepsilon}{3} \end{aligned}$$

for all $\alpha \in A$ and all N sufficiently large due to (2.5). Since $\{\mu_\alpha : \alpha \in A\}$ is a λ -family we know there exists an M such that for all $\alpha \in A$

$$(2.6) \quad \mu_\alpha \left\{ x \in l_p : \sum_{M+1}^{\infty} |x_i|^p < \frac{\delta}{2} \right\} \geq 1 - \frac{2\varepsilon}{3}.$$

By (i) we have $x_1, \dots, x_L \in P_M(l_p)$ such that

$$\mu_\alpha \left(P_M^{-1} \left(\bigcup_{k=1}^L S \left(x_k, \frac{\delta}{2} \right) \right) \right) > 1 - \frac{\varepsilon}{3}$$

for all $\alpha \in A$ where $S(x, \gamma) = \{y \in l_p : \|x - y\|^p < \gamma\}$. Combining (2.6) with the above we see that for all $\alpha \in A$

$$\mu_\alpha \left\{ \bigcup_{k=1}^L S(x_k, \delta) \right\} > 1 - \varepsilon.$$

Thus $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact [7, p. 170] and the lemma is proved.

Proof of Theorem 2.2. If $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact, Lemmas 2.2 and 2.3 demonstrate that (a), (b), (c), and (d) hold. Now if (a), (b), and (c) hold then (a) and (b) imply (i) of Lemma 2.4 and (a) and (c) imply (ii) since

$$\begin{aligned} & \lim_N \sup_\alpha \lim_k J_{N,k}[(T_{\alpha,\varepsilon}x, x) + \varepsilon] \\ &= \varepsilon + \lim_N \sup_\alpha \lim_k \int_{(P_{N+k} - P_N)l_p^*} (T_{\alpha,\varepsilon}x, x) \lambda(N, k, dx) \\ &= \varepsilon + \lim_N \sup_\alpha \lim_k \sum_{i,j=N+1}^{N+k} \int_{(P_{N+k} - P_N)l_p^*} t_{ij}^{(\alpha,\varepsilon)} x_i x_j \lambda(N, k, dx) \\ &= \varepsilon + \lim_N \sup_\alpha \sum_{i=N+1}^{\infty} \lambda_i t_{ii}^{(\alpha,\varepsilon)}. \end{aligned}$$

That is, since $\lambda = (\lambda_1, \lambda_2, \dots)$ is an element in the positive cone of $l_{p/2}^*$ we have, by (c), that

$$\lim_N \sup_\alpha \sum_{i=N+1}^{\infty} \lambda_i t_{ii}^{(\alpha,\varepsilon)}.$$

and hence (ii) holds. Finally, (i), (ii), and (d) imply that $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact so Theorem 2.2 is proved.

We now turn to Bochner's theorem on l_p , $2 \leq p < \infty$.

THEOREM 2.3. *A function $\phi(x)$ defined on l_p^* is the Fourier transform of a probability measure on the Borel subsets of l_p , $2 \leq p < \infty$, if and only if*

- (i) $\phi(0) = 1$ and ϕ is positive definite on l_p^* .
- (ii) ϕ is continuous at zero in the τ_p -topology on l_p^* .
- (iii) The family of measures $\{\mu_n\}$ corresponding to $\phi(P_n(\cdot))$ is a λ -family for some λ in the positive cone of $l_{p/2}^*$.

Proof. If μ is a probability measure on l_p with Fourier transform ϕ then ϕ clearly satisfies (i). Let $\varepsilon > 0$ be given and choose a compact set K of l_p such that $\mu(K) > 1 - \varepsilon/2$. Then for $x \in l_p^*$

$$\begin{aligned} (2.7) \quad \int_{l_p} [1 - \cos(x, y)] d\mu(y) &\leq \frac{1}{2} \int_K (x, y)^2 d\mu(y) + \varepsilon \\ &\leq \frac{1}{2} \sum_{i,j=1}^{\infty} x_i x_j \int_K y_i y_j d\mu(y) + \varepsilon. \end{aligned}$$

Now let $T = (t_{ij})$ where $t_{ij} = \int_K y_i y_j d\mu(y)$. Then

$$\sum_{i=1}^{\infty} (t_{ii})^{p/2} = \sum_{i=1}^{\infty} \left(\int_K y_i^2 d\mu(y) \right)^{p/2} \leq \sum_{i=1}^{\infty} \int_K |y_i|^p d\mu(y) = \int_K \|y\|^p d\mu(y) < \infty$$

and this implies, using (2.7),

$$|1 - \phi(x)|^2 \leq 2[1 - \operatorname{Re} \phi(x)] \leq (Tx, x) + \varepsilon$$

for any $x \in l_p^*$. Thus ϕ is τ_p -continuous at zero and (ii) holds. We now need only show that $\{\mu_n\}$ satisfies (iii). Let f be any bounded continuous function on l_p and notice that $\mu_n(A) = \mu(P_n^{-1}(A))$ for every Borel set A . Thus

$$\lim_n \int_{l_p} f d\mu_n = \lim_n \int_{l_p} f(P_n(x)) d\mu(x) = \int_{l_p} f d\mu$$

and hence $\{\mu_n\}$ converges weakly to μ . By Lemma 2.2 $\{\mu_n\}$ satisfies (iii).

If (i), (ii), and (iii) hold we show that Theorem 2.2 is applicable to the sequence $\{\mu_n\}$. Let $\varepsilon > 0$ be given. Using (ii) there exists an S_p -operator T^ε on l_p^* such that

$$1 - \operatorname{Re} \phi(x) \leq (T^\varepsilon x, x) + \varepsilon.$$

Since $\phi(x, \mu_n) = \phi(P_n x)$ we see that

$$1 - \operatorname{Re} \phi(x, \mu_n) \leq (T^\varepsilon P_n x, P_n x) + \varepsilon = (T_n^\varepsilon x, x) + \varepsilon$$

for $n = 1, 2, \dots$, where $T_n^\varepsilon = P_n T^\varepsilon P_n$. Thus (a), (b), (c) of Theorem 2.2 hold and (iii) is equivalent to (d) so $\{\mu_n\}$ is a conditionally compact sequence of probability measures on l_p . Hence if $\{\mu_{n_k}\}$ is a convergent subsequence of $\{\mu_n\}$ converging to the measure μ on l_p then $\phi(x, \mu_{n_k}) = \phi(P_{n_k}(x))$ so $\lim_k \phi(x, \mu_{n_k}) = \phi(x)$. This implies the Fourier transform of μ is $\phi(x)$ and the theorem is proved.

We now consider some easy examples which indicate the independence of the conditions (i), (ii), and (iii) in Theorem 2.3. The first example shows that (i) and (iii) do not imply (ii). Let $2 \leq p < \infty$ and assume $\psi(x)$, $x \in l_p^*$, is the Fourier transform of a probability measure μ on l_p . Let $\phi(x) = \psi(x)$ if $x \in \bigcup_{k=1}^\infty P_k(l_p^*)$ and be zero elsewhere. Then ϕ satisfies (i) and since $\phi(P_n(\cdot)) = \psi(P_n(\cdot)) = \psi(\cdot, \mu_n)$ and $\{\mu_n\}$ converges weakly to μ it follows that $\{\mu_n\}$ is a λ -family. Furthermore, ϕ is not continuous at zero in the norm topology of l_p^* and hence not in the τ_p -topology so (ii) does not hold. The second example shows that (i) and (ii) do not imply (iii). Take $p = 4$ and consider the probability measure μ on l such that the coordinates are independent functions each with distribution

$$\mu\{y \in l : y_k = \pm 1\} = 1/k, \quad \mu\{y \in l : y_k = 0\} = 1 - 2/k.$$

Now $l_4^* = l_{4/3}$ so for $x \in l_{4/3}$ we have

$$\begin{aligned} \phi(x) &= \int_l \exp\{i(x, y)\} d\mu(y) = \prod_{k=1}^\infty \left[\int_l \exp\{ix_k y_k\} d\mu(y) \right]^{(*)} \\ (2.8) \qquad &= \prod_{k=1}^\infty \left[1 - \frac{2}{k} (1 - \cos x_k) \right] \end{aligned}$$

(*) This integration will be justified in the next section after Lemma 3.1.

and since $0 \leq 1 - \cos t \leq t^2/2$ we have

$$\sum_{k=1}^{\infty} \frac{[1 - \cos x_k]}{k} \leq \sum_{k=1}^{\infty} \frac{x_k^2}{k} < \infty$$

because $x \in l_{4/3} \subset l_2$. Thus ϕ satisfies (i) and is continuous in the norm topology on $l_{4/3}$. For $x \in P_N l_{4/3}$ we have

$$|\phi(x) - 1|^2 \leq 2 \int_l (1 - \cos(x, y)) d\mu(y) \leq \int_l (x, y)^2 d\mu(y) = \sum_{i=1}^N \frac{x_i^2}{i} \leq \sum_{i=1}^{\infty} \frac{x_i^2}{i}$$

and the last term is finite for $x \in l_{4/3} \subset l_2$. Let $T = (t_{ij})$ where $t_{ii} = 1/i$ and $t_{ij} = 0$ for $i \neq j$. Then $(Tx, x) = \sum_{i=1}^{\infty} x_i^2/i$ for $x \in l_{4/3}$ and $\sum_{i=1}^{\infty} (1/i)^{p/2} < \infty$, so T is an S_p -operator such that for $x \in P_N l_{4/3}$

$$|\phi(x) - 1|^2 \leq (Tx, x) \quad (N = 1, 2, \dots).$$

Now both sides are norm continuous functions on $l_{4/3}$ and since $\bigcup_{N=1}^{\infty} P_N l_{4/3}$ is dense in $l_{4/3}$ we have $|\phi(x) - 1| \leq (Tx, x)$ for all $x \in l_{4/3}$. Thus ϕ satisfies (ii) of Theorem 2.3. Using the Borel-Cantelli lemma it is clear that

$$\mu \left\{ y \in l : \sum_{k=1}^{\infty} |y_k|^p < \infty \right\} = 0$$

for all $p \geq 2$ and in particular for $p = 4$ so (iii) cannot hold.

This last example contradicts a conjecture of Ju. V. Prohorov and V. Sazonov. In view of the results of [10] and [11] they seem to imply in [8] that continuity in the S_p -topology would be sufficient for a positive definite function ϕ , with $\phi(0) = 1$, to be the Fourier transform of a probability measure on l_p . Our example shows that this is not enough.

3. For each p , $0 < p \leq 2$, we denote by λ_p the probability measure on the Borel subsets \mathcal{B} of l formed by taking the product measure on l such that the coordinate functions have independent symmetric stable laws with Fourier transform $\exp \{-\frac{1}{2}|t|^p\}$.

LEMMA 3.1. *If μ is a probability measure on the Borel sets \mathcal{C} of l_p , $0 < p \leq 2$, then the function*

$$(x, y) = \lim_N \sum_{k=1}^N x_k y_k$$

is a $\mathcal{B} \times \mathcal{C}$ -measurable function on $l \times l_p$ where $x = (x_1, x_2, \dots)$, $y = (y_1, y_2, \dots)$, and $(\lambda_p \times \mu)\{(x, y) | < \infty\} = 1$.

Proof. Let $F(x, y) = (x, y)$ if (x, y) exists and is finite, and be infinity otherwise. Then $F(x, y)$ is $\mathcal{B} \times \mathcal{C}$ -measurable (it is the limit of $\mathcal{B} \times \mathcal{C}$ measurable functions), and if $E = \{|F(x, y)| < \infty\}$ then we claim $(\lambda_p \times \mu)(E) = 1$. To see this notice that for each $y \in l_p$ we have $Z_1(x) = x_1 y_1, \dots, Z_k(x) = x_k y_k, \dots$ a sequence of independent

stable random variables such that

$$\int_I \exp \{itZ_k(x)\} d\lambda_p(x) = \exp \{-\frac{1}{2}|y_k t|^p\}.$$

Hence $\sum_{k=1}^{\infty} Z_k$ converges in distribution, and hence almost everywhere, to a stable random variable with Fourier transform

$$\exp \left\{ -\frac{1}{2} \sum_{k=1}^{\infty} |y_k|^p |t|^p \right\}.$$

Thus for fixed $y \in l_p$, $F(x, y)$ is finite for almost all $x \in I$. Since F is jointly measurable the set E is jointly measurable and

$$(\lambda_p \times \mu)(E) = \int_{l_p} \left[\int_I I_E(x, y) d\lambda_p(x) \right] d\mu(y) = 1$$

since the inner integral is one for all $y \in l_p$.

Henceforth we will use (x, y) in both the usual sense when $x, y \in l_2$ or $x \in l_p$ and $y \in l_q$, or in a "stochastic sense" such as defined in Lemma 3.1. The task of deciding which way the inner product is being used is trivial and hence will not always be mentioned. For example, the stochastic inner product used in (2.8) can be rigorously obtained by repeating the proof of Lemma 3.1, and the second equality is then only an application of the bounded convergence theorem.

If μ is a measure on the Borel sets of l_p , $0 < p \leq 2$, and

$$\tilde{\phi}(x) = \int_{l_p} \exp \{i(x, y)\} d\mu(y) \quad (x \in I)$$

then $\tilde{\phi}$ is a Borel measurable function on I which is finite almost everywhere with respect to the measure λ_p and which is equal to

$$\phi(x) = \int_{l_p} \exp \{i(x, y)\} d\mu(y)$$

for all $x \in l_p^*$. Thus $\tilde{\phi}$ is an extension of ϕ from l_p^* to I . When dealing with $\tilde{\phi}$ we will always choose a version which agrees with ϕ on l_p^* and we will refer to $\tilde{\phi}$ as an *extended Fourier transform*.

We now state the continuity theorem for l_p , $0 < p \leq 2$.

THEOREM 3.1. *If $0 < p \leq 2$ and $\{\mu_k\}$ is a sequence of probability measures on l_p with Fourier transforms $\{\phi_k\}$ defined on l_p^* , then $\{\mu_k\}$ converges weakly to a measure μ with Fourier transform ϕ if and only if $\tilde{\phi}_k$ converges in probability to $\tilde{\phi}$ with respect to the measure λ_p and $\{\phi_k\}$ converges to ϕ on l_p^* .*

Actually, Theorem 3.1 is not quite equivalent to L. Gross' result (the case $p=2$), and in §4 we will obtain the equivalent result for $p=2$. We wish to point out that our proof will depend only on the stochastic inner product we have defined, and

the great smoothness of τ -continuous functions (see Lemma 4.2). This contrasts very much with the highly analytical proof given by Gross. We first proceed with several lemmas and the proof of Theorem 3.1. The first lemma uses an idea from J. Feldman's [2].

LEMMA 3.2. *If $\{\mu_\alpha : \alpha \in A\}$ is a family of probability measures on l_p , $0 < p \leq 2$ such that*

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \sup_{\alpha} J_{N,\gamma}(\mu_\alpha) = 0$$

where

$$J_{N,\gamma}(\mu_\alpha) = \int_{l_p} [1 - \exp \{-\frac{1}{2} \|\gamma P_N x + Q_N x\|^p\}] d\mu_\alpha(x)$$

then $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact.

Proof. Let $\varepsilon > 0$, $0 < \delta \leq 1$ be given and define

$$E = \{x \in l_p : \|\gamma P_N x + Q_N x\| \geq \delta/2\}.$$

If $0 \leq t \leq 1$ then $t/2 \leq 1 - e^{-t}$ so we find

$$\mu_\alpha(E) \leq 4 \left(\frac{2}{\delta}\right)^p \int_{l_p} [1 - \exp \{-\frac{1}{2} \|\gamma P_N x + Q_N x\|^p\}] d\mu_\alpha(x) = \frac{2^{p+2}}{\delta^p} J_{N,\gamma}(\mu_\alpha).$$

By our condition there exists $N_0(\varepsilon, \delta)$ and $\gamma_0(\varepsilon, \delta)$ such that $N \geq N_0$, $\gamma \leq \gamma_0$ implies

$$J_{N,\gamma}(\mu_\alpha) \leq \varepsilon(\delta^p/2^{p+2}) \quad (\alpha \in A).$$

Thus

$$\mu_\alpha\{x \in l_p : \|\gamma P_N x + Q_N x\| < \delta/2\} \geq 1 - \varepsilon$$

for $\gamma \leq \gamma_0$, $N \geq N_0$, and all μ_α . Further, $x \in E^c$ implies $\|P_N x\| < (\delta/2\gamma)$, so we let x_1, \dots, x_r be in $P_N l_p$ such that $\|x_j\| < \delta/2\gamma$ and such that for all $x \in l_p$ with $\|P_N x\| < \delta/2\gamma$ we have

$$\min_{1 \leq j \leq r} \|P_N x - x_j\| < \delta/2.$$

Then $\mu_\alpha\{\bigcup_{j=1}^r S(x_j, \delta)\} \geq 1 - \varepsilon$ for all μ_α where

$$S(x, \delta) = \{y \in l_p : \|y - x\| < \delta\}.$$

That is,

$$S(x_j, \delta) = \{y \in l_p : \|y - x_j\| < \delta\} \supseteq \{y \in l_p : \|P_N y - x_j\| < \delta/2, \|Q_N y\| < \delta/2\}$$

and hence $E^c \subseteq \bigcup_{j=1}^r S(x_j, \delta)$. Thus $\{\mu_\alpha : \alpha \in A\}$ is conditionally compact.

LEMMA 3.3. *If $\{\mu_k\}$ is a sequence of probability measures on l_p such that $\{\hat{f}_k\}$ converges in λ_p -measure to $\hat{\phi}$ where ϕ is the Fourier transform of a measure μ in l_p , then*

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \sup_k J_{N,\gamma}(\mu_k) = 0.$$

Proof. First of all observe that

$$\begin{aligned} J_{N,\gamma}(\mu_k) &= \int_{l_p} [1 - \exp \{-\frac{1}{2} \|\gamma P_N y + Q_N y\|^p\}] d\mu_k(y) \\ &= \int_l [1 - \tilde{\phi}_k(\gamma P_N x + Q_N x)] d\lambda_p(x). \end{aligned}$$

Thus

$$\lim_{k \rightarrow \infty} J_{N,\gamma}(\mu_k) = \int_l [1 - \tilde{\phi}(\gamma P_N x + Q_N x)] d\lambda_p(x)$$

since $\{\tilde{\phi}_k\}$ converging in λ_p -measure to $\tilde{\phi}$ implies $\{\tilde{\phi}_k \circ T_{N,\gamma}\}$ converges in measure to $\tilde{\phi} \circ T_{N,\gamma}$ where $T_{N,\gamma}(x) = \gamma P_N x + Q_N x$. Hence

$$\begin{aligned} \lim_{N \rightarrow \infty; \gamma \downarrow 0} \sup_k J_{N,\gamma}(\mu_k) &= \lim_{N \rightarrow \infty; \gamma \downarrow 0} \int_l [1 - \tilde{\phi}(\gamma P_N x + Q_N x)] d\lambda_p(x) \\ &= \lim_{N \rightarrow \infty; \gamma \downarrow 0} \int_l \int_{l_p} [1 - \exp \{i(\gamma P_N x + Q_N x, y)\}] d\mu(y) d\lambda_p(x) \\ &= \lim_{N \rightarrow \infty; \gamma \downarrow 0} \int_{l_p} [1 - \exp \{-\frac{1}{2} \|\gamma P_N y + Q_N y\|^p\}] d\mu(y) = 0 \end{aligned}$$

as was to be shown.

Proof of Theorem 3.1. We first assume $\{\mu_k\}$ converges weakly to μ on l_p . Then $\lim_k \phi_k(x) = \phi(x)$ on l_p^* and

$$\int_l |\tilde{\phi}_k - \tilde{\phi}|^2 d\lambda_p = \int_l |\tilde{\phi}_k|^2 d\lambda_p - \int_l \tilde{\phi}_k \tilde{\psi} d\lambda_p - \int_l \tilde{\psi}_k \tilde{\phi} d\lambda_p + \int_l |\tilde{\phi}|^2 d\lambda_p$$

where $\tilde{\psi}, \tilde{\psi}_k$ are the conjugates of $\tilde{\phi}$ and $\tilde{\phi}_k$. Furthermore, $\{\mu_k\}$ converging weakly to μ implies $\{\mu_k \times \mu_k\}$ converges weakly to $\mu \times \mu$ and hence

$$\begin{aligned} \lim_k \int_l |\tilde{\phi}_k|^2 d\lambda_p &= \lim_k \int_l \int_{l_p} e^{i(x,y)} d\mu_k(y) \cdot \int_{l_p} e^{-i(x,z)} d\mu_k(z) d\lambda_p(x) \\ &= \lim_k \int_{l_p} \int_{l_p} \int_l e^{i(x,y-z)} d\lambda_p(x) d\mu_k(y) d\mu_k(z) \\ &= \lim_k \int_{l_p} \int_{l_p} \exp(-\frac{1}{2} \|y-z\|^p) d\mu_k(y) d\mu_k(z) \\ &= \int_{l_p} \int_{l_p} \exp(-\frac{1}{2} \|y-z\|^p) d\mu(y) d\mu(z) \\ &= A. \end{aligned}$$

Similarly, $\int_l \phi_k \tilde{\psi} d\lambda_p$ and $\int_l \tilde{\psi}_k \phi d\lambda_p$ both converge to A as k approaches infinity. Thus we have $\{\tilde{\phi}_k\}$ converging in mean-square to $\tilde{\phi}$, and hence $\{\tilde{\phi}_k\}$ converges to $\tilde{\phi}$ in λ_p -measure.

On the other hand, if $\{\tilde{\phi}_k\}$ converges to $\tilde{\phi}$ in λ_p -measure on l where ϕ is the Fourier transform of a probability measure on l_p we know, by applying Lemmas 3.2 and 3.3, that $\{\mu_k\}$ is conditionally compact. Hence there is a subsequence $\{\mu_{k_j}\}$ converging weakly to a probability measure ν with Fourier transform ψ . Then $\psi = \lim_j \phi_{k_j} = \phi$

and hence $\{\mu_{k_i}\}$ converges weakly to μ by the uniqueness of the Fourier transforms for measure on l_p . Furthermore, this shows that any convergent subsequence of $\{\mu_{k_i}\}$ must converge to μ and hence $\{\mu_{k_i}\}$ actually converges to μ because every subsequence has a convergent subsequence going weakly to μ .

4. We now turn to the special case $p=2$ and prove the continuity theorem given by L. Gross in [5]. As mentioned previously, our proof mainly depends on the stochastic inner product and the smoothness of τ -continuous functions on l_2 . This is in contrast to the analytical approach of L. Gross [5] and J. Feldman [2]. The Gaussian measure λ_2 on l will be denoted by $P(\cdot)$ throughout this section.

LEMMA 4.1. *Let T be an S -operator on l_2 . Then $T(x) = \sum_i \gamma_i(x, \alpha_i)\alpha_i$ where $\gamma_i > 0$, $\sum_i \gamma_i < \infty$, $\{\alpha_1, \alpha_2, \dots\}$ is an orthonormal family, and*

$$(Tx, x) = \sum_i \gamma_i(x, \alpha_i)^2$$

is finite on a linear subset of l of P -measure one. Furthermore, if

$$Z_N(x) = (TQ_Nx, Q_Nx)$$

then $\lim_N Z_N(x) = 0$ on some linear subset \mathcal{E} of l such that $P(\mathcal{E}) = 1$.

Proof. Since T is an S -operator the given representation for T is well known. That (Tx, x) is finite on a subset of measure one follows since $\{(x, \alpha_i)\}$ is a sequence of independent Gaussian functionals with mean zero and variance one, and since $\sum_i \gamma_i < \infty$. The linearity of this subset follows since

$$(T(x+y), x+y) \leq \{(Tx, x)^{1/2} + (Ty, y)^{1/2}\}^2.$$

The linearity of the subset of l where $\lim_N Z_N(x) = 0$ follows in the same way. Hence all that remains is to show that $\lim_N Z_N(x) = 0$ for almost all $x \in l$. First we observe that

$$E(Z_N(x)) = \sum_i \gamma_i E(Q_Nx, \alpha_i)^2 = \sum_i \gamma_i (Q_N\alpha_i, \alpha_i)$$

so $\lim_N E(Z_N(x)) = 0$.

Furthermore, $\{Z_N\}_N$ is a reversed submartingale so the convergence to zero almost everywhere (P) follows. That is, for all $A \in \mathcal{B}(Z_{N+1}, Z_{N+2}, \dots)$

$$\begin{aligned} \int_A E(Z_N | Z_{N+1}, Z_{N+2}, \dots) dP &= \int_A Z_N dP = \int_A \sum_i \gamma_i (Q_Nx, \alpha_i)^2 dP \\ &= \sum_i \int_A \gamma_i [(\mathcal{P}_{N+1}x, \alpha_i) + (Q_{N+1}x, \alpha_i)]^2 dP \\ &= \sum_i \gamma_i \int_A \{(\mathcal{P}_{N+1}x, \alpha_i)^2 + (Q_{N+1}x, \alpha_i)^2\} dP \\ &\geq \sum_i \gamma_i \int_A (Q_{N+1}x, \alpha_i)^2 dP \\ &= \int_A Z_{N+1} dP \end{aligned}$$

where the equality preceding the inequality follows since $(\mathcal{P}_{N+1}x, \alpha_i)$ and $(Q_{N+1}x, \alpha_i)$ are independent with mean zero and $A \in \mathcal{B}(Z_{N+1}, Z_{N+2}, \dots)$. Thus

$$E(Z_N \mid Z_{N+1}, Z_{N+2}, \dots) \geq Z_{N+1}$$

so $\{Z_N\}_N$ is a reversed submartingale [6, p. 393].

LEMMA 4.2. *If $\phi(x)$ is uniformly τ -continuous on l_2 then $\phi(x)$ can be uniquely extended, call the extension $\tilde{\phi}$, to be uniformly continuous on a linear subset \mathcal{E} of l such that*

- (i) $P(\mathcal{E}) = 1$.
- (ii) For $x \in \mathcal{E}$ we have $\tilde{\phi}(x) = \lim_N \phi(P_N x)$.
- (iii) $\lim_{N \rightarrow \infty; \gamma \rightarrow 0} \tilde{\phi}(\gamma P_N x + Q_N x) = \tilde{\phi}(0) = \phi(0)$ for $x \in \mathcal{E}$.

Proof. Since $\phi(x)$ is uniformly τ -continuous there exists, for each integer k , an S -operator T_k such that if $E_k = \{x \in l_2 : (T_k x, x) < 1\}$ then

$$|\phi(y+x) - \phi(y)| < 1/k \quad (x \in E_k, y \in l_2).$$

Choose at least one T_k such that $(T_k x, x) > 0$ unless $x = 0$. Let

$$\mathcal{E}_k = \left\{ x \in l : (T_k Q_N x, Q_N x) < \infty \text{ for } N = 0, 1, 2, \dots; \lim_N (T_k Q_N x, Q_N x) = 0 \right\}$$

for $k = 1, 2, \dots$

Then $P(\mathcal{E}_k) = 1$ and \mathcal{E}_k is a linear subset of l . If $\mathcal{E} = \bigcap_{k=1}^{\infty} \mathcal{E}_k$ then $P(\mathcal{E}) = 1$ and \mathcal{E} is also linear. We define a topology on \mathcal{E} by taking as a base \mathcal{S} translates of sets of the type

$$\mathcal{U} = \{x \in \mathcal{E} : ((T_{i_1} + \dots + T_{i_r})x, x) < r\}$$

where $0 < r < \infty$ and $\{i_1, \dots, i_r\}$ runs over finite subsets of the integers. That \mathcal{S} is actually a base is easily verified. By Lemma 4.1 it follows that l_2 is dense in \mathcal{E} , and since $\phi(x)$ is uniformly continuous on l_2 with respect to the topology induced by \mathcal{S} it follows that $\phi(\cdot)$ has the unique uniformly continuous extension $\tilde{\phi}(x) = \lim_N \phi(P_N x)$ for $x \in \mathcal{E}$. Furthermore, property (iii) holds for $\tilde{\phi}$ since

$$\lim_N (T_k Q_N x, Q_N x) = 0$$

for $k = 1, 2, \dots$ and $x \in \mathcal{E}$.

Now if ϕ is the Fourier transform of a probability measure μ on l_2 then ϕ is uniformly τ -continuous and hence by Lemma 4.2 ϕ can be extended to be uniformly continuous on a linear subset \mathcal{E} of l such that $P(\mathcal{E}) = 1$. However, we have considered another extension of ϕ which we called the extended Fourier transform of μ . In this sense $\tilde{\phi}$ was defined to be

$$\tilde{\phi}(x) = \int_{l_2} \exp \{i(x, y)\} d\mu(y) \quad (x \in l).$$

The next lemma shows that these two definitions agree with probability one.

LEMMA 4.3. Let ϕ be the Fourier transform of a probability measure μ on l_2 . If $\tilde{\phi}$ and \mathcal{E} are as in Lemma 4.2, and if

$$\psi(x) = \int_{l_2} \exp \{i(x, y)\} d\mu(y) \quad (x \in l),$$

then $\psi(x) = \tilde{\phi}(x)$ on a linear subset V of l such that $P(V) = 1$.

Proof. First of all we see that $\lim_N (P_N x, y) = \lim_N (x, P_N y) = (x, y)$ almost everywhere ($P \times \mu$) so by the bounded convergence theorem $\lim_N \phi(P_N x) = \psi(x)$ on a linear set E of l such that $P(E) = 1$. However, by Lemma 4.2 $\tilde{\phi}(x) = \lim_N \phi(P_N x)$ for all $x \in \mathcal{E}$ where $P(\mathcal{E}) = 1$. Thus $\psi(x) = \tilde{\phi}(x)$ for all $x \in V = E \cap \mathcal{E}$ and, clearly, V is a linear subset of l such that $P(V) = 1$.

The next lemma is contained in the work of L. Gross [5] but for the sake of completeness we include its proof.

LEMMA 4.4. If μ and ν are probability measures on l_2 then $\mu = \nu$ if and only if their extended Fourier transforms $\tilde{\phi}$ and $\tilde{\psi}$ are equal with probability one on l .

Proof. If $\mu = \nu$ it is obvious that $\tilde{\phi} = \tilde{\psi}$ almost everywhere (P) on l . The converse goes as follows. If $\mu \neq \nu$ and ϕ and ψ are the usual Fourier transforms of μ and ν , then there exists an $x_0 \in l_2$ such that $\phi(x_0) \neq \psi(x_0)$. Let $h = \phi - \psi$ and assume without loss of generality that $h(x_0) = 2a > 0$. Let

$$I = \{x \in l_2 : h(x) > a\}.$$

Now h is τ -continuous on l_2 so there exists an S -operator T such that if

$$H = \{x \in l_2 : T(x - x_0, x - x_0) < 1\}$$

then $H \subseteq I$. If $T(x, x) = \sum_i \gamma_i(x, \alpha_i)^2$ where $\{\alpha_1, \alpha_2, \dots\}$ are orthonormal, $\gamma_i \geq 0$, and $\sum_i \gamma_i < \infty$ we define

$$\tilde{H} = \left\{x \in l : \sum_i \gamma_i(x - x_0, \alpha_i)^2 < 1\right\}.$$

Since

$$\tilde{h}(x) = \tilde{\phi}(x) - \tilde{\psi}(x) = \lim_N [\phi(P_N x) - \psi(P_N x)] = \lim_N h(P_N x)$$

it follows that $h(x) = \tilde{\phi}(x) - \tilde{\psi}(x) \geq a > 0$ for almost all $x \in \tilde{H}$. This is a contradiction if $P(\tilde{H}) > 0$. Now

$$\begin{aligned} & P\left\{x \in l : \sum_i \gamma_i(x - x_0, \alpha_i)^2 < 1\right\} \\ & \geq P\left\{\sum_{i=1}^N \gamma_i(x - x_0, \alpha_i)^2 < \frac{1}{2} \text{ and } \sum_{i=N+1}^{\infty} \gamma_i(x - x_0, \alpha_i)^2 < \frac{1}{2}\right\} \\ & = P\left\{\sum_{i=1}^N \gamma_i(x - x_0, \alpha_i)^2 < \frac{1}{2}\right\} P\left\{\sum_{i=N+1}^{\infty} \gamma_i(x - x_0, \alpha_i)^2 < \frac{1}{2}\right\} \\ & \geq P\left\{\sum_{i=1}^N \gamma_i(x - x_0, \alpha_i)^2 < \frac{1}{2}\right\} \left[1 - 2 \sum_{i=N+1}^{\infty} \gamma_i\right] \end{aligned}$$

where the equality holds because $\{(x - x_0, \alpha_i)\}$ is a sequence of independent Gaussian random variables and the last inequality follows by Čebyšev's inequality and the fact that $E[(x - x_0, \alpha_i)^2] = 1$. Hence for N such that $[1 - 2 \sum_{i=N+1}^{\infty} \gamma_i]$ is positive, we have the last quantity as a product of two positive numbers and we see that $P(\tilde{H}) > 0$. This is a contradiction so $\mu = \nu$ as was to be proved.

We now prove Gross' result for l_2 .

THEOREM 4.1. *If $\{\mu_k\}$ is a sequence of probability measures on l_2 with Fourier transforms $\{\phi_k\}$, then $\{\mu_k\}$ converges weakly to a measure μ if and only if $\{\tilde{\phi}_k\}$ converges in probability to $\tilde{\phi}$ for some τ -continuous ϕ on l_2 such that $\phi(0) = 1$.*

Proof. If $\{\mu_k\}$ converges weakly to μ then $\{\phi_k\}$ converges in probability to ϕ , the extended Fourier transform of μ , by the first part of Theorem 3.1. Furthermore, $\phi(0) = 1$ and ϕ is τ -continuous on l_2 . Conversely, we assume $\{\tilde{\phi}_k\}$ converges to $\{\tilde{\phi}\}$ in probability, where $\phi(0) = 1$ and ϕ is τ -continuous on l_2 (here, of course, $\tilde{\phi}$ is the continuous extension of ϕ as given in Lemma 4.2). Let $J_{N,\gamma}(\mu_k)$ be as in Lemma 3.3 with $p = 2$. We now verify that

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \sup_k J_{N,\gamma}(\mu_k) = 0$$

under the present hypothesis. In fact, the proof proceeds exactly as in Lemma 3.3 except that

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \int_l [1 - \tilde{\phi}(\gamma P_N x + Q_N x)] dP(x) = 0$$

because $\tilde{\phi}$ is a continuous function on a linear subset of l of probability one and by Lemma 4.2

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \tilde{\phi}(\gamma P_N x + Q_N x) = \tilde{\phi}(0) = \phi(0) = 1.$$

Thus by Lemma 3.2 $\{\mu_k\}$ is conditionally compact, and proceeding as in Theorem 3.1 (along with Lemma 4.4) we see $\{\mu_k\}$ converges weakly to a measure μ with extended Fourier transform $\tilde{\phi}$.

5. The continuity theorem on l_p , $2 \leq p < \infty$, involves the concept of a λ -family of measures as defined in §2.

If λ is in the positive cone of $l_{p/2}^*$ we will denote by P_λ the probability measure on l which has the coordinate functions as independent Gaussian random variables with mean zero and $E(\mathcal{P}_k x)^2 = \lambda_k$. For $p = 2$ we have $l_{p/2}^* = l_\infty$ and we can choose $\lambda = (1, 1, \dots)$ so P_λ then denotes the canonical Gaussian distribution on l used in §§3 and 4 and in [3], [4], [5].

If μ is a probability measure on l_p with Fourier transform ϕ defined on l_p^* , then the P_λ -extended Fourier transform is defined on l as follows:

$$\tilde{\phi}(x) = \int_{l_p} \exp \{i(x, y)\} d\mu(y) \quad (x \in l).$$

Using the ideas of Lemma 3.1 and the fact that $\lambda \in l_{p/2}^*$ it is easy to see that $\tilde{\phi}(x)$ is a measurable function on l . When dealing with a version of $\tilde{\phi}$ we will always assume that it agrees with ϕ on l_p^* .

THEOREM 5.1. *Let $\{\mu_k\}$ be a sequence of probability measures on l_p , $2 \leq p < \infty$, with Fourier transforms $\{\phi_k\}$. Then $\{\mu_k\}$ converges weakly to a measure μ with Fourier transform ϕ if and only if $\{\mu_k\}$ is a λ -family for some λ in the positive cone of $l_{p/2}^*$, $\{\tilde{\phi}_k\}$ converges in P_λ -measure to $\tilde{\phi}$, and $\{\phi_k\}$ converges to ϕ on l_p^* .*

Proof. Suppose $\{\mu_k\}$ converges weakly to μ . Then $\{\mu_k\}$ is conditionally compact and by Lemma 2.2 it is a λ -family for any λ in the positive cone of $l_{p/2}^*$. Repeating the argument given in the proof of Theorem 3.1 it follows that $\{\tilde{\phi}_k\}$ converges in mean-square to $\tilde{\phi}$ and hence $\{\tilde{\phi}_k\}$ converges in P_λ -probability to $\tilde{\phi}$. That $\{\phi_k\}$ converges to ϕ on l_p^* is easy.

Now assume $\{\tilde{\phi}_k\}$ converges in P_λ -probability to $\tilde{\phi}$. If

$$\begin{aligned} H_{N,\gamma}(\mu_k) &= \int_{l_p} \left[1 - \exp \left\{ -\frac{1}{2} \left[\sum_{j=1}^N \lambda_j \gamma^2 y_j^2 + \sum_{j=N+1}^\infty \lambda_j y_j^2 \right] \right\} \right] d\mu_k(y) \\ &= \int_l [1 - \tilde{\phi}_k(\gamma P_N x + Q_N x)] dP_\lambda(x) \end{aligned}$$

then

$$\begin{aligned} \lim_{k \rightarrow \infty} H_{N,\gamma}(\mu_k) &= \int_l [1 - \tilde{\phi}(\gamma P_N x + Q_N x)] dP_\lambda(x) \\ &= \int_{l_p} \left[1 - \exp \left\{ -\frac{1}{2} \left[\sum_{j=1}^N \lambda_j \gamma^2 y_j^2 + \sum_{j=N+1}^\infty \lambda_j y_j^2 \right] \right\} \right] d\mu(y) \end{aligned}$$

and hence

$$\lim_{N \rightarrow \infty; \gamma \downarrow 0} \sup_k H_{N,\gamma}(\mu_k) = 0.$$

Thus for $\varepsilon, \delta > 0$ there exists $\gamma_0 > 0$, N_0 such that $\gamma \leq \gamma_0$ and $N \geq N_0$ implies

$$\mu_k \left\{ x \in l_p : \gamma^2 \sum_{j=1}^N \lambda_j x_j^2 + \sum_{j=N+1}^\infty \lambda_j x_j^2 < \delta \right\} \geq 1 - \varepsilon$$

for $k = 1, 2, \dots$ (see Lemma 3.2 for this type of argument). Since $\{\mu_k\}$ is a λ -family we thus have for γ sufficiently small (γ may depend on N) that

$$\mu_k \{x \in l_p : \|\gamma P_N x + Q_N x\|^p < \delta\} \geq 1 - (2\varepsilon + \varepsilon_N)$$

where $\lim_N \varepsilon_N = 0$. Then, as in Lemma 3.2, we see that $\{\mu_k\}$ is conditionally compact and since $\{\phi_k\}$ converges to ϕ on l_p^* . We have $\{\mu_k\}$ converging weakly to μ .

As a final remark we mention that using the above techniques it is possible to prove a central limit theorem for independent identically distributed random variables in l_p , $0 < p < \infty$. In the case $1 \leq p \leq 2$ certain results are given in [7] and [11].

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