

## SUPPORTS OF CERTAIN INFINITELY DIVISIBLE PROBABILITY MEASURES ON LOCALLY CONVEX SPACES<sup>1</sup>

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Let  $\mathbf{B}$  be a separable Banach space and let  $\mu$  be a centered Poisson probability measure on  $\mathbf{B}$  with Lévy measure  $M$ . Assume that  $M$  admits a polar decomposition in terms of a finite measure  $\sigma$  on the unit sphere of  $\mathbf{B}$  and a Lévy measure  $\rho$  on  $(0, \infty)$ . The main result of this paper provides a complete description of the structure of  $\mathcal{S}_\mu$ , the support of  $\mu$ . Specifically, it is shown that: (i) if  $\int_{(0,1]} s\rho(ds) = \infty$ , then  $\mathcal{S}_\mu$  is a linear space and is equal to the closure of the semigroup generated by  $\mathcal{S}_M$  (the support of  $M$ ) and the negative of the barycenter of  $\sigma$ ; and (ii) if  $\int_{(0,1]} s\rho(ds) < \infty$  and zero is in the support of  $\rho$ , then  $\mathcal{S}_\mu$  is a convex cone and is equal to the closure of the semigroup generated by  $\mathcal{S}_M$ . The result (i) yields an affirmative answer to the question, open for some time, of whether the support of a stable probability measure of index  $1 \leq \alpha < 2$  on  $B$  is a translate of a linear space. Analogs of these results, for both Poisson and stable probability measures defined on general locally convex spaces, are also provided.

**1. Introduction.** This paper provides a complete description of the structure of the supports of general (not necessarily symmetric) Poisson probability measures [defined on general locally convex (l.c.) spaces] whose Lévy measures admit a polar-type decomposition. This work is inspired by and completes a result of de Acosta [3] and several results of Tortrat [10], [11].

The core result is Theorem 3.1(a). It shows that if the Lévy measure  $M$  of a centered Poisson probability measure  $\mu$  on a separable Banach space  $\mathbf{B}$  admits a polar decomposition in terms of a finite measure  $\sigma$  on the unit sphere of  $\mathbf{B}$  and a (Lévy) measure  $\rho$  on  $(0, \infty)$  which satisfies  $\int_{(0,1]} s\rho(ds) = \infty$ , then  $\mathcal{S}_\mu$ , the support of  $\mu$ , is a (closed) linear space and it is equal to the closure of the semigroup generated by  $\mathcal{S}_M$ , the support of  $M$ , and the negative of the barycenter of  $\sigma$ . This result completes and encompasses several results of Tortrat ([10], Theorems 2(i), 3, and 4 and Corollary 1; [11], Proposition 1 and Theorem 3'). It also settles two of his conjectures, which state that, under the hypotheses of Theorem 3.1(a),  $\mathcal{S}_\mu$  may fail to be a linear space in the infinite-dimensional case even when  $\mathbf{B}$  is a Hilbert space ([10], page 41) and that  $\mathcal{S}_\mu$  “presque sûr” appears to be a linear space in the finite-dimensional case ([11], page 294). We also prove a companion result to Theorem 3.1(a) [Theorem 3.1(b)]. It shows that if  $\int_{(0,1]} s\rho(ds) < \infty$  and if 0 belongs to the

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support of  $\rho$ , then  $\mathcal{S}_\mu$  is a translate of a (closed) convex cone and that this cone is equal to the closure of the semigroup generated by  $\mathcal{S}_M$ . In this result we also provide a necessary and sufficient condition in order for this cone to be a linear space. In addition, we provide, in Theorems 4.1(a) and 4.1(b), appropriate analogs of Theorems 3.1(a) and 3.1(b), respectively, for probability measures defined on general l.c. spaces.

Let  $\mu$  be an  $\alpha$ -stable,  $1 \leq \alpha < 2$ , probability measure on  $\mathbf{B}$ ; the question of whether  $\mathcal{S}_\mu$  is a translate of a linear space (without any restrictive hypotheses on the space or on the measure) has been open for some time (see [3], page 874, Theorem 5.2; [10], pages 38–39, Theorems 2 and 3; and [11], pages 294–295, Corollary and Remark). The core result and its analog in l.c. spaces yield two corollaries which answer this question in the affirmative for measures defined not only on  $\mathbf{B}$  but also on general l.c. spaces. These complete the results of de Acosta and Tortrat just noted as well as pertinent results obtained in [2], [5] and [6].

The methods of proof used are refinements of those exploited earlier in [4], [7] and [10–12]. The organization of the rest of the paper is as follows: Section 2 contains preliminaries. Section 3 contains the core result, its companion result and a few of their corollaries. Section 4 contains the analogs of results of Section 3 for measures defined on general l.c. spaces.

**2. Preliminaries.** Throughout,  $\mathbf{B}$  and  $\mathbf{E}$  will denote, respectively, a separable Banach space and a l.c. space; further,  $\mathbf{B}^*$  and  $\mathbf{E}^*$  will denote, respectively, the topological duals of  $\mathbf{B}$  and  $\mathbf{E}$ . For a subset  $A$  of  $\mathbf{B}$ , the closure of the linear space (resp., the convex cone) generated by  $A$  will be denoted by  $\mathbf{L}(A)$  [resp., by  $\mathbf{C}(A)$ ]. Unless stated otherwise, all measures on a topological space  $X$  are assumed to be defined on the Borel  $\sigma$ -algebra of  $X$ ; for a set  $A$  in  $X$ ,  $\bar{A}$  will denote the closure of  $A$ .

Let  $\nu$  be a finite or infinite measure on a separable metric space  $X$ . Then the *support* of  $\nu$ , denoted throughout by  $\mathcal{S}_\nu$ , is, by definition, the intersection of all closed sets  $F$  with  $\nu(F^c) = 0$ , where, throughout, for a set  $A$ ,  $A^c$  denotes the complement of  $A$ . Clearly,  $\mathcal{S}_\nu = \{x \in X: \nu(V) > 0, \text{ for every open neighborhood } V \text{ of } x\}$ . If  $X = \mathbf{B}$ , then the *linear support* of  $\nu$ , denoted throughout by  $\mathcal{L}\mathcal{S}(\nu)$ , is the intersection of all closed linear subspaces  $G$  of  $\mathbf{B}$  with  $\nu(G^c) = 0$ . It is easy to see that

$$(2.1) \quad \mathcal{L}\mathcal{S}(\nu) = \mathbf{L}(\mathcal{S}_\nu); \quad \mathcal{L}\mathcal{S}(\nu) = \mathcal{S}_\nu, \text{ if } \mathcal{S}_\nu \text{ is linear.}$$

Let  $M$  be a finite or infinite measure on  $\mathbf{B}$ . If  $M$  admits a polar decomposition, that is,  $M = (\sigma \times \rho) \circ \Psi^{-1}$ , where  $\Psi$  is the topological isomorphism from  $\partial\Delta \times (0, \infty)$  onto  $\mathbf{B} \setminus \{\theta\}$  defined by  $\Psi(u, t) = tu$ , and where  $\sigma$  and  $\rho$  are, respectively, a finite measure on  $\partial\Delta$  (the boundary of  $\Delta \equiv \{x \in \mathbf{B}: \|x\| \leq 1\}$ ) and a measure on  $(0, \infty)$  ( $\theta$  being the zero element of  $\mathbf{B}$ ), then we shall write, throughout,  $M = \Psi(\sigma \times \rho)$ .

Now let  $M = \Psi(\sigma \times \rho)$  be a Lévy measure [therefore,  $\rho$  is also Lévy on  $\mathbf{R}^+ \equiv (0, \infty)$ ]. Following [1], the  $\tau$ -centered Poisson probability measure on  $\mathbf{B}$  with Lévy measure  $M$  will be denoted, throughout, by  $c_\tau\text{-Pois}(M)$  or by

$c_\tau$ -Pois( $\sigma, \rho$ ),  $0 \leq \tau \leq \infty$ . If  $\tau = 0$  (resp.,  $\tau = \infty$ ), it is assumed here and elsewhere that  $\int_\Delta \|x\| dM < \infty$  (resp.,  $\int_{\Delta^c} \|x\| dM < \infty$ ).

Let  $M = \Psi(\sigma \times \rho)$  be a Lévy measure on  $\mathbf{B}$ . The barycenter  $\int_{\partial\Delta} u \sigma(du)$  and, for a fixed  $\tau > 0$ , the function  $\lambda \mapsto \int_{(\lambda, \tau]} s \rho(ds)$ ,  $0 < \lambda < \tau$ , we shall denote, throughout, by  $a_\sigma$  ( $\equiv a$ ) and  $\varphi_{\tau, \rho}$  ( $\equiv \varphi_\tau$ ), respectively. Further, throughout, we shall use the notation  $\mathbf{I}_j$ ,  $j = 1, 2$ , to denote the following integral conditions on  $M$ :

$$(2.2) \quad \begin{aligned} \mathbf{I}_1: & \int_\Delta \|x\| dM = \infty \quad (\Leftrightarrow \varphi_{1, \rho}(0^+) = \infty); \\ \mathbf{I}_2: & \int_\Delta \|x\| dM < \infty \quad (\Leftrightarrow \varphi_{1, \rho}(0^+) < \infty). \end{aligned}$$

Finally, we shall use the notation  $\tilde{\sigma}$  ( $\equiv \tilde{\sigma}_a$ ) and  $\tilde{M}$  ( $\equiv \tilde{M}_a$ ) for the measures defined, respectively, by

$$(2.3) \quad \tilde{\sigma} = \begin{cases} \sigma, & \text{if } a = \theta, \\ \sigma + \|a\| \delta_{\{-a/\|a\|\}}, & \text{if } a \neq \theta, \end{cases}$$

$$\tilde{M} = (\tilde{\sigma} \times \rho) \circ \Psi^{-1},$$

where  $\delta_{\{ \cdot \}}$  is the Dirac measure. The Hahn–Banach theorem yields

$$(2.4) \quad a_\sigma \in \mathbf{C}(\mathcal{S}_\sigma), \quad \text{hence, by (2.3), } \mathbf{L}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_{\tilde{\sigma}}).$$

Therefore, if  $\mathbf{C}(\mathcal{S}_\sigma)$  is a linear space, then

$$(2.5) \quad \mathbf{C}(\mathcal{S}_\sigma) = \dot{\mathbf{L}}(\mathcal{S}_\sigma) \quad [= \mathbf{L}(\mathcal{S}_{\tilde{\sigma}}) = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})].$$

Now we state two preliminary results [Propositions 2.1(a) and 2.1(b)]; for a proof of (a) see [10], page 39; a proof of (b) can be provided using (2.1) and standard techniques. We will need one additional notation: For a measure  $\gamma$  on  $\mathbf{B}$ , following [10], page 34, we shall denote by  $\Sigma(\mathcal{S}_\gamma)$  the closure of the semigroup generated by  $\mathcal{S}_\gamma$  and  $\{\theta\}$ .

**PROPOSITION 2.1.** *Let  $M = \Psi(\sigma \times \rho)$  by a Lévy measure on  $\mathbf{B}$  and let  $\mu_\tau \equiv c_\tau$ -Pois( $\sigma, \rho$ ),  $0 \leq \tau \leq \infty$ .*

(a) *If  $0 < \tau \leq \infty$ , then, for any sequence  $\tau > \lambda_n \downarrow 0$ ,*

$$(2.6) \quad \mathcal{S}_{\mu_\tau} = \bigcap_{m \geq 1} \left[ \bigcup_{n \geq m} \{ \bar{A}_{\lambda_n} - a_\sigma \varphi_\tau(\lambda_n) \} \right] = \bigcap_{m \geq 1} \left[ \bigcup_{n \geq m} \{ A_{\lambda_n} - a_\sigma \varphi_\tau(\lambda_n) \} \right];$$

*if  $\tau = 0$ , then, for any sequence  $\lambda_n \downarrow 0$ ,*

$$(2.7) \quad \mathcal{S}_{\mu_0} = \bigcap_{m \geq 1} \left[ \bigcup_{n \geq m} \bar{A}_{\lambda_n} \right] = \bigcap_{m \geq 1} \left[ \bigcup_{n \geq m} A_{\lambda_n} \right],$$

*where  $A_{\lambda_n}$  is the semigroup generated by  $\mathcal{S}_{M/\{\|x\| \geq \lambda_n\}} \cup \{\theta\}$ .*

(b) For any  $0 \leq \tau \leq \infty$ , set  $\mathbf{B}_0(\mu_\tau) = \{x \in \mathbf{B}: y(x) = 0, \text{ for all } y \in B^* \text{ with } \mu_\tau \circ y^{-1} = \delta_{\{0\}}\}$ . Then

$$(2.8) \quad \begin{aligned} \mathbf{B}_0(\mu_\tau) &= \mathcal{L}\mathcal{S}(\mu_\tau) (= \mathbf{L}(\mathcal{S}_{\mu_\tau})) = \mathcal{L}\mathcal{S}(\sigma) (= \mathbf{L}(\mathcal{S}_\sigma)) \\ &= \mathcal{L}\mathcal{S}(M) (= \mathbf{L}(\mathcal{S}_M)). \end{aligned}$$

**3. The supports of Poisson probability measures on  $\mathbf{B}$ .** We first state our main theorem and several of its implications; then we state several lemmas necessary to prove the theorem. Next we present our proof of the theorem; the proofs of the lemmas are given after the proof of the theorem.

**THEOREM 3.1.** *Let  $M = \Psi(\sigma \times \rho)$  be a Lévy measure on  $\mathbf{B}$ , let  $0 \leq \tau \leq \infty$  and let  $\mu_\tau = c_\tau\text{-Pois}(M) [\equiv c_\tau\text{-Pois}(\sigma, \rho)]$ . Then the following hold:*

(a) *If  $M$  satisfies  $\mathbf{I}_1$  [see (2.2)], then  $\Sigma(\mathcal{S}_{\tilde{M}})$ ,  $\mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$  and  $\mathcal{S}_{\mu_\tau}$ ,  $0 < \tau \leq \infty$ , are linear spaces and*

$$(3.1) \quad \mathcal{S}_{\mu_\tau} = \Sigma(\mathcal{S}_{\tilde{M}}) = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}}) = \mathbf{B}_0(\mu_\tau) = \mathbf{L}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_M),$$

*for every  $0 < \tau \leq \infty$ ; in particular, all the supports  $\mathcal{S}_{\mu_\tau}$ 's are identical.*

(b) *If  $M$  satisfies  $\mathbf{I}_2$  and  $0 \in \mathcal{S}_\rho$ , then  $\mathcal{S}_{\mu_0}$  and  $\Sigma(\mathcal{S}_M)$  are convex cones; moreover,*

$$(3.2) \quad \mathcal{S}_{\mu_0} = \Sigma(\mathcal{S}_M) = \mathbf{C}(\mathcal{S}_\sigma) \quad \text{and} \quad \mathcal{S}_{\mu_\tau} = -a_\sigma \varphi_\tau(0^+) + \Sigma(\mathcal{S}_M),$$

*for all  $0 < \tau \leq \infty$  (recall  $\varphi_\tau = \varphi_{\tau, \rho}$ ). Further,  $\mathcal{S}_{\mu_\tau}$ , for one (equivalently, for all)  $\tau \in [0, \infty]$ , is a linear space if and only if  $\mathcal{S}_\sigma$  (equivalently,  $\mathcal{S}_M$ ) is not contained in the half-space  $\{y \geq 0\}$ , for any  $y \in \mathbf{B}^*$  with  $y \neq 0$  on  $\mathbf{L}(\mathcal{S}_\sigma)$ . If this condition is satisfied, then  $\forall \tau$ ,*

$$(3.3) \quad \mathcal{S}_{\mu_\tau} = \Sigma(\mathcal{S}_M) = \mathbf{C}(\mathcal{S}_\sigma) = \mathbf{B}_0(\mu_\tau) = \mathbf{L}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_M),$$

*in particular, all supports  $\mathcal{S}_{\mu_\tau}$  are identical.*

Let  $\nu$  be an infinitely divisible (i.d.) probability measure on  $\mathbf{B}$  (without a Gaussian component) with Lévy measure  $M = \Psi(\sigma \times \rho)$ . Then, for any  $0 \leq \tau \leq \infty$ ,

$$(3.4) \quad \nu = \delta_{\{x_\tau\}} * c_\tau\text{-Pois}(M) = \delta_{\{x_\tau\}} * c_\tau\text{-Pois}(\sigma, \rho),$$

for some  $x_\tau \in \mathbf{B}$ , where, recall, it is assumed that the condition (i)  $\int_{[1, \infty)} s\rho(ds) < \infty$  [resp., (ii)  $\int_{(0, 1]} s\rho(ds) < \infty$ ] is satisfied if  $\tau = \infty$  (resp.,  $\tau = 0$ ). For  $t > 0$ , the measure  $\nu^t = \delta_{\{tx_\tau\}} * \text{Pois}(tM) = \delta_{\{tx_\tau\}} * \text{Pois}(\sigma, t\rho)$  is called the  $t$ -th root of  $\nu$ . If  $\nu$  is symmetric, then one can take  $M$  and  $\sigma$  symmetric and, hence, one can write  $\nu = c_1\text{-Pois}(M) = c_1\text{-Pois}(\sigma, \rho)$  [resp.,  $c_0\text{-Pois}(M) = c_0\text{-Pois}(\sigma, \rho)$ ] under the condition  $\mathbf{I}_1$  (resp.,  $\mathbf{I}_2$ ).

Recall that a measure  $\nu_\alpha$  on  $\mathbf{B}$  is called  $\alpha$ -stable,  $0 < \alpha < 2 \Leftrightarrow \nu_\alpha$  admits the representation (3.4), with  $\rho_\alpha(ds) = s^{-(1+\alpha)} ds$ . Clearly, since  $\rho_\alpha$  satisfies the above condition (i) [resp., (ii)] if  $1 < \alpha < 2$  (resp.,  $0 < \alpha < 1$ ), one can take  $\tau = \infty$  (resp.,  $\tau = 0$ ), in (3.4), if  $1 < \alpha < 2$  (resp.,  $0 < \alpha < 1$ ). Recall also that  $\nu_\alpha$

is called *strictly  $\alpha$ -stable*  $\Leftrightarrow \nu_\alpha^t(\cdot) = \nu_\alpha(t^{-1/\alpha} \cdot)$  for all  $t > 0$ ; in the case  $1 < \alpha < 2$  (resp.,  $0 < \alpha < 1$ ), this is equivalent to taking  $x_\infty = \theta$  (resp.,  $x_0 = \theta$ ) in (3.4).

In view of the above observations and the facts that  $\mathcal{S}_{t\gamma} = \mathcal{S}_\gamma$  for a measure  $\gamma$  on  $\mathbf{B}$ , that  $t\widetilde{M} = t\bar{M}$  and that  $\alpha_\sigma = \theta$  (and, hence,  $\tilde{\sigma} = \sigma$ ,  $\bar{M} = M$ ) when  $\sigma$  is symmetric, Theorem 3.1 immediately yields the following.

**COROLLARY 3.1.** *Let  $\nu$  be an i.d. probability measure on  $\mathbf{B}$  with representation (3.4). Then we have the following:*

(a) *If  $\rho$  satisfies  $\mathbf{I}_1$  and  $0 < \tau \leq \infty$ , then  $\mathcal{S}_{\nu^t}$  is the translate by  $tx_\tau$  of any one of the equal linear spaces appearing in (3.1). This applies, in particular, when  $\nu$  is  $\alpha$ -stable  $1 \leq \alpha < 2$ ; further, if  $\nu$  is strictly  $\alpha$ -stable,  $1 < \alpha < 2$ , then  $\mathcal{S}_{\nu^t}$  is equal to any one of these linear spaces.*

(b) *If  $\rho$  satisfies  $\mathbf{I}_2$  (and  $0 \leq \tau \leq \infty$ ), then  $\mathcal{S}_{\nu^t}$  is a translate by  $t(x_\tau - \alpha_\sigma \varphi_\tau(0^+))$  of the convex cone  $\mathbf{C}(\mathcal{S}_\sigma) [\equiv \Sigma(\mathcal{S}_\nu)] [\varphi_0(0^+) \equiv 0]$ ; further,  $\mathcal{S}_{\nu^t}$  is the translate by  $tx_\tau$  of any one of the equal linear spaces appearing in (3.3)  $\Leftrightarrow$  the condition for the linearity for  $\mathcal{S}_{\mu_\tau}$  in Theorem 3.1 is satisfied. These apply, in particular, if  $\nu$  is  $\alpha$ -stable,  $0 < \alpha < 1$ . Further, if  $\nu$  is strictly  $\alpha$ -stable,  $0 < \alpha < 1$ , then  $\mathcal{S}_{\nu^t} = \mathbf{C}(\mathcal{S}_\sigma)$ , and the obvious analogous statement holds for the linearity of  $\mathcal{S}_{\nu^t}$ .*

(c) *If  $\nu$  is symmetric (with  $\sigma$  and  $M$  also taken symmetric) then, under either one of the conditions  $\mathbf{I}_1$  or  $\mathbf{I}_2$ ,  $\mathbf{C}(\mathcal{S}_\sigma)$  and  $\Sigma(\mathcal{S}_M)$  are linear spaces and  $\mathcal{S}_{\nu^t} = \mathbf{C}(\mathcal{S}_\sigma) = \Sigma(\mathcal{S}_M) = \mathbf{L}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_M)$ . All of these apply, in particular, to symmetric  $\alpha$ -stable measures,  $\forall \alpha$ .*

Whenever we encounter several Lévy measures having the *same* first component  $\sigma$  but *varying* second component we will write, for clarity,  $M(\rho_r)$  for  $M$ , if the second component is  $\rho_r$ . Further, we will write  $\mu(\rho_r)$  for the probability measure  $c_1$ -Pois( $\sigma, \rho_r$ ). In addition, for a given Lévy measure  $\rho_r$  on  $R^+$ , we shall write  $\psi_{\rho_r}$  for the function  $\psi_{\rho_r}(\lambda) = \rho_r([\lambda, \infty))$ ,  $\lambda > 0$ ; finally, we shall also use the notation  $\varphi_{\rho_r}$  for  $\varphi_{1, \rho_r}$ .

**LEMMA 3.1.** *Let  $M = \Psi(\sigma \times \rho)$  be a finite (or infinite) Lévy measure on  $\mathbf{B}$ . Assume that  $0 \in \mathcal{S}_\rho$ . Then  $\Sigma(\mathcal{S}_M)$  is a convex cone and,  $\forall m_0 \in N \equiv \{1, 2, \dots\}$ ,*

$$(3.5) \quad \mathbf{C}(\mathcal{S}_\sigma) = \Sigma(\mathcal{S}_M) = \bigcap_{n \geq 1} \left[ \bigcup_{k \geq n} A_{\lambda_k} \right] = \left[ \bigcup_{k \geq m_0} A_{\lambda_k} \right],$$

where  $\{\lambda_k\}$  is any sequence with  $1 > \lambda_k \downarrow 0$  and the  $A_{\lambda_k}$ 's are as defined in Proposition 2.1.

From (3.5) one notes that  $\Sigma(\mathcal{S}_M)$  is independent of any other property of  $\rho$  except that  $0 \in \mathcal{S}_\rho$ ; this will be used in the following without any specific reference.

**LEMMA 3.2.** *Let  $M(\rho_i) = \Psi(\sigma \times \rho_i)$ ,  $i = 1, 2$ , be two Lévy measures on  $\mathbf{B}$ . Assume  $\mathcal{S}_{\rho_1} = \mathcal{S}_{\rho_2}$  and  $\varphi_{\rho_1}(\lambda) - \varphi_{\rho_2}(\lambda) \rightarrow 0$  as  $\lambda \downarrow 0$ . Then  $\mathcal{S}_{\mu(\rho_1)} = \mathcal{S}_{\mu(\rho_2)}$ .*

LEMMA 3.3. Let  $\rho_d$  be a discrete Lévy measure on  $R^+$  with  $\varphi_{\rho_d}(0^+) = \infty$ . Then one can construct a discrete Lévy measure  $\rho_{d0} = \sum_n p_n \delta_{\{t_n\}}$  on  $R^+$  such that  $1 > t_1 > \dots > t_n > \dots > t_n \downarrow 0$ ,  $\varepsilon_n = t_n p_n \rightarrow 0$ , as  $n \rightarrow \infty$ ,  $0 < p_n \leq \rho_d(\{t_n\})$ ,  $\forall n$ , and that  $\varphi_{\rho_{d0}}(0^+) = \infty$ .

LEMMA 3.4. Let  $\rho_{d0} = \sum_{n=1}^\infty p_n \delta_{\{t_n\}}$  be the Lévy measure on  $R^+$  as in the conclusion of Lemma 3.3. Then one can construct Lévy measures  $\rho_1, \rho_2, \rho_{20}$  and  $\rho_{21}$  on  $R^+$  which satisfy the following properties:

(a)  $\rho_1$  is continuous on  $R^+$ ,  $\varphi_{\rho_1}(0^+) = \infty$  and

$$(3.6) \quad \psi_{\rho_1}(\lambda) \leq \psi_{\rho_{d0}}(2^{-1}\lambda), \quad \text{for all } \lambda > 0;$$

(b)  $\mathcal{S}_{\rho_2} = \mathcal{S}_{\rho_1}$  and  $\varphi_{\rho_2}(\lambda) - \varphi_{\rho_1}(\lambda) \rightarrow 0$ , as  $\lambda \rightarrow 0$  [hence, also  $\varphi_{\rho_2}(0^+) = \infty$ ];

(c)  $\rho_2 = \rho_{20} + \rho_{21}$ ,  $\rho_{21}(R^+) < \infty$ ,  $\rho_{20} \leq \rho_{d0}$ ,  $0 \in \mathcal{S}_{\rho_{20}} \cap \mathcal{S}_{\rho_{21}}$  and  $(\rho_{d0} - \rho_{20})(R^+) < \infty$ .

PROOF OF THEOREM 3.1(a). Clearly,  $\mathbf{I}_1$  implies that  $0 \in \mathcal{S}_\rho$ ; hence, from (3.5),  $\Sigma(\mathcal{S}_M) = \mathbf{C}(\mathcal{S}_\sigma)$ . Thus, it is sufficient to prove that  $\mathbf{C}(\mathcal{S}_\sigma)$  is a linear space and that  $\mathcal{S}_{c_1\text{-Pois}(\sigma, \rho)} = \mathbf{C}(\mathcal{S}_\sigma)$ . In fact, once this is done, the proof of (3.1) follows from (2.1), (2.8) and the fact that  $\mu_\tau = \delta_{\{k(\tau)\alpha_\sigma\}} * c_1\text{-Pois}(\sigma, \rho)$ , where  $k(\tau) = -\int_{R^+} sI_{(\tau, 1]}(s)\rho(ds)$ , if  $0 < \tau < 1$ , and  $k(\tau) = -\int_{R^+} sI_{(1, \tau]}(s)\rho(ds)$ , if  $1 < \tau \leq \infty$ ; and the fact that  $\alpha_\sigma \in \mathbf{C}(\mathcal{S}_\sigma) \subseteq \mathbf{C}(\mathcal{S}_\sigma)$  [see (2.4)]. Set  $\mu \equiv c_1\text{-Pois}(\sigma, \rho)$ ; now we will prove (i)  $\mathbf{C}(\mathcal{S}_\sigma)$  is a linear space and (ii)  $\mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_\sigma)$ . The proof is divided into two parts.

Part I. In this part we prove the above two assertions under the additional assumption that  $\rho$  has no atom near 0. First consider the case when  $\alpha_\sigma = \theta$ ; in this case we have  $\tilde{\sigma} = \sigma$  and  $\tilde{M} = M$  [see (2.3)]. Hence, using (2.4) and (3.5), we get (ii), that is,

$$(3.7) \quad \mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}}) = \mathbf{C}(\mathcal{S}_\sigma).$$

Next we show that if  $y \in \mathbf{B}^*$  and  $y(x_0) \neq 0$ , for some  $x_0 \in \mathbf{L}(\mathcal{S}_\sigma)$ , then  $\mathcal{S}_{\mu \circ y^{-1}} = R$ . As  $y \neq 0$  on  $\mathbf{L}(\mathcal{S}_\sigma)$ , we can find an  $u_0 \in \mathcal{S}_\sigma$  with  $y(u_0) \neq 0$ ; hence there exists an open neighborhood  $V$  of  $u_0$  such that  $y$  is nonzero on  $V$ . Thus, since  $u_0 \in \mathcal{S}_\sigma$ , we have  $\sigma(W) > 0$  and  $|y| > 0$  on  $W$ , where  $W = V \cap \mathcal{S}_\sigma$ . It is easy to verify that  $\mu \circ y^{-1}$  is i.d. with Lévy measure  $M_y \equiv M \circ y^{-1} / R \setminus \{0\} = (\sigma \times \rho) \circ (y \circ \Psi)^{-1} / R \setminus \{0\}$ . Therefore, we have

$$\begin{aligned} \int_{\{0 < |s| < 1\}} |s| dM_y &= \int_{\{x: y(x) \neq 0\}} |y(x)| I_{\{0 < |s| \leq 1\}}(y(x)) dM \\ &\geq \int_W |y(u)| \left( \int_{(0, |y(u)|^{-1}] } s \rho(ds) \right) \sigma(du) = \infty, \end{aligned}$$

since the integral  $\int_{(0, |y(u)|^{-1}] } s \rho(ds) = \infty$ , by  $\mathbf{I}_1$ , and  $|y(u)| > 0$ , for every  $u \in W$  and  $\sigma(W) > 0$ . Therefore, by [11], page 293,  $\mathcal{S}_{\mu \circ y^{-1}} = R$ .

Now we prove that  $\mathcal{S}_\mu = \mathbf{L}(\mathcal{S}_\sigma)$ . We already know  $\mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_\sigma) \subseteq \mathbf{L}(\mathcal{S}_\sigma)$ ; if there is a strict inclusion, then we can find  $x_0 \in \mathbf{L}(\mathcal{S}_\sigma) \setminus \mathbf{C}(\mathcal{S}_\sigma)$ , and, by the Hahn–Banach theorem,

$$(3.8) \quad y_0 \in \mathbf{B}^* \quad \text{with} \quad y_0(x_0) < r \equiv \inf\{y_0(x) : x \in \mathbf{C}(\mathcal{S}_\sigma)\}.$$

Then, since  $\theta \in \mathbf{C}(\mathcal{S}_\sigma)$ , we have  $r \leq 0$ ; hence  $y_0(x_0) < 0$ . Therefore, from above,  $\mathcal{S}_{\mu \circ y_0^{-1}} = R$ . On the other hand, by (3.7) and (3.8),  $\mu\{x \in B : y_0(x) \geq r\} = 1$ ; this is a contradiction. This completes the proof of the two assertions, under the condition  $a_\sigma = \theta$ .

Now let  $a_\sigma \neq \theta$ , and set  $\tilde{\mu} = c_1\text{-Pois}(\tilde{M}) = c_1\text{-Pois}(\tilde{\sigma}, \rho)$ . Now, since  $a_{\tilde{\sigma}} = \int_{\partial\Delta} u \tilde{\sigma}(du) = a_\sigma - a_\sigma = \theta$ , it follows, from what we proved above, that  $\mathcal{S}_{\tilde{\mu}}$  is a linear space and that

$$(3.9) \quad \mathcal{S}_{\tilde{\mu}} = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}}) [= \mathbf{L}(\mathcal{S}_{\tilde{\sigma}})].$$

Next let  $\gamma_1 = c_1\text{-Pois}(\|a_\sigma\| \delta_{\{1\}}, \rho)$  and  $\nu_1 = \gamma_1 \circ f^{-1}$ , where  $f: R \rightarrow B$  is the map  $f(s) = -s(a_\sigma/\|a_\sigma\|)$ . Then, from  $\mathbf{I}_1$  and [11], page 293,  $\mathcal{S}_{\gamma_1} = R$ ; hence, using Lemma 1 of [7],  $\mathcal{S}_{\nu_1} = a_\sigma R$ . Now observe that  $\nu_1 = c_1\text{-Pois}(\|a_\sigma\| \delta_{\{-a_\sigma/\|a_\sigma\|\}}, \rho)$  and that

$$\tilde{\mu} = c_1\text{-Pois}(\sigma, \rho) * c_1\text{-Pois}(\|a_\sigma\| \delta_{\{-a_\sigma/\|a_\sigma\|\}}, \rho) = \mu * \nu_1.$$

Therefore, by (3.9) and [5], page 307, we have

$$(3.10) \quad \mathbf{C}(\mathcal{S}_{\tilde{\sigma}}) = \mathcal{S}_{\tilde{\mu}} = \overline{\mathcal{S}_\mu + \mathcal{S}_{\nu_1}} = \overline{\mathcal{S}_\mu + a_\sigma R}.$$

Now observing that  $a_\sigma \in \mathbf{C}(\mathcal{S}_\sigma) [= \Sigma(\mathcal{S}_M)]$  [see (2.4) and (3.5)], we have, by [12], page 352,  $\mathcal{S}_\mu + ta_\sigma \subseteq \mathcal{S}_\mu$ , for all  $t \geq 0$ , and, by [10], page 38 (where it is additionally required that  $\rho$  is continuous on  $R^+$  and that  $\mathcal{S}_\rho = [0, \infty)$ , but these are not required for the proof of  $\mathcal{S}_\mu - ta_\sigma \subseteq \mathcal{S}_\mu$ , for  $t \geq 0$ ),  $\mathcal{S}_\mu - ta_\sigma \subseteq \mathcal{S}_\mu$ , for all  $t \geq 0$ . Consequently, since  $\mathcal{S}_\mu \subseteq \mathbf{L}(\mathcal{S}_\sigma) \subseteq \mathbf{L}(\mathcal{S}_{\tilde{\sigma}}) = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$  [see (2.3), (2.8) and (3.9)], we have, from (3.10), that  $\mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$ , completing the proof of Part I.

*Part II.* Now let  $\rho$  be arbitrary and let  $\rho_c$  and  $\rho_d$  denote the continuous and discrete parts of  $\rho$ ; then, since  $M(\rho_c)$  and  $M(\rho_d)$  are dominated by  $M$  [ $\equiv M(\rho)$ ], both of these are Lévy measures. Since  $M = M(\rho_c) + M(\rho_d)$ , we have  $\mu = \mu(\rho_c) * \mu(\rho_d)$ . If  $\varphi_{\rho_c}(0^+) = \infty$ , then, from Part I,  $\mathcal{S}_{\mu(\rho_c)}$  is a linear space and  $\mathcal{S}_{\mu(\rho_c)} = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$ . Therefore, since

$$\mathcal{S}_\mu = \overline{\mathcal{S}_{\mu(\rho_c)} + \mathcal{S}_{\mu(\rho_d)}} = \overline{\mathbf{C}(\mathcal{S}_{\tilde{\sigma}}) + \mathcal{S}_{\mu(\rho_d)}},$$

and since  $\mathcal{S}_{\mu(\rho_d)} \subseteq \mathbf{L}(\mathcal{S}_\sigma)$  [by (2.8)]  $\subseteq \mathbf{L}(\mathcal{S}_{\tilde{\sigma}}) = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$ , we have  $\mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_{\tilde{\sigma}})$  ( $= \mathcal{S}_{\mu(\rho_c)}$ ). Thus, the only remaining case of interest is when  $\varphi_{\rho_c}(0^+) < \infty$  and  $\varphi_{\rho_d}(0^+) = \infty$ . To take care of this case, let  $\rho_{d0}$  be the Lévy measure on  $R^+$  constructed in Lemma 3.3 for the measure  $\rho_d$ . Then, observing that  $\rho = \rho_{d0} + (\rho_c + (\rho_d - \rho_{d0}))$  (recall that  $\rho_{d0} \leq \rho_d$ ), we have that  $M(\rho_{d0})$  and  $M(\rho_{d'0})$  are

both Lévy measures on  $\mathbf{B}$ , where  $\rho_{d'0} = \rho_c + (\rho_d - \rho_{d0})$ . Hence  $\mathcal{S}_\mu = \mathcal{S}_{\mu(\rho_{d0})} + \mathcal{S}_{\mu(\rho_{d'0})}$ ; thus, making use of similar arguments to those above, it follows that, in order to prove that  $\mathcal{S}_\mu$  is a linear space and that  $\mathcal{S}_\mu = \mathbf{C}(\mathcal{S}_\sigma)$ , we need only prove these two assertions for  $\mathcal{S}_{\mu(\rho_{d0})}$ . We do this in the following:

Let  $\rho_1, \rho_2, \rho_{20}$  and  $\rho_{21}$  be the Lévy measures on  $R^+$  constructed in Lemma 3.4 for the measure  $\rho_{d0}$ . Using (3.6) and the fact that  $M(\rho_{d0})$  is a Lévy measure on  $\mathbf{B}$ , we have, from the contraction principle ([8], page 64; we thank Professor Jan Rosinski for bringing this contraction principle to our attention) that  $M(\rho_1)$  is also a Lévy measure on  $\mathbf{B}$ . Thus, since  $\rho_1$  is continuous on  $R^+$  and  $\varphi_{\rho_1}(0^+) = \infty$ , we have, from Part I, that  $\mathcal{S}_{\mu(\rho_1)}$  is a linear space and that  $\mathcal{S}_{\mu(\rho_1)} = \mathbf{C}(\mathcal{S}_\sigma)$ . Since  $\rho_{21}(R^+) < \infty$  and  $\rho_{20} \leq \rho_{d0}$ , both  $M(\rho_{21})$  and  $M(\rho_{20})$  [and, hence, also  $M(\rho_2)$ ] are Lévy measures on  $\mathbf{B}$ . Hence, since  $\mathcal{S}_{\rho_1} = \mathcal{S}_{\rho_2}$  and  $\varphi_{\rho_2}(\lambda) - \varphi_{\rho_1}(\lambda) \rightarrow 0$  as  $\lambda \rightarrow 0$ , we have, by Lemma 3.2, that  $\mathcal{S}_{\mu(\rho_2)} = \mathcal{S}_{\mu(\rho_1)}$ . Thus  $\mathcal{S}_{\mu(\rho_2)}$  is a linear space and is equal to  $\mathbf{C}(\mathcal{S}_\sigma)$ . Since  $\rho_{21}(R^+) < \infty$ , we have that  $\mathcal{S}_{\mu(\rho_{21})} = \Sigma(\mathcal{S}_{M(\rho_{21})}) - \varphi_{\rho_{21}}(0^+)a_\sigma$  ([6], page 333); further, since  $0 \in \mathcal{S}_{\rho_{21}} \cap \mathcal{S}_{\rho_{20}}$  we have, from Lemma 3.1, that  $\Sigma(\mathcal{S}_{M(\rho_{21})}) = \Sigma(\mathcal{S}_{M(\rho_{20})})$ . Therefore, since, by [12], page 352,  $\mathcal{S}_{\mu(\rho_{20})} + \Sigma(\mathcal{S}_{M(\rho_{20})}) \subseteq \mathcal{S}_{\mu(\rho_{20})}$ , and since  $\mathcal{S}_{\mu(\rho_2)} = \mathcal{S}_{\mu(\rho_{20})} + \mathcal{S}_{\mu(\rho_{21})}$  [recall  $M(\rho_2) = M(\rho_{20}) + M(\rho_{21})$ ], we have that  $\mathcal{S}_{\mu(\rho_2)} \subseteq \mathcal{S}_{\mu(\rho_{20})} - \varphi_{\rho_{21}}(0^+)a_\sigma$ . Consequently, since  $\mathcal{S}_{\mu(\rho_2)} = \mathbf{C}(\mathcal{S}_\sigma)$  is a linear space and  $a_\sigma \in \mathbf{C}(\mathcal{S}_\sigma) \subseteq \mathbf{C}(\mathcal{S}_\sigma)$  [recall (2.3) and (2.4)], we have  $\mathbf{C}(\mathcal{S}_\sigma) \subseteq \mathcal{S}_{\mu(\rho_{20})}$ . But, by (2.4) and (2.8),  $\mathcal{S}_{\mu(\rho_{20})} \subseteq \mathbf{L}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_\sigma)$ ; therefore, since  $\mathbf{C}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_\sigma)$ , we have  $\mathcal{S}_{\mu(\rho_{20})} = \mathbf{C}(\mathcal{S}_\sigma)$  [ $\equiv \mathbf{L}(\mathcal{S}_\sigma)$ ]. Finally, since  $\rho_{20} \leq \rho_{d0}$ , we have  $\mathcal{S}_{\mu(\rho_{d0})} = \mathcal{S}_{\mu(\rho_{20})} + \mathcal{S}_{\mu(\rho_{d0} - \rho_{20})}$ . But, since  $\mathcal{S}_{\mu(\rho_{20})} = \mathbf{C}(\mathcal{S}_\sigma) = \mathbf{L}(\mathcal{S}_\sigma)$  and  $\mathcal{S}_{\mu(\rho_{d0} - \rho_{20})} \subseteq \mathbf{L}(\mathcal{S}_\sigma)$ , we have  $\mathcal{S}_{\mu(\rho_{d0})} = \mathcal{S}_{\mu(\rho_{20})}$ . This completes the proof of part (a) of the theorem.  $\square$

**PROOF OF THEOREM 3.1(b).** Clearly, we have  $\mu_\tau = \delta_{\{-\varphi_\tau(0^+)a_\sigma\}} * \mu_0$ ; also, by (2.7), we have  $\mathcal{S}_{\mu_0} = \bigcap_{m \geq 1} [\bigcup_{n \geq m} A_{\lambda_n}]$ . Therefore, by Lemma 3.1,  $\mathcal{S}_{\mu_0}$  and  $\Sigma(\mathcal{S}_M)$  are convex cones and (3.2) holds. The proof of the stated equivalence now follows easily using the Hahn-Banach theorem as in the proof of Theorem 3.1(a) and is omitted. The proof of (3.3) follows trivially from (3.2), (2.5) and (2.8).  $\square$

**PROOF OF LEMMA 3.1.** Trivially,  $\mathcal{S}_M \cup \{\theta\} \supseteq \mathcal{S}_{M_{\lambda_k}} \cup \{\theta\}, \forall k$ , where  $M_{\lambda_k} = M / \{\|x\| \geq \lambda_k\}$ ; consequently,  $\Sigma(\mathcal{S}_M) \supseteq [\bigcup_{k \geq n} A_{\lambda_k}]$ , for every  $n \in N$ . Further, since  $\mathcal{S}_M \cup \{\theta\} \subseteq [\bigcup_{k \geq n} A_{\lambda_k}], \forall n$ , we also have  $\Sigma(\mathcal{S}_M) \subseteq [\bigcup_{k \geq n} A_{\lambda_k}]$ . Thus  $\Sigma(\mathcal{S}_M) = \bigcap_{n \geq 1} [\bigcup_{k \geq n} A_{\lambda_k}] = [\bigcup_{k \geq m_0} A_{\lambda_k}], \forall m_0 \in N$ . Up to now no special property of  $M$  is used, and in fact, there is nothing new in this proof; it is included in [10], page 34. To see that  $\Sigma(\mathcal{S}_M) = \mathbf{C}(\mathcal{S}_\sigma)$ , we first observe that  $\mathcal{S}_M = \{tu: t \in \mathcal{S}_\rho, u \in \mathcal{S}_\sigma\}$  (use Lemma 1 of [7], page 302). Therefore, as  $0 \in \mathcal{S}_\rho$ ,  $\Sigma(\mathcal{S}_M) = \{\sum t_i u_i: t_i \in \mathcal{S}_\rho, u_i \in \mathcal{S}_\sigma\}$ , where  $\Sigma$  denotes the sum over the finite elements. Thus, since, clearly,  $\mathbf{C}(\mathcal{S}_\sigma) = \{\sum t_i u_i: t_i \geq 0, u_i \in \mathcal{S}_\sigma\}$ , we have that  $\mathbf{C}(\mathcal{S}_\sigma) \supseteq \Sigma(\mathcal{S}_M)$ . Now, let  $t_0 > 0$  and  $u_0 \in \mathcal{S}_\sigma$ ; then, as  $0 \in \mathcal{S}_\rho$ , we



can choose  $t_n \in \mathcal{S}_\rho$  and  $k_n \in N$  such that  $t_n k_n \rightarrow t_0$ . Therefore, since  $k_n t_n u_0 \in \Sigma(\mathcal{S}_M)$  and  $k_n t_n u_0 \rightarrow t_0 u_0$ , we have  $t_0 u_0 \in \Sigma(\mathcal{S}_M)$ . Thus, as  $\Sigma(\mathcal{S}_M)$  is a closed semigroup,  $\{\sum t_i u_i : t_i \geq 0, u_i \in \mathcal{S}_\sigma\}$  and, hence,  $C(\mathcal{S}_\sigma) = \{\sum t_i u_i : t_i \geq 0, u_i \in \mathcal{S}_\sigma\} \subseteq \Sigma(\mathcal{S}_M)$ , proving that  $C(\mathcal{S}_\sigma) = \Sigma(\mathcal{S}_M)$ .  $\square$

PROOF OF LEMMA 3.2. The hypothesis  $\mathcal{S}_{\rho_1} = \mathcal{S}_{\rho_2}$  clearly implies that, for any sequence  $\{\lambda_k\}$ , with  $1 > \lambda_1 > \dots > \lambda_n > \dots$  and  $\lambda_n \downarrow 0$ ,  $A_{\lambda_k}(M(\rho_1)) = A_{\lambda_k}(M(\rho_2))$  for all  $k$  (see Proposition 2.1 for the definition of  $A_{\lambda_k}$ 's). Now let  $x \in \mathcal{S}_{\mu(\rho_1)}$  and  $\varepsilon > 0$ ; then, by (2.6), there exists a subsequence  $\{\lambda_{k(r)}\}$  of  $\{\lambda_k\}$  and  $x_r \in A_{\lambda_{k(r)}}(M(\rho_1))$  such that  $\|x - (x_r - \alpha_\sigma \varphi_{\rho_1}(\lambda_{k(r)}))\| < \varepsilon/2$ , for large  $r$ . Then, observing that

$$\begin{aligned} \|x - (x_r - \alpha_\sigma \varphi_{\rho_2}(\lambda_{k(r)}))\| &\leq \|x - (x_r - \alpha_\sigma \varphi_{\rho_1}(\lambda_{k(r)}))\| \\ &\quad + \|\alpha_\sigma\| |\varphi_{\rho_1}(\lambda_{k(r)}) - \varphi_{\rho_2}(\lambda_{k(r)})| \end{aligned}$$

and using the hypothesis  $\varphi_{\rho_1}(\lambda) - \varphi_{\rho_2}(\lambda) \rightarrow 0$  as  $\lambda \rightarrow 0$ , we get that  $\|x - (x_r - \alpha_\sigma \varphi_{\rho_2}(\lambda_{k(r)}))\| < \varepsilon$ , for large  $r$ ; therefore, since  $x_r \in A_{\lambda_{k(r)}}(M(\rho_2))$ , for every  $r$ , we have  $x \in \mathcal{S}_{\mu(\rho_2)}$ . Thus  $\mathcal{S}_{\mu(\rho_1)} \subseteq \mathcal{S}_{\mu(\rho_2)}$ ; similarly,  $\mathcal{S}_{\mu(\rho_2)} \subseteq \mathcal{S}_{\mu(\rho_1)}$ , showing that  $\mathcal{S}_{\mu(\rho_1)} = \mathcal{S}_{\mu(\rho_2)}$ .  $\square$

PROOF OF LEMMA 3.3. Since  $\varphi_{\rho_d}(0^+) = \infty$ , we can choose distinct  $0 < u_j^{(1)} < 1$ ,  $j = 1, \dots, m_1$ , such that  $\sum_{j=1}^{m_1} u_j^{(1)} \rho_d(\{u_j^{(1)}\}) \geq 1$ . Let  $u_{j(1)}^{(1)} > \dots > u_{j(m_1)}^{(1)}$  be the natural order of  $u_j^{(1)}$ 's and let  $v_r = u_{j(r)}^{(1)}$ ,  $r = 1, 2, \dots, m_1$ . Now noting that  $\int_{(0, v_{m_1})} s \rho_d(ds) = \infty$ , with a similar argument, we get  $v_{m_1+1} > \dots > v_{m_2}$  with  $\sum_{j=1}^{m_2} v_{m_1+j} \rho_d(\{v_{m_1+j}\}) \geq 2$ . Continuing this process inductively, we get a sequence  $\{v_r\}$  such that  $1 > v_1 > \dots > v_r > \dots$ ,  $v_r \downarrow 0$  and  $\sum_{r=1}^\infty v_r \rho_d(\{v_r\}) = \infty$ . Now if  $v_r \rho_d(\{v_r\}) \rightarrow 0$  as  $r \rightarrow \infty$ , then we are done by taking  $s_r = v_r$  and  $p_r = \rho_d(\{v_r\})$ . If not, then there must exist a subsequence  $\{r_n\}$  of  $\{r\}$  and  $\varepsilon > 0$  such that  $v_{r_n} \rho_d(\{v_{r_n}\}) \geq \varepsilon$ ,  $\forall n$ . In this case, it suffices to take  $s_n = v_{r_n}$  and  $p_n = \varepsilon(v_{r_n} n)^{-1}$ ,  $\forall n$ .  $\square$

PROOF OF LEMMA 3.4. (The reader is advised to draw rough sketches of the functions  $\varphi$ 's and  $\psi$ 's to understand the proof.) Let

$$0 < \alpha_n < \min\left\{\frac{t_n}{3}, \frac{t_{n-1} - t_n}{2}, \frac{t_n - t_{n+1}}{2}\right\}, \text{ with } t_0 = 1;$$

set  $f(s) = \sum_{n=1}^\infty (2\alpha_n)^{-1} p_n I_{T_n}(s)$  and define  $\rho_1(ds) = f(s) ds$ ,  $0 < s < \infty$ , where  $T_n = [t_n - \alpha_n, t_n + \alpha_n]$ ,  $n \in N$ . Next let  $0 < \beta_n < \min\{1, p_n/4\}$ ,  $q_n = p_n - 2\alpha_n \beta_n$ ,  $\forall n \in N$  (note that  $0 < q_n < p_n$ ). Let  $\rho_{20}$  and  $\rho_{21}$  denote, respectively, the measures on  $R^+$  defined by  $\sum_{n=1}^\infty q_n \delta_{\{t_n\}}$  and  $g(s) ds$ , where  $g(s) = \sum_{n=1}^\infty \beta_n I_{T_n}(s)$ ,  $s \in R^+$ ; finally, define  $\rho_2 = \rho_{20} + \rho_{21}$ .

Clearly,  $\rho_1$  is a continuous measure on  $R^+$ ,  $\mathcal{S}_{\rho_2} = \mathcal{S}_{\rho_1}$  ( $= \cup_n T_n \cup \{0\}$ ),  $0 \in \mathcal{S}_{\rho_{21}}$ . Trivially,  $\rho_i(C) = 0$ ,  $i = 1, 2$ , where  $C = R^+ \setminus \cup_n V_n$  and  $V_n = (t_n - \alpha_n, t_n + \alpha_n)$ . Next, observing that  $\int_{T_n} \rho_{21}(ds) = 2\alpha_n \beta_n$ , we have

$\rho_{21}(R^+) = 2\sum_{n=1}^{\infty} \alpha_n \beta_n < \infty$  (recall that  $0 < \beta_n < 1$  and  $\sum_{n=1}^{\infty} \alpha_n < \infty$ ). This, along with  $0 < q_n < p_n$ , shows that  $\rho_{20} \leq \rho_{d0}$ , that  $\rho_{20}$  is a Lévy measure on  $R^+$  and that  $(\rho_{d0} - \rho_{20})(R^+) < \infty$ . Now note that

$$(3.11) \quad \int_{T_n} s \rho_1(ds) = t_n p_n \quad \text{and} \\ \int_{T_n} s^2 \rho_1(ds) = (2\alpha_n)^{-1} p_n [2\alpha_n^3 + 6t_n^2 \alpha_n] \leq 4t_n^2 p_n;$$

therefore, since  $\rho_{d0} = \sum_{n=1}^{\infty} p_n \delta_{\{t_n\}}$  is a Lévy measure on  $R^+$  and  $\sum_{n=1}^{\infty} t_n p_n = \infty$ , we have that  $\rho_2$  is a Lévy measure on  $R^+$  [recalling that  $\rho_2(C) = 0$ ] and that  $\varphi_{\rho_1}(0^+) = \infty$ . Further, the first equation in (3.11), the facts that  $\int_{T_n} s \rho_2(ds) = q_n t_n + \int_{T_n} s g(s) ds = q_n t_n + 2\alpha_n \beta_n t_n = t_n p_n$  and  $\int_{T_n} s \rho_1(ds) = p_n$  and that  $\rho_i(C) = 0$  together imply

$$(3.12) \quad \varphi_{\rho_{d0}}(\lambda) = \varphi_{\rho_1}(\lambda) = \varphi_{\rho_2}(\lambda) \quad \text{and} \quad \psi_{\rho_{d0}}(\lambda) = \psi_{\rho_1}(\lambda), \quad \forall \lambda \in C$$

and

$$(3.13) \quad \varphi_{\rho_{d0}}(t_n + \alpha_n) < \varphi_{\rho_1}(\lambda), \quad \varphi_{\rho_2}(\lambda) < \varphi_{\rho_{d0}}(t_n + \alpha_n) + t_n p_n, \\ \forall \lambda \in V_n.$$

Therefore, since  $\varepsilon_n = t_n p_n \rightarrow 0$ , it follows, from (3.13) and the first equation in (3.12), that  $\lim_{\lambda \rightarrow 0} (\varphi_{\rho_1}(\lambda) - \varphi_{\rho_2}(\lambda)) = 0$ . The proof is thus completed except for the last part in (a), which we prove now.

From the second equation in (3.12) and the nonincreasing property of  $\psi_{\rho_{d0}}$ , we have that, if  $\lambda \in C$ , then  $\psi_{\rho_1}(\lambda) \leq \psi_{\rho_{d0}}(2^{-1}\lambda)$ . If  $\lambda \in (t_n - \alpha_n, t_n]$ , then  $\psi_{\rho_1}(\lambda) \leq \psi_{\rho_1}(t_n - \alpha_n) = \psi_{\rho_{d0}}(t_n - \alpha_n) = \psi_{\rho_{d0}}(\lambda) \leq \psi_{\rho_{d0}}(2^{-1}\lambda)$ , by the nonincreasing property of  $\psi_{\rho_1}$  and  $\psi_{\rho_{d0}}$ , (3.12) and the facts that  $t_{n-1} < (t_n - \alpha_n)$  and that  $\rho_{d0}$  has jumps only at  $t_n$ 's. Next, if  $t_n < \lambda < (t_n + \alpha_n)$ , then, since  $2^{-1}(t_n + \alpha_n) < (t_n - \alpha_n)$  (recall  $\alpha_n < t_n/3$ ), we have  $2^{-1}\lambda < (t_n - \alpha_n)$ . Therefore, using the nonincreasing property of  $\psi_{\rho_{d0}}$  and  $\psi_{\rho_1}$  and (3.12) again,  $\psi_{\rho_{d0}}(2^{-1}\lambda) \geq \psi_{\rho_{d0}}(t_n - \alpha_n) = \psi_{\rho_1}(t_n - \alpha_n) \geq \psi_{\rho_1}(\lambda)$ .  $\square$

**4. The supports of Poisson probability measures on  $E$ .** Let  $\mu$  be a nondegenerate  $\tau$ -regular probability measure on  $E$  ( $\mu$  is defined on the Borel  $\sigma$ -algebra of the topology of  $E$ ); and let  $E_0 \equiv E_0(\mu)$  be defined as in Proposition 2.1(b). Define  $\mathcal{S}_\mu$  the support of  $\mu$ , as in Section 2. Let  $\mathcal{P}_E$  be a fixed family of seminorms generating the (l.c.) topology of  $E$ , and let  $\mathcal{P} = \{p \in \mathcal{P}_E, p \neq 0 \text{ on } E_0\}$ . Then,  $\forall p \in \mathcal{P}$ , the normed space  $(E_0 / (\{p = 0\} \cap E_0), \|\cdot\|_p)$  and, hence, also its completion  $B_p$ , is separable, where  $\|\cdot\|_p$  is the natural induced norm. Furthermore

$$(4.1) \quad \mu(E_0) = 1, \quad \mathcal{S}_\mu = \bigcap_{\mathcal{P}} \pi_p^{-1}(\mathcal{S}_{\mu_p}),$$

$\mathcal{S}_{\mu_p} \equiv$  the support of  $\mu_p \equiv \mu_0 \circ \pi_p^{-1}$ , where  $\mu_0 = \mu/E_0$  and  $\pi_p$  is the natural projection of  $E_0$  into  $B_p$  (see [7], pages 299–300, and [10], page 29, for details on these results).

Now assume that  $\mu$  is a weakly Poisson type (resp., an  $\alpha$ -stable, a strictly  $\alpha$ -stable) probability measure, that is, each finite-dimensional projection of  $\mu$  is a translated Poisson (resp., an  $\alpha$ -stable, a strictly  $\alpha$ -stable) probability measure. It follows that  $\mu_p$  is a translated centered Poisson (resp., an  $\alpha$ -stable, a strictly  $\alpha$ -stable) probability measure on  $\mathbf{B}_p$  (the result in [9], page 313, is needed here). In the following, the Lévy measure of  $\mu_p$  will be denoted by  $M_p$ , the statement that  $M_p \equiv \Psi(\sigma_p \times \rho_p)$  will have a similar meaning as in Section 2, and the notations  $\tilde{\sigma}_p$  and  $\tilde{M}_p$  will denote measures defined as in (2.3) with  $\sigma$  and  $M$  replaced, respectively, by  $\sigma_p$  and  $M_p$ . Further, if  $\mu_p$  is symmetric, we will take  $M_p$  and  $\sigma_p$  also symmetric. Finally, if  $\mu_p$  is degenerate, we will take, by convention,  $\sigma_p = M_p = \tilde{\sigma}_p = \tilde{M}_p \equiv 0$  and  $\Sigma(\mathcal{L}_{M_p}) = \mathbf{C}(\mathcal{L}_{\sigma_p}) = \Sigma(\mathcal{L}_{\tilde{M}_p}) = \mathbf{C}(\mathcal{L}_{\tilde{\sigma}_p}) \equiv \{\theta\}$ .

Using the above notation and conventions, we now state the main result and a corollary. The proofs of these follow immediately using the above facts, (4.1), Theorem 3.1 and Corollary 3.1, and are, therefore, omitted.

**THEOREM 4.1.** *Let  $\mu$  be a  $\tau$ -regular weakly Poisson type probability measure on  $\mathbf{E}$  and assume  $M_p = \Psi(\sigma_p \times \rho_p)$ ,  $\forall p \in \mathcal{P}$ . Then we have the following:*

(a) *If  $\varphi_{\rho_p}(0^+) [\equiv \int_{(0,1]} s\rho_p(s) ds] = \infty$ , whenever  $\mu_p \equiv \delta_{\{x_p\}} * c_1\text{-Pois}(\sigma_p, \rho_p)$  is nondegenerate, then  $\bigcap_{\mathcal{P}} \pi_p^{-1}(\Sigma(\mathcal{L}_{M_p}))$  and  $\bigcap_{\mathcal{P}} \pi_p^{-1}(\mathbf{C}(\mathcal{L}_{\tilde{\sigma}_p}))$  are equal linear spaces (denoted  $\mathcal{L}_\mu$ ) and*

$$(4.2) \quad \mathcal{L}_\mu = b_\mu + \mathcal{L}_\mu, \quad \text{for any } b_\mu \in \bigcap_{\mathcal{P}} \pi_p^{-1}(x_p + \Sigma(\mathcal{L}_{M_p})).$$

(b) *If  $\mu_p = c_0\text{-Pois}(\sigma_p, \rho_p)$ ,  $\forall p \in \mathcal{P}$ , and if  $\varphi_{\rho_p}(0^+) < \infty$  and  $0 \in \mathcal{L}_{\rho_p}$  whenever  $\mu_p$  is nondegenerate, then  $\mathcal{L}_\mu$  is a convex cone and*

$$(4.3) \quad \mathcal{L}_\mu = \bigcap_{\mathcal{P}} \pi_p^{-1}(\Sigma(\mathcal{L}_{M_p})) = \bigcap_{\mathcal{P}} \pi_p^{-1}(\mathbf{C}(\mathcal{L}_{\sigma_p})),$$

and  $\mathcal{L}_\mu = \mathbf{E}_0(\mu) \Leftrightarrow \forall y \in \mathbf{E}^*$ , which is nonzero on  $\mathbf{E}_0$ ,  $\mathcal{L}_{\mu \circ y^{-1}} = \mathbf{R}$ .

(c) *If  $\mu$  is symmetric and if,  $\forall p$  for which  $\mu_p$  is nondegenerate, either  $\rho_p(0^+) = \infty$  or  $\rho_p(0^+) < \infty$  and  $0 \in \mathcal{L}_{\rho_p}$ , then the convex cones in (4.3) are linear spaces and each is equal to  $\mathcal{L}_\mu$ .*

**COROLLARY 4.1.** *Let  $\mu$  be a  $\tau$ -regular weakly  $\alpha$ -stable probability measure on  $\mathbf{E}$  so that  $M_p = \psi(\sigma_p \times s_{d_s}^{-(1+\alpha)})$ . Then we have the following:*

(a) *If  $1 \leq \alpha < 2$ , then  $\mathcal{L}_\mu = b_\mu + \mathcal{L}_\mu$  [see (4.2)]; further, if  $1 < \alpha < 2$  and  $\mu$  is strictly stable, then  $\mathcal{L}_\mu = \mathcal{L}_\mu$ .*

(b) *If  $0 < \alpha < 1$  and if  $\mu$  is strictly stable, then  $\mathcal{L}_\mu$  is equal to either of the equal cones appearing in (4.3), otherwise it is a translate (by  $x_0$ ) of any of these equal cones, where  $x_0$  is the centering element of  $\mu$ . Further, the analog of the last statement in Theorem 4.1(b) holds.*

(c) *If  $\mu$  is symmetric, then the convex cones in (4.3) are linear spaces and each is equal to  $\mathcal{L}_\mu$ .*

REMARK 4.2. Using the same methods as noted above, versions of both of these results can be proved for weakly Poisson type and weakly  $\alpha$ -stable probability measures  $\mu$  which are defined on the *cylinder  $\sigma$ -algebra* (rather than on the Borel  $\sigma$ -algebra) of  $\mathbf{E}$ . In this case, the hypothesis of  $\tau$ -regularity of  $\mu$  is replaced by “ $\mathbf{E}$  is  $\mu$ -reducible”, and the support of  $\mu$  is defined as in [9], pages 27–28. These results were included in an earlier version of the paper, but are excluded from this version for brevity. Theorem 3.1(b) and the just-noted versions of Theorem 4.1(b) and Corollary 4.1(b) complement Proposition 4 and Theorem 1 of [10]. However, since the proof of these [except perhaps for proving that  $\mathcal{S}_\mu$  is a convex cone in Theorem 3.1(b)] follow directly from Proposition 4 of [10], we do not claim any credit for these.

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