

## LECTURE 19: INFINITELY DIVISIBLE PROBABILITY MEASURES

The convolution product turns  $\mathbf{M}_1(\mathbb{R}^N)$  into a commutative ring in which  $\delta_0$  is the identity. A  $\mu \in \mathbf{M}_1(\mathbb{R}^N)$  is said to be *infinitely divisible* in this ring if, for each  $n \geq 1$ , there exists a  $\mu_{\frac{1}{n}} \in \mathbf{M}_1(\mathbb{R}^N)$  such that

$$\mu = \mu_{\frac{1}{n}}^{*n} \equiv \underbrace{\mu_{\frac{1}{n}} * \cdots * \mu_{\frac{1}{n}}}_{n \text{ times}},$$

and the set  $\mathcal{I}(\mathbb{R}^N)$  of infinitely divisible measures is an important source of building blocks for constructions in probability theory.

For probabilists, an element of  $\mathcal{I}(\mathbb{R}^N)$  is the distribution of a random variable which, for each  $n \geq 1$ , can be written as the sum of  $n$  identically distributed random variables. Using commutativity, it is easy to check that set  $\mathcal{I}(\mathbb{R}^N)$  of infinitely divisible measures is a subring of  $\mathbf{M}_1(\mathbb{R}^N)$ .

A famous theorem of Lévy and A. Khinchine describes the characteristic function of any element of  $\mathcal{I}(\mathbb{R}^N)$ . Namely,  $\mu \in \mathcal{I}(\mathbb{R}^N)$  if and only if

$$(19.1) \quad \hat{\mu}(\boldsymbol{\xi}) = \exp\left(i(\mathbf{b}, \boldsymbol{\xi})_{\mathbb{R}^N} - \frac{1}{2}(\boldsymbol{\xi}, A\boldsymbol{\xi})_{\mathbb{R}^N} + \int \left(e^{i(\boldsymbol{\xi}, y)_{\mathbb{R}^N}} - 1 - i\mathbf{1}_{B(0,1)}(y)(\boldsymbol{\xi}, y)_{\mathbb{R}^N}\right) M(dy)\right),$$

for some  $\mathbf{b} \in \mathbb{R}^N$ , non-negative definite, symmetric  $A \in \text{Hom}(\mathbb{R}^N; \mathbb{R}^N)$ , and Borel measure  $M$  on  $\mathbb{R}^N$  such that  $M(\{0\}) = 0$  and  $\int \frac{|y|^2}{1+|y|^2} M(d\mathbf{y}) < \infty$ . The expression in (19.1) is called the *Lévy–Khinchine formula*, a measure  $M$  satisfying the stated conditions is called a *Lévy measure*, and the triple  $(\mathbf{b}, A, M)$  is called a *Lévy system*. It is clear that if the right hand side of (19.1) is a characteristic function for every Lévy system, then these are characteristic functions of infinitely divisible laws. Indeed, if  $\mu$  corresponds to  $(b, A, M)$  and  $\mu_{\frac{1}{n}}$  corresponds to  $(\frac{\mathbf{b}}{n}, \frac{A}{n}, \frac{M}{n})$ , then  $\hat{\mu} = (\widehat{\mu_{\frac{1}{n}}})^n$ .

Proving that the function  $f_{(\mathbf{b}, A, M)}$  on the right hand side of (19.1) is a characteristic function is a relatively easy. To wit,  $f_{(0, \mathbf{I}, 0)} = \hat{\gamma}$ , where  $\gamma$  is the standard Gaussian measure on  $\mathbb{R}^N$ , and so it is easy to check that  $f_{\mathbf{b}, A, 0}$  is the characteristic function of the distribution of  $\mathbf{x} \rightsquigarrow \mathbf{b} + A^{\frac{1}{2}}\mathbf{x}$  under  $\gamma$ . Also, if the Lévy measure  $M$  is finite and  $\pi_M$  is the Poisson measure given by

$$(19.2) \quad \pi_M = e^{-M(\mathbb{R}^N)} \sum_{n=0}^{\infty} \frac{M^{*n}}{n!},$$

then

$$\widehat{\pi_M}(\boldsymbol{\xi}) = e^{-M(\mathbb{R}^N)} \sum_{n=0}^{\infty} \frac{\hat{M}(\boldsymbol{\xi})^n}{n!} = e^{-M(\mathbb{R}^N) + \hat{M}(\boldsymbol{\xi})} = \exp\left(\int (e^{i(\boldsymbol{\xi}, y)_{\mathbb{R}^N}} - 1) M(dy)\right),$$

and so  $\widehat{\pi_M} = f_{(\mathbf{b}_M, 0, M)}$ , where  $b_M = \int_{B(0,1)} \mathbf{y} M(d\mathbf{y})$ . Hence, when  $M$  is finite,  $f_{(\mathbf{b}, A, M)}$  is the characteristic function of  $\gamma_{\mathbf{b} - \mathbf{b}_M, A} * \pi_M$ . Finally, for general Lévy measures  $M$ , set  $M_k(dy) = \mathbf{1}_{[\frac{1}{k}, \infty)}(|y|) M(dy)$ . Then  $M_k$  is finite, and so  $f_{(\mathbf{b}, A, M_k)}$  is a characteristic function. Therefore, since  $f_{(\mathbf{b}, A, M_k)} \rightarrow f_{(\mathbf{b}, A, M)}$  uniformly on compact subsets, Theorem 18.2 says that  $f_{(\mathbf{b}, A, M)}$  is a characteristic function.

There are no easy proofs that the characteristic function of any  $\mu \in \mathcal{I}(\mathbb{R}^N)$  is given by (19.1). The first step is to show that if  $\mu \in \mathcal{I}(\mathbb{R}^N)$ , then there is a unique  $\ell \in C(\mathbb{R}^N; \mathbb{C})$  such that  $\ell(\mathbf{0}) = 0$ ,  $\frac{|\ell(\boldsymbol{\xi})|}{1+|\boldsymbol{\xi}|^2}$  is bounded, and  $\hat{\mu}(\boldsymbol{\xi}) = e^{\ell(\boldsymbol{\xi})}$ . Showing that  $\ell$  exists and is unique comes down to showing that  $\hat{\mu}$  never vanishes. To do that, choose  $r > 0$  so that  $|1 - \hat{\mu}(\boldsymbol{\xi})| \leq \frac{1}{2}$  when  $|\boldsymbol{\xi}| \leq r$ . Then there is an  $\ell$  for which  $\ell(0) = 0$ ,  $|\ell(\boldsymbol{\xi})| \leq 2$ , and  $\hat{\mu}(\boldsymbol{\xi}) = e^{\ell(\boldsymbol{\xi})}$  if  $|\boldsymbol{\xi}| \leq r$ . Using  $\log z = -\sum_{n=1}^{\infty} \frac{(1-z)^n}{n}$  when  $|1-z| < 1$ , one sees that  $|\ell(\boldsymbol{\xi})| \leq 2$  for  $|\boldsymbol{\xi}| < r$ .

Since  $\widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi})^n = \hat{\mu}(\boldsymbol{\xi})$ ,  $\widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi}) \neq 0$  when  $|\boldsymbol{\xi}| \leq r$ , and so, by uniqueness, it must be that  $\widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi}) = e^{\frac{\ell(\boldsymbol{\xi})}{n}}$  for  $|\boldsymbol{\xi}| \leq r$ , and therefore  $|1 - \widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi})| \leq \frac{2}{n}$  when  $|\boldsymbol{\xi}| \leq r$ . Hence, by (18.2), for any  $R > 0$ ,

$$\mu_{\frac{1}{n}}(\{\mathbf{y} : |\mathbf{y}| \geq R\}) \leq \frac{2N}{ns(rN^{-\frac{1}{2}}R)},$$

and so

$$|1 - \widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi})| \leq \int |1 - e^{i\boldsymbol{\xi} \cdot \mathbf{y}}| \mu_{\frac{1}{n}}(d\mathbf{y}) \leq |\boldsymbol{\xi}|R + 2\mu_{\frac{1}{n}}(\{\mathbf{y} : |\mathbf{y}| \geq R\}) \leq |\boldsymbol{\xi}|R + \frac{2N}{ns(rN^{-\frac{1}{2}}R)}.$$

Given  $\boldsymbol{\xi} \neq \mathbf{0}$ , take  $R = \frac{1}{4r|\boldsymbol{\xi}|}$ , choose  $n$  so that  $\frac{2N}{ns(rN^{-\frac{1}{2}}R)} \leq \frac{1}{4}$ , and conclude that  $|1 - \widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi})| \leq \frac{1}{2}$  and therefore  $|\hat{\mu}(\boldsymbol{\xi})| \geq 2^{-n}$ . This proves that  $\hat{\mu}$  never vanishes and therefore that  $\hat{\mu} = e^{\ell}$ . In addition, by using the fact that  $\lim_{t \searrow 0} \frac{s(t)}{t^2} = \frac{1}{6}$ , the preceding line of reasoning shows that there is a  $C < \infty$  such that  $|1 - e^{\frac{\ell(\boldsymbol{\xi})}{n}}| \leq \frac{1}{2}$  when  $n \geq C|\boldsymbol{\xi}|^2$ , and therefore  $\frac{|\ell(\boldsymbol{\xi})|}{1+|\boldsymbol{\xi}|^2}$  is bounded.

Knowing that  $\widehat{\mu_{\frac{1}{n}}} = e^{\frac{\ell}{n}}$  and that  $\ell$  has at most quadratic growth, one knows that

$$\ell(\boldsymbol{\xi}) = \lim_{n \rightarrow \infty} n(\widehat{\mu_{\frac{1}{n}}}(\boldsymbol{\xi}) - 1).$$

Thinking of  $\ell$  as a tempered distribution, the challenge is to describe the distribution of which it is the Fourier transform. Thus, set  $u = \check{\ell}$ . Then

$$\begin{aligned} (2\pi)^N \langle \varphi, u \rangle &= \langle \hat{\varphi}, \ell \rangle = \lim_{n \rightarrow \infty} n \int \hat{\varphi}(\boldsymbol{\xi}) \left( \int (e^{-i\boldsymbol{\xi} \cdot \mathbf{x}} - 1) \mu_{\frac{1}{n}}(d\mathbf{x}) \right) d\boldsymbol{\xi} \\ &= \lim_{n \rightarrow \infty} n \int \left( \int (e^{-i\boldsymbol{\xi} \cdot \mathbf{x}} - 1) \hat{\varphi}(\boldsymbol{\xi}) d\boldsymbol{\xi} \right) \mu_{\frac{1}{n}}(d\mathbf{x}) \\ &= (2\pi)^N \lim_{n \rightarrow \infty} n \int (\varphi(\mathbf{x}) - \varphi(\mathbf{0})) \mu_{\frac{1}{n}}(d\mathbf{x}), \end{aligned}$$

and so

$$\langle \varphi, u \rangle = \lim_{n \rightarrow \infty} n \int (\varphi(\mathbf{x}) - \varphi(\mathbf{0})) \mu_{\frac{1}{n}}(d\mathbf{x}).$$

In particular,  $u$  satisfies the obvious  $\mathbb{R}^N$  analog of the minimum principle in (14.4). Thus, by the  $\mathbb{R}^N$ -analog of Theorem 14.7, we know that

$$\begin{aligned} \langle \varphi, u \rangle &= \frac{1}{2} \sum_{i,j=1}^N A_{i,j} \partial_{x_i} \partial_{x_j} \varphi(\mathbf{0}) + \sum_{i=1}^N b_i \partial_{x_i} \varphi(\mathbf{0}) \\ &\quad + \int \left( \varphi(y) - \varphi(\mathbf{0}) - \mathbf{1}_{B(0,1)}(y) (y, \nabla \varphi(\mathbf{0}))_{\mathbb{R}^N} \right) M(dy), \end{aligned}$$

where  $(\mathbf{b}, A, M)$  is a Lévy system.

To compute the Fourier transform of  $u$ , introduce the operator

$$\begin{aligned} \mathcal{L}_{(\mathbf{b}, A, M)} \varphi(\mathbf{x}) &= \frac{1}{2} \sum_{i,j=1}^N A_{i,j} \partial_{x_i} \partial_{x_j} \varphi(\mathbf{x}) + \sum_{i=1}^N b_i \partial_{x_i} \varphi(\mathbf{x}) \\ &\quad + \int \left( \varphi(\mathbf{x} + \mathbf{y}) - \varphi(\mathbf{x}) - (\mathbf{b}, \nabla \varphi(\mathbf{x}))_{\mathbb{R}^N} \right) M(d\mathbf{y}). \end{aligned}$$

What we have shown is that  $\langle \varphi, u \rangle = \mathcal{L}_{(\mathbf{b}, A, M)} \varphi(\mathbf{0})$ . Using  $\widehat{\partial_{x_j} \varphi}(\boldsymbol{\xi}) = -i \xi_j \widehat{\varphi}(\boldsymbol{\xi})$  and Fubini's theorem, one sees that

$$\widehat{\mathcal{L}_{(\mathbf{b}, A, M)} \varphi}(\boldsymbol{\xi}) = \widehat{\varphi}(\boldsymbol{\xi}) \ell_{(\mathbf{b}, A, M)}(-\boldsymbol{\xi}),$$

where

$$\begin{aligned} \ell_{(\mathbf{b}, A, M)}(\boldsymbol{\xi}) &= \log f_{(\mathbf{b}, A, M)} \\ &= -\frac{1}{2} (\boldsymbol{\xi}, A \boldsymbol{\xi}) + i (\mathbf{b}, \boldsymbol{\xi})_{\mathbb{R}^N} + \int \left( e^{i(\boldsymbol{\xi}, \mathbf{y})} - 1 - i \mathbf{1}_{B(\mathbf{0}, 1)}(\mathbf{y}) (\boldsymbol{\xi}, \mathbf{y}) \right) M(d\mathbf{y}). \end{aligned}$$

Hence, by Parseval's identity,

$$\langle \widehat{\varphi}, \ell \rangle = (2\pi)^N \langle \varphi, u \rangle = (2\pi)^N \mathcal{L}_{(\mathbf{b}, A, M)} \varphi(\mathbf{0}) = \langle \widehat{\varphi}, \ell_{(\mathbf{b}, A, M)}(\boldsymbol{\xi}) \rangle,$$

and so  $\ell = \ell_{(\mathbf{b}, A, M)}$ .

We will now use (19.1) to prove some properties of the associated measures based on properties of the Lévy system. Use  $\mu_{(\mathbf{b}, A, M)} \in \mathcal{S}(\mathbb{R}^N; \mathbb{C})^*$  to denote the probability measure of which  $f_{(\mathbf{b}, A, M)}$  is the Fourier transform, and set  $\mu_t = \mu_{(t\mathbf{b}, tA, tM)}$  for  $t > 0$ . Then

$$(2\pi)^N \partial_t \langle \varphi, \mu_t \rangle = \langle \widehat{\varphi}, \ell_{(\mathbf{b}, A, M)} f_{(t\mathbf{b}, tA, tM)} \rangle = (2\pi)^N \langle \mathcal{L}_{(\mathbf{b}, A, M)} \varphi, \mu_t \rangle.$$

That is, we have shown that

$$(19.3) \quad \partial_t \langle \varphi, \mu_{(t\mathbf{b}, tA, tM)} \rangle = \langle \mathcal{L}_{(\mathbf{b}, A, M)} \varphi, \mu_{(t\mathbf{b}, tA, tM)} \rangle.$$

**Theorem 19.1.** *If either  $A$  is non-degenerate or  $M(G) > 0$  for all non-empty open sets  $G \subseteq \mathbb{R}^N \setminus \{\mathbf{0}\}$ , then  $\mu_{(\mathbf{b}, A, M)}(G) > 0$  for all non-empty open sets  $G \subseteq \mathbb{R}^N$ .*

*Proof.* First observe that  $\mu_{(\mathbf{b}, A, M)} = \delta_{\mathbf{b}} * \mu_{(\mathbf{0}, A, M)}$ , and therefore we can assume that  $\mathbf{b} = \mathbf{0}$ . Next note that  $\mu_{(\mathbf{0}, A, M)} = \gamma_A * \mu_{(\mathbf{0}, 0, M)}$  where  $\gamma_A$  is the distribution of  $x \rightsquigarrow A^{\frac{1}{2}}x$  under  $\gamma$ , and so, if  $A$  is non-degenerate and therefore  $\gamma_A$  has a strictly positive density,  $\mu_{(\mathbf{0}, A, M)}$  does also.

Now assume that  $\mathbf{b} = \mathbf{0}$ ,  $A = 0$ , and  $M(G) > 0$  for all open  $\emptyset \neq G \subseteq \mathbb{R}^N \setminus \{\mathbf{0}\}$ . Given an open  $G \neq \emptyset$ , choose an  $\eta \in C^\infty(\mathbb{R}^N; [0, 1])$  which is strictly positive on  $G$  and vanishes off of  $G$ . Then

$$\begin{aligned} \mathcal{L}_{(\mathbf{0}, 0, M)} \eta(\mathbf{x}) &= \int \left( \eta(\mathbf{x} + \mathbf{y}) - \eta(\mathbf{x}) - \mathbf{1}_{B(\mathbf{0}, 1)}(\mathbf{y}) (\nabla \eta(\mathbf{x}), \mathbf{y})_{\mathbb{R}^N} \right) M(d\mathbf{y}) \\ &= \int \eta(\mathbf{x} + \mathbf{y}) M(d\mathbf{y}) > 0 \end{aligned}$$

if  $\mathbf{x} \notin G$ . Hence, if  $f(t) = \langle \eta, \mu_{(\mathbf{0}, 0, tM)} \rangle$ , then  $f \geq 0$  and, by (19.3),  $\mu_{(\mathbf{0}, 0, tM)}(G) = 0 \implies f'(t) > 0$ . But  $\mu_{(\mathbf{0}, 0, tM)}(G) = 0$  also implies that  $f(t) = 0$ , which, by the first derivative test, is possible only if  $f'(t) = 0$ . Hence  $f(t) > 0$  for all  $t > 0$ , and so  $\mu_{(\mathbf{0}, 0, M)}(G) > 0$ .  $\square$

**Theorem 19.2.** *If  $N = 1$ , then  $\mu_{(b,A,M)}((-\infty, 0)) = 0$  if and only if*

$$(19.4) \quad A = 0, \quad M((-\infty, 0)) = 0, \quad \text{and} \quad \int_{|y| < 1} y M(dy) \leq b.$$

*Proof.* Observe that,

$$\{\mathbf{x} \in \mathbb{R}^n : x_j < 0 \text{ for } 1 \leq j \leq n\} \subseteq \left\{ \mathbf{x} \in \mathbb{R}^n : \sum_{j=1}^n x_j < 0 \right\},$$

and therefore  $\mu_{\frac{1}{n}}((-\infty, 0))^n \leq \mu^{*n}((-\infty, 0))$  for any  $\mu \in \mathbf{M}_1(\mathbb{R})$ .

Now assume that  $\mu_{(b,A,M)}((-\infty, 0)) = 0$ . Since  $\mu_{(b,A,M)} = \gamma_A * \mu_{(b,0,M)}$  and  $\gamma_A(G) > 0$  for all open  $G \neq \emptyset$  unless  $A = 0$ , it follows that  $A = 0$ . Next observe that  $f_{(b,0,M)}$  has a bounded analytic extension to  $\{\zeta \in \mathbb{C} : \Re \zeta < 0\}$ , and therefore  $M((-\infty, 0))$  must be 0. Finally, to prove the inequality in (19.4), set  $\mu_{\frac{1}{n}} = \mu_{(\frac{b}{n}, 0, \frac{M}{n})}$ . Since  $\mu_1 = \mu_{\frac{1}{n}}^{*n}$ , the observation above shows that  $\mu_{\frac{1}{n}}((-\infty, 0)) = 0$ , and therefore, if  $\varphi \geq 0$  on  $[0, \infty)$  and  $\varphi(0) = 0$ , then, by (19.3),

$$\mathcal{L}_{(b,0,M)}\varphi(0) = \lim_{n \rightarrow \infty} n(\langle \varphi, \mu_{\frac{1}{n}} \rangle - \varphi(0)) \geq 0,$$

and so

$$b\varphi'(0) + \int (\varphi(y) - \mathbf{1}_{(-1,1)}(y)y\varphi'(0)) M(dy) \geq 0.$$

Now choose  $\eta \in C^\infty(\mathbb{R}; [0, 1])$  so that  $\eta = 1$  on  $[-\frac{1}{2}, \frac{1}{2}]$  and  $\eta = 0$  off  $(-1, 1)$ , and, for  $r \in (0, 1)$ , set  $\varphi_r(x) = y\eta_r(y)$  where  $\eta_r(y) = \eta(\frac{y}{r})$ . By the preceding applied to  $\varphi_r$ ,

$$b - \int (\mathbf{1}_{(-1,1)}(y) - \eta_r(y))y M(dy) \geq 0,$$

and so

$$\int_{(r,1)} y M(dy) \leq b \text{ for all } r \in (0, 1).$$

Finally, assume that (19.4) holds, and set  $M_r(dy) = \mathbf{1}_{[r,\infty)}(y) M(dy)$  and  $b_r = b - \int y M_r(dy)$  for  $r > 0$ . Then (19.4) holds for  $(b, 0, M_r)$  and (cf. (19.2))  $\mu_{(b,0,M_r)} = \delta_{b_r} * \pi_{M_r}$ , from which it is clear that  $\mu_{(b,0,M_r)}((-\infty, 0)) = 0$ . Therefore, since  $\mu_{(b,0,M_r)} \xrightarrow{w} \mu_{(b,0,M)}$ ,  $\mu_{(b,0,M)}((-\infty, 0)) = 0$ .  $\square$

**Exercise 19.3.** If  $M$  is symmetric, show that the integral in (19.1) can be replaced by

$$\int (\cos(\xi, \mathbf{y})_{\mathbb{R}^N} - 1) M(d\mathbf{y}).$$

If  $M$  is invariant under orthogonal transformations, show that the integral in (19.1) is equal to

$$\int_{\mathbb{S}^{N-1}} (\cos(\mathbf{e}, \omega)_{\mathbb{R}^N} - 1) \lambda_{\mathbb{S}^{N-1}}(d\omega) |\xi|^\alpha,$$

where  $\mathbf{e} \in \mathbb{S}^{N-1}$  and  $\alpha \in (0, 2)$ . In particular, by combining this with part (iii) of Exercise 4.3, conclude that  $e^{-|\xi|^\alpha}$  is a characteristic function if and only if  $\alpha \in [0, 2]$ .

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