

Exercises Chapter 6: Infinitely Divisible Distributions

① X, Y independ. rand. variables with densities f_X and f_Y , resp. Show that $X+Y$ has density

$$f_{X+Y}(x) = \int_{\mathbb{R}^d} f_X(x-y) f_Y(y) dy, \quad x \in \mathbb{R}^d.$$

We assume that X, Y take values in \mathbb{R}^d . Consider the $\mathbb{R}^d \times \mathbb{R}^d$ -valued rand. variable $(U, V) := (X+Y, Y) = g(X, Y)$.

Thanks to the independence of X and Y , we have that

$$f_{X,Y}(x,y) = f_X(x) f_Y(y).$$

Next note that the function $g: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}^d \times \mathbb{R}^d$ is a bijection and $g^{-1}: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}^d \times \mathbb{R}^d$. Moreover,

$$\mathcal{J}_{g^{-1}}(u,v) = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \Rightarrow |\det \mathcal{J}_{g^{-1}}(u,v)| = 1.$$

Hence,

$$\begin{aligned} f_{U,V}(u,v) &= f_{X,Y}(g^{-1}(u,v)) |\det \mathcal{J}_{g^{-1}}(u,v)| \mathbb{1}_{g(\mathbb{R}^d \times \mathbb{R}^d)}(u,v) \\ &= f_X(u-v) f_Y(v) \mathbb{1}_{\mathbb{R}^d \times \mathbb{R}^d}(u,v) \end{aligned}$$

We are interested in the marginal density of U , which is obtained by integrating $f_{U,V}$ with respect to v .

$$f_U(u) = \int_{\mathbb{R}^d} f_X(u-v) f_Y(v) dv$$

which is the desired result.

2) $X \sim N(0, 1)$. Show $\phi_X(u) = e^{-\frac{u^2}{2}}$.

As $f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$ is symmetric with respect to 0, we have

$$\begin{aligned} \phi_X(u) &= E[e^{i u X}] = \int_{\mathbb{R}} (\cos(ux) + i \sin(ux)) f_X(x) dx \\ &= \int_{\mathbb{R}} \cos(ux) f_X(x) dx \end{aligned}$$

Taking the derivative with respect to u (under the integral sign)

we get

$$\left\{ \begin{aligned} \phi'_X(u) &= - \int_{\mathbb{R}} x \sin(ux) f_X(x) dx = \begin{cases} w = \sin(ux), dw = u \cos(ux) dx \\ dv = -x f_X(x) dx, v = f_X(x) \end{cases} \\ &= \left[\sin(ux) f_X(x) \right]_{-\infty}^{+\infty} - \int_{-\infty}^{+\infty} u \cos(ux) f_X(x) dx \\ &= -u \phi_X(u) \end{aligned} \right.$$

$$\phi_X(0) = 1$$

$$\Rightarrow \left\{ \begin{aligned} \frac{\phi'_X(u)}{\phi_X(u)} &= -u \Rightarrow \log(\phi_X(u)) = K - \frac{u^2}{2} \\ \phi_X(0) &= 1 \end{aligned} \right. \Downarrow \phi_X(u) = e^K e^{-\frac{u^2}{2}} = e^{-\frac{u^2}{2}}$$

$\phi_X(0) = 1 \Rightarrow K = 0$

3) $X \sim \text{Pois}(\lambda)$. Show that $\varphi_X(u) = \exp(\lambda(e^{iu} - 1))$.

Recall that the law of a Poisson distribution is supported on the non-negative integers \mathbb{Z}_+ and it is given by

$$P_X(k) = e^{-\lambda} \frac{\lambda^k}{k!} \quad k=0, 1, 2, \dots$$

$$\varphi_X(u) = E[e^{iuX}] = \sum_{k \geq 0} e^{iu k} P_X(k)$$

$$= \sum_{k \geq 0} e^{iu k} e^{-\lambda} \frac{\lambda^k}{k!}$$

$$= e^{-\lambda} \sum_{k \geq 0} \frac{(e^{iu} \lambda)^k}{k!}$$

$$= e^{-\lambda} e^{e^{iu} \lambda} = e^{\lambda(e^{iu} - 1)}$$

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$$\boxed{4} \quad X \sim \Gamma(m, \lambda), m \in \mathbb{N}, \lambda > 0, f_X(x) = \frac{\lambda^m x^{m-1} e^{-\lambda x}}{(m-1)!}, x > 0$$

Show that

$$(P_X)^{1/m} = P_Y \quad \text{with } Y \sim \text{Exp}(\lambda).$$

$(P_X)^{1/m}$ is such that $(P_X^{1/m})^{*m} = P_X$.

Hence, we only need to show that $P_Y^{*m} = P_X$.

When $m=1$, the result is clear because $\Gamma(1, \lambda) \sim \text{Exp}(\lambda)$.

The result follows by induction. Let $A=(0, a], a > 0$.

$$\begin{aligned} P_Y^{*m}(A) &= P_Y^{*(m-1)} * P_Y(A) = \int_{\mathbb{R} \times \mathbb{R}} \mathbb{1}_A(x+y) P_Y^{*(m-1)} \otimes P_Y(dx, dy) \\ &\quad \uparrow \\ &\quad \text{def of } * \text{ product} \\ &= \int_{\mathbb{R} \times \mathbb{R}} \mathbb{1}_A(x+y) \frac{\lambda^{m-1} x^{m-2} e^{-\lambda x}}{(m-2)!} \mathbb{1}_{(0, +\infty)}(x) \lambda e^{-\lambda y} \mathbb{1}_{(0, +\infty)}(y) dx \otimes dy \\ &= \int_D \frac{\lambda^m x^{m-2} e^{-\lambda(x+y)}}{(m-2)!} dx \otimes dy \quad \left(\begin{aligned} D &= \{(x, y) \in \mathbb{R}^2 : 0 \leq x+y \leq a\} \\ &= \{0 \leq x \leq a, 0 \leq y \leq a-x\} \end{aligned} \right) \\ &= \int_0^a \frac{\lambda^{m-1} x^{m-2} e^{-\lambda x}}{(m-2)!} \left(\int_0^{a-x} \lambda e^{-\lambda y} dy \right) dx \\ &= \int_0^a \frac{\lambda^{m-1} x^{m-2} e^{-\lambda x}}{(m-2)!} (1 - e^{-\lambda(a-x)}) dx = (*) \end{aligned}$$

To get the density of the law P_Y^{*m} we need to differentiate with respect to a .

$$\frac{d}{da} (*) = \frac{\lambda^{m-1} a^{m-2} e^{-\lambda a}}{(m-2)!} (1 - e^{-\lambda(a-a)}) + \int_0^a \frac{\lambda^{m-1} x^{m-2} e^{-\lambda x}}{(m-2)!} \cdot \lambda e^{-\lambda(a-x)} dx$$

$$= \int_0^a \lambda^m \frac{x^{m-2}}{(m-2)!} e^{-\lambda x} dx = \lambda^m e^{-\lambda a} \int_0^a \frac{x^{m-2}}{(m-2)!} dx$$

$$= \lambda^m e^{-\lambda a} \frac{a^{m-1}}{(m-1)!}$$

which is the density of $X \sim \Gamma(m, \lambda)$

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μ is inf. div. (\iff) For each $n \in \mathbb{N}$, $\exists (\mu^{1/n}) \in \mathcal{M}_2(\mathbb{R}^d)$
for which $\int_{\mu} (u) = \left(\int_{\mu^{1/n}} (u) \right)^n \quad u \in \mathbb{R}^d$.

$\bar{\mu}$ is inf. div. $(\stackrel{\text{Def}}{\iff})$ For each $n \in \mathbb{N}$, $\exists \mu^{1/n} \in \mathcal{M}_2(\mathbb{R}^d)$
such that $(\mu^{1/n})^{*n} = \mu$.

Therefore we only need to prove that

$$\int_{(\mu^{1/n})^{*n}} (u) = \left(\int_{\mu^{1/n}} (u) \right)^n$$

Using Proposition 4.3. and induction it follows that

$$\forall n \in \mathbb{N}, \int_{\mathbb{R}^d} f(z) (\mu_1 * \dots * \mu_n)(dz) = \int_{\mathbb{R}^d \times \dots \times \mathbb{R}^d} f(x_1 + \dots + x_n) \mu_1 \otimes \dots \otimes \mu_n(dx_1, \dots, dx_n)$$

for $f \in \mathcal{B}_c(\mathbb{R}^d)$.

Applying this result to $e^{i\langle u, z \rangle}$ and $\mu_i = \mu^{1/n} \quad i=1, \dots, n$.

we get

$$\begin{aligned} \int_{(\mu^{1/n})^{*n}} (u) &= \int_{\mathbb{R}^d} e^{i\langle u, z \rangle} (\mu^{1/n})^{*n}(dz) \\ &= \int_{\mathbb{R}^d \times \dots \times \mathbb{R}^d} e^{i\langle u, x_1 + \dots + x_n \rangle} (\mu^{1/n})^{\otimes n}(dx_1, \dots, dx_n) \end{aligned}$$

$$\begin{aligned} \text{Fubini} \quad &\searrow = \int_{\mathbb{R}^d} e^{i\langle u, x_1 \rangle} \mu^{1/n}(dx_1) \dots \int_{\mathbb{R}^d} e^{i\langle u, x_n \rangle} \mu^{1/n}(dx_n) \\ &= \int_{\mu^{1/n}} (u) \dots \int_{\mu^{1/n}} (u) = \left(\int_{\mu^{1/n}} (u) \right)^n. \end{aligned}$$

6) Let X and Y be two indep. inf. divis. r.v.
 Prove that $X+Y$ is also inf. divisible.

By Corollary 4.5 with $f = \mathbb{1}_A$, $A \in \mathcal{B}(\mathbb{R}^d)$ we can conclude that

$$P_{X+Y} = P_X * P_Y.$$

As X and Y are inf. divis., for each $n \in \mathbb{N}$ there exists $P_X^{1/n}, P_Y^{1/n} \in M_2(\mathbb{R}^d)$ such that $P_X = (P_X^{1/n})^{*n}$ and $P_Y = (P_Y^{1/n})^{*n}$.

Therefore,

$$P_{X+Y} = P_X * P_Y = (P_X^{1/n})^{*n} * (P_Y^{1/n})^{*n} = (P_X^{1/n} * P_Y^{1/n})^{*n}$$

By Prop. 4.3 (2) and (3) commutativity and associativity of the conv. product.

By Prop. 4.2 we know that $P_X^{1/n} * P_Y^{1/n} \in M_2(\mathbb{R}^d)$.

Hence, we can conclude by using Prop. 4.7, which shows the equivalence between Z being inf. divis. and P_Z having a convolution with itself that is itself a random variable.

In our case $Z = X+Y$ and $P_{X+Y}^{1/n} = P_X^{1/n} * P_Y^{1/n}$.

7) ν a Lévy measure on \mathbb{R}^d . Then, for all $\varepsilon > 0$
 $\nu(\{y \in \mathbb{R}^d : |y|^2 > \varepsilon\}) < +\infty$ and conclude
 that ν is σ -finite

Since ν is a Lévy measure we know that

$$\int_{\mathbb{R}^d} (1 \wedge |y|^2) \nu(dy) < +\infty \iff (1) \int_{|y|^2 < 1} |y|^2 \nu(dy) < +\infty \text{ and } (2) \int_{|y|^2 \geq 1} \nu(dy) < +\infty.$$

If $\varepsilon \geq 1$ then (2) implies the result.

If $\varepsilon < 1$ then

$$\nu(\{y \in \mathbb{R}^d : |y|^2 > \varepsilon\}) = \int_{|y|^2 > \varepsilon} \nu(dy) = \int_{\varepsilon < |y|^2 < 1} \nu(dy) + \int_{|y|^2 \geq 1} \nu(dy)$$

The second integral is finite by (2).

$$\int_{\varepsilon < |y|^2 < 1} \nu(dy) = \frac{1}{\varepsilon} \int_{\varepsilon < |y|^2 < 1} \varepsilon \nu(dy) < \frac{1}{\varepsilon} \int_{\varepsilon < |y|^2 < 1} |y|^2 \nu(dy)$$

$$\overset{|y|^2 \geq 0}{\uparrow} < \frac{1}{\varepsilon} \int_{|y|^2 < 1} |y|^2 \nu(dy) < +\infty \quad \uparrow \text{ by (2).}$$

and $\{\varepsilon < |y|^2 < 1\} \subset \{|y|^2 < 1\}$

$$\mathbb{R}^d_0 = \bigcup_{n \geq 1} A_n, \text{ where } A_1 = \{|y|^2 > 1\}, A_n = \{\frac{1}{n} < |y|^2 \leq \frac{1}{n-2}\}, n \geq 2$$

Clearly $\nu(A_1) < +\infty$ and
 $\nu(A_n) \leq \nu(\{|y|^2 > 1/n\}) < +\infty$.
 $A_n \subset \{|y|^2 > 1/n\}$

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Show that the Lévy symbol ρ of an inf-divis. distribution μ is continuous at every $u \in \mathbb{R}^d$.

$$\rho(u) = i \langle \xi, u \rangle - \frac{1}{2} \langle u, Au \rangle + \int_{\mathbb{R}^d} (e^{i \langle u, y \rangle} - 1 - i \langle u, y \rangle) \mathbb{1}_{B_2^c(c)}(y) \nu(dy)$$

As $\langle \cdot, \cdot \rangle$ is continuous we only need to prove that

$$\Psi(u) = \int_{\mathbb{R}^d} (e^{i \langle u, y \rangle} - 1 - i \langle u, y \rangle) \mathbb{1}_{B_2^c(c)}(y) \nu(dy)$$

is continuous with respect to u , that is, for any $u_n \xrightarrow[n \rightarrow \infty]{} u$

$$\Psi(u) = \lim_{n \rightarrow \infty} \Psi(u_n)$$

As $\xi(u, y)$ is continuous with respect to u , the result follows from dominated convergence. In order to apply dominated convergence we need to bound

$$|\xi(u_n, y)| \text{ by any } \tilde{\xi}(u) \in L^1(\nu)$$

In the proof of Theorem 4.14 we saw that

$$\begin{aligned} |\xi(u_n, y)| &\leq \frac{1}{2} |u_n|^2 |y|^2 \mathbb{1}_{B_2(c)}(y) + 2 \mathbb{1}_{B_2^c(c)}(y) \\ &\leq \frac{1}{2} \sup_n |u_n|^2 |y|^2 \mathbb{1}_{B_2(c)}(y) + 2 \mathbb{1}_{B_2^c(c)}(y) \in L^1(\nu) \end{aligned}$$

Note that $\sup_n |u_n|^2 < +\infty$, because $u_n \xrightarrow[n \rightarrow \infty]{} u$.

9 Show that the Lévy symbol η of an inf. div. distribution μ satisfies

$$|\eta(u)| \leq C(1 + |u|^2) \quad u \in \mathbb{R}^d, \quad C > 0.$$

We have

$$\eta(u) = i \langle \gamma, u \rangle - \frac{1}{2} \langle u, Au \rangle + \int_{\mathbb{R}_0^d} (e^{i \langle u, y \rangle} - 1 - i \langle u, y \rangle \mathbb{1}_{B_1(c)}(y)) \nu(dy)$$

By the triangle inequality and Cauchy-Schwarz

$$|i \langle \gamma, u \rangle - \frac{1}{2} \langle u, Au \rangle| \leq |\langle \gamma, u \rangle| + \frac{1}{2} |\langle u, Au \rangle|$$

$$\begin{aligned} \circ \quad |u| &\leq 1 + |u|^2 && \leq |\gamma| |u| + \frac{1}{2} |u| |Au| \\ \circ \quad |Au| &\leq \|A\|_2 |u| && \leq |\gamma| + \left(\frac{1}{2} \|A\|_2 + 1\right) |u|^2 \end{aligned}$$

In Theorem 4.14 we showed that

$$|e^{i \langle u, y \rangle} - 1 - i \langle u, y \rangle \mathbb{1}_{B_1(c)}(y)| \leq \frac{1}{2} |u|^2 |y|^2 \mathbb{1}_{B_1(c)}(y) + 2 \mathbb{1}_{B_1^c(c)}(y)$$

and, therefore,

$$\left| \int_{\mathbb{R}_0^d} (e^{i \langle u, y \rangle} - 1 - i \langle u, y \rangle \mathbb{1}_{B_1(c)}(y)) \nu(dy) \right| \leq \frac{1}{2} \left(\int_{B_1(c)} |y|^2 \nu(dy) \right) |u|^2 + 2 \int_{B_1^c(c)} \nu(dy)$$

The result follows with

$$C = \max \left(|\gamma| + 2 \int_{B_1^c(c)} \nu(dy), \frac{1}{2} (\|A\|_2 + 1) + \frac{1}{2} \int_{B_1(c)} |y|^2 \nu(dy) \right)$$

16 If μ is I.D. then, for every $t \in \mathbb{R}_+$, μ^t is well defined and I.D.

We have a distribution $\mu^{1/n}$ for any $n \in \mathbb{N}$, because μ is I.D.

It is I.D., since $\int \mu^{1/n}(z) = \int \mu(z)^{1/n} = (\int \mu(z)^{1/(nk)})^k$ for any $k \in \mathbb{N}$.

Hence for any $m, n \in \mathbb{N}$, $\mu^{m/n}$ is also I.D.

because

$$\mu^{m/n} = \mu^{1/n} * \dots * \mu^{1/n} \quad (m \text{ times})$$

and the convolution product of I.D. distributions is I.D. (See Exercise 6).

For any real number t , choose rational numbers $r_n \rightarrow t$.

Then,

$$\int \mu^{r_n}(z) = \int \mu(z)^{r_n} \xrightarrow{n \rightarrow \infty} \int \mu(z)^t \quad \forall z \in \mathbb{R}^d.$$

$\int \mu(z)^t$ is well defined but does not need to be the characteristic function of a distribution.

Since $\int \mu(z)$ is a continuous function (due to the construction of the distinguished logarithm and $\int \mu(z) \neq 0$), it is continuous at 0, and, by Levy's continuity theorem

Theorem 2.28, we know that $\int \mu(z)^t$ is indeed a charact. function. By Bochner's theorem $\mu^{r_n} \xrightarrow{n \rightarrow \infty} \mu^t$ and μ^t is I.D. by Prop. 6.20.