Probability and Measure, Hand-out 4 Fourier transforms: uniqueness and inversion

A finite Borel measure is determined uniquely by its Fourier transform. There is a formula for the measure in terms of its transform, and this is quite simple in the integrable case.

Let X be a random variable taking values in \mathbb{R}^n . Its characteristic function ϕ_X is the Fourier transform of its law μ_X , which is to say that

$$\phi_X(u) = \mathbb{E}(e^{i\langle u, X \rangle}), \qquad u \in \mathbb{R}^n.$$

The 'heat kernel' function is given by

$$p(t, x, y) = \frac{1}{(2\pi t)^{n/2}} e^{-\frac{1}{2}|y-x|^2/t}, \qquad t > 0, \ x, y \in \mathbb{R}^n.$$

Recall the identity

$$e^{-\frac{1}{2}w^2} = \int_{\mathbb{R}} e^{iuw} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} du.$$

Write $w = (x - y)/\sqrt{t}$ and make a change of variable to find when n = 1 that

$$p(t, x, y) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{iux} e^{-\frac{1}{2}u^2 t} e^{-iuy} \, du,$$

whence, for $n \ge 1$,

$$p(t,x,y) = \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{i\langle u,x\rangle} e^{-\frac{1}{2}|u|^2 t} e^{-i\langle u,y\rangle} du.$$

Theorem. Let X be a random variable taking values in \mathbb{R}^n . The law μ_X of X is uniquely determined by knowledge of its characteristic function ϕ_X . Moreover, if ϕ_X is integrable, then X has a density function f_X where

$$f_X(z) = \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} \phi_X(u) e^{-i\langle u, z \rangle} \, du, \qquad z \in \mathbb{R}^n.$$

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Proof. Let Y be a standard Gaussian random variable in \mathbb{R}^n , independent of X, and let $g: \mathbb{R}^n \to \mathbb{R}$ be bounded and measurable. Then, for t > 0,

$$\begin{split} \mathbb{E}(g(X+Y\sqrt{t})) &= \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} g(x+y\sqrt{t}) \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2}|y|^2} \, dy \, \mu_X(dx) \\ &= \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} p(t,x,z) g(z) \, dz \, \mu_X(dx) \\ &= \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^n} \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{i\langle u,x \rangle} e^{-\frac{1}{2}|u|^2 t} e^{-i\langle u,z \rangle} \, du \, \mu_X(dx) \right) g(z) \, dz \\ &= \int_{\mathbb{R}^n} \left(\frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} \phi_X(u) e^{-\frac{1}{2}|u|^2 t} e^{-i\langle u,z \rangle} \, du \right) g(z) \, dz, \end{split}$$

where we have used Fubini's theorem and the change of variables $z = x + y\sqrt{t}$.

By this formula, ϕ_X determines $\mathbb{E}(g(X+Y\sqrt{t}))$ for any given g. For any bounded continuous function g, we have by bounded convergence that

$$\mathbb{E}(g(X+Y\sqrt{t})) \to \mathbb{E}(g(X)) \qquad \text{as } t \downarrow 0,$$

whence ϕ_X determines $\mathbb{E}(g(X))$ for any given g. It follows that ϕ_X determines μ_X .

If ϕ_X is integrable and if g is continuous with compact support, then $|\phi_X(u)| \cdot |g(z)| \in L^1(\mathbb{R}^{2n}, \mathcal{B}, \lambda)$. By dominated convergence,

$$\mathbb{E}(g(X+Y\sqrt{t})) = \int_{\mathbb{R}^n} \left(\frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} \phi_X(u) e^{-\frac{1}{2}|u|^2 t} e^{-i\langle u, z \rangle} \, du\right) g(z) \, dz$$
$$\to \int_{\mathbb{R}^n} \left(\frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} \phi_X(u) e^{-i\langle u, z \rangle} \, du\right) g(z) \, dz,$$

as $t \downarrow 0$. Hence X has the claimed density function.