

Characteristic Functions

Def: Let $\mu \in \text{Prob}(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$.

For $\xi \in \mathbb{R}^d$, define

$$\hat{\mu}(\xi) := \int_{\mathbb{R}^d} e^{i\xi \cdot x} \mu(dx)$$

the Fourier transform of μ .

If $X: (\Omega, \mathcal{F}) \rightarrow (\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$ is a random vector,
its characteristic function $\varphi_X: \mathbb{R}^d \rightarrow \mathbb{C}$ is

$$\varphi_X(\xi) = \hat{\mu}_X(\xi)$$

Prop: $\mu \mapsto \hat{\mu}$ is injective: if $\hat{\mu}(\xi) = \hat{\nu}(\xi) \forall \xi \in \mathbb{R}^d$, then $\mu = \nu$.

Pf. The final corollary of [Lecture 24.1]:

If $\int e_{\xi} d\mu = \int e_{\xi} d\nu \forall \xi \in \mathbb{R}^d$, then $\mu = \nu$



Thus, in principle, we can recover μ from $\hat{\mu}$.

I.e. μ_{Σ} is determined by φ_{Σ} (and vice versa).

/ Compare: the moment generating function over \mathbb{R} :

$$M_X(t) = \mathbb{E}[e^{tX}], \quad \varphi_X(\xi) = \mathbb{E}[e^{i\xi X}]$$

↑
requires exponential
integrability to be defined

↑
always defined,
and carries at least as much info.

Prop: (Basic Properties of the Fourier Transform $\hat{\mu}$ of $\mu \in \text{Prob}(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$)

1. $\hat{\mu}(0) = 1$ and $|\hat{\mu}(\xi)| \leq 1 \quad \forall \xi \in \mathbb{R}^d$.

2. $\hat{\mu} \in C_0(\mathbb{R}^d)$

3. $\hat{\mu}(\xi) = \hat{\mu}(-\xi) \quad \forall \xi \in \mathbb{R}^d$. In particular, $\hat{\mu}$ is \mathbb{R} -valued

iff μ is symmetric ($\mu(B) = \mu(-B) \quad \forall B \in \mathcal{B}(\mathbb{R}^d)$)

(if $\mu = \mu_{\Sigma}$, $\Sigma = -\Sigma$.)

4. If $\int_{\mathbb{R}^d} |x|^k \mu(dx) < \infty$ then $\hat{\mu} \in C_0^k$ and $\frac{\partial}{\partial \xi_{j_1}} \cdots \frac{\partial}{\partial \xi_{j_k}} \hat{\mu}(\xi) = \int_{\mathbb{R}^d} (ix_{j_1}) \cdots (ix_{j_k}) e^{i\xi \cdot x} \mu(dx)$

Pf. 1. $\hat{\mu}(\xi) = \int e^{i\xi \cdot x} \mu(dx)$

$$|\hat{\mu}(\xi)| = \left| \int e^{i\xi \cdot x} \mu(dx) \right|$$

2. If $\xi_n \rightarrow \xi$ in \mathbb{R}^d , then $e_{\xi_n}(x) = e^{i\xi_n \cdot x}$

3. $\overline{\hat{\mu}(\xi)} = \int e^{i\xi \cdot x} \mu(dx)$

\therefore If μ is symmetric, let $\bar{\mu} \stackrel{d}{=} \mu$

Conversely, let $\nu(B) := \mu(-B)$. If $\hat{\mu}(\xi) = \hat{\nu}(\xi)$

$$4. \frac{\partial}{\partial z_{j_1}} \dots \frac{\partial}{\partial z_{j_k}} e^{i\zeta \cdot x} = (ix_{j_1}) \dots (ix_{j_k}) e^{i\zeta \cdot x} \quad \forall x \in \mathbb{R}^d$$

[Lecture 10.1] Differentiate under the \int

• need $\left| \frac{\partial}{\partial z_{j_1}} \dots \frac{\partial}{\partial z_{j_k}} e^{i\zeta \cdot x} \right| \leq g_{j_k}(x)$ for some $g_{j_k} \in L^1(\mu)$,
 $\forall \zeta \in K$

Prop. If $\mu, \nu \in \text{Prob}(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$, then

$$\widehat{\mu * \nu} = \widehat{\mu} \cdot \widehat{\nu}$$

Ex. If $\underline{X}, \underline{Y}$ are independent random vectors in \mathbb{R}^d , then

$$\varphi_{\underline{X} + \underline{Y}}(\xi) = \varphi_{\underline{X}}(\xi) \cdot \varphi_{\underline{Y}}(\xi) \quad \forall \xi \in \mathbb{R}^d.$$

Moreover: if $a \in \mathbb{R}$, $v \in \mathbb{R}^d$, then $\varphi_{a\underline{X} + v}(\xi) = e^{i\xi \cdot v} \varphi_{\underline{X}}(a\xi)$.

Pf. $\varphi_{\underline{X} + \underline{Y}}(\xi) = \mathbb{E}[e^{i\xi \cdot (\underline{X} + \underline{Y})}]$

Eg. $U \stackrel{d}{=} \text{Unif}([a, b])$

$$\varphi_U(\xi) = \int_a^b \frac{1}{b-a} e^{i\xi x} dx = \frac{e^{ib\xi} - e^{ia\xi}}{i(b-a)\xi}$$

Special case: if $a = -b$, $\varphi_U(\xi) = \frac{e^{ib\xi} - e^{-ib\xi}}{2ib\xi}$

Eg. $N \stackrel{d}{=} \text{Poisson}(\lambda)$

$$\varphi_N(\xi) = \mathbb{E}[e^{i\xi N}] = \sum_{n=0}^{\infty} e^{i\xi n} \cdot e^{-\lambda} \frac{\lambda^n}{n!}$$

Eg. $T \stackrel{d}{=} \text{Exp}(\lambda)$

$$\varphi_T(\xi) = \int_0^{\infty} \lambda e^{-\lambda x} e^{ix\xi} dx$$

Variant: bilateral exponential T_{\pm} with density $f(x) = \frac{\lambda}{2} e^{-\lambda|x|}$.

$$\varphi_{T_{\pm}}(\xi) = \int_{\mathbb{R}} \frac{\lambda}{2} e^{-\lambda|x|} e^{ix\xi} dx$$

Eg. $Y \stackrel{d}{=} \text{Rademacher} : P(Y = \pm 1) = \frac{1}{2}$

$$\varphi_X(\xi) = \mathbb{E}[e^{i\xi Y}] =$$

So, if Y_1, \dots, Y_n are iid Rademachers, $S_n = Y_1 + \dots + Y_n$

$$\varphi_{S_n}(\xi) =$$

$$\text{Rescale} : \varphi_{S_n/b_n}(\xi) =$$

Eg. $X \stackrel{d}{=} \mathcal{N}(0, 1)$

$$\varphi_X(\xi) = \mathbb{E}[e^{i\xi X}] = \int_{\mathbb{R}} e^{i\xi x} (2\pi)^{-1/2} e^{-x^2/2} dx$$

$$\varphi_X'(\xi) = \mathbb{E}[iX e^{i\xi X}]$$