

Lecture. 19. Char F and weak conv

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9:26 AM

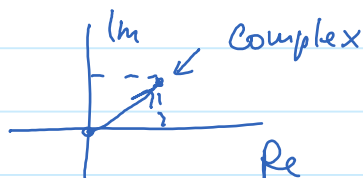
Characteristic functions

$$X \quad \varphi_X(t) = \mathbb{E}\{e^{itX}\}$$

$$i = \sqrt{-1}$$

Euler $e^{itx} = \cos x + i \sin x$

Complex plain



$$|e^{itX}| = 1$$

φ_X always exists, because $|\varphi_X| \leq 1$

Looking forward to results like

$$\varphi_n \rightarrow \varphi \Rightarrow X_n \rightsquigarrow X$$

Statistics :

Consistency :

$$X_n \xrightarrow{p} c \xrightarrow{LLN}$$

$$\text{Weak convergence } \tilde{X}_n \rightsquigarrow X \xrightarrow{CLT}$$

usually $N(\cdot, \cdot)$

Convolution

$$X \perp Y$$

$$\varphi_{X+Y}(t) = \varphi_X(t) \cdot \varphi_Y(t)$$

$\square \{X_k\}_{k=1}^n$ i.i.d. r.v.

$$\varphi_{\sum_{k=1}^n X_k}(t) = \varphi_{X_1}^n(t)$$

Moments

X r.v.

$\varphi_X(t)$ - char. funct.

$$\varphi_X^{(k)}(0) = i^k \underbrace{\mathbb{E}(X^k)}_{M_k}$$

Taylor expansion

$$\varphi_X(t) = \sum_{k=1}^n \frac{(it)^k}{k!} M_k + o(|t|^n)$$

Examples

① \square X is degenerate : $P(X=c)=1$

non-random

$$\varphi_X(t) = \mathbb{E}\{e^{itX}\} = e^{itc}$$

Consequence : if $\varphi_n \rightarrow e^{itc}$
then $X_n \xrightarrow{P} c$

②

N distribution ; $X \sim N(0,1)$

$$f_X(x) = e^{-x^2/2} \cdot \frac{1}{\sqrt{2\pi}}$$

Characteristic function

$$\begin{aligned}\varphi_X(t) &= \mathbb{E}\{e^{itX}\} = \\ &= \int_{-\infty}^{\infty} e^{itx - \frac{x^2}{2}} dx \cdot \frac{1}{\sqrt{2\pi}}\end{aligned}$$

Write

φ for φ_X for brevity

$$\int_{-\infty}^{\infty} e^{itx - \frac{x^2}{2}} dx$$

$$\varphi'(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} ix e^{itx - \frac{x^2}{2}} dx =$$

$$= -\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{itx} \cdot i \cdot d \underbrace{e^{-\frac{x^2}{2}}}_{-e^{-\frac{x^2}{2}} \cdot x} = \quad \swarrow \text{Integrating by parts}$$

$$= -\frac{1}{\sqrt{2\pi}} i e^{itx} \cdot e^{-\frac{x^2}{2}} \Big|_{x=-\infty}^{\infty} - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} t \cdot e^{itx - \frac{x^2}{2}} dx =$$

$$i d e^{itx} = \underbrace{i^2}_{-1} t \cdot e^{itx} dx \quad \nearrow$$

$$= -t \varphi(t)$$

$$\varphi'(t) = -t \varphi(t)$$

$$\varphi'(t) = \frac{d\varphi}{dt}$$

$$\frac{d\varphi}{\varphi} = -t dt = -d\frac{t^2}{2}$$

$$d \log \varphi = d\left(-\frac{t^2}{2}\right)$$

$$\varphi = C \cdot e^{-\frac{t^2}{2}}$$

$$\varphi(0) = 1 = \mathbb{E}\{e^{itx}\} \Big|_{t=0} \Rightarrow C = 1$$

$$\varphi_{N(0,1)}(t) = e^{-t^2/2}$$

$$X \sim N(\mu, \sigma^2)$$

$$\varphi_X(t) = e^{it\mu - \frac{t^2\sigma^2}{2}}$$

Multivariate Characteristic functions

Multivariate Characteristic functions

] X is a r. vector

(DF)

$$\varphi_X(t) = \mathbb{E} \left\{ e^{it^T \cdot X} \right\}$$

↑
a vector

Example:

$$X \sim N(\mu, \Sigma) \quad \varphi_X(t) = e^{it^T \mu - \frac{1}{2} t^T \Sigma t}$$

↓ ↓
vector matrix

(TH)

φ uniquely determines the distribution

$$(1) \quad \varphi_X \equiv \varphi_Y := \varphi \quad \Leftrightarrow \quad X \text{ and } Y \text{ have same distributions}$$

↓
uniformly over the argument

(2) Inversion formula for densities

$$f_X(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-itx} \varphi(t) dt$$

$$(3) \quad F_X(y) - F_X(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{-itx} - e^{-ity}}{it} \varphi(t) dt$$

Proof:

Informally, $\int_{-\infty}^y e^{-itx} dx = \frac{-1}{it} \cdot e^{-itx} \Big|_{x=-\infty}^y = \frac{e^{-ity} - e^{-i\infty}}{it}$

$$\int_{-\infty}^y \dots dx - \int_{-\infty}^x \dots dx = \frac{e^{-itx} - e^{-ity}}{it}$$

$$(2) \Rightarrow (3)$$

Proof of (i): $Z \sim N(0,1)$, $Z \perp X$

Consider $\tilde{X} = X + \sigma \cdot Z$

$$f_{\tilde{X}}(x) = \int_{-\infty}^{\infty} f_{\sigma Z}(x-y) \cdot \underbrace{f_X(y)}_{dF_X(y)} dy =$$

convolution

Fubini

$$= \mathbb{E} \left\{ f_{\sigma Z}(x-X) \right\} = \frac{1}{\sqrt{2\pi}\sigma} \mathbb{E} \left\{ e^{-\frac{(x-X)^2}{2\sigma^2}} \right\}$$

$\sigma Z \sim N(0, \sigma^2)$

Consider

$$\int e^{-itx} \varphi_{\tilde{X}}(t) dt = \mathbb{E} \left\{ \int e^{-it(X-x) - \frac{t^2\sigma^2}{2}} dt \right\} =$$

$$\varphi_{\tilde{X}}(t) = \varphi_X(t) \cdot \varphi_{\sigma Z}(t) = \mathbb{E} \left\{ e^{itX} \right\} \cdot \underbrace{e^{-\frac{t^2\sigma^2}{2}}}_{\varphi_{N(0, \sigma^2)}}$$

$$= \mathbb{E} \left\{ \int e^{it(X-x)} \cdot \underbrace{e^{-\frac{t^2}{2(\frac{1}{\sigma})^2}} \frac{1}{\sqrt{2\pi} \frac{1}{\sigma}}}_{\tilde{Z} \sim \text{pdf of } N(0, \frac{1}{\sigma^2})} dt \right\} \sqrt{2\pi} \frac{1}{\sigma} =$$

$$= \mathbb{E} \left\{ e^{i(X-x)\tilde{Z}} \right\} \sqrt{2\pi} \frac{1}{\sigma} =$$

$$\varphi_{\tilde{Z}}(X-x) = e^{i(X-x) \cdot 0 - \frac{(X-x)^2 \left(\frac{1}{\sigma}\right)^2}{2}} = e^{-\frac{(X-x)^2}{2\sigma^2}}$$

$$= \frac{1}{\sigma} \sqrt{2\pi} \mathbb{E} \left\{ e^{-\frac{(X-x)^2}{2\sigma^2}} \right\}$$

$$\int e^{-itx} \varphi_{\tilde{X}}(t) dt = \frac{\sqrt{2\pi}}{\sigma} \mathbb{E} \left\{ e^{-\frac{(X-x)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi}\sigma} \right\} \cdot \sqrt{2\pi}\sigma =$$

$$\int e^{-itx} \varphi_{\tilde{X}}(t) dt = \frac{\sqrt{2\pi}}{\sigma} \int \left\{ e^{-\frac{(x-y)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi}\sigma} \right\} \cdot \sqrt{2\pi}\sigma =$$

$$= 2\pi \cdot f_{\tilde{X}}(x)$$

↑
based on the form of $f_{\tilde{X}}$ obtained as a
convolution pdf

i.e.

$$\forall \text{ r.v. } X \perp \sigma Z \sim N(0, \sigma^2), \quad \tilde{X} = X + \sigma Z$$

$$f_{\tilde{X}}(x) = \frac{1}{2\pi} \int e^{-itx} \varphi_{\tilde{X}}(t) dt$$

By Slutsky

$$\tilde{X} = X + \sigma Z \rightsquigarrow X, \quad \sigma \rightarrow 0$$

$$\text{and } \varphi_{\tilde{X}} \rightarrow \varphi_X, \quad \sigma \rightarrow 0$$

If X, Y have same char. funct. \Rightarrow

$$\Rightarrow X + \sigma Z, Y + \sigma Z \text{ have same lin in distribution}$$

$$X \stackrel{d}{=} Y \quad \square$$

(TH) Lévy

$$(1) \quad X_n \rightsquigarrow X \Leftrightarrow \varphi_n(t) \rightarrow \varphi(t), \quad \forall t$$

$$(2) \quad \text{If } \varphi_n(t) \rightarrow \varphi(t), \quad \forall t \text{ and } \varphi \text{ is continuous at } t=0$$

$\Rightarrow \varphi$ is a characteristic function of
some r.v. X , and
 $X_n \rightsquigarrow X$

w/o proof